Power Density Spectrum - Continuous Time

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Based on Probability, Random Variables and Random Signal Principles, P.Z. Peebles, Jr. and B. Shi

Outline

Energy, Average Power – deterministic, time domain

a deterministic signal x(t)

$$x_T(t) = \begin{cases} x(t) & -T < t < T \\ 0 & otherwise \end{cases}$$

the energy

$$E(T) = \int_{-T}^{+T} x^2(t) dt = \int_{-\infty}^{+\infty} x_T^2(t) dt$$

the average power

$$P(T) = \frac{1}{2T} \int_{-T}^{+T} x^2(t) dt = \frac{1}{2T} \int_{-\infty}^{+\infty} x_T^2(t) dt$$

Fourier Transform Pair $x(t) \iff X(\mathbf{o})$

Fourier transform

$$X(\mathbf{\omega}) = \int_{-\infty}^{\infty} x(t) e^{-j\mathbf{\omega}t} dt$$

a deterministic signal x(t)

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} x(\omega) e^{j\omega t} d\omega$$

bounded duration, bounded variation

for a finite T, $x_T(t)$ is assumed to have bounded variation

$$\int_{-T}^{+T} |x(t)| dt < \infty$$

the Fourier transform of $x_T(t)$

$$X_{T}(\mathbf{\omega}) = \int_{-\infty}^{+\infty} x_{T}(t) e^{-j\mathbf{\omega}t} dt$$

$$= \int_{-T}^{+T} x(t) e^{-j\omega t} dt$$

Fourier transforms of $x_T(t)$ and $X_T(t)$ for continuous time signals

deterministic $X_T(\omega)$ v.s. random $X_T(\omega)$

a deterministic sample signal $x_T(t)$

$$X_T(t) \Longleftrightarrow X_T(\omega)$$

a random process signal $X_T(t)$

$$X_T(t) \Longleftrightarrow X_T(\omega)$$

Parseval's theorem (I) for continuous time signals

for a deterministic $x_T(t)$

a deterministic sample signal $x_T(t)$

$$\int_{-\infty}^{+\infty} x_T(\tau) x_T^*(\tau) d\tau = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X_T(\mathbf{\omega}) X_T^*(\mathbf{\omega}) d\omega$$

$$\int_{-\infty}^{+\infty} |x_T(\tau)|^2 d\tau = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |X_T(\omega)|^2 d\omega$$

for a deterministic $x_T(t)$ v.s. a random $X_T(t)$

• a deterministic signal $x_T(t) \iff X_T(\omega)$

$$\int_{-\infty}^{+\infty} |x_T(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |X_T(\omega)|^2 d\omega$$

• a random signal $X_T(t) \iff X_T(\omega)$

$$\int_{-\infty}^{+\infty} \mathbf{E} \left[|X_T(t)|^2 \right] dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \mathbf{E} \left[|X_T(\boldsymbol{\omega})|^2 \right] d\omega$$

Energy and average power in frequency domain for continuous time signals

Energy, Average Power - Parseval's theorem applied

a deterministic signal $x_T(t)$

$$x_T(t) = \begin{cases} x(t) & -T < t < T \\ 0 & \text{otherwise} \end{cases} x_T(t) \iff X_T(\omega)$$

the energy by Parseval's theorem

$$E(T) = \int_{-T}^{+T} x^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |X_T(\omega)|^2 d\omega$$

the average power by Parseval's theorem

$$P(T) = \frac{1}{2T} \int_{-T}^{+T} x^{2}(t) dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{|X_{T}(\omega)|^{2}}{2T} d\omega$$

E(T) and P(T) in frequency domain – deterministic case for continuous time signals

deterministic $x_T(t) \iff X_T(\omega)$

the energy for the deterministic $X_T(\omega)$ in $x_T(t) \iff X_T(\omega)$

$$E(T) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} |X_T(\omega)|^2 d\omega$$

the average power for the deterministic $X_T(\omega)$

$$P(T) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{|X_T(\omega)|^2}{2T} d\omega$$

the power density spectrum for the deterministic $X_T(\omega)$

$$\lim_{T\to\infty}\frac{|X_T(\omega)|^2}{2T}$$



E(T) and P(T) in frequency domain – random case for continuous time signals

random $X_T(t) \iff X_T(\omega)$

the energy for the random $X_T(\omega)$ in $X_T(t) \iff X_T(\omega)$

$$E(T) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} E[|X_T(\omega)|^2] d\omega$$

the average power for the random $X_T(\omega)$

$$P(T) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{E\left[|X_T(\omega)|^2\right]}{2T} d\omega$$

the power density spectrum for the random $X_T(\omega)$

$$\lim_{T\to\infty}\frac{E\left[|X_T(\omega)|^2\right]}{2T}$$



Average power P(T) – bounded duration (-T, +T) for continuous time signals

$$P(T) = \frac{1}{2T} \int_{-T}^{+T} x^2(t) dt$$

- not the average power in a random process only the power in one sample function
 - to obtain the average power over all possible realizations, replace x(t) by X(t) take the expected value of $x^2(t)$, that is $E[X^2(t)]$
 - then, the average power is a random variable with respect to the random process X(t)
- not the average power in an entire sample function
 - take $T \to \infty$ to include all power in the **ensemble** member



Average power P_{XX} – unbounded duration $(-\infty, +\infty)$ for continuous time signals

$$P(T) = \frac{1}{2T} \int_{-T}^{+T} x^2(t) dt$$

- replace x(t) by the random variable X(t)
- take the expected value of $x^2(t)$, that is $E[X^2(t)]$

$$P(T) = \frac{1}{2T} \int_{-T}^{+T} \mathbf{E} \left[X^2(t) \right] dt$$

• take $T \to \infty$ to include all power

$$\boxed{P_{XX} = \lim_{T \to \infty} P(T)} = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{+T} \mathbf{E} \left[X^2(t) \right] dt$$



Average power P_{XX} – time average $A[\bullet]$ for continuous time signals

The time average

$$A_{T}[\bullet] = \frac{1}{2T} \int_{-T}^{T} [\bullet] dt \qquad A[\bullet] = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} [\bullet] dt$$

time average and sample average operations

$$\begin{bmatrix}
P_{XX} = \lim_{T \to \infty} P(T) \\
 = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{+T} E[X^{2}(t)] dt$$

$$= \lim_{T \to \infty} A_{T} [E[X^{2}(t)]]$$

$$= A[E[X^{2}(t)]]$$

for deterministic and random signals

the average power P(T) for a deterministic signal x(t)

$$P(T) = \frac{1}{2T} \int_{-T}^{+T} x^2(t) dt$$

the average power P_{XX} for a random process X(t)

$$P_{XX} = \lim_{T \to \infty} P(T)$$

$$= \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{+T} E[X^{2}(t)] dt$$

$$= A[E[X^{2}(t)]]$$

Power density spectrum $S_{XX}(\omega)$ for continuous time signals

the average power via power density

the average power P_{XX} for the <u>random process</u> $X_T(\omega)$

$$P_{XX} = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left[\lim_{T \to \infty} \frac{\mathbf{E} \left[|X_T(\omega)|^2 \right]}{2T} \right] d\omega$$
$$= \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left[S_{XX}(\omega) \right] d\omega$$

the power density spectrum $S_{XX}(\omega)$

$$\boxed{S_{XX}(\boldsymbol{\omega})} = \lim_{T \to \infty} \frac{E\left[|X_T(\boldsymbol{\omega})|^2\right]}{2T}$$

Properties of Power Spectrum

for continuous time signals

•
$$S_{XX}(\omega) \geq 0$$

•
$$S_{XX}(-\omega) = S_{XX}(\omega)$$

$$X(t)$$
 real

- $S_{XX}(\omega)$ real
- $\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) d\omega = A \left[E \left[X^2(t) \right] \right]$
- $S_{\dot{X}\dot{X}}(\omega) = \omega^2 S_{XX}(\omega)$
- $\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) e^{j\omega t} d\omega = A[R_{XX}(t,t+\tau)]$
- $S_{XX}(\omega) = \int_{-\infty}^{+\infty} A[R_{XX}(t,t+\tau)] e^{-j\omega\tau} d\tau$

the average power P_{xx} and the inverse Fourier transform of $S_{XX}(\omega)$

the average power related equation

$$\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) d\omega = A \left[\mathbf{E} \left[X^2(t) \right] \right]$$

the autocorrelation related equation

$$\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) e^{j\omega t} d\omega = A[R_{XX}(t, t+\tau)]$$

the average power P_{xx}

the average power related equation

$$\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) d\omega = A \left[\mathbf{E} \left[X^2(t) \right] \right]$$

- a random process X(t) in time domain
- a random process $X(\omega)$ in frequency domain
- Parseval's theorem over $X_T(t) \iff X_T(\omega)$

$$X(t) = \lim_{T \to \infty} X_T(t)$$
 $X(\omega) = \lim_{T \to \infty} X_T(\omega)$

Average power P_{XX} in time / frequency domain for continuous time signals

Definition

Using a random process X(t) in time domain

$$P_{XX} = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{+T} E\left[X^{2}(t)\right] dt$$
$$= \lim_{T \to \infty} A_{T} \left[E\left[X^{2}(t)\right]\right] = A\left[E\left[X^{2}(t)\right]\right]$$

Using a random process $X_T(\omega)$ in frequency domain

$$P_{XX} = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \left[\lim_{T \to \infty} \frac{E\left[|X_T(\omega)|^2 \right]}{2T} \right] d\omega$$
$$= \left[\frac{1}{2\pi} \int_{-\infty}^{+\infty} \left[S_{XX}(\omega) \right] d\omega \right]$$

the Inverse Fourier transform of $S_{XX}(\mathbf{o})$

the average power related equation

$$\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\boldsymbol{\omega}) e^{j\omega t} d\omega = A[R_{XX}(t, t+\tau)]$$

auto-correlation function

$$R_{XX}(t, t+\tau) = E[X(t)X(t+\tau)] \Rightarrow R_{XX}(\tau)$$

- a random process X(t) in time domain
- a random process $X_T(\omega)$ in frequency domain

Fourier transforms of autocorrelation functions for continuous time signals

Definition

Fourier transform of an autocorrelation functions

$$S_{XX}(\mathbf{\omega}) = \int_{-\infty}^{+\infty} R_{XX}(\tau) e^{-j\mathbf{\omega}\tau} d\tau$$
$$S_{\dot{X}\dot{X}}(\mathbf{\omega}) = \int_{-\infty}^{+\infty} R_{\dot{X}\dot{X}}(\tau) e^{-j\mathbf{\omega}\tau} d\tau$$

auto-correlation function

$$R_{XX}(t,t+\tau) = E[X(t)X(t+\tau)] \Rightarrow R_{XX}(\tau)$$

$$R_{\dot{X}\dot{X}}(t,t+\tau) = E[\dot{X}(t)\dot{X}(t+\tau)] \Rightarrow R_{\dot{X}\dot{X}}(\tau)$$

- a random process X(t) in time domain
- $\dot{X}(t) = \frac{d}{dt}X(t)$: the derivative of X(t)



Fourier transform of a derivative function for continuous time signals

Definition

Fourier transform of an autocorrelation functions

$$x(t) \Longleftrightarrow X(\omega)$$
$$\frac{d^n}{dt^n}x(t) \Longleftrightarrow (j\omega)^n X(\omega)$$

Power Density Spectrum and Auto-correlation for continuous time signals

Definition

$$S_{XX}(\omega) = \int_{-\infty}^{+\infty} A[R_{XX}(t, t+\tau)] e^{-j\omega\tau} d\tau$$

$$\frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) e^{j\omega t} d\omega = A[R_{XX}(t, t+\tau)]$$

for a WSS
$$X(t)$$
, $A[R_{XX}(t,t+ au)]=R_{XX}(au)$

$$S_{XX}(\omega) = \int_{-\infty}^{+\infty} R_{XX}(\tau) e^{-j\omega\tau} d\tau$$

$$R_{XX}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) e^{+j\omega\tau} d\omega$$

Power Spectrum and Auto-Correlation Functions for continuous time signals

Definition

the power spectrum

$$S_{XX}(\omega) = \int_{-\infty}^{+\infty} R_{XX}(\tau) e^{-j\omega\tau} d\tau$$

the auto-correlation function

$$R_{XX}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} S_{XX}(\omega) e^{+j\omega\tau} d\omega$$

Definition

the standard deviation is a measure of the spread in a density function. the analogous quantity for the normalized power spectrum is a measure of its spread that we call the rms bandwidth (root-mean-square)

$$W_{rms}^{2} = \frac{\int_{-\infty}^{+\infty} \omega^{2} S_{XX}(\omega) d\omega}{\int_{-\infty}^{+\infty} S_{XX}(\omega) d\omega}$$

Definition

the mean frequence $\bar{\omega}_0$

$$\bar{\omega}_0 = \frac{\int_{-\infty}^{+\infty} {}_{\omega} S_{XX}(\omega) d\omega}{\int_{-\infty}^{+\infty} S_{XX}(\omega) d\omega}$$

the rms bandwidth

$$W_{rms}^{2} = \frac{4 \int_{-\infty}^{+\infty} (\boldsymbol{\omega} - \bar{\omega}_{0})^{2} S_{XX}(\boldsymbol{\omega}) d\boldsymbol{\omega}}{\int_{-\infty}^{+\infty} S_{XX}(\boldsymbol{\omega}) d\boldsymbol{\omega}}$$