

LMIs in Control

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1 Basic Matrix Theory

1.1 Basic Matrix Notation

Consider the complex matrix $A \in \mathbb{C}^{n \times m}$.

$$A = \begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix} \in \mathbb{C}^{n \times m}$$

Transpose of a Matrix

The transpose of A , denoted as A^T or A' is:

$$A^T = \begin{bmatrix} a_{11} & \dots & a_{n1} \\ \vdots & \ddots & \vdots \\ a_{1m} & \dots & a_{nm} \end{bmatrix} \in \mathbb{C}^{m \times n}.$$

Adjoint of a Matrix

The adjoint or hermitian conjugate of A , denoted as A^* is:

$$A^* = \begin{bmatrix} a_{11}^* & \dots & a_{n1}^* \\ \vdots & \ddots & \vdots \\ a_{1m}^* & \dots & a_{nm}^* \end{bmatrix} \in \mathbb{C}^{m \times n}.$$

Where a_{nm}^* is the complex conjugate of matrix element a_{nm} .

Notice that for a real matrix $A \in \mathbb{R}^{n \times m}$, $A^* = A^T$.

1.2 Important Properties of Matricies

Hermitian, Self-Adjoint, and Symmetric Matrices

A square matrix $A \in \mathbb{C}^{n \times n}$ is called Hermitian or self-adjoint if $A = A^*$.

If $A \in \mathbb{R}^{n \times n}$ is Hermitian then it is called symmetric.

Unitary Matrices

A square matrix $A \in \mathbb{C}^{n \times n}$ is called unitary if $A^* = A^{-1}$ or $A^*A = I$.

1.3 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

2 Notion of Matrix Positivity

2.1 Notation of Positivity

A symmetric matrix $A \in \mathbb{R}^{n \times n}$ is defined to be:

positive semidefinite, ($A \geq 0$), if $x^T Ax \geq 0$ for all $x \in \mathbb{R}^n, x \neq \mathbf{0}$.

positive definite, ($A > 0$), if $x^T Ax > 0$ for all $x \in \mathbb{R}^n, x \neq \mathbf{0}$.

negative semidefinite, ($-A \geq 0$).

negative definite, ($-A > 0$).

indefinite if A is neither positive semidefinite nor negative semidefinite.

2.2 Properties of Positive Matrices

- For any matrix M , $M^T M > 0$.
- Positive definite matrices are invertible and the inverse is also positive definite.
- A positive definite matrix $A > 0$ has a square root, $A^{1/2} > 0$, such that $A^{1/2} A^{1/2} = A$.
- For a positive definite matrix $A > 0$ and invertible M , $M^T A M > 0$.
- If $A > 0$ and $M > 0$, then $A + M > 0$.
- If $A > 0$ then $\mu A > 0$ for any scalar $\mu > 0$.

2.3 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

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1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599>

3 <https://web.stanford.edu/~boyd/lmibook/>

3 KYP Lemma (Bounded Real Lemma)

KYP Lemma (Bounded Real Lemma)

The Kalman–Popov–Yakubovich (KYP) Lemma is a widely used lemma in control theory. It is sometimes also referred to as the Bounded Real Lemma. The KYP lemma can be used to determine the H_∞ norm of a system and is also useful for proving many LMI results.

3.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t) \\ x(0) &= x_0\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$, $u(t) \in \mathbb{R}^q$, at any $t \in \mathbb{R}$.

3.2 The Data

The matrices A, B, C, D are known.

3.3 The Optimization Problem

The following optimization problem must be solved.

$$\begin{array}{ll}\text{minimize}_{\gamma, X} & \gamma \\ \text{subject to} & X > 0 \\ & \begin{bmatrix} A^T X + XA & XB \\ B^T X & -\gamma I \end{bmatrix} + \frac{1}{\gamma} \begin{bmatrix} C^T \\ D^T \end{bmatrix} \begin{bmatrix} C & D \end{bmatrix} < 0\end{array}$$

3.4 The LMI: The KYP or Bounded Real Lemma

Suppose $\hat{G}(s)(A, B, C, D)$ is the system. Then the following are equivalent.

- 1) $\|G\|_{H_\infty} \leq \gamma$
- 2) There exists a $X > 0$ such that

$$\begin{bmatrix} A^T X + XA & XB \\ B^T X & -\gamma I \end{bmatrix} + \frac{1}{\gamma} \begin{bmatrix} C^T \\ D^T \end{bmatrix} \begin{bmatrix} C & D \end{bmatrix} < 0$$

3.5 Conclusion:

The KYP Lemma can be used to find the bound γ on the H_∞ norm of a system. Note from the (1,1) block of the LMI we know that A is Hurwitz.

3.6 Implementation

Since the KYP lemma shown above is nonlinear in gamma, in order to implement it in MATLAB we must first linearize it by using the Schur Complement to arrive at the dual formulation below:

$$\begin{bmatrix} A^T X + XA & XB & C^T \\ B^T X & -\gamma I & D^T \\ C & D & -\gamma I \end{bmatrix} < 0$$

This dual KYP LMI is now linear in both X and γ .

This implementation requires the use of Yalmip and Sedumi.

https://github.com/eoskowro/LMI/blob/master/KYP_Lemma_LMI.m

3.7 Related LMIs

Positive Real Lemma¹

3.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/KYP_Lemmas/Positive_Real_Lemma

² http://control.asu.edu/MAE598_frame.htm

³ <https://arxiv.org/abs/1903.08599/>

⁴ <https://web.stanford.edu/~boyd/lmibook/>

4 Positive Real Lemma

Positive Real Lemma

The Positive Real Lemma is a variation of the Kalman–Popov–Yakubovich (KYP) Lemma. The Positive Real Lemma can be used to determine if a system is passive (positive real).

4.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t) \\ x(0) &= x_0\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$, $u(t) \in \mathbb{R}^q$, at any $t \in \mathbb{R}$.

4.2 The Data

The matrices A, B, C, D are known.

4.3 The LMI: The Positive Real Lemma

Suppose $\hat{G}(s)(A, B, C, D)$ is the system. Then the following are equivalent.

- 1) G is passive, i.e. $\langle u, Gu \rangle_{L_2} \geq 0$ ($\hat{G}(s) + \hat{G}(s)^* \geq 0$)
- 2) There exists a $X > 0$ such that

$$\begin{bmatrix} A^T X + XA & XB - C^T \\ B^T X - C & -D^T - D \end{bmatrix} \leq 0$$

4.4 Conclusion:

The Positive Real Lemma can be used to determine if the system G is passive. Note from the (1,1) block of the LMI we know that A is Hurwitz.

4.5 Implementation

This implementation requires Yalmip and Sedumi. https://github.com/eoskowro/LMI/blob/master/Positive_Real_Lemma.m

4.6 Related LMIs

KYP Lemma (Bounded Real Lemma)¹

4.7 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

1 [https://en.wikibooks.org/wiki/LMIs_in_Control/KYP_Lemmas/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/KYP_Lemmas/KYP_Lemma_(Bounded_Real_Lemma))
2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599/>
4 <https://web.stanford.edu/~boyd/lmibook/>

5 KYP Lemma for QSR Dissipative Systems

5.1 The Concept

In systems theory the concept of dissipativity was first introduced by Willems which describes dynamical systems by input-output properties. Considering a dynamical system described by its state $x(t)$, its input $u(t)$ and its output $y(t)$, the input-output correlation is given a supply rate $w(u(t), y(t))$. A system is said to be dissipative with respect to a supply rate if there exists a continuously differentiable storage function $V(x(t))$ such that $V(0) = 0$, $V(x(t)) \geq 0$ and

$$\dot{V}(x(t)) \leq w(u(t), y(t))$$

As a special case of dissipativity, a system is said to be passive if the above dissipativity inequality holds with respect to the passivity supply rate $w(u(t), y(t)) = u(t)^T y(t)$.

The physical interpretation is that $V(x)$ is the energy stored in the system, whereas $w(u(t), y(t))$ is the energy that is supplied to the system.

This notion has a strong connection with Lyapunov stability, where the storage functions may play, under certain conditions of controllability and observability of the dynamical system, the role of Lyapunov functions.

Roughly speaking, dissipativity theory is useful for the design of feedback control laws for linear and nonlinear systems. Dissipative systems theory has been discussed by Vasile M. Popov, Jan Camiel Willems, D.J. Hill, and P. Moylan. In the case of linear invariant systems, this is known as positive real transfer functions, and a fundamental tool is the so-called Kalman–Yakubovich–Popov lemma which relates the state space and the frequency domain properties of positive real systems. Dissipative systems are still an active field of research in systems and control, due to their important applications.

5.2 The System

Consider a continuous-time LTI system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization $(\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$, where $\mathcal{A} \in \mathbb{R}^{n \times n}$, $\mathcal{B} \in \mathbb{R}^{n \times m}$, $\mathcal{C} \in \mathbb{R}^{p \times n}$, and $\mathcal{D} \in \mathbb{R}^{p \times m}$.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

5.3 The Data

The matrices A, B, C and D which defines the state space of the system

5.4 The Optimization Problem

The system \mathcal{G} is **QSR** dissipative if

$$\int_0^T (y^T(t)Qy(t) + 2y^T(t)Su(t) + u^T(t)Ru(t)) dt \geq 0, \forall u \in \mathcal{L}_{2e}, \forall T \geq 0$$

where $u(t)$ is the input to \mathcal{G} , $y(t)$ is the output of \mathcal{G} , $Q \in \mathcal{S}^p$, $S \in \mathcal{R}^{p \times m}$, and $R \in \mathcal{S}^m$.

5.5 LMI : KYP Lemma for QSR Dissipative Systems

The system \mathcal{G} is also **QSR** dissipative if and only if there exists $P \in \mathcal{S}^n$, where $P > 0$, such that

$$\begin{bmatrix} PA + A^T P - C^T QC & PB - C^T S - C^T QD \\ (PB - C^T S - C^T QD)^T & -D^T QD - (D^T S + S^T D) - R \end{bmatrix} \leq 0.$$

5.6 Conclusion:

If there exist a positive definite P for the selected **Q, S** and **R** matrices then the system \mathcal{G} is **QSR** dissipative.

5.7 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

5.8 Related LMIs

KYP Lemma¹

5.9 References

1. J. C. Willems, “Dissipative dynamical systems - part I: General theory,” Archive Rational Mechanics and Analysis, vol. 45, no. 5, pp. 321–351, 1972.

¹ https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

2. D. J. Hill and P. J. Moylan, "The stability of nonlinear dissipative systems," IEEE Transactions on Automatic Control, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²

6 KYP Lemma without Feedthrough

6.1 The Concept

It is assumed in the Lemma that the state-space representation $(\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$ is minimal. Then Positive Realness (PR) of the transfer function $C(SI - A)^{-1}B + D$ is equivalent to the solvability of the set of LMIs given in this page. Consider now the following scalar example, where $(A, B, C, D) = (-a, 0, 0, 1)$, with $a > 0$. The transfer function is $H(s) = 0$ that is PR

6.2 The System

Consider a continuous-time LTI system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization $(\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{0})$, where $\mathcal{A} \in \mathcal{R}^{n \times n}$, $\mathcal{B} \in \mathcal{R}^{n \times m}$, and $\mathcal{C} \in \mathcal{R}^{m \times n}$.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t)\end{aligned}$$

6.3 The Data

The matrices A, B and C

6.4 LMI : KYP Lemma without Feedthrough

The system \mathcal{G} is positive real (PR) under either of the following equivalent necessary and sufficient conditions.

1. There exists $P \in \mathcal{S}^n$, where $P > 0$ such that

$$\begin{aligned}PA + A^T P &\geq 0 \\ PB &= C^T\end{aligned}$$

2. There exists $Q \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{aligned}AQ + QA^T &\geq 0 \\ B &= QC^T\end{aligned}$$

This is a special case of the KYP Lemma for QSR dissipative systems with $\mathbf{Q} = \mathbf{0}$, $\mathbf{Q} = \mathbf{0.5}$ and $\mathbf{R} = \mathbf{0}$.

The system \mathcal{G} is strictly positive real (SPR) under either of the following equivalent necessary and sufficient conditions.

1. There exists $P \in \mathcal{S}^n$, where $p > 0$ such that

$$\begin{aligned} PA + A^T P &< 0 \\ PB &= C^T \end{aligned}$$

2. There exists $Q \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{aligned} AQ + QA^T &< 0 \\ B &= QC^T \end{aligned}$$

This is a special case of the KYP Lemma for QSR dissipative systems with $\mathbf{Q} = \epsilon \mathbf{1}$, $\mathbf{Q} = \mathbf{0.5}$ and $\mathbf{R} = \mathbf{0}$. where $\epsilon \in \mathcal{R}_{>0}$.

6.5 Conclusion:

If there exist a positive definite P for the selected \mathbf{Q}, \mathbf{S} and \mathbf{R} matrices then the system \mathcal{G} is **Positive Real**.

6.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

6.7 Related LMIs

KYP Lemma¹

State Space Stability²

Discrete Time KYP Lemma with Feedthrough³

6.8 References

1. J. C. Willems, “Dissipative dynamical systems - part I: General theory,” Archive Rational Mechanics and Analysis, vol. 45, no. 5, pp. 321–351, 1972.

1 https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

2 https://en.wikipedia.org/wiki/State-Space_Stability

3 https://en.wikipedia.org/wiki/Discrete_Time_KYP_Lemma_with_Feedthrough

2. D. J. Hill and P. J. Moylan, "The stability of nonlinear dissipative systems," IEEE Transactions on Automatic Control, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Brogliato B., Maschke B., Lozano R., Egeland O. (2007) Kalman-Yakubovich-Popov Lemma. In: Dissipative Systems Analysis and Control. Communications and Control Engineering. Springer, London

7 KYP Lemma for Descriptor Systems

7.1 The Concept

Descriptor system descriptions frequently appear when solving computational problems in the analysis and design of standard linear systems. The numerically reliable solution of many standard control problems like the solution of Riccati equations, computation of system zeros, design of fault detection and isolation filters (FDI), etc. relies on using descriptor system techniques.

Many algorithm for standard systems as for example stabilization techniques, factorization methods, minimal realization, model reduction, etc. have been extended to the more general descriptor system descriptions. An important application of these algorithms is the numerically reliable computation with rational and polynomial matrices via equivalent descriptor representations. Recall that each rational matrix $R(s)$ can be seen as the transfer-function matrix of a continuous- or discrete-time descriptor system. Thus, each $R(s)$ can be equivalently realized by a descriptor system quadruple $(A-sE, B, C, D)$ satisfying $R(S) = C(SE-A)^{-1}B+D$

Many operations on standard matrices (e.g., finding the rank, determinant, inverse or generalized inverses), or the solution of linear matrix equations can be performed for rational matrices as well using descriptor system techniques. Other important applications of descriptor techniques are the computation of inner-outer and spectral factorisations, or minimum degree and normalized coprime factorisations of polynomial and rational matrices. More explanation can be found in the website of Institute of System Dynamics and control¹

7.2 The System

Consider a square, continuous-time linear time-invariant (LTI) system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization $(\mathbf{E}, \mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$, where $\mathcal{E}, \mathcal{A} \in \mathcal{R}^{n \times n}, \mathcal{B} \in \mathcal{R}^{n \times m}, \mathcal{C} \in \mathcal{R}^{p \times n}$, and $\mathcal{D} \in \mathcal{R}^{p \times m}$.

$$\begin{aligned} E\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t) \end{aligned}$$

7.3 The Data

The matrices E, A, B, C and D

¹ https://www.dlr.de/rm/en/desktopdefault.aspx/tabid-3844/6357_read-9125/

7.4 LMI : KYP Lemma for Descriptor Systems

The system \mathcal{G} is extended strictly positive real (ESPR) if and only if there exists $X \in \mathbb{R}^{n \times n}$ and $W \in \mathbb{R}^{n \times m}$ such that

$$E^T X = X^T E \geq 0$$

$$E^T W = 0$$

$$\begin{bmatrix} X^T A + A^T X & A^T W + X^T B - C^T \\ (A^T W + X^T B - C^T)^T & W^T B + B^T W - (D^T + D) \end{bmatrix} < 0.$$

The system is also *ESPR* if there exists $X \in \mathbb{R}^{n \times n}$ such that

$$E^T X = X^T E \geq 0$$

$$\begin{bmatrix} X^T A + A^T X & X^T B - C^T \\ (X^T B - C^T)^T & -(D^T + D) \end{bmatrix} < 0.$$

7.5 Conclusion:

If there exist a X and W matrix satisfying above LMIs then the system \mathcal{G} is **Extended Strictly Positive Real**.

7.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

7.7 Related LMIs

KYP Lemma²

State Space Stability³

Discrete Time KYP Lemma with Feedthrough⁴

2 https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

3 https://en.wikipedia.org/wiki/State-Space_Stability

4 https://en.wikipedia.org/wiki/Discrete_Time_KYP_Lemma_with_Feedthrough

7.8 References

1. J. C. Willems, "Dissipative dynamical systems - part I: General theory," *Archive Rational Mechanics and Analysis*, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, "The stability of nonlinear dissipative systems," *IEEE Transactions on Automatic Control*, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Brogliato B., Maschke B., Lozano R., Egeland O. (2007) Kalman-Yakubovich-Popov Lemma. In: Dissipative Systems Analysis and Control. Communications and Control Engineering. Springer, London
5. Numerical algorithms and software tools for analysis and modelling of descriptor systems. Prepr. of 2nd IFAC Workshop on System Structure and Control, Prague, Czechoslovakia, pp. 392-395, 1992.

8 Generealized KYP (GKYP) Lemma for conic Sectors

8.1 The Concept

The conic sector theorem is a powerful input-output stability analysis tool, providing a fine balance between generality and simplicity of system characterisations that is conducive to practical stability analysis and robust controller synthesis.

8.2 The System

Consider a square, continuous-time linear time-invariant (LTI) system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization $(\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$, where $\mathcal{E}, \mathcal{A} \in \mathcal{R}^{n \times n}$, $\mathcal{B} \in \mathcal{R}^{n \times m}$, $\mathcal{C} \in \mathcal{R}^{p \times n}$, and $\mathcal{D} \in \mathcal{R}^{p \times m}$.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

Also consider $\pi_c(a, b) \in \mathcal{S}^m$, which is defined as

$$\pi_c(a, b) = \begin{bmatrix} -\frac{1}{b}I & \frac{1}{2}(1 + \frac{a}{b})I \\ (\frac{1}{2}(1 + \frac{a}{b})I)^T & -aI \end{bmatrix}$$

where $a \in \mathcal{R}, b \in \mathcal{R}_{>0}$ and $a < b$.

8.3 The Data

The matrices A, B, C and D . The values of a and b

8.4 LMI : Generalized KYP (GKYP) Lemma for Conic Sectors

The following generalized *KYP* Lemmas give conditions for \mathcal{G} to be inside the cone $[a, b]$ within finite frequency bandwidths.

1. (*Low Frequency Range*) The system \mathcal{G} is inside the cone $[a, b]$ for all $\omega \in \mathcal{R} | |\omega| < \omega_1, \det(j\omega I - A) \neq 0$, where $\omega_1 \in \mathcal{R}_{>0}, a \in \mathcal{R}, b \in \mathcal{R}_{>0}$ and $a < b$, if there exist $P, Q \in \mathcal{S}^n$ and $\bar{\omega}_1 \in \mathcal{R}_{>0}$, where $Q \geq 0$, such that

$$\begin{bmatrix} A & B \\ I & 0 \end{bmatrix}^T \begin{bmatrix} -Q & P \\ P^T & (\omega_1 - \bar{\omega}_1)^2 Q \end{bmatrix} \begin{bmatrix} A & B \\ I & 0 \end{bmatrix} - \begin{bmatrix} C & D \\ 0 & I \end{bmatrix}^T \pi_c(a, b) \begin{bmatrix} C & D \\ 0 & I \end{bmatrix} < 0$$

If $\omega_1 \rightarrow \infty$, $P > 0$. and $Q = 0$, then the traditional Conic Sector Lemma is recovered. The parameter $\bar{\omega}_1$ is incuded in the above LMI to effectively transform $|\omega| \leq (\omega_1 - \bar{\omega}_1)$ into the strict inequality $|\omega| < \omega_1$

2. (*Intermediate Frequency Range*) The system \mathcal{G} is inside the cone $[a, b]$ for all $\omega \in \mathcal{R} | \omega_1 \leq |\omega| < \omega_2, \det(j\omega I - A) \neq 0$, where $\omega_1, \omega_2 \in \mathcal{R}_{>0}, a \in \mathcal{R}, b \in \mathcal{R}_{>0}$ and $a < b$, if there exist $P, Q \in \mathcal{C}^n$ and $\bar{\omega}_2 \in \mathcal{R}_{>0}$ and $\hat{\omega}_2 = (\omega_1 + \frac{(\omega_2 - \bar{\omega}_2)}{2})$, where $P^H = P, Q^H = Q$ and $Q \geq 0$, such that

$$\begin{bmatrix} A & B \\ I & 0 \end{bmatrix}^T \begin{bmatrix} -Q & P + |\hat{\omega}_2| Q \\ P - |\hat{\omega}_2| Q & \omega_1(\omega_2 - \hat{\omega}_2 - 2)Q \end{bmatrix} \begin{bmatrix} A & B \\ I & 0 \end{bmatrix} - \begin{bmatrix} C & D \\ 0 & I \end{bmatrix}^T \pi_c(a, b) \begin{bmatrix} C & D \\ 0 & I \end{bmatrix} < 0$$

The parameter $\bar{\omega}_2$ is incuded in the above LMI to effectively transform $\omega_1 \leq |\omega| \leq (\omega_2 - \bar{\omega}_2)$ into the strict inequality $\omega_1 \leq |\omega| < \omega_2$.

3. (*High Frequency Range*) The system \mathcal{G} is inside the cone $[a, b]$ for all $\omega \in \mathcal{R} | \omega_2 < |\omega|, \det(j\omega I - A) \neq 0$, where $\omega_2 \in \mathcal{R}_{>0}, a \in \mathcal{R}, b \in \mathcal{R}_{>0}$ and $a < b$, if there exist $P, Q \in \mathcal{S}^n$, where $Q \geq 0$, such that

$$\begin{bmatrix} A & B \\ I & 0 \end{bmatrix}^T \begin{bmatrix} -Q & P \\ P^T & \omega_2^2 Q \end{bmatrix} \begin{bmatrix} A & B \\ I & 0 \end{bmatrix} - \begin{bmatrix} C & D \\ 0 & I \end{bmatrix}^T \pi_c(a, b) \begin{bmatrix} C & D \\ 0 & I \end{bmatrix} < 0$$

If **(A, B, C, D)** is a minimal realization, then the matrix inequalities in all of the above LMI, then it can be nostrict.

8.5 Conclusion:

If there exist a positive definite q matrix satisfying above LMIs for the given frequency bandwidths then the system \mathcal{G} is inside the cone $[a, b]$.

8.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

8.7 Related LMIs

KYP Lemma¹

¹ https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

State Space Stability²

Exterior Conic Sector Lemma³

Modified Exterior Conic Sector Lemma⁴

8.8 References

1. J. C. Willems, "Dissipative dynamical systems - part I: General theory," *Archive Rational Mechanics and Analysis*, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, "The stability of nonlinear dissipative systems," *IEEE Transactions on Automatic Control*, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Bridgeman, Leila Jasmine, and James Richard Forbes. "The exterior conic sector lemma." *International Journal of Control* 88.11 (2015): 2250-2263.

² https://en.wikipedia.org/wiki/State-Space_Stability

³ https://en.wikipedia.org/wiki/Exterior_Conic_Sector_Lemma

⁴ https://en.wikipedia.org/wiki/Modified_Exterior_Conic_Sector_Lemma

9 Discrete time Bounded Real Lemma

Discrete-Time Bounded Real Lemma

A discrete time system operates on a discrete time signal input and produces a discrete time signal output. They are used in digital signal processing, such as digital filters for images or sound. The class of discrete time systems that are both linear and time invariant, known as discrete time LTI systems.

Discrete-Time Bounded Real Lemma or the $H\infty$ norm can be found by solving a LMI.

9.1 The System

Discrete-Time LTI System with state space realization (A_d, B_d, C_d, D_d)

$$A_d \in \mathbf{R}^{n*n}, \quad B_d \in \mathbf{R}^{n*m}, \quad C_d \in \mathbf{R}^{p*n}, \quad D_d \in \mathbf{R}^{p*m}$$

9.2 The Data

The matrices: System $(A_d, B_d, C_d, D_d), P$.

9.3 The Optimization Problem

The following feasibility problem should be optimized:

γ is minimized while obeying the LMI constraints.

9.4 The LMI:

Discrete-Time Bounded Real Lemma

The LMI formulation

$$H\infty \text{norm} < \gamma$$

$$\begin{aligned} P \in S^n; \gamma \in R_{>0} \\ P > 0, \\ \begin{bmatrix} A_d^T P A_d - P & A_d^T P B_d & C_d^T \\ * & B_d^T P B_d - \gamma I & D_d^T \\ * & * & -\gamma I \end{bmatrix} < 0, \end{aligned}$$

9.5 Conclusion:

The H_∞ norm is the minimum value of $\gamma \in R_{>0}$ that satisfies the LMI condition. If (A_d, B_d, C_d, D_d) is the minimal realization then the inequalities can be non-strict.

9.6 Implementation

A link to CodeOcean or other online implementation of the LMI
MATLAB Code¹

9.7 Related LMIs

[https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma)) - Continuous time KYP_Lemma_(Bounded_Real_Lemma)

9.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

1 <https://github.com/Harishankar-Prabhakaran/LMIs/blob/master/A2.m>

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599/>

4 <https://web.stanford.edu/~boyd/lmibook/>

10 Discrete Time KYP Lemma for QSR Dissipative System

10.1 The Concept

In systems theory the concept of dissipativity was first introduced by Willems which describes dynamical systems by input-output properties. Considering a dynamical system described by its state $x(t)$, its input $u(t)$ and its output $y(t)$, the input-output correlation is given a supply rate $w(u(t), y(t))$. A system is said to be dissipative with respect to a supply rate if there exists a continuously differentiable storage function $V(x(t))$ such that $V(0) = 0$, $V(x(t)) \geq 0$ and

$$\dot{V}(x(t)) \leq w(u(t), y(t))$$

As a special case of dissipativity, a system is said to be passive if the above dissipativity inequality holds with respect to the passivity supply rate $w(u(t), y(t)) = u(t)^T y(t)$.

The physical interpretation is that $V(x)$ is the energy stored in the system, whereas $w(u(t), y(t))$ is the energy that is supplied to the system.

This notion has a strong connection with Lyapunov stability, where the storage functions may play, under certain conditions of controllability and observability of the dynamical system, the role of Lyapunov functions.

Roughly speaking, dissipativity theory is useful for the design of feedback control laws for linear and nonlinear systems. Dissipative systems theory has been discussed by Vasile M. Popov, Jan Camiel Willems, D.J. Hill, and P. Moylan. In the case of linear invariant systems, this is known as positive real transfer functions, and a fundamental tool is the so-called Kalman–Yakubovich–Popov lemma which relates the state space and the frequency domain properties of positive real systems. Dissipative systems are still an active field of research in systems and control, due to their important applications.

10.2 The System

Consider a discrete-time LTI system, $\mathcal{G} : \mathbb{J}_{2e} \rightarrow \mathbb{J}_{2e}$, with minimal state-space realization $(\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d, \mathcal{D}_d)$, where $\mathcal{A}_d \in \mathbb{R}^{n \times n}$, $\mathcal{B}_d \in \mathbb{R}^{n \times m}$, $\mathcal{C}_d \in \mathbb{R}^{p \times n}$, and $\mathcal{D}_d \in \mathbb{R}^{p \times m}$.

$$x(k+1) = \mathcal{A}_d x(k) + \mathcal{B}_d u(k)$$

$$y(k) = \mathcal{C}_d x(k) + \mathcal{D}_d u(k), k = 0, 1, \dots$$

10.3 The Data

The matrices $\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d$ and \mathcal{D}_d

10.4 The Optimization Problem

The system \mathcal{G} is **QSR** dissipative if

$$\sum_{i=0}^K (\dagger_i^T Q \dagger_i + 2\dagger_i^T S \sqcap_i + \sqcap_i^T R \sqcap_i) dt \geq 0, \forall u \in \mathbb{J}_{2e}, \forall k \in \mathcal{Z} \geq 0$$

where \sqcap_k is the input to \mathcal{G} , \dagger_k is the output of \mathcal{G} , $Q \in \mathcal{S}^p, S \in \mathcal{R}^{p \times m}$, and $R \in \mathcal{S}^m$.

10.5 LMI : Discrete-Time KYP Lemma for QSR Dissipative Systems

The system \mathcal{G} is also **QSR** dissipative if and only if there exists $P \in \mathcal{S}^n$, where $P > 0$, such that

$$\begin{bmatrix} A_d^T P A_d - P - C_d^T Q C_d & A_d^T P B_d - C_d^T S - C_d^T Q D_d \\ (A_d^T P B_d - C_d^T S - C_d^T Q D_d)^T & B_d^T P B_d - D_d^T Q D_d - (D_d^T S + S^T D_d) - R \end{bmatrix} \leq 0.$$

10.6 Conclusion:

If there exist a positive definite P for the selected **Q, S** and **R** matrices then the system \mathcal{G} is **QSR** dissipative.

10.7 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

10.8 Related LMIs

KYP Lemma¹

KYP Lemma for continuous Time QSR Dissipative system²

¹ https://en.wikipedia.org/wiki/Kalman%20%93Yakubovich%20%93Popov_lemma

² https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_QSR#The_Data

10.9 References

1. J. C. Willems, "Dissipative dynamical systems - part I: General theory," *Archive Rational Mechanics and Analysis*, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, "The stability of nonlinear dissipative systems," *IEEE Transactions on Automatic Control*, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²

11 Discrete Time KYP Lemma with Feedthrough

11.1 The Concept

It is assumed in the Lemma that the state-space representation $(\mathbf{A}_d, \mathbf{B}_d, \mathbf{C}_d, \mathbf{D}_d)$ is minimal. Then Positive Realness (PR) of the transfer function $C_d(SI - A_d)^{-1}B_d + D_d$ is equivalent to the solvability of the set of LMIs given in this page. Consider now the following scalar example, where $(A_d, B_d, C_d, D_d) = (-\alpha, 0, 0, 1)$, with $\alpha > 0$. The transfer function is $H(s) = 0$ that is PR

11.2 The System

Consider a discrete-time LTI system, $\mathcal{G} : \mathbb{F}_{2e} \rightarrow \mathbb{F}_{2e}$, with minimal state-space realization $(\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d, \mathcal{D}_d)$, where $\mathcal{A}_d \in \mathbb{R}^{n \times n}$, $\mathcal{B}_d \in \mathbb{R}^{n \times m}$, $\mathcal{C}_d \in \mathbb{R}^{p \times n}$, and $\mathcal{D}_d \in \mathbb{R}^{p \times m}$.

$$x(k+1) = \mathcal{A}_d x(k) + \mathcal{B}_d u(k)$$

$$y(k) = \mathcal{C}_d x(k) + \mathcal{D}_d u(k), k = 0, 1, \dots$$

11.3 The Data

The matrices $\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d$ and \mathcal{D}_d

11.4 LMI : Discrete-Time KYP Lemma with Feedthrough

The system \mathcal{G} is positive real (PR) under either of the following equivalent necessary and sufficient conditions.

1. There exists $P \in \mathcal{S}^n$, where $P > 0$ such that

$$\begin{bmatrix} A_d^T P A_d - P & A_d^T P B_d - C_d^T \\ (A_d^T P B_d - C_d^T)^T & B_d^T P B_d - (D_d^T + D_d) \end{bmatrix} \leq 0.$$

2. There exists $Q \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{bmatrix} A_d Q A_d^T - Q & A_d Q C_d^T - B_d \\ (A_d Q C_d^T - B_d)^T & C_d P C_d^T - (D_d^T + D_d) \end{bmatrix} \leq 0.$$

3. There exists $P \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{bmatrix} P & PA_d & PB_d \\ (PA_d)^T & P & C_d^T \\ (PB_d)^T & C_d & D_d^T + D_d \end{bmatrix} \geq 0.$$

4. There exists $Q \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{bmatrix} Q & A_d Q & B_d \\ (A_d Q)^T & Q & QC_d^T \\ (B_d)^T & (QC_d^T)^T & D_d^T + D_d \end{bmatrix} \geq 0.$$

This is a special case of the *KYP* Lemma for QSR dissipative systems with $\mathbf{Q} = \mathbf{0}$, $\mathbf{Q} = \mathbf{0.5}$ and $\mathbf{R} = \mathbf{0}$.

The system \mathcal{G} is strictly positive real (SPR) under either of the following equivalent necessary and sufficient conditions.

1. There exists $P \in \mathcal{S}^n$, where $P > 0$ such that

$$\begin{bmatrix} A_d^T P A_d - P & A_d^T P B_d - C_d^T \\ (A_d^T P B_d - C_d^T)^T & B_d^T P B_d - (D_d^T + D_d) \end{bmatrix} < 0.$$

2. There exists $Q \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{bmatrix} A_d Q A_d^T - Q & A_d Q C_d^T - B_d \\ (A_d Q C_d^T - B_d)^T & C_d P C_d^T - (D_d^T + D_d) \end{bmatrix} < 0.$$

3. There exists $P \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{bmatrix} P & PA_d & PB_d \\ (PA_d)^T & P & C_d^T \\ (PB_d)^T & C_d & D_d^T + D_d \end{bmatrix} > 0.$$

4. There exists $Q \in \mathcal{S}^n$, where $Q > 0$ such that

$$\begin{bmatrix} Q & A_d Q & B_d \\ (A_d Q)^T & Q & QC_d^T \\ (B_d)^T & (QC_d^T)^T & D_d^T + D_d \end{bmatrix} > 0.$$

This is a special case of the *KYP* Lemma for QSR dissipative systems with $\mathbf{Q} = \varepsilon \mathbf{1}$, $\mathbf{Q} = \mathbf{0.5}$ and $\mathbf{R} = \mathbf{0}$. where $\varepsilon \in \mathcal{R}_{>0}$.

11.5 Conclusion:

If there exist a positive definite P for the selected \mathbf{Q}, \mathbf{S} and \mathbf{R} matrices then the system \mathcal{G} is **Positive Real**.

11.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

11.7 Related LMIs

KYP Lemma¹

State Space Stability²

KYP Lemma without Feedthrough³

11.8 References

1. J. C. Willems, “Dissipative dynamical systems - part I: General theory,” Archive Rational Mechanics and Analysis, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, “The stability of nonlinear dissipative systems,” IEEE Transactions on Automatic Control, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Brogliato B., Maschke B., Lozano R., Egeland O. (2007) Kalman-Yakubovich-Popov Lemma. In: Dissipative Systems Analysis and Control. Communications and Control Engineering. Springer, London

¹ https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

² https://en.wikipedia.org/wiki/State-Space_Stability

³ https://en.wikipedia.org/wiki/KYP_Lemma_without_Feedthrough

12 Schur Complement

An important tool for proving many LMI theorems is the Schur Compliment. It is frequently used as a method of LMI linearization.

12.0.1 The Schur Compliment

Consider the matrices Q , M , and R where Q and M are self-adjoint. Then the following statements are equivalent:

1. $Q > 0$ and $M - RQ^{-1}R^* > 0$ both hold.
2. $M > 0$ and $Q - R^*M^{-1}R > 0$ both hold.
3. $\begin{bmatrix} M & R \\ R^* & Q \end{bmatrix} > 0$ is satisfied.

More concisely:

$$\begin{bmatrix} M & R \\ R^* & Q \end{bmatrix} > 0 \iff \begin{bmatrix} M & 0 \\ 0 & Q - R^*M^{-1}R \end{bmatrix} > 0 \iff \begin{bmatrix} M - RQ^{-1}R^* & 0 \\ 0 & Q \end{bmatrix} > 0$$

12.1 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

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1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599>

3 <https://web.stanford.edu/~boyd/lmibook/>

13 LMI for Eigenvalue Minimization

LMI for Minimizing Eigenvalue of a Matrix

Synthesizing the eigenvalues of a matrix plays an important role in designing controllers for linear systems. The eigenvalues of the state matrix of a linear time-invariant system determine if the system is stable or not. The system is stable if all the eigenvalues of the state matrix are located in the left half of the complex plane. Thus, we may desire to minimize the maximal eigenvalue of the state matrix such that the minimized eigenvalue is placed in the left half-plane, which guarantees that the system is stable.

13.1 The System

Assume that we have a matrix function of variables x :

$$A(x) = A_0 + A_1x_1 + \dots + A_nx_n$$

where $A_i, i = 1, 2, \dots, n$ are symmetric matrices.

13.2 The Data

The symmetric matrices $A_i (A_0, A_1, \dots, A_n)$ are given.

13.3 The Optimization Problem

The optimization problem is to find the variables $x = [x_1 \ x_2 \dots x_n]$ to minimize the following cost function:

$$J(x) = \lambda_{\max}(A(x))$$

where $J(x)$ is the cost function and $\lambda_{\max}(\cdot)$ indicates the maximum eigenvalue of a matrix.

According to Lemma 1.1 in LMI in Control Systems Analysis, Design and Applications¹ (page 10), the following statements are equivalent

$$\lambda_{\max}(A(x)) \leq t \iff A(x) - tI \leq 0$$

where t is defined as the maximum eigenvalue of the matrix A .

¹ <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

13.4 The LMI: LMI for eigenvalue minimization

This optimization problem can be converted to an LMI problem.

The mathematical description of the LMI formulation can be written as follows:

$$\begin{aligned} \min \quad & t \\ \text{s.t.} \quad & A(x) - tI \leq 0 \end{aligned}$$

13.5 Conclusion:

As a result, the variables $x_i, i = 1, 2, \dots, n$ after solving this LMI problem.

Moreover, we obtain the maximum eigenvalue, t , of the matrix $A(x)$.

13.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Minimizing-the-Maximum-Eigenvalue-of-Matrix>

13.7 Related LMIs

LMI for Generalized Eigenvalue Problem²

LMI for Matrix Norm Minimization³

LMI for Maximum Singular Value of a Complex Matrix⁴

LMI for Matrix Positivity⁵

13.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- State-space Representation of a System⁶
- Eigenvalues and Eigenvectors of a Matrix

² https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Generalized_eigenvalue_problem

³ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Matrix_Norm_Minimization

⁴ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools/Maximum_Singular_Value_of_a_Complex_Matrix

⁵ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools/Notion_of_Matrix_Positivity

⁶ https://en.wikipedia.org/wiki/State-space_representation

14 LMI for Matrix Norm Minimization

LMI for Matrix Norm Minimization

This problem is a slight generalization of the eigenvalue minimization problem for a matrix. Calculating norm of a matrix is necessary in designing an H_2 or an H_∞ optimal controller for linear time-invariant systems. In those cases, we need to compute the norm of the matrix of the closed-loop system. Moreover, we desire to design the controller so as to minimize the closed-loop matrix norm.

14.1 The System

Assume that we have a matrix function of variables x :

$$A(x) = A_0 + A_1x_1 + \dots + A_nx_n$$

where $A_i, i = 1, 2, \dots, n$ are symmetric matrices.

14.2 The Data

The symmetric matrices $A_i (A_0, A_1, \dots, A_n)$ are given.

14.3 The Optimization Problem

The optimization problem is to find the variables $x = [x_1 \ x_2 \dots x_n]$ in order to minimize the following cost function:

$$J(x) = \|A(x)\|_2$$

where $J(x)$ is the cost function and $\|\cdot\|_2$ indicates the norm of the matrix function A .

According to Lemma 1.1 in LMI in Control Systems Analysis, Design and Applications¹ (page 10), the following statements are equivalent:

$$A^T A - t^2 I \leq 0 \iff \begin{bmatrix} -tI & A \\ A^T & -tI \end{bmatrix} \leq 0$$

¹ <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

14.4 The LMI: LMI for matrix norm minimization

This optimization problem can be converted to an LMI problem.

The mathematical description of the LMI formulation can be written as follows:

$$\begin{aligned} \min \quad & t \\ \text{s.t.} \quad & \begin{bmatrix} -tI & A(x) \\ A(x)^T & -tI \end{bmatrix} \leq 0 \end{aligned}$$

14.5 Conclusion:

As a result, the variables $x_i, i = 1, 2, \dots, n$ after solving this LMI problem and we obtain t that is the norm of matrix function $A(x)$.

14.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Matrix-Norm-Minimization>

14.7 Related LMIs

LMI for Matrix Norm Minimization²

LMI for Generalized Eigenvalue Problem³

LMI for Maximum Singular Value of a Complex Matrix⁴

LMI for Matrix Positivity⁵

14.8 External Links

A list of references documenting and validating the LMI.

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

² https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/MatrixEigenValueMinimization&stable=0#Related_LMIs
³ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Generalized_eigenvalue_problem
⁴ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools/Maximum_Singular_Value_of_a_Complex_Matrix
⁵ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools/Notion_of_Matrix_Positivity

15 LMI for Generalized Eigenvalue Problem

LMI for Generalized Eigenvalue Problem

Technically, the generalized eigenvalue problem considers two matrices, like A and B , to find the generalized eigenvector, x , and eigenvalues, λ , that satisfies $Ax = \lambda Bx$. If the matrix B is an identity matrix with the proper dimension, the generalized eigenvalue problem is reduced to the eigenvalue problem.

15.1 The System

Assume that we have three matrix functions which are functions of variables $x = [x_1 \ x_2 \ \dots \ x_n]^T \in \mathbb{R}^n$ as follows:

$$A(x) = A_0 + A_1x_1 + \dots + A_nx_n$$

$$B(x) = B_0 + B_1x_1 + \dots + B_nx_n$$

$$C(x) = C_0 + C_1x_1 + \dots + C_nx_n$$

where A_i , B_i , and C_i ($i = 1, 2, \dots, n$) are the coefficient matrices.

15.2 The Data

The $A(x)$, $B(x)$, and $C(x)$ are matrix functions of appropriate dimensions which are all linear in the variable x and A_i , B_i , C_i are given matrix coefficients.

15.3 The Optimization Problem

The problem is to find $x = [x_1 \ x_2 \dots x_n]$ such that:

$A(x) < \lambda B(x)$, $B(x) > 0$, and $C(x) < 0$ are satisfied and λ is a scalar variable.

15.4 The LMI: LMI for Schur stabilization

A mathematical description of the LMI formulation for the generalized eigenvalue problem can be written as follows:

$$\begin{aligned} \min \quad & \lambda \\ \text{s.t.} \quad & A(x) < \lambda B(x) \\ & B(x) > 0 \\ & C(x) < 0 \end{aligned}$$

15.5 Conclusion:

The solution for this LMI problem is the values of variables x such that the scalar parameter, λ , is minimized. In practical applications, many problems involving LMIs can be expressed in the aforementioned form. In those cases, the objective is to minimize a scalar parameter that is involved in the constraints of the problem.

15.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Schur-Stability>

15.7 Related LMIs

LMI for Generalized Eigenvalue Problem¹

LMI for Matrix Norm Minimization²

LMI for Maximum Singular Value of a Complex Matrix³

15.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Generalized_eigenvalue_problem
2 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/MatrixEigenValueMinimization&stable=0#Related_LMIs
3 https://en.wikibooks.org/wiki/LMIs_in_Control/Tools/Maximum_Singular_Value_of_a_Complex_Matrix

16 LMI for Linear Programming

LMI for Linear Programming

Linear programming has been known as a technique for the optimization of a linear objective function subject to linear equality or inequality constraints. The feasible region for this problem is a convex polytope. This region is defined as a set of the intersection of many finite half-spaces which are created by the inequality constraints. The solution for this problem is to find a point in the polytope of existing solutions where the objective function has its extremum (minimum or maximum) value.

16.1 The System

We define the objective function as:

$$c_1x_1 + c_2x_2 + \dots + c_nx_n$$

and constraints of the problem as:

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n < b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n < b_2$$

.

.

$$a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n < b_m$$

16.2 The Data

Suppose that $c_j \in \mathbb{R}^n$, $a_{ij} \in \mathbb{R}^n$, and $b_i \in \mathbb{R}^m$ are given parameters where $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$. Moreover, $x = [x_1 \quad x_2 \quad \dots \quad x_n]^T$ is an $n \times 1$ vector of positive variables.

16.3 The Optimization Problem

The optimization problem is to minimize the objective function, $c^T x$ when the aforementioned linear constraints are satisfied.

16.4 The LMI: LMI for linear programming

The mathematical description of the optimization problem can be readily written in the following LMI formulation:

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & a_i^T x \leq b_i \end{aligned}$$

16.5 Conclusion:

Solving this problem results in the values of variables x which minimize the objective function. It is also worthwhile to note that if $m \geq n$, the computational cost for solving this problem would be mn^2 .

There does not exist an analytical formulation to solve a general linear programming problem. Nonetheless, there are some efficient algorithms, like the Simplex algorithm, for solving a linear programming problem.

16.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Linear-Programming>

16.7 Related LMIs

LMI for Feasibility Problem¹

16.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Feasibility_Problem

17 LMI for Feasibility Problem

LMI for Feasibility Problem in Optimization

The feasibility problem is to find any feasible solutions for an optimization problem without regard to the objective value. This problem can be considered as a special case of an optimization problem where the objective value is the same for all the feasible solutions. Many optimization problems have to start from a feasible point in the range of all possible solutions. One way is to add a slack variable to the problem in order to relax the feasibility condition. By adding the slack variable the problem any start point would be a feasible solution. Then, the optimization problem is converted to find the minimum value for the slack variable until the feasibility is satisfied.

17.1 The System

Assume that we have two matrices as follows:

$$A(x) = A_0 + A_1x_1 + \dots + A_nx_n \quad i = 1, 2, \dots, n$$

$$B(x) = B_0 + B_1x_1 + \dots + B_nx_n \quad i = 1, 2, \dots, n$$

which are matrix functions of variables $x = [x_1 \quad x_2 \quad \dots \quad x_n]^T \in \mathbb{R}^n$.

17.2 The Data

Suppose that the matrices A_0, A_1, \dots, A_n and B_0, B_1, \dots, B_n are given.

17.3 The Optimization Problem

The optimization problem is to find variables $x = [x_1 \quad x_2 \dots x_n]$ such that the following constraint is satisfied:

$$A(x) < B(x)$$

17.4 The LMI: LMI for Feasibility Problem

This optimization problem can be converted to a standard LMI problem by adding a slack variable, t .

The mathematical description for this problem is to minimize t in the following form of the LMI formulation:

$$\begin{aligned} \min \quad & t \\ \text{s.t.} \quad & A(x) < B(x) + tI \end{aligned}$$

17.5 Conclusion:

In this problem, x and t are decision variables of the LMI problem.

As a result, these variables are determined in the optimization problem such that the minimum value of t is found while the inequality constraint is satisfied.

17.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Feasibility-Problem-of-Convex-Optimization>

17.7 Related LMIs

LMI for Linear Programming¹

17.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Linear_Programming

18 Structured Singular Value

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The LMI can be used to find a Θ that belongs to the set of scalings $P\Theta$. Minimizing γ allows to minimize the squared norm of $\Theta M \Theta^{-1}$.

18.1 The System

M with transfer function $\hat{M}(s) = C(sI - A)^{-1}B + D$, $\hat{M} \in H_\infty$

18.2 The Data

The matrices $A \in R^{n \times n}, B \in R^{n \times m}, C \in R^{o \times n}, D \in R^{o \times m}$.

18.3 The Optimization Problem

There exists $\Theta \in \Theta$ such that $\|\Theta M \Theta^{-1}\|^2 < \gamma$.

18.4 The LMI:

$$\begin{aligned} & \text{Find } X > 0 : \\ & \begin{bmatrix} A^T X + XA & XB \\ B^T X & -\Theta \end{bmatrix} + \gamma^{-2} \begin{bmatrix} C^T \\ D^T \end{bmatrix} \Theta \begin{bmatrix} C & D \end{bmatrix} < 0 \end{aligned}$$

18.5 Conclusion:

The optimization problem and the LMI are equivalent. γ must be optimized using bisection.

18.6 Implementation

<https://github.com/mcavorsi/LMI>

18.7 Related LMIs

Eigenvalue Problem¹

18.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Matrix_and_LMI_Properties_and_Tools/Continuous_Time/Eigenvalue_Problem
2 http://control.asu.edu/MAE598_frame.htm

19 Eigenvalue Problem

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The maximum eigenvalue of a matrix is going to have the most impact on system performance. This LMI allows for minimization of the maximum eigenvalue by minimizing γ .

19.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t)\end{aligned}$$

19.2 The Data

The matrices $A \in R^{n \times n}, B \in R^{n \times m}, C \in R^{o \times n}$.

19.3 The Optimization Problem

Minimize γ subject to the LMI below.

19.4 The LMI:

$$\begin{aligned}\text{Find } P > 0 : \\ \begin{bmatrix} -A^T P - PA - C^T C & PB \\ B^T P & \gamma I \end{bmatrix} &> 0\end{aligned}$$

19.5 Conclusion:

The eigenvalue problem can be utilized to minimize the maximum eigenvalue of a matrix that depends affinely on a variable.

19.6 Implementation

<https://github.com/mcavorsi/LMI>

19.7 Related LMIs

Structured Singular Value¹

19.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Matrix_and_LMI_Properties_and_Tools/Continuous_Time/Structured_Singular_Value

2 http://control.asu.edu/MAE598_frame.htm

3 <https://web.stanford.edu/~boyd/lmibook/>

20 LMI for Minimizing Condition Number of Positive Definite Matrix

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20.1 The System:

A related problem is minimizing the condition number of a positive-definite matrix M that depends affinely on the variable x , subject to the LMI constraint $F(x) > 0$. This problem can be reformulated as the GEVP.

20.2 The Optimization Problem:

The GEVP can be formulated as follows:

minimize γ

subject to $F(x) > 0$;

$\mu > 0$;

$\mu I < M(x) < \gamma \mu I$.

We can reformulate this GEVP as an EVP as follows. Suppose,

$$M(x) = M_0 + \sum_{i=1}^m x_i M_i, \quad F(x) = F_0 + \sum_{i=1}^m x_i F_i$$

20.3 The LMI:

Defining the new variables $\nu = 1/\mu$, $\tilde{x} = x/\mu$ we can express the previous formulation as the EVP with variables \tilde{x}, ν and γ :

minimize γ

subject to $\nu F_0 + \sum_{i=1}^m x_i F_i > 0$; $I < \nu M_0 + \sum_{i=1}^m x_i M_i < \gamma I$

20.4 Conclusion:

The LMI is feasible.

20.5 Implementation

20.6 References

- LMIs in Systems and Control Theory¹ - A downloadable book on LMIs by Stephen Boyd.

¹ <https://web.stanford.edu/~boyd/lmibook/>

21 Continuous Quadratic Stability

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To study stability of a LTI system, we first ask whether all trajectories of system converge to zero as $t \rightarrow \infty$. A sufficient condition for this is the existence of a quadratic function $V(\xi) = \xi^T P \xi$, $P > 0$ that decreases along every nonzero trajectory of system . If there exists such a P , we say the system is quadratically stable and we call V a quadratic Lyapunov function.

21.1 The System

$$\dot{x}(t) = A(\delta(t))x(t)$$

21.2 The Data

The system coefficient matrix takes the form of

$$\dot{x}(t) = A_0 + \Delta A(\delta(t))x(t)$$

where $A_0 \in \mathbb{R}$ is a known matrix, which represents the nominal system matrix, while $\Delta A(\delta(t))x(t) = \delta_1(t)A_1 + \delta_2(t)A_2 + \dots + \delta_k(t)A_k$ is the system matrix perturbation, where $A_i \in \mathbb{R}^{n \times n}, i = 1, 2, \dots, k$, are known matrices, which represent the perturbation matrices.

$\delta_i(t), i = 1, 2, \dots, k$, which represent the uncertain parameters in the system.

$\delta(t) = [\delta_1(t) \delta_2(t) \dots \delta_k(t)]^T$ is the uncertain parameter vector, which is often assumed to be within a certain compact and convex set : Δ that is

$$\delta(t) = [\delta_1(t) \delta_2(t) \dots \delta_k(t)]^T \in \Delta$$

21.3 The LMI: Continuous-Time Quadratic Stability

The uncertain system is quadratically stable if and only if there exists $P \in \mathbb{S}^n$, where $P > 0$, such that

$$(A_0 + \Delta A(\delta(t))x(t))^T + P(A_0 + \Delta A(\delta(t))x(t)) < 0 \quad \delta(t) \in \Delta$$

The following statements can be made for particular sets of perturbations.

21.3.1 Case 1: Regular Polyhedron

Consider the case where the set of perturbation parameters is defined by a regular polyhedron as

$$\Delta = \delta(t) = [\delta_1(t) \delta_2(t) \dots \delta_k(t)] \in \mathbb{R}^k \mid \underline{\delta}_i(t), \overline{\delta}_i(t), \underline{\delta}_i \leq \delta_i(t) \leq \overline{\delta}_i$$

The uncertain system is quadratically stable if and only if there exists $P \in \mathbb{S}^n$, where $P > 0$, such that

$$(A_0 + \Delta A(\delta(t))x(t))^T + P(A_0 + \Delta A(\delta(t))x(t)) < 0 \quad \underline{\delta}_i(t) \in \underline{\delta}_i, \overline{\delta}_i, i = 1, 2, \dots, k.$$

21.3.2 Case 2: Polytope

Consider the case where the set of perturbation parameters is defined by a polytope as

$$\Delta = \delta(t) = [\delta_1(t) \delta_2(t) \dots \delta_k(t)] \in \mathbb{R}^k \mid \delta_i(t) \in \mathbb{R}_{\geq 0}, \sum_{i=1}^k \delta_i(t) = 1$$

The uncertain system is quadratically stable if and only if there exists $P \in \mathbb{S}^n$, where $P > 0$, such that

$$(A_0 + A_i)^T P + P(A_0 + A_i) < 0, i = 1, 2, \dots, k.$$

21.4 Conclusion:

If feasible, System is Quadratically stable for any $x \in \mathbb{R}^n$

21.5 Implementation

<https://github.com/Ricky-10/coding107/blob/master/PolytopicUncertainties>

21.6 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.

¹ http://control.asu.edu/MAE598_frame.htm

² <https://arxiv.org/abs/1903.08599/>

22 Exterior Conic Sector Lemma

22.1 The Concept

The conic sector theorem is a powerful input-output stability analysis tool, providing a fine balance between generality and simplicity of system characterisations that is conducive to practical stability analysis and robust controller synthesis.

22.2 The System

Consider a square, continuous-time linear time-invariant (LTI) system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization $(\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$, where $\mathcal{E}, \mathcal{A} \in \mathcal{R}^{n \times n}$, $\mathcal{B} \in \mathcal{R}^{n \times m}$, $\mathcal{C} \in \mathcal{R}^{p \times n}$, and $\mathcal{D} \in \mathcal{R}^{p \times m}$.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

22.3 The Data

The matrices A, B, C and D

22.4 LMI : Exterior Conic Sector Lemma

The system \mathcal{G} is in the exterior cone of radius r centered at c (i.e. $\mathcal{G} \in \text{excone}_r(c)$), where $r \in \mathcal{R}_{>0}$ and $c \in \mathcal{R}$, under either of the following equivalent necessary and sufficient conditions.

1. There exists $\mathbf{P} \in \mathcal{S}^n$, where $\mathbf{P} \geq 0$, such that

$$\begin{bmatrix} PA + A^T P - C^T C & PB - C^T (D - CI) \\ (PB - C^T (D - CI))^T & r^2 I - (D - cI)^T (D - cI) \end{bmatrix} \leq 0.$$

2. There exists $\mathbf{P} \in \mathcal{S}^n$, where $\mathbf{P} \geq 0$, such that

$$\begin{bmatrix} PA + A^T P - C^T C & PB - C^T (D - CI) & 0 \\ (PB - C^T (D - CI))^T & -(D - cI)^T (D - cI) & rI \\ 0 & (rI)^T & -I \end{bmatrix} \leq 0.$$

Proof, Applying the Schur complement lemma to the $r^2 I$ terms in (1) gives (2).

22.5 Conclusion:

If there exist a positive definite P matrix satisfying above LMIs then the system \mathcal{G} is in the exterior cone of radius r centered at c .

22.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

22.7 Related LMIs

KYP Lemma¹

State Space Stability²

22.8 References

1. J. C. Willems, “Dissipative dynamical systems - part I: General theory,” Archive Rational Mechanics and Analysis, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, “The stability of nonlinear dissipative systems,” IEEE Transactions on Automatic Control, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Bridgeman, Leila Jasmine, and James Richard Forbes. ”The exterior conic sector lemma.” International Journal of Control 88.11 (2015): 2250-2263.

1 https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

2 https://en.wikipedia.org/wiki/State-Space_Stability

23 Modified Exterior Conic Sector Lemma

23.1 The Concept

The conic sector theorem is a powerful input-output stability analysis tool, providing a fine balance between generality and simplicity of system characterisations that is conducive to practical stability analysis and robust controller synthesis.

23.2 The System

Consider a square, continuous-time linear time-invariant (LTI) system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization $(\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$, where $\mathcal{E}, \mathcal{A} \in \mathcal{R}^{n \times n}$, $\mathcal{B} \in \mathcal{R}^{n \times m}$, $\mathcal{C} \in \mathcal{R}^{p \times n}$, and $\mathcal{D} \in \mathcal{R}^{p \times m}$.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

23.3 The Data

The matrices A, B, C and D

23.4 LMI : Modified Exterior Conic Sector Lemma

The system \mathcal{G} is in the exterior cone of radius r centered at c (i.e. $\mathcal{G} \in \text{excone}_r(c)$), where $r \in \mathcal{R}_{>0}$ and $c \in \mathcal{R}$, under either of the following sufficient conditions.

1. There exists $\mathbf{P} \in \mathcal{S}^n$, where $\mathbf{P} \geq 0$, such that

$$\begin{bmatrix} PA + A^T P & PB - C^T(D - CI) \\ (PB - C^T(D - CI))^T & r^2 I - (D - cI)^T(D - cI) \end{bmatrix} \leq 0.$$

Proof. The term $-C^T C$ in the Actual Exterior Conic Sector Lemma¹ makes the matrix inequality more negative definite.

Therefore,

¹ https://en.wikipedia.org/wiki/Exterior_Conic_Sector_Lemma

$$\begin{bmatrix} PA + A^T P - C^T C & PB - C^T (D - CI) \\ (PB - C^T (D - CI))^T & r^2 I - (D - cI)^T (D - cI) \end{bmatrix} \leq \begin{bmatrix} PA + A^T P & PB - C^T (D - CI) \\ (PB - C^T (D - CI))^T & r^2 I - (D - cI)^T (D - cI) \end{bmatrix}$$

2. There exists $\mathbf{P} \in \mathcal{S}^n$, where $\mathbf{P} \geq 0$, such that

$$\begin{bmatrix} PA + A^T P & PB - C^T (D - CI) & 0 \\ (PB - C^T (D - CI))^T & -(D - cI)^T (D - cI) & rI \\ 0 & (rI)^T & -I \end{bmatrix} \leq 0.$$

Proof. Applying the Schur complement lemma to the $r^2 I$ terms in (1) gives (2).

23.5 Conclusion:

If there exist a positive definite P matrix satisfying above LMIs then the system \mathcal{G} is in the exterior cone of radius r centered at c .

23.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

23.7 Related LMIs

KYP Lemma²

State Space Stability³

Exterior Conic Sector Lemma⁴

23.8 References

1. J. C. Willems, “Dissipative dynamical systems - part I: General theory,” Archive Rational Mechanics and Analysis, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, “The stability of nonlinear dissipative systems,” IEEE Transactions on Automatic Control, vol. 21, no. 5, pp. 708–711, 1976.

2 https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

3 https://en.wikipedia.org/wiki/State-Space_Stability

4 https://en.wikipedia.org/wiki/Exterior_Conic_Sector_Lemma

3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²

4. Bridgeman, Leila Jasmine, and James Richard Forbes. "The exterior conic sector lemma." International Journal of Control 88.11 (2015): 2250-2263.

24 DC Gain of a Transfer Matrix

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The continuous-time DC gain is the transfer function value at the frequency $s = 0$.

24.1 The System

Consider a square continuous time Linear Time invariant system, with the state space realization (A, B, C, D) and $\gamma \in \mathbb{R}_{>0}$

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y &= Cx(t) + Du(t)\end{aligned}$$

24.2 The Data

$$A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, C \in \mathbb{R}^{p \times n}, D \in \mathbb{R}^{p \times m}$$

24.3 The LMI: LMI for DC Gain of a Transfer Matrix

The transfer matrix is given by $G(s) = C(sI - A)^{-1}B + D$

The DC Gain of the system is strictly less than γ if the following LMIs are satisfied.

$$\begin{bmatrix} \gamma I & -CA^{-1}B + D \\ (-CA^{-1}B + D)' & \gamma I \end{bmatrix} > 0$$

OR

$$\begin{bmatrix} \gamma I & -B^T A^{-T} C^T + DT \\ (-B^T A^{-T} C^T + DT)' & \gamma I \end{bmatrix} > 0$$

24.4 Conclusion

The DC Gain of the continuous-time LTI system, whose state space realization is give by (A, B, C, D) , is

$$K = D - CA^{-1}B$$

- Upon implementation we can see that the value of γ obtained from the LMI approach and the value of K obtained from the above formula are the same

24.5 Implementation

A link to the Matlab code for a simple implementation of this problem in the Github repository:

https://github.com/yashgvd/LMI_wikibooks

24.6 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- <https://www.mathworks.com/help/control/ref/dcgain.html> -Mathworks reference to DC Gain

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599/>

25 Discrete Time H2 Norm

Discrete-Time H2 Norm

A discrete time system operates on a discrete time signal input and produces a discrete time signal output. They are used in digital signal processing, such as digital filters for images or sound. The class of discrete time systems that are both linear and time invariant, known as discrete time LTI systems.

Discrete-Time LTI systems' H2 norm can be found by solving a LMI.

25.1 The System

Discrete-Time LTI System with state space realization (A_d, B_d, C_d, D_d)

$$A_d \in \mathbf{R}^{n \times n}, \quad B_d \in \mathbf{R}^{n \times m}, \quad C_d \in \mathbf{R}^{p \times n}, \quad D_d \in \mathbf{R}^{p \times m}$$

25.2 The Data

The matrices: System $(A_d, B_d, C_d, D_d), P, Z$.

25.3 The Optimization Problem

The following feasibility problem should be optimized:

μ is minimized while obeying the LMI constraints.

25.4 The LMI:

Discrete-Time Bounded Real Lemma

The LMI formulation

H2 norm $< \mu$

$$P \in S^n; Z \in S^p; \mu \in R_{>0}$$

$$P > 0, \quad Z > 0$$

$$\begin{bmatrix} P & A_d P & B_d \\ * & P & 0 \\ * & * & I \end{bmatrix} > 0,$$

$$\begin{bmatrix} Z & C_d P \\ * & P \end{bmatrix} > 0,$$

$$trZ < \mu^2$$

25.5 Conclusion:

The H2 norm is the minimum value of $\mu \in R_{>0}$ that satisfies the LMI condition.

25.6 Implementation

A link to CodeOcean or other online implementation of the LMI
MATLAB Code¹

25.7 Related LMIs

https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_System_H2_Norm -
Continuous time H2 norm.

25.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.
- <https://www.mathworks.com/help/control/ref/norm.html> - MATLAB function for the H2 norm.

1 <https://github.com/Harishankar-Prabhakaran/LMIs/blob/master/A3.m>

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599/>

4 <https://web.stanford.edu/~boyd/lmibook/>

26 Discrete Time Minimum Gain Lemma

26.1 The Concept

The output of the system $y(t)$ is fed back through a sensor measurement F to a comparison with the reference value $r(t)$. The controller C then takes the error e (difference) between the reference and the output to change the inputs u to the system under control P . This is shown in the figure. This kind of controller is a closed-loop controller or feedback controller.

This is called a single-input-single-output (*SISO*) control system; *MIMO* (i.e., Multi-Input-Multi-Output) systems, with more than one input/output, are common. In such cases variables are represented through vectors instead of simple scalar values. For some distributed parameter systems the vectors may be infinite-dimensional (typically functions).

If we assume the controller C , the plant P , and the sensor F are linear and time-invariant (i.e., elements of their transfer function $C(s)$, $P(s)$, and $F(s)$ do not depend on time), the systems above can be analysed using the Laplace transform on the variables. This gives the following relations:

$$Y(s) = P(s)U(s)$$

$$U(s) = C(s)E(s)$$

$$E(s) = R(s) - F(s)Y(s).$$

Solving for $Y(s)$ in terms of $R(s)$ gives

$$Y(s) = \left(\frac{P(s)C(s)}{1 + P(s)C(s)F(s)} \right) R(s) = H(s)R(s).$$

The expression $H(s) = \frac{P(s)C(s)}{1 + P(s)C(s)F(s)}$ is referred to as the *closed-loop transfer function* of the system. The numerator is the forward (open-loop) gain from r to y , and the denominator is one plus the gain in going around the feedback loop, the so-called loop gain. If $|P(s)C(s)| \gg 1$, i.e., it has a large norm with each value of s , and if $|F(s)| \approx 1$, then $Y(s)$ is approximately equal to $R(s)$ and the output closely tracks the reference input. This page gives an LMI to reduce the gain so that the output closely tracks the reference input.

26.2 The System

Consider a discrete-time LTI system, $\mathcal{G} : \mathbb{Z}_{2e} \rightarrow \mathbb{Z}_{2e}$, with minimal state-space realization $(\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d, \mathcal{D}_d)$, where $\mathcal{A}_d \in \mathbb{R}^{n \times n}$, $\mathcal{B}_d \in \mathbb{R}^{n \times m}$, $\mathcal{C}_d \in \mathbb{R}^{p \times n}$, and $\mathcal{D}_d \in \mathbb{R}^{p \times m}$.

$$x(k+1) = \mathcal{A}_d x(k) + \mathcal{B}_d u(k)$$

$$y(k) = \mathcal{C}_d x(k) + \mathcal{D}_d u(k), k = 0, 1, \dots$$

26.3 The Data

The matrices $\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d$ and \mathcal{D}_d

26.4 LMI : Discrete-Time Minimum Gain Lemma

The system \mathcal{G} has minimum gain γ under any of the following equivalent sufficient conditions.

1. There exists $P \in \mathcal{S}^n$, and $\gamma \in \mathbb{R}_{\geq 0}$ where $P \geq 0$ such that

$$\begin{bmatrix} A_d^T P A_d - P - C_D^T C_D & A_d^T P B_d - C_d^T D_d \\ (A_d^T P B_d - C_d^T D_d)^T & B_d^T P B_d + \gamma^2 I - (D_d^T + D_d) \end{bmatrix} \leq 0.$$

2. There exists $P \in \mathcal{S}^n$, and $\gamma \in \mathbb{R}_{\geq 0}$ where $P \geq 0$ such that

$$\begin{bmatrix} A_d^T P A_d - P - C_D^T C_D & A_d^T P B_d - C_d^T D_d & 0 \\ (A_d^T P B_d - C_d^T D_d)^T & B_d^T P B_d - (D_d^T + D_d) & \gamma I \\ 0 & \gamma I & I \end{bmatrix} \leq 0.$$

proof : Applying the Schur complement lemma to the γ^2 term in equation 1 gives equation 2.

26.5 Conclusion:

If there exist a positive definite P for the system \mathcal{G} , then the minimum gain of the system is γ can be obtained from above defined LMIs.

26.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

26.7 Related LMIs

KYP Lemma¹

State Space Stability²

KYP Lemma without Feedthrough³

26.8 References

1. J. C. Willems, “Dissipative dynamical systems - part I: General theory,” Archive Rational Mechanics and Analysis, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, “The stability of nonlinear dissipative systems,” IEEE Transactions on Automatic Control, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Brogliato B., Maschke B., Lozano R., Egeland O. (2007) Kalman-Yakubovich-Popov Lemma. In: Dissipative Systems Analysis and Control. Communications and Control Engineering. Springer, London

1 https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

2 https://en.wikipedia.org/wiki/State-Space_Stability

3 https://en.wikipedia.org/wiki/KYP_Lemma_without_Feedthrough

27 Modified Discrete Time Minimum Gain Lemma

27.1 The Concept

The output of the system $y(t)$ is fed back through a sensor measurement F to a comparison with the reference value $r(t)$. The controller C then takes the error e (difference) between the reference and the output to change the inputs u to the system under control P . This is shown in the figure. This kind of controller is a closed-loop controller or feedback controller.

This is called a single-input-single-output (*SISO*) control system; *MIMO* (i.e., Multi-Input-Multi-Output) systems, with more than one input/output, are common. In such cases variables are represented through vectors instead of simple scalar values. For some distributed parameter systems the vectors may be infinite-dimensional (typically functions).

If we assume the controller C , the plant P , and the sensor F are linear and time-invariant (i.e., elements of their transfer function $C(s)$, $P(s)$, and $F(s)$ do not depend on time), the systems above can be analysed using the Laplace transform on the variables. This gives the following relations:

$$Y(s) = P(s)U(s)$$

$$U(s) = C(s)E(s)$$

$$E(s) = R(s) - F(s)Y(s).$$

Solving for $Y(s)$ in terms of $R(s)$ gives

$$Y(s) = \left(\frac{P(s)C(s)}{1 + P(s)C(s)F(s)} \right) R(s) = H(s)R(s).$$

The expression $H(s) = \frac{P(s)C(s)}{1 + P(s)C(s)F(s)}$ is referred to as the *closed-loop transfer function* of the system. The numerator is the forward (open-loop) gain from r to y , and the denominator is one plus the gain in going around the feedback loop, the so-called loop gain. If $|P(s)C(s)| \gg 1$, i.e., it has a large norm with each value of s , and if $|F(s)| \approx 1$, then $Y(s)$ is approximately equal to $R(s)$ and the output closely tracks the reference input. This page gives an LMI to reduce the gain so that the ouput closely tracks the reference input.

27.2 The System

Consider a discrete-time LTI system, $\mathcal{G} : \mathbb{Z}_{2e} \rightarrow \mathbb{Z}_{2e}$, with minimal state-space realization $(\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d, \mathcal{D}_d)$, where $\mathcal{A}_d \in \mathbb{R}^{n \times n}$, $\mathcal{B}_d \in \mathbb{R}^{n \times m}$, $\mathcal{C}_d \in \mathbb{R}^{p \times n}$, and $\mathcal{D}_d \in \mathbb{R}^{p \times m}$.

$$x(k+1) = \mathcal{A}_d x(k) + \mathcal{B}_d u(k)$$

$$y(k) = \mathcal{C}_d x(k) + \mathcal{D}_d u(k), k = 0, 1, \dots$$

27.3 The Data

The matrices $\mathcal{A}_d, \mathcal{B}_d, \mathcal{C}_d$ and \mathcal{D}_d

27.4 LMI : Discrete-Time Modified Minimum Gain Lemma

The system \mathcal{G} has minimum gain γ under any of the following equivalent sufficient conditions.

1. There exists $P \in \mathcal{S}^n$, and $\gamma \in \mathbb{R}_{\geq 0}$ where $P \geq 0$ such that

$$\begin{bmatrix} A_d^T P A_d - P & A_d^T P B_d - C_d^T D_d \\ (A_d^T P B_d - C_d^T D_d)^T & B_d^T P B_d + \gamma^2 I - (D_d^T + D_d) \end{bmatrix}$$

Proof. The term $-C_d^T C_d$ in Discrete Time Minimum Gain Lemma¹ makes the matrix inequality more negative definite. Therefore,

$$\begin{bmatrix} A_d^T P A_d - P - C_D^T C_D & A_d^T P B_d - C_d^T D_d \\ (A_d^T P B_d - C_d^T D_d)^T & B_d^T P B_d + \gamma^2 I - (D_d^T + D_d) \end{bmatrix} \leq \begin{bmatrix} A_d^T P A_d - P & A_d^T P B_d - C_d^T D_d \\ (A_d^T P B_d - C_d^T D_d)^T & B_d^T P B_d + \gamma^2 I - (D_d^T + D_d) \end{bmatrix}$$

2. There exists $P \in \mathcal{S}^n$, and $\gamma \in \mathbb{R}_{\geq 0}$ where $P \geq 0$ such that

$$\begin{bmatrix} A_d^T P A_d - P & A_d^T P B_d - C_d^T D_d & 0 \\ (A_d^T P B_d - C_d^T D_d)^T & B_d^T P B_d - (D_d^T + D_d) & \gamma I \\ 0 & \gamma I & I \end{bmatrix} \leq 0.$$

proof : Applying the Schur complement lemma to the γ^2 term in equation 1 gives equation 2.

27.5 Conclusion:

If there exist a positive definite P for the system \mathcal{G} , then the minimum gain of the system γ can be obtained from above defined LMIs.

¹ https://en.wikipedia.org/wiki/Discrete_Time_Minimum_Gain_Lemma

27.6 Implementation

Code for implementation of this LMI using MATLAB. <https://github.com/VJanand25/LMI>

27.7 Related LMIs

KYP Lemma²

State Space Stability³

KYP Lemma without Feedthrough⁴

Discrete Time Minimum Gain Lemma⁵

27.8 References

1. J. C. Willems, "Dissipative dynamical systems - part I: General theory," *Archive Rational Mechanics and Analysis*, vol. 45, no. 5, pp. 321–351, 1972.
2. D. J. Hill and P. J. Moylan, "The stability of nonlinear dissipative systems," *IEEE Transactions on Automatic Control*, vol. 21, no. 5, pp. 708–711, 1976.
3. LMI Properties and Applications in Systems, Stability, and Control Theory, by Ryan James Caverly¹ and James Richard Forbes²
4. Brogliato B., Maschke B., Lozano R., Egeland O. (2007) Kalman-Yakubovich-Popov Lemma. In: Dissipative Systems Analysis and Control. Communications and Control Engineering. Springer, London

² https://en.wikipedia.org/wiki/Kalman%25E2%2580%2593Yakubovich%25E2%2580%2593Popov_lemma

³ https://en.wikipedia.org/wiki/State-Space_Stability

⁴ https://en.wikipedia.org/wiki/KYP_Lemma_without_Feedthrough

⁵ https://en.wikipedia.org/wiki/Discrete_Time_Minimum_Gain_Lemma

28 Discrete-Time Algebraic Riccati Equation

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28.1 The System

Consider a Discrete-Time LTI system

$$x_{k+1} = A_d x_k + B_d u_k$$

$$y_k = C_d x_k$$

Consider $A_d \in \mathbb{R}^{nxn}$; $B_d \in \mathbb{R}^{nxm}$

28.2 The Data

The Matrices A_d , B_d , C_d , Q , R are given

Q and R should necessarily be **Hermitian** Matrices.

A square matrix is Hermitian if it is equal to its complex conjugate transpose.

28.3 The Optimization Problem

Our aim is to find

P - Unique solution to the discrete-time algebraic Riccati equation, returned as a matrix.

K - State-feedback gain, returned as a matrix.

The algorithm used to evaluate the State-feedback gain is given by

$$K = (R + B_d^T P B_d)^{-1} B_d^T P A_d$$

L - Closed-loop eigenvalues, returned as a matrix.

28.4 The LMI: Discrete-Time Algebraic Riccati Inequality (DARE)

An algebraic Riccati equation is a type of nonlinear equation that arises in the context of infinite-horizon optimal control problems in continuous time or discrete time

The Discrete-Time Algebraic Riccati Inequality is given by

$$A_d^T P A_d - A_d^T P B_d (R + B_d^T P B_d)^{-1} B_d^T P A_d + Q - P \geq 0$$

$$P, Q \in \mathbb{S}^n \text{ and } R \in \mathbb{S}^m \text{ where } P > 0, Q \geq 0, R > 0$$

P is the unknown n by n symmetric matrix and A, B, Q, R are known real coefficient matrices.

The above equation can be rewritten using the Schur Complement Lemma as :

$$\begin{bmatrix} A_d^T P A_d - P + Q & A_d^T P B_d \\ B_d^T P A_d & R + B_d^T P B_d \end{bmatrix} \geq 0$$

28.5 Conclusion:

Algebraic Riccati Inequalities play a key role in LQR/LQG control, H₂- and H_∞ control and Kalman filtering. We try to find the unique stabilizing solution, if it exists. A solution is stabilizing, if controller of the system makes the closed loop system stable.

Equivalently, this Discrete-Time algebraic Riccati Inequality is satisfied under the following *necessary and sufficient condition*:

$$\begin{bmatrix} Q & 0 & A_d^T P & P \\ 0 & R & B_d^T P & 0 \\ P A_d & P B_d & P & 0 \\ P & 0 & 0 & P \end{bmatrix} \geq 0$$

28.6 Implementation

(X in the output corresponds to P in the LMI)

A link to the Matlab code for a simple implementation of this problem in the Github repository:

https://github.com/yashgvd/LMI_wikibooks

28.7 Related LMIs

LMI for Continuous-Time Algebraic Riccati Inequality

LMI for Schur Stabilization

28.8 External Links

A list of references documenting and validating the LMI.

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

29 Deduced LMI Conditions for H-infinity Index

H-infinity Index Deduced LMI

Although the KYP Lemma, also known as the Bounded Real Lemma¹, is a basic condition to evaluate an upper bound on the H_∞ , the verification of the bound on the H_∞ -gain of the system can be done via the deduced condition.

29.1 The System

A state-space representation of a linear system as given below:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bw(t) \\ y(t) &= Cx(t) + Dw(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and $w(t) \in \mathbb{R}^r$ are the system state, output, and the disturbance vector respectively. The transfer function of such a system can be evaluated as:

$$G(s) = C(sI - A)^{-1}B + D$$

29.2 The Data

Number of states n , number of outputs m and number of external noise channels r need to be known. Moreover, the system matrices A, B, C, D are also required to be known.

29.3 The Feasibility LMI

For an arbitrary γ , the transfer function $G(s)$ satisfies

$$\|G(s)\|_\infty < \gamma$$

¹ [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

if and only if there exists a symmetric matrix $X > 0$ and a matrix Ω , such that:

$$\begin{aligned} \text{Find } X, \Omega : \\ \Theta + \Phi^\top \Omega \Psi + \Psi^\top \Omega^\top \Phi < 0 \end{aligned}$$

where:

$$\begin{aligned} \Theta &= \begin{bmatrix} 0 & X & 0 & 0 & 0 \\ X & -X & 0 & 0 & 0 \\ 0 & 0 & -\gamma I_m & 0 & 0 \\ 0 & 0 & 0 & -X & 0 \\ 0 & 0 & 0 & 0 & \gamma I_r \end{bmatrix} \\ \Phi &= \begin{bmatrix} -I_n & A^\top & C^\top & I_n & 0 \\ 0 & B^\top & D^\top & 0 & -\gamma I_r \end{bmatrix} \\ \Psi &= \begin{bmatrix} I_n & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & I_r \end{bmatrix} \end{aligned}$$

The above LMI can be combined with the bisection method to find minimum γ to find the minimum upper bound on the H_∞ gain of $G(s)$.

29.4 Conclusion:

If there is a feasible solution to the aforementioned LMI, then the γ upper bounds the infinity norm of the system $G(s)$.

29.5 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/Deduced_hinf_example.m

29.6 Related LMIs

Bounded Real Lemma²

² [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

29.7 External Links

A list of references documenting and validating the LMI.

- LMIs in Control Systems: Analysis, Design and Applications³ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 5.2.2 pp. 153–156.

³ [https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/
Duan-Yu/p/book/9781466582996](https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996)

30 Deduced LMI Conditions for H_2 Index

H_2 Index Deduced LMI

Although there are ways to evaluate an upper bound on the H_2 , the verification of the bound on the H_2 -gain of the system can be done via the deduced condition.

30.1 The System

We consider the generalized Continuous-Time LTI system with the state space realization of (A, B, C, D)

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and $u(t) \in \mathbb{R}^r$ are the system state, output, and the input vectors respectively.

The transfer function of such a system can be evaluated as:

$$G(s) = C(sI - A)^{-1}B + D$$

30.2 The Data

The system matrices A, B, C are known.

30.3 The Optimization Problem

For an arbitrary $\gamma > 0$ (a given scalar), the transfer function satisfies

$$\|C(sI - A)^{-1}B + D\|_2 < \gamma$$

The H_2 -norm condition on Transfer function holds only when the matrix A is stable. And this can be conveniently converted to an LMI problem

if and only if 1. There exists a symmetric matrix $X > 0$ such that:

$$AX + XA^T + BB^T < 0, \text{trace}(CXC^T) < \gamma^2$$

2. There exists a symmetric matrix $Y > 0$ such that:

$$AY + YA^T + C^T C < 0, \text{trace}(B^T Y B) < \gamma^2$$

30.4 The LMI - Deduced Conditions for H2-norm

These deduced condition can be derived from the above equations. According to this

For an arbitrary $\gamma > 0$ (a given scalar), the transfer function satisfies

$$\|G(s)\|_2 < \gamma$$

if and only if there exists symmetric matrices Z and P ; and a matrix V such that

$$\begin{bmatrix} -Z & B^T \\ B & -P \end{bmatrix} < 0$$

$$\begin{bmatrix} -(V+V^T) & V^T A^T + P & V^T C^T & V^T \\ AV + P & -P & 0 & 0 \\ CV & 0 & -I & 0 \\ V & 0 & 0 & -P \end{bmatrix} < 0$$

The above LMI can be combined with the bisection method to find minimum γ to find the minimum upper bound on the H₂ gain of $G(s)$.

30.5 Conclusion:

If there is a feasible solution to the aforementioned LMI, then the γ upper bounds the norm of the system $G(s)$.

30.6 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the problem, and SeDuMi or MOSEK is required to solve the problem. The following link showcases an example of the problem:

<https://github.com/yashgvd/ygovada>

30.7 Related LMIs

Bounded Real Lemma¹

¹ [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

Deduced LMIs for H-infinity index

30.8 External Links

A list of references documenting and validating the LMI.

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599/>

31 Dissipativity of Systems

Dissipativity of Systems

The dissipativity of systems is associated with their supply function. In general, a linear system is dissipative if the accumulated sum (integration) of the supply function is non-negative over all the duration of $T \geq 0$.

31.1 The System

A state-space representation of a linear system as given below:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and $u(t) \in \mathbb{R}^r$ are the system state, output, and the input vector respectively. A , B , C and D are system coefficient matrices of appropriate dimensions. The control input u is restricted to be a piece-wise continuous vector function defined on $[0, \infty)$.

The transfer function of such a system can be evaluated as:

$$G(s) = C(sI - A)^{-1}B + D$$

For such a system, a general quadratic supply function is defined as:

$$s(u, y) = [y \ u] Q \begin{bmatrix} y \\ u \end{bmatrix}$$

where Q is a real symmetric matrix of $(m+r)$ dimensions. Q need not be either symmetric positive or negative definite.

31.2 The Data

Number of states n , number of outputs m and number of control inputs r need to be known. Moreover, the system matrices A, B, C, D are also required to be known. The system should also be controllable.

31.3 The Feasibility LMI

The system $G(s)$ defined can be evaluated to be dissipative with respect to a supply function $s(u, y)$ iff there exist $P \geq 0$ and a Q (defining $s(u, y)$) such that the following is feasible:

$$\begin{aligned} & \text{Find } P, Q : \\ & P \geq 0 \\ & \begin{bmatrix} A^\top P + PA & PB \\ B^\top P & 0 \end{bmatrix} - \begin{bmatrix} C & D \\ 0 & I \end{bmatrix}^\top Q \begin{bmatrix} C & D \\ 0 & I \end{bmatrix} \leq 0. \end{aligned}$$

31.4 Conclusion:

If there is a feasible solution to the aforementioned LMI, then there exists a supply function $s(u, y)$ for which the system $G(s)$ is dissipative. Since the assumption of the system being controllable is required for it to be dissipative, this check can be used of as a sufficient condition to check the controllability of the linear system, just like the feasibility for Lyapunov stability.

31.5 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/Dissipativity_example.m

31.6 Related LMIs

Continuous_Time_Lyapunov_Inequality¹

31.7 External Links

A list of references documenting and validating the LMI.

- LMIs in Control Systems: Analysis, Design and Applications² - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 6.3 pp. 178–184.

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/Lyapunov_Inequality

² <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

32 D-Stabilization

$\mathbb{D}_{(q,r)}$ -Stabilization

There are a wide variety of control design problems that are addressed in a wide variety of different ways. One of the most important control design problem is that of state feedback stabilization. One such state feedback problem, which will be the main focus of this article, is that of $\mathbb{D}_{(q,r)}$ -Stabilization, a form of D-Stabilization where the closed-loop poles are located on the left-half of the complex plane.

32.1 The System

For this particular problem, suppose that we were given a linear system in the form of:

$$\rho x = Ax + Bu,$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, and ρ represents either the differential operator (in the continuous-time case) or the one-step forward operator (for the discrete-time system case). Then the LMI for determining the $\mathbb{D}_{(q,r)}$ -stabilization case would be obtained as described below.

32.2 The Data

In order to obtain the LMI, we need the following 2 matrices: A and B .

32.3 The Optimization Problem

Suppose - for the linear system given above - we were asked to design a state-feedback control law where $u = Kx$ such that the closed-loop system:

$$\rho x = (A + BK)x$$

is $\mathbb{D}_{(q,r)}$ -stable, then the system would be stabilized as follows.

32.4 The LMI: $\mathbb{D}_{(q,r)}$ -Stabilization

From the given pieces of information, it is clear that the optimization problem only has a solution if there exists a matrix W and a symmetric matrix P that satisfies the following:

$$\begin{bmatrix} -rP & qP + AP + BW \\ qP + PA^T + W^T B^T & -rP \end{bmatrix} < 0$$

32.5 Conclusion:

Given the resulting controller matrix $K = WP^{-1}$, it can be observed that the matrix is $\mathbb{D}_{(q,r)}$ -stable.

32.6 Implementation

- Example Code¹ - A GitHub link that contains code (titled "DStability.m") that demonstrates how this LMI can be implemented using MATLAB-YALMIP.

32.7 Related LMIs

- ..//H stabilization/² - Equivalent LMI for $\mathbb{H}_{(\alpha,\beta)}$ -stabilization.
- ..//Continuous Time D-Stability Controller/³ - LMI for deriving a Controller using D-Stability.
- ..//Continuous Time D-Stability Observer/⁴ - LMI for deriving an Observer using D-Stability.

32.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁵ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁶ - A List of LMIs by Ryan Caverly and James Forbes.

1 <https://github.com/aramani3/MAE-598-LMI-Codes>

2 Chapter 33 on page 93

3 <https://en.wikibooks.org/wiki/..%2FContinuous%20Time%20D-Stability%20Controller%2F>

4 <https://en.wikibooks.org/wiki/..%2FContinuous%20Time%20D-Stability%20Observer%2F>

5 http://control.asu.edu/MAE598_frame.htm

6 <https://arxiv.org/abs/1903.08599/>

- LMIs in Systems and Control Theory⁷ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications - A book co-authored by Guang-Ren Duan and Hai-Hua Yu.

⁷ <https://web.stanford.edu/~boyd/lmibook/>

33 H-Stabilization

$\mathbb{H}_{(\alpha,\beta)}$ -Stabilization

There are a wide variety of control design problems that are addressed in a wide variety of different ways. One of the most important control design problem is that of state feedback stabilization. One such state feedback problem, which will be the main focus of this article, is that of $\mathbb{H}_{(\alpha,\beta)}$ -Stabilization, a form of \mathbb{D} -Stabilization where the real components are located on the left-half of the complex plane.

33.1 The System

For this particular problem, suppose that we were given a linear system in the form of:

$$\dot{x} = Ax + Bu,$$

where $x \in \mathbb{R}^n$ and $u \in \mathbb{R}^r$. Then the LMI for determining the $\mathbb{H}_{(\alpha,\beta)}$ -stabilization case would be obtained as described below.

33.2 The Data

In order to obtain the LMI, we need the following 2 matrices: A and B .

33.3 The Optimization Problem

Suppose - for the linear system given above - we were asked to design a state-feedback control law where $u = Kx$ such that the closed-loop system:

$$\dot{x} = (A + BK)x$$

is $\mathbb{H}_{(\alpha,\beta)}$ stable, then the system would be stabilized as follows.

33.4 The LMI: $\mathbb{H}_{(\alpha,\beta)}$ -Stabilization

From the given pieces of information, it is clear that the optimization problem only has a solution if there exists a matrix W and a symmetric matrix $P > 0$ that satisfy the following:

$$\begin{cases} AP + PA^T + BW + W^T B^T + 2\alpha P & < 0 \\ -AP - PA^T - BW - W^T B^T - 2\beta P & < 0 \end{cases}$$

33.5 Conclusion:

Given the resulting controller matrix $K = WX^{-1}$, it can be observed that the matrix is $\mathbb{H}_{(\alpha,\beta)}$ -stable.

33.6 Implementation

- Example Code¹ - A GitHub link that contains code (titled "HStability.m") that demonstrates how this LMI can be implemented using MATLAB-YALMIP.

33.7 Related LMIs

- ..//D stabilization/² - Equivalent LMI for $\mathbb{D}_{(q,r)}$ -stabilization.

33.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications - A book co-authored by Guang-Ren Duan and Hai-Hua Yu.

1 <https://github.com/aramani3/MAE-598-LMI-Codes>

2 Chapter 32 on page 89

3 http://control.asu.edu/MAE598_frame.htm

4 <https://arxiv.org/abs/1903.08599/>

5 <https://web.stanford.edu/~boyd/lmibook/>

34 H-2 Norm of the System

H₂-norm of System

The H_2 -norm is conceptually identical to the Frobenius (aka Euclidean) norm on a matrix. It can be used to determine whether the system **representation** can be reduced to its simplest form, thereby allowing its use in performing effective block-diagram algebra.

34.1 The System

Suppose we define the **state-space** system $G : L_2 \rightarrow L_2$ by $y = Gu$ if:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

where $A \in \mathbb{R}^{mxm}$, $B \in \mathbb{R}^{mxn}$, $C \in \mathbb{R}^{pxm}$, and $D \in \mathbb{R}^{qxn}$ for any $t \in \mathbb{R}$. Then the H_2 -norm of the system can be determined as described below.

34.2 The Data

In order to determine the H_2 -norm of the system, we need the matrices A , B , and C .

34.3 The Optimization Problem

Suppose we wanted to infer properties of the system behaviour (which is represented in the form **(A,B,C,D)**). Then it becomes necessary to ensure that the overall system forms an algebra, as the standard use of block-diagram algebra would otherwise be invalid. The only way this is possible is by calculating H_2 and/or H_∞ -norms - both of which are signal norms that (in a certain sense) measure the size of the transfer function.

34.4 The LMI: The H_2 Norm

Assuming that $\hat{P}(s) = C(sI - A)^{-1}B$, this means that the following are equivalent:

1) A is Hurwitz and $\|\hat{P}\|_{H_2}^2 < \gamma$

$$2) \begin{cases} \text{trace}(CXC^T) & < \gamma \\ AX + XA^T + BB^T & < 0 \\ X & > 0 \end{cases}$$

34.5 Conclusion:

The LMI can be used to minimize the H_2 -norm of the system. It is worth noting that a finite H_2 -norm does not guarantee finite H_∞ -norm, and that in order for the block diagram algebra to be valid, H_∞ -norm must be finite.

34.6 Implementation

- Example Code¹ - A GitHub link that contains code (titled "H2Norm.m") that demonstrates how this LMI can be implemented using MATLAB-YALMIP.

34.7 Related LMIs

- H_2 -Filtering² - LMI for H_2 -Filtering
- Discrete-Time H_2 Norm³ - LMI for H_2 -norm in the Discrete-Time case.

34.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁵ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁶ - A downloadable book on LMIs by Stephen Boyd.

1 <https://github.com/aramani3/MAE-598-LMI-Codes>

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H-2_filtering

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete-Time_H2_Norm/

4 http://control.asu.edu/MAE598_frame.htm

5 <https://arxiv.org/abs/1903.08599/>

6 <https://web.stanford.edu/~boyd/lmibook/>

35 Algebraic Riccati Equation

Algebraic Riccati Equations are particularly significant in Optimal Control, filtering and estimation problems. The need to solve such equations is common in the analysis and linear quadratic Gaussian control along with general Control problems. In one form or the other, Riccati Equations play significant roles in optimal control of multivariable and large-scale systems, scattering theory, estimation, and detection processes. In addition, closed forms solution of Riccati Equations are intractable for two reasons namely; one, they are nonlinear and two, are in matrix forms.

35.1 The System

$$\dot{x}(t) = Ax(t) + Bu(t)$$

35.2 The Data

Following matrices are needed as Inputs:

$$A, B, N$$

35.3 The Optimization Problem

In control systems theory, many analysis and design problems are closely related to Riccati algebraic equations or inequalities. Find

35.4 The LMI: Algebraic Riccati Inequality

Title and mathematical description of the LMI formulation.

$$\begin{aligned} P &> 0, \\ Q &> 0, \\ R &> 0. \end{aligned}$$

The Algebraic Riccati inequality is given by

$$A^T P + PA - (PB + N^T)R^{-1}(B^T P + N) + Q \geq 0$$

can be written using the Schur complement lemma as

$$\begin{bmatrix} A^T P + PA + Q & PB + N^T \\ * & R \end{bmatrix} \geq 0$$

35.5 Conclusion:

If the solution exists, LMIs give a unique, stabilizing, symmetric matrix P.

35.6 Implementation:

Matlab code for this LMI in the Github repository:

1. REDIRECT https://github.com/Ricky-10/coding107/blob/master/LMI_Algebraic_Riccati_Equations¹ https://- CODE

35.7 External links

- <https://www.intechopen.com/books/matlab-a-fundamental-tool-for-scientific-computing-and-engineering-applications-volume-3/optimal-solution-to-matrix-riccati-equation-for-kalman-filter-implementation>-Optimal Solution to Matrix Riccati Equation
- <https://arxiv.org/abs/1903.08599>/ LMI Properties and Applications in Systems, Stability, and Control Theory.²- - A List of LMIs by Ryan Caverly and James Forbes

¹ Chapter 1 on page 3

² <https://arxiv.org/abs/1903.08599>%20LMI%20Properties%20and%20Applications%20in%20Systems,%20Stability,%20and%20Control%20Theory.

36 System Zeros without feedthrough

Let's say we have a transfer function defined as a ratio of two polynomials:
 $H(s) = \frac{N(s)}{D(s)}$ Zeros are the roots of $N(s)$ (the numerator of the transfer function)
obtained by setting $N(s) = 0$ and solving for s . The values of the poles and the zeros of a system determine whether the system is stable, and how well the system performs. Similarly, the system zeros are either real or appear in complex conjugate pairs. In the case of system zeros without feedthrough, we take the assumption that $D = 0$.

36.1 The System

Consider a continuous-time LTI system, G , with minimal statespace representation $(A, B, C, 0)$

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

36.2 The Data

The matrices:

$$A \in \mathbb{R}^{n \times n}$$

$$M \in \mathbb{R}^{n \times q}$$

$$N \in \mathbb{R}^{q \times n}$$

36.3 The LMI: System Zeros without feedthrough

The transmission zeros of $G(s) = C(sI - A)^{-1}B$ are the eigenvalues of NAM , where $N \in \mathbb{R}^{q \times n}, M \in \mathbb{R}^{n \times q}, CM = 0, NB = 0, NM = 1..$. Therefore, $G(s)$ is a minimum phase if and only if there exists $P \in \mathbb{S}^q$, where $P > 0$ such that

$$PNAM + M^T A^T N^T P < 0$$

36.4 Conclusion:

If P exists, it ensures non-minimum phase. Eigenvalues of NAM then gives the zeros of the system.

36.5 Implementation

[https://github.com/Ricky-10/coding107/blob/master/
Systemzeroswithoutfeedthrough](https://github.com/Ricky-10/coding107/blob/master/Systemzeroswithoutfeedthrough)

36.6 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.
- Control_Systems/Poles_and_Zeros⁴

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599/>

3 <https://web.stanford.edu/~boyd/lmibook/>

4 https://en.wikibooks.org/wiki/Control_Systems%2FPoles_and_Zeros

37 System zeros with feedthrough

Let's say we have a transfer function defined as a ratio of two polynomials:
 $H(s) = \frac{N(s)}{D(s)}$ Zeros are the roots of $N(s)$ (the numerator of the transfer function)
obtained by setting $N(s) = 0$ and solving for s . The values of the poles and the zeros of a system determine whether the system is stable, and how well the system performs. Similarly, the system zeros are either real or appear in complex conjugate pairs. In the case of system zeros with feedthrough, we take D as full rank.

37.1 The System

Consider a continuous-time LTI system, G , with minimal statespace representation (A, B, C, D)

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du\end{aligned}$$

37.2 The Data

The matrices needed as inputs are:

$$\begin{aligned}A &\in \mathbb{R}^{n \times n} \\ B &\in \mathbb{R}^{n \times m} \\ N &\in \mathbb{R}^{p \times n}\end{aligned}$$

In this case, $m \leq p$

37.3 The LMI: System Zeros with feedthrough

The transmission zeros of $G(s) = C(sI - A)^{-1}B + D$ are the eigenvalues of $A - B(D^T D)^{-1}D^T C$. Therefore, $G(s)$ is a minimum phase if and only if there exists $P \in \mathbb{S}^q$, where $P > 0$ such that

$$P(A - B(D^T D)^{-1}D^T C) + (A - B(D^T D)^{-1}D^T C)^T P < 0$$

37.4 Conclusion:

If P exists, it ensures non-minimum phase. Eigenvalues of $A - B(D^T D)^{-1} D^T C$ then gives the zeros of the system.

37.5 Related LMIs

LMIs_in_Controls/pages/systemzeroswithoutfeedthrough¹

37.6 Implementation

[https://github.com/Ricky-10/coding107/blob/master/
systemzeroswithfeedthrough](https://github.com/Ricky-10/coding107/blob/master/systemzeroswithfeedthrough)

37.7 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.
- Control_Systems/Poles_and_Zeros⁵

1 https://en.wikibooks.org/wiki/LMIs_in_Controls%2Fpages%2Fsystemzeroswithoutfeedthrough
2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599/>
4 <https://web.stanford.edu/~boyd/lmibook/>
5 https://en.wikibooks.org/wiki/Control_Systems%2FPoles_and_Zeros

38 Negative Imaginary Lemma

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Positive real systems are often related to systems involving energy dissipation. But the standard Positive real theory will not be helpful in establishing closed-loop stability. However transfer functions of systems with a degree more than one can be satisfied with the negative imaginary conditions for all frequency values and such systems are called "systems with negative imaginary frequency response" or "negative imaginary systems".

38.1 The System

Consider a square continuous time Linear Time invariant system, with the state space realization (A, B, C, D)

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y &= Cx(t) + Du(t)\end{aligned}$$

38.2 The Data

$$A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, C \in \mathbb{R}^{m \times n}, D \in \mathbb{S}^m$$

38.3 The LMI: LMI for Negative Imaginary Lemma

According to the Lemma, The aforementioned system is negative imaginary under either of the following equivalent necessary and sufficient conditions

- There exists a $P \in \mathbb{S}^n$, where $P \geq 0$, such that,

$$\begin{bmatrix} A^T P + PA & PB - A^T C^T \\ B^T P - CA & -(CB + B^T C^T) \end{bmatrix} \leq 0$$

- There exists a $Q \in \mathbb{S}^n$, where $Q \geq 0$, such that,

$$\begin{bmatrix} QA^T + AQ & B - QA^T C^T \\ B^T - CAQ & -(CB + B^T C^T) \end{bmatrix}$$

$$\leq 0$$

38.4 Conclusion

The system is strictly negative if $\det(A) \neq 0$ and either of the above LMIs are feasible with resulting $P > 0$ or $Q > 0$

38.5 Implementation

A link to the Matlab code for a simple implementation of this problem in the Github repository:

<https://github.com/yashgvd/ygovada>

38.6 Related LMIs

Positive Real Lemma

38.7 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- Xiong, Junlin & Petersen, Ian & Lanzon, Alexander. (2010). A Negative Imaginary Lemma and the Stability of Interconnections of Linear Negative Imaginary Systems.

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599/>

39 Small Gain Theorem

LMIs in Control/Matrix and LMI Properties and Tools/Small Gain Theorem

The Small Gain Theorem provides a sufficient condition for the stability of a feedback connection.

39.1 Theorem

Suppose B is a Banach Algebra and $Q \in B$. If $\|Q\| < 1$, then $(I - Q)^{-1}$ exists, and furthermore,

$$(I - Q)^{-1} = \sum_{k=0}^{\infty} Q^k$$

39.2 Proof

Assuming we have an interconnected system (G, K) :

$$y_1 = G(u_1 - y_2) \text{ and, } y_2 = K(u_2 - y_1)$$

The above equations can be represented in matrix form as

$$\begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 0 & -G \\ -K & 0 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} + \begin{bmatrix} G & 0 \\ 0 & K \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

Making $\begin{bmatrix} y_1 & y_2 \end{bmatrix}^T$ the subject, we then have:

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} I & G \\ K & I \end{bmatrix}^{-1} \begin{bmatrix} G & 0 \\ 0 & K \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} (I - GK)^{-1}G & -G(I - KG)^{-1}K \\ -K(I - GK)^{-1}G & (I - KG)^{-1}K \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

If $(I - GK)^{-1}$ is well-behaved, then the interconnection is stable. For $(I - GK)^{-1}$ to be well-behaved, $\|(I - GK)^{-1}\|$ must be finite.

Hence, we have $\|(I - GK)^{-1}\| < \infty$

$\|\cdot\| = \|\cdot\|$ and $\|\cdot\| < I$ for the higher exponents of $\|\cdot\|$ to converge to 0.

39.3 Conclusion

If $\|\cdot\| < 1$, then this implies stability, since the higher exponents of Q in the summation of $\sum_{k=0}^{\infty} Q^k$ will converge to 0, instead of blowing up to infinity.

39.4 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://web.stanford.edu/~boyd/lmibook/>

40 Tangential Nevanlinna-Pick Interpolation

40.1 Tangential Nevanlinna-Pick

The Tangential Nevanlinna-Pick arises in multi-input, multi-output (MIMO) control theory, particularly H_∞ robust and optimal control.

The problem is to try and find a function $H : C \rightarrow C^{pxq}$ which is analytic in C_+ and satisfies

$$H(\lambda_i)u_i = v_i,$$

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$$i = 1, \dots, m$$

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$$\text{with } \|H\|_\infty \leq 1$$

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(1)

40.2 The System

$N_{(ij)}$ is a set of pxq matrix valued Nevanlinna functions. The matrix valued function $H(\{\lambda\})$ analytic on the open upper half plane is a Nevanlinna function if $\text{Im}(H(\lambda)) \geq 0$ ($\lambda \in \pi^+$).

40.3 The Data

Given:

Initial sequence of data points on real axis $\lambda_1, \dots, \lambda_m$ with $\lambda_i \in C_+ \equiv (s | \text{Re}(s) > 0)$,
And two sequences of row vectors containing distinct target points u_1, \dots, u_m with
 $u_i \in C^q$, and v_1, \dots, v_m with $v_i \in C^p, i = 1, \dots, m$.

40.4 The LMI:Tangential Nevanlinna- Pick

Problem (1) has a solution if and only if the following is true:

Nevanlinna-Pick Approach

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$$N_{ij} = \frac{u_i^* u_j - v_i^* v_j}{\lambda_i^* + \lambda_j}$$

Lyapunov Approach

N can also be found using the Lyapunov equation:

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$$A^* N + N A - (U^* U - V^* V) = 0$$

where $A = \text{diag}(\lambda_1, \dots, \lambda_m)$, $U = [u_1 \dots u_m]$, $V = [v_1 \dots v_m]$

The tangential Nevanlinna-Pick problem is then solved by confirming that $N \geq 0$.

40.5 Conclusion:

If $N_{(ij)}$ is positive (semi)-definite, then there exists a norm-bounded analytic function, H which satisfies $H(\lambda_i)u_i = v_i$,

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$$i = 1, \dots, m$$

UNKNOWN TEMPLATE spaces

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with $\|H\|_\infty \leq 1$

40.6 Implementation

Implementation requires YALMIP and a linear solver such as sedumi.
https://github.com/nnbeauli/Optimal-Control-LMI/blob/main/Tan_Nevanlinna_Pick.m - MATLAB code for Tangential Nevanlinna-Pick Problem.

40.7 Related LMIs

Nevanlinna-Pick Interpolation with Scaling¹

40.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.
- Generalized Interpolation in H_∞ by Donald Sarason.⁴
- Tangential Nevanlinna-Pick Interpolation Problem With Boundary Nodes in the Nevanlinna Class And The Related Moment Problem by Yong Jian Hu and Xiu Ping Zhang.⁵

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Matrix_and_LMI_Properties_and_Tools/Nevanlinna_Pick_Interpolation_with_Scaling

2 http://control.asu.edu/MAE598_frame.htm

3 <https://web.stanford.edu/~boyd/lmibook/>
<https://www.ams.org/journals/tran/1967-127-02/S0002-9947-1967-0208383-8/S0002-9947-1967-0208383-8.pdf>

5 <http://www.math.ualberta.ca/ijiss/SS-Volume-4-2008/No-1-08/SS-08-01-07.pdf>

41 Nevanlinna-Pick Interpolation with Scaling

41.1 Nevanlinna-Pick Interpolation with Scaling

The Nevanlinna-Pick problem arises in multi-input, multi-output (MIMO) control theory, particularly H_∞ robust and optimal controller synthesis with structured perturbations.

The problem is to try and find $\gamma_{opt} = \inf(\|DHD^{-1}\|_\infty)$ such that H is analytic in C_+ ,

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1em

$D = D^* > 0$, and $D \in \mathbb{D}$ define the scaling, and finally,

$$H(\lambda_i)u_i = v_i$$

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2em

$$i = 1, \dots, m$$

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(1)

41.2 The System

The scaling factor \mathbb{D} is given as a set of $m \times m$ block-diagonal matrices with specified block structure. The matrix valued function $H(\{\lambda\})$ analytic on the open upper half plane is a Nevanlinna function if $\text{Im}(H(\lambda)) \geq 0$ ($\lambda \in \pi^+$). The Nevanlinna LMI matrix N is defined as $N = G_{in} - G_{out}$. The matrix A is a diagonal matrix of the given sequence of data points $\lambda_i \in \mathbb{C}$ ($A = \text{diag}(\lambda_1, \dots, \lambda_m)$)

41.3 The Data

Given:

Initial sequence of data points in the complex plane $\lambda_1, \dots, \lambda_m$ with $\lambda_i \in$

$C_+ \hat{=} (s | Re(s) > 0)$.

Two sequences of row vectors containing distinct target points u_1, \dots, u_m with $u_i \in C^q$, and v_1, \dots, v_m with $v_i \in C^p, i = 1, \dots, m$.

41.4 The LMI: Nevanlinna- Pick Interpolation with Scaling

First, implement a change of variables for $P = D^*D$ and $N = G_{in} - G_{out}$.

From this substitution it can be concluded that γ_{opt} is the smallest positive γ such that there exists a $P > 0, P \in \mathbb{D}$ such that the following is true:

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$$A^*G_{in} + G_{in}A - U^*PU = 0,$$

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$$A^*G_{out} + G_{out}A - V^*PV = 0,$$

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$$\gamma^2 G_{in} - G_{out} \geq 0$$

41.5 Conclusion:

If the LMI constraints are met, then there exists a H_∞ norm-bounded optimal gain γ which satisfies the scaled Nevanlinna-Pick interpolation objective defined above in Problem (1).

41.6 Implementation

Implementation requires YALMIP and Mosek. https://github.com/nbeauli/Optimal-Control-LMI/blob/main/Nevanlinna_Pick_Inter.m - MATLAB code for Nevanlinna-Pick Interpolation.

41.7 Related LMIs

Nevalinna-Pick Interpolation¹

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/Matrix_and_LMI_Properties_and_Tools/Tangential_Nevanlinna_Pick

41.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.
- Generalized Interpolation in H_∞ .⁴

² http://control.asu.edu/MAE598_frame.htm

³ <https://web.stanford.edu/~boyd/lmibook/>

⁴ <https://www.ams.org/journals/tran/1967-127-02/S0002-9947-1967-0208383-8/S0002-9947-1967-0208383-8.pdf>

42 Generalized H_2 Norm

42.1 Generalized H_2 Norm

The H_2 norm characterizes the average frequency response of a system. To find the H_2 norm, the system must be strictly proper, meaning the state space represented D matrix must equal zero. The H_2 norm is frequently used in optimal control to design a stabilizing controller which minimizes the average value of the transfer function, G as much as possible. This optimal control problem is also called the Linear Quadratic Gaussian.

42.2 The System

Consider a continuous-time, linear, time-invariant system $G : L_{2e} \rightarrow L_{2e}$ with state space realization $(A, B, C, 0)$ where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $C \in \mathbb{R}^{p \times n}$, and A is Hurwitz. The generalized H_2 norm of G is:

$$\|G\|_{2,\infty} = \sup_{u \in L_2, u \neq \text{zero}} \frac{\|Gu\|_\infty}{\|u\|_2}$$

42.3 The Data

The transfer function G , and system matrices A, B, C are known and A is Hurwitz.

42.4 The LMI: Generalized H_2 Norm LMIs

The inequality $\|G\|_{2,\infty} < \mu$ holds under the following conditions:

1. There exists $P \in \mathbb{S}^n$ and $\mu \in \mathbb{R}_{>0}$ where $P > 0$ such that:

$$\begin{bmatrix} A^T P + PA & PB \\ * & -\mu I \end{bmatrix} < 0$$

$$\begin{bmatrix} P & C^T \\ * & \mu I \end{bmatrix} > 0$$

2. There exists $Q \in \mathbb{S}^n$ and $\mu \in \mathbb{R}_{>0}$ where $Q > 0$ such that:

$$\begin{bmatrix} QA^T + AQ & B \\ * & -\mu I \end{bmatrix} < 0$$

$$\begin{bmatrix} Q & QC^T \\ * & \mu I \end{bmatrix} > 0$$

3. There exists $P \in \mathbb{S}^n, V \in \mathbb{R}^{n \times n}$ and $\mu \in \mathbb{R}_{>0}$ where $P > 0$ such that:

$$\begin{bmatrix} -(V + V^T) & V^T A + P & V^T B & V^T \\ * & -P & 0 & 0 \\ * & ** -\mu I & 0 & 0 \\ * & * & * & -P \end{bmatrix} < 0$$
$$\begin{bmatrix} P & C^T \\ * & \mu I \end{bmatrix} > 0$$

42.5 Conclusion:

The generalized H_2 norm of G is the minimum value of $\mu \in \mathbb{R}_{>0}$ that satisfies the LMIs presented in this page.

42.6 Implementation

This implementation requires Yalmip and Sedumi.

Generalized H_2 Norm¹ - MATLAB code for Generalized H_2 Norm.

42.7 Related LMIs

LMI for System H_{∞} Norm²

42.8 External Links

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.

1 https://github.com/nbeauli/Optimal-Control-LMI/blob/main/Gen_H2_Norm.m

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_System_H2_Norm

3 http://control.asu.edu/MAE598_frame.htm

4 <https://arxiv.org/abs/1903.08599/>

- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications⁶ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013.

⁵ <https://web.stanford.edu/~boyd/lmibook/>

⁶ <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

43 Passivity and Positive Realness

This section deals with passivity of a system.

43.1 The System

Given a state-space representation of a linear system

$$\begin{aligned}\dot{x} &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

$x \in \mathbb{R}^n, y \in \mathbb{R}^m, u \in \mathbb{R}^r$ are the state, output and input vectors respectively.

43.2 The Data

A, B, C, D are system matrices.

43.3 Definition

The linear system with the same number of input and output variables is called passive if

UNKNOWN TEMPLATE NumBlk

$$:\int_0^T u^T y(t) dt \geq 0$$

UNKNOWN TEMPLATE EquationRef

1

hold for any arbitrary $T \geq 0$, arbitrary input $u(t)$, and the corresponding solution $y(t)$ of the system with $x(0) = 0$. In addition, the transfer function matrix

UNKNOWN TEMPLATE NumBlk

$$:G(s) = C(sI - A)^{-1}B + D$$

UNKNOWN TEMPLATE EquationRef

2

of system is called is positive real if it is square and satisfies

UNKNOWN TEMPLATE NumBlk

$$: G^H(s) + G(s) \geq 0 \forall s \in \mathbb{C}, Re(s) > 0$$

UNKNOWN TEMPLATE EquationRef

3

43.4 LMI Condition

Let the linear system be controllable. Then, the system is passive if and only if there exists $P > 0$ such that

UNKNOWN TEMPLATE NumBlk

$$: \begin{bmatrix} A^T P + PA & PB - C^T \\ B^T P - C & -D^T - D \end{bmatrix} \leq 0$$

UNKNOWN TEMPLATE EquationRef

4

43.5 Implementation

This implementation requires Yalmip and Mosek.

- <https://github.com/ShenoyVaradaraya/LMI--master/blob/main/Passivity.m>

43.6 Conclusion

Thus, it is seen that passivity and positive-realness describe the same property of a linear system, one gives the time-domain feature and the other provides frequency-domain feature of this property.

43.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm

2 <https://web.stanford.edu/~boyd/lmibook/>

- LMIs in Control Systems: Analysis, Design and Applications³ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013

³ [https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/
Duan-Yu/p/book/9781466582996](https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996)

44 Non-expansivity and Bounded Realness

This section studies the non-expansivity and bounded-realness of a system.

44.1 The System

Given a state-space representation of a linear system

$$\begin{aligned}\dot{x} &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

$x \in \mathbb{R}^n, y \in \mathbb{R}^m, u \in \mathbb{R}^r$ are the state, output and input vectors respectively.

44.2 The Data

A, B, C, D are system matrices.

44.3 Definition

The linear system with the same number of input and output variables is called non-expansive if

UNKNOWN TEMPLATE NumBlk

$$:\int_0^T y^T y(t) dt \geq \int_0^T u^T u(t) dt$$

UNKNOWN TEMPLATE EquationRef

1

hold for any arbitrary $T \geq 0$, arbitrary input $u(t)$, and the corresponding solution $y(t)$ of the system with $x(0) = 0$. In addition, the transfer function matrix

UNKNOWN TEMPLATE NumBlk

$$:G(s) = C(sI - A)^{-1}B + D$$

UNKNOWN TEMPLATE EquationRef

2

of system is called is positive real if it is square and satisfies

UNKNOWN TEMPLATE NumBlk

$$: G^H(s) + G(s) \geq I \forall s \in \mathbb{C}, \operatorname{Re}(s) > 0$$

UNKNOWN TEMPLATE EquationRef

3

44.4 LMI Condition

Let the linear system be controllable. Then, the system is bounded-real if and only if there exists $P > 0$ such that

UNKNOWN TEMPLATE NumBlk

$$: \begin{bmatrix} A^T P + PA & PB & C^T \\ B^T P & -I & D^T \\ C & D & -I \end{bmatrix} < 0$$

UNKNOWN TEMPLATE EquationRef

4

and

UNKNOWN TEMPLATE NumBlk

$$: \begin{bmatrix} PA + A^T P + C^T C & PB + C^T D \\ B^T P + D^T C & D^T D - I \end{bmatrix} < 0$$

UNKNOWN TEMPLATE EquationRef

5

44.5 Implementation

This implementation requires Yalmip and Mosek.

- https://github.com/ShenoyVaradaraya/LMI--master/blob/main/bounded_realness.m

44.6 Conclusion:

Thus, it is seen that passivity and positive-realness describe the same property of a linear system, one gives the time-domain feature and the other provides frequency-domain feature of this property.

44.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications³ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013

1 http://control.asu.edu/MAE598_frame.htm

2 <https://web.stanford.edu/~boyd/lmibook/>

3 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

45 Change of Subject

LMIs in Control/Matrix and LMI Properties and Tools/Change of Subject

A Bilinear Matrix Inequality (BMI) can sometimes be converted into a Linear Matrix Inequality (LMI) using a change of variables. This is a basic mathematical technique of changing the position of variables with respect to equal signs and the inequality operators. The change of subject will be demonstrated by the example below.

45.1 Example

Consider $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $K \in \mathbb{R}^{m \times n}$, and $Q \in \mathbb{S}^n$, where $Q > 0$.

The matrix inequality given by:

$QA^T + AQ - QK^T BT - BKQ < 0$ is bilinear in the variables Q and K .

Defining a change of variable as $F = KQ$ to obtain

$$QA^T + AQ - F^T B^T - BF < 0,$$

which is an LMI in the variables Q and F .

Once this LMI is solved, the original variable can be recovered by $K = FQ^{-1}$.

45.2 Conclusion

It is important that a change of variables is chosen to be a one-to-one mapping in order for the new matrix inequality to be equivalent to the original matrix inequality. The change of variable $F = KQ$ from the above example is a one-to-one mapping since Q^{-1} is invertible, which gives a unique solution for the reverse change of variable $K = FQ^{-1}$.

45.3 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.

¹ http://control.asu.edu/MAE598_frame.htm

- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A downloadable book on LMIs by Ryan James Caverly and James Richard Forbes.

2 <https://web.stanford.edu/~boyd/lmibook/>
3 <https://arxiv.org/abs/1903.08599v2/>

46 Congruence Transformation

LMIs in Control/Matrix and LMI Properties and Tools/Congruence Transformation

This methods uses change of variable and some matrix properties to transform Bilinear Matrix Inequalities to Linear Matrix Inequalities. This method preserves the definiteness of the matrices that undergo the transformation.

46.1 Theorem

Consider $Q \in \mathbb{S}^n, W \in \mathbb{R}^{n \times n}$, where $\text{rank}(W) = n$. The matrix inequality $Q < 0$ is satisfied if and only if $WQW^T < 0$ or equivalently, $W^T Q W < 0$.

46.2 Example

Consider $A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, K \in \mathbb{R}^{m \times p}, C^T \in \mathbb{R}^{n \times p}, P \in \mathbb{S}^n$ and $V \in \mathbb{S}^p$, where $P > 0$ and $V > 0$. The matrix inequality given by

$$Q = \begin{bmatrix} A^T P + PA & -PBK + C^T V \\ * & -2V \end{bmatrix} < 0,$$

is linear in variable V and bilinear in the variable pair (P, K) . Choose the matrix $\text{diag}(P^{-1}, V^{-1})$ to obtain the equivalent BMI given by

$$WQW^T = \begin{bmatrix} P^{-1}A^T + AP^{-1} & -BKV^{-1} + P^{-1}C^T \\ * & -2V^{-1} \end{bmatrix} < 0,$$

Using a change of variable $X = P^{-1}, U = V^{-1}$ and $F = KV^{-1}$, the above equation becomes

$$WQW^T = \begin{bmatrix} XA^T + AX & -BF + XC^T \\ * & -2U \end{bmatrix} < 0,$$

which is an LMI of variables X, U and F . The original variable K is recovered by doing a reverse change of variable $K = FU^{-1}$.

46.3 Conclusion

A congruence transformation preserves the definiteness of a matrix by ensuring that $Q < 0$ and $W^T Q W < 0$ are equivalent. A congruence transformation is related, but not equivalent to a similarity transformation TQT^{-1} , which preserves not only the

definiteness, but also the eigenvalues of a matrix. A congruence transformation is equivalent to a similarity transformation in the special case when $W^T = W^{-1}$.

46.4 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A downloadable book on LMIs by Ryan James Caverly and James Richard Forbes.

1 http://control.asu.edu/MAE598_frame.htm

2 <https://web.stanford.edu/~boyd/lmibook/>

3 <https://arxiv.org/abs/1903.08599v2/>

47 Finsler's Lemma

LMIs in Control/Matrix and LMI Properties and Tools/Finsler's Lemma

This method It states equivalent ways to express the positive definiteness of a quadratic form Q constrained by a linear form L . It is equivalent to other lemmas used in optimization and control theory, such as Yakubovich's S-lemma, Finsler's lemma and it is widely used in Linear Matrix Inequalities

47.1 Theorem

Consider $\Psi \in \mathbb{S}^n, G \in \mathbb{R}^{n \times m}, A \in \mathbb{R}^{m \times p}, H \in \mathbb{R}^{n \times p}$ and $\sigma \in \mathbb{R}$. There exists A such that

$$\Psi + GAH^T + HA^T G^T < 0,$$

if and only if there exists σ such that

$$\Psi - \sigma GG^T < 0$$

$$\Psi - \sigma HH^T < 0$$

47.2 Alternative Forms of Finsler's Lemma

Consider $\Psi \in \mathbb{S}^n, Z \in \mathbb{R}^{p \times n}, x \in \mathbb{R}^n$ and $\sigma \in \mathbb{R}_{>0}$. If there exists Z such that

$$x^T \Psi x, 0$$

holds for all $x \neq 0$ satisfying $Zx = 0$, then there exists σ such that

$$\Psi - \sigma Z^T Z < 0$$

47.3 Modified Finsler's Lemma

Consider $\Psi \in \mathbb{S}^n, G \in \mathbb{R}^{n \times m}, A \in \mathbb{R}^{m \times p}, H \in \mathbb{R}^{n \times p}$ and $\epsilon \in \mathbb{R}_{>0}$, where $A^T A$ is less than or equal to \mathbb{R} , and $R > 0$. There exists A such that

$$\Psi + GAH^T + HA^T G^T < 0,$$

there exists ϵ such that

$$\Psi + \epsilon^{-1} GG^T + \epsilon HRH^T < 0.$$

47.4 Conclusion

In summary, a number of identical methods have been stated above to determine the positive definiteness of LMIs.

47.5 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A downloadable book on LMIs by Ryan James Caverly and James Richard Forbes.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://web.stanford.edu/~boyd/lmibook/>
3 <https://arxiv.org/abs/1903.08599v2/>

48 D-Stability

1. Continuous Time D-Stability Observer¹

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Continuous_Time_D-Stability_Observer

49 Time-Delay Systems

1. Delay Dependent Time-Delay Stabilization¹
2. Delay Independent Time-Delay Stabilization²

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Delay_Dependent_Time-Delay_Stabilization
2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Delay_Independent_Time-Delay_Stabilization

50 Parametric, Norm-Bounded Uncertain System Quadratic Stability

UNKNOWN TEMPLATE FULLPAGENAME

Given a system with matrices A, M, N, Q the quadratic stability of the system with parametric, norm-bounded uncertainty can be determined by the following LMI. The feasibility of the LMI tells if the system is quadratically stable or not.

50.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Mp(t), \quad p(t) = \Delta(t)q(t), \\ q(t) &= Nx(t) + Qp(t), \quad \Delta \in \Delta := \{\Delta \in \mathbb{R}^{n \times n} : \|\Delta\| \leq 1\}\end{aligned}$$

50.2 The Data

The matrices A, M, N, Q .

50.3 The LMI:

$$\begin{aligned}&\text{Find } P > 0, \mu \geq 0 : \\ \begin{bmatrix} AP + PA^T & PN^T \\ NP & 0 \end{bmatrix} &+ \mu \begin{bmatrix} MM^T & MQ^T \\ QM^T & QQ^T - I \end{bmatrix} < 0\end{aligned}$$

50.4 Conclusion:

The system above is quadratically stable if and only if there exists some $\mu \geq 0$ and $P > 0$ such that the LMI is feasible.

50.5 Implementation

<https://github.com/mcavorsi/LMI>

50.6 Related LMIs

Stability of Structured, Norm-Bounded Uncertainty¹

Stability under Arbitrary Switching²

Quadratic Stability Margins³

50.7 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Stability_of_Structured,_Norm-Bounded_Uncertainty
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Stability_under_Arbitrary_Switching
3 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Quadratic_Stability_Margins
4 http://control.asu.edu/MAE598_frame.htm

51 Stability of Structured, Norm-Bounded Uncertainty

UNKNOWN TEMPLATE FULLPAGENAME

Given a system with matrices A, M, N, Q with structured, norm-bounded uncertainty, the stability of the system can be found using the following LMI. The LMI takes variables P and Θ and checks for a feasible solution.

51.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Mp(t), \quad p(t) = \Delta(t)q(t), \\ q(t) &= Nx(t) + Qp(t), \quad \Delta \in \Delta, \|\Delta\| \leq 1\end{aligned}$$

51.2 The Data

The matrices A, M, N, Q .

51.3 The LMI:

$$\text{Find } P > 0 : \\ \begin{bmatrix} AP + PA^T & PN^T \\ NP & 0 \end{bmatrix} + \begin{bmatrix} M\Theta M^T & M\Theta Q^T \\ Q\Theta M^T & Q\Theta Q^T - \Theta \end{bmatrix} < 0$$

51.4 Conclusion:

The system above is quadratically stable if and only if there exists some $\Theta \in P\Theta$ and $P > 0$ such that

51.5 Implementation

<https://github.com/mcavorsi/LMI>

51.6 Related LMIs

Parametric, Norm-Bounded Uncertain System Quadratic Stability¹

Stability under Arbitrary Switching²

Quadratic Stability Margins³

51.7 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Parametric,_Norm-Bounded_Uncertain_System_Quadratic_Stability
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Stability_under_Arbitrary_Switching
3 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Quadratic_Stability_Margins
4 http://control.asu.edu/MAE598_frame.htm

52 Stability under Arbitrary Switching

LMIs in Control/Stability Analysis/Continuous Time/Stability under Arbitrary Switching

Using the LMI below, find a P matrix that fits the constraints. If there exists one, then the system can switch between subsystems A_1 and A_2 arbitrarily and remain stable.

52.1 The System

$$\dot{x}(t) \in \{A_1x(t), A_2x(t)\}$$

52.2 The Data

The matrices $A_1 \in R^{n \times n}, A_2 \in R^{n \times n}$.

52.3 The LMI

$$\begin{aligned} \text{Find } P &> 0 : \\ A_1^T P + PA_1 &< 0 \text{ and} \\ A_2^T P + PA_2 &< 0 \end{aligned}$$

52.4 Conclusion

The switched system is stable under arbitrary switching if there exists some $P > 0$ such that the LMIs hold.

52.5 Implementation

<https://github.com/mcavorsi/LMI>

52.6 Related LMIs

Parametric, Norm-Bounded Uncertain System Quadratic Stability¹

Stability of Structured, Norm-Bounded Uncertainty²

Quadratic Stability Margins³

52.7 External links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Parametric,_Norm-Bounded_Uncertain_System_Quadratic_Stability
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Stability_of_Structured,_Norm-Bounded_Uncertainty
3 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Quadratic_Stability_Margins
4 http://control.asu.edu/MAE598_frame.htm

53 Quadratic Stability Margins

UNKNOWN TEMPLATE FULLPAGENAME

The quadratic stability margin of the system is defined as the largest $\alpha \geq 0$ for which the system is

53.1 The System

$$\dot{x}(t) = Ax(t) + B_p p(t), \quad p^T p \leq \alpha^2 x^T C_q^T C_q x$$

53.2 The Data

The matrices A, B_p, C_q .

53.3 The Optimization Problem

Maximize $\beta = \alpha^2$ subject to the LMI constraint.

53.4 The LMI:

$$\begin{aligned} & \text{Find } P, \lambda, \text{ and } \beta = \alpha^2 : \\ & \begin{bmatrix} A^T P + PA + \beta \lambda C_q^T C_q & PB_p \\ B_p^T P & -\lambda I \end{bmatrix} < 0 \end{aligned}$$

53.5 Conclusion:

If there exists an $\alpha \geq 0$ then the system is quadratically stable, and the stability margin is the largest such α .

53.6 Implementation

<https://github.com/mcavorsi/LMI>

53.7 Related LMIs

Parametric, Norm-Bounded Uncertain System Quadratic Stability¹

Stability of Structured, Norm-Bounded Uncertainty²

Stability under Arbitrary Switching³

53.8 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Parametric,_Norm-Bounded_Uncertain_System_Quadratic_Stability

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Stability_of_Structured,_Norm-Bounded_Uncertainty

3 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Stability_under_Arbitrary_Switching

4 http://control.asu.edu/MAE598_frame.htm

5 <https://web.stanford.edu/~boyd/lmibook/>

54 Stability of Linear Delayed Differential Equations

54.1 The System

$$\dot{x}(t) = Ax(t) + \sum_{i=1}^L A_i x(t - \tau_i),$$

where $x(t) \in \mathbb{R}^n$ and $\tau_i > 0$.

54.2 The Data

The matrices $A, \{A_i, \tau_i\}_{i=1}^L$.

54.3 The LMI:

Solve the following LMIP

Find $\{P \succ 0, P_1 \succ 0, \dots, P_L \succ 0\}$:

$$s.t. \begin{bmatrix} A^\top P + PA + \sum_{i=1}^L P_i & PA_1 & \dots & PA_L \\ A_1^\top P & -P_1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ A_L^\top P & 0 & \dots & -P_L \end{bmatrix} \prec 0, P_1 \succ 0, \dots, P_L \succ 0.$$

54.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/50fc71737b69f2cf57d15634f2f19d091bf37d02>

54.5 Conclusion

The stability of the above linear delayed differential equation is proved, using Lyapunov functionals of the form $V(x, t) = x(t)^\top Px(t) + \sum_{i=1}^L \int_0^L x^\top(t-s)P_i x(t-s) ds$, if the provided LMIP is feasible.

54.6 Remark

The techniques for proving stability of norm-bound LDIs [Boyd, ch.5] can also be used.

54.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

55 H infinity Norm for Affine Parametric Varying Systems

55.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_w w(t), \\ z(t) &= C_z(\theta)x(t) + D_{zw}(\theta)w(t),\end{aligned}$$

where C_z and D_{zw} depend affinity on parameter $\theta \in \mathbb{R}^p$.

55.2 The Data

The matrices $A, B_w, C_z(\cdot), D_{zw}(\cdot)$.

55.3 The Optimization Problem:

Solve the following semi-definite program

$$\begin{aligned}\min_{\{P\succ 0, \gamma \geq 0\}} \quad & \gamma \\ \text{s.t.} \quad & \begin{bmatrix} A^\top P + PA & PB_w \\ B_w^\top P & -\gamma^2 I \end{bmatrix} + \begin{bmatrix} C_z^\top(\theta) \\ D_{zw}^\top(\theta) \end{bmatrix} \begin{bmatrix} C_z(\theta) & D_{zw}(\theta) \end{bmatrix} \preceq 0.\end{aligned}$$

55.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/5462bc1dc441bc298d50a2c35075e9466eba8355>

55.5 Conclusion

The value function of the above semi-definite program returns the \mathcal{H}_∞ norm of the system.

55.6 Remark

It is assumed that A is stable and (A, B_w) is controllable and the semi-infinite convex constraint $\theta(j\omega) < \gamma$ for all $\omega \in \mathbb{R}$, is converted into a finite-dimensional convex LMI.

55.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

56 Entropy Bond for Affine Parametric Varying Systems

56.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_w w(t), \\ z(t) &= C_z(\theta)x(t) + D_{zw}(\theta)w(t),\end{aligned}$$

where C_z and D_{zw} depend affinely on parameter $\theta \in \mathbb{R}^p$.

56.2 The Data

The matrices $A, B_w, C_z(\cdot), D_{zw}(\cdot)$.

56.3 The Optimization Problem:

Solve the following semi-definite program

$$\begin{aligned}& \min_{\{P\succeq 0, \gamma^2, \lambda, \theta\}} \gamma^2 \\ \text{s.t. } & D_{zw}(\theta) = 0, \begin{bmatrix} A^\top P + PA & PB_w & C_z(\theta)^\top \\ B_w^\top P & -\gamma^2 I & 0 \\ C_z(\theta) & 0 & -I \end{bmatrix} \preceq 0, \quad \text{Tr}(B_w^\top P B_w) \leq \lambda.\end{aligned}$$

56.4 Implementation

<https://github.com/mkhajenjad/Mohammad-Khajenjad/commit/02f31a2d7a22b2464dfc9212eb76409bda9439b1>

56.5 Conclusion

The value function of the above semi-definite program returns a bound for γ -entropy of the system, which is defined as

$$I_\gamma(H_\theta) \triangleq \begin{cases} \frac{-\gamma^2}{2\pi} \int_{-\infty}^{\infty} \log \det(I - \gamma^2 H_\theta(i\omega) H_\theta(i\omega)^*) d\omega, & \text{if } \|\theta\|_\infty < \gamma \\ \infty, & \text{otherwise.} \end{cases}$$

56.6 Remark

When it is finite, $I_\gamma(H_\theta)$ is given by $\text{Tr}(B_w^\top P B_w)$ where P , is asymmetric matrix with the smallest possible maximum singular value among all solutions of a Riccati equation.

56.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599>

3 <https://web.stanford.edu/~boyd/lmibook/>

57 Dissipativity of Affine Parametric Varying Systems

57.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_w w(t), \\ z(t) &= C_z(\theta)x(t) + D_{zw}(\theta)w(t),\end{aligned}$$

where C_z and D_{zw} depend affinely on parameter $\theta \in \mathbb{R}^p$.

57.2 The Data

The matrices $A, B_w, C_z(\cdot), D_{zw}(\cdot)$.

57.3 The Optimization Problem:

Solve the following semi-definite program

$$\begin{aligned}\min_{\{P\succeq 0, \gamma, \theta\}} \quad & \gamma \\ \text{s.t.} \quad & \begin{bmatrix} A^\top P + PA & PB_w - C_z(\theta)^\top \\ B_w^\top P - C_z(\theta) & 2\gamma I - D_{zw}(\theta) - D_{zw}(\theta)^\top \end{bmatrix} \preceq 0.\end{aligned}$$

57.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/b6cd6b81f75be4a2052ba3fa76cad1a2f9c49caa>

57.5 Conclusion

The dissipativity of H_θ (see [Boyd, eq:6.59]) exceeds γ if and only if the above LMI holds and the value function returns the minimum provable dissipativity.

57.6 Remark

It is worth noticing that passivity corresponds to zero dissipativity.

57.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

58 Hankel Norm of Affine Parameter Varying Systems

58.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_w w(t), \\ z(t) &= C_z(\theta)x(t) + D_{zw}(\theta)w(t),\end{aligned}$$

where C_z and D_{zw} depend affinely on parameter $\theta \in \mathbb{R}^p$.

58.2 The Data

The matrices $A, B_w, C_z(\cdot), D_{zw}(\cdot)$.

58.3 The Optimization Problem:

Solve the following semi-definite program

$$\begin{aligned}\min_{\{Q \succeq 0, \gamma^2, \theta\}} \quad & \gamma^2 \\ \text{s.t.} \quad & D_{zw}(\theta) = 0, \quad A^\top Q + QA + C_z(\theta)C_z(\theta)^\top \preceq 0, \quad \gamma^2 I - W_c^{1/2} Q W_c^{1/2} \succeq 0,\end{aligned}$$

where W_c is the controllability Gramian, i.e., $W_c \triangleq \int_0^\infty e^{At} B_w B_w^\top e^{A^\top t} dt$.

58.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/0faedcdd9fba92bc27a318d80159c04a0b342f35>

58.5 Conclusion

The Hanakel norm (i.e., the square root of the maximum eigenvalue) of H_θ is less than γ if and only if the above LMI holds and the value function returns the maximum provable Hankel norm.

58.6 Remark

D_{zw} is assumed to be zero.

58.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

59 Positive Orthant Stabilizability

Positive Orthant Stabilizability

The positive orthant stability of a linear system refers to the property of the system states being real and positive for all $t \geq 0$ and decaying down to zero over time. In this section, the feasibility problem for systems to be positive orthant stable, and the stabilizability conditions to make the system positive orthant stable will be covered.

59.1 The System

Consider a linear state-space representation of a system as:

$$\dot{x}(t) = Ax(t) + Bu(t)$$

where $x(t) \in \mathbb{R}^n$ and $u(t) \in \mathbb{R}^r$ are the system state and the input vector respectively. A and B are system coefficient matrices of appropriate dimensions.

59.2 The Data

Number of states n and number of control inputs r need to be known. Moreover, the system matrices A, B are also required to be known.

59.3 The Feasibility LMI

An LTI system is positive orthant stable if $x(0) \in \mathbb{R}_+^n$ implies that $\forall t \geq 0, x(t) \in \mathbb{R}_+^n$. Moreover, as $t \rightarrow \infty$, $x(t) \rightarrow 0$. This is possible if and only if the following conditions hold:

$$\begin{aligned} A_{ij} &\geq 0, \forall i \neq j, \\ \exists P > 0 \text{ s.t. } PA^\top + AP &< 0, \end{aligned}$$

The above LMI feasibility is the positive orthant stability criteria. To convert it into a *positive orthant stabilizability* check, the problem can be modified so that we check if $\dot{x} = (A + BK)x$ is positive orthant stable. As K is also a design variable here, the second inequality in the above LMI will result in bilinearity. A simple change of

variables can overcome that to result in the following LMI feasibility problem for checking positive orthant stabilizability of the LTI system:

Find Q, Y subj. to:

$$\begin{aligned} Q &> 0, (AQ + BY)_{ij} \geq 0, \forall i \neq j, \\ \text{s.t. } & QA^\top + AQ + BY + Y^\top B^\top < 0, \end{aligned}$$

If the above LMI is feasible, the LTI system is stabilizable with controller $K = YQ^{-1}$.

59.4 Conclusion:

The feasibility of the above LMIs guarantees that the system is positive orthant stable if the first LMI is feasible or stabilizable with a controller if the second LMI holds.

59.5 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/Positive_Orthant_LMI.m

59.6 External Links

A list of references documenting and validating the LMI.

- LMIs in Systems and Control Theory¹ - A downloadable book on LMIs by Stephen Boyd.

¹ <https://web.stanford.edu/~boyd/lmibook/>

60 LMI For Stabilization Condition for Systems With Unsymmetrical Saturated Control

LMI For Stabilization Condition for Systems With Unsymmetrical Saturated Control

The LMI in this page gives the feasibility conditions which, if satisfied, imply that the corresponding system can be stabilized.

60.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + BSat(u(t)), \\ x(0) &= x_0, \\ Sat(u)_i &=,\end{aligned}$$

where $x \in \mathbb{R}^n$ is the state, $u \in \mathbb{R}^m$ is the control input.

For the system given as above, its symmetrical saturated control form can be derived by following the procedure in the original article. The new system will have the form:

$$\dot{x}(t) = Ax(t) + \tilde{B}sat(z(t)) + Ew$$

$$\text{where } w_i = u_i - \frac{\alpha_i - \beta_i}{2}, z_i = w_i \frac{2}{\alpha_i + \beta_i}$$

60.2 The Data

The system matrices (A, \tilde{B}, E) , the saturation bounds (α_i, β_i) of the control inputs. Positive scalars ρ, η .

60.3 The LMI: The Stabilization Feasibility Condition

$$\begin{aligned}
 & \text{Find } X, Y, Z : \\
 & \text{subj. to: } X > 0, \\
 & \quad [AX + \tilde{B}(D_s Y + \hat{D}_s^- Z)] + [AX + \tilde{B}(D_s Y + \hat{D}_s^- Z)]^\top + \frac{\eta}{\rho} X + \frac{1}{\eta} E E^\top < \\
 & \quad \begin{bmatrix} \mu & Z_i \\ *X & \end{bmatrix} > 0, i = 1, \dots, \bar{m}
 \end{aligned}$$

Here D_s is a diagonal matrix with a component either 0 or 1, and $D_s + D_s^- = \frac{\Lambda + T}{2}$ and $\hat{D}_s^- = e_{f_m} \times D_s^-$

60.4 Conclusion:

The feasibility of the given LMI implies that the system is stabilizable with control gains $K = Y X^{-1}, H = Z X^{-1}$.

60.5 Implementation

A link to CodeOcean or other online implementation of the LMI

60.6 Related LMIs

60.7 External Links

- Stabilization of Systems with Unsymmetrical Saturated Control: An LMI Approach¹ Link to the original article.

¹ <https://link.springer.com/content/pdf/10.1007/s00034-014-9786-5.pdf>

61 LMI Condition For Exponential Stability of Linear Systems With Interval Time-Varying Delays

LMI Condition For Exponential Stability of Linear Systems With Interval Time-Varying Delays

For systems experiencing time-varying delays where the delays are bounded, the feasibility LMI in this section can be used to determine if the system is α -exponentially stable.

61.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Dx(t - h(t)), & t \in \mathbb{R}^+, \\ x(t) &= \phi(t), & t \in [-h_2, 0],\end{aligned}$$

where $x(t) \in \mathbb{R}^n$ is the state, $A, D \in \mathbb{R}^{n \times n}$ are the matrices of delay dynamics, and $\phi(t) \in \mathbb{R}^n$ is the initial function with norm $\|\phi\| = \sup_{-\bar{h} \leq t \leq 0} \{\|\phi(t)\|, \|\dot{\phi}(t)\|\}$ and it is continuously differentiable function on $[-h_2, 0]$. The time-varying delay function $h(t)$ satisfies:

$$0 \leq h_1 \leq h(t) \leq h_2, t \in \mathbb{R}^+,$$

61.2 The Data

The matrices (A, D) are known, as well as the bounds (h_1, h_2) of the time-varying delay.

61.3 The Optimization Problem

For a given $\alpha > 0$, the zero solution of the system described above is α -exponentially stable if there exists a positive number $N > 0$ such that every solution $x(t, \phi)$ satisfies the following condition:

$$(t, \phi) \| \leq Ne^{-\alpha t} \|\phi\|, \forall t \in \mathbb{R}^+$$

61.4 The LMI: α -Stability Condition

The following feasibility LMI can be used to check if the system is α -exponentially stable or not for a given $\alpha > 0$:

Find P, Q, R, U, S_i , where $i = 1, 2, \dots, 5$:

$$\begin{bmatrix} M_{11} & M_{12} & M_{13} & M_{14} & M_{15} \\ * & M_{22} & 0 & M_{24} & S_2 \\ * & * & M_{33} & M_{34} & S_3 \\ * & * & * & M_{44} & S_4 - S_5 D \\ * & * & * & * & M_{55} \end{bmatrix} < 0$$

where:

$$\begin{aligned} M_{11} &= A^\top P + PA + 2\alpha P - (e^{-2\alpha h_1} + e^{-2\alpha h_2})R + 0.5S_1(I - A) + 0.5(I - A^\top)S_1^\top + 2Q, \\ M_{12} &= e^{-2\alpha h_1}R - S_2A, \quad M_{13} = e^{-2\alpha h_2}R - S_3A, \\ M_{14} &= PD - S_1D - S_4A, \quad M_{15} = S_1 - S_5A, \\ M_{22} &= -e^{-2\alpha h_1}(Q + R), \quad M_{24} = S_2D + e^{-2\alpha h_2}U, \\ M_{33} &= -e^{-2\alpha h_1}(Q + R + U), \quad M_{34} = -S_3D + e^{-2\alpha h_2}U, \\ M_{44} &= 0.5(S_4D + D^\top S_4^\top) - e^{-2\alpha h_2}U, \\ M_{55} &= S_5 + S_5^\top + (h_1^2 + h_2^2)R + (h_2 - h_1)^2U, \end{aligned}$$

The above LMI can be combined with the bisection method to find α .

61.5 Conclusion:

For systems with time-varying delays with intervals, the LMI in this section can be used to check if the system is exponentially stable with a certain α . The bisection algorithm can be additionally used to compute α .

61.6 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/Intervaled_Delay_Sys_Stability_example.m

61.7 External Links

A list of references documenting and validating the LMI.

- LMI approach to exponential stability of linear systems with interval time-varying delays¹ - Original Article by Phat et al.

61.8 Return to Main Page:

¹ <https://core.ac.uk/download/pdf/82311937.pdf>

62 Conic Sector Lemma

Conic Sector Lemma

For general input-output systems, sector conditions are formulated to verify or enforce the feedback stability. One of these sector conditions is the **conic sector lemma**, and the problem that designs the feedback controller is the conic sector theorem.

62.1 The System

Consider a square, continuous-time linear time-invariant (LTI) system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization (A, B, C, D) , where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $C \in \mathbb{R}^{m \times n}$, and $D \in \mathbb{R}^{m \times m}$. The state-space representation is:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and $u(t) \in \mathbb{R}^m$ are the system state, output, and the input vector respectively.

62.2 The Data

The system coefficient matrices (A, B, C, D) are required. Optionally, the parameters to define a cone, either in the form of $[a, b]$ where $a, b \in \mathbb{R}, a < b$ or a radius $r \in \mathbb{R}_+$ and center $c \in \mathbb{R}$.

62.3 The Feasibility LMI

The system \mathcal{G} is inside the given cone $[a, b]$ if the following is feasible:

$$\begin{aligned}&\text{Find: } P \\ &\text{subj. to: } P > 0 \\ &\left[\begin{array}{cc} PA + A^\top P + C^\top C & PB - \frac{a+b}{2} C^\top + C^\top D \\ (PB - \frac{a+b}{2} C^\top + C^\top D)^\top & D^\top D - \frac{a+b}{2} (D + D^\top) + abI \end{array} \right] \leq 0.\end{aligned}$$

The above LMI can be used to also determine the cone parameters by setting a as a variable along with the condition $a < b$, and use the bisection method to find b .

If the given cone is represented by a center c and radius r , then the following feasibility problem can be evaluated to check if \mathcal{G} is inside the given cone:

$$\begin{aligned} \text{Find: } P \\ \text{subj. to: } P > 0 \\ \begin{bmatrix} PA + A^\top P + C^\top C & PB - cC^\top + C^\top D \\ (PB - cC^\top + C^\top D)^\top & D^\top D - c(D + D^\top) + (c^2 - r^2)I \end{bmatrix} \leq 0. \end{aligned}$$

In order to also find the cone parameters, substituting $\gamma = r^2$ as a decision variable with additional constraint $\gamma \geq 0$ and then solving for c via the bisection method will give the cone in which the system \mathcal{G} resides if the problem is feasible.

62.4 Conclusion:

The aforementioned LMIs can be utilized to either check if \mathcal{G} is in the specified cone or not, or can be used to check the stability of \mathcal{G} by finding if a feasible cone can be obtained that encloses \mathcal{G} . An important point to note here is that the Conic Sector Lemma is a special case of the KYP Lemma¹ for QSR dissipative systems with:

$$Q = -1, S = \frac{a+b}{2}I = cI, R = -abI = (r^2 - c^2)I.$$

62.5 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/Conic_sector_example.m

62.6 Related LMIs

Exterior Conic Sector Lemma².

KYP Lemma³

1 [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Exterior_Conic_Sector_Lemma

3 [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

62.7 External Links

A list of references documenting and validating the LMI.

- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.

62.8 Return to Main Page:

⁴ <https://arxiv.org/abs/1903.08599/>

63 Polytopic Quadratic Stability

An important result to determine the stability of the system with uncertainties

63.1 The System:

Consider the system with Affine Time-Varying uncertainty (No input)

$$\dot{x}(t) = (A_0 + \Delta A(t))x(t)$$

where

$$\Delta A(t) = A_1\delta_1(t) + \dots + A_k\delta_k(t)$$

where $\delta_i(t)$ lies in either the intervals

$$\delta_i \in [\delta_i^-, \delta_i^+]$$

or the simplex

$$\delta(t) \in \delta : \sum \alpha_i = 1, \alpha \geq 0$$

where $x \in \mathbb{R}^m$ and $A \in \mathbb{R}^{m \times m}$

63.2 The Data

The matrix A and $\Delta_{A(t)}$ are known

63.3 The Optimization

The Definitions: Quadratic Stability for Dynamic Uncertainty

The system

$$\dot{x}(t) = (A_0 + \Delta A(t))x(t)$$

is Quadratically Stable over Δ if there exists a $P > 0$

Theorem

$(A + \Delta, \Delta)$ is quadratically stable over $\Delta := Co(A_1, \dots, A_k)$ if and only if there exists a $P > 0$ such that

$$(A + A_i)^T P + P(A + A_i) < 0 \quad \text{for all } i = 1, \dots, k$$

The theorem says the LMI only needs to hold at the EXTREMAL POINTS or VERTICES of the polytope.

- Quadratic Stability MUST be expressed as an LMI

63.4 The LMI

$$(A + \Delta)^T P + P(A + \Delta) < 0 \quad \text{for all } \Delta \in \Delta$$

63.5 Conclusion:

Quadratic Stability Implies Stability of trajectories for any Δ with $\Delta \in \Delta$ for all $t \geq 0$

Quadratic Stability is CONSERVATIVE.

There are Stable System which are not Quadratically stable.

Quadratic Stability is sometimes referred to as an "infinite-dimensional LMI"

- Meaning it represents an infinite number of LMI constraints.
- One for each possible value Δ with $\Delta \in \Delta$
- Also called a parameterized LMI
- Such LMIs are not tractable.
- For polytopic sets, the LMI can be made finite.

63.6 Implementation

A link to implementation of the LMI

<https://github.com/JalpeshBhadra/LMI/blob/master/polytopicstability.m>

63.7 Related LMIs

- Parametric Norm Bounded Uncertain System Quadratic Stability¹

63.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

63.9 Return to Main Page:

1 Chapter 50 on page 137

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599>

4 <https://web.stanford.edu/~boyd/lmibook/>

64 Mu Analysis

Mu Synthesis. The technique of μ synthesis extends the methods of H synthesis to design a robust controller for an uncertain plant. You can perform μ synthesis on plants with parameter uncertainty, dynamic uncertainty, or both using the "musyn" command in MATLAB. μ analysis is an extremely powerful multivariable technique which has been applied to many problems in the almost every industry including Aerospace, process industry etc.

64.1 The System:

Consider the continuous-time generalized LTI plant with minimal states-space realization

$$\begin{aligned}\dot{x} &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

where it is assumed that D is Invertible.

64.2 The Data

The matrices needed as inputs are only, A and D .

64.3 The LMI: μ - Analysis

The inequality $\bar{\sigma}(DAD^{-1}) < \gamma$ holds if and only if there exist $X \in \mathbb{S}^n$ and $\gamma \in \mathbb{R}_{>0}$, where $X > 0$, satisfying:

$$A^T X A - \gamma^2 X < 0$$

64.4 Conclusion:

The inequality $\bar{\sigma}(DAD^{-1}) < \gamma$ holds for $D = X^{1/2}$, where X satisfies the above Inequality.

64.5 Implementation

- <https://github.com/Ricky-10/coding107/blob/master/Mu%20Analysis> - μ -Analysis

64.6 External links

- <https://www.mathworks.com/help/robust/mu-synthesis.html>-MATLAB μ -synthesis Implementation
- LMI Properties and Applications in Systems, Stability, and Control Theory¹ - A List of LMIs by Ryan Caverly and James Forbes.

¹ <https://arxiv.org/abs/1903.08599/>

65 Optimization Over Affine Family of Linear Systems

65.1 Optimization over an Affine Family of Linear Systems

Presented in this page is a general framework for optimizing various convex functionals for a system which depends affinely, or linearly, on a parameter using linear matrix inequalities. The optimization problem presented on this page generalizes an LMI which can be applied to various problems within linear systems and control. Some examples of these applications are finding the H_2 and H_∞ norms, entropy, dissipativity, and the Hankel norm of an affinely parametric system.

65.2 The System

Consider a family of linear systems

$$\begin{aligned}\dot{x} &= Ax + B_w w, \\ z &= C_z(\theta)x + D_{zw}(\theta)w\end{aligned}$$

with state space realization (A, B_w, C_z, D_{zw}) where C_z and D_{zw} depend affinely on the parameter $\theta \in \mathbb{R}^p$.

We assume A is stable and (A, B_w) is controllable.

The transfer function, $H_\theta(s) \hat{=} C_z(\theta)(sI - A)^{-1}B_w + D_{zw}(\theta)$ depends affinely on θ .

65.3 The Data

The transfer function H , and system matrices A, B_w, C_z, D_{zw} are known. φ represents the convex functionals, α and ψ represent some auxiliary variables dependent on the problem being solved.

65.4 The LMI:Generalized Optimization for Affine Linear Systems

Several control theory problems, mentioned earlier, take the following form:

$$\begin{aligned}&\text{minimize } \psi_0(H_\theta) \\ &\text{subject to } \psi_i(H_\theta) < \alpha_i, i = 1, \dots, p\end{aligned}$$

Problems of this nature can be formulated as an LMI by representing $\psi_i(H_\theta) < \alpha_i$ as an LMI in θ, α_i , and possibly ψ_i such that $F_i(\theta, \alpha_i, \psi_i) > 0$

Thus, the general optimization problem to be applied to an affine family of linear systems is as follows:

$$\begin{aligned} & \text{minimize } \alpha_0 \\ & \text{subject to } F_i(\theta, \alpha_i, \psi_i) > 0, i = 0, 1, \dots, P \end{aligned}$$

65.5 Conclusion:

The LMI for this generalized optimization problem may be extended to various convex functionals for affine parametric systems. For extensions of this LMI, see the related LMIs section.

65.6 Implementation

Implementation of LMI's of this form require Yalmip and a linear solver such as Sedumi or SDPT3.

H_∞ Norm for Affine Parametric Systems¹ - MATLAB code for an extension of this generalized LMI.

Entropy Bond for Affine Parametric Systems² - MATLAB code for an extension of this generalized LMI.

LMI can be applied to other extensions in stability and controller analysis. Please see the related LMI pages in the section below.

65.7 Related LMIs

H_∞ Norm for Affine Parametric Systems³

Entropy Bond for Affine Parametric Systems⁴

Dissipativity for Affine Parametric Systems⁵

for Affine Parametric Systems⁶

1 <https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/5462bc1dc441bc298d50a2c35075e9466eba8355>
 2 <https://github.com/mkhajenejad/Mohammad-Khajenejad/blob/master/LMIs%20for%20Entropy%20Bound%20for%20Affine%20Parameter%20Varying%20Systems>
 3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H_infinity_Norm_for_Affine_Parametric_Varying_Systems
 4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Entropy_Bound_for_Affine_Parametric_Varying_Systems
 5 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Dissipativity_of_Affine_Parametric_Varying_Systems
 6 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Hankel_Norm_for_Affine_Parametric_Varying_Systems&stable=0

65.8 External Links

- LMI Methods in Optimal and Robust Control⁷ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory⁸ - A downloadable book on LMIs by Stephen Boyd.* LMI Properties and Applications in Systems, Stability, and Control Theory⁹ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Control Systems: Analysis, Design and Applications¹⁰ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013.

65.9 Return to Main Page:

LMIs in Control: https://en.wikibooks.org/wiki/LMIs_in_Control

7 http://control.asu.edu/MAE598_frame.htm

8 <https://web.stanford.edu/~boyd/lmibook/>

9 <https://arxiv.org/abs/1903.08599/>

10 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

66 Hurwitz Stabilizability

This section studies the stabilizability properties of the control systems.

66.1 The System

Given a state-space representation of a linear system

$$\begin{aligned}\rho x &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

Where ρ represents the differential operator (when the system is continuous-time) or one-step forward shift operator (Discrete-Time system). $x \in \mathbb{R}^n, y \in \mathbb{R}^m, u \in \mathbb{R}^r$ are the state, output and input vectors respectively.

66.2 The Data

A, B, C, D are system matrices.

66.3 Definition

The system , or the matrix pair (A, B) is Hurwitz Stabilizable if there exists a real matrix K such that $(A + BK)$ is Hurwitz Stable. The condition for Hurwitz Stabilizability of a given matrix pair (A, B) is given by the PBH criterion:

UNKNOWN TEMPLATE NumBlk

$$: \text{rank} [sI - A \ B] = n, \forall s \in \mathbb{C}, \text{Re}(s) \geq 0$$

UNKNOWN TEMPLATE EquationRef

1

The PBH criterion shows that the system is Hurwitz stabilizable if all uncontrollable modes are Hurwitz stable.

66.4 LMI Condition

The system, or matrix pair (A, B) is Hurwitz stabilizable if and only if there exists symmetric positive definite matrix P and W such that:

UNKNOWN TEMPLATE NumBlk

$$:AP + PA^T + BW + W^T B^T < 0$$

UNKNOWN TEMPLATE EquationRef

2

Following definition of Hurwitz Stabilizability and Lyapunov Stability theory, the PBH criterion is true if and only if , a matrix K and a matrix $P > 0$ satisfying:

UNKNOWN TEMPLATE NumBlk

$$:(A + BK)P + P(A + BK)^T < 0$$

UNKNOWN TEMPLATE EquationRef

3

Letting

UNKNOWN TEMPLATE NumBlk

$$:W = KP$$

UNKNOWN TEMPLATE EquationRef

4

Putting (4) in (3) gives us (2).

66.5 Implementation

This implementation requires Yalmip and Mosek.

- https://github.com/ShenoyVaradaraya/LMI--master/blob/main/Hurwitz_Stabilizability.m

66.6 Conclusion

Compared with the second rank condition, LMI has a computational advantage while also maintaining numerical reliability.

66.7 References

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications³ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013

66.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm

2 <https://web.stanford.edu/~boyd/lmibook/>

3 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

67 Quadratic Hurwitz Stabilization for Polytopic Systems

This section studies the Quadratic Hurwitz stabilization for polytopic systems.

67.1 The System

Given a state-space representation of a linear system

UNKNOWN TEMPLATE NumBlk

$$\begin{aligned}\dot{x}(t) &= A(\delta(t))x(t) + B\delta(t)u(t) \\ :A(\delta(t)) &= A_0 + \delta_1(t)A_1 + \delta_2(t)A_2 \\ B(\delta(t)) &= B_0 + \delta_1(t)B_1 + \delta_2(t)B_2\end{aligned}$$

UNKNOWN TEMPLATE EquationRef

1

67.2 LMI Condition

With $\Delta = \Delta_p$, the quadratic Hurwitz Stabilization problem has a solution if and only if there exists a symmetric positive definite matrix P and a matrix W satisfying the below LMI :

UNKNOWN TEMPLATE NumBlk

$$:(A_0 + A_i)P + P(A_0 + A_i)^T + (B_0 + B_i)W + W(B_0 + B_i)^T < 0, i = 1, 2, \dots, k$$

UNKNOWN TEMPLATE EquationRef

2

In this case, a solution to the problem is given by

UNKNOWN TEMPLATE NumBlk

$$:K = WP^{-1}$$

UNKNOWN TEMPLATE EquationRef

3

67.3 Conclusion

Stability of a system does not guarantee quadratic stability. Since quadratic stability can represent infinite LMI constraints, it is not tractable. Therefore, to make it feasible and tractable, polytopic sets are helpful.

67.4 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications³ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013

1 http://control.asu.edu/MAE598_frame.htm

2 <https://web.stanford.edu/~boyd/lmibook/>

3 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

68 Discrete-Time Lyapunov Stability

Discrete-Time Lyapunov Stability

A discrete time system operates on a discrete time signal input and produces a discrete time signal output. They are used in digital signal processing, such as digital filters for images or sound. The class of discrete time systems that are both linear and time invariant, known as discrete time LTI systems.

Stability of DT LTI systems can be determined by solving Lyapunov Inequality.

68.1 The System

Discrete-Time System

$$x(t)_{k+1} = A_d x(t)_k, \quad A_d \in \mathbf{R}^{n \times n}$$

68.2 The Data

The matrices: System A_d, P .

68.3 The Optimization Problem

The following feasibility problem should be optimized:

Find P obeying the LMI constraints.

68.4 The LMI:

Discrete-Time Bounded Real Lemma

The LMI formulation

$$\begin{aligned} P &\in \mathbf{S}^n \\ \text{Find } P &> 0, \\ [A_d^T P A_d - P] &< 0 \end{aligned}$$

68.5 Conclusion:

If there exists a $P \in \mathbf{S}^n$ satisfying the LMI then, $|\lambda_i(A_d)| \leq 1, \forall i = 1, 2, \dots, n$; and the equilibrium point $\bar{x} = 0$ of the system is Lyapunov stable.

68.6 Implementation

A link to CodeOcean or other online implementation of the LMI
MATLAB Code¹

68.7 Related LMIs

[Continuous_Time_Lyapunov_Inequality²](#) - Lyapunov_Inequality

68.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

68.9 Return to Main Page:

1 <https://github.com/Harishankar-Prabhakaran/LMIs/blob/master/A1.m>
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Lyapunov_Inequality
3 http://control.asu.edu/MAE598_frame.htm
4 <https://arxiv.org/abs/1903.08599/>
5 <https://web.stanford.edu/~boyd/lmibook/>

69 LMI for Schur Stabilization

LMI for Schur Stabilization

Similar to the stability of continuous-time systems, one can analyze the stability of discrete-time systems. A discrete-time system is said to be stable if all roots of its characteristic equation lie in the open unit disk. This provides a condition for the stability of discrete-time linear systems and a linear time-invariant system with this property is called a Schur stable system.

69.1 The System

We consider the following system:

$$x(k+1) = Ax(k) + Bu(k)$$

where the matrices $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times r}$, $x \in \mathbb{R}^n$, and $u \in \mathbb{R}^r$ are the state matrix, input matrix, state vector, and the input vector, respectively.

Moreover, k represents time in the discrete-time system and $k+1$ is the next time step.

The state feedback control law is defined as follows:

$$u(k) = Kx(k)$$

where $K \in \mathbb{R}^{n \times r}$ is the controller gain. Thus, the closed-loop system is given by:

$$x(k+1) = (A + BK)x(k)$$

69.2 The Data

The matrices A and B are given.

We define the scalar as γ with the range of $0 < \gamma \leq 1$.

69.3 The Optimization Problem

The optimization problem is to find a matrix $K \in \mathbb{R}^{r \times n}$ such that:

$$\|A + BK\|_2 < \gamma$$

According to the definition of the spectral norms of matrices, this condition becomes equivalent to:

$$(A + BK)^T (A + BK) < \gamma^2 I$$

Using the Lemma 1.2 in LMI in Control Systems Analysis, Design and Applications¹ (page 14), the aforementioned inequality can be converted into:

$$\begin{bmatrix} -\gamma I & (A + BK) \\ (A + BK)^T & -\gamma I \end{bmatrix} < 0$$

69.4 The LMI: LMI for Schur stabilization

The LMI for Schur stabilization can be written as minimization of the scalar, γ , in the following constraints:

$$\begin{array}{ll} \min & \gamma \\ \text{s.t.} & \begin{bmatrix} -\gamma I & (A + BK) \\ (A + BK)^T & -\gamma I \end{bmatrix} < 0 \end{array}$$

69.5 Conclusion:

After solving the LMI problem, we obtain the controller gain K and the minimized parameter γ . This problem is a special case of Intensive Disk Region Design (page 230 in [1]). This problem may not have a solution even when the system is stabilizable. In other words, once there exists a solution, the solution is robust in the sense that when there are parameter perturbations, the closed-loop system's eigenvalues are not easy to go outside of a circle region within the unit circle [1].

69.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Schur-Stability>

69.7 Related LMIs

LMI for Hurwitz stability²

1 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Hurwitz_Stability

69.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

69.9 Return to Main Page

LMIs in Control/Tools³

³ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

70 L2-Gain of Systems with Multiplicative noise

70.1 The System

$$x(k+1) = Ax(k) + B_w w(k) + \sum_{i=1}^L (A_i x(k) + B_{w,i} w(k)) p_i(k), \quad x(0) = 0,$$

$$z(k) = C_z x(k) + D_{zw} w(k) + \sum_{i=1}^L (C_{z,i} x(k) + D_{zw,i} w(k)) p_i(k),$$

where $p(0), p(1), \dots$, are independent, identically distributed random variables with $E p(k) = 0$, $E p(k) p^\top(k) = \Sigma = \text{diag}(\sigma_1, \dots, \sigma_L)$ and $x(0)$ is independent of the process p .

70.2 The Data

The matrices $A, B_w, \{A_i, B_{w,i}\}_{i=1}^L, C_z, D_{zw}, \{C_{z,i}, D_{zw,i}\}_{i=1}^L, \{\sigma_i\}_{i=1}^L$.

70.3 The LMI:

$$\begin{aligned} & \min_{\{P \succ 0, \gamma^2\}} \gamma^2 \\ \text{s.t. } & \begin{bmatrix} A & B_w \\ C_z & D_{zw} \end{bmatrix}^\top \begin{bmatrix} P & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} A & B_w \\ C_z & D_{zw} \end{bmatrix} - \begin{bmatrix} P & 0 \\ 0 & \gamma^2 I \end{bmatrix} + \sum_{i=1}^L \sigma_i^2 \begin{bmatrix} A_i & B_{w,i} \\ C_{z,i} & D_{zw,i} \end{bmatrix}^\top \begin{bmatrix} P & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} A_i & B_{w,i} \\ C_{z,i} & D_{zw,i} \end{bmatrix} \preceq 0 \end{aligned}$$

70.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/a34713575cd8ae9831cb5b7eb759d0fd45a8c37f>

70.5 Conclusion

The optimal γ returns an upper bound on the \mathcal{L}_2 gain of the system. .

70.6 Remark

It is straightforward to apply scaling method [Boyd, sec.6.3.4] to obtain component-wise results.

70.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

70.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

71 Discrete-Time Quadratic Stability

71.1 Discrete-Time Quadratic Stability

Stability is an important property, stability analysis is necessary for control theory. For robust control, this criterion is applicable for the uncertain discrete-time linear system. It is based on the Discrete Time Lyapunov Stability.

71.2 The System

$$x_{k+1} = A_d(\alpha)x_k$$

Where :

$$A_d(\alpha) = A_d + \Delta A_d(\delta(t))$$

$$\Delta A_d(\delta(t)) = \sum_{k=1}^n \delta_k(t) A_{d;k} \in \mathbb{R}^{n \times n}$$

$\delta(t) = [\delta_1(t), \dots, \delta_n(t)]$ – The set of perturbation parameters

$$\delta(t) \in \mathbb{R}^n, A_{d;i} \in \mathbb{R}^{n \times n}$$

71.3 The Data

The matrices $A \in \mathbb{R}^{n \times n}, A_{d;i} \in \mathbb{R}^{n \times n}$.

71.4 The Optimization Problem

The following feasibility problem should be solved:

Find $P > 0$:

$$(A_{d;0} + \Delta A_d(\delta(t)))^T P (A_{d;0} + \Delta A_d(\delta(t))) - P < 0 \text{ for all } \delta$$

Where $P \in \mathbb{R}^{n,n}$.

In case of polytopic uncertainty:

Find $P > 0$:

$$(A_{d;0} + A_{d;i})^T P (A_{d;0} + A_{d;i}) - P < 0 \text{ for all } i = 1, \dots, n$$

71.5 Conclusion:

This LMI allows us to investigate stability for the robust control problem in the case of polytopic uncertainty and gives on the controller for this case

71.6 Implementation:

- https://github.com/Talitsky/LMI/blob/main/Discrete_Time_Quadratic_Stability.m - Matlab implementation using the YALMIP framework and Mosek solver

71.7 Related LMIs:

- - Discrete Time Stabilizability¹
- Polytopic stability for continuous time case²
- Quadratic polytopic stabilization³
- Discrete Time Lyapunov Stability⁴

71.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁵ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁶ - A List of LMIs by Ryan Caverly and James Forbes. (3.20.2 page 64)
- LMIs in Systems and Control Theory⁷ - A downloadable book on LMIs by Stephen Boyd.

71.9 Return to Main Page:

- Main Page.⁸

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_Stabilizability

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/polytopicstability/

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Quadratic_polytopic_stabilization

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_Lyapunov_Stability

5 http://control.asu.edu/MAE598_frame.htm

6 <https://arxiv.org/abs/1903.08599>

7 <https://web.stanford.edu/~boyd/lmibook/>

8 https://en.wikibooks.org/wiki/LMIs_in_Control/

72 Stability of Lure's Systems

72.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_p p(t) + B_w w(t), \\ z(t) &= C_z x(t) \\ p_i(t) &= \phi_i(q_i(t)), i = 1, \dots, n_p \\ q &= C_q x, \\ 0 \leq \sigma \phi_i(\sigma) &\leq \sigma^2 \quad \forall \sigma \in \mathbb{R}\end{aligned}$$

72.2 The Data

The matrices A, B_p, B_w, C_q, C_z .

72.3 The LMI: The Lure's System's Stability

The following feasibility problem should be solved as sufficient condition for the stability of the above Lur'e system.

$$\begin{aligned}\text{Find } P > 0, \Lambda = \text{diag}(\lambda_1, \dots, \lambda_{n_p}) \succeq 0, T = \text{diag}(\tau_1, \dots, \tau_{n_p}) \succeq 0 : \\ \begin{bmatrix} A^\top P + PA & PB_p + A^\top C_q^\top \Lambda + C_q^\top T \\ B_p^\top P + \Lambda C_q A + T C_q & \Lambda C_q B_p + B_p^\top C_q^\top \Lambda - 2T \end{bmatrix} &\prec 0\end{aligned}$$

72.4 Implementation

<https://codeocean.com/capsule/0232754/tree>

72.5 Conclusion

If the feasibility problem with LMI constraints has solution, then the Lure's system is stable.

72.6 Remark

The LMI is only a sufficient condition for the existence of a Lur'e Lyapunov function that proves stability of Lur'e system . It is also necessary when there is only one nonlinearity, i.e., when $n_p = 1$.

72.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

72.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

73 L2 Gain of Lure's Systems

73.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_p p(t) + B_w w(t), \\ z(t) &= C_z x(t) \\ p_i(t) &= \phi_i(q_i(t)), i = 1, \dots, n_p \\ q &= C_q x, \\ 0 \leq \sigma \phi_i(\sigma) &\leq \sigma^2 \quad \forall \sigma \in \mathbb{R}\end{aligned}$$

73.2 The Data

The matrices A, B_p, B_w, C_q, C_z .

73.3 The Optimization Problem:

The following semi-definite problem should be solved.

$$\begin{aligned}\min_{\{P\succeq 0, \Lambda=diag(\lambda_1, \dots, \lambda_{n_p})\succeq 0, T=diag(\tau_1, \dots, \tau_{n_p})\succeq 0\}} \quad & \gamma^2 \\ \text{s.t.} \quad & \begin{bmatrix} A^\top P + PA + C_z^\top C_z & PB_p + A^\top C_q^\top \Lambda + C_q^\top T & PB_w \\ B_p^\top P + \Lambda C_q A + T C_q & \Lambda C_q B_p + B_p^\top C_q^\top \Lambda - 2T & \Lambda C_q B_w \\ B_w^\top P & B_w^\top C_q^\top \Lambda & -\gamma^2 I \end{bmatrix} \preceq 0\end{aligned}$$

73.4 Implementation

<https://github.com/mkhajenjad/Mohammad-Khajenjad/commit/12a7039f9e3d966e24b43fd58a3cce3725282ed2>

73.5 Conclusion

The value function returns the square of the smallest provable upper bound on the \mathcal{L}_2 gain of the Lure's system.

73.6 Remark

The Lyapunov function which is used to proof is similar to the one for the systems with unknown parameters.

73.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

73.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

74 Output Energy Bound for Lure's Systems

74.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_p p(t) + B_w w(t), \\ z(t) &= C_z x(t) \\ p_i(t) &= \phi_i(q_i(t)), i = 1, \dots, n_p \\ q(t) &= C_q x(t), \\ 0 \leq \sigma \phi_i(\sigma) &\leq \sigma^2 \quad \forall \sigma \in \mathbb{R}\end{aligned}$$

74.2 The Data

The matrices $A, B_p, B_w, C_q, C_z, x(0)$.

74.3 The Optimization Problem:

The following optimization problem should be to find the tightest upper bound for the output energy of the above Lur'e system.

$$\begin{aligned}\min_{P\succeq 0, \Lambda=diag(\lambda_1, \dots, \lambda_{n_p})\succeq 0, T=diag(\tau_1, \dots, \tau_{n_p})\succeq 0} \quad & x^\top(0)(P + C_q^\top \Lambda C_q)x(0) \\ \left[\begin{array}{cc} A^\top P + PA & PB_p + A^\top C_q^\top \Lambda + C_q^\top T \\ B_p^\top P + \Lambda C_q A + T C_q & \Lambda C_q B_p + B_p^\top C_q^\top \Lambda - 2T \end{array} \right] & \preceq 0\end{aligned}$$

74.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/blob/master/LMIs%20for%20Output%20Energy%20Bounds%20of%20Lure's%20Systems>

74.5 Conclusion

The value function returns the lowest bound for the energy function of the Lure's systems, i.e., $J = \int_0^\infty z^\top z dt$ with initial conditions $x(0)$.

74.6 Remark

The key step in the proof is to satisfy $\frac{d}{dt}V(x) + z^\top z \leq 0$, where $V(\cdot)$ is Lyapunov function in a special form.

74.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

74.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

75 Stability of Quadratic Constrained Systems

75.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_p p(t) + B_u u(t) + B_w w(t), \\ q(t) &= C_q x(t) + D_{qp} p(t) + D_{qu} u(t) + D_{qw} w(t), \\ z(t) &= C_z x(t) + D_{zp} p(t) + D_{zu} u(t) + D_{zw} w(t) \\ \int_0^t p^\top(\tau)p(\tau) d\tau &\leq \int_0^t q^\top(\tau)q(\tau) d\tau.\end{aligned}$$

75.2 The Data

The matrices $A, B_p, B_w, C_q, C_z, D_{qp}, D_{zw}$.

75.3 The LMI:

The following feasibility problem should be solved.

$$\begin{aligned}&\text{Find } \{P \succ 0, \lambda \geq 0\} : \\ &\text{s.t. } \begin{bmatrix} A^\top P + PA + \lambda C_q^\top C_q & PB_p + \lambda C_q^\top D_{qp} \\ (PB_p + \lambda C_q^\top D_{qp})^\top & \lambda(I - D_{qp}^\top D_{qp}) \end{bmatrix} \prec 0.\end{aligned}$$

75.4 Implementation

<https://github.com/mkhajenejad/Mohammad-Khajenejad/commit/38f3b55ca7060a1260384a96e9dc31142af07a9a>

75.5 Conclusion

The integral quadratic constrained system is stable if the provided LMI is feasible

75.6 Remark

The key point of the proof is to satisfy $\dot{V} < 0$ whenever $p^\top p \leq q^\top q$, using \mathcal{S} -procedure.

75.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

75.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

76 Conic Sector Lemma

Conic Sector Lemma

For general input-output systems, sector conditions are formulated to verify or enforce the feedback stability. One of these sector conditions is the **conic sector lemma**, and the problem that designs the feedback controller is the conic sector theorem.

76.1 The System

Consider a square, continuous-time linear time-invariant (LTI) system, $\mathcal{G} : \mathcal{L}_{2e} \rightarrow \mathcal{L}_{2e}$, with minimal state-space realization (A, B, C, D) , where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $C \in \mathbb{R}^{m \times n}$, and $D \in \mathbb{R}^{m \times m}$. The state-space representation is:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and $u(t) \in \mathbb{R}^m$ are the system state, output, and the input vector respectively.

76.2 The Data

The system coefficient matrices (A, B, C, D) are required. Optionally, the parameters to define a cone, either in the form of $[a, b]$ where $a, b \in \mathbb{R}, a < b$ or a radius $r \in \mathbb{R}_+$ and center $c \in \mathbb{R}$.

76.3 The Feasibility LMI

The system \mathcal{G} is inside the given cone $[a, b]$ if the following is feasible:

Find: P

subj. to: $P > 0$

$$\begin{bmatrix} PA + A^\top P + C^\top C & PB - \frac{a+b}{2}C^\top + C^\top D \\ (PB - \frac{a+b}{2}C^\top + C^\top D)^\top & D^\top D - \frac{a+b}{2}(D + D^\top) + abI \end{bmatrix} \leq 0.$$

The above LMI can be used to also determine the cone parameters by setting a as a variable along with the condition $a < b$, and use the bisection method to find b .

If the given cone is represented by a center c and radius r , then the following feasibility problem can be evaluated to check if \mathcal{G} is inside the given cone:

$$\begin{aligned} \text{Find: } P \\ \text{subj. to: } P > 0 \\ \begin{bmatrix} PA + A^\top P + C^\top C & PB - cC^\top + C^\top D \\ (PB - cC^\top + C^\top D)^\top & D^\top D - c(D + D^\top) + (c^2 - r^2)I \end{bmatrix} \leq 0. \end{aligned}$$

In order to also find the cone parameters, substituting $\gamma = r^2$ as a decision variable with additional constraint $\gamma \geq 0$ and then solving for c via the bisection method will give the cone in which the system \mathcal{G} resides if the problem is feasible.

76.4 Conclusion:

The aforementioned LMIs can be utilized to either check if \mathcal{G} is in the specified cone or not, or can be used to check the stability of \mathcal{G} by finding if a feasible cone can be obtained that encloses \mathcal{G} . An important point to note here is that the Conic Sector Lemma is a special case of the KYP Lemma¹ for QSR dissipative systems with:

$$Q = -1, S = \frac{a+b}{2}I = cI, R = -abI = (r^2 - c^2)I.$$

76.5 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/Conic_sector_example.m

76.6 Related LMIs

Exterior Conic Sector Lemma².

KYP Lemma³

1 [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Exterior_Conic_Sector_Lemma

3 [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma))

76.7 External Links

A list of references documenting and validating the LMI.

- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.

76.8 Return to Main Page:

⁴ <https://arxiv.org/abs/1903.08599/>

77 State Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³
4. Stabilization of Second-Order Systems⁴
5. LQ Regulation via H2 Control⁵
6. Controller to achieve the desired Reachable set; Polytopic uncertainty⁶
7. Controller to achieve the desired Reachable set; Norm bound uncertainty⁷
8. Controller to achieve the desired Reachable set; Diagonal Norm-bound uncertainty⁸

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>

2 <https://en.wikibooks.org/wiki/%2FH-2%2F>

3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

4 <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2Fpages%2Fstabilization%20of%20second%20order%20systems>

5 <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2Fpages%2FLQ%20Regulation%20via%20H2%20Control>

6 <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2Fpages%2FReachable%20set%20polytopic>

7 <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2Fpages%2FReachable%20set%20normbounded>

8 <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2Fpages%2FReachable%20set%20diagonalNB>

78 D-Stability

1. Continuous Time D-Stability Controller¹

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Continuous_Time_D-Stability_Controller

79 Optimal State Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>
2 <https://en.wikibooks.org/wiki/%2FH-2%2F>
3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

80 Output Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>
2 <https://en.wikibooks.org/wiki/%2FH-2%2F>
3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

81 Static Output Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³
4. Continuous-Time Static Output Feedback Stabilizability⁴

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>

2 <https://en.wikibooks.org/wiki/%2FH-2%2F>

3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/CT-SOFS

82 Optimal Output Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>
2 <https://en.wikibooks.org/wiki/%2FH-2%2F>
3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

83 Stabilizability LMI

Stabilizability LMI

A system is stabilizable if all unstable modes of the system are controllable. This implies that if the system is controllable, it will also be stabilizable. Thus, stabilizability is a essentially a weaker version of the controllability condition. The LMI condition for stabilizability of pair (A, B) is shown below.

83.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ x(0) &= x_0,\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^m$, at any $t \in \mathbb{R}$.

83.2 The Data

The matrices necessary for this LMI are A and B . There is no restriction on the stability of A .

83.3 The LMI: Stabilizability LMI

(A, B) is stabilizable if and only if there exists $X > 0$ such that

$$AX + XA^T + BB^T << 0$$

where the stabilizing controller is given by

$$u(t) = -\frac{1}{2}B^T X^{-1}x(t)$$

83.4 Conclusion:

If we are able to find $X > 0$ such that the above LMI holds it means the matrix pair (A, B) is stabilizable. In words, a system pair (A, B) is stabilizable if for any

initial state $x(0) = x_0$ an appropriate input $u(t)$ can be found so that the state $x(t)$ asymptotically approaches the origin. Stabilizability is a weaker condition than controllability in that we only need to *approach* $x(t) = 0$ as $t \rightarrow \infty$ whereas controllability requires that the state must *reach* the origin in a *finite* time.

83.5 Implementation

This implementation requires Yalmip and Sedumi.

https://github.com/eoskowro/LMI/blob/master/Stabilizability_LMI.m

83.6 Related LMIs

Hurwitz Stability LMI¹

Detectability LMI²

Controllability Grammian LMI³

Observability Grammian LMI⁴

83.7 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁵ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁶ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁷ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications⁸ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 6.1.1 and Table 6.1 pp. 166–170, 192.
- A Course in Robust Control Theory: a Convex Approach,⁹ - by Geir E. Dullerud and Fernando G. Paganini, Springer, 2011, Section 2.2.3, pp. 71–73.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Hurwitz_Stability

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Detectability_LMI

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_the_Controllability_Grammian

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_the_Observability_Grammian

5 http://control.asu.edu/MAE598_frame.htm

6 <https://arxiv.org/abs/1903.08599/>

7 <https://web.stanford.edu/~boyd/lmibook/>

8 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

9 <https://link.springer.com/book/10.1007/978-1-4757-3290-0#toc>

83.8 Return to Main Page:

84 LMI for the Controllability Grammian

LMI to Find the Controllability Grammian

Being able to adjust a system in a desired manor using feedback and sensors is a very important part of control engineering. However, not all systems are able to be adjusted. This ability to be adjusted refers to the idea of a "controllable" system and motivates the necessity of determining the "controllability" of the system. Controllability refers to the ability to accurately and precisely manipulate the state of a system using inputs. Essentially if a system is controllable then it implies that there is a control law that will transfer a given initial state $x(t_0) = x_0$ and transfer it to a desired final state $x(t_f) = x_f$. There are multiple ways to determine if a system is controllable, one of which is to compute the rank "controllability grammian". If the grammian is full rank, the system is controllable and a state transferring control law exists.

84.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ x(0) &= x_0,\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^m$, at any $t \in \mathbb{R}$.

84.2 The Data

The matrices necessary for this LMI are A and B . A must be stable for the problem to be feasible.

84.3 The LMI: LMI to Determine the Controllability Grammian

(A, B) is controllable if and only if $W > 0$ is the unique solution to

$$AW + WA^T - BB^T < 0$$

where W is the Controllability Grammian.

84.4 Conclusion:

The LMI above finds the controllability grammian W of the system (A, B) . If the problem is feasible and a unique W can be found, then we also will be able to say the system is controllable. The controllability grammian of the system (A, B) can also be computed as: $W = \int_0^{\infty} e^{As} BB^T e^{A^T s} ds$, with control law $u(t) = B^T W^{-1} x(t)$ that will transfer the given initial state $x(t_0) = x_0$ to a desired final state $x(t_f) = x_f$.

84.5 Implementation

This implementation requires Yalmip and Sedumi.

https://github.com/eoskowro/LMI/blob/master/Controllability_Gram_LMI.m

84.6 Related LMIs

Stabilizability LMI¹

Hurwitz Stability LMI²

Detectability LMI³

Observability Grammian LMI⁴

84.7 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁵ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁶ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁷ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications⁸ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 6.1.1 and Table 6.1 pp. 166–170, 192.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Stabilizability_LMI

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Hurwitz_Stability

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Detectability_LMI

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_the_Observability_Grammian

5 http://control.asu.edu/MAE598_frame.htm

6 <https://arxiv.org/abs/1903.08599/>

7 <https://web.stanford.edu/~boyd/lmibook/>

8 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

- A Course in Robust Control Theory: a Convex Approach,⁹ - by Geir E. Dullerud and Fernando G. Paganini, Springer, 2011, Section 2.2.3, pp. 71-73.

84.8 Return to Main Page:

⁹ <https://link.springer.com/book/10.1007/978-1-4757-3290-0#toc>

85 LMI for Decentralized Feedback Control

LMI for Decentralized Feedback Control

In large-scale systems like a multi-agent robotic system, national economies, or chemical refineries, an actuator should act based on local information, which necessitates a decentralized or distributed control strategy. In a decentralized control framework, the controllers are distributed and each controller has only access to a subset of local measurements. We describe LMI formulations for a general decentralized control framework and then provide an illustrative example of a decentralized control design.

85.1 The System

In a decentralized controller design, the state feedback controller $u = Kx$ can be divided into n sub-controllers $u_i = K_i x_i, \quad i = 1, 2, \dots, n$.

85.2 The Data

A general state space representation of a linear time-invariant system is as follows:

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

where x is a $n \times n$ vector of state variables, B is the input matrix, C is the output matrix, and D is called the feedforward matrix. We assume that all the four matrices, A , B , C , and D are given.

85.3 The Optimization Problem

We aim to solve the H_∞ -optimal full-state feedback control problem using a controller $u = Kx$.

In a decentralized fashion, the control input u can be divided into sub-controllers u_1, u_2, \dots, u_j and can be written as $u = [u_1 \ u_2 \ \dots \ u_j]_{1 \times n}^T$.

For instance, let $j = 3$ and $n = 6$. Thus, there are three control inputs u_1, u_2 , and u_3 . We also assume that $u_{\{1\}}$ only depends on the first and the second states, while

u_2 and u_3 only depend on third to sixth states. For this example, the controller gain matrix can be described by:

$$K = \begin{bmatrix} k_1 & k_2 & 0 & 0 & 0 & 0 \\ 0 & 0 & k_3 & k_4 & k_5 & k_6 \\ 0 & 0 & k_7 & k_8 & k_9 & k_{10} \end{bmatrix}$$

Thus, the decentralized controller gain consists of sub-matrices of gains.

85.4 The LMI: LMI for decentralized feedback controller

The mathematical description of the LMI formulation for a decentralised optimal full-state feedback controller can be described by:

$$\min \gamma$$
$$\begin{bmatrix} YA^T + AY + Z^T B_2^T + B_2 Z & *^T & *^T \\ B_1^T & -\gamma I & *^T \\ YC_1^T + Z^T D_{12} & D_{11} & -\gamma I \end{bmatrix}$$

where $Y > 0$ is a positive definite matrix and Z such that the aforementioned constraints in LMIs are satisfied.

85.5 Conclusion:

The controller gain matrix is defined as:

$$K = \begin{bmatrix} 0 & 0 \\ 0 & F \end{bmatrix}$$

where F can be found after solving the LMIs and obtaining the variables matrices Y and Z . Thus,

$$F = ZY^{-1}.$$

85.6 Implementation

A link to Matlab codes for this problem in the Github repository:

https://github.com/asalimil/LMI_for_decentralized_feedback_controller/tree/master

85.7 Related LMIs

85.8 External Links

A list of references documenting and validating the LMI.

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

85.9 Return to Main Page

LMIs in Control/Tools¹

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

86 LMI for Mixed H_2/H_∞ Output Feedback Controller

LMI for Mixed H_2/H_∞ Output Feedback Controller

The mixed H_2/H_∞ output feedback control has been known as an example of a multi-objective optimal control problem. In this problem, the control feedback should respond properly to several specifications. In the H_2/H_∞ controller, the H_∞ channel is used to improve the robustness of the design while the H_2 channel guarantees good performance of the system.

86.1 The System

We consider the following state-space representation for a linear system:

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

where A , B , C , and D are the state matrix, input matrix, output matrix, and feedforward matrix, respectively.

These are the system (plant) matrices that can be shown as $P = (A, B, C, D)$.

86.2 The Data

We assume that all the four matrices of the plant, A, B, C, D , are given.

86.3 The Optimization Problem

In this problem, we use an LMI to formulate and solve the optimal output-feedback problem to minimize both the $\langle \cdot \rangle$ and $\langle \cdot \rangle$ norms. Giving equal weights to each of the norms, we will have the optimization problem in the following form:

$$\min \quad \|S(P, K)\|_{H_2}^2 + \|S(P, K)\|_{H_\infty}^2$$

86.4 The LMI: LMI for mixed H_2/H_∞

Mathematical description of the LMI formulation for a mixed H_2/H_∞ optimal output-feedback problem can be written as follows:

$$\min \quad \gamma_1^2 + \gamma_2^2$$

s.t.

$$\begin{bmatrix} X_1 & I \\ I & Y_1 \end{bmatrix} > 0$$

$$\begin{bmatrix} AY_1 + Y_1A^\top + B_2C_n + C_nB_2^\top & *^\top & *^\top & *^\top \\ A^\top + A_n + (B_2D_nC_2)^\top & X_1A + A^\top + B_nC_2 + C_2^\top B_n^\top & *^\top & *^\top \\ (B_1 + B_2D_nD_{21})^\top & (X_1B_1 + B_nD_{21})^\top & -\gamma I & *^\top \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix} < 0$$

$$\begin{bmatrix} Y_1 & I & (C_1Y_1 + D_{12}C_n)^\top & \\ I & X_1 & (C_1 + D_{12}D_nC_2)^\top & \\ (C_1Y_1 + D_{12}C_n) & (C_1 + D_{12}D_nD_{21}) & Z & \\ C_1Y_1 + D_{12}C_n & C_1 + D_{12}D_nC_2 & D_{11} + D_{12}D_nD_{21} & -\gamma I \end{bmatrix} > 0$$

$$\begin{bmatrix} AY_1 + Y_1A^\top + B_2C_n + C_nTB_2^\top & *^\top & *^\top & *^\top \\ (A^\top + A_n + (B_2 * D_n * C_2)^\top) & X_1A + A^\top X_1 + B_nC_2 + C_2^\top B_n^\top & *^\top & *^\top \\ (B_1 + B_2D_nD_{21})^\top & (X_1B_1 + B_nD_{21})^\top & -\gamma_2^2 I & *^\top \\ (C_1Y_1 + D_{12}C_n) & (C_1 + D_{12}D_nC_2) & (D_{11} + D_{12}D_n * D_{21}) & -I \end{bmatrix} < 0$$

$$\text{trace}(Z) < \gamma_1^2$$

$$D_{11} + D_{12}D_nD_{21} = 0$$

where γ_1^2 and γ_2^2 are defined as the H_2 and H_∞ norm of the system:

$$\|S(P, K)\|_{H_2}^2 = \gamma_1^2$$

$$\|S(P, K)\|_{H_\infty}^2 = \gamma_2^2$$

Moreover, X_1 , Y_1 , A_n , B_n , C_n , and D_n are variable matrices with appropriate dimensions that are found after solving the LMIs.

86.5 Conclusion:

The calculated scalars γ_1^2 and γ_2^2 are the H_2 and H_∞ norms of the system, respectively. Thus, the norm of mixed H_2/H_∞ is defined as $\beta = \gamma_1^2 + \gamma_2^2$. The results for each individual H_2 norm and H_∞ norms of the system show that a bigger value of norms are found in comparison with the case they are solved separately.

86.6 Implementation

A link to Matlab codes for this problem in the Github repository:

https://github.com/asalimil/LMI_for_Mixed_H2_Hinf_Output_Feedback_Controller

86.7 Related LMIs

86.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

86.9 Return to Main Page

LMIs in Control/Tools¹

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

87 Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty

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If the system is quadratically stable, then there exists some $\mu \geq 0, P > 0$, and Z such that the LMI is feasible. The Z and P matrices can also be used to create a quadratically stabilizing controller.

87.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + Mp(t), & p(t) &= \Delta(t)q(t), \\ q(t) &= Nx(t) + Qp(t) + D_{12}u(t), & \Delta \in \Delta &:= \{\Delta \in \mathbb{R}^{n \times n} : \|\Delta\| \leq 1\}\end{aligned}$$

87.2 The Data

The matrices A, B, M, N, Q, D_{12} .

87.3 The LMI:

$$\text{Find } P > 0, \mu \geq 0, \text{ and } Z : \quad \begin{bmatrix} AP + BZ + PA^T + Z^T B^T & PN^T + Z^T D_{12}^T \\ NP + D_{12}Z & 0 \end{bmatrix} + \mu \begin{bmatrix} MM^T & MQ^T \\ QM^T & QQ^T - I \end{bmatrix} < 0$$

87.4 Conclusion:

There exists a controller for the system with $u(t) = Kx(t)$ where $K = ZP^{-1}$ is the quadratically stabilizing controller, if the above LMI is feasible.

87.5 Implementation

<https://github.com/mcavorsi/LMI>

87.6 Related LMIs

H-infinity Optimal Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty¹

Stabilizing State-Feedback Controllers with Structured Norm-Bounded Uncertainty²

Optimal State-Feedback Controllers with Structured Norm-Bounded Uncertainty³

87.7 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

87.8 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/H-inf_Optimal_Quadratically_Stabilizing.Controllers_with_Parametric_Norm-Bounded_Uncertainty
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Stabilizing_State-Feedback.Controllers_with_Structured_Norm-Bounded_Uncertainty
3 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Optimal_State-Feedback.Controllers_with_Structured_Norm-Bounded_Uncertainty
4 http://control.asu.edu/MAE598_frame.htm

88 H-inf Optimal Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty

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If there exists some $\mu \geq 0, P > 0$, and Z such that the LMI holds, then the system satisfies $\|L_2 \leq \gamma\|_{L_2}$. There also exists a controller with $u(t) = Kx(t)$.

88.1 The System

$$\begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t) + Mp(t) + B_2w(t), & p(t) &= \Delta(t)q(t), \\ q(t) &= Nx(t) + D_{12}u(t), & \Delta \in \Delta &:= \{\Delta \in \mathbb{R}^{n \times n} : \|\Delta\| \leq 1\} \\ y(t) &= Cx(t) + D_{22}u(t) \end{aligned}$$

88.2 The Data

The matrices $A, B, M, B_2, N, D_{12}, C, D_{22}$.

88.3 The Optimization Problem

Minimize γ subject to the LMI constraints below.

88.4 The LMI:

$$\text{Find } P > 0, \mu \geq 0, \text{ a } Z \text{ s.t.}$$

$$\begin{bmatrix} AP + BZ + PA^T + Z^T B^T + B_2 B_2^T + \mu M M^T & (CP + D_{22}Z)^T & PN^T + Z^T D_{12}^T \\ CP + D_{22}Z & -\gamma^2 I & 0 \\ NP + D_{12}Z & 0 & -\mu I \end{bmatrix} < 0$$

88.5 Conclusion:

The controller gains, K , are calculated by $K = ZP^{-1}$.

88.6 Implementation

<https://github.com/mcavorsi/LMI>

88.7 Related LMIs

Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty¹

Stabilizing State-Feedback Controllers with Structured Norm-Bounded Uncertainty²

Optimal State-Feedback Controllers with Structured Norm-Bounded Uncertainty³

88.8 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

88.9 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Quadratically_Stabilizing.Controllers_with_Parametric_Norm-Bounded_Uncertainty
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Stabilizing.State-Feedback.Controllers_with_Structured_Norm-Bounded_Uncertainty
3 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Optimal.State-Feedback.Controllers_with_Structured_Norm-Bounded_Uncertainty
4 http://control.asu.edu/MAE598_frame.htm

89 Stabilizing State-Feedback Controllers with Structured Norm-Bounded Uncertainty

UNKNOWN TEMPLATE FULLPAGENAME

The system is quadratically stable if and only if there exists some $\Theta \in P\Theta, P > 0$, and Z such that

89.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + Mp(t), & p(t) &= \Delta(t)q(t), \\ q(t) &= Nx(t) + Qp(t) + D_{12}u(t), & \Delta \in \Delta, \|\Delta\| \leq 1\end{aligned}$$

89.2 The Data

The matrices A, B, M, N, Q, D_{12} .

89.3 The LMI:

$$\text{Find } P > 0, Z : \quad \begin{bmatrix} AP + BZ + PA^T + Z^T B^T & PN^T + Z^T D_{12}^T \\ NP + D_{12}Z & 0 \end{bmatrix} + \begin{bmatrix} M\Theta M^T & M\Theta Q^T \\ Q\Theta M^T & Q\Theta Q^T - \Theta \end{bmatrix} < 0$$

89.4 Conclusion:

If the LMI is feasible, the controller, K , is calculated by $K = ZP^{-1}$.

89.5 Implementation

<https://github.com/mcavorsi/LMI>

89.6 Related LMIs

Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty¹

H-infinity Optimal Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty²

Optimal State-Feedback Controllers with Structured Norm-Bounded Uncertainty³

89.7 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

89.8 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Quadratically_Stabilizing.Controllers_with_Parametric_Norm-Bounded_Uncertainty
https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/H-inf_Optimal_Quadratically_Stabilizing.Controllers_with_Parametric_Norm-Bounded_Uncertainty

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Optimal_State-Feedback.Controllers_with_Structured_Norm-Bounded_Uncertainty

3 http://control.asu.edu/MAE598_frame.htm

90 Optimal State-Feedback Controllers with Structured Norm-Bounded Uncertainty

UNKNOWN TEMPLATE FULLPAGENAME

If there exists some $\Theta \in P\Theta, P > 0$, and Z such that the LMI is feasible, then the system satisfies |

90.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + Mp(t) + B_2w(t), & p(t) &= \Delta(t)q(t), \\ q(t) &= Nx(t) + D_{12}u(t), & \Delta \in \Delta, \|\Delta\| \leq 1 \\ y(t) &= Cx(t) + D_{22}u(t)\end{aligned}$$

90.2 The Data

The matrices $A, B, M, B_2, N, D_{12}, C, D_{22}$.

90.3 The Optimization Problem

Minimize γ subject to the LMI constraints.

90.4 The LMI:

$$\text{Find } P > 0, Z : \quad \begin{bmatrix} AP + BZ + PA^T + Z^T B^T + B_2B_2^T + M\Theta M^T & (CP + D_{22}Z)^T & PN^T + Z^T D_{12}^T \\ CP + D_{22}Z & -\gamma^2 I & 0 \\ NP + D_{12}Z & 0 & -\Theta \end{bmatrix} < 0$$

90.5 Conclusion:

The controller is $K = ZP^{-1}$.

90.6 Implementation

<https://github.com/mcavorsi/LMI>

90.7 Related LMIs

Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty¹

H-infinity Optimal Quadratically Stabilizing Controllers with Parametric Norm-Bounded Uncertainty²

Stabilizing State-Feedback Controllers with Structured Norm-Bounded Uncertainty³

90.8 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.

90.9 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Quadratically_Stabilizing.Controllers_with_Parametric_Norm-Bounded_Uncertainty
https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/H-inf_Optimal_Quadratically_Stabilizing.Controllers_with_Parametric_Norm-Bounded_Uncertainty

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Controller_Synthesis/Continuous_Time/Stabilizing.State-Feedback.Controllers_with_Structured_Norm-Bounded_Uncertainty

3 http://control.asu.edu/MAE598_frame.htm

91 H_∞ Optimal Output Controllability for Systems With Transients

H_∞ Optimal Output Controllability for Systems With Transients

This LMI provides an H_∞ optimal output controllability problem to check if such controllers for systems with unknown exogenous disturbances and initial conditions can exist or not.

91.1 The System

$$\begin{aligned}\dot{x} &= Ax + B_1v + B_2u, x(0) = x_0, \\ z &= C_1x + D_{11}v + D_{12}u, \\ y &= C_2x + D_{21}v,\end{aligned}$$

where $x \in \mathbb{R}^n$ is the state, $v \in \mathbb{R}^r$ is the exogenous input, $u \in \mathbb{R}^m$ is the control input, $y \in \mathbb{R}^p$ is the measured output and $z \in \mathbb{R}^s$ is the regulated output.

91.2 The Data

System matrices $(A, B_1, B_2, C_1, C_2, D_{11}, D_{12}, D_{21}, D_{22})$ need to be known. It is assumed that $v \in L_2[0, \infty)$. N_1, N_2 are matrices with their columns forming the basis of kernels of C_2D_{21} and C_2D_{12} respectively.

91.3 The Optimization Problem

For a given γ , the following H_∞ condition needs to be fulfilled:

$$\gamma_w = \sup_{\|x\|_\infty + x_0^\top Rx_0 \neq 0} \frac{\|x\|_\infty}{(\|x\|_\infty + x_0^\top Rx_0)^{1/2}} < \gamma_w,$$

91.4 The LMI: H_∞ Output Feedback Controller for Systems With Transients

$\min_{\gamma, X_{11}, Y_{11}} : \gamma$
subj. to: $X_{11} > 0, Y_{11} > 0,$

$$\begin{bmatrix} N_1 & 0 \\ 0 & I \end{bmatrix}^\top \begin{bmatrix} A^\top X_{11} + X_{11}A & X_{11}B_1 & C_1^\top \\ * & -\gamma^2 I & D_{11}^\top \\ * & * & -I \end{bmatrix} \begin{bmatrix} N_1 & 0 \\ 0 & I \end{bmatrix} < 0,$$

$$\begin{bmatrix} N_2 & 0 \\ 0 & I \end{bmatrix}^\top \begin{bmatrix} AY_{11} + Y_{11}A^\top & Y_{11}C_1^\top & B_1 \\ * & -I & D_{11} \\ * & * & -\gamma^2 I \end{bmatrix} \begin{bmatrix} N_2 & 0 \\ 0 & I \end{bmatrix} < 0,$$

$$\begin{bmatrix} X_{11} & I \\ I & Y_{11} \end{bmatrix} \geq 0, X_{11} < \gamma^2 R,$$

91.5 Conclusion:

Solution of the above LMI gives a check to see if an H_∞ optimal output controller for systems with transients can exist or not.

91.6 Implementation

A link to CodeOcean or other online implementation of the LMI

91.7 Related LMIs

Links to other closely-related LMIs

91.8 External Links

- LMI-based H_∞ -optimal control with transients¹ Link to the original article.

91.9 Return to Main Page:

¹ <https://www.tandfonline.com/doi/pdf/10.1080/00207179.2010.487222?needAccess=true>

92 Quadratic Polytopic Stabilization

A Quadratic Polytopic Stabilization Controller Synthesis can be done using this LMI, requiring the information about $A, \Delta_{A(t)}, B$ and $\Delta_{B(t)}$ matrices.

92.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ x(0) &= x_0,\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^m$, at any $t \in \mathbb{R}$.

The system consist of uncertainties of the following form

$$\begin{aligned}\Delta_{A(t)} &= A_1\delta_1(t) + \dots + A_k\delta_k(t) \\ \Delta_{B(t)} &= B_1\delta_1(t) + \dots + B_k\delta_k(t)\end{aligned}$$

where $x \in \mathbb{R}^m, u \in \mathbb{R}^n, A \in \mathbb{R}^{mxm}$ and $B \in \mathbb{R}^{mxn}$

92.2 The Data

The matrices necessary for this LMI are $A, \Delta_{A(t)}$ ie A_i, B and $\Delta_{B(t)}$ ie B_i

92.3 The Optimization and LMI:LMI for Controller Synthesis using the theorem of Polytopic Quadratic Stability

There exists a K such that

$$\dot{x}(t) = (A + \Delta_A + (B + \Delta_B)K)x(t)$$

is quadratically stable for $(\Delta_A, \Delta_B) \in C_0((A_1, B_2), \dots, (A_k, B_k))$ if and only if there exists some $P > 0$ and Z such that

$$(A + A_i)P + P(A + A_i)^T + (B + B_i)Z + Z^T(B + B_i)^T < 0 \quad \text{for } i = 1, \dots, k.$$

92.4 Conclusion:

The Controller gain matrix is extracted as $K = ZP^{-1}$

Note that here the controller doesn't depend on Δ

- If you want K to depend on Δ , the problem is harder.
- But this would require sensing Δ in real-time.

92.5 Implementation

This implementation requires Yalmip and Sedumi. <https://github.com/JalpeshBhadra/LMI/blob/master/quadraticpolytopicstabilization.m>

92.6 Related LMIs

Quadratic Polytopic H_∞ Controller¹

Quadratic Polytopic H_2 Controller²

92.7 External Links

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Quadratic_Polytopic_Hinf_Optimal_State_Feedback_Control/
2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/quadratic_polytopic_h2_optimal_state_feedback_control
3 http://control.asu.edu/MAE598_frame.htm
4 <https://arxiv.org/abs/1903.08599>
5 <https://web.stanford.edu/~boyd/lmibook/>

93 Quadratic D-Stabilization

Continuous-Time D-Stability Controller

This LMI will let you place poles at a specific location based on system performance like rising time, settling time and percent overshoot, while also ensuring the stability of the system.

93.1 The System

Suppose we were given the continuous-time system

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

whose stability was not known, and where $A \in \mathbb{R}^{m \times m}$, $B \in \mathbb{R}^{m \times n}$, $C \in \mathbb{R}^{p \times m}$, and $D \in \mathbb{R}^{q \times n}$ for any $t \in \mathbb{R}$.

Adding uncertainty to the system

$$\dot{x}(t) = (A + A_i)x(t) + (B + B_i)u(t)$$

93.2 The Data

In order to properly define the acceptable region of the poles in the complex plane, we need the following pieces of data:

- matrices A , B , A_i , B_i
- rise time (t_r)
- settling time (t_s)
- percent overshoot (M_p)

Having these pieces of information will now help us in formulating the optimization problem.

93.3 The Optimization Problem

Using the data given above, we can now define our optimization problem. In order to do that, we have to first define the acceptable region in the complex plane that the poles can lie on using the following inequality constraints:

$$\text{Rise Time: } \omega_n \leq \frac{1.8}{t_r}$$

$$\text{Settling Time: } \sigma \leq \frac{-4.6}{t_s}$$

$$\text{Percent Overshoot: } \sigma \leq \frac{-\ln(M_p)}{\pi} |\omega_d|$$

Assume that z is the complex pole location, then:

$$\begin{aligned} \omega_n^2 &= \|z\|^2 = z^* z \\ \omega_d &= \operatorname{Im} z = \frac{(z - z^*)}{2} \\ \sigma &= \operatorname{Re} z = \frac{(z + z^*)}{2} \end{aligned}$$

This then allows us to modify our inequality constraints as:

$$\text{Rise Time: } z^* z - \frac{1.8^2}{t_r^2} \leq 0$$

$$\text{Settling Time: } \frac{(z+z^*)}{2} + \frac{4.6}{t_s} \leq 0$$

$$\text{Percent Overshoot: } z - z^* + \frac{\pi}{\ln(M_p)} |z + z^*| \leq 0$$

which not only allows us to map the relationship between complex pole locations and inequality constraints but it also now allows us to easily formulate our LMIs for this problem.

93.4 The LMI: An LMI for Quadratic D-Stabilization

Suppose there exists $X > 0$ and Z such that

$$\begin{bmatrix} -rP & AP+BZ \\ (AP+BZ)^T & -rP \end{bmatrix} + \begin{bmatrix} 0 & A_i P + B_i Z \\ (A_i P + B_i Z)^T & 0 \end{bmatrix} < 0$$

$$AP + BZ + (AP + BZ)^T + A_i P + B_i Z + (A_i P + B_i Z)^T + 2\alpha P < 0, \text{ and}$$

$$\begin{bmatrix} AP + BZ + (AP + BZ)^T & c(AP + BZ - (AP + BZ)^T) \\ c((AP + BZ)^T - (AP + BZ)) & AP + BZ + (AP + BZ)^T \end{bmatrix} + \begin{bmatrix} A_i P + B_i Z + (A_i P + B_i Z)^T & c((A_i P + B_i Z)^T - (A_i P + B_i Z)) \\ c((A_i P + B_i Z)^T - (A_i P + B_i Z)) & AP + BZ + (AP + BZ)^T \end{bmatrix} < 0$$

for $i = 1, \dots, k$

93.5 Conclusion:

Given the resulting controller $K = ZP^{-1}$, we can now determine that the pole locations $z \in \mathbb{C}$ of $A(\Delta) + B(\Delta)K$ satisfies the inequality constraints $|x| \leq r$, $\text{Re}(x) \leq -\alpha$ and $z + z^* \leq -c|z - z^*|$ for all $\Delta \in C_0(\Delta_1, \dots, \Delta_k)$

93.6 Implementation

The implementation of this LMI requires Yalmip and Sedumi <https://github.com/JalpeshBhadra/LMI/blob/master/quadraticDstabilization.m>

93.7 Related LMIs

- [.. / Continuous Time D-Stability Observer](#)¹ - Equivalent D-stability LMI for a continuous-time observer.

93.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

93.9 Return to Main Page:

1 <https://en.wikibooks.org/wiki/..%2FContinuous%20Time%20D-Stability%20Observer%2F>

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599/>

4 <https://web.stanford.edu/~boyd/lmibook/>

94 Quadratic Polytopic Full State Feedback Optimal H_∞ Control

94.1 Quadratic Polytopic Full State Feedback Optimal H_∞ Control

For a system having polytopic uncertainties, Full State Feedback is a control technique that attempts to place the system's closed-loop system poles in specified locations based off of performance specifications given. H_∞ methods formulate this task as an optimization problem and attempt to minimize the H_∞ norm of the system.

94.2 The System

Consider System with following state-space representation.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1q(t) + B_2w(t) \\ p(t) &= C_1x(t) + D_{11}q(t) + D_{12}w(t) \\ z(t) &= C_2x(t) + D_{21}q(t) + D_{22}w(t)\end{aligned}$$

where $x \in \mathbb{R}^m$, $q \in \mathbb{R}^n$, $w \in \mathbb{R}^g$, $A \in \mathbb{R}^{mxm}$, $B_1 \in \mathbb{R}^{mxn}$, $B_2 \in \mathbb{R}^{mxg}$, $p \in \mathbb{R}^p$, $C_1 \in \mathbb{R}^{pxm}$, $D_{11} \in \mathbb{R}^{pxn}$, $D_{12} \in \mathbb{R}^{pxg}$, $z \in \mathbb{R}^s$, $C_2 \in \mathbb{R}^{sxm}$, $D_{21} \in \mathbb{R}^{sxn}$, $D_{22} \in \mathbb{R}^{sxt}$ for any $t \in \mathbb{R}$.

Add uncertainty to system matrices

$$A, B_1, B_2, C_1, C_2, D_{11}, D_{12}$$

New state-space representation

$$\begin{aligned}\dot{x}(t) &= (A + A_i)x(t) + (B_1 + B_i)q(t) + (B_2 + B_i)w(t) \\ p(t) &= (C_1 + C_i)x(t) + (D_{11} + D_i)q(t) + (D_{12} + D_i)w(t) \\ z(t) &= C_2x(t) + D_{21}q(t) + D_{22}w(t)\end{aligned}$$

94.3 The Optimization Problem:

Recall the closed-loop in state feedback is:

$$S(P, K) = \begin{bmatrix} A + B_2 F & B_1 \\ C_1 + D_{12} F & D_{11} \end{bmatrix}$$

This problem can be formulated as H_∞ optimal state-feedback, where K is a controller gain matrix.

94.4 The LMI:

An LMI for Quadratic Polytopic H_∞ Optimal State-Feedback Control
 $\|S(P(\Delta), K(0, 0, 0, F))\|_{H_\infty} \leq \gamma$

$$Y > 0$$

$$\begin{bmatrix} Y(A + A_i)^T + (A + A_i)Y + Z^T(B_2 + B_{1,i})^T + (B_2 + B_{1,i})Z & *^T & *^T \\ (B_1 + B_{1,i})^T & -\gamma I & *^T \\ (C_1 + C_{1,i})Y + (D_{12} + D_{12,i})Z & (D_{11} + D_{11,i}) & -\gamma I \end{bmatrix} < 0$$

94.5 Conclusion:

The H_∞ Optimal State-Feedback Controller is recovered by $F = ZY^{-1}$

Controller will determine the bound γ on the H_∞ norm of the system.

94.6 Implementation:

<https://github.com/JalpeshBhadra/LMI/tree/master>

94.7 Related LMIs

Full State Feedback Optimal H_∞ Controller¹

¹ https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Full-State_Feedback_Optimal_Control_Hinf_LMI&action=submit

94.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

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2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599>

4 <https://web.stanford.edu/~boyd/lmibook/>

95 Quadratic Polytopic Full State Feedback Optimal H_2 Control

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95.1 Quadratic Polytopic Full State Feedback Optimal H_2 Control

For a system having polytopic uncertainties, Full State Feedback is a control technique that attempts to place the system's closed-loop system poles in specified locations based on performance specifications given, such as requiring stability or bounding the overshoot of the output. By minimizing the H_2 norm of this system we are minimizing the effect noise has on the system as part of the performance specifications.

95.2 The System

Consider System with following state-space representation.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1q(t) + B_2w(t) \\ p(t) &= C_1x(t) + D_{11}q(t) + D_{12}w(t) \\ z(t) &= C_2x(t) + D_{21}q(t) + D_{22}w(t)\end{aligned}$$

where $x \in \mathbb{R}^m$, $q \in \mathbb{R}^n$, $w \in \mathbb{R}^g$, $A \in \mathbb{R}^{mxm}$, $B_1 \in \mathbb{R}^{mxn}$, $B_2 \in \mathbb{R}^{mxg}$, $p \in \mathbb{R}^p$, $C_1 \in \mathbb{R}^{pxm}$, $D_{11} \in \mathbb{R}^{pxn}$, $D_{12} \in \mathbb{R}^{pxg}$, $z \in \mathbb{R}^s$, $C_2 \in \mathbb{R}^{sxm}$, $D_{21} \in \mathbb{R}^{sxn}$, $D_{22} \in \mathbb{R}^{sxg}$ for any $t \in \mathbb{R}$.

Add uncertainty to system matrices

$$A, B_1, B_2, C_1, C_2, D_{11}, D_{12}$$

New state-space representation

$$\begin{aligned}\dot{x}(t) &= (A + A_i)x(t) + (B_1 + B_i)q(t) + (B_2 + B_i)w(t) \\ p(t) &= (C_1 + C_i)x(t) + (D_{11} + D_i)q(t) + (D_{12} + D_i)w(t) \\ z(t) &= C_2x(t) + D_{21}q(t) + D_{22}w(t)\end{aligned}$$

95.3 The Data

The matrices necessary for this LMI are

95.4 The Optimization Problem:

Recall the closed-loop in state feedback is:

$$S(P, K) = \begin{bmatrix} A + B_{22}F & B_1 \\ C_1 + D_{12}F & D_{11} \end{bmatrix}$$

This problem can be formulated as H_2 optimal state-feedback, where K is a controller gain matrix.

95.5 The LMI: An LMI for Quadratic Polytopic H_2 Optimal State-Feedback Control

$$\|S(P(\Delta), K(0, 0, 0, F))\|_{H_2} \leq \gamma$$

$$X > 0$$

$$\begin{bmatrix} AX + B_2Z + XA^T + Z^T B_2^T & B_1 \\ B_1^T & -I \end{bmatrix} + \begin{bmatrix} A_i X + B_{2,i}Z + XA_i^T + Z^T B_{2,I}^T & B_{1,i} \\ B_{1,i}^T & 0 \end{bmatrix} < 0 \quad i = 1, \dots, k$$

$$\begin{bmatrix} X & (C_1 X + D_{12}Z)^T \\ C_1 X + D_{12}Z & W \end{bmatrix} + \begin{bmatrix} 0 & (C_{1,i} X + D_{12,i}Z)^T \\ C_{1,i} X + D_{12,i}Z & 0 \end{bmatrix} > 0 \quad i = 1, \dots, k$$

$$\text{Trace}W < \gamma$$

95.6 Conclusion:

The H_2 Optimal State-Feedback Controller is recovered by $F = ZX^{-1}$

95.7 Implementation:

https://github.com/JalpeshBhadra/LMI/blob/master/H2_optimal_statefeedback_controller.m

95.8 Related LMIs

H_2 Optimal State-Feedback Controller¹

95.9 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

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1 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Full-State_Feedback_Optimal_Control_H2_LMI&stable=0
2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599/>
4 <https://web.stanford.edu/~boyd/lmibook/>

96 Continuous-Time Static Output Feedback Stabilizability

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In view of applications, static feedback of the full state is not feasible in general: only a few of the state variables (or a linear combination of them, $y = Cx(t)$, called the output) can be actually measured and re-injected into the system.

So, we are led to the notion of static output feedback

96.1 The System

Consider the continuous-time LTI system, with generalized state-space realization $(A, B, C, 0)$

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

96.2 The Data

- $A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, C \in \mathbb{R}^{p \times n}$
- $x \in \mathbb{R}^n, y \in \mathbb{R}^p, u \in \mathbb{R}^m$

96.3 The Optimization Problem

This system is static output feedback stabilizable (SOFS) if there exists a matrix F such that the closed-loop system

$$\dot{x} = (A - BKC)x$$

(obtained by replacing $u = -Ky$ which means applying static output feedback) is asymptotically stable at the origin

96.4 The LMI: LMI for Continuous Time - Static Output Feedback Stabilizability

The system is static output feedback stabilizable if and only if it satisfies any of the following conditions:

- There exists a $K \in \mathbb{R}^{m \times p}$ and $P \in \mathbb{S}^n$, where $P > 0$, such that

$$\begin{bmatrix} A^T P + PA - PBB^T P & PB + C^T K^T \\ KC + B^T P & -1 \end{bmatrix} < 0$$

- There exists a $K \in \mathbb{R}^{m \times p}$ and $Q \in \mathbb{S}^n$, where $Q > 0$, such that

$$\begin{bmatrix} QA^T + AQ - QC^T CQ & BK + QC^T \\ CQ^T + K^T B^T & -1 \end{bmatrix} < 0$$

- There exists a $K \in \mathbb{R}^{m \times p}$ and $Q \in \mathbb{S}^n$, where $Q > 0$, such that

$$\begin{bmatrix} QA^T + AQ - BB^T & B + QC^T K^T \\ B^T + KCQ^T & -1 \end{bmatrix} < 0$$

- There exists a $K \in \mathbb{R}^{m \times p}$ and $P \in \mathbb{S}^n$, where $P > 0$, such that

$$\begin{bmatrix} A^T P + PA - C^T C & PBK + C^T \\ K^T B^T P & -1 \end{bmatrix} < 0$$

96.5 Conclusion

On implementation and optimization of the above LMI using YALMIP and MOSEK (or) SeDuMi we get 2 output matrices one of which is the Symmetric matrix P (or Q) and K

96.6 Implementation

A link to the Matlab code for a simple implementation of this problem in the Github repository:

https://github.com/yashgvd/LMI_wikibooks

96.7 Related LMIs

Discrete time Static Output Feedback Stabilizability

Static Feedback Stabilizability

96.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- D. d. S. Madeira and J. Adamy, "Static output feedback: An LMI condition for stabilizability based on passivity indices," 2016 IEEE Conference on Control Applications (CCA), Buenos Aires, 2016, pp. 960-965.

96.9 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599/>

97 Multi-Criterion LQG

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The Multi-Criterion Linear Quadratic Gaussian (LQG) linear matrix inequality will allow one to form an optimized controller, similar to that in an LQR framework, for a state space system with gaussian noise based on several different criterions defined in the Q and R matrices, that are optimized as a part of the arbitrary cost function. Just like traditional LQR, the cost matrices must be tuned in much a similar fashion as traditional gains in classical control. In the LQR and LQG framework however, the gains are more intuitive as each correlates directly to a state or an input.

97.1 The System

The system is a linear time-invariant system, that can be represented in state space as shown below:

$$\begin{aligned}\dot{x} &= Ax + Bu + w, \\ y &= Cx + v, \\ z &= \begin{bmatrix} Q^{1/2} & 0 \\ 0 & R^{1/2} \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix}\end{aligned}$$

where $x \in R^n$, $y \in R^l$, $z \in R^m$ represent the state vector, the measured output vector, and the output vector of interest, respectively, $w \in R^p$ is the disturbance vector, and A, B, C, Q, R are the system matrices of appropriate dimension. To further define: x is $\in R^n$ and is the state vector, A is $\in R^{n*n}$ and is the state matrix, B is $\in R^{n*r}$ and is the input matrix, w is $\in R^r$ and is the exogenous input, C, Q, R is $\in R^{m*n}$ and are the output matrices, and y and z are $\in R^m$ and are the output and the output of interest, respectively.

$Q \geq 0$ and $R > 0$, and the system is controllable and observable.

97.2 The Data

The matrices A, B, C, Q, R, W, V and the noise signals w, v .

97.3 The Optimization Problem

In the Linear Quadratic Gaussian (LQG) control problem, the goal is to minimize a quadratic cost function while the plant has random initial conditions and suffers white noise disturbance on the input and measurement.

There are multiple outputs of interest for this problem. They are defined by

$$z = \begin{bmatrix} Q^{1/2} & 0 \\ 0 & R^{1/2} \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix}, Q_i \geq 0, R_i > 0, i = 0, \dots, p.$$

For each of these outputs of interest, we associate a cost function:

$$J_{LQG}^i = \lim_{t \rightarrow \infty} E z_i(t)^T z_i(t), i = 0, \dots, p.$$

Additionally, the matrices X_{LQG} and Y_{LQG} must be found as the solutions to the following Riccati equations:

$$\begin{aligned} A^T X_{LQG} + X_{LQG} A &= X_{LQG} B R^{-1} B^T X_{LQG} + Q = 0 \\ A Y_{LQG} + Y_{LQG} A^T - Y_{LQG} C^T V^{-1} C Y_{LQG} + W &= 0 \end{aligned}$$

The optimization problem is to minimize J_{LQG}^0 over u subject to the measurability condition and the constraints $J_{LQG}^i < \gamma_i, i = 0, \dots, p..$ This optimization problem can be formulated as:

$$\max \operatorname{trace}(X_{LQG} U + Q Y_{LQG}) - \sum_{i=1}^p \gamma_i \tau_i,$$

over τ_1, \dots, τ_p , with:

$$\begin{aligned} Q &= Q_0 + \sum_{i=1}^p \tau_i Q_i, \\ R &= R_0 + \sum_{i=1}^p \tau_i R_i. \end{aligned}$$

97.4 The LMI: Multi-Criterion LQG

$$\max : \operatorname{trace}(X U + (Q_0 + \sum_{i=1}^p \tau_i Q_i) Y_{LQG}) - \sum_{i=1}^p \gamma_i \tau_i,$$

over X, τ_1, \dots, τ_p , subject to the following constraints:

$$\begin{aligned} X &> 0, \\ \tau_1 &\geq 0, \dots, \tau_p \geq 0, \\ A^T X + X A - X B (R_0 + \sum_{i=1}^p \tau_i R_i)^{-1} B^T X + Q_0 + \sum_{i=1}^p \tau_i Q_i &\geq 0. \end{aligned}$$

97.5 Conclusion:

The result of this LMI is the solution to the aforementioned Riccati equations:

$$\begin{aligned} A^T X_{LQG} + X_{LQG} A - X_{LQG} B R^{-1} B^T X_{LQG} + Q &= 0 \\ A Y_{LQG} + Y_{LQG} A^T - Y_{LQG} C^T V^{-1} C Y_{LQG} + W &= 0 \end{aligned}$$

97.6 Implementation

This implementation requires Yalmip and Sedumi.

<https://github.com/rezajamesahmed/LMImatlabcode/blob/master/multicriterionquadraticproblems.m>

97.7 Related LMIs

1. Inverse Problem of Optimal Control¹

97.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

97.9 Return to Main Page:

1 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Inverse_Problem_of_Optimal_Control
 2 http://control.asu.edu/MAE598_frame.htm
 3 <https://arxiv.org/abs/1903.08599/>
 4 <https://web.stanford.edu/~boyd/lmibook/>

98 Inverse Problem of Optimal Control

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In some cases, it is needed to solve the inverse problem of optimal control within an LQR framework. In this inverse problem, a given controller matrix needs to be verified for the system by assuring that it is the optimal solution to some LQR optimization problem that is controllable and detectable. In other words: in this inverse problem, the controller is known and the LQR gain matrices are to be calculated such that the controller is the optimal solution.

98.1 The System

The system is a linear time-invariant system, that can be represented in state space as shown below:

$$\begin{aligned}\dot{x} &= Ax + Bu, \\ z &= \begin{bmatrix} Q^{1/2} & 0 \\ 0 & R^{1/2} \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix}\end{aligned}$$

where $x \in R^n$, $y \in R^l$, $z \in R^m$ represent the state vector, the measured output vector, and the output vector of interest, respectively, $w \in R^p$ is the disturbance vector, and A, B, C, Q, R are the system matrices of appropriate dimension. To further define: x is $\in R^n$ and is the state vector, A is $\in R^{n*n}$ and is the state matrix, B is $\in R^{n*r}$ and is the input matrix, w is $\in R^r$ and is the exogenous input, C, Q, R is $\in R^{m*n}$ and are the output matrices, and y and z are $\in R^m$ and are the output and the output of interest, respectively.

98.2 The Data

The matrices A, B, C that define the system, and a given controller K for which the inverse problem is to be solved.

98.3 The Optimization Problem

In this LMI, the following cost function is to be minimized for a given controller K by finding an optimal input:

$$\int_0^\infty z^T z dt$$

the solution of the problem can be formulated as a state feedback controller given as:

$$K = -R^{-1}B^T P,$$
$$A^T P + PA - PBR^{-1}B^T P + Q = 0$$

98.4 The LMI: Inverse Problem of Optimal Control

the inverse problem of optimal control is the following: Given a matrix K , determine if there exist $Q \geq 0$ and $R > 0$, such that (Q, A) is detectable and $u = Kx$ is the optimal control for the corresponding LQR problem. Equivalently, we seek $R > 0$ and $Q \geq 0$ such that there exist P nonnegative and P_1 positive-definite satisfying

$$(A + BK)^T P + P(A + BK) + K^T RK + Q = 0$$
$$B^T P + RK = 0$$
$$A^T P_1 + P_1 A < Q$$

98.5 Conclusion:

If the solution exists, then K is the optimal controller for the LQR optimization on the matrices Q and R

98.6 Implementation

This implementation requires Yalmip and Sedumi.

<https://github.com/rezajamesahmed/LMImatlabcode/blob/master/inverseprob.m>

98.7 Related LMIs

1. Multi-Criterion LQG^{1]}

¹ https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Multi-Criterion_LQG&stable=0

98.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

98.9 Return to Main Page:

2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599>
4 <https://web.stanford.edu/~boyd/lmibook/>

99 Nonconvex Multi-Criterion Quadratic Problems

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The Non-Concex Multi-Criterion Quadratic linear matrix inequality will allow one to form an optimized controller, similar to that in an LQR framework, for a non-convex state space system based on several different criterions defined in the Q and R matrices, that are optimized as a part of the arbitrary cost function. Just like traditional LQR, the cost matrices must be tuned in much a similar fashion as traditional gains in classical control. In the LQR and LQG framework however, the gains are more intuitive as each correlates directly to a state or an input.

99.1 The System

The system for this LMI is a linear time invariant system that can be represented in state space as shown below:

$$\dot{x} = Ax + Bw, x(0) = x_0$$

where the system is assumed to be controllable.

where $x \in R^n$ represents the state vector, respectively, $w \in R^p$ is the disturbance vector, and A, B are the system matrices of appropriate dimension. To further define: x is $\in R^n$ and is the state vector, A is $\in R^{n*n}$ and is the state matrix, B is $\in R^{n*r}$ and is the input matrix, w is $\in R^r$ and is the exogenous input.

for any input, we define a set $p+1$ cost indices J_0, \dots, J_P by

$$J_i(u) = \int_0^\infty [x^T \ u^T] \begin{bmatrix} Q_i & C_i \\ C_i^T & R_i \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix} dt, \quad i = 0, \dots, p$$

Here the symmetric matrices,

$$\begin{bmatrix} Q_i & C_i \\ C_i^T & R_i \end{bmatrix}, i = 0, \dots, p$$

are not necessarily positive-definite.

99.2 The Data

The matrices A, B, C .

99.3 The Optimization Problem

The constrained optimal control problem is:

$$\max : J_0,$$

subject to

$$J_i \leq \gamma_i, i = 1, \dots, p, x \rightarrow 0, t \rightarrow \infty$$

99.4 The LMI: Nonconvex Multi-Criterion Quadratic Problems

The solution to this problem proceeds as follows: We first define

$$\begin{aligned} Q &= Q_0 + \sum_{i=1}^p \tau_i Q_i, \\ R &= R_0 + \sum_{i=1}^p \tau_i R_i, \\ C &= C_0 + \sum_{i=1}^p \tau_i C_i, \end{aligned}$$

where $\tau_i \geq 0$ and for every τ_i , we define

$$S = J_0 + \sum_{i=1}^p \tau_i J_i - \sum_{i=1}^p \tau_i \gamma_i$$

then, the solution can be found by:

$$\max : x(0)^T P x(0) - \sum_{i=1}^p \tau_i \gamma_i$$

subject to

$$\begin{bmatrix} A^T P + PA + Q & PQ + C^T \\ B^T P + C & R \end{bmatrix} \geq 0$$
$$\tau_i \geq 0$$

99.5 Conclusion:

If the solution exists, then K is the optimal controller and can be solved for via an EVP in P.

99.6 Implementation

This implementation requires Yalmip and Sedumi.

<https://github.com/rezajamesahmed/LMImatlabcode/blob/master/multicriterionquadraticproblems.m>

99.7 Related LMIs

1. Multi-Criterion LQG¹
2. Inverse Problem of Optimal Control²
3. Nonconvex Multi-Criterion Quadratic Problems³
4. Static-State Feedback Problem⁴

99.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁵ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁶ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁷ - A downloadable book on LMIs by Stephen Boyd.

99.9 Return to Main Page:

1 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Multi-Criterion_LQG&stable=0
 2 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/Inverse_Problem_of_Optimal_Control
 3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Nonconvex_Multi-Criterion_Quadratic_Problems
 4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/SSFP
 5 http://control.asu.edu/MAE598_frame.htm
 6 <https://arxiv.org/abs/1903.08599/>
 7 <https://web.stanford.edu/~boyd/lmibook/>

100 Static-State Feedback Problem

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We are attempting to stabilize The Static State-Feedback Problem

100.1 The System

Consider a continuous time Linear Time invariant system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

100.2 The Data

A, B are known matrices

100.3 The Optimization Problem

The Problem's main aim is to find a feedback matrix such that the system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

and

$$u(t) = Kx(t)$$

is stable Initially we find the K matrix such that $(A + BK)$ is Hurwitz.

100.4 The LMI: Static State Feedback Problem

This problem can now be formulated into an LMI as Problem 1:

$$X(A + BK) + (A + BK)^T X < 0$$

From the above equation $X > 0$ and we have to find K

The problem as we can see is bilinear in K, X

- The bilinear in X and K is a common paradigm
- Bilinear optimization is not Convex. To Convexify the problem, we use a change of variables.

Problem 2:

$$AP + BZ + PA^T + Z^T B^T < 0$$

where $P > 0$ and we find Z

$$K = ZP^{-1}$$

The Problem 1 is equivalent to Problem 2

100.5 Conclusion

If the (A, B) are controllable, We can obtain a controller matrix that stabilizes the system.

100.6 Implementation

A link to the Matlab code for a simple implementation of this problem in the Github repository:

<https://github.com/yashgvd/ygovada>

100.7 Related LMIs

Hurwitz Stability

100.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- <https://www.mathworks.com/help/control/ref/dcgain.html> - Mathworks reference to DC Gain

¹ http://control.asu.edu/MAE598_frame.htm

² <https://arxiv.org/abs/1903.08599/>

100.9 Return to Main Page:

101 Mixed H₂ H_∞ with desired pole location control

LMI for Mixed H_2/H_∞ with desired pole location Controller

The mixed H_2/H_∞ output feedback control has been known as an example of a multi-objective optimal control problem. In this problem, the control feedback should respond properly to several specifications. In the H_2/H_∞ controller, the H_∞ channel is used to improve the robustness of the design while the H_2 channel guarantees good performance of the system and additional constraint is used to place poles at desired location.

101.1 The System

We consider the following state-space representation for a linear system:

$$\begin{aligned}\dot{x} &= Ax + B_1u + B_2w \\ z_\infty &= C_\infty + D_{\infty 1}u + D_{\infty 2}w \\ z_2 &= C_2x + D_{21}u\end{aligned}$$

where

- $x \in \mathbb{R}^n$, $z_2, z_\infty \in \mathbb{R}^m$ are the state vector and the output vectors, respectively
- $w \in \mathbb{R}^p$, $u \in \mathbb{R}^r$ are the disturbance vector and the control vector
- $A, B_1, B_2, C_\infty, C_2, D_{\infty 1}, D_{\infty 2}$, and D_{21} are the system coefficient matrices of appropriate dimensions

101.2 The Data

We assume that all the four matrices of the plant, $A, B_1, B_2, C_\infty, C_2, D_{\infty 1}, D_{\infty 2}$, and D_{21} are given.

101.3 The Optimization Problem

For the system with the following feedback law:

$$u = Kx$$

The closed loop system can be obtained as:

$$\begin{aligned}\dot{x} &= (A + B_1 K)x + B_2 w \\ z_\infty &= (C_\infty + D_{\infty 1} K)x + D_{\infty 2} w \\ z_2 &= (C_2 + D_{21} K)u\end{aligned}$$

the transfer function matrices are $G_{z_\infty w}(s)$ and $G_{z_2 w}(s)$

Thus the H_∞ performance and the H_2 performance requirements for the system are, respectively

$$\|G_{z_\infty w}(s)\|_\infty < \gamma_\infty$$

and

$$\|G_{z_2 w}(s)\|_2 < \gamma_2$$

. For the performance of the system response, we introduce the closed-loop eigenvalue location requirement. Let

$$D = \{s \mid s \in C, L + sM + sM^T < 0\},$$

It is a region on the complex plane, which can be used to restrain the closed-loop eigenvalue locations. Hence a state feedback control law is designed such that,

- The H_∞ performance and the H_2 performance are satisfied.
- The closed-loop eigenvalues are all located in D , that is,

$$\lambda(A + B_1 K) \subset D.$$

101.4 The LMI: LMI for mixed H_2/H_∞ with desired Pole locations

The optimization problem discussed above has a solution if there exist two symmetric matrices X, Z and a matrix W , satisfying

$$\min c_2 \gamma_2^2 + c_\infty \gamma_\infty$$

s.t

$$\begin{bmatrix} (AX + B_1W)^T + AX + B_1W & B_2 & (C_\infty X + D_{\infty 1}W)^T \\ B_2^T & -\gamma_\infty I & D_{\infty 2}^T \\ C_\infty X + D_{\infty 1}W & D_{\infty 2} & -\gamma_\infty I \end{bmatrix} < 0$$

$$AX + B_1W + (AX + B_1W)^T + B_2B_2^T < 0$$

$$\begin{bmatrix} -Z & C_2X + D_{21}W \\ (C_2X + D_{21}W)^T & -X \end{bmatrix} > 0$$

$$\text{trace}(Z) < \gamma_2^2$$

$$L \otimes M \otimes (AX + B_1W) + M^T \otimes (AX + B_1W)^T < 0$$

where $c_2 > 0$ and $c_\infty > 0$ are the weighting factors.

101.5 Conclusion:

The calculated scalars γ_∞ and γ_2 are the H_2 and H_∞ norms of the system, respectively. The controller is extracted as $K = WX^{-1}$

101.6 Implementation

A link to Matlab codes for this problem in the Github repository:

101.7 Related LMIs

Mixed H2 Hinf with desired pole location for perturbed system¹

101.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

101.9 Return to Main Page

LMIs in Control/Tools⁵

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/mixh2hinfdesiredpole4perturbed

² http://control.asu.edu/MAE598_frame.htm

³ <https://arxiv.org/abs/1903.08599/>

⁴ <https://web.stanford.edu/~boyd/lmibook/>

⁵ https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

102 Mixed H₂ H_∞ with desired pole location control for perturbed systems

LMI for Mixed H_2/H_∞ with desired pole location Controller for perturbed system case

The mixed H_2/H_∞ output feedback control has been known as an example of a multi-objective optimal control problem. In this problem, the control feedback should respond properly to several specifications. In the H_2/H_∞ controller, the H_∞ channel is used to improve the robustness of the design while the H_2 channel guarantees good performance of the system and additional constraint is used to place poles at desired location.

102.1 The System

We consider the following state-space representation for a linear system:

$$\begin{aligned}\dot{x} &= (A + \Delta A)x + (B_1 + \Delta B_1)u + B_2w \\ z_\infty &= C_\infty + D_{\infty 1}u + D_{\infty 2}w \\ z_2 &= C_2x + D_{21}u\end{aligned}$$

where

- $x \in \mathbb{R}^n$, $z_2, z_\infty \in \mathbb{R}^m$ are the state vector and the output vectors, respectively
- $w \in \mathbb{R}^p$, $u \in \mathbb{R}^r$ are the disturbance vector and the control vector
- $A, B_1, B_2, C_\infty, C_2, D_{\infty 1}, D_{\infty 2}$, and D_{21} are the system coefficient matrices of appropriate dimensions.
- ΔA and ΔB_1 are real valued matrix functions which represent the time varying parameters uncertainties.

Furthermore, the parameter uncertainties ΔA and ΔB_1 are in the form of $[\Delta A \ \Delta B_1] = HF[E_1 \ E_2]$ where

- H, E_1 and E_2 are known matrices of appropriate dimensions.
- F is a matrix containing the uncertainty, which satisfies

$$F^T F < I$$

102.2 The Data

We assume that all the four matrices of the plant, A , ΔA , B_1 , ΔB_1 , B_2 , C_∞ , C_2 , $D_{\infty 1}$, $D_{\infty 2}$, and D_{21} are given.

102.3 The Optimization Problem

For the system with the following feedback law:

$$u = Kx$$

The closed loop system can be obtained as:

$$\begin{aligned}\dot{x} &= ((A + \Delta A) + (B_1 + \Delta B_1)K)x + B_2w \\ z_\infty &= (C_\infty + D_{\infty 1}K)x + D_{\infty 2}w \\ z_2 &= (C_2 + D_{21}K)u\end{aligned}$$

the transfer function matrices are $G_{z_\infty w}(s)$ and $G_{z_2 w}(s)$

Thus the H_∞ performance and the H_2 performance requirements for the system are, respectively

$$\|G_{z_\infty w}(s)\|_\infty < \gamma_\infty$$

and

$$\|G_{z_2 w}(s)\|_2 < \gamma_2$$

. For the performance of the system response, we introduce the closed-loop eigenvalue location requirement. Let

$$D = \{s | s \in C, L + sM + sM^T < 0\},$$

It is a region on the complex plane, which can be used to restrain the closed-loop eigenvalue locations. Hence a state feedback control law is designed such that,

- The H_∞ performance and the H_2 performance are satisfied.
- The closed-loop eigenvalues are all located in D , that is,

$$\lambda(A + B_1K) \subset D.$$

102.4 The LMI: LMI for mixed H_2/H_∞ with desired Pole locations

The optimization problem discussed above has a solution if there exist scalars α, β two symmetric matrices X, Z and a matrix W , satisfying

$$\min c_2\gamma_2^2 + c_\infty\gamma_\infty$$

s.t

$$\begin{aligned} & \begin{bmatrix} \Psi(X, W) & B_2 & (C_\infty X + D_{\infty 1} W)^T & (E_1 X + E_2 W)^T \\ B_2^T & -\gamma_\infty I & D_{\infty 2}^T & 0 \\ C_\infty X + D_{\infty 1} W & D_{\infty 2} & -\gamma_\infty I & 0 \\ (E_1 X + E_2 W) & 0 & 0 & -\alpha I \end{bmatrix} < 0 \\ & \begin{bmatrix} \langle AX + B_1 W \rangle + B_2 B_2^T + \beta H H^T & (E_1 X + E_2 W)^T \\ E_1 X + E_2 W & -\beta I \end{bmatrix} < 0 \\ & \begin{bmatrix} -Z & C_2 X + D_{21} W \\ (C_2 X + D_{21} W)^T & -X \end{bmatrix} > 0 \\ & \text{trace}(Z) < \gamma_2^2 \\ & L \otimes M \otimes (AX + B_1 W) + M^T \otimes (AX + B_1 W)^T < 0 \end{aligned}$$

where $\Psi(X, W) = \langle AX + B_1 W \rangle + \alpha H H^T$

$c_2 > 0$ and $c_\infty > 0$ are the weighting factors.

102.5 Conclusion:

The calculated scalars γ_∞ and γ_2 are the H_2 and H_∞ norms of the system, respectively. The controller is extracted as $K = W X^{-1}$

102.6 Implementation

A link to Matlab codes for this problem in the Github repository:

102.7 Related LMIs

Mixed H2 Hinf with desired poles controller¹

¹ https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/mixedhinfh2desiredpole/&stable=0

102.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

102.9 Return to Main Page

LMIs in Control/Tools⁵

2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599/>
4 <https://web.stanford.edu/~boyd/lmibook/>
5 https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

103 Robust H₂ State Feedback Control

103.1 Robust H₂ State Feedback Control

For the uncertain linear system given below, and a scalar $\gamma > 0$. The goal is to design a state feedback control $u(t)$ in the form of $u(t) = Kx(t)$ such that the closed-loop system is asymptotically stable and satisfies.

$$\|G_{zw}(s)\|_2 < \gamma$$

103.2 The System

Consider System with following state-space representation.

$$\begin{aligned}\dot{x}(t) &= (A + \Delta A)x(t) + (B_1 + \Delta B_1)u(t) + B_2w(t) \\ z(t) &= Cx(t) + D_1u(t) + D_2w(t)\end{aligned}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^r$, $w \in \mathbb{R}^p$, $z \in \mathbb{R}^m$. For H_2 state feedback control $D_2 = 0$

ΔA and ΔB_1 are real valued matrix functions that represent the time varying parameter uncertainties and of the form

$$[\Delta A \quad \Delta B_1] = HF [E_1 \quad E_2]$$

where matrices E_1, E_2 and H are some known matrices of appropriate dimensions, while F is a matrix which contains the uncertain parameters and satisfies.

$$F^T F \leq I$$

For the perturbation, we obviously have

$$[\Delta A \quad \Delta B_1] = [0 \quad 0], \text{ for } F = 0$$

$$[\Delta A \quad \Delta B_1] = H [E_1 \quad E_2], \text{ for } F = 0$$

103.3 The Problem Formulation:

The H_2 state feedback control problem has a solution if and only if there exist a scalar β , a matrix W , two symmetric matrices Z and X satisfying the following LMI's problem.

103.4 The LMI:

$$\min \gamma^2 ::$$

$$\begin{bmatrix} \langle AX + B_1 W \rangle_s + B_2 B_2^T + \beta H H^T & (E_1 X + E_2 W)^T \\ E_1 X + E_2 W & -\beta I \end{bmatrix} < 0$$

$$\begin{bmatrix} -Z & CX + D_1 W \\ (CX + D_1 W)^T & -X \end{bmatrix} < 0$$

$$\text{trace}(Z) < \gamma^2$$

where $\langle M \rangle_s = (M + M^T)$ is the definition that is need for the above LMI.

103.5 Conclusion:

In this case, an H_2 state feedback control law is given by $u(t) = WX^{-1}x(t)$.

103.6 External Links

- LMIs in Control Systems Analysis, Design and Applications - Duan and Yu
- A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory - A downloadable book on LMIs by Stephen Boyd.

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104 LQ Regulation via H₂ control

104.1 LQ Regulation via H_2 Control

The LQR design problem is to build an optimal state feedback controller $u = Kx$ for the system $\dot{x} = Ax + Bu, x(0) = x_0$ such that the following quadratic performance index.

$$J(x, u) = \int_0^\infty (x^T Q x + u^T R u) dt$$

is minimized, where

$$Q = Q^T \geq 0, R = R^T > 0$$

The following assumptions should hold for a traditional solution.

A1. (A, B) is stabilizable.

A2. (A, L) is observable, with $L = Q^{1/2}$.

104.2 Relation to H_2 performance

For the system given above an auxiliary system is constructed

$$\dot{x} = Ax + Bu + x_0\omega, y = Cx + Du$$

where

$$C = \begin{bmatrix} Q^{1/2} \\ 0 \end{bmatrix}, D = \begin{bmatrix} 0 \\ R^{1/2} \end{bmatrix}$$

Where ω represents an impulse disturbance. Then with state feedback controller $u = Kx$ the closed loop transfer function from disturbance ω to output y is

$$G_{y\omega}(s) = (C + DK)[sI - (A + BK)]^{-1}x_0$$

Then the LQ problem and the H_2 norm of $G_{y\omega}$ are related as

$$J(x, u) = \|G_{y\omega}(s)\|_2^2$$

Then H_2 norm minimization leads minimization of J .

104.3 Data

The state-representation of the system is given and matrices Q, R are chosen for the optimal LQ problem.

104.4 The Problem Formulation:

Let assumptions $A1$ and $A2$ hold, then the state feedback control of the form $u = Kx$ exists such that $J(x, u) < \gamma$ if and only if there exist $X \in \mathbb{S}^n$, $Y \in \mathbb{S}^r$ and $W \in \mathbb{R}^{rxn}$. Then K can be obtained by the following LMI.

104.5 The LMI:

$$\min \gamma ::$$

$$(AX + BW) + (AX + BW)^T + x_0 x_0^T < 0$$

$$\text{trace}(Q^{1/2}X(Q^{1/2})) + \text{trace}(Y) < \gamma$$

$$\begin{bmatrix} -Y & R^{1/2}W \\ (R^{1/2}W)^T & -X \end{bmatrix} < 0$$

104.6 Conclusion:

In this case, a feedback control law is given as $K = WX^{-1}$.

104.7 External Links

- LMIs in Control Systems Analysis, Design and Applications - Duan and Yu
- A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory - A downloadable book on LMIs by Stephen Boyd.

UNKNOWN TEMPLATE bookcat

105 State Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³
4. /Closed-Loop Robust Stability and Controller synthesis of Discrete-Time System with Polytopic Uncertainty/⁴

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>

2 <https://en.wikibooks.org/wiki/%2FH-2%2F>

3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

4 <https://en.wikibooks.org/wiki/%2FClosed-Loop%20Robust%20Stability%20and%20Controller%20synthesis%20of%20Discrete-Time%20System%20with%20Polytopic%20Uncertainty%2F>

106 Optimal State Feedback

1. Discrete Time Hinf Optimal Full State Feedback Control¹
2. Discrete Time H2 Optimal Full State Feedback Control²
3. Discrete Time Mixed H2-Hinf Optimal Full State Feedback Control³

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_H%E2%88%9E_Optimal_Full_State_Feedback_Control
2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_H2_Optimal_Full_State_Feedback_Control
3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_Mixed_H2-H%E2%88%9E_Optimal_Full_State_Feedback_Control

107 Output Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>
2 <https://en.wikibooks.org/wiki/%2FH-2%2F>
3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

108 Static Output Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³
4. Discrete-Time Static Output Feedback Stabilizability⁴

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>

2 <https://en.wikibooks.org/wiki/%2FH-2%2F>

3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/DT-SOFS

109 Optimal Output Feedback

1. /H-infinity/¹
2. /H-2/²
3. /Mixed/³

1 <https://en.wikibooks.org/wiki/%2FH-infinity%2F>
2 <https://en.wikibooks.org/wiki/%2FH-2%2F>
3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

110 Optimal Dynamic Output Feedback

1. Discrete Time Hinf Optimal Dynamic Output Feedback Control¹
2. Discrete Time H2 Optimal Dynamic Output Feedback Control²
3. /Mixed/³

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_H%E2%88%9E_Optimal_Dynamic_Output_Feedback_Control
2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_H2_Optimal_Dynamic_Output_Feedback_Control
3 <https://en.wikibooks.org/wiki/%2FMixed%2F>

111 Discrete Time Stabilizability

Discrete-Time Stabilizability

A discrete time system operates on a discrete time signal input and produces a discrete time signal output. They are used in digital signal processing, such as digital filters for images or sound. The class of discrete time systems that are both linear and time invariant, known as discrete time LTI systems.

Discrete-Time LTI systems can be made stable using controller gain K , which can be found using LMI optimization, such that the close loop system is stable.

111.1 The System

Discrete-Time LTI System with state space realization (A_d, B_d, C_d, D_d)

$$A_d \in \mathbf{R}^{n*n}, \quad B_d \in \mathbf{R}^{n*m}, \quad C_d \in \mathbf{R}^{p*n}, \quad D_d \in \mathbf{R}^{p*m}$$

111.2 The Data

The matrices: System $(A_d, B_d, C_d, D_d), P, W$.

111.3 The Optimization Problem

The following feasibility problem should be optimized:

Maximize P while obeying the LMI constraints.

Then K is found.

111.4 The LMI:

Discrete-Time Stabilizability

The LMI formulation

$$P \in S^n; W \in R^{m*n}$$

$$P > 0$$

$$\begin{bmatrix} P & A_d P + B_d W \\ * & P \end{bmatrix} > 0,$$
$$K_d = W P^{-1}$$

111.5 Conclusion:

The system is stabilizable iff there exists a P , such that $P > 0$. The matrix $A_d + B_d K_d$ is Schur with $K_d = W P^{-1}$

111.6 Implementation

A link to CodeOcean or other online implementation of the LMI
MATLAB Code¹

111.7 Related LMIs

https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Stabilizability_LMI - Continuous Time Stabilizability

111.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

111.9 Return to Main Page:

1 <https://github.com/Harishankar-Prabhakaran/LMIs/blob/master/A4.m>

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599/>

4 <https://web.stanford.edu/~boyd/lmibook/>

112 Quadratic Schur Stabilization

LMI for Quadratic Schur Stabilization A discrete-time system is said to be stable if all roots of its characteristic equation lie in the open unit disk. This provides a condition for the stability of discrete-time linear systems with polytopic uncertainties and a linear time-invariant system with this property is called a Schur stable system.

112.1 The System

Consider discrete time system

$$x_{k+1} = Ax_k + Bu_k,$$

where $x_k \in \mathbb{R}^n$, $u_k \in \mathbb{R}^m$, at any $t \in \mathbb{R}$.

The system consist of uncertainties of the following form

$$\begin{aligned}\Delta_{A(t)} &= A_1\delta_1(t) + \dots + A_k\delta_k(t) \\ \Delta_{B(t)} &= B_1\delta_1(t) + \dots + B_k\delta_k(t)\end{aligned}$$

where $x \in \mathbb{R}^m$, $u \in \mathbb{R}^n$, $A \in \mathbb{R}^{mxm}$ and $B \in \mathbb{R}^{mxn}$

112.2 The Data

The matrices necessary for this LMI are $A, \Delta_{A(t)}$ ie A_i , B and $\Delta_{B(t)}$ ie B_i

112.3 The LMI:

There exists some $X > 0$ and Z such that

$$\begin{bmatrix} X & AX+BZ \\ (AX+BZ)^T & X \end{bmatrix} + \begin{bmatrix} 0 & A_iX+B_iZ \\ (A_iX+B_iZ)^T & 0 \end{bmatrix} > 0 \quad i = 1, \dots, k$$

112.4 The Optimization Problem

The optimization problem is to find a matrix $K \in \mathbb{R}^{r \times n}$ such that:

$$\|A + BK\|_2 < \gamma$$

According to the definition of the spectral norms of matrices, this condition becomes equivalent to:

$$(A + BK)^T (A + BK) < \gamma^2 I$$

Using the Lemma 1.2 in LMI in Control Systems Analysis, Design and Applications¹ (page 14), the aforementioned inequality can be converted into:

$$\begin{bmatrix} -\gamma I & (A + BK) \\ (A + BK)^T & -\gamma I \end{bmatrix} < 0$$

112.5 Conclusion:

The Controller gain matrix is extracted as $F = ZX^{-1}$

$$u_k = Fx_k$$

$$\begin{aligned} x_{k+1} &= Ax_k + Bu_k, \\ &= Ax_k + BFx_k \\ &= (A + BF)x_k \end{aligned}$$

It follows that the trajectories of the closed-loop system $(A + BK)$ are stable for any $\Delta \in C_0(\Delta_1, \dots, \Delta_k)$

112.6 Implementation

https://github.com/JalpeshBhadra/LMI/blob/master/quadratic_schur_stabilization.m

112.7 Related LMIs

Schur Complement²

Schur Stabilization³

1 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Schur_Complement

3 https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/SchurStabilization&stable=0#External_Links

112.8 External Links

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁵ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁶ - A downloadable book on LMIs by Stephen Boyd.
- LMI in Control Systems Analysis, Design and Applications⁷

UNKNOWN TEMPLATE bookcat

4 http://control.asu.edu/MAE598_frame.htm

5 <https://arxiv.org/abs/1903.08599/>

6 <https://web.stanford.edu/~boyd/lmibook/>

7 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/>

Duan-Yu/p/book/9781466582996

113 Generic Insensitive Strip Region Design

Insensitive Strip Region Design

Suppose if one were interested in robust stabilization where closed-loop eigenvalues are placed in particular regions of the complex plane where the said regions has an inner boundary that is insensitive to perturbations of the system parameter matrices. This would be accomplished with the help of 2 design problems: the **insensitive strip region design** and **insensitive disk region design** (see link below for the latter).

113.1 The System

Suppose we consider the following continuous-time linear system that needs to be controlled:

$$\begin{cases} \dot{x} = Ax + Bu, \\ y = Cx \end{cases}$$

where $x \in \mathbb{R}^n$, $y \in \mathbb{R}^m$, and $u \in \mathbb{R}^r$ are the state, output and input vectors respectively. Then the steps to obtain the LMI for insensitive strip region design would be obtained as follows.

113.2 The Data

Prior to obtaining the LMI, we need the following matrices: A , B , and C .

113.3 The Optimization Problem

Consider the above linear system as well as 2 scalars γ_1 and γ_2 . Then the output feedback control law $u = Ky$ would be such that $\gamma_1 < \lambda_i(A_c^s) < \gamma_2$, where:

$$A_c^s \triangleq \frac{1}{2} \langle A_c \rangle_s = \frac{(A + BKC)^T + (A + BKC)}{2}$$

Letting K being the solution to the above problem, then

$$\gamma_1 < \alpha_1 \leq \operatorname{Re}(\lambda_i(A + BKC)) \leq \alpha_2 < \gamma_2, \quad i = 1, 2, \dots, n$$

where

$$\begin{cases} \alpha_1 &= \lambda_{\min}(A_c^s) \\ \alpha_2 &= \lambda_{\max}(A_c^s) \end{cases}$$

113.4 The LMI: Insensitive Strip Region Design

Using the above info, we can simplify the problem by setting γ_1 to $-\infty$ for all practical applications. This then simplifies our problem and results in the following LMI:

$$\begin{cases} \min \gamma \\ \text{s.t. } (A + BKC)^T + (A + BKC) < \gamma I \end{cases}$$

113.5 Conclusion:

If the resulting solution from the LMI above produces a negative γ , then the output feedback controller K is Hurwitz-stable. However, if γ is a really small positive number, then $\alpha_2 = \lambda_{\max}(A_c^s)$ must be negative for the controller to be Hurwitz-stable.

113.6 Implementation

- Example Code¹ - A GitHub link that contains code (titled "InsensitiveStripRegion.m") that demonstrates how this LMI can be implemented using MATLAB-YALMIP.

113.7 Related LMIs

- Insensitive Disk Region Design² - Equivalent LMI for Insensitive Disk Region Design
- H_2 Strip Region Design³ - LMI for H_2 Strip Region Design

1 <https://github.com/aramani3/MAE-598-LMI-Codes>

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Disk_Region_Design

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H2_StripRegion

113.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁵ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁶ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications - A book co-authored by Guang-Ren Duan and Hai-Hua Yu.

113.9 Return to Main Page:

4 http://control.asu.edu/MAE598_frame.htm
5 <https://arxiv.org/abs/1903.08599>
6 <https://web.stanford.edu/~boyd/lmibook/>

114 Generic Insensitive Disk Region Design

Insensitive Disk Region Design

Similar to the **insensitive strip region design** problem, **insensitive disk region design** is another way with which robust stabilization can be achieved where closed-loop eigenvalues are placed in particular regions of the complex plane where the said regions has an inner boundary that is insensitive to perturbations of the system parameter matrices.

114.1 The System

Suppose we consider the following linear system that needs to be controlled:

$$\begin{cases} \rho x &= Ax + Bu, \\ y &= Cx \end{cases}$$

where $x \in \mathbb{R}^n$, $y \in \mathbb{R}^m$, and $u \in \mathbb{R}^r$ are the state, output and input vectors respectively, and ρ represents the differential operator (in the continuous-time case) or one-step shift forward operator (i.e., $\rho x(k) = x(k+1)$) (in the discrete-time case). Then the steps to obtain the LMI for insensitive strip region design would be obtained as follows.

114.2 The Data

Prior to obtaining the LMI, we need the following matrices: A , B , and C .

114.3 The Optimization Problem

Consider the above linear system as well as 2 positive scalars γ and q . Then the output feedback control law $u = Ky$ would be designed such that:

$$\eta = \|A + BKC + qI\| < \gamma$$

Recalling the definition, we have:

$$\begin{aligned}\mathbb{D}_{q,\eta} &= \{s | s \in \mathbb{C}, |s+q| < \eta\} \\ &= \{x+jy | x, y \in \mathbb{R}, (x+q)^2 + y^2 < \eta^2\}\end{aligned}$$

and

$$\begin{aligned}\mathbb{D}_{q,r} &= \{s | s \in \mathbb{C}, |s+q| < r\} \\ &= \{x+jy | x, y \in \mathbb{R}, (x+q)^2 + y^2 < r^2\}\end{aligned}$$

Letting K being the solution to the above problem, then

$$\lambda_i(A+BKC) \in \mathbb{D}_{q,\eta} \subset \mathbb{D}_{q,r}, \quad i = 1, 2, \dots, n$$

114.4 The LMI: Insensitive Strip Region Design

Using the above info, we can convert the given problem into an LMI, which - after using Schur compliment Lemma - results in the following:

$$\begin{cases} \min \gamma \\ \text{s.t. } \begin{bmatrix} -\gamma I & (A+BKC+qI) \\ (A+BKC+qI)^T & -\gamma I \end{bmatrix} < 0 \end{cases}$$

114.5 Conclusion:

For Schur stabilization, we can choose to solve the problem with $q = 0$. Schur stability is achieved when $\gamma \leq 1$. Alternately, if γ is greater than (but very close to) 1, then Schur stability is also achieved when $\eta = \|A+BKC+qI\|_2 \leq 1$.

114.6 Implementation

- Example Code¹ - A GitHub link that contains code (titled "InsensitiveDiskRegion.m") that demonstrates how this LMI can be implemented using MATLAB-YALMIP.

114.7 Related LMIs

- H_2 Disk Region Design² - LMI for Disk Region Design with minimal H_2 gain

¹ <https://github.com/aramani3/MAE-598-LMI-Codes>

² https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H2_DiskRegion

- Insensitive Strip Region Design³ - Equivalent LMI for Insensitive Strip Region Design

114.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁵ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁶ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications - A book co-authored by Guang-Ren Duan and Hai-Hua Yu.

114.9 Return to Main Page:

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Strip_Region_Design

4 http://control.asu.edu/MAE598_frame.htm

5 <https://arxiv.org/abs/1903.08599/>

6 <https://web.stanford.edu/~boyd/lmibook/>

115 Design for Insensitive Strip Region

Insensitive Strip Region Design with Minimum H_2 Gain

When designing controllers with insensitive region conditions, the aim is to place the closed-loop poles of the system in a particular region defined by its inner boundary. These regions are specified based on their insensitivity to perturbations to the system parameter matrices.

One type of such design is the Insensitive Strip Region Design¹. In this section, building upon that, optimization problems will be provided that ensure that the conditions for insensitive strip region design are satisfied with some bounds on the H_2 gain of the closed-loop system.

115.1 The System

A state-space representation of a linear system as given below:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$ and $u(t) \in \mathbb{R}^r$ are the system state, output, and the input vector respectively.

115.2 The Data

To solve the design optimization problem, the linear system matrices A, B, C are required. Furthermore, to define the strip region on the eigenvalue-space, two parameters γ_1 and γ_2 are required.

115.3 The Optimization Problem

The problem of designing an H_2 optimal controller that results in the closed loop system insensitive to a certain strip region involves two sub-problems:

- Finding a control gain K such that: $\|K\|_2 < \gamma$.

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Strip_Region_Design

- The conditions for insensitive strip region design for the closed-loop system, as provided in the section *Insensitive Strip Region Design*² are fulfilled.
- The optimization goal is to minimize γ such that above two hold.

115.4 The LMI: H_2 Optimal Control Design for Insensitive Strip Region

The problem above has a solution if and only if the following optimization problem has a solution (K, γ) :

$$\begin{aligned} & \min \gamma \\ \text{s.t. } & \begin{bmatrix} -\gamma I & K \\ K^\top & -\gamma I \end{bmatrix} < 0 \\ & 2\gamma_1 I < (A + BKC)^\top + (A + BKC) < 2\gamma_2 I \end{aligned}$$

115.5 Conclusion:

By using the design problem provided here, an optimal H_2 controller is designed to make the closed-loop system robust to perturbations in the system matrices.

115.6 Implementation

To solve the optimization problem with LMI presented here, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/H2_Strip_example.m

115.7 Related LMIs

Insensitive Strip Region Design³

115.8 External Links

A list of references documenting and validating the LMI.

² https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Strip_Region_Design
³ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Strip_Region_Design

- LMIs in Control Systems: Analysis, Design and Applications⁴ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 10.1.1 pp. 322–323.

115.9 Return to Main Page:

⁴ [https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/
Duan-Yu/p/book/9781466582996](https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996)

116 Design for Insensitive Disk Region

Insensitive Disk Region Design with Minimum H_2 Gain

Apart from the design for the insensitive strip region with minimum H_2 gain¹, another type of such design is the Insensitive Disk Region Design². In this section, optimization problems will be provided that ensure that the conditions for insensitive disk region design are satisfied with some bounds on the H_2 gain of the closed-loop system.

116.1 The System

A state-space representation of a linear system as given below:

$$\begin{aligned}\rho x &= Ax + Bu \\ y &= Cx\end{aligned}$$

where $x \in \mathbb{R}^n$, $y \in \mathbb{R}^m$ and $u \in \mathbb{R}^r$ are the system state, output, and the input vector respectively. ρ represents the differential operation for continuous time systems, or the one-step shift forward operator for discrete time case.

116.2 The Data

To solve the design optimization problem, the linear system matrices A, B, C are required. Furthermore, to define the disk region on the eigenvalue-space, its radius γ_0 is required.

116.3 The Optimization Problem

The problem of designing an H_2 optimal controller that results in the closed loop system insensitive to a certain disk region involves two sub-problems:

- Finding a control gain K such that: $\|K\|_2 < \gamma$.
- The conditions for insensitive disk region design for the closed-loop system, as provided in the section *Insensitive Disk Region Design*³ are fulfilled.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H2_StripRegion

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Disk_Region_Design

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Disk_Region_Design

- The optimization goal is to minimize γ such that above two hold.

116.4 The LMI: H_2 Optimal Control Design for Insensitive Disk Region

The problem above has a solution if and only if the following optimization problem has a solution (K, γ) :

$$\begin{aligned} & \min \gamma \\ \text{s.t. } & \begin{bmatrix} -\gamma I & K \\ K^\top & -\gamma I \end{bmatrix} < 0 \\ & \begin{bmatrix} -\gamma_0 I & A + BKC + qI \\ (A + BKC + qI)^\top & -\gamma_0 I \end{bmatrix} < 0 \end{aligned}$$

116.5 Conclusion:

By using the design problem provided here, an optimal H_2 controller is designed to make the closed-loop system robust to perturbations in the system matrices.

116.6 Implementation

To solve the optimization problem with LMI presented here, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/H2_Disk_example.m

116.7 Related LMIs

Insensitive Disk Region Design⁴

116.8 External Links

A list of references documenting and validating the LMI.

- LMIs in Control Systems: Analysis, Design and Applications⁵ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 10.1.2 pp. 323–325.

⁴ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Insensitive_Disk_Region_Design

⁵ <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

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117 Quadratic Stability

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117.1 The System:

A TS fuzzy model allows the representation of a non-linear model as a set of local LTI (Linear Time Invariant) models [1,p.10], each one called subsystem. A subsystem is the local representation of the system in the space of premise variables $z(t) = [z_1(t) z_2(t) \dots z_p(t)]$ which are known and could depend on the state variables and input variables.

117.2 The Optimization Problem:

Let consider an autonomous system $\dot{x} = Ax$ with A being a constant matrix. If we define the Lyapunov function $V(x) = x^T Px$, then the system is stable if there exist $P > 0$ such that condition is satisfied.

$$A^T P + PA < 0$$

If we have a family of matrices $A(\delta(t))$ (where $\delta(t)$ is a parameter that is bounded by a polytope Δ) instead of a single matrix A , then the system equation becomes $\dot{x} = A(\delta(t))x$ and condition should be satisfied for all possible values of $\delta(t)$. If exists $P > 0$ such that following condition is satisfied then the system is quadratically stable.

$$A(\delta(t))^T P + PA(\delta(t)) < 0 \quad \forall \delta(t) \in \Delta.$$

Since there are an infinite number of matrices $A(\delta(t))$ there is also an infinite number of constraints like that for quadratic stability mentioned previously that should be fulfilled. From a practical point of view this makes the problem impossible to be solved. Let consider now that the system $\dot{x} = A(\delta(t))x$ can be written in a polytopic form as a Takagi-Sugeno (TS) polytopic system with premise variables $z(t)$ and a set of r subsystems A_i for $i = 1, \dots, r$.

$$x(t) = \sum_{i=1}^r (h_i(z(t)) A_i x(t)).$$

It can be proven that a polytopic autonomous system is quadratically stable if previous condition is satisfied in the vertices (subsystems) of the polytope. Therefore there is no need to check stability in an infinite number of matrices, but only in subsystems matrices A_i .

$$A_i^T P + PA_i < 0 \quad \forall i = 1, \dots, r.$$

Stability conditions can be applied to the closed-loop system and the following set of conditions are obtained.

$$G_{ii}^T P + PG_{ii} < 0 \quad \forall i = 1, \dots, r.$$

$$((G_{ij} + G_{ji})/2)^T P + P((G_{ij} + G_{ji})/2) <= 0 \quad \forall i, j \in \{1, \dots, r\}, i < j.$$

where $G_{ij} = A_i + B_i K_j$ and $h_i(z(t))h_j(z(t)) \neq 0$.

In the special case where matrices B_i are constant (i.e. $B_i = B$), the first set of inequalities are enough to prove stability. Therefore, assuming constant B for all the subsystems, if there exist $P > 0$ such that conditions are fulfilled, then the polytopic TS model (2.2) with state feedback control is quadratically stable inside the polytope.

$$(A_i + BK_i)^T P + P(A_i + BK_i) < 0 \quad \forall i = 1, \dots, r.$$

The assumption of constant B can be achieved using a prefiltering of the input. This change is not restrictive and the main consequence is the addition of some new state variables (the ones from the filter) to the TS model.

117.3 The LMI:

The design of the controller that stabilizes the closed-loop system boils down to solve the Linear Matrix Inequality (LMI) problem of finding a positive definite matrix P and a set of matrices K_i such that conditions are fulfilled. However, since the constraints should be linear combinations of the unknown variable, the following change of variables is applied: $W_i = K_i Q$ where $Q = P^{-1}$. The solution of the LMI problem is the set of matrices W_i such that conditions are fulfilled.

$$Q > 0$$

$$A_i Q + Q A_i^T + B W_i + W_i^T B^T < 0. \quad \forall i = 1, \dots, r.$$

The i -th controller is computed from the solution as $K_i = W_i Q^{-1}$

117.4 Conclusion:

The LMI is feasible.

117.5 Implementation

117.6 References

- Control, A. (2016). Gain-scheduling Control of a Quadrotor Using the Takagi-Sugeno Approach.

118 Apkarian Filter and State Feedback

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118.1 The System:

The number of LMI constraints needed to check quadratic stability is reduced if all the subsystems in the polytopic model has the same matrix B . This can be achieved by adding an Apkarian filter in the input of the system.

118.2 The Optimization Problem:

Apkarian Filter

Let consider our TS-LIA model. This can be re written in linear form as:

$$\dot{x} = A(z(t))x + B(z(t))u$$

The filter should be such that the equilibrium of the states are the input values and the dynamics should be fast, so we could assume the dynamics of the filter negligible (i.e. the input of the filter is equivalent to the input of the quadrotor). One possible filter is shown , where $A_F = -100I_4$, $B_F = 100I_4$ and $I_4 \in R^{4 \times 4}$ is the identity matrix.

$$\dot{x}_F = A_F x_F + B_F u_F; y_f = x_f$$

When applying the filter, we are imposing that the output of the filter is the new input of the TS-LIA model (i.e. $u = y_F$). Then, the extended model is:

$$\dot{x}_c = \begin{bmatrix} A(z(t)) & B(z(t)) \\ 0 & A_F \end{bmatrix} x_c + \begin{bmatrix} 0 \\ B_F \end{bmatrix} u_F = A_e(z(t))x_e + B_e u_f; x_e = \begin{bmatrix} x \\ x_F \end{bmatrix}$$

This prefiltering does not affect the procedure followed to obtain the TS-LIA model, so the premise variables, membership functions and activations functions remains the same.

State Feedback Controller Design

Let consider the state feedback control law for the extended TS-LIA model: $\dot{x}_e = \sum_{i=1}^{32} h_i(z(t)) [A_{ei}x_e + B_{ei}u_F]$, where the state feedback control laws are : $u_F = \sum_{i=1}^{32} h_i(z(t)) K_i x(t)$, we get the closed loop system : $\dot{x}_e = \sum_{i=1}^{32} \sum_{j=1}^{32} h_j(z(t)) [A_{ei}x_e + B_{ei}K_j] x_e$

118.3 The LMI:

The design of the controller is done by solving an LMI problem involving the quadratic stability constraints. In case we want D- stabilization, the following set of LMI constraints are needed:

$$L \otimes P + M \otimes P(A_{ei} + B_{ei}K_i) + MT \otimes (A_{ei} + B_{ei}K_i)^T P < 0 \quad \forall i = 1, \dots, 32.$$

118.4 Conclusion:

The LMI is feasible.

118.5 Related LMIs

- LMI for Natural Frequency in State Feedback. https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Maximum_Natural_Frequency_in_State_Feedback#The_LMI%3A
- LMI for Minimum Decay Rate in State Feedback. https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Minimum_Decay_Rate_in_State_Feedback#The_LMI%3A

118.6 References

- Control, A. (2016). Gain-scheduling Control of a Quadrotor Using the Takagi-Sugeno Approach.

119 Minimum Decay Rate in State Feedback

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119.1 The System:

The number of LMI constraints needed to check quadratic stability is reduced if all the subsystems in the polytopic model has the same matrix B . This can be achieved by adding an Apkarian filter in the input of the system.

119.2 The Optimization Problem:

Apkarian Filter

Let consider our TS-LIA model. This can be re written in linear form as:

$$\dot{x} = A(z(t))x + B(z(t))u$$

The filter should be such that the equilibrium of the states are the input values and the dynamics should be fast, so we could assume the dynamics of the filter negligible (i.e. the input of the filter is equivalent to the input of the quadrotor). One possible filter is shown , where $A_F = -100I_4$, $B_F = 100I_4$ and $I_4 \in R^{4 \times 4}$ is the identity matrix.

$$\dot{x}_F = A_F x_F + B_F u_F; y_F = x_F$$

When applying the filter, we are imposing that the output of the filter is the new input of the TS-LIA model (i.e. $u = y_F$). Then, the extended model is:

$$\dot{x}_c = \begin{bmatrix} A(z(t)) & B(z(t)) \\ 0 & A_F \end{bmatrix} x_c + \begin{bmatrix} 0 \\ B_F \end{bmatrix} u_F = A_e(z(t))x_e + B_e u_f; x_e = \begin{bmatrix} x \\ x_F \end{bmatrix}$$

This prefiltering does not affect the procedure followed to obtain the TS-LIA model, so the premise variables, membership functions and activations functions remains the same.

State Feedback Controller Design

Let consider the state feedback control law for the extended TS-LIA model: $\dot{x}_e = \sum_{i=1}^{32} h_i(z(t)) [A_{ei}x_e + B_{ei}u_F]$, where the state feedback control laws are : $u_F = \sum_{i=1}^{32} h_i(z(t)) K_i x(t)$, we get the closed loop system : $\dot{x}_e = \sum_{i=1}^{32} \sum_{j=1}^{32} h_j(z(t)) [A_{ei}x_e + B_{ei}K_j] x_e$

119.3 The LMI:

The design of the controller is done by solving an LMI problem involving the quadratic stability constraints. In case we want D- stabilization, the following set of LMI constraints are needed:

$$L \otimes P + M \otimes P(A_{ei} + B_{ei}K_i) + MT \otimes (A_{ei} + B_{ei}K_i)^T P < 0 \quad \forall i = 1, \dots, 32.$$

A pair of conjugate complex poles s of the closed loop system can be written as $s = -\xi\omega_n \pm j\omega_d$ where ξ is the damping ratio, ω_n is the undamped natural frequency and ω_d is the frequency response defined as $\omega_d = \omega_n \sqrt{1 - \xi^2}$. Three different LMI regions have been considered, each one related with a performance specification regarding $\alpha = \xi\omega_n$, ω_n and ξ :

Minimum Decay Rate:

If we want to set a minimum decay rate α in the closed loop system response, the poles should be inside the LMI region defined in : $S_\alpha = \{s = x + jy \mid x < -\alpha\}$. where $\alpha > 0$. In this case $L_\alpha = 2\alpha$ and $M_\alpha = 1$.

Applying condition to the closed-loop system , the LMI condition associated to this LMI region is:

$$2\alpha P + (A_{ei} + B_{ei}K_i)^T P + P(A_{ei} + B_{ei}K_i) < 0 \quad \forall i = 1, \dots, 32.$$

119.4 Conclusion:

The LMI is feasible.

119.5 Related LMIs

- Apkarian Filter and State Feedback. https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Apkarian_Filter-and_State_Feedback
- Maximum Natural Frequency in State Feedback https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Maximum_Natural_Frequency_in_State_Feedback#The_LMI%3A

119.6 References

- Control, A. (2016). Gain-scheduling Control of a Quadrotor Using the Takagi-Sugeno Approach.

120 Maximum Natural Frequency in State Feedback

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120.1 The System:

The number of LMI constraints needed to check quadratic stability is reduced if all the subsystems in the polytopic model has the same matrix B . This can be achieved by adding an Apkarian filter in the input of the system.

120.2 The Optimization Problem:

Apkarian Filter

Let consider our TS-LIA model. This can be re written in linear form as:

$$\dot{x} = A(z(t))x + B(z(t))u$$

The filter should be such that the equilibrium of the states are the input values and the dynamics should be fast, so we could assume the dynamics of the filter negligible (i.e. the input of the filter is equivalent to the input of the quadrotor). One possible filter is shown , where $A_F = -100I_4$, $B_F = 100I_4$ and $I_4 \in R^{4 \times 4}$ is the identity matrix.

$$\dot{x}_F = A_F x_F + B_F u_F; y_F = x_F$$

When applying the filter, we are imposing that the output of the filter is the new input of the TS-LIA model (i.e. $u = y_F$). Then, the extended model is:

$$\dot{x}_c = \begin{bmatrix} A(z(t)) & B(z(t)) \\ 0 & A_F \end{bmatrix} x_c + \begin{bmatrix} 0 \\ B_F \end{bmatrix} u_F = A_e(z(t))x_e + B_e u_f; x_e = \begin{bmatrix} x \\ x_F \end{bmatrix}$$

This prefiltering does not affect the procedure followed to obtain the TS-LIA model, so the premise variables, membership functions and activations functions remains the same.

State Feedback Controller Design

Let consider the state feedback control law for the extended TS-LIA model: $\dot{x}_e = \sum_{i=1}^{32} h_i(z(t)) [A_{ei}x_e + B_{ei}u_F]$, where the state feedback control laws are $:u_F = \sum_{i=1}^{32} h_i(z(t)) K_i x(t)$, we get the closed loop system : $\dot{x}_e = \sum_{i=1}^{32} \sum_{j=1}^{32} h_j(z(t)) [A_{ei}x_e + B_{ei}K_j] x_e$

120.3 The LMI:

The design of the controller is done by solving an LMI problem involving the quadratic stability constraints. In case we want D- stabilization, the following set of LMI constraints are needed:

$$L \otimes P + M \otimes P(A_{ei} + B_{ei}K_i) + MT \otimes (A_{ei} + B_{ei}K_i)^T P < 0 \quad \forall i = 1, \dots, 32.$$

A pair of conjugate complex poles s of the closed loop system can be written as $s = -\xi\omega_n \pm j\omega_d$ where ξ is the damping ratio, ω_n is the undamped natural frequency and ω_d is the frequency response defined as $\omega_d = \omega_n \sqrt{1 - \xi^2}$. Three different LMI regions have been considered, each one related with a performance specification regarding $\alpha = \xi\omega_n, \omega_n$ and ξ :

Maximizing Natural Frequency:

Natural frequency is related with the maximum frequency response in the undamped case ($\xi = 0$). If we want to set a maximum ω_n condition, the LMI region associated is $S_\rho = [s = x + jy \mid |x + jy| < \rho]$, :: $L_\rho = \begin{bmatrix} -\rho & 0 \\ 0 & \rho \end{bmatrix}, M_\rho = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$. Resulting LMI condition is:

$$\begin{bmatrix} -\rho P & P(A_{ei} + B_{ei}K_i) \\ (A_{ei} + B_{ei}K_i)^T P & -\rho P \end{bmatrix} < 0 \quad \forall i = 1, \dots, 32.$$

120.4 Conclusion:

The LMI is feasible.

120.5 Related LMIs

- Apkarian Filter and State Feedback. https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Apkarian_Filter-and_State_Feedback
- Minimum Decay Rate in State Feedback. https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Minimum_Decay_Rate_in_State_Feedback#The_LMI%3A

120.6 References

- Control, A. (2016). Gain-scheduling Control of a Quadrotor Using the Takagi-Sugeno Approach.

121 Optimal Observer and State Estimation

122 LMI for the Observability Grammian

LMI for the Observability Grammian

Observability is a system property which says that the state of the system $x(T)$ can be reconstructed using the input $u(t)$ and output $y(t)$ on an interval $[0, T]$. This is necessary when knowledge of the full state is not available. If observable, estimators or observers can be created to reconstruct the full state. Observability and controllability are dual concepts. Thus in order to investigate the observability of a system we can study the controllability of the dual system. Although system observability can be determined with multiple methods, one is to compute the rank of the observability grammian.

122.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t), \\ x(0) &= x_0,\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^m$, at any $t \in \mathbb{R}$.

122.2 The Data

The matrices necessary for this LMI are A and C .

122.3 The LMI:LMI to Determine the Observability Grammian

(A, B) is observable if and only if $Y > 0$ is the unique solution to

$$AY + YA^T - C^T C < 0$$

where Y is the observability grammian.

122.4 Conclusion:

The above LMI attempts to find the observability grammian Y of the system (A, C) . If the problem is feasible and a unique Y is found, then the system is also observable. The observability grammian can also be computed as: $Y = \int_0^\infty e^{AT} s C^T C e^{As} ds$. Due to the dual nature of observability and controllability this LMI can be determined by determining the controllability of the dual nature, which results in the above LMI. The Observability and Controllability matrixies are written as \mathcal{O} and \mathcal{C} respectively. They are related as follows:

$$\mathcal{O}(C, A) = \mathcal{C}(A^T, C^T)^T$$

$$\mathcal{C}(A, B) = \mathcal{O}(B^T, A^T)^T$$

Hence (C, A) is observable if and only if (A^T, C^T) is controllable. Please refer to the section on controllability grammians.

122.5 Implementation

This implementation requires Yalmip and Sedumi.

https://github.com/eoskowro/LMI/blob/master/Observability_Gram_LMI.m

122.6 Related LMIs

Stabilizability LMI¹

Hurwitz Stability LMI²

Detectability LMI³

Controllability Grammian LMI⁴

122.7 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁵ - A course on LMIs in Control by Matthew Peet.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Stabilizability_LMI

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Hurwitz_Stability

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Detectability_LMI

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_the_Controllability_Grammian

5 http://control.asu.edu/MAE598_frame.htm

- LMI Properties and Applications in Systems, Stability, and Control Theory⁶ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁷ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications⁸ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013, Section 6.1.1 and Table 6.1 pp. 166–170, 192.
- A Course in Robust Control Theory: a Convex Approach,⁹ - by Geir E. Dullerud and Fernando G. Paganini, Springer, 2011, Section 2.2.3, pp. 71-73.

122.8 Return to Main Page:

6 <https://arxiv.org/abs/1903.08599/>

7 <https://web.stanford.edu/~boyd/lmibook/>
<https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/>

8 Duan-Yu/p/book/9781466582996

9 <https://link.springer.com/book/10.1007/978-1-4757-3290-0#toc>

123 H-infinity filtering

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For systems that have disturbances, filtering can be used to reduce the effects of these disturbances. Described on this page is a method of attaining a filter that will reduce the effects of the disturbances as completely as possible. To do this, we look to find a set of new coefficient matrices that describe the filtered system. The process to achieve such a new system is described below. The H-infinity-filter tries to minimize the maximum magnitude of error.

123.1 The System

For the application of this LMI, we will look at linear systems that can be represented in state space as

$$\begin{aligned}\dot{x} &= Ax + Bw, x(0) = x_0 \\ y &= Cx + Dw \\ z &= Lx\end{aligned}$$

where $x \in R^n$, $y \in R^l$, $z \in R^m$ represent the state vector, the measured output vector, and the output vector of interest, respectively, $w \in R^p$ is the disturbance vector, and A, B, C, D and L are the system matrices of appropriate dimension.

To further define: x is $\in R^n$ and is the state vector, A is $\in R^{n*n}$ and is the state matrix, B is $\in R^{n*r}$ and is the input matrix, w is $\in R^r$ and is the exogenous input, C is $\in R^{m*n}$ and is the output matrix, D and L are $\in R^{m*r}$ and are feedthrough matrices, and y and z are $\in R^m$ and are the output and the output of interest, respectively.

123.2 The Data

The data are w (the disturbance vector), and A, B, C, D and L (the system matrices). Furthermore, the A matrix is assumed to be stable

123.3 The Optimization Problem

We need to design a filter that will eliminate the effects of the disturbances as best we can. For this, we take a filter of the following form:

$$\begin{aligned}\dot{\sigma} &= A_f \sigma + B_f \sigma, \sigma(0) = \sigma_0 \\ \hat{z} &= C_f \sigma + D_f y,\end{aligned}$$

where $\sigma \in R^n$ is the state vector, $\hat{z} \in R^m$ is the estimation vector of z , and $A_f, B_f, C_f, \text{and } D_f$ are the coefficient matrices of appropriate dimensions.

Note that the combined complete system can be represented as

$$\begin{aligned}\dot{x}_e &= \tilde{A}x_e + \tilde{B}w, x_e(0) = x_{e0} \\ \tilde{z} &= \tilde{C}x_e + \tilde{D}w,\end{aligned}$$

where $\tilde{z} = z - \hat{z} \in R^m$ is the estimation error,

$$x_e = \begin{bmatrix} x \\ \sigma \end{bmatrix}$$

is the state vector of the system, and $\tilde{A}, \tilde{B}, \tilde{C}, \tilde{D}$ are the coefficient matrices, defined as:

$$\begin{aligned}\tilde{A} &= \begin{bmatrix} A & 0 \\ B_f C & A_f \end{bmatrix}, \tilde{B} = \begin{bmatrix} B \\ B_f D \end{bmatrix}, \\ \tilde{C} &= [L - D_f C \quad -C_f], \tilde{D} = -D_f D\end{aligned}$$

In other words, for the system defined above we need to find $A_f, B_f, C_f, \text{and } D_f$ such that

$$|G_{\tilde{z}w}(s)|_{\inf} < \gamma,$$

where γ is a positive constant, and

$$G_{\tilde{z}w}(s) = \tilde{C}(sI - \tilde{A})^{-1}\tilde{B} + \tilde{D}.$$

123.4 The LMI: H-inf Filtering

The solution can be obtained by finding matrices R, X, M, N, Z, D_f that obey the following LMIs:

$$\begin{array}{c}
 X > 0 \\
 R - X > 0 \\
 \left[\begin{array}{cccc}
 RA + A^T R + ZC + C^T Z^T & * & * & * \\
 M^T + ZC + XA & M^T + M & * & * \\
 B^T R + D^T Z^T & B^T X + D^T Z^T & -\gamma I & * \\
 L - D_f C & -N & -D_f D & -\gamma I
 \end{array} \right] < 0
 \end{array}$$

123.5 Conclusion:

To find the corresponding filter, use the optimized matrices from the solution to find:

$$A_f = X^{-1}M, B_f = X^{-1}Z, C_f = N, D_f = D_f$$

These matrices can then be used to produce $\tilde{A}, \tilde{B}, \tilde{C}, \tilde{D}$ to construct the filter described above, that will best eliminate the disturbances of the system.

123.6 Implementation

This implementation requires Yalmip and Sedumi.

https://github.com/rezajamesahmed/LMImatlabcode/blob/master/hinf_filtering.m

123.7 Related LMIs

H-2 _filtering¹

123.8 External Links

This LMI comes from

- ² - "LMIs in Control Systems: Analysis, Design and Applications" by Guang-Ren Duan and Hai-Hua Yu

Other resources:

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H-2_filtering

² <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/>

³ [Duan-Yu/p/book/9781466582996](http://control.asu.edu/MAE598_frame.htm)

- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

123.9 References

Duan, G. (2013). LMIs in control systems: analysis, design and applications. Boca Raton: CRC Press, Taylor & Francis Group.

123.10 Return to Main Page:

4 <https://arxiv.org/abs/1903.08599>
5 <https://web.stanford.edu/~boyd/lmibook/>

124 H2 filtering

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For systems that have disturbances, filtering can be used to reduce the effects of these disturbances. Described on this page is a method of attaining a filter that will reduce the effects of the disturbances as completely as possible. To do this, we look to find a set of new coefficient matrices that describe the filtered system. The process to achieve such a new system is described below. The H2-filter tries to minimize the average magnitude of error.

124.1 The System

For the application of this LMI, we will look at linear systems that can be represented in state space as

$$\begin{aligned}\dot{x} &= Ax + Bw, x(0) = x_0 \\ y &= Cx + Dw \\ z &= Lx\end{aligned}$$

where $x \in R^n$, $y \in R^l$, $z \in R^m$ represent the state vector, the measured output vector, and the output vector of interest, respectively, $w \in R^p$ is the disturbance vector, and A, B, C, D and L are the system matrices of appropriate dimension. To further define: x is $\in R^n$ and is the state vector, A is $\in R^{n*n}$ and is the state matrix, B is $\in R^{n*r}$ and is the input matrix, w is $\in R^r$ and is the exogenous input, C is $\in R^{m*n}$ and is the output matrix, D and L are $\in R^{m*r}$ and are feedthrough matrices, and y and z are $\in R^m$ and are the output and the output of interest, respectively.

124.2 The Data

The data are w (the disturbance vector), and A, B, C, D and L (the system matrices). Furthermore, the A matrix is assumed to be stable

124.3 The Optimization Problem

We need to design a filter that will eliminate the effects of the disturbances as best we can. For this, we take a filter of the following form:

$$\begin{aligned}\dot{\sigma} &= A_f \sigma + B_f y, \sigma(0) = \sigma_0 \\ \hat{z} &= C_f \sigma,\end{aligned}$$

where $\sigma \in R^n$ is the state vector, $\hat{z} \in R^m$ is the estimation vector, and A_f, B_f, C_f are the coefficient matrices of appropriate dimensions.

Note that the combined complete system can be represented as

$$\begin{aligned}\dot{x}_e &= \tilde{A}x_e + \tilde{B}w, x_e(0) = x_{e0} \\ \tilde{z} &= \tilde{C}x_e,\end{aligned}$$

where $\tilde{z} = z - \hat{z} \in R^m$ is the estimation error,

$$x_e = \begin{bmatrix} x \\ \sigma \end{bmatrix}$$

is the state vector of the system, and $\tilde{A}, \tilde{B}, \tilde{C}$ are the coefficient matrices, defined as:

$$\begin{aligned}\tilde{A} &= \begin{bmatrix} A & 0 \\ B_f C & A_f \end{bmatrix}, \tilde{B} = \begin{bmatrix} B \\ B_f D \end{bmatrix}, \\ \tilde{C} &= \begin{bmatrix} L & -C_f \end{bmatrix}\end{aligned}$$

In other words, for the system defined above we need to find A_f, B_f, C_f such that

$$\|G_{\tilde{z}w}(s)\|_2 < \gamma,$$

where γ is a positive constant, and

$$G_{\tilde{z}w}(s) = \tilde{C}(sI - \tilde{A})^{-1}\tilde{B}$$

124.4 The LMI: H-2 Filtering

For this LMI, the solution exists if one of the following sets of LMIs hold:

Matrices R, X, M, N, Z, Q exist that obey the following LMIs:

$$\begin{aligned}
 & R - X > 0, \\
 & \text{trace}(Q) < \gamma^2, \\
 & \begin{bmatrix} -Q & * & * \\ L^T & -R & * \\ -N^T & -X & -X \end{bmatrix} < 0, \\
 & \begin{bmatrix} RA + A^T R + ZC + C^T Z^T & * & * \\ M^T + ZC + XA & M^T + M & * \\ B^T R + D^T Z^T & B^T X + D^T Z^T & -I \end{bmatrix} < 0.
 \end{aligned}$$

or

Matrices $\bar{R}, \bar{X}, \bar{M}, \bar{N}, \bar{Z}, \bar{Q}$ exist that obey the following LMIs:

$$\begin{aligned}
 & \bar{R} - \bar{X} > 0, \\
 & \text{trace}(\bar{Q}) < \gamma^2, \\
 & \begin{bmatrix} -\bar{Q} & * & * \\ \bar{R}B + \bar{Z}D & -\bar{R} & * \\ \bar{X}B + \bar{Z}D & -\bar{X} & -I \end{bmatrix} < 0, \\
 & \begin{bmatrix} \bar{R}A + A^T \bar{R} + \bar{Z}C + C^T \bar{Z}^T & * & * \\ \bar{M}^T + \bar{Z}C + \bar{X}A & \bar{M}^T + \bar{M} & * \\ L & -\bar{N} & -I \end{bmatrix} < 0.
 \end{aligned}$$

124.5 Conclusion:

To find the corresponding filter, use the optimized matrices from the first solution to find:

$$A_f = X^{-1}M, B_f = X^{-1}Z, C_f = N$$

Or the second solution to find:

$$A_f = \bar{X}^{-1}\bar{M}, B_f = \bar{X}^{-1}\bar{Z}, C_f = \bar{N}$$

These matrices can then be used to produce $\tilde{A}, \tilde{B}, \tilde{C}$ to construct the final filter below, that will best eliminate the disturbances of the system.

$$\begin{aligned}
 \dot{x}_e &= \tilde{A}x_e + \tilde{B}w, x_e(0) = x_{e0} \\
 \tilde{z} &= \tilde{C}x_e,
 \end{aligned}$$

124.6 Implementation

This implementation requires Yalmip and Sedumi.

https://github.com/rezajamesahmed/LMImatlabcode/blob/master/H2_Filtering.m

124.7 Related LMIs

H-infinity filtering¹

124.8 External Links

This LMI comes from

- ² - "LMIs in Control Systems: Analysis, Design and Applications" by Guang-Ren Duan and Hai-Hua Yu

Other resources:

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

124.9 References

Duan, G. (2013). LMIs in control systems: analysis, design and applications. Boca Raton: CRC Press, Taylor & Francis Group.

124.10 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H-infinity_filtering#Conclusion%3A

2 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

3 http://control.asu.edu/MAE598_frame.htm

4 <https://arxiv.org/abs/1903.08599/>

5 <https://web.stanford.edu/~boyd/lmibook/>

125 H₂ Optimal Observer

State observer is a system that provides estimates of internal states of a given real system, from measurements of the inputs and outputs of the real system. The goal of H_2 -optimal state estimation is to design an observer that minimizes the H_2 norm of the closed-loop transfer matrix from w to z . Kalman filter is a form of Optimal Observer.

125.1 The System

Consider the continuous-time generalized plant P with state-space realization

$$\begin{aligned}\dot{x} &= Ax + B_1 w(t), \\ y &= C_2 x + D_{21} w\end{aligned}$$

125.2 The Data

The matrices needed as input are A, B, C, D .

125.3 The Optimization Problem

The task is to design an observer of the following form:

$$\begin{aligned}\dot{\hat{x}} &= A\hat{x} + L(y - \hat{y}), \\ \hat{y} &= C_2 \hat{x}\end{aligned}$$

125.4 The LMI: H_2 Optimal Observer

LMIs in the variables P, G, Z, ν are given by:

$$\begin{bmatrix} PA + A^T P - GC_2 - C_2^T G^T & PB_1 - GD_{21} \\ * & -1 \\ & trZ < \nu \end{bmatrix} < 0$$

125.5 Conclusion:

The H_2 -optimal observer gain is recovered by $L = P^{-1}G$ and the H_2 norm of $T(s)$ is $\mu = \sqrt{\nu}$

125.6 Implementation

<https://github.com/Ricky-10/coding107/blob/master/H2%20Optimal%20Observer>

126 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.
- <https://onlinelibrary.wiley.com/doi/10.1002/rnc.1310>

126.1 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

127 H_{Inf} Optimal Observer

H -Optimal observers yield robust estimates of some or all internal plant states by processing measurement data. Robust observers are increasingly demanded in industry as they may provide state and parameter estimates for monitoring and diagnosis purposes even in the presence of large disturbances such as noise etc. It is there where Kalman filters may tend to fail. State observer is a system that provides estimates of internal states of a given real system, from measurements of the inputs and outputs of the real system. The goal of H -optimal state estimation is to design an observer that minimizes the H norm of the closed-loop transfer matrix from w to z .

127.1 The System

Consider the continuous-time generalized plant P with state-space realization

$$\begin{aligned}\dot{x} &= Ax + B_1 w, \\ y &= C_2 x + D_{21} w\end{aligned}$$

127.2 The Data

The matrices needed as input are $A, B_1, B_2, C_2, D_{21}, D_{11}$.

127.3 The Optimization Problem

The observer gain L is to be designed such that the H of the transfer matrix from w to z , given by

$$T(s) = C_1(sI - (A - LC_2))^{-1}(B_1 - LD_{21}) + D_{11}$$

is minimized. The form of the observer would be:

$$\begin{aligned}\dot{\hat{x}} &= A\hat{x} + L(y - \hat{y}), \\ \hat{y} &= C_2\hat{x}\end{aligned}$$

127.4 The LMI: H Optimal Observer

The H -optimal observer gain is synthesized by solving for $P \in \mathbb{S}^{n_x}$, $G \in \mathbb{R}^{n_x \times n_y}$, and $\gamma \in \mathbb{R}_{>0}$ that minimize $\zeta(\gamma) = \gamma$ subject to $P > 0$ and

$$\begin{bmatrix} PA + A^T P - GC_2 - C_2^T G^T & PB_1 - GD_{21} & C_1^T \\ * & -\gamma I & D_{11}^T \\ * & * & -\gamma I \end{bmatrix} < 0$$

127.5 Conclusion:

The H -optimal observer gain is recovered by $L = P^{-1}G$ and the H norm of $T(s)$ is γ .

127.6 Implementation

Link to the MATLAB code designing H - Optimal Observer

[https://github.com/Ricky-10/coding107/blob/master/
HinfinityOptimalobserver](https://github.com/Ricky-10/coding107/blob/master/HinfinityOptimalobserver)

128 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.
- https://www.researchgate.net/publication/322245033_On_structural_approaches_to_H-infinity_observer_design
- https://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-241j-dynamic-systems-and-control-spring-2011/lecture-notes/MIT6_241JS11_lec25.pdf

128.1 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

129 Mixed H₂ H_{Inf} Optimal Observer

The goal of mixed $H_2 - H$ -optimal state estimation is to design an observer that minimizes the H_2 norm of the closed-loop transfer matrix from w_1 to z_1 , while ensuring that the H norm of the closed-loop transfer matrix from w_2 to z_2 is below a specified bound.

129.1 The System

Consider the continuous-time generalized plant P with state-space realization

$$\begin{aligned}\dot{x} &= Ax + B_{1,1}w_1 + B_{1,2}w_2, \\ y &= C_2x + D_{21,1}w_1 + D_{21,2}w_2\end{aligned}$$

where it is assumed that (A, C_2) is detectable.

129.2 The Data

The matrices needed as input are $A, B_1, B_2, C_2, D_{21}, D_{11}$.

129.3 The Optimization Problem

The observer gain L is to be designed to minimize the H_2 norm of the closed-loop transfer matrix $T_{11}(s)$ from the exogenous input w_1 to the performance output z_1 while ensuring the H norm of the closed-loop transfer matrix $T_{22}(s)$ from the exogenous input w_2 to the performance output z_2 is less than γ_d , where

$$\begin{aligned}T_{11}(s) &= C_{1,1}(sI - (A - LC_2))^{-1}(B_{1,1} - LD_{21,1}) \\ T_{22}(s) &= C_{1,2}(sI - (A - LC_2))^{-1}(B_{1,2} - LD_{21,2}) + D_{11,22}\end{aligned}$$

is minimized. The form of the observer would be:

$$\begin{aligned}\dot{\hat{x}} &= A\hat{x} + L(y - \hat{y}), \\ \hat{y} &= C_2\hat{x}\end{aligned}$$

is to be designed, where $L \in \mathbb{R}^{n_x \times n_y}$ is the observer gain.

129.4 The LMI: H Optimal Observer

The mixed $H_2 - H$ -optimal observer gain is synthesized by solving for $P \in \mathbb{S}^{n_x}, G \in \mathbb{R}^{n_x \times n_y}$, and $\nu \in \mathbb{R}_{>0}$ that minimize $\zeta(\nu) = \nu$ subject to $P > 0, Z > 0$,

$$\begin{aligned} & \begin{bmatrix} PA + A^T P - GC_2 - C_2^T G^T & PB_1 - GD_{21} \\ \star & -1 \end{bmatrix} < 0 \\ & \begin{bmatrix} PA + A^T P - GC_2 - C_2^T G^T & PB_1 - GD_{21} & C_1 \\ \star & -\gamma I & D_{11}^T \\ \star & \star & -\gamma I \end{bmatrix} < 0 \\ & \begin{bmatrix} P & C_{1,1}^T \\ \star & Z \end{bmatrix} > 0 \\ & trZ < \nu \end{aligned}$$

129.5 Conclusion:

The mixed $H_2 - H$ -optimal observer gain is recovered by $L = P^{-1}G$, the H_2 norm of $T_{11}(s)$ is less than $\mu = \sqrt{\nu}$ and the H norm of $T(s)$ is less than γ_d .

129.6 Implementation

Link to the MATLAB code designing $H_2 - H$ - Optimal Observer

Code $H_2 - H$ Optimal Observer¹

129.7 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

1 <https://github.com/Ricky-10/coding107/blob/master/Mixed%20H2-Hinfinity%20Optimal%20Observer>
2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599/>
4 <https://web.stanford.edu/~boyd/lmibook/>

129.8 Related LMIs

- H_2 Optimal observer⁵
- H Optimal observer⁶

129.9 Return to Main Page:

5 <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2FH2Optimalobserver>

6 https://en.wikibooks.org/wiki/LMIs_in_Control%2Fpages%2Finfinityoptimalobserver

130 H2 Optimal Filter

Optimal filtering is a means of adaptive extraction of a weak desired signal in the presence of noise and interfering signals. Optimal filters normally are free from stability problems. There are simple operational checks on an optimal filter when it is being used that indicate whether it is operating correctly. Optimal filters are probably easier to make adaptive to parameter changes than suboptimal filters. The goal of optimal filtering is to design a filter that acts on the output z of the generalized plant and optimizes the transfer matrix from w to the filtered output.

130.1 The System:

Consider the continuous-time generalized LTI plant with minimal states-space realization

$$\begin{aligned}\dot{x} &= Ax + B_1 w \\ z &= C_1 x + D_{11} w, \\ y &= C_2 x + D_{21} w,\end{aligned}$$

where it is assumed that A is Hurwitz.

130.2 The Data

The matrices needed as inputs are A, B_1, C_2, D_{21} .

130.3 The Optimization Problem:

An H_2 -optimal filter is designed to minimize the H_2 norm of $\tilde{P}(s)$ in following equation.

$$\tilde{P}(s) = \tilde{C}_1(sI - \tilde{A})^{-1}\tilde{B}_1 + \tilde{D}_{11},$$

where

$$\tilde{A} = \begin{bmatrix} A & 0 \\ B_f C_2 & A_f \end{bmatrix} < 0$$

$$\tilde{B}_1 = \begin{bmatrix} B_1 \\ B_f D_{21} \end{bmatrix} < 0$$

$$\tilde{C}_1 = [C_1 - D_f C_2 - C_f] < 0$$

$$\tilde{D}_{11} = D_{11} - D_f D_{21}$$

To ensure that $\tilde{P}(s)$ has a finite H_2 norm, it is required that $D_f = D_{11}$, which results in $\tilde{D}_{11} = D_{11} - D_f = 0$

130.4 The LMI: H_2 - Optimal filter

Solve for $A_n \in \mathbb{R}^{n_x \times n_x}, B_n \in \mathbb{R}^{n_x \times n_y}, C_f \in \mathbb{R}^{n_x \times n_x}, X, Y \in \mathbb{S}^{n_x}, Z \in \mathbb{S}^{n_z}$ and $\nu \in \mathbb{R}_{>0}$ that minimize $\zeta(\nu) = \nu$ subject to $X > 0, Y > 0, Z > 0$.

$$\begin{bmatrix} YA + A^T Y + B_n C_2 & A_n + C_2^T B_n^T + A^T X & Y B_1 + B_n D_{21} \\ * & A_n + A_n^T & X B_1 + B_n D_{21} \\ * & * & -1 \end{bmatrix} < 0$$

$$\begin{bmatrix} -Z & C_1 - D_f C_2 & -C_f \\ * & -Y & -X \\ * & * & -X \end{bmatrix} < 0$$

$$Y - X > 0$$

$$tr Z < /nu$$

130.5 Conclusion:

The filter is recovered by $A_f = X^{-1}A_n$ and $B_f = X^{-1}B_n$.

130.6 Implementation

MATLAB code of H_2 Optimal filter¹

¹ <https://github.com/Ricky-10/coding107/blob/master/H2Optimal%20Filter>

130.7 External links

- <http://home.eng.iastate.edu/~julied/classes/ee524/LectureNotes/110.pdf>- Optimal Filtering
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- [<http://users.cecs.anu.edu.au/~john/papers/BOOK/B02.PDF>]- Optimal Filtering by Brian D.O. Anderson

² <https://arxiv.org/abs/1903.08599/>

131 H_{Inf} Optimal Filter

Optimal filtering is a means of adaptive extraction of a weak desired signal in the presence of noise and interfering signals. The goal of optimal filtering is to design a filter that acts on the output z of the generalized plant and optimizes the transfer matrix from w to the filtered output.

131.1 The System:

Consider the continuous-time generalized LTI plant with minimal states-space realization

$$\begin{aligned}\dot{x} &= Ax + B_1 w \\ z &= C_1 x + D_{11} w, \\ y &= C_2 x + D_{21} w,\end{aligned}$$

where it is assumed that A is Hurwitz.

131.2 The Data

The matrices needed as inputs are $A, B_1, C_2, C_1, , D_{11}, D_{21}$.

131.3 The Optimization Problem:

An H -optimal filter is designed to minimize the H norm of $\tilde{P}(s)$ in following equation.

$$\begin{aligned}\tilde{P}(s) &= \tilde{C}_1(sI - \tilde{A})^{-1}\tilde{B}_1 + \tilde{D}_{11}, \\ &\quad \text{where} \\ \tilde{A} &= \begin{bmatrix} A & 0 \\ B_f C_2 & A_f \end{bmatrix} < 0 \\ \tilde{B}_1 &= \begin{bmatrix} B_1 \\ B_f D_{21} \end{bmatrix} < 0 \\ \tilde{C}_1 &= [C_1 - D_f C_2 - C_f] < 0 \\ \tilde{D}_{11} &= D_{11} - D_f D_{21}\end{aligned}$$

131.4 The LMI: H - Optimal filter

Solve for $A_n \in \mathbb{R}^{n_x \times n_x}, B_n \in \mathbb{R}^{n_x \times n_y}, C_f \in \mathbb{R}^{n_x \times n_x}$, $X, Y \in \mathbb{S}^{n_x}$ and $\nu \in \mathbb{R}_{>0}$ that minimize $\zeta(\nu) = \nu$ subject to $X > 0, Y > 0$.

$$\begin{bmatrix} YA + A^T Y + B_n C_2 & A_n + C_2^T B_n^T + A^T X & Y B_1 + B_n D_{21} & C_1^T - C_2^T D_f^T \\ * & A_n + A_n^T & X B_1 + B_n D_{21} & -C_f^T \\ * & * & -\gamma I & D_1 1^T - D_2 1^T D_f^T \\ * & * & * & -\gamma I \end{bmatrix} < 0$$

$$Y - X > 0$$

131.5 Conclusion:

The filter is recovered by $A_f = X^{-1}A_n$ and $B_f = X^{-1}B_n$.

131.6 Implementation

- <https://github.com/Ricky-10/coding107/blob/master/Hinfinity%20optimal%20filter-H-Optimal%20Filter>

131.7 External links

- <http://home.eng.iastate.edu/~julied/classes/ee524/LectureNotes/110.pdf> - Optimal Filtering
- LMI Properties and Applications in Systems, Stability, and Control Theory¹ - A List of LMIs by Ryan Caverly and James Forbes.

¹ <https://arxiv.org/abs/1903.08599>

132 FDI Filter Design For Systems With Sensor Faults: an LMI

FDI Filter Design For Systems With Sensor Faults: an LMI

Systems with faulty sensors are a very common type of systems. In many cases, redundancy is added in the form of additional sensors, but in certain cases it could be a costly solution. For general linear system models, the LMI in this section can be utilized to design state estimators which can detect and isolate faulty sensor readings in order to mitigate their effects.

132.1 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1w(t) + B_2u(t) \\ y(t) &= Cx(t) + v(t)\end{aligned}$$

where $x(t) \in \mathbb{R}^n$ is the state, $u(t) \in \mathbb{R}^m$ is the control input, $w(t) \in \mathbb{R}^r$ is the process noise, $y(t) \in \mathbb{R}^p$ is the output and $v(t) \in \mathbb{R}^q$ is the measurement noise.

132.2 The Data

The state space matrices (A, B_1, B_2, C) are required to be known.

132.3 The Optimization LMI

The following LMI is used to design the Fault Detection and Isolation (FDI) filter:

$$\begin{aligned}& \min_{\Phi, \Theta, \gamma} \gamma, \\ & \text{subj. to: } \Phi > 0, \\ & \left[\begin{array}{cccc} A^\top \Phi + \Phi A + C^\top \Theta^\top + \Theta C & \Phi B_1 & \Theta & I \\ * & -\gamma I & 0 & 0 \\ * & * & -\gamma I & 0 \\ * & * & * & -\gamma I \end{array} \right] < 0\end{aligned}$$

Then the filter is $K = \Phi^{-1}\Theta$.

132.4 Conclusion:

The LMI designed in this section is used to design filters that can work on systems that are prone to sensors getting damaged or faulty.

132.5 Implementation

To solve the feasibility LMI, YALMIP toolbox is required for setting up the feasibility problem, and SeDuMi is required to solve the problem. The following link showcases an example of the feasibility problem:

https://github.com/smhassaan/LMI-Examples/blob/master/FDI_Filter_example.m

132.6 Related LMIs

H-infinity Optimal Filter¹

H-infinity Optimal Observer²

132.7 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

132.8 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/HinfinityOptimalfilter

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/HinfinityOptimalObserver

3 http://control.asu.edu/MAE598_frame.htm

4 <https://arxiv.org/abs/1903.08599/>

5 <https://web.stanford.edu/~boyd/lmibook/>

133 H_2 Optimal State estimation

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The H_2 norm of a stable system H is the root-mean-square of the impulse response of the system. The H_2 norm measures the steady-state covariance (or power) of the output response to unit noise input. In this module, the goal of H_2 optimal state estimation is to design an observer that minimizes the H_2 norm of the closed loop transfer matrix

133.1 The System

Consider the continuous-time generalized plant \mathbf{P} with state-space realization

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1u(t) + B_2w(t) & x(0) &= x_0 \\ y(t) &= C_1x(t) + D_1u(t) + D_2w(t) \\ z(t) &= C_2x(t)\end{aligned}$$

where it is assumed that (A, C_2) is detectable. An observer of the form

133.2 The Data

- $x \in \mathbb{R}^n$, $y \in \mathbb{R}^l$, $z \in \mathbb{R}^m$ are respectively the state vector, the measured output vector, and the output vector of interests
- $w \in \mathbb{R}^p$ and $u \in \mathbb{R}^r$ are the disturbance vector and the control vector, respectively
- \mathbf{A} , \mathbf{B}_1 , \mathbf{B}_2 , \mathbf{C}_1 , \mathbf{C}_2 , \mathbf{D}_1 , and \mathbf{D}_2 are the system coefficient matrices of appropriate dimensions

133.3 The Optimization Problem

Given the system and a positive scalar γ we have to find the matrix \mathbf{L} such that

$$\|G_{zw}(s)\|_2 < \gamma$$

An observer of the form

$$\begin{aligned}\dot{x}(t) &= Ax(t) + (Ly - Ly) + B_1u(t) + B_2w(t) \\ \dot{x}(t) &= (A + LC_1)x(t) - Ly + (B_1 + LD_1)u(t) \\ &\quad + (B_2 + LD_2)w(t)\end{aligned}$$

is to be designed, where L is the observer gain.

Defining the error state as

$$e = x - \hat{x}$$

The break dynamics are found to be

$$\begin{aligned}\dot{e} &= (A + LC_1)e + (B_2 + LD_2)w \\ \bar{z}(t) &= C_2e\end{aligned}$$

For the system we introduce a full state observer in the following form:

$\dot{\hat{x}} = (A + LC_1)\hat{x} - Ly + (B_1 + LD_1)u$ \hat{x} , L are the observation vector and the observer gain.

The transfer function for this case is $G_{zw}(s) = C_2(sI - A - LC_1)^{-1}(B_2 + LD_2)$ and thus the problem of H_2 state observer design is to find L such that $\|G_{zs}(s)\|_2 < \gamma$

133.4 The LMI: LMI for H_2 Observer estimation

The H_2 state observer problem has a solution if and only if there exists a matrix W , a symmetric matrix Q and a symmetric matrix X such that

$$\begin{bmatrix} XA + WC_1 + (XA + WC_1)^T & XB_2 + WD_2 \\ (XB_2 + WD_2)^T & -I \end{bmatrix}$$

$$< 0$$

$$\begin{bmatrix} -Q & C_2 \\ C_2^T & -X \end{bmatrix}$$

$$< 0$$

$$\text{trace}(Q) < \gamma^2$$

and from the solution of the above LMIs we can obtain the observer matrix as

$$L = X^{-1}W$$

133.5 Conclusion

Thus by formulation, we have converted the problem of H_2 state observer design into an LMI feasibility problem by optimizing the above LMIs. In application we are often concerned with the problem of finding the minimal attenuation level γ

On implementation and optimization of the above LMI using YALMIP and MOSEK (or) SeDuMi we get 3 matrices as output, X, W and Q and also ρ which is used to calculate γ which is the H_2 norm of the system.

There exists another set of LMIs which holds true for the same optimization problem as above.

$$A^T Y + C_1^T V^T + YA + VC_1 + C_2^T C_2 < 0$$

$$\begin{bmatrix} -Z & YB_2 + VD_2 \\ (YB_2 + VD_2)^T & -Y \end{bmatrix} < 0$$

$$\text{trace}(Z) < \gamma^2$$

When a minimal ρ is obtained, the minimal attenuation level is $\gamma = \sqrt{\rho}$

133.6 Implementation

A link to the Matlab code for a simple implementation of this problem in the Github repository:

<https://github.com/yashgvd/ygovada>

133.7 Related LMIs

H_∞ State Observer Design

Discrete time H_2 State Observer Design

133.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.

133.9 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599/>

134 Hurwitz Detectability

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134.1 Hurwitz Detectability

Hurwitz detectability is a dual concept of Hurwitz stabilizability and is defined as the matrix pair (A, C) , is said to be Hurwitz detectable if there exists a real matrix L such that $A + LC$ is Hurwitz stable.

134.2 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t) \\ x(0) &= x_0\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$, $u(t) \in \mathbb{R}^q$, at any $t \in \mathbb{R}$.

134.3 The Data

- The matrices A, B, C, D are system matrices of appropriate dimensions and are known.

134.4 The Optimization Problem

There exist a symmetric positive definite matrix P and a matrix W satisfying

$$A^T P + PA + W^T C + C^T W < 0$$

There exists a symmetric positive definite matrix P satisfying

$$N_c^T (A^T P + PA) N_c < 0$$

with N_c being the right orthogonal complement of C .

There exists a symmetric positive definite matrix P such that

$$A^T P + PA < \gamma C^T C$$

for some scalar $\gamma > 0$

134.5 The LMI:

Matrix pair (A, C) , is Hurwitz detectable if and only if following LMI holds

- $A^T P + PA + W^T C + C^T W < 0$.
- $N_c^T (A^T P + PA) N_c < 0$
- $A^T P + PA - \gamma C^T C < 0$

134.6 Conclusion:

Thus by proving the above conditions we prove that the matrix pair (A, C) is Hurwitz Detectable.

134.7 Implementation

Find the MATLAB implementation at this link below<br \> Hurwitz detectability¹

134.8 Related LMIs

Links to other closely-related LMIs

LMI for Hurwitz stability²

LMI for Schur stability³

Schur Detectability⁴

1 https://github.com/JalpeshBhadra/LMI/blob/master/hurwitz_detect.m
2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Hurwitz_Stability
3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/SchurStabilization
4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Schur_Detectability#Related_LMIs

134.9 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory - A downloadable book on LMIs by Stephen Boyd.

134.10 Return to Main Page:

- https://en.wikibooks.org/wiki/LMIs_in_Control

135 Full-Order State Observer

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135.1 Full-Order State Observer

The problem of constructing a simple full-order state observer directly follows from the result of Hurwitz detectability LMI's, Which essentially is the dual of Hurwitz stabilizability. If a feasible solution to the first LMI for Hurwitz detectability exist then using the results we can back out a full state observer L such that $A + LC$ is Hurwitz stable.

135.2 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t) \\ x(0) &= x_0\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$, $u(t) \in \mathbb{R}^q$, at any $t \in \mathbb{R}$.

135.3 The Data

- The matrices A, B, C, D are system matrices of appropriate dimensions and are known.

135.4 The Optimization Problem

The full-order state observer problem essential is finding a positive definite P such that the following LMI conclusions hold.

135.5 The LMI:

- 1) The full-order state observer problem has a solution if and only if there exist a symmetric positive definite Matrix P and a matrix W that satisfy

- $A^T P + PA + W^T C + C^T W < 0$.

Then the observer can be obtained as $L = P^{-1}W$

2) The full-state state observer can be found if and only if there is a symmetric positive definite Matrix P that satisfies the below Matrix inequality

- $A^T P + PA - C^T C < 0$

In this case the observer can be reconstructed as $L = -\frac{1}{2}P^{-1}C^T$. It can be seen that the second relation can be directly obtained by substituting $W = -\frac{1}{2}C^T$ in the first condition.

135.6 Conclusion:

Hence, both the above LMI's result in a full-order observer L such that $A + LC$ is Hurwitz stable.

135.7 External Links

A list of references documenting and validating the LMI.

- LMIs in Control Systems Analysis, Design and Applications - Duan and Yu
- LMI Methods in Optimal and Robust Control - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory - A downloadable book on LMIs by Stephen Boyd.

135.8 Return to Main Page:

136 Full-Order H-infinity State Observer

In this section, we design full order $H\infty$ state observer.

136.1 The System

Given a state-space representation of a linear system

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1u(t) + B_2w(t), x(0) = x_0 \\ y(t) &= C_1x(t) + D_1u(t) + D_2w(t) \\ z(t) &= C_2x(t)\end{aligned}$$

- $x \in \mathbb{R}^n, y \in \mathbb{R}^l, z \in \mathbb{R}^m$ are the state vector, measured output vector and output vectors of interest.
- $w \in \mathbb{R}^p, u \in \mathbb{R}^r$, are the disturbance vector and control vector respectively.

136.2 The Data

$A, B_1, B_2, C_1, C_2, D_1, D_2$ are system matrices

136.3 Definition

For the system , a full order state observer of the form of equation (1) is introduced and the estimate of interested output is given by .

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$$:\dot{\hat{x}} = (A + LC_1)\hat{x} - Ly + (B_1 + LD_1)u$$

UNKNOWN TEMPLATE EquationRef

1

The estimate of interested output is

UNKNOWN TEMPLATE NumBlk

$$:\hat{z}(t) = C_2\hat{x}(t)$$

UNKNOWN TEMPLATE EquationRef

2

Given the system and a positive scalar γ , L is found such that

UNKNOWN TEMPLATE NumBlk

$$:\|G_{zw}(s)\|_\infty = \gamma$$

UNKNOWN TEMPLATE EquationRef

3

136.4 LMI Condition

The H_∞ state observers problem has a solution if and only if there exists a symmetric positive definite matrix P and a matrix W satisfying the below LMI

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$$:\begin{bmatrix} A^T P + C_1^T W^T + PA + WC_1 & PB_2 + WD_2 & C_2^T \\ (PB_2 + WD_2)^T & -\gamma I & 0 \\ C_2 & 0 & -\gamma I \end{bmatrix}$$

UNKNOWN TEMPLATE EquationRef

4

When such a pair of matrices is found, the solution is

UNKNOWN TEMPLATE NumBlk

$$:L = P^{-1}W$$

UNKNOWN TEMPLATE EquationRef

5

136.5 Implementation

This implementation requires Yalmip and Mosek.

- https://github.com/ShenoyVaradaraya/LMI--master/blob/main/hinf_obs.m

136.6 Conclusion

Thus, an H_∞ state observer is designed such that the output vectors of interest are accurately estimated.

136.7 External Links

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications³ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013

1 http://control.asu.edu/MAE598_frame.htm

2 <https://web.stanford.edu/~boyd/lmibook/>

3 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

137 Reduced-Order State Observer

UNKNOWN TEMPLATE FULLPAGENAME

137.1 Reduced Order State Observer

The Reduced Order State Observer design paradigm follows naturally from the design of Full Order State Observer.

137.2 The System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t) + Du(t) \\ x(0) &= x_0\end{aligned}$$

where $x(t) \in \mathbb{R}^n$, $y(t) \in \mathbb{R}^m$, $u(t) \in \mathbb{R}^q$, at any $t \in \mathbb{R}$.

137.3 The Data

- The matrices A, B, C, D are system matrices of appropriate dimensions and are known.

137.4 The Problem Formulation

Given a State-space representation of a system given as above. First an arbitrary matrix $R \in \mathbb{R}^{(n-m) \times n}$ is chosen such that the vertical augmented matrix given as

$$T = \begin{bmatrix} C \\ R \end{bmatrix}$$

is nonsingular, then

$$CT^{-1} = \begin{bmatrix} I_m & 0 \end{bmatrix}$$

Furthermore, let

$$TAT^{-1} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, A_{11} \in \mathbb{R}^{m \times m}$$

then the matrix pair (A_{22}, A_{12}) is detectable if and only if (A, C) is detectable, then let

$$Tx = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, TB = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix}$$

then a new system of the form given below can be obtained

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} + \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} u, y = x_1$$

once an estimate of x_2 is obtained the the full state estimate can be given as

$$\hat{x} = T^{-1} \begin{bmatrix} y \\ \hat{x}_2 \end{bmatrix}$$

the the reduced order observer can be obtained in the form.

$$\begin{aligned} \dot{z} &= Fz + Gy + Hu, \\ \hat{x}_2 &= Mz + Ny \end{aligned}$$

Such that for arbitrary control and arbitrary initial system values, There holds

$$\lim_{t \rightarrow \infty} (x_2 - \hat{x}_2) = 0$$

The value for F, G, H, M, N can be obtain by solving the following LMI.

137.5 The LMI:

The reduced-order observer exists if and only if one of the two conditions holds.

1) There exist a symmetric positive definite Matrix P and a matrix W that satisfy

$$\bullet A_{22}^T P + PA_{22} + W_{12}^A + A_{12}^T W < 0.$$

Then $L = P^{-1}W$

2) There exist a symmetric positive definite Matrix P that satisfies the below Matrix inequality

- $A_{22}^T P + PA_{22} - A_{12}^T A_{12} < 0$

Then $L = -\frac{1}{2}P^{-1}A_{12}^T$.

By using this value of L we can reconstruct the observer state matrices as

$$F = A_{22} + LA_{12}, G = (A_{21} + LA_{11}) - (A_{22} + LA_{12})L, H = B_2 + LB_1, M = I, N = -L,$$

137.6 Conclusion:

Hence, we are able to form a reduced-order observer using which we can back off full state information as per the equation given at the end of the problem formulation given above.

137.7 External Links

A list of references documenting and validating the LMI.

- LMIs in Control Systems Analysis, Design and Applications - Duan and Yu
- LMI Methods in Optimal and Robust Control - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory - A downloadable book on LMIs by Stephen Boyd.

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138 Optimal Observer; Mixed

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In many applications, perhaps even most, the state of the system cannot be directly known. In this case, you will need to strategically measure key system outputs that will make the system states indirectly observable. Observers need to converge much faster than the system dynamics in order for their estimations to be accurate. Optimal observer synthesis is therefore advantageous. In this LMI, we seek to optimize both H₂ and H_{inf} norms, to minimize both the average and the maximum error of the observer.

138.1 The System

$$\begin{aligned}x_{k+1} &= A_d x_k + B_{d1,1} w_{1,k} + B_{d1,2} w_{2,k}, \\y_k &= C_{c2} x_k + D_{d21,1} w_{1,k} + D_{d21,2} w_{2,k}\end{aligned}$$

where $x \in R^n$ and is the state vector, $A \in R^{n*n}$ and is the state matrix, $B \in R^{n*r}$ and is the input matrix, $w \in R^r$ and is the exogenous input, $C \in R^{m*n}$ and is the output matrix, $D \in R^{m*r}$ and is the feedthrough matrix, $y \in R^m$ and is the output, and it is assumed that (A_d, C_{d2}) is detectable.

$$A \in R^{n*n}$$

138.2 The Data

The matrices $A_d, B_{d1}, C_{cd2}, C_{cd1}, D_{d21}$.

138.3 The Optimization Problem

An observer of the form:

$$\begin{aligned}\hat{x}_{k+1} &= A_d \hat{x}_k + L_d (y_k - \hat{y}_k), \\\hat{y}_k &= C_{d2} \hat{x}_k\end{aligned}$$

is to be designed, where $L_d \in R^{n_x * n_y}$ is the observer gain.

Defining the error state $e_k = x_k - \hat{x}_k$, the error dynamics are found to be

$$e_{k+1} = (A_d - L_d C_{d2}) e_k + (B_{d1,1} - L_d D_{d21,1}) w_{1,k} + (B_{d1,2} - L_d D_{d21,2}) w_{2,k},$$

and the performance output is defined as

$$\begin{bmatrix} Z_{1,k} \\ Z_{2,k} \end{bmatrix} = \begin{bmatrix} C_{d1,1} \\ C_{d1,2} \end{bmatrix} e_k + \begin{bmatrix} 0 & D_{d11,12} \\ D_{d11,21} & D_{d11,22} \end{bmatrix} \begin{bmatrix} w_{1,k} \\ w_{2,k} \end{bmatrix}.$$

The observer gain L_d is to be designed to minimize the H_2 norm of the closed loop transfer matrix $T_{11}(z)$ from the exogenous input $w_{2,k}$ to the performance output $z_{2,k}$ is less than γ_d , where

$$T_{11}(z) = C_{d1,1}(zI - (A_d - L_d C_{d2}))^{-1}(B_{d1,1} - L_d D_{d21,1}),$$

$$T_{22}(z) = C_{d1,2}(zI - (A_d - L_d C_{d2}))^{-1}(B_{d1,2} - L_d D_{d21,2}) + D_{d11,22}$$

138.4 The LMI: Discrete-Time Mixed H2-Hinf-Optimal Observer

The discrete-time mixed- H_2H_{inf} -optimal observer gain is synthesized by solving for $P \in S^{n_x}$, $Z \in S^{n_z}$, $G_d \in R^{n_x * n_y}$, and $v \in R_{>0}$ that minimize $J(v) = v$ subject to $P > 0, Z > 0$,

$$\begin{aligned} & \begin{bmatrix} P & PA_d - G_d C_{d2} & PB_{d1,1} - G_d D_{d21,1} \\ * & P & 0 \\ * & * & 1 \end{bmatrix} > 0, \\ & \begin{bmatrix} P & PA_d - G_d C_{d2} & PB_{d1,2} - G_d D_{d21,2} & 0 \\ * & P & 0 & C_{d1,2}^T \\ * & * & \gamma_d I & D_{d11,22}^T \\ * & * & * & \gamma_d I \end{bmatrix} > 0, \\ & \begin{bmatrix} Z & PC_{d1,1} \\ * & P \end{bmatrix} > 0, \\ & tr Z < v \end{aligned}$$

where tr refers to the trace of a matrix.

138.5 Conclusion:

The mixed- H_2H_{inf} -optimal observer gain is recovered by $L_d = P^{-1}G_d$, the H_2 norm of $T_{11}(z)$ is less than $\mu = \sqrt{v}$, and the H_{inf} norm of $T_{22}(z)$ is less than γ_d . This result gives us a matrix of observer gains L_d that allow us to optimally observe the states of the system indirectly as:

$$\begin{aligned} \hat{x}_{k+1} &= A_d \hat{x}_k + L_d(y_k - \hat{y}_k), \\ \hat{y}_k &= C_{d2} \hat{x}_k \end{aligned}$$

138.6 Implementation

This implementation requires Yalmip and Sedumi.

<https://github.com/rezajamesahmed/LMImatlabcode/blob/master/mixedh2hinfobsdiscretetime.m>

138.7 Related LMIs

Discrete-Time_Hinfinity-Optimal_Observer¹

Discrete-Time_H2-Optimal_Observer²

138.8 External Links

This LMI comes from Ryan Caverly's text on LMI's (Section 5.3.2):

- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.

Other resources:

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

138.9 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete-Time_HInf-Optimal_Observer

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete-Time_H2-Optimal_Observer

3 <https://arxiv.org/abs/1903.08599>

4 http://control.asu.edu/MAE598_frame.htm

5 <https://web.stanford.edu/~boyd/lmibook/>

139 Optimal Observer; H2

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In many applications, perhaps even most, the state of the system cannot be directly known. In this case, you will need to strategically measure key system outputs that will make the system states indirectly observable. Observers need to converge much faster than the system dynamics in order for their estimations to be accurate. Optimal observer synthesis is therefore advantageous. In this LMI, we seek to optimize the H2 norm, which conceptually is minimizing the average magnitude of error in the observer.

139.1 The System

$$\begin{aligned}x_{k+1} &= A_d x_k + B_{d1} w_k, \\y_k &= C_{c2} x_k + D_{d21} w_k\end{aligned}$$

where $x \in R^n$ and is the state vector, $A \in R^{n \times n}$ and is the state matrix, $B \in R^{n \times r}$ and is the input matrix, $w \in R^r$ and is the exogenous input, $C \in R^{m \times n}$ and is the output matrix, $D \in R^{m \times r}$ and is the feedthrough matrix, $y \in R^m$ and is the output, and it is assumed that (A_d, C_{d2}) is detectable.

139.2 The Data

The matrices $A_d, B_{d1}, C_{cd2}, C_{cd1}, D_{d21}$.

139.3 The Optimization Problem

An observer of the form:

$$\begin{aligned}\hat{x}_{k+1} &= A_d \hat{x}_k + L_d (y_k - \hat{y}_k), \\ \hat{y}_k &= C_{d2} \hat{x}_k\end{aligned}$$

is to be designed, where $L_d \in R^{n_x \times n_y}$ is the observer gain.

Defining the error state $e_k = x_k - \hat{x}_k$, the error dynamics are found to be

$$e_{k+1} = (A_d - L_d C_{d2}) e_k + (B_{d1} - L_d D_{d21}) w_k,$$

and the performance output is defined as

$$z_k = C_{d1}e_k.$$

The observer gain L_d is to be designed such that the H_2 of the transfer matrix from w_k to z_k , given by

$$L_d T(z) = C_{d1}(zI - (A_d - L_d C_{d2}))^{-1}(B_{d1} - L_d D_{d21}),$$

is minimized.

139.4 The LMI: Discrete-Time H2-Optimal Observer

The discrete-time H_2 -optimal observer gain is synthesized by solving for $P \in S^{n_x}$, $Z \in S^{n_z}$, $G_d \in R^{n_x \times n_y}$, and $v \in R_{>0}$ that minimize $J(v) = v$ subject to $P > 0$, $Z > 0$,

$$\begin{aligned} \begin{bmatrix} P & PA_d - G_d C_{d2} & PB_{d1} - G_d D_{d21} \\ * & P & 0 \\ * & * & 1 \end{bmatrix} &> 0, \\ \begin{bmatrix} Z & PC_{d1} \\ * & P \end{bmatrix} &> 0, \\ \text{tr } Z < v \end{aligned}$$

where tr refers to the trace of a matrix.

139.5 Conclusion:

The H_2 -optimal observer gain is recovered by $L_d = P^{-1}G_d$ and the H_2 norm of $T(z)$ is $\mu = \sqrt{v}$. The L_d matrix is the observer gains that can be used to form the optimal observer:

$$\begin{aligned} \hat{x}_{k+1} &= A_d \hat{x}_k + L_d(y_k - \hat{y}_k), \\ \hat{y}_k &= C_{d2} \hat{x}_k \end{aligned}$$

139.6 Implementation

This implementation requires Yalmip and Sedumi.

https://github.com/rezajamesahmed/LMImatlabcode/blob/master/Discrete_Time_H2_Optimal_Observer_LMIs_Wikibook_Example.m

139.7 Related LMIs

Mixed H2-Hinfinity discrete time observer¹

Discrete-Time_Hinfinity-Optimal_Observer²

139.8 External Links

This LMI comes from Ryan Caverly's text on LMI's (Section 5.1.2):

- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.

Other resources:

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

139.9 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/discrete_time_mixed_h2_hinf_optimal_observer#Implementation

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete-Time_HInf-Optimal_Observer

3 <https://arxiv.org/abs/1903.08599/>

4 http://control.asu.edu/MAE598_frame.htm

5 <https://web.stanford.edu/~boyd/lmibook/>

140 Optimal Observer; H_{inf}

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In many applications, perhaps even most, the state of the system cannot be directly known. In this case, you will need to strategically measure key system outputs that will make the system states indirectly observable. Observers need to converge much faster than the system dynamics in order for their estimations to be accurate. Optimal observer synthesis is therefore advantageous. In this LMI, we seek to optimize the H-infinity norm, which conceptually is minimizing the maximum magnitude of error in the observer.

140.1 The System

The system needed for this LMI is a discrete-time LTI plant G , which has the state space realization:

$$\begin{aligned}x_{k+1} &= A_d x_k + B_{d1} w_k, \\y_k &= C_{d2} x_k + D_{d21} w_k\end{aligned}$$

where $x \in R^n$ and is the state vector, $A \in R^{n*n}$ and is the state matrix, $B \in R^{n*r}$ and is the input matrix, $w \in R^r$ and is the exogenous input, $C \in R^{m*n}$ and is the output matrix, $D \in R^{m*r}$ and is the feedthrough matrix, $y \in R^m$ and is the output, and it is assumed that (A_d, C_{d2}) is detectable.

140.2 The Data

The matrices $A_d, B_{d1}, C_{d2}, D_{d21}, D_{d11}$.

140.3 The Optimization Problem

An observer of the form:

$$\begin{aligned}\hat{x}_{k+1} &= A_d \hat{x}_k + L_d (y_k - \hat{y}_k), \\ \hat{y}_k &= C_{d2} \hat{x}_k\end{aligned}$$

is to be designed, where $L_d \in R^{n_x * n_y}$ is the observer gain.

Defining the error state $e_k = x_k - \hat{x}_k$, the error dynamics are found to be

$$e_{k+1} = (A_d - L_d C_{d2}) e_k + (B_{d1} - L_d D_{d21}) w_k,$$

and the performance output is defined as

$$z_k = C_{d1} e_k + D_{d11} w_k.$$

The observer gain L_d is to be designed such that the H_{inf} of the transfer matrix from w_k to z_k , given by

$$T(z) = C_{d1}(zI - (A_d - L_d C_{d2}))^{-1}(B_{d1} - L_d D_{d21}) + D_{d11},$$

is minimized.

140.4 The LMI: Discrete-Time Hinfin-Optimal Observer

The discrete-time H_{inf} -optimal observer gain is synthesized by solving for $P \in S^{n_x}$, $G_d \in R^{n_x \times n_y}$, and $\gamma \in R_{>0}$ that minimize $J(\gamma) = \gamma$ subject to $P > 0$, and

$$\begin{bmatrix} P & PA_d - G_d C_{d2} & PB_{d1} - G_d D_{d21} & 0 \\ * & P & 0 & C_{d1}^T \\ * & * & \gamma * 1 & D_{d11}^T \\ * & * & * & \gamma * 1 \end{bmatrix} > 0.$$

140.5 Conclusion:

The H_{inf} -optimal observer gain is recovered by $L_d = P^{-1}G_d$ and the H_{inf} norm of $T(z)$ is γ . This matrix of observer gains can then be used to form the optimal observer formulated by:

$$\begin{aligned} \hat{x}_{k+1} &= A_d \hat{x}_k + L_d(y_k - \hat{y}_k), \\ \hat{y}_k &= C_{d2} \hat{x}_k \end{aligned}$$

140.6 Implementation

This implementation requires Yalmip and Sedumi.

<https://github.com/rezajamesahmed/LMImatlabcode/blob/master/Hinfobsdiscretetime.m>

140.7 Related LMIs

Mixed H2-Hinfinity discrete time observer¹

Discrete-Time_H2-Optimal_Observer²

140.8 External Links

This LMI comes from Ryan Caverly's text on LMI's (Section 5.2.2):

- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.

Other resources:

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

140.9 Return to Main Page:

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/discrete_time_mixed_h2_hinf_optimal_observer#Implementation
2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete-Time_H2-Optimal_Observer
3 <https://arxiv.org/abs/1903.08599>
4 http://control.asu.edu/MAE598_frame.htm
5 <https://web.stanford.edu/~boyd/lmibook/>

141 Discrete Time Detectability

Discrete-Time Detectability

A discrete time system operates on a discrete time signal input and produces a discrete time signal output. They are used in digital signal processing, such as digital filters for images or sound. The class of discrete time systems that are both linear and time invariant, known as discrete time LTI systems.

Discrete-Time LTI systems can be made detectable using observer gain L , which can be found using LMI optimization, such that the close loop system is detectable.

141.1 The System

Discrete-Time LTI System with state space realization (A_d, B_d, C_d, D_d)

$$A_d \in \mathbf{R}^{n*n}, \quad B_d \in \mathbf{R}^{n*m}, \quad C_d \in \mathbf{R}^{p*n}, \quad D_d \in \mathbf{R}^{p*m}$$

141.2 The Data

The matrices: System $(A_d, B_d, C_d, D_d), P, W$.

141.3 The Optimization Problem

The following feasibility problem should be optimized:

Maximize P while obeying the LMI constraints.

Then L is found.

141.4 The LMI:

Discrete-Time Detectability

The LMI formulation

$$P \in S^n; W \in R^{m*n}$$

$$P > 0$$

$$\begin{bmatrix} P & A_d^T P + C_d^T W \\ * & P \end{bmatrix} > 0,$$
$$L = P^{-1}W$$

141.5 Conclusion:

The system is detectable iff there exists a P , such that $P > 0$. The matrix $A_d + LC_d$ is Schur with $L = P^{-1}W$

141.6 Implementation

A link to CodeOcean or other online implementation of the LMI
MATLAB Code¹

141.7 Related LMIs

https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Detectability_LMI - Continuous time Detectability

141.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.

141.9 Return to Main Page:

1 <https://github.com/Harishankar-Prabhakaran/LMIs/blob/master/A5.m>

2 http://control.asu.edu/MAE598_frame.htm

3 <https://arxiv.org/abs/1903.08599/>

4 <https://web.stanford.edu/~boyd/lmibook/>

142 Schur Detectability

Schur Detectability

Schur detectability is a dual concept of Schur stabilizability and is defined as follows, the matrix pair (A, C) , is said to be Schur detectable if there exists a real matrix L such that $A + LC$ is Schur stable.

142.1 The System

We consider the following system:

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) \\y(k) &= Cx(k) + Du(k)\end{aligned}$$

where the matrices $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times r}$, $C \in \mathbb{R}^{m \times n}$, $D \in \mathbb{R}^{m \times r}$, $x \in \mathbb{R}^n$, $y \in \mathbb{R}^m$, and $u \in \mathbb{R}^r$ are the state matrix, input matrix, state vector, and the input vector, respectively.

Moreover, k represents time in the discrete-time system and $k + 1$ is the next time step.

The state feedback control law is defined as follows:

$$u(k) = Kx(k)$$

where $K \in \mathbb{R}^{n \times r}$ is the controller gain. Thus, the closed-loop system is given by:

$$x(k+1) = (A + BK)x(k)$$

142.2 The Data

- The matrices A, B, C, D are system matrices of appropriate dimensions and are known.

142.3 The Optimization Problem

There exist a symmetric matrix P and a matrix W satisfying

$$\begin{bmatrix} -P & A^T P + C^T W^T \\ PA + WC & P \end{bmatrix} < 0$$

There exists a symmetric matrix P satisfying

$$\begin{bmatrix} -N_c^T P N_c & N_c^T A^T P \\ P A N_c & -P \end{bmatrix} < 0$$

with N_c being the right orthogonal complement of C .

There exists a symmetric matrix P such that

$$\begin{bmatrix} -P & PA \\ A^T P & -P - \gamma C^T C \end{bmatrix} < 0$$

$$\gamma > 1$$

142.4 The LMI:

The LMI for Schur detectability can be written as minimization of the scalar, γ , in the following constraints:

$$\min \quad \gamma$$

s.t.

$$\begin{bmatrix} -P & A^T P + C^T W^T \\ PA + WC & P \end{bmatrix} < 0$$

$$\begin{bmatrix} -N_c^T P N_c & N_c^T A^T P \\ P A N_c & -P \end{bmatrix} < 0$$

$$\begin{bmatrix} -P & PA \\ A^T P & -P - \gamma C^T C \end{bmatrix} < 0$$

142.5 Conclusion:

Thus by proving the above conditions we prove that the matrix pair (A, C) is Schur Detectable.

142.6 Implementation

A link to Matlab codes for this problem in the Github repository: Schur Detectability¹

¹ https://github.com/JalpeshBhadra/LMI/blob/master/schur_detect.m

142.7 Related LMIs

LMI for Hurwitz stability²

LMI for Schur stability³

Hurwitz Detectability⁴

142.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

142.9 Return to Main Page

LMIs in Control/Tools⁵

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Hurwitz_Stability

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/SchurStabilization

4 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Hurwitz_detectability

5 https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

143 Robust Stabilization of Second-Order Systems

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Stabilization is a vastly important concept in controls, and is no less important for second order systems with perturbations. Such a second order system can be conceptualized most simply by the model of a mass-spring-damper, with added perturbations. Velocity and position are of course chosen as the states for this system, and the state space model can be written as it is below. The goal of stabilization in this context is to design a control law that is made up of two controller gain matrices K_p , and K_d . These allow the construction of a stabilized closed loop controller.

143.1 The System

In this LMI, we have an *uncertain* second-order linear system, that can be modeled in state space as:

$$\begin{aligned}(A_2 + \Delta A_2)\ddot{x} + (A_1 + \Delta A_1)\dot{x} + (A_0 + \Delta A_0)x &= Bu \\ y_d &= C_d\dot{x} \\ y_p &= C_p x\end{aligned}$$

where $x \in R^n$ and $u \in R^r$ are the state vector and the control vector, respectively, $y_d \in R^{m_p}$ and $y_d \in R^{m_p}$ are the derivative output vector and the proportional output vector, respectively, and $A_2, A_1, A_0, B, C_d, C_p$ are the system coefficient matrices of appropriate dimensions.

$\Delta A_2, \Delta A_1$, and ΔA_0 are the perturbations of matrices A_2, A_1 , and A_0 , respectively, are bounded, and satisfy

$$|\Delta A_2|_2 \leq \epsilon_2, |\Delta A_1|_2 \leq \epsilon_1, |\Delta A_0|_2 \leq \epsilon_0,$$

or

$$\max\{\|\Delta a_{2ij}\|\} \leq \eta_2, \max\{\|\Delta a_{1ij}\|\} \leq \eta_1, \max\{\|\Delta a_{0ij}\|\} \leq \eta_0,$$

where $\epsilon_2, \epsilon_1, \epsilon_0$ and η_2, η_1, η_0 are two sets of given positive scalars, $\Delta a_{2ij}, \Delta a_{1ij}$, and Δa_{0ij} are the i-th row and j-th column elements of matrices $\Delta A_2, \Delta A_1$, and ΔA_0 ,

respectively. Also, the perturbation notations also satisfy the assumption that $\Delta A_2, \Delta A_0 \in S^n$ and $A_2 + \Delta A_2 > 0$.

To further define: $x \in R^n$ and is the state vector, A_0 is $\in R^{n*n}$ and is the state matrix on x , A_1 is $\in R^{n*n}$ and is the state matrix on \dot{x} , A_2 is $\in R^{n*n}$ and is the state matrix on \ddot{x} , B is $\in R^{n*r}$ and is the input matrix, u is $\in R^r$ and is the input, C_d and C_p are $\in R^{m*n}$ and are the output matrices, y_d is $\in R^m$ and is the output from C_d , and y_p is $\in R^m$ and is the output from C_p .

143.2 The Data

The matrices $A_2, A_1, A_0, B, C_d, C_p$ and perturbations $\Delta A_2, \Delta A_1, \Delta A_0$, describing the second order system with perturbations.

143.3 The Optimization Problem

For the system defined as shown above, we need to design a feedback control law such that the following system is Hurwitz stable. In other words, we need to find the matrices K_p and K_d in the below system.

$$(A_2 + \Delta A_2)\ddot{x} + (A_1 - BK_pC_p + \Delta A_1)\dot{x} + (A_0 - BK_dC_d + \Delta A_0)x = 0$$

However, to do proceed with the solution, first we need to present a Lemma. This Lemma comes from Appendix A.6 in "LMI's in Control systems" by Guang-Ren Duan and Hai-Hua Yu. This Lemma states the following: if $A_2 > 0, A_1 + A_1^T > 0, A_0 > 0$, then the following is also true for the system described above:

The system is hurwitz stable if

$$\lambda_{min}(A_2) > |\Delta A_2|_2, \lambda_{min}(A_1 + A_1^T) > |\Delta A_1|_2, \lambda_{min}(A_0) > |\Delta A_0|_2$$

or

the system is hurwitz stable if

$$\lambda_{min}(A_2) > \sqrt{l_2} \max\{\|\Delta a_{2ij}\|\}, \lambda_{min}(A_1 + A_1^T) > \sqrt{l_1} \max\{\|\Delta a_{1ij}\|\}, \lambda_{min}(A_0) > \sqrt{l_0} \max\{\|\Delta a_{0ij}\|\}$$

, where l_2, l_1, l_0 are the numbers of nonzero elements in matrices $\Delta A_2, \Delta A_1, \Delta A_0$, respectively.

143.4 The LMI: Robust Stabilization of Second Order Systems

This problem is solved by finding matrices $K_p \in R^{r*m_p}$ and $K_d \in R^{r*m_d}$ that satisfy either of the following sets of LMIs.

$$A_0 - BK_d C_d > \epsilon_0 I,$$

$$(A_1 - BK_p C_p) + (A_1 - BK_p C_p)^T > \epsilon_1 I.$$

or

$$A_0 - BK_d C_d > \eta_0 \sqrt{l_0} I,$$

$$(A_1 - BK_p C_p) + (A_1 - BK_p C_p)^T > \eta_1 \sqrt{l_1} I.$$

143.5 Conclusion:

Having solved the above problem, the matrices K_p and K_d can be substituted into the input as $u = K_p C_p \dot{x} + K_d C_d x$ to robustly stabilize the second order uncertain system. The new system is stable in closed loop.

143.6 Implementation

This implementation requires Yalmip and Sedumi.

<https://github.com/rezajamesahmed/LMImatlabcode/blob/master/R0Bstab2ndorder.m>

143.7 Related LMIs

Stabilization of Second-Order Systems¹

143.8 External Links

This LMI comes from

- ² - "LMIs in Control Systems: Analysis, Design and Applications" by Guang-Ren Duan and Hai-Hua Yu

Other resources:

¹ https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/pages/stabilization_of_second_order_systems&stable=0

² <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory⁴ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁵ - A downloadable book on LMIs by Stephen Boyd.

143.9 References

Duan, G. (2013). LMIs in control systems: analysis, design and applications. Boca Raton: CRC Press, Taylor & Francis Group.

143.10 Return to Main Page:

3 http://control.asu.edu/MAE598_frame.htm
4 <https://arxiv.org/abs/1903.08599/>
5 <https://web.stanford.edu/~boyd/lmibook/>

144 Robust Stabilization of $H\infty$ Optimal State Feedback Control

144.1 Robust Full State Feedback Optimal $H\infty$ Control

Additive uncertainty

Full State Feedback is a control technique which places a given system's closed loop system poles in locations specified by desired performance specifications. One can use $H\infty$ methods to turn this into an optimization problem with the goal to minimize the impact of uncertain perturbations in a closed loop system while maintaining system stability. This is done by minimizing the $H\infty$ norm of the uncertain closed loop system, which minimizes the worst case effect of the system disturbance or perturbation. This can be done for single-input single-output (SISO) or multiple-input multiple-output (MIMO) systems. Here we consider the case of a MIMO system with additive uncertainties.

144.2 The System

Consider linear system with uncertainty below:

$$\begin{bmatrix} \dot{x} \\ z \end{bmatrix} = \begin{bmatrix} (A + \Delta A) & (B_1 + \Delta B_1) & B_2 \\ C & D_1 & D_2 \end{bmatrix} \begin{bmatrix} x \\ u \\ w \end{bmatrix}$$

Where $x(t) \in \mathbb{R}^n$ is the state, $z(t) \in \mathbb{R}^m$ is the output, $w(t) \in \mathbb{R}^p$ is the exogenous input or disturbance vector, and $u(t) \in \mathbb{R}^r$ is the actuator input or control vector, at any $t \in \mathbb{R}$

ΔA and ΔB_1 are real-valued matrices which represent the time-varying parameter uncertainties in the form:

$$[\Delta A \quad \Delta B_1] = HF [E_1 \quad E_2]$$

Where

H, E_1, E_2 are known matrices with appropriate dimensions and F is the uncertain parameter matrix which satisfies: $F^T F \leq I$

For additive perturbations: $\Delta A = \delta_1 A_1 + \delta_2 A_2 + \dots + \delta_k A_k$

Where

$A_i, i = 1, 2, \dots, k$ are the known system matrices and

$\delta_i, i = 1, 2, \dots, k$ are the perturbation parameters which satisfy $|\delta_i| < r_i, i = 1, 2, \dots, k$

Thus, $\Delta A = HFE$ with

$$H = [A_1 \ A_2 \ \dots \ A_k]$$

$$E = (\sum_{i=1}^k r_i^2)^{1/2}$$

$$F = (\sum_{i=1}^k r_i^2)^{-1/2} \begin{bmatrix} \delta_1 I \\ \delta_2 I \\ \vdots \\ \delta_k I \end{bmatrix}$$

144.3 The Data

$A, B_1, B_2, C, D_1, D_2, E_1, E_2, \gamma$ are known.

144.4 The LMI:Full State Feedback Optimal H_∞ Control LMI

There exists $X > 0$ and W and scalar α such that

$$\begin{bmatrix} \Psi(X, W) & B_2 & (CX + D_1 W)^T & (E_1 X + E_2 W)^T \\ B_2^T & -\gamma I & D_2^T & 0 \\ CX + D_1 W & D_2 & -\gamma I & 0 \\ E_1 X + E_2 W & 0 & 0 & -\alpha I \end{bmatrix} < 0$$

Where $\Psi(X, W) = (AX + B_1 W)_s + \alpha HH^T$

And $K = WX^{-1}$.

144.5 Conclusion:

Once K is found from the optimization LMI above, it can be substituted into the state feedback control law $u(t) = Kx(t)$ to find the robustly stabilized closed loop system as shown below:

$$\begin{bmatrix} \dot{x} \\ z \end{bmatrix} = \begin{bmatrix} (A + \Delta A) + (B_1 + \Delta B_1)K & B_2 \\ (C + D_1)K & D_2 \end{bmatrix} \begin{bmatrix} x \\ w \end{bmatrix}$$

where $x(t) \in \mathbb{R}^n$ is the state, $z(t) \in \mathbb{R}^m$ is the output, $w(t) \in \mathbb{R}^p$ is the exogenous input or disturbance vector, and $u(t) \in \mathbb{R}^r$ is the actuator input or control vector, at any $t \in \mathbb{R}$

Finally, the transfer function of the system is denoted as follows:

$$G_{zw}(s) = (C + D_1 K)(sI - [(A + \Delta A) + (B_1 + \Delta B_1)K])^{-1} B_2 + D_2$$

144.6 Implementation

This implementation requires Yalmip and Sedumi. <https://github.com/rubindan/LMIcontrol/blob/master/HinfFilter.m>

144.7 Related LMIs

Full State Feedback Optimal H_{inf} LMI¹

144.8 External Links

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd.
- LMIs in Control Systems: Analysis, Design and Applications⁵ - by Guang-Ren Duan and Hai-Hua Yu, CRC Press, Taylor & Francis Group, 2013.

144.9 Return to Main Page:

LMIs in Control: https://en.wikibooks.org/wiki/LMIs_in_Control

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Full-State_Feedback_Optimal_Control_Hinf_LMI
 2 http://control.asu.edu/MAE598_frame.htm
 3 <https://arxiv.org/abs/1903.08599/>
 4 <https://web.stanford.edu/~boyd/lmibook/>
 5 <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996>

145 Robust H inf State Feedback Control

1. REDIRECT LMIs in Control/pages/Robust H inf State Feedback Control¹

¹ Chapter 144 on page 407

146 LMI for Time-Delay system on delay Independent Condition

146.1 The System

The problem is to check the stability of the following linear time-delay system

$$\begin{cases} \dot{x}(t) = Ax(t) + A_d x(t-d) \\ x(t) = \phi(t), t \in [-d, 0], 0 < d \leq \bar{d}, \end{cases}$$

where

$A, A_d \in \mathbb{R}^{n \times n}$, $A \in \mathbb{R}^{n \times r}$ are the system coefficient matrices,

$\phi(t)$ is the initial condition

d represents the time-delay

\bar{d} is a known upper-bound of d

146.2 The Data

The matrices A, A_d are known

146.3 The LMI: *The Time-Delay systems (Delay Independent Condition)*

From the given pieces of information, it is clear that the optimization problem only has a solution if there exists two symmetric matrices $P, S \in \mathbb{S}^n$ such that

$$P > 0$$

$$\begin{bmatrix} A^T P + PA + S & PA_d \\ A_d^T P & -S \end{bmatrix} < 0$$

This LMI has been derived from the Lyapunov function for the system. By Schur Complement we can see that the above matrix inequality is equivalent to the Riccati inequality

$$A^T P + PA + PA_d S^{-1} A_d^T P + S < 0$$

146.4 Conclusion:

We can now implement these LMIs to do stability analysis for a Time delay system on the delay independent condition

146.5 Implementation

The implementation of the above LMI can be seen here

https://github.com/yashgvd/LMI_wikibooks

146.6 Related LMIs

Time Delay systems (Delay Dependent Condition)

146.7 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- D. d. S. Madeira and J. Adamy, "Static output feedback: An LMI condition for stabilizability based on passivity indices," 2016 IEEE Conference on Control Applications (CCA), Buenos Aires, 2016, pp. 960-965.

146.8 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599/>

147 LMI for Time-Delay system on delay Dependent Condition

147.1 The System

The problem is to check the stability of the following linear time-delay system on a delay dependent condition

$$\begin{cases} \dot{x}(t) = Ax(t) + A_d x(t-d) \\ x(t) = \phi(t), t \in [-d, 0], 0 < d \leq \bar{d}, \end{cases}$$

where

$A, A_d \in \mathbb{R}^{n \times n}, A \in \mathbb{R}^{n \times r}$ are the system coefficient matrices,

$\phi(t)$ is the initial condition

d represents the time-delay

\bar{d} is a known upper-bound of d

For the purpose of the delay dependent system we rewrite the system as

$$\begin{cases} \dot{x}(t) = Ax(t) + A_d x(t-d) \\ \dot{x}(t) = (A + A_d)x(t) - A_d(x(t) - x(t-d)) \end{cases}$$

147.2 The Data

The matrices A, A_d are known

147.3 The LMI: *The Time-Delay systems (Delay Dependent Condition)*

From the given pieces of information, it is clear that the optimization problem only has a solution if there exists a symmetric positive definite matrix X and a scalar $0 < \beta < 1$ such that

$$\begin{bmatrix} \Phi(X) & \bar{d}XA^T & \bar{d}XA_d^T \\ \bar{d}AX & -d\beta I & 0 \\ \bar{d}A_dX & 0 & -\bar{d}(1-\beta)I \end{bmatrix} < 0$$

Here $\Phi(X) = X)A + A_d)^T + (A + A_d)X + \bar{d}A_dA_d^T < 0$

This LMI has been derived from the Lyapunov function for the system. It follows that the system is asymptotically stable if

$$P(A + A_d) + (A + A_d)^T P + \bar{d}PA_dA_d^T P + \frac{\bar{d}}{\beta}A^T A + \frac{\bar{d}}{1-\beta}A_d^T A_d < 0$$

This is obtained by replacing X with P^{-1}

147.4 Conclusion:

We can now implement these LMIs to do stability analysis for a Time delay system on the delay dependent condition

147.5 Implementation

The implementation of the above LMI can be seen here

https://github.com/yashgvd/LMI_wikibooks

147.6 Related LMIs

Time Delay systems (Delay Independent Condition)

147.7 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications
- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.

¹ http://control.asu.edu/MAE598_frame.htm

² <https://arxiv.org/abs/1903.08599/>

- D. d. S. Madeira and J. Adamy, "Static output feedback: An LMI condition for stabilizability based on passivity indices," 2016 IEEE Conference on Control Applications (CCA), Buenos Aires, 2016, pp. 960-965.

147.8 Return to Main Page:

148 LMI for Stability of Retarded Differential Equation with Slowly-Varying Delay

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This page describes an LMI for stability analysis of a continuous-time system with a time-varying delay. In particular, a delay-independent condition is provided to test uniform asymptotic stability of a retarded differential equation through feasibility of an LMI. The system under consideration pertains a single discrete delay, with the extent of the delay at any time bounded by some known value. Moreover, the delay is assumed to vary only slowly in time, with a temporal derivative bounded by a value less than one. Solving the LMI for a particular value of this bound, uniform asymptotic stability can be shown for any time-delay satisfying this bound.

148.1 The System

The system under consideration is one of the form:

$$\dot{x}(t) = Ax(t) + A_1x(t - \tau(t)) \quad t \geq t_0, \quad 0 \leq \tau(t) \leq h, \quad \dot{\tau}(t) \leq d < 1$$

In this description, A and A_1 are matrices in $\mathbb{R}^{n \times n}$. The variable $\tau(t)$ denotes a delay in the state at time $t \geq t_0$, assuming a value no greater than some $h \in \mathbb{R}_+$. Moreover, we assume that the function $\tau(t)$ is differentiable at any time, with the derivative bounded by some value $d < 1$, assuring the delay to be slowly-varying in time.

148.2 The Data

To determine stability of the system, the following parameters must be known:

$$A \in \mathbb{R}^{n \times n}$$

$$A_1 \in \mathbb{R}^{n \times n}$$

$$d \in [0, 1)$$

148.3 The Optimization Problem

Based on the provided data, uniform asymptotic stability can be determined by testing feasibility of the following LMI:

148.4 The LMI: Delay-Independent Uniform Asymptotic Stability for Continuous-Time TDS

Find :

$$P, Q \in \mathbb{R}^{n \times n}$$

such that:

$$P > 0, \quad Q > 0$$

$$\begin{bmatrix} A^T P + PA + Q & PA_1 \\ A_1^T P & -(1-d)Q \end{bmatrix} < 0$$

148.5 Conclusion:

If the presented LMI is feasible, the system will be uniformly asymptotically stable for any delay function $\tau(t)$ satisfying $\dot{\tau}(t) \leq d < 1$. That is, independent of the values of the delays $\tau(t)$ and the starting time $t_0 \in \mathbb{R}$:

- For any real number $\epsilon > 0$, there exists a real number $\delta > 0$ such that:

$${}_{t_0}\|c < \delta \Rightarrow \|c(t)\| < \epsilon \quad \forall t \geq t_0$$

- There exists a real number $\delta_a > 0$ such that for any real number $\eta > 0$, there exists a time $T(\delta_a, \eta)$ such that:

$${}_{t_0}\|c < \delta_a \Rightarrow \|c(t)\| < \eta \quad \forall t \geq t_0 + T(\delta_a, \eta)$$

Here, we let $x_{t_0}(\theta) = x(t_0 + \theta)$ for $\theta \in [-\tau(t_0), 0]$ denote the delayed state function at time t_0 . The norm ${}_{t_0}\|c$ of this function is defined as the maximal value of the vector norm assumed by the state over the delayed time interval, given by:

$${}_{t_0}\|c := \max_{\theta \in [-\tau(t_0), 0]} \|x(t_0 + \theta)\|$$

Obtaining a feasible point for the LMI, this result can be proven using a Lyapunov-Krasovskii functional:

$$V(t, x_t) = x^T(t)Px(t) + \int_{t-\tau(t)}^t x^T(s)Qx(s)ds$$

Notably, if matrices $P > 0, Q > 0$ prove feasibility of the LMI for the pair (A, A_1) , these same matrices will also prove feasibility of the LMI for the pair $(A, -A_1)$. As such, feasibility of this LMI proves uniform asymptotic stability of both systems:

$$\dot{x}(t) = Ax(t) \pm A_1x(t - \tau(t)) \quad t \geq t_0, \quad 0 \leq \tau(t) \leq h \quad \dot{\tau}(t) \leq d < 1$$

Moreover, since the result is independent of the value of the delay, it will also hold for a delay $\tau(t) \equiv 0$. Hence, if the LMI is feasible, the matrices $A \pm A_1$ will be Hurwitz.

148.6 Implementation

An example of the implementation of this LMI in Matlab is provided on the following site:

- https://github.com/djagt/LMI_Codes/blob/main/stblty_cTDS_SlowVarying.m

Note that this implementation requires packages for YALMIP with solver mosek, though a different solver can also be implemented.

148.7 Related LMIs

- https://en.wikibooks.org/wiki/LMIs_in_Control/pages/TDSDC - Delay-dependent stability LMI for continuous-time TDS
- https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/Discrete_Time/Stability_Condition_for_Discrete-Time_TDS - Stability LMI for delayed discrete-time system

148.8 External Links

The presented results have been obtained from:

- Fridman E. 2014. Introduction to Time-Delay Systems, Analysis and Control. Springer. ISBN: 978-3-319-09392-5.

Additional information on LMI's in control theory can be obtained from the following resources:

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.

¹ http://control.asu.edu/MAE598_frame.htm

- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

148.9 Return to Main Page:

2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

149 LMI for Robust Stability of Retarded Differential Equation with Norm-Bounded Uncertainty

UNKNOWN TEMPLATE FULLPAGENAME

This page describes an LMI for stability analysis of an uncertain continuous-time system with a time-varying delay. In particular, a delay-independent condition is provided to test uniform asymptotic stability of a retarded differential equation with uncertain matrices through feasibility of an LMI. The system under consideration pertains a single discrete delay, with the extent of the delay at any time bounded by some known value. The matrices describing the system are assumed to be uncertain, with the norm of the uncertainty bounded by a value of one. In addition, the delay is assumed to vary only slowly in time, with a temporal derivative bounded by a value less than one. Solving the LMI for a particular value of this bound, uniform asymptotic stability can be shown for any time-delay satisfying this bound, independent of the value of the uncertainty function.

149.1 The System

The system under consideration is one of the form:

$$\dot{x}(t) = (A + H\Delta(t)E)x(t) + (A_1 + H\Delta(t)E_1)x(t - \tau(t)) \quad t \geq t_0, \quad 0 \leq \tau(t) \leq h, \quad \dot{\tau}(t) \leq d < 1$$

In this description, A and A_1 are matrices in $\mathbb{R}^{n \times n}$. The variable $\tau(t)$ denotes a delay in the state at time $t \geq t_0$, assuming a value no greater than some $h \in \mathbb{R}_+$. Moreover, we assume that the function $\tau(t)$ is differentiable at any time, with the derivative bounded by some value $d < 1$, assuring the delay to be slowly-varying in time. The uncertainty $\Delta(t) \in \mathbb{R}^{r_1 \times r_2}$ is also allowed to vary in time, but at any time $t \geq t_0$ must satisfy the inequality:

$$\Delta^T(t)\Delta(t) \leq I$$

The uncertainty affects the system through matrices $H \in \mathbb{R}^{n \times r_1}$ and $E, E_1 \in \mathbb{R}^{r_2 \times n}$, which are constant in time and assumed to be known.

149.2 The Data

To determine stability of the system, the following parameters must be known:

$$A, A_1 \in \mathbb{R}^{n \times n}$$

$$H \in \mathbb{R}^{n \times r_1}$$

$$E, E_1 \in \mathbb{R}^{r_2 \times n}$$

$$d \in [0, 1)$$

149.3 The Optimization Problem

Based on the provided data, uniform asymptotic stability can be determined by testing feasibility of the following LMI:

149.4 The LMI: Delay-Independent Robust Uniform Asymptotic Stability for Continuous-Time TDS

Find :

$$\epsilon \in \mathbb{R}, \quad P, Q \in \mathbb{R}^{n \times n}$$

such that:

$$\epsilon > 0, \quad P > 0, \quad Q > 0$$

$$\begin{bmatrix} A^T P + PA + Q & PA_1 & PH & \epsilon E^T \\ A_1^T P & -(1-d)Q & 0 & \epsilon E_1^T \\ H^T P & 0 & -\epsilon I & 0 \\ \epsilon E & \epsilon E_1 & 0 & -\epsilon I \end{bmatrix} < 0$$

149.5 Conclusion:

If the presented LMI is feasible, the system will be uniformly asymptotically stable for any delay function $\tau(t)$ satisfying $\dot{\tau}(t) \leq d < 1$, and any uncertainty $\Delta(t)$ satisfying $\Delta^T(t)\Delta(t) \leq I$. That is, independent of the values of the delays $\tau(t)$, uncertainties $\Delta(t)$, and the starting time $t_0 \in \mathbb{R}$:

- For any real number $\epsilon > 0$, there exists a real number $\delta > 0$ such that:

$$t_0 \|c|_C < \delta \Rightarrow \|c(t)\| < \epsilon \quad \forall t \geq t_0$$

- There exists a real number $\delta_a > 0$ such that for any real number $\eta > 0$, there exists a time $T(\delta_a, \eta)$ such that:

$$t_0 \|c|_C < \delta_a \Rightarrow \|c(t)\| < \eta \quad \forall t \geq t_0 + T(\delta_a, \eta)$$

Here, we let $x_{t_0}(\theta) = x(t_0 + \theta)$ for $\theta \in [-\tau(t_0), 0]$ denote the delayed state function at time t_0 . The norm ${}_{t_0}\| \cdot \|_{\mathcal{C}}$ of this function is defined as the maximal value of the vector norm assumed by the state over the delayed time interval, given by:

$${}_{t_0}\| \cdot \|_{\mathcal{C}} := \max_{\theta \in [-\tau(t_0), 0]} \| (t_0 + \theta) \|$$

The proof of this result relies on the fact that the following inequality holds for any value $\epsilon > 0$ and constant matrices α, β of appropriate dimensions:

$$\alpha \Delta(t) \beta + \beta^T \Delta^T(t) \alpha^T \leq \epsilon^{-1} \alpha \alpha^T + \epsilon \beta^T \beta$$

Using this inequality with $\alpha^T = [H^T P \ 0]$ and $\beta = [E \ E_1]$, the described LMI can then be derived from that presented in https://en.wikibooks.org/wiki/LMIs_in_Control/Time-Delay_Systems/Continuous_Time/LMI_for_Stability_of_Retarded_Differential_Equation_with_Slowly-Varying_Delay, corresponding to a situation without uncertainty.

149.6 Implementation

An example of the implementation of this LMI in Matlab is provided on the following site:

- https://github.com/djagt/LMI_Codes/blob/main/Rstblty_cTDS_SlowVarying.m

Note that this implementation requires packages for YALMIP with solver mosek, though a different solver can also be implemented.

149.7 Related LMIs

- https://en.wikibooks.org/wiki/LMIs_in_Control/Time-Delay_Systems/Continuous_Time/LMI_for_Stability_of_Retarded_Differential_Equation_with_Slowly-Varying_Delay - Stability LMI for continuous-time RDE with slowly-varying delay without uncertainty
- https://en.wikibooks.org/wiki/LMIs_in_Control/Stability_Analysis/Continuous_Time/Parametric,_Norm-Bounded_Uncertain_System_Quadratic_Stability - LMI for quadratic stability of continuous-time system with norm-bounded uncertainty
- https://en.wikibooks.org/w/index.php?title=LMIs_in_Control/Discrete_Time/Stability_Condition_for_Discrete-Time_TDS - Stability LMI for delayed discrete-time system

149.8 External Links

The presented results have been obtained from:

- Fridman E. 2014. Introduction to Time-Delay Systems, Analysis and Control. Springer. ISBN: 978-3-319-09392-5.

Additional information on LMI's in control theory can be obtained from the following resources:

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

149.9 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599>
3 <https://web.stanford.edu/~boyd/lmibook/>

150 Bounded Real Lemma under Slowly-Varying Delay

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This page describes a bounded real lemma for a continuous-time system with a time-varying delay. In particular, a condition is provided to obtain a bound on the L_2 -gain of a retarded differential system through feasibility of an LMI. The system under consideration pertains a single discrete delay, with the extent of the delay at any time bounded by some known value. This delay is only present in the state, with no direct delay in the effects of exogenous inputs on the state. In addition, the delay is assumed to vary only slowly in time, with a temporal derivative bounded by a value less than one, although results can also be attained if no bound is known. Solving the LMI for a particular value of the bound, while minimizing a scalar variable, an upper limit on the L_2 -gain of the system can be shown for any time-delay satisfying this bound.

150.1 The System

The system under consideration is one of the form:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + A_1x(t - \tau(t)) + B_0w(t) \quad t \geq t_0, \quad 0 \leq \tau(t) \leq h, \quad \dot{\tau}(t) \leq d < 1 \\ z(t) &= C_0x(t) + C_1x(t - \tau(t))\end{aligned}$$

In this description, A and A_1 are constant matrices in $\mathbb{R}^{n \times n}$. In addition, B_0 is a constant matrix in $\mathbb{R}^{n \times n_w}$, and C_0, C_1 are constant matrices in $\mathbb{R}^{n_z \times n}$ where $n_w, n_z \in \mathbb{N}$ denote the number of exogenous inputs and regulated outputs respectively. The variable $\tau(t)$ denotes a delay in the state at time $t \geq t_0$, assuming a value no greater than some $h \in \mathbb{R}_+$. Moreover, we assume that the function $\tau(t)$ is differentiable at any time, with the derivative bounded by some value $d < 1$, assuring the delay to be slowly-varying in time.

150.2 The Data

To obtain a bound on the L_2 -gain of the system, the following parameters must be known:

$$A, A_1 \in \mathbb{R}^{n \times n}$$

$$B_0 \in \mathbb{R}^{n \times n_w}$$

$$C_0, C_1 \in \mathbb{R}^{n_z \times n}$$

$$h \in \mathbb{R}_+$$

$$d \in [0, 1)$$

150.3 The Optimization Problem

Based on the provided data, we can obtain a bound on the L_2 -gain of the system by testing feasibility of an LMI. In particular, the bounded real lemma states that if the LMI presented below is feasible for some $\gamma > 0$, the L_2 -gain of the system is less than or equal to this γ . To attain a bound that is as small as possible, we minimize the value of γ while solving the LMI:

150.4 The LMI: L2-gain for TDS with Slowly-Varying Delay

Solve :

$$\min \gamma$$

such that there exist :

$$P, P_2, P_3, R, S, S_{12}, Q \in \mathbb{R}^{n \times n}$$

for which :

$$P > 0, \quad R > 0, \quad S > 0, \quad Q > 0$$

$$\left[\begin{array}{ccc|cc} \Phi_{11} & \Phi_{12} & S_{12} & R - S_{12} + P_2^T A_1 & P_2^T B_0 & C_0^T \\ * & \Phi_{22} & 0 & P_3^T A_1 & P_3^T B_0 & 0 \\ * & * & -S - R & R - S_{12}^T & 0 & 0 \\ * & * & * & -2R - S_{12} - S_{12}^T - (1-d)Q & 0 & C_1^T \\ * & * & * & * & -\gamma^2 I & 0 \\ * & * & * & * & * & -I \end{array} \right] < 0$$

where :

$$\Phi_{11} = A^T P_2 + P_2^T A + S + Q - R$$

$$\Phi_{12} = P - P_2^T + A^T P_3$$

$$\Phi_{22} = -P_3 - P_3^T + h^2 R$$

In this notation, the symbols * are used to indicate appropriate matrices to assure the overall matrix is symmetric.

150.5 Conclusion:

If the presented LMI is feasible for some γ , the system is internally stable, and will have an L_2 -gain less than γ . That is, independent of the values of the delays $\tau(t)$:

$$\|L_2 < \gamma\|_{L_2}$$

It should be noted that this result is conservative. That is, even when minimizing the value of γ , there is no guarantee that the bound obtained on the L_2 -gain is sharp.

In a scenario where no bound d on the change in the delay is known, the above LMI can still be used to obtain a bound on the L_2 -gain. In particular, setting $Q = 0$ in the above LMI, a bound can be attained independent of the value of the derivative of the delay.

150.6 Implementation

An example of the implementation of this LMI in Matlab is provided on the following site:

- https://github.com/djagt/LMI_Codes/blob/main/L2gain_cTDS.m

Note that this implementation requires packages for YALMIP with solver mosek, though a different solver can also be implemented.

150.7 Related LMIs

- [https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_\(Bounded_Real_Lemma\)](https://en.wikibooks.org/wiki/LMIs_in_Control/pages/KYP_Lemma_(Bounded_Real_Lemma)) - Bounded real lemma for continuous-time system without delay
- https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Discrete_Time_Bounded_Real_Lemma - Bounded real lemma for discrete-time system without delay
- https://en.wikibooks.org/wiki/LMIs_in_Control/Time-Delay_Systems/Continuous_Time/LMI_for_Stability_of_Retarded_Differential_Equation_with_Slowly-Varying_Delay - Stability LMI for continuous-time RDE with slowly-varying delay

150.8 External Links

The presented results have been obtained from:

- Fridman E. 2014. Introduction to Time-Delay Systems, Analysis and Control. Springer. ISBN: 978-3-319-09392-5.

Additional information on LMI's in control theory can be obtained from the following resources:

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

150.9 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm
2 <https://arxiv.org/abs/1903.08599/>
3 <https://web.stanford.edu/~boyd/lmibook/>

151 LMI for L₂-Optimal State-Feedback Control under Time-Varying Input Delay

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This page describes a method for constructing a full-state-feedback controller for a continuous-time system with a time-varying input delay. In particular, a condition is provided to obtain a bound on the L_2 -gain of closed-loop system under time-varying delay through feasibility of an LMI. The system under consideration pertains a single discrete delay in the actuator input, with the extent of the delay at any time bounded by some known value. Moreover, the delay is assumed to vary only slowly in time, with a temporal derivative bounded by a value less than one, although results may also be attained if no bound is known. Solving the LMI for a particular value of the bound, while minimizing a scalar variable, an upper limit on the L_2 -gain of the system can be shown for any time-delay satisfying this bound.

151.1 The System

The system under consideration is one of the form:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_2u(t - \tau(t)) + B_1w(t) \quad t \geq t_0, \quad 0 \leq \tau(t) \leq h, \quad \dot{\tau}(t) \leq d < 1 \\ z(t) &= C_1x(t) + D_{12}u(t - \tau(t))\end{aligned}$$

In this description, A and A_1 are constant matrices in $\mathbb{R}^{n \times n}$. In addition, B_1 is a constant matrix in $\mathbb{R}^{n \times n_w}$, and B_2 is a constant matrix in $\mathbb{R}^{n \times n_u}$, where $n_w, n_u \in \mathbb{N}$ denote the number of exogenous and actuator inputs respectively. Finally, C_1 and D_{12} are constant matrices in $\mathbb{R}^{n_z \times n}$ and $\mathbb{R}^{n_z \times n_u}$ respectively, where $n_z \in \mathbb{N}$ denotes the number of regulated outputs. The variable $\tau(t)$ denotes a delay in the actuator input at time $t \geq t_0$, assuming a value no greater than some $h \in \mathbb{R}_+$. Moreover, we assume that the function $\tau(t)$ is differentiable at any time, with the derivative bounded by some value $d < 1$, assuring the delay to be slowly-varying in time.

151.2 The Data

To construct an L_2 -optimal controller of the system, the following parameters must be known:

$$A \in \mathbb{R}^{n \times n}$$

$$B_1 \in \mathbb{R}^{n \times n_w}$$

$$B_2 \in \mathbb{R}^{n \times n_u}$$

$$C_1 \in \mathbb{R}^{n_z \times n}$$

$$D_{12} \in \mathbb{R}^{n_z \times n_u}$$

$$h \in \mathbb{R}_+$$

$$d \in [0, 1)$$

In addition to these parameters, a tuning scalar $\epsilon > 0$ is also implemented in the LMI.

151.3 The Optimization Problem

Based on the provided data, we can construct an L_2 -optimal full-state-feedback controller of the system by testing feasibility of an LMI. In particular, we note that if the LMI presented below is feasible for some $\gamma > 0$ and matrices $\bar{P}_2^{-1} > 0$ and Y , implementing the state-feedback $u(t) = Kx(t)$ with $K = Y\bar{P}_2^{-1}$, the L_2 -gain of the closed-loop system will be less than or equal to γ . To attain a bound that is as small as possible, we minimize the value of γ while solving the LMI:

151.4 The LMI: L2-Optimal Full-State-Feedback for TDS with Slowly-Varying Input Delay

Solve :

$$\min \gamma$$

such that there exist :

$$\bar{P}, \bar{P}_2, \bar{R}, \bar{S}, \bar{S}_{12}, \bar{Q} \in \mathbb{R}^{n \times n}, \quad Y \in \mathbb{R}^{n_u \times n}$$

for which :

$$\begin{aligned} \bar{P} &> 0, \quad \bar{P}_2 > 0, \quad \bar{R} > 0, \quad \bar{S} > 0 \\ \left[\begin{array}{ccc|c} \bar{\Phi}_{11} & \bar{\Phi}_{12} & \bar{S}_{12} & B_2 Y + \bar{R} - \bar{S}_{12} \\ * & \bar{\Phi}_{22} & 0 & \epsilon B_2 Y \\ * & * & -\bar{S} - \bar{R} & \bar{R} - \bar{S}_{12}^T \\ * & * & * & -(1-d)\bar{Q} - 2\bar{R} + \bar{S}_{12} + \bar{S}_{12}^T \\ \hline * & * & * & * \\ * & * & * & * \end{array} \right] &< 0 \\ \left[\begin{array}{cc} B_1 & \bar{P}_2^T C_1^T \\ \epsilon B_1 & 0 \\ 0 & 0 \\ 0 & Y^T D_{12}^T \\ \hline -\gamma^2 I & 0 \\ * & -I \end{array} \right] & \end{aligned}$$

where :

$$\Phi_{11} = A\bar{P}_2 + \bar{P}_2^T A^T + \bar{S} + \bar{Q} - \bar{R}$$

$$\Phi_{12} = \bar{P} - \bar{P}_2 + \epsilon \bar{P}_2^T A^T$$

$$\Phi_{22} = -\epsilon \bar{P}_2 - \epsilon \bar{P}_2^T + h^2 R$$

In this notation, the symbols $*$ are used to indicate appropriate matrices to assure the overall matrix is symmetric.

151.5 Conclusion:

If the presented LMI is feasible for some $\gamma, Y, \bar{P}_2 x(t)$, implementing the full-state-feedback controller $u(t) = Kx(t) = Y\bar{P}_2^{-1}$, the closed-loop system will be asymptotically stable, and will have an L_2 -gain less than γ . That is, independent of the values of the delays $\tau(t)$, the system:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_2 Kx(t - \tau(t)) + B_1 w(t) \\ z(t) &= C_1 x(t) + D_{12} Kx(t - \tau(t))\end{aligned}$$

with:

$$\|_{L_2} < \gamma \|_{L_2} K = Y\bar{P}_2^{-1}$$

will satisfy:

$$\|_{L_2} < \gamma \|_{L_2}$$

Here we note that $\bar{P}_2^{-1}x(t)$ is guaranteed to exist as P_2 is positive definite, and thus nonsingular.

It should be noted that the obtained result is conservative. That is, even when minimizing the value of γ , there is no guarantee that the bound obtained on the L_2 -gain is sharp, meaning that the actual L_2 -gain of the closed-loop can be (significantly) smaller than γ .

In a scenario where no bound d on the change in the delay is known, or this bound is greater than one, the above LMI may still be used to construct a controller. In particular, if the presented LMI is feasible with $\bar{Q} = 0$, the closed-loop system imposing $u(t) = Kx(t) = Y\bar{P}_2^{-1}$ will be internally exponentially stable with an L_2 -gain less than γ independent of the value of $\dot{\tau}(t)$.

151.6 Implementation

An example of the implementation of this LMI in Matlab is provided on the following site:

- https://github.com/djagt/LMI_Codes/blob/main/L2_OptStateFdbck_cTDS.m

Note that this implementation requires packages for YALMIP with solver mosek, though a different solver can also be implemented.

151.7 Related LMIs

- https://en.wikibooks.org/wiki/LMIs_in_Control/Time-Delay_Systems/Continuous_Time/Bounded_Real_Lemma_under_Slowly-Varying_Delay - Bounded real lemma for continuous-time system with slowly-varying delay
- https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Full-State_Feedback_Optimal_Control_Hinf_LMI - LMI for Hinf-optimal full-state-feedback control in a non-delayed continuous-time system
- https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Optimal_Output_Feedback_Hinf_LMI - LMI for Hinf-optimal output-feedback control in a non-delayed continuous-time system

151.8 External Links

The presented results have been obtained from:

- Fridman E. 2014. Introduction to Time-Delay Systems, Analysis and Control. Springer. ISBN: 978-3-319-09392-5.

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- LMI Properties and Applications in Systems, Stability, and Control Theory² - A List of LMIs by Ryan Caverly and James Forbes.
- LMIs in Systems and Control Theory³ - A downloadable book on LMIs by Stephen Boyd.

151.9 Return to Main Page:

1 http://control.asu.edu/MAE598_frame.htm

2 <https://arxiv.org/abs/1903.08599/>

3 <https://web.stanford.edu/~boyd/lmibook/>

152 Discrete Time

1. REDIRECT LMIs in Control/Discrete Time/Stability Condition for Discrete-Time TDS¹

¹ <https://en.wikibooks.org/wiki/LMIs%20in%20Control%2FDiscrete%20Time%2FStability%20Condition%20for%20Discrete-Time%20TDS>

153 LMI for Attitude Control of Nonrotating Missiles, Pitch Channel

LMI for Attitude Control of Nonrotating Missiles, Pitch Channel

The dynamic model of a missile is very complicated and a simplified model is used. To do so, we consider a simplified attitude system model for the pitch channel in the system. We aim to achieve a non-rotating motion of missiles. It is worthwhile to note that the attitude control design for the pitch channel and the yaw/roll channel can be solved exactly in the same way while representing matrices of the system are different.

153.1 The System

The state-space representation for the pitch channel can be written as follows:

$$\begin{aligned}\dot{x}(t) &= A(t)x(t) + B_1(t)u(t) + B_2(t)d(t) \\ y(t) &= C(t)x(t) + D_1(t)u(t) + D_2(t)d(t)\end{aligned}$$

where $x = [\alpha \quad w_z \quad \delta_z]^T$, $u = \delta_{zc}$, $y = [\alpha \quad n_y]^T$, and $d = [\beta \quad w_y]^T$ are the state variable, control input, output, and disturbance vectors, respectively. The parameters α , w_z , δ_z , δ_{zc} , n_y , β , and w_y stand for the attack angle, pitch angular velocity, the elevator deflection, the input actuator deflection, the overload on the side direction, the sideslip angle, and the yaw angular velocity, respectively.

153.2 The Data

In the aforementioned pitch channel system, the matrices $A(t)$, $B_1(t)$, $B_2(t)$, $C(t)$, $D_1(t)$, and $D_2(t)$ are given as:

$$\begin{aligned}A(t) &= \begin{bmatrix} -a_4(t) & 1 & -a_5(t) \\ -\dot{\alpha}_1(t)a_4(t) - a_2(t) & \dot{\alpha}_1(t) - a_1(t) & \dot{\alpha}_1(t)a_5(t) - a_3(t) \\ 0 & 0 & -1/\tau_z \end{bmatrix} \\ B_1(t) &= \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \quad B_2(t) = \frac{w_x}{57.3} \begin{bmatrix} -1 & 0 \\ -\dot{\alpha}_1(t) & \frac{J_x - J_y}{J_z} \\ 0 & 0 \end{bmatrix} \\ C(t) &= \frac{w_x}{57.3} \begin{bmatrix} 57.3g & 0 & 0 \\ V(t)a_4(t) & 0 & V(t)a_5(t) \end{bmatrix}\end{aligned}$$

$$D_1(t) = 0, \quad D_2(t) = \frac{1}{57.3g} \begin{bmatrix} 0 & 0 \\ V(t)b_7(t) & 0 \end{bmatrix}$$

where $a_1(t) \sim a_6(t)$, $b_1(t) \sim b_7(t)$, $\dot{a}_1(t)$, $\dot{b}_1(t)$ and $c_1(t) \sim c_4(t)$ are the system parameters. Moreover, V is the speed of the missile and J_x , J_y , and J_z are the rotary inertia of the missile corresponding to the body coordinates.

153.3 The Optimization Problem

The optimization problem is to find a state feedback control law $u = Kx$ such that:

1. The closed-loop system:

$$\dot{x} = (A + B_1 K)x + B_2 d$$

$$z = (C + D_1 K)x + D_2 d$$

is stable.

2. The H_∞ norm of the transfer function:

$$G_{zd}(s) = (C + D_1 K)(sI - (A + B_1 K))^{-1} B_2 + D_2$$

is less than a positive scalar value, γ . Thus:

$$\|G_{zd}(s)\|_\infty < \gamma$$

153.4 The LMI: LMI for non-rotating missile attitude control

Using Theorem 8.1 in [1], the problem can be equivalently expressed in the following form:

$$\min \gamma$$

$$\text{s.t. } X > 0$$

$$\begin{bmatrix} (AX + B_1 W)^T + AX + B_1 W & B_2 & (CX + D_1 W)^T \\ B_2^T & -\gamma I & D_2^T \\ CX + D_1 W & D_2 & -\gamma I \end{bmatrix} < 0$$

153.5 Conclusion:

As mentioned, the aim is to attenuate the disturbance on the performance of the missile. The parameter γ is the disturbance attenuation level. When the matrices W and X are determined in the optimization problem, the controller gain matrix can be computed by:

$$K = WX^{-1}$$

153.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Non-rotating-Missle-Attitude-Control>

153.7 Related LMIs

LMI for Attitude Control of Nonrotating Missles, Yaw/Roll Channel¹

153.8 External Links

- <https://www.crcpress.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMI in Control Systems Analysis, Design and Applications

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LMIs in Control/Tools²

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Nonrotating_Missles_Attitude_Control_yaw_roll_channel

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

154 LMI for Attitude Control of Nonrotating Missiles, Yaw/Roll Channel

LMI for Attitude Control of Nonrotating Missiles, Yaw/Roll Channel

Deriving the exact dynamic modeling of a missile is a very complicated procedure. Thus, a simplified model is used to model the missile dynamics. To do so, we consider a simplified attitude system model for the yaw/roll channel of the system. We aim to achieve a non-rotating motion of missiles. Note that the attitude control design for the yaw/roll channel and the pitch channel can be solved exactly in the same way except for different representing matrices of the system.

154.1 The System

The state-space representation for the yaw/roll channel can be written as follows:

$$\begin{aligned}\dot{x}(t) &= A(t)x(t) + B_1(t)u(t) + B_2(t)d(t) \\ y(t) &= C(t)x(t) + D_1(t)u(t) + D_2(t)d(t)\end{aligned}$$

where $x = [\beta \quad w_y \quad w_x \quad \delta_x \quad \delta_y]^T$, $u = [\delta_{xc} \quad \delta_{yc}]^T$, $y = [n_z \quad w_x]^T$, and $d = \delta_z$ are the state variable, control input, output, and disturbance vectors, respectively. The parameters α , w_z , δ_z , δ_{zc} , n_y , β , and w_y stand for the attack angle, pitch angular velocity, the elevator deflection, the input actuator deflection, the overload on the side direction, the sideslip angle, and the yaw angular velocity, respectively.

154.2 The Data

In the aforementioned yaw/roll channel system, the matrices $A(t)$, $B_1(t)$, $B_2(t)$, $C(t)$, $D_1(t)$, and $D_2(t)$ are given as:

$$A(t) = \begin{bmatrix} A_{11}(t) & A_{12}(t) \\ 0 & A_{22}(t) \end{bmatrix}$$

where

$$A_{11}(t) = \begin{bmatrix} -b_4(t) & 1 & \frac{\alpha(t)}{57.3} \\ -\dot{b}_1(t)b_4(t) - b_2(t) & -\dot{b}_1(t) - \dot{b}_1(t) & \frac{J_y - J_z}{57.3J_x}w_z(t) - \frac{\dot{b}_1\alpha(t)}{57.3} \\ c_2(t) & \frac{J_y - J_z}{57.3J_x}w_z(t) & -c_1(t) \end{bmatrix}$$

$$A_{12}(t) = \begin{bmatrix} 0 & -b_5(t) \\ 0 & -\dot{b}_1(t)\dot{b}_5 - b_3(t) \\ -c_3(t) & c_4(t) \end{bmatrix}$$

$$A_{22}(t) = -\frac{1}{\tau_x \tau_y} \begin{bmatrix} \tau_y & 0 \\ 0 & \tau_x \end{bmatrix}$$

and

$$B_1(t) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ \frac{1}{\tau_x} & 0 \\ 0 & \frac{1}{\tau_y} \end{bmatrix}, \quad B_2(t) = \begin{bmatrix} a_6(t) \\ -\dot{b}_1(t)a_6(t) \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$C(t) = -\frac{1}{57.3g} \begin{bmatrix} V(t)b_4(t) & 0 & 0 & 0 & V(t)b_5(t) \\ 0 & 0 & -57.3g & 0 & 0 \end{bmatrix}$$

$$D_1(t) = 0, \quad D_2(t) = -\frac{V(t)}{57.3g} \begin{bmatrix} b_6(t) \\ 0 \end{bmatrix}$$

where $a_1(t) \sim a_6(t)$, $b_1(t) \sim b_7(t)$, $\dot{a}_1(t)$, $\dot{b}_1(t)$ and $c_1(t) \sim c_4(t)$ are the system parameters. Moreover, V is the speed of the missile and J_x , J_y , and J_z are the rotary inertia of the missile corresponding to the body coordinates.

154.3 The Optimization Problem

The optimization problem is to find a state feedback control law $u = Kx$ such that:

1. The closed-loop system:

$$\dot{x} = (A + B_1 K)x + B_2 d$$

$$z = (C + D_1 K)x + D_2 d$$

is stable.

2. The H_∞ norm of the transfer function:

$$G_{zd}(s) = (C + D_1 K)(sI - (A + B_1 K))^{-1} B_2 + D_2$$

is less than a positive scalar value, γ . Thus:

$$\|G_{zd}(s)\|_\infty < \gamma$$

154.4 The LMI: LMI for non-rotating missile attitude control

Using Theorem 8.1 in [1], the problem can be equivalently expressed in the following form:

$$\begin{aligned} \min \quad & \gamma \\ \text{s.t.} \quad & X > 0 \\ & \begin{bmatrix} (AX + B_1W)^T + AX + B_1W & B_2 & (CX + D_1W)^T \\ B_2^T & -\gamma I & D_2^T \\ CX + D_1W & D_2 & -\gamma I \end{bmatrix} < 0 \end{aligned}$$

154.5 Conclusion:

As mentioned, the aim is to attenuate the disturbance on the performance of the missile. The parameter γ is the disturbance attenuation level. When the matrices W and X are determined in the optimization problem, the controller gain matrix can be computed by:

$$K = WX^{-1}$$

154.6 Implementation

A link to Matlab codes for this problem in the Github repository:

<https://github.com/asalimil/LMI-for-Attitude-Control-Nonrotating-Missle-Yaw-Roll-Channel>

154.7 Related LMIs

LMI for Attitude Control of Nonrotating Missles, Pitch Channel¹

154.8 External Links

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LMIs in Control/Tools²

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/LMI_for_Attitude_Control_of_Nonrotating_Missles

2 https://en.wikibooks.org/wiki/LMIs_in_Control/Tools

155 LMI for H₂/H_∞ Polytopic Controller for Robot Arm on a Quadrotor

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155.1 The System:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1w(t) + B_2u(t) \\ z(t) &= C_1x(t) + D_{11}w(t) + D_{12}u(t) \\ y(t) &= C_2x(t) + D_{21}w(t) + D_{22}u(t) \\ \dot{x}_K(t) &= A_Kx_K(t) + B_Ky(t) \\ u(t) &= C_Kx_K(t) + D_Ky(t)\end{aligned}$$

155.2 The Optimization Problem:

Given a state space system of

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B_1w(t) + B_2u(t) \\ z(t) &= C_1x(t) + D_{11}w(t) + D_{12}u(t) \\ y(t) &= C_2x(t) + D_{21}w(t) + D_{22}u(t) \\ \dot{x}_K(t) &= A_Kx_K(t) + B_Ky(t) \\ u(t) &= C_Kx_K(t) + D_Ky(t)\end{aligned}$$

where A_K , B_K , C_K , and D_K , form the K matrix as defined in below. This, therefore, means that the Regulator system can be re-written as:

$$\left[\begin{array}{c} \dot{x}(t) \\ z_1(t) \\ z_2(t) \\ y(t) \end{array} \right] = \left[\begin{array}{c|cc|c} A & B & 0 & B \\ C & D & 0 & D \\ 0 & 0 & 0 & I \\ C & D & I & D \end{array} \right] \left[\begin{array}{c} \dot{x}(t) \\ w_1(t) \\ w_2(t) \\ u(t) \end{array} \right]$$

With the above 9-matrix representation in mind, the we can now derive the controller needed for solving the problem, which in turn will be accomplished through the use of LMI's. Firstly, we will be taking our H_2/H_∞ state-feedback control and

make some modifications to it. More specifically, since the focus is modeling for worst-case scenario of a given parameter, we will be modifying the LMI's such that the mixed H_2/H_∞ controller is polytopic.

155.3 The LMI:

H_2/H_∞ Polytopic Controller for Quadrotor with Robotic Arm.

Recall that from the 9-matrix framework , $w_1(t)$ and $w_2(t)$ represent our process and sensor noises respectively and $u(t)$ represents our input channel. Suppose we were interested in modeling noise across all three of these channels. Then the best way to model uncertainty across all three cases would be modifying the D matrix to D_i , where ($i = 1, \dots, k$ parameters, $D_i = nI$, and n is a constant noise value). This, in turn results in our D_{11} - D_{22} matrices to be modified to $D_{11,i}$ - $D_{22,i}$.

Using the LMI's given for optimal H_2/H_∞ -optimal state-feedback controller from Peet Lecture 11 as reference, our resulting polytopic LMI becomes:

$$\begin{aligned} & \min_{\gamma_1, \gamma_2, X_1, Y_1, Z, A_n, B_n, C_n, D_n} \gamma_1^2 + \gamma_2^2 \\ & \begin{bmatrix} AA_i & AB_i^T & AC_i^T \\ AB_i & BB_i & BC_i^T \\ AC_i & BC_i & -I \end{bmatrix} < 0 \\ & \begin{bmatrix} AA_i & AB_i^T & AC_i^T & AD_i^T \\ AB_i & BB_i & BC_i^T & BD_i^T \\ AC_i & BC_i & -I & CD_i^T \\ AD_i & BD_i & CD_i & -\gamma_2^2 I \end{bmatrix} < 0 \\ & \begin{bmatrix} Y_1 & I & AD_i^T \\ I & X_1 & BD_i^T \\ AD_i & BD_i & Z \end{bmatrix} > 0 \end{aligned}$$

CD=0

$$\text{trace}(Z) < \gamma_1^2$$

where $i=1,\dots,k$, $\|S(K, P)\|_{H_2} < \gamma_1$ and $\|S(K, P)\|_{H_\infty} < \gamma_2$ and:

$$\begin{aligned} AA_i &= AY_1 + Y_1A^T + B_2C_n + C_n^TB_2^T \\ AB_i &= A^T + A_n + [B_2D_nC_2]^T \\ AC_i &= [B_1 + B_2D_nD_{21,i}]^T \\ AD_i &= C_1Y_1 + D_{12,i}C_n \\ BB_i &= X_1A + A^TX_1 + B_nC_2 + C_2^TB_n^T \\ BC_i &= [X_1B_1 + B_nD_{21,i}]^T \\ BD_i &= C_1 + D_{12,i}D_nC_2 \\ CD_i &= D_{11,i} + D_{12,i}D_nD_{21,i} \end{aligned}$$

After solving for both the optimal H_2 and H_∞ gain ratios as well as $X_1, Y_1, Z, A_n, B_n, C_n, D_n$, we can then construct our worst-case scenario controller by setting our D matrix (and consequently our $D_{11}, D_{12}, D_{21}, D_{22}$ matrices) to the highest n value. This results in the controller:

$$K = \left[\begin{array}{c|c} A_K & B_K \\ \hline C_K & D_K \end{array} \right]$$

which is constructed by setting:

$$\begin{aligned} D_K &= (I + D_{K_2} D_{22})^{-1} D_{K_2} \\ B_K &= B_{K_2} (I + D_{22} D_K) \\ C_K &= (I - D_K D_{22}) C_{K_2} \\ A_K &= A_{K_2} - B_K (I - D_{22} D_K)^{-1} D_{22} C_K \end{aligned}$$

where:

$$\begin{aligned} X_2 Y_2^T &= I - X_1 Y_1 \\ \left[\begin{array}{c|c} A_{K_2} & B_{K_2} \\ \hline C_{K_2} & D_{K_2} \end{array} \right] &= \left[\begin{array}{cc} X_2 & X_1 B_2 \\ 0 & I \end{array} \right]^{-1} \left[\begin{array}{cc} A_n - X_1 A Y_1 & B_n \\ C_n & D_n \end{array} \right] \left[\begin{array}{cc} Y_2^T & 0 \\ C_2 Y_1 & I \end{array} \right] \end{aligned}$$

155.4 Conclusion:

The LMI is feasible and the resulting controller is found to be stable under normal noise disturbances for all states.

155.5 Implementation

155.6 References

1. An LMI-Based Approach for Altitude and Attitude Mixed H2/Hinf-Polytopic Regulator Control of a Quadrotor Manipulator by Aditya Ramani and Sudhanshu Katareay.

156 An LMI for the Kalman Filter

UNKNOWN TEMPLATE FULLPAGENAME

This is a **An LMI for the Kalman Filter**. The Kalman Filter is one of the most widely used state-estimation techniques. It has applications in multiple aspects of navigation (inertial, terrain-aided, stellar.)

156.1 The System

Continuous Time:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + w(t), \\ y(t) &= Cx(t) + v(t)\end{aligned}$$

The process and sensor noises are given by $w(t)$ and $v(t)$ respectively.

Discrete Time:

$$\begin{aligned}x_{k+1} &= Ax_k + w_k \\ y_k &= Cx_k + v_k\end{aligned}$$

The process and sensor noises are given by w_k and v_k respectively.

156.2 The Data

The data required for the Kalman Filter include a model of the system that the states are trying to be output and a measurement that is the output of the system dynamics being estimated.

156.3 The Filter

The Filter and Estimator equations can be written as:

Continuous Time

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L(y(t) - \hat{y}(t))$$

Discrete Time

$$\hat{x}_{k+1} = A\hat{x}_k + L(C\hat{x}_k - y_k)$$

156.4 The Error

The error dynamics evolve according to the following expression

Continuous Time

$$\dot{e}(t) = (A + LC)e(t) + w(t) + Lv(t)$$

Discrete Time

$$e_{k+1} = (A + LC)e_k + w_k + Lv_k$$

156.5 The Optimization Problem

The Kalman Filtering (or LQE) problem is a Dual to the LQR problem. Replace the matrices (A, B, Q, R, K) from LQR with $(A^T, C^T, V_1, V_2, L^T)$.

The Kalman Filter chooses L to minimize the cost $J = E[e^T e]$. This cost can be thought of as the covariance of the state error between the actual and estimated state. When the state error covariance is low the filter has converged and the estimate is good.

The Luenberger or Kalman gain can be computed from $L = \Sigma C^T V_2^{-1}$

The process and measurement noise covariances for the Kalman filter are given by

$$V_1 = E[w(t)w(t)^T] \text{ aka: } Q$$

$$V_2 = E[v(t)v(t)^T] \text{ aka: } R$$

The matrix Σ satisfies the following equality

$$A\Sigma + \Sigma A^T + V_1 = \Sigma C^T V_2^{-1} C \Sigma$$

We also cover the discrete Kalman Filter formulation which is more useful for real-life computer implementations.

The discrete Kalman filter chooses the gain $L = A\Sigma C^T (C\Sigma C^T + V)^{-1}$ where the PSDs of the process and sensor noises are given by

$$W = E[w_k w_k^T] \text{ aka: } Q$$

$$V = E[v_k v_k^T] \text{ aka: } R$$

The steady-state covariance of the error in the estimated state is given by Σ and satisfies the following Riccati equation.

$$\Sigma = A\Sigma A^T + W - A\Sigma C^T(C\Sigma C^T + V)^{-1}C\Sigma A^T$$

- Objective: State Estimate Error Covariance
- Variables: Observer Gains
- Constraints: Dynamics of System to be Estimated

156.6 The LMI: H2-Optimal Control Full-State Feedback to LQR to Kalman Filter

The Kalman Filter is a dual to the LQR problem which has been shown to be equivalent to a special case of H2-static state feedback.

Start with the H2-Optimal Control Full-State Feedback.

The following are equivalent

$$1. ||S(K, P)||_{H2} < \gamma$$

$$2. K = ZX^{-1} \text{ for some } Z \text{ and } X > 0 \text{ where}$$

$$\begin{aligned} \begin{bmatrix} X \\ Z \end{bmatrix} + [X \ Z^T] \begin{bmatrix} A^T \\ B_2^T \end{bmatrix} + B_1 B_1^T &< 0 \\ \begin{bmatrix} X & (C_1 X + D_{12} Z)^T \\ C_1 X + D_{12} Z & W \end{bmatrix} &> 0 \\ \text{Trace}W &< \gamma \end{aligned}$$

To solve the LQR problem using H2 optimal state-feedback control the following variable substitutions are required.

$$\begin{aligned} C_1 &= \begin{bmatrix} Q^{1/2} \\ 0 \end{bmatrix}, D_{12} = \begin{bmatrix} R^{1/2} \\ 0 \end{bmatrix}, D_{11} = 0, \\ B_2 &= B, B_1 = I \end{aligned}$$

Then

$$S(P, K) = \begin{bmatrix} A + B_2 K & B_1 \\ C_1 + D_{12} K & D_{11} \end{bmatrix} = \begin{bmatrix} A + B K & I \\ Q^{1/2} & 0 \\ K & 0 \end{bmatrix}$$

This results in the following LMI.

$$\begin{aligned} & \begin{bmatrix} X \\ Z \end{bmatrix} + [X \ Z^T] \begin{bmatrix} A^T \\ B^T \end{bmatrix} + I < 0 \\ & \begin{bmatrix} X & (\begin{bmatrix} Q^{1/2} \\ 0 \end{bmatrix} X + \begin{bmatrix} R^{1/2} \\ 0 \end{bmatrix} Z)^T \\ \begin{bmatrix} Q^{1/2} \\ 0 \end{bmatrix} + \begin{bmatrix} R^{1/2} \\ 0 \end{bmatrix} Z & W \end{bmatrix} > 0 \\ & \text{Trace}W < \gamma \end{aligned}$$

To solve the Kalman Filtering problem using the LQR LMI replace A, B, Q, R, K with A^T, C^T, V_1, V_2 , and L^T . This results in the following LMI.

$$\begin{aligned} & \begin{bmatrix} X \\ Z \end{bmatrix} + [X \ Z^T] \begin{bmatrix} A \\ C \end{bmatrix} + I < 0 \\ & \begin{bmatrix} X & (\begin{bmatrix} V_1^{1/2} \\ 0 \end{bmatrix} X + \begin{bmatrix} V_2^{1/2} \\ 0 \end{bmatrix} Z)^T \\ \begin{bmatrix} V_1^{1/2} \\ 0 \end{bmatrix} + \begin{bmatrix} V_2^{1/2} \\ 0 \end{bmatrix} Z & W \end{bmatrix} > 0 \\ & \text{Trace}W < \gamma \end{aligned}$$

The discrete-time Kalman Filtering LMI is saved for another page as it requires derivation of the Discrete-Time LQR LMI problem which was not covered in class.

156.7 Conclusion:

The LMI for the Kalman Filter allows us to calculate the optimal gain for state estimation. It is shown that it can be found as a special case of the H2-optimal state feedback with the appropriate substitution of matrices. The LMI gives us a different way of computing the optimal Kalman gain.

156.8 Implementation

A link to CodeOcean or other online implementation of the LMI

156.9 Related LMIs

Links to other closely-related LMIs

- H2OptimalObserver¹

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/H2OptimalObserver

156.10 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.
- LMI Properties and Applications in Systems, Stability, and Control Theory³ - A List of LMIs by Ryan Caverly and James Forbes
- LMIs in Systems and Control Theory⁴ - A downloadable book on LMIs by Stephen Boyd
- <http://users.isr.ist.utl.pt/~pjcro/temp/Applied%20Optimal%20Estimation%20-%20Gelb.pdf> - Applied Optimal Estimation, Arthur Gelb (Kalman Filtering and Optimal Estimation Classic)
- <https://us.artechhouse.com/All-Source-Positioning-Navigation-and-Timing-P2082.aspx> - All Source Position, Navigation and Timing. New textbook that has a good development and description of the Kalman Filter and some of its very practical uses and implementations by Boeing Senior Technical Fellow Ken Li.

156.11 Return to Main Page:

2 http://control.asu.edu/MAE598_frame.htm
3 <https://arxiv.org/abs/1903.08599>
4 <https://web.stanford.edu/~boyd/lmibook/>

157 Hinf Optimal Model Reduction

Given a full order model and an initial estimate of a reduced order model it is possible to obtain a reduced order model optimal in H_∞ sense. This methods uses LMI techniques iteratively to obtain the result.

157.1 The System

Given a state-space representation of a system $G(s)$ and an initial estimate of reduced order model $\hat{G}(s)$.

$$G(s) = C(sI - A)B + D,$$
$$\hat{G}(s) = \hat{C}(sI - \hat{A})\hat{B} + \hat{D},$$

Where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $C \in \mathbb{R}^{p \times n}$, $D \in \mathbb{R}^{p \times m}$, $\hat{A} \in \mathbb{R}^{k \times k}$, $\hat{B} \in \mathbb{R}^{k \times m}$, $\hat{C} \in \mathbb{R}^{p \times k}$ and $\hat{D} \in \mathbb{R}^{p \times m}$. Where n, k, m, p are full order, reduced order, number of inputs and number of outputs respectively.

157.2 The Data

The full order state matrices A, B, C, D and the reduced model order k .

157.3 The Optimization Problem

The objective of the optimization is to reduce the H_∞ norm distance of the two systems. Minimizing $-\hat{G}\|_\infty$ with respect to \hat{G} .

157.4 The LMI: The Lyapunov Inequality

Objective: $\min \gamma$.

Subject to:: $P = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} > 0,$

$$\begin{bmatrix} A^T P_{11} + P_{11}A & A^T P_{12} + P_{12}\hat{A} & P_{11}B - P_{12}\hat{B} & C^T \\ \hat{A}^T P_{12}^T + P_{12}^T A & \hat{A}^T P_{22} + P_{22}\hat{A} & P_{12}^T B - P_{22}\hat{B} & \hat{C}^T \\ B^T P_{11} - \hat{B}^T P_{12}^T & B^T P_{12} - \hat{B}^T P_{22} & -\gamma I & D^T - \hat{D}^T \\ C & \hat{C} & D - \hat{D} & -\gamma I \end{bmatrix} > 0$$

It can be seen from the above LMI that the second matrix inequality is not linear in $\hat{A}, \hat{B}, \hat{C}, \hat{D}, P$. But making \hat{A}, \hat{B} constant it is linear in \hat{C}, \hat{D}, P . And if P_{12}, P_{22} are constant it is linear in $\hat{A}, \hat{B}, \hat{C}, \hat{D}, P_{11}$. Hence the following iterative algorithm can be used.

- (a) Start with initial estimate \hat{G} obtained from techniques like Hankel-norm reduction/Balanced truncation.
- (b) Fix \hat{A}, \hat{B} and optimize with respect to \hat{C}, \hat{D}, P .
- (c) Fix P_{12}, P_{22} and optimize with respect to $\hat{A}, \hat{B}, \hat{C}, \hat{D}, P_{11}$.
- (d) Repeat steps (b) and (c) until the solution converges.

157.5 Conclusion:

The LMI techniques results in model reduction close to the theoretical limits set by the largest removed hankel singular value. The improvements are often not significant to that of Hankel-norm reduction. Due to high computational load it is recommended to only use this algorithm if optimal performance becomes a necessity.

157.6 External Links

A list of references documenting and validating the LMI.

- Model order Reduction using LMIs - A conference paper by Helmersson, Anders, Proceedings of the 33rd IEEE Conference on Decision and Control, 1994, p. 3217-3222 vol.4

157.7 Return to Main Page:

158 An LMI for Multi-Robot Systems

An LMI for Multi-Robot Systems

1. /Consensus for Multi-Agent Systems/¹

¹ <https://en.wikibooks.org/wiki/%2FConsensus%20for%20Multi-Agent%20Systems%2F>

159 Helicopter Inner Loop LMI

UNKNOWN TEMPLATE FULLPAGENAME

This is a **Helicopter Inner Loop LMI**. Optimization methods and optimal control have had difficulty gaining traction in the rotorcraft control law community. However, this LMI derived in the referenced paper attempts to address the issues with a LMI for Robust, Optimal Control.

159.1 The System

Continuous Time:

$$\begin{aligned}\dot{x}(t) &= \hat{A}x(t) + B_1w(t) + \hat{B}_2u(t), \\ z(t) &= C_1x(t) + D_{12}u(t)\end{aligned}$$

The Helicopter model is given by knowledge of the stability and control derivatives which populate the elements of the \hat{A}, \hat{B}_2 matrices in the dynamic equations above.

The state vector is given by the typical elements of a rigid 6-DOF body model. $x = [u, w, q, \theta, v, p, \psi, r]^T$. The input vector is given by $u = [\delta_0, \delta_{1s}, \delta_{1c}, \delta_T]^T$ which pertain to the main rotor collective, longitudinal/lateral cyclic and tail rotor collective blade angles in radians.

The gust disturbance is denoted by $w(t)$ and is assumed to be random in nature. The stability and control derivative matrices are modeled with uncertainty as follows:

$$\hat{A} = A + \Delta A, \hat{B}_2 = B_2 + \Delta B$$

The Δ terms represent the uncertainties in the helicopter system model.

159.2 The Data

The Data required for this LMI are the stability and control derivatives that populate the A and B-matrices of the system above which can be obtained from linearizing non-linear models. It can also be obtained from experimental methods such as step responses and swept sines (System Identification.)

159.3 The Control Architecture

A control architecture for the inner loop of the helicopter model mentioned above is designed using a state feedback control law.

$$u = Kx(t)$$

The objective for the inner loop control is to design a full state feedback law such that the closed-loop helicopter system satisfies the following 3 performance specifications.

159.4 The Optimization Problem

Objective 1: The closed-loop system is internally stable for any admissible uncertainty.

Objective 2: Poles of the close-loop system lie within the disk $D(-q, r)$ with center $-q + jo$ and radius $r, q > r > 0$, for any admissible uncertainty.

Objective 3: Given gust disturbance suppression index γ , for any admissible uncertainty, the effect of the gust disturbance to selected flight states and control input is in the given level, i.e.

$$\int_0^\infty \{x^T(t)Qx(t) + u^T(t)Ru(t)\} dt < \gamma^2 \int_0^\infty w(t)^T w(t) dt.$$

where $w(t) \in L_2(0, \infty)$. Q and R are weighting matrices with appropriate dimensions and $Q = Q^T \geq 0, R = R^T > 0$.

It can be shown that the inner loop performance specifications listed in Objectives 1-3 can be met with a state feedback control law if the LMI described in the following section is true.

- Objective: Objectives listed above
- Variables: Controller Gains
- Constraints: Rotorcraft Dynamics and Modeled Actuator Limits

159.5 The LMI: H-Inf Inner Loop D-Stabilization Optimization

The paper derives and LMI of the form below and asserts that the if there exists a constant ϵ , matrix Z with appropriate dimensions and a symmetric positive matrix P , such that

$$\begin{bmatrix} \Psi_{11} & B_1 & XC_1^T + Z^T D_1 2^T & A_\alpha X + B_2 Z & XF_1^T + Z^T F_2^T & \epsilon H \\ B_1^T & -\gamma^2 I & 0 & 0 & 0 & 0 \\ C_1 X + D_{12} Z & 0 & -I & 0 & 0 & 0 \\ X A_\alpha^T + Z^T B_2^T & 0 & 0 & -rX & XF_1^T + Z^T F_2^T & 0 \\ F_1 X + F_2 Z & 0 & 0 & F_1 X + F_2 Z & -\epsilon I & 0 \\ \epsilon H^T & 0 & 0 & 0 & 0 & -\epsilon I \end{bmatrix} < 0$$

where, $\Psi_{11} = A_\alpha X + X A_\alpha^T + B_2 Z + Z^T B_2^T$, $A_\alpha = A + \alpha I$

This LMI is shown to satisfy Objectives 1, 2,3, and the control law is given by

$$\begin{aligned} u(t) &= Kx(t) \\ K &= ZX^{-1} \end{aligned}$$

159.6 Conclusion:

The LMI for Helicopter Inner Loop Control design provides an optimization-based approach towards achieving Level 1 Handling Qualities per ADS-33E. This is an interesting way to approach a very difficult problem that has usually been approached through classical control methods and with extensive piloted simulation and flight test.

159.7 Implementation

A link to CodeOcean or other online implementation of the LMI

159.8 Related LMIs

Links to other closely-related LMIs

- Optimal_Output_Feedback_Hinf_LMI¹

159.9 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control² - A course on LMIs in Control by Matthew Peet.

¹ https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Optimal_Output_Feedback_Hinf_LMI
² http://control.asu.edu/MAE598_frame.htm

- <https://scialert.net/fulltext/?doi=srj.2008.39.52> - Multi-Mode Flight Control for an Unmanned Helicopter Based on Robust H ∞ D-Stabilization and PI Tracking Configuration

159.10 Return to Main Page:

160 Hinf LMI Satellite Attitude Control

UNKNOWN TEMPLATE FULLPAGENAME

This is a H_∞ LMI for Satellite Attitude Control. Satellite attitude control is necessary to allow satellites in orbit accomplish their mission. Poor satellite attitude control results in poor pointing performance which can result in increased cost, delayed service, and reduced lifetime of the satellite.

160.1 The System

The full derivation of the system from first principles is accomplished in the companion LMI for H_2 Satellite Attitude Control. The link to that page is at the bottom with the references.

Continuous Time:

$$\begin{aligned} I_x \ddot{\phi} + 4(I_y - I_z)\omega_0^2 \phi + (I_y - I_z - I_z)\omega_0 \dot{\psi} &= T_{cx} + T_{dx} \\ I_y \ddot{\theta} + 3(I_x - I_z)\omega_0^2 \theta &= T_{cy} + T_{dy} \\ I_z \ddot{\psi} + (I_x + I_z - I_y)\omega_0 \dot{\psi} &= T_{cz} + T_{dz} \end{aligned}$$

The above model was derived by substituting satellite attitude kinematics into the attitude dynamics of a satellite. The following are definitions of the variables above:

- Moments of inertia about the corresponding axis: I_x, I_y, I_z
- Euler Angles: ϕ, θ, ψ
- Disturbance Torques (flywheel, gravitational, and disturbance): T_c, T_g, T_d
- Rotational-angular velocity of the Earth: $\omega_0 = 7.292115 \times 10^{-5}$ [rad/s]

The state-space representation of the system can be found by the following steps.
Let

$$x = \begin{bmatrix} q & \dot{q} \end{bmatrix}^T, \quad z_\infty = 10^{-3} M \ddot{q}, z_2 = q,$$

Introduce the notations

$$I_{\alpha\beta} = I_\alpha - I_\beta, \quad I_{\alpha\beta\gamma} = I_\alpha - I_\beta - I_\gamma$$

where α, β, γ stand for any element in x, y, z . Then the state-space system is:

$$\begin{aligned}\dot{x} &= Ax + B_1 u + B_2 d \\ z_\infty &= C_1 x + D_1 u + D_2 d \\ z_2 &= C_2 x\end{aligned}$$

where the matrices in the above state-space representation are defined as follows:

$$A = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ \frac{-4\omega_0^2 I_{yz}}{I_x} & 0 & 0 & 0 & 0 & \frac{-\omega_0 I_{yzz}}{I_x} \\ 0 & \frac{-3\omega_0^2 I_{xz}}{I_y} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{-\omega_0^2 I_{yx}}{I_z} & \frac{\omega_0 I_{yzz}}{I_x} & 0 & 0 \end{bmatrix}$$

$$B_1 = B_2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ \frac{1}{I_x} & 0 & 0 \\ 0 & \frac{1}{I_y} & 0 \\ 0 & 0 & \frac{1}{I_z} \end{bmatrix}$$

$$C_1 = 10^{-3} \times \begin{bmatrix} -4\omega_0^2 I_{yz} & 0 & 0 & 0 & -\omega_0 I_{yxz} \\ 0 & -3\omega_0^2 I_{xz} & 0 & 0 & 0 \\ 0 & 0 & -\omega_0^2 I_{yx} & \omega_0 I_{yzz} & 0 \end{bmatrix}$$

$$C_2 = [I_{3x3} \quad 0_{3x3}]$$

$$D_1 = 10^{-3} \times L_1, D_2 = 10^{-3} \times L_2$$

160.2 The Data

Data required for this LMI include moments of inertia of the satellite being controlled and the angular velocity of the earth. Any knowledge of the disturbance torques would also facilitate solution of the problem.

160.3 The Optimization Problem

The idea is to design a state feedback control law for the previous satellite state-space system of the form

$$u = Kx$$

This control law is designed so that the closed-loop system is stable and the transfer function matrix from disturbance to output

$$G_{z_\infty d}(s) = (C_1 + D_1 K)(sI - (A + B_1 K))^{-1} B_2 + D_2$$

satisfies

$$\|G_{z_\infty d}(s)\|_\infty = (C_1 + D_1 K)(sI - (A + B_1 K))^{-1} B_2 + D_2$$

for a minimal positive scalar γ_∞ which represents the minimum attenuation level.

The idea here is to attenuate the disturbances as much as possible while still maintaining the ability of the satellite to track. This minimum attenuation level is found from the LMI in the following section.

- Objective: H_{inf} norm
- Variables: Controller Gains
- Constraints: Satellite Attitude Dynamics and Kinematics. Maximum safe rotational rate of Satellite, maximum jet pulse thrust

160.4 The LMI: H_∞ Feedback Control of the Satellite System

Duan and Yu approach the H_∞ satellite system as follows. The minimum attenuation level from disturbance to output can be found by solving the following LMI optimization problem.

$$\min \gamma_\infty$$

$$\text{s.t. } X > 0$$

$$\begin{bmatrix} (AX + B_1 W)^T + AX + B_1 W & B_2 & (C_1 X + D_1 W)^T \\ B_2^T & -\gamma_{H_\infty} I & D_2^T \\ C_1 X + D_1 W & D_2 & -\gamma_{H_\infty} I \end{bmatrix} < 0$$

which is the same as Theorem 8.1 in Duan and Yu's Book, the solution to the H_∞ problem.

160.5 Conclusion:

The Duan and Yu textbook takes as typical values of the satellite moment of inertias as:

$$I_x = 1030.17 \text{ kg} \cdot \text{m}^2, I_y = 3015.65 \text{ kg} \cdot \text{m}^2, I_z = 3030.43 \text{ kg} \cdot \text{m}^2$$

They then proceed to solve the optimization problem to find a controller gain that yields an attenuation level of 0.0010. Though this value is very small and represents very good attenuation the optimized controller pushes the poles of the closed loop system very close to the imaginary axis, resulting in slow oscillatory behavior with a very long settling time.

To address this a second approach was used by the authors which involves modifying the final LMI in the expression above and requiring that it be constrained as follows

$$-I < \begin{bmatrix} (AX + B_1W)^T + AX + B_1W & B_2 & (C_1X + D_1W)^T \\ B_2^T & -\gamma_{H\infty}I & D_2^T \\ C_1X + D_1W & D_2 & -\gamma_{H\infty}I \end{bmatrix} < 0$$

These results are planned verified in the linked code implementation using YALMIP, whereas the authors took advantage of the MATLAB LMI Toolbox to achieve their results.

160.6 Implementation

A link to CodeOcean or other online implementation of the LMI

160.7 Related LMIs

Links to other closely-related LMIs

- H2_LMI_SatelliteAttitudeControl¹
- Optimal_Output_Feedback_Hinf_LMI²
- Full-State_Feedback_Optimal_Control_Hinf_LMI³

160.8 External Links

A list of references documenting and validating the LMI.

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Applications/H2_LMI_SatelliteAttitudeControl

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Optimal_Output_Feedback_Hinf_LMI

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Full-State_Feedback_Optimal_Control_Hinf_LMI

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- <https://www.routledge.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMIs in Control Systems: Analysis, Design, and Applications - Duan and Yu

⁴ http://control.asu.edu/MAE598_frame.htm

161 H2 LMI Satellite Attitude Control

UNKNOWN TEMPLATE FULLPAGENAME

This is a **H2 LMI for Satellite Attitude Control**

Satellite attitude control is important for military, civil, and scientific activities. Attitude control of a satellite involves fast maneuvering and accurate pointing in the presence of all kinds of disturbances and parameter uncertainties.

161.1 The System

The satellite state-space formulation is given in the H_∞ LMI page for Satellite Attitude Control which is also in the applications section of this WikiBook. This section discusses the derivation of that state-space formulation based on first principles.

The attitude dynamics of a satellite in an inertial coordinate system can be described in terms of the time rate of change of its angular momentum and the sum of the external torques and moments acting on the system. That is:

$$\begin{aligned}\dot{H} &= T_c + T_g + T_d, \\ H &= I_b \omega\end{aligned}$$

where the following variables are defined as follows:

- T_c, T_g , and T_d are the flywheel torque, the gravitational torque, and the disturbance torque.
- H is the total momentum/torque acting on the satellite
- I_b is the inertia matrix/tensor for the satellite
- ω is the angular velocity vector of the satellite.

The time derivative of the total angular momentum in an arbitrary rotating reference frame (such as the body frame of the satellite) is given by:

$$\dot{H} = I_b \dot{\omega} + \omega \times (I_b \omega),$$

which takes into the account of the angular velocity of the rotating reference frame relative to the inertial reference frame where Newton's laws are valid.

Combining equations, collecting terms and choosing the principle axes of the space-craft so that the Inertia Tensor is diagonalized yields the following equations of motion:

$$I_b \dot{\omega} + \omega \times (I_b \omega) = T_c + T_g + T_d,$$

Using the small angle approximation, the angular velocity of the satellite in the inertial coordinate system represented in the body coordinate system can be written as

$$\omega = \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} \dot{\phi} - \omega_0 \psi \\ \dot{\theta} - \omega_0 \\ \dot{\psi} + \omega_0 \psi \end{bmatrix}$$

where $\omega_0 = 7.292115 \times 10^{-5}$ [rad/s]

These equations form the basis of the state-space representation used in the H-inf LMI for satellite attitude control. For clarity, they are repeated below.

$$\begin{aligned} \dot{x} &= Ax + B_1 u + B_2 d \\ z_\infty &= C_1 x + D_1 u + D_2 d \\ z_2 &= C_2 x \end{aligned}$$

$$A = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ \frac{-4\omega_0^2 I_{yz}}{I_x} & 0 & 0 & 0 & 0 & \frac{-\omega_0 I_{yzz}}{I_x} \\ 0 & \frac{-3\omega_0^2 I_{xz}}{I_y} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{-\omega_0^2 I_{yx}}{I_z} & \frac{\omega_0 I_{yzz}}{I_x} & 0 & 0 \end{bmatrix}$$

$$B_1 = B_2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ \frac{1}{I_x} & 0 & 0 \\ 0 & \frac{1}{I_y} & 0 \\ 0 & 0 & \frac{1}{I_z} \end{bmatrix}$$

$$C_1 = 10^{-3} \times \begin{bmatrix} -4\omega_0^2 I_{yz} & 0 & 0 & 0 & -\omega_0 I_{yzz} & 0 \\ 0 & -3\omega_0^2 I_{xz} & 0 & 0 & 0 & 0 \\ 0 & 0 & -\omega_0^2 I_{yx} & \omega_0 I_{yzz} & 0 & 0 \end{bmatrix}$$

$$C_2 = \begin{bmatrix} I_{3x3} & 0_{3x3} \end{bmatrix}$$

$$D_1 = 10^{-3} \times L_1, D_2 = 10^{-3} \times L_2$$

161.2 The Data

Data required for this LMI include moments of inertia of the satellite being controlled and the angular velocity of the earth. Any knowledge of the disturbance torques would also facilitate solution of the problem.

161.3 The Optimization Problem

The optimization problem seeks to minimize the H₂ norm of the transfer function from disturbance to output. Thus, we expect slightly different results than the H-inf case. Deriving the H₂ control problem and setup also serves for useful setup for the mixed H-inf/H₂ optimization that the book follows up with later.

- Objective: H₂ norm
- Variables: Controller Gains
- Constraints: Satellite Attitude Dynamics and Kinematics. Maximum safe rotational rate of Satellite, maximum jet pulse thrust

161.4 The LMI: H-2 Satellite Attitude Control

Duan and Yu use the following H-2 Satellite Attitude Control LMI to minimize the attenuation level from disturbance to output. Note that in the H₂-case we are minimizing the integral of the magnitude of the bode plot transfer function whereas in the H-inf case the optimization is minimizing the maximum value of the bode plot magnitude.

To design an optimizing controller of the form

$$u = Kx$$

such that the closed-loop system is stable and the transfer function matrix

$$G_{z2w}(s) = C_2(sI - (A + B_1K))^{-1}B_2$$

satisfies

$$\|G_{z2w}(s)\|_2 < \gamma_2$$

for a minimal positive scalar γ_2 .

This scalar is found from the solution of the following LMI

$$\min \rho$$

$$\text{s.t. } AX + B_1W + (AX + B_1W)^T + B_2B_2^T < 0$$

$$\begin{bmatrix} -Z & C_2X \\ (C_2X)^T & -X \end{bmatrix} < 0$$

$$\text{trace}(Z) < \rho$$

$$\text{where } \rho = \gamma_2^2.$$

and the controller is given by $K = WX^{-1}$

161.5 Conclusion:

The LMI for H-2 Satellite Attitude Control comes up with a different attenuation value for the disturbance vs the H-inf problem which is expected. It also serves for good preparation for the mixed H2/H-inf problem that Duan and Yu cover in a later section. Though no implementation is included for the mixed H2/H-inf optimization problem it is interesting to compare the results of all three cases for the satellite attitude control problem.

161.6 Implementation

A link to CodeOcean or other online implementation of the LMI

161.7 Related LMIs

Links to other closely-related LMIs

- Hinf_LMI_SatelliteAttitudeControl¹
- H2OptimalOutputFeedback²
- Full-State_Feedback_Optimal_Control_H2³

1 https://en.wikibooks.org/wiki/LMIs_in_Control/Applications/Hinf_LMI_SatelliteAttitudeControl

2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Optimal_Output_Feedback_H2_LMI

3 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Full-State_Feedback_Optimal_Control_H2_LMI

161.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control⁴ - A course on LMIs in Control by Matthew Peet.
- <https://www.routledge.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMIs in Control Systems: Analysis, Design, and Applications - Duan and Yu

⁴ http://control.asu.edu/MAE598_frame.htm

162 Problem of Space Rendezvous and LMI Approaches

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This is a **Problem of Space Rendezvous and LMI Approaches**

In Section 12.4 of their book LMIs in Control Systems: Analysis, Design, and Applications, Duan and Yu discuss the problem of space rendezvous and how it can be formulated into an LMI problem. Modeling and simulating space rendezvous is of importance because it is used for any cargo or passenger spacecraft traveling to and from earth-orbiting space stations and also for satellites servicing aging in-orbit satellites, and for potential missions to mine asteroids.

162.1 The System

Though Duan and Yu first mention space rendezvous in Example 7.14 of their book. In this example, they show that the relative orbital dynamic model of spacecraft rendezvous can be described by the famous Clohessy-Wiltshire equations.

$$\begin{aligned} m\ddot{r}_x - 2m\omega_0\dot{r}_y - 3m\omega_0^2r_x &= T_x + d_x \\ m\ddot{r}_y + 2m\omega_0\dot{r}_x &= T_y + d_y \\ m\ddot{r}_z + m\omega_0^2r_z &= T_z + d_z \end{aligned}$$

where

- r_x, r_y, r_z are the components of the relative position between chaser and target
- $\omega_0 = \pi/12$ [rad/h] is the orbital angular velocity of the target satellite
- m is the mass of the chaser
- $T_i, (i = x, y, z)$ is the i-th component of the control input force acting on the relative motion dynamics
- $d_i, (i = x, y, z)$ is the i-th component of the external disturbance

The C-W equations give a first-order approximation of the chaser's motion in a target-centered coordinate system and is often used in planning space rendezvous problems (ISS, Salyut, and Tiangong space stations are just some examples.)

With appropriate definitions of states and variables the dynamic equations of motion for space-rendezvous can be converted into standard state-space form for LMI optimization as follows:

$$\begin{aligned}\dot{x} &= Ax + B_1 u + B_2 d \\ y &= Cx\end{aligned}$$

where the vectors in the above state-space representation are defined as follows:

$$x = [r_x \ r_y \ r_z \ \dot{r}_x \ \dot{r}_y \ \dot{r}_z]^T$$

$$y = [r_z \ r_y \ r_z]^T$$

$$u = [T_x \ T_y \ T_z]^T$$

$$d = [d_x \ d_y \ d_z]^T$$

and the matrices in the above state-space representation are defined as follows:

$$A = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 3\omega_0^2 & 0 & 0 & 0 & 2\omega_0^2 & 0 \\ 0 & 0 & 0 & -2\omega_0 & 0 & 0 \\ 0 & 0 & -\omega_0^2 & 0 & 0 & 0 \end{bmatrix}$$

$$B_1 = B_2 = \begin{bmatrix} 0_{3 \times 3} \\ I_3 \end{bmatrix}$$

$$C = [I_3 \ 0_{3 \times 3}]$$

162.2 The Data

The data required are the mass properties of both the target and chaser vehicles for space rendezvous. Also required is the orbital angular velocities of the target and chasers and measurements of relative kinematics between the two.

162.3 The Optimization Problem

The optimization problem is trying to attenuate the disturbance to output transfer function using either the H-inf or H2 norm.

- Objective: Hinf or H2 norm
- Variables: Controller Gains
- Constraints: Relative Dynamics/Kinematics between Chaser and Target in Orbit

162.4 The LMI: Space Rendezvous LMI Optimization

The space rendezvous problem can be approached with either H-inf or H-2 optimization formulations. Both formulations can achieve closed-loop stability which ensures that rendezvous occurs because the relative distance between target and chaser eventually approaches zero. The LMIs for the H-inf and H2 optimization problem are shown below which are easily solvable because the matrices for the space rendezvous problem are available above in standard form.

Duan and Yu approach the H_∞ . The minimum attenuation level from disturbance to output can be found by solving the following LMI optimization problem.

$$\min \gamma_\infty$$

$$\text{s.t. } X > 0$$

$$\begin{bmatrix} (AX + B_1W)^T + AX + B_1W & B_2 & (C_1X + D_1W)^T \\ B_2^T & -\gamma_{H_\infty}I & D_2^T \\ C_1X + D_1W & D_2 & -\gamma_{H_\infty}I \end{bmatrix} < 0$$

which is the same as Theorem 8.1 in Duan and Yu's Book, the solution to the H_∞ problem.

162.5 Conclusion:

The LMI for Space Rendezvous is a useful and interesting method to model and simulate practical problems in spacecraft engineering. Space Rendezvous usually requires very good vision-based navigation or an exceptional human operator that can close the gap for final mating of the two docking adapters.

162.6 Implementation

A link to CodeOcean or other online implementation of the LMI

162.7 Related LMIs

Links to other closely-related LMIs

- Optimal_Output_Feedback_Hinf_LMI¹
- H2OptimalOutputFeedback²

162.8 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control³ - A course on LMIs in Control by Matthew Peet.
- <https://www.routledge.com/LMIs-in-Control-Systems-Analysis-Design-and-Applications/Duan-Yu/p/book/9781466582996> - LMIs in Control Systems: Analysis, Design, and Applications - Duan and Yu

1 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Optimal_Output_Feedback_Hinf_LMI
2 https://en.wikibooks.org/wiki/LMIs_in_Control/pages/Optimal_Output_Feedback_H2_LMI
3 http://control.asu.edu/MAE598_frame.htm

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This methods uses LMI techniques iteratively to obtain the result.

163.1 The System

Given a state-space representation of a system $G(s)$ and an initial estimate of reduced order model $\hat{G}(s)$.

$$G(s) = C(sI - A)B + D,$$
$$\hat{G}(s) = \hat{C}(sI - \hat{A})\hat{B} + \hat{D},$$

Where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, $C \in \mathbb{R}^{p \times n}$, $D \in \mathbb{R}^{p \times m}$, $\hat{A} \in \mathbb{R}^{k \times k}$, $\hat{B} \in \mathbb{R}^{k \times m}$, $\hat{C} \in \mathbb{R}^{p \times k}$ and $\hat{D} \in \mathbb{R}^{p \times m}$.

163.2 The Data

The full order state matrices A, B, C, D .

163.3 The Optimization Problem

The objective of the optimization is to reduce the H_∞ norm .

163.4 The LMI: The Lyapunov Inequality

Objective: $\min \gamma$.

Subject to:: $P = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} > 0$,

$$\begin{bmatrix} A^T P_{11} + P_{11}A & A^T P_{12} + P_{12}\hat{A} & P_{11}B - P_{12}\hat{B} & C^T \\ \hat{A}^T P_{12}^T + P_{12}^T A & \hat{A}^T P_{22} + P_{22}\hat{A} & P_{12}^T B - P_{22}\hat{B} & \hat{C}^T \\ B^T P_{11} - \hat{B}^T P_{12}^T & B^T P_{12} - \hat{B}^T P_{22} & -\gamma I & D^T - \hat{D}^T \\ C & \hat{C} & D - \hat{D} & -\gamma I \end{bmatrix} > 0$$

163.5 Conclusion:

The LMI techniques results in model reduction close to the theoretical bounds.

163.6 External Links

A list of references documenting and validating the LMI.

- LMI Methods in Optimal and Robust Control¹ - A course on LMIs in Control by Matthew Peet.
- LMIs in Systems and Control Theory² - A downloadable book on LMIs by Stephen Boyd.

1 http://control.asu.edu/MAE598_frame.htm
2 <https://web.stanford.edu/~boyd/lmibook/>

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³⁵ Chapter 165 on page 487

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