

# DLTI Convolution (1B)

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- Circular Convolution
- Numerical Convolution
- Moving Average Filter

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# Based on

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Introduction to Signal Processing

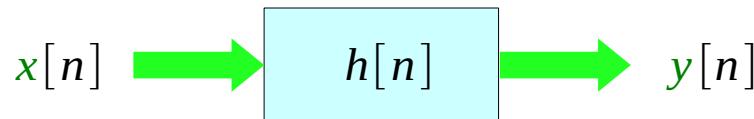
S. J. Ofranidis

The necessities in DSP C Programming

FIR Filter (A.pdf) 20191114

# Linear Convolution using the DFT

$$y[n] = \sum_{k=-\infty}^{+\infty} h[k] x[n-k]$$

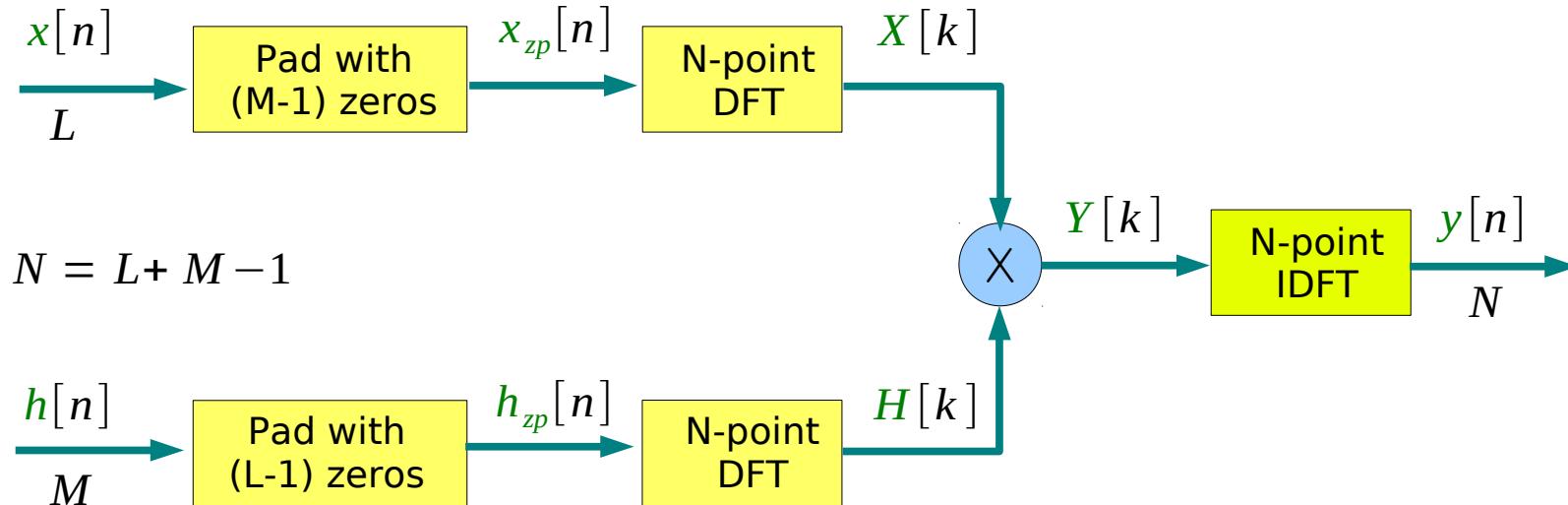


$$x[n] \quad 0 \leq n \leq L-1$$

$$h[n] \quad 0 \leq n \leq M-1$$

$$y[n] \quad 0 \leq n \leq L+M-2$$

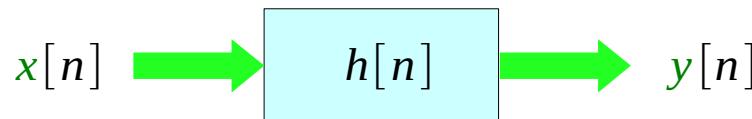
$$Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega})$$



# Circular Convolution using the DFT

$$y[n] = \sum_{k=-\infty}^{+\infty} h[k] x[n-k]$$

**Linear Convolution**



$$\begin{aligned} x[n] & \quad 0 \leq n \leq L-1 \\ h[n] & \quad 0 \leq n \leq M-1 \\ y[n] & \quad 0 \leq n \leq L+M-2 \end{aligned}$$

$$Y(e^{j\hat{\omega}}) = H(e^{j\hat{\omega}})X(e^{j\hat{\omega}})$$

DTFT

$\omega$  continuous frequency variable

$$Y[k] = Y(e^{j2\pi k/N}) \quad 0 \leq k \leq N-1$$

$$H[k] = H(e^{j2\pi k/N}) \quad N-M \text{ 0's}$$

$$X[k] = X(e^{j2\pi k/N}) \quad N-L \text{ 0's}$$

N-point  
DFT

$k$  sampled frequency variable

$$Y[k] = H[k]X[k]$$

$$y[n] = \sum_{m=0}^{N-1} h[m] x[\langle n - m \rangle_N] \quad 0 \leq n \leq N-1$$

**Circular Convolution**

$$y[n] = h[n] \circledcirc_N x[n]$$

# Circular Convolution using the DFT

## Linear Convolution

$$y[n] = h[n] * x[n]$$

DTFT

$$Y(e^{j\hat{\omega}}) = H(e^{j\hat{\omega}})X(e^{j\hat{\omega}})$$

$$y[n] = \sum_{k=-\infty}^{+\infty} h[k] x[n-k]$$

$\omega$  continuous frequency variable

## Circular Convolution

$$y[n] = h[n] \circledcirc_N x[n]$$

N-point  
DFT

$$Y[k] = H[k]X[k]$$

$$y[n] = \sum_{m=0}^{N-1} h[m] x[\langle n - k \rangle_N]$$

$$0 \leq n \leq N-1$$

$k$  sampled frequency variable

# Circular Convolution (1)

$$\begin{aligned}y[n] &= \frac{1}{N} \sum_{k=0}^{N-1} H[k] X[k] W_N^{-kn} \\&= \frac{1}{N} \sum_{k=0}^{N-1} \left\{ \sum_{m=0}^{N-1} h[m] W_N^{+k m} \right\} \left\{ \sum_{l=0}^{N-1} x[l] W_N^{+k l} \right\} W_N^{-kn} \\&= \sum_{m=0}^{N-1} h[m] \sum_{l=0}^{N-1} x[l] \left\{ \frac{1}{N} \sum_{k=0}^{N-1} W_N^{+k(m+l-n)} \right\} \\&\left\{ \frac{1}{N} \sum_{k=0}^{N-1} W_N^{+k(m+l-n)} \right\} = \begin{cases} 1, & m+l-n = rN \\ 0, & otherwise \end{cases}\end{aligned}$$

$$l = n - m + rN = (n - m) \bmod N = \langle n - m \rangle_N$$

# Circular Convolution (1)

$$X_1[k] = \sum_{n=0}^{N-1} x_1[n] e^{-j2\pi n k / N} \quad k = 0, 1, \dots, N-1$$

$$X_2[k] = \sum_{n=0}^{N-1} x_2[n] e^{-j2\pi n k / N} \quad k = 0, 1, \dots, N-1$$

$$X_3[k] = X_1[k]X_2[k] \quad k = 0, 1, \dots, N-1$$

$$\begin{aligned} x_3[n] &= \frac{1}{N} \sum_{k=0}^{N-1} X_3[k] e^{+j2\pi k m / N} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} X_1[k]X_2[k] e^{+j2\pi k m / N} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \left( \sum_{n=0}^{N-1} x_1[n] e^{-j2\pi n k / N} \right) \left( \sum_{l=0}^{N-1} x_2[l] e^{-j2\pi l k / N} \right) e^{+j2\pi k m / N} \\ &= \frac{1}{N} \sum_{n=0}^{N-1} x_1[n] \sum_{l=0}^{N-1} x_2[l] \sum_{k=0}^{N-1} e^{-j2\pi n k / N} e^{-j2\pi l k / N} e^{+j2\pi k m / N} \\ &= \frac{1}{N} \sum_{n=0}^{N-1} x_1[n] \sum_{l=0}^{N-1} x_2[l] \sum_{k=0}^{N-1} e^{+j2\pi k(m-n-l)/N} \end{aligned}$$

# Circular Convolution (2)

$$X_1[k] = \sum_{n=0}^{N-1} x_1[n] e^{-j2\pi n k/N}$$

$$X_2[k] = \sum_{n=0}^{N-1} x_2[n] e^{-j2\pi n k/N} \quad k = 0, 1, \dots, N-1$$

$$X_3[k] = X_1[k]X_2[k] \quad k = 0, 1, \dots, N-1$$

$$x[\textcolor{blue}{k}] = \frac{1}{N} \sum_{\textcolor{green}{n}=0}^{N-1} x_1[\textcolor{green}{n}] \sum_{\textcolor{violet}{l}=0}^{N-1} x_2[\textcolor{violet}{l}] \sum_{\textcolor{blue}{k}=0}^{N-1} e^{+j2\pi k(m-\textcolor{green}{n}-\textcolor{violet}{l})/N}$$

$$\sum_{k=0}^{N-1} a^k = N \quad (a = 1)$$

$$a = e^{+j2\pi(m-\textcolor{green}{n}-\textcolor{violet}{l})/N} \quad = \frac{1-a^N}{1-a} \quad (a \neq 1)$$

# Circular Convolution (2)

$$\frac{(m-\textcolor{green}{n}-\textcolor{violet}{l})}{N} \in \left\{ \begin{array}{cccc} \frac{0}{N}, & \frac{1}{N}, & \dots, & \frac{N-1}{N}, \\ \frac{0}{N}+1, & \frac{1}{N}+1, & \dots, & \frac{N-1}{N}+1, \\ \frac{0}{N}+2, & \frac{1}{N}+2, & \dots, & \frac{N-1}{N}+2, \\ \vdots & \vdots & & \vdots \end{array} \right\}$$

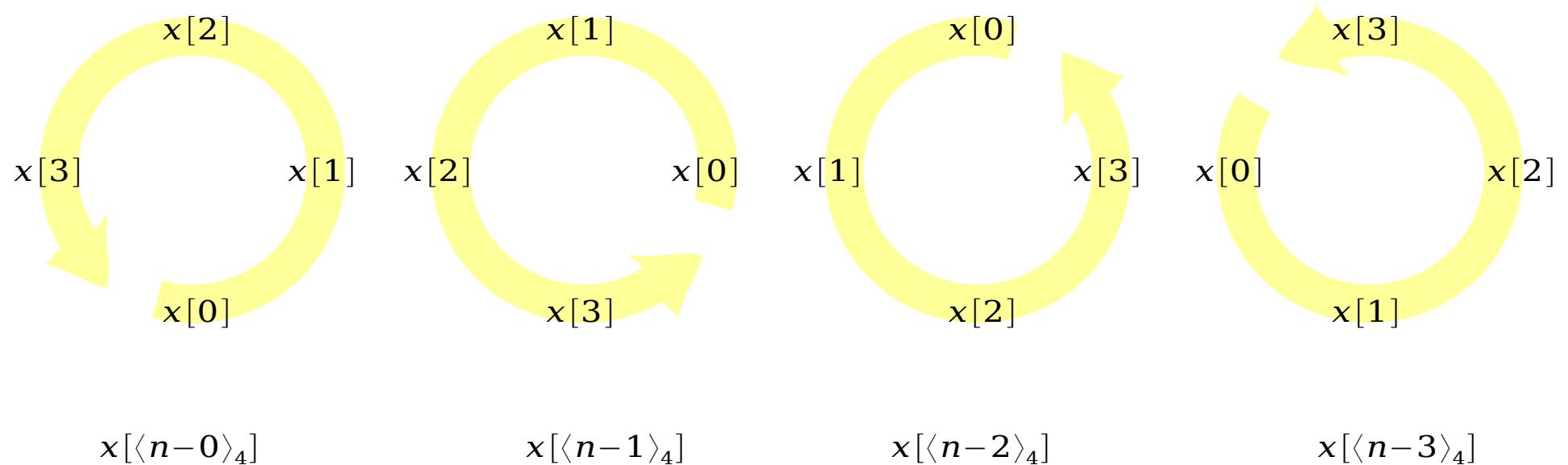
$$e^{j\frac{2\pi(m-\textcolor{green}{n}-\textcolor{violet}{l})}{N}} \in \left\{ e^{j2\pi \cdot 0/N}, e^{j2\pi \cdot 1/N}, \dots, e^{j2\pi(N-1)/N} \right\} \quad e^{j\frac{2\pi(m-\textcolor{green}{n}-\textcolor{violet}{l})}{N}} = e^{j\frac{2\pi((m-\textcolor{green}{n}-\textcolor{violet}{l}))_N}{N}}$$

$((m-\textcolor{green}{n}-\textcolor{violet}{l}))_N = 0 \rightarrow e^{j\frac{2\pi(m-\textcolor{green}{n}-\textcolor{violet}{l})}{N}} = e^{j\frac{2\pi((m-\textcolor{green}{n}-\textcolor{violet}{l}))_N}{N}} = 1$

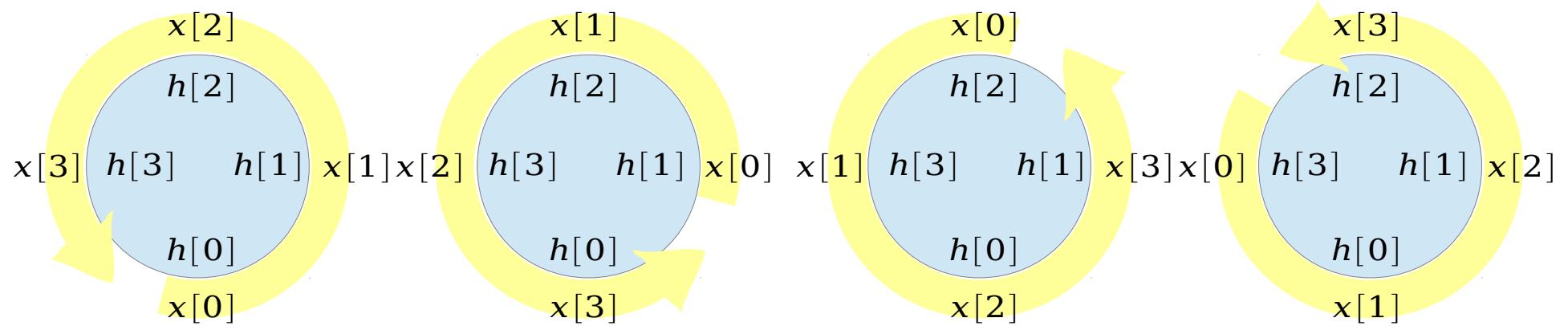
$\sum_{\textcolor{green}{k}=0}^{N-1} e^{j\frac{2\pi \textcolor{green}{k}(m-\textcolor{green}{n}-\textcolor{violet}{l})}{N}} = \sum_{\textcolor{green}{k}=0}^{N-1} e^{j\frac{2\pi \textcolor{green}{k}((m-\textcolor{green}{n}-\textcolor{violet}{l}))_N}{N}} = N$

$((m-\textcolor{green}{n}))_N = \textcolor{violet}{l}$

# Circular Convolution



# Circular Convolution

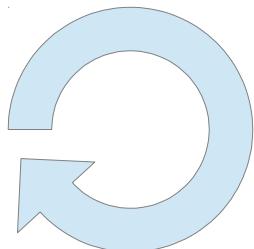


$x[\langle n-0 \rangle_4]$

$x[\langle n-1 \rangle_4]$

$x[\langle n-2 \rangle_4]$

$x[\langle n-3 \rangle_4]$



# Circular Convolution

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# Octave conv()

`conv (a, b)`

`conv (a, b, shape)`

Convolve two vectors a and b.

The length of the output vector : `length (a) + length (b) - 1`

When a and b are the coefficient vectors of two polynomials,  
the convolution represents the coefficient vector of the product polynomial.

The optional shape argument may be

`shape = "full"`

Return the full convolution. (default)

`shape = "same"`

Return the central part of the convolution with the same size as a.

See also: `deconv` `conv2` `convn` `fftconv`

# Octave fftconv()

`fftconv (x, y)`

`fftconv (x, y, n)`

Convolve two vectors using the FFT for computation.

`c = fftconv (x, y)` returns

a vector of length equal to `length (x) + length (y) - 1`.

If `x` and `y` are the coefficient vectors of two polynomials,  
the returned value is the coefficient vector of the product polynomial.

The computation uses the FFT by calling the function `fftfilt`.

If the optional argument `n` is specified, an `N`-point FFT is used.

`fftfilt (b, x, n)`

With two arguments, `fftfilt` filters `x` with the FIR filter `b` using the FFT.

Given the optional third argument, `n`,

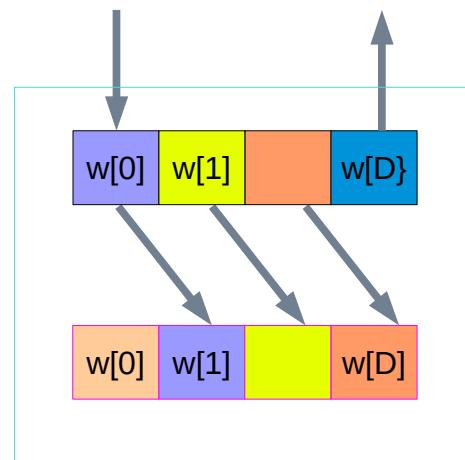
`fftfilt` uses the overlap-add method to filter `x` with `b` using an `N`-point FFT.

If `x` is a matrix, filter each column of the matrix.

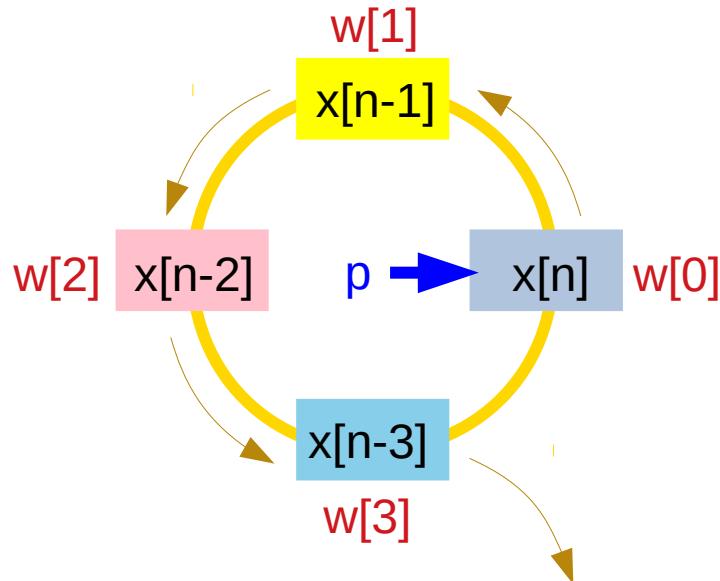
# Linear delay-line buffer

At each time instant,  
the data in the delay line are shifted  
one memory location ahead.

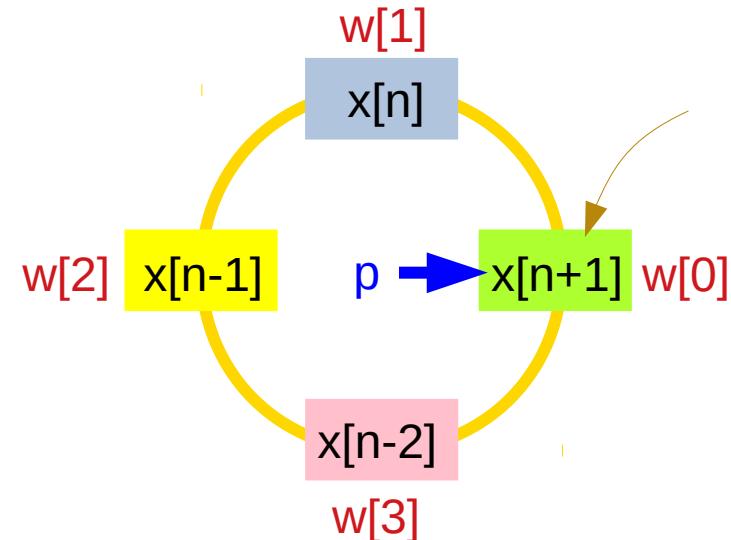
shifting the data forward  
while holding the buffer addresses fixed



# Wrapped linear delay-line buffer



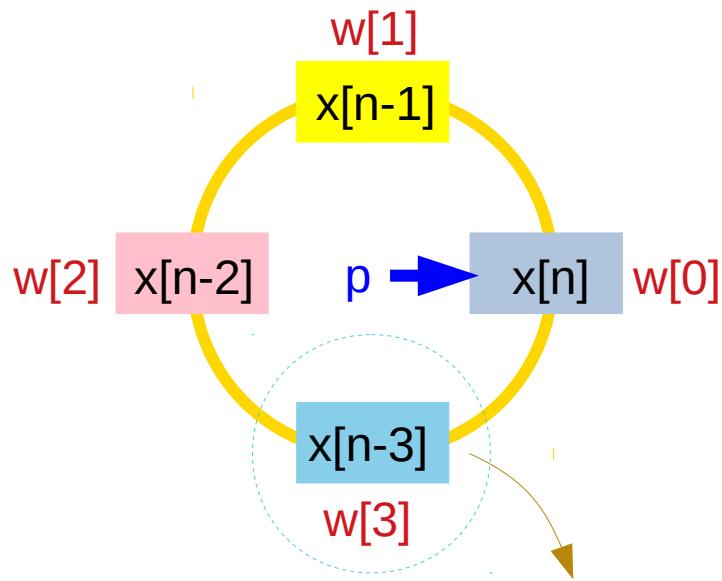
the **data** are **shifted** one location **forward**  
the buffer **addresses** are **fixed**



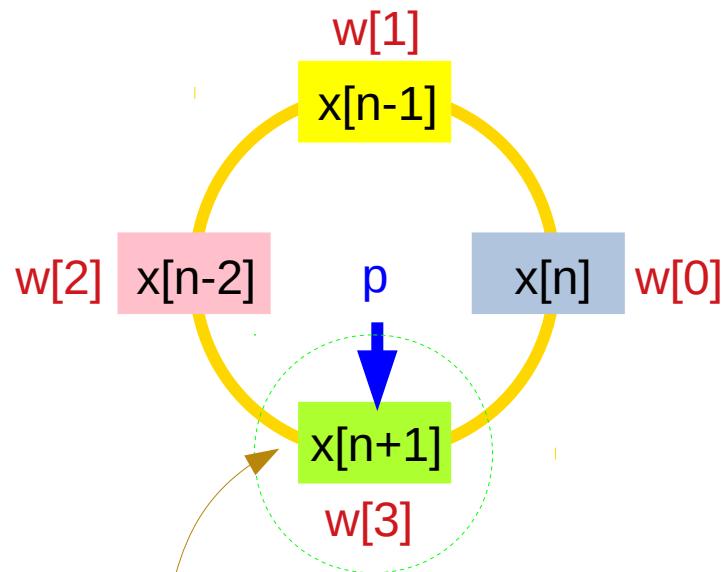
the **starting** address **p** is  
**fixed**

0	1	2	3	index
p	$p+1$	$p+2$	$p+3$	pointer at time <b>n</b>
p	$p+1$	$p+2$	$p+3$	pointer at time <b>n+1</b>
$x[n]$	$x[n-1],$	$x[n-2],$	$x[n-3]$	data at time <b>n</b>
$x[n+1],$	$x[n],$	$x[n-1],$	$x[n-2]$	data at time <b>n+1</b>

# Circular delay-line buffer



the **data** are kept **fixed**  
the **addresses** are **shifted backwards**



the **starting** address  **$p$**  is  
shifted **backward**

0	1	2	3	location index
$p$	$p+1$	$p+2$	$p+3$	pointer at time $n$
$p+1$	$p+2$	$p+3$	$p$	pointer at time $n+1$
$x[n]$	$x[n-1],$	$x[n-2],$	$x[n-3]$	data at time $n$
$x[n],$	$x[n-1],$	$x[n-2],$	$x[n+1]$	data at time $n+1$

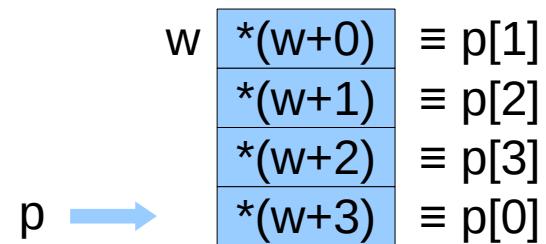
# Pointer variable p

A pointer variable **p**  
always points to the current input

A buffer address **w** (or w register)

At each time instant, the buffer (register)  
pointed to by **p** gets loaded  
with the current input sample

$$*p = x \quad p[0] = x$$



# Offset integer q

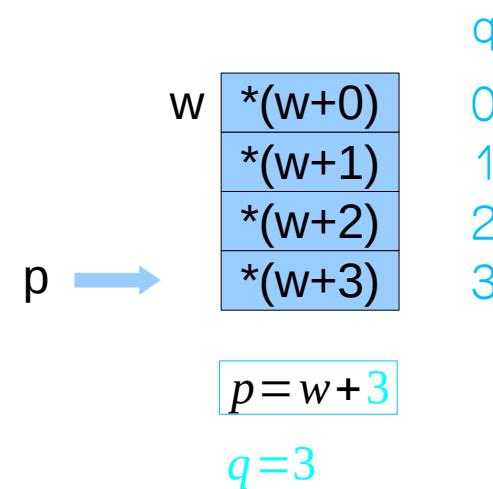
The pointer **p** is restricted to lie within the pointer range of the linear buffer **w** ,

$$w \leq p \leq w + M$$

always point at somewhere in the **w** buffer, **w[q]**

$$p = w + q \quad \Rightarrow$$

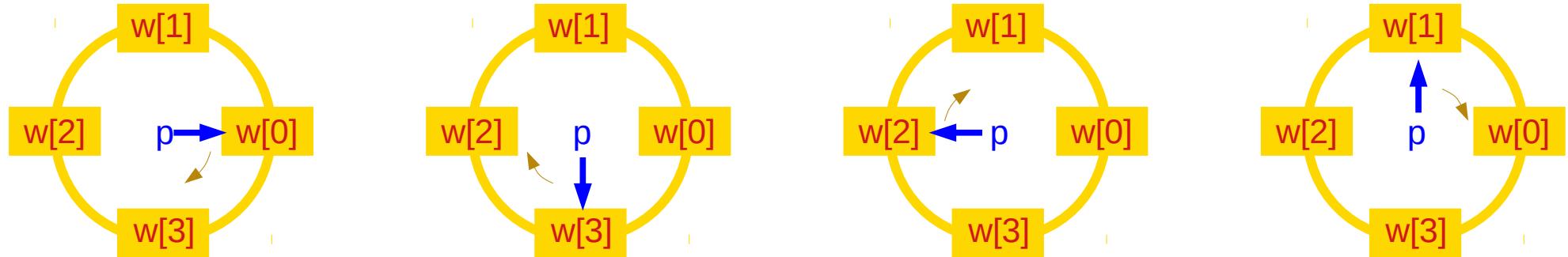
$$*p = p[0] = w[q]$$



**q** is an integer that gives the offset of **p** from the fixed beginning of the **w** buffer.

$$0 \leq q \leq M$$

# Address pointer p and index q



$n =$	0, 4, 8, 12, $\dots$	1, 5, 9, 13,...	2, 6, 10, 14, $\dots$	3, 7, 11, 15, $\dots$
$w[0] = *(w+0)$		$w[3] = *(w+3)$	$w[2] = *(w+2)$	$w[1] = *(w+1)$
$p$ when $q = 0$		$p$ when $q = 3$	$p$ when $q = 2$	$p$ when $q = 1$

$M = 3$   
modulo  $(M+1) = \text{modulo } 4$

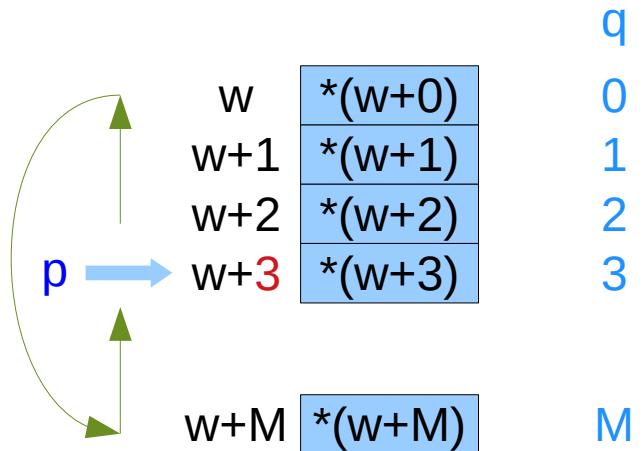
# Address pointer p (moving) and w (fixed)

modulo (M+1)

$$w \leq p \leq w+M$$

$$p = w + q \quad *p = * (w + q) = w[q]$$

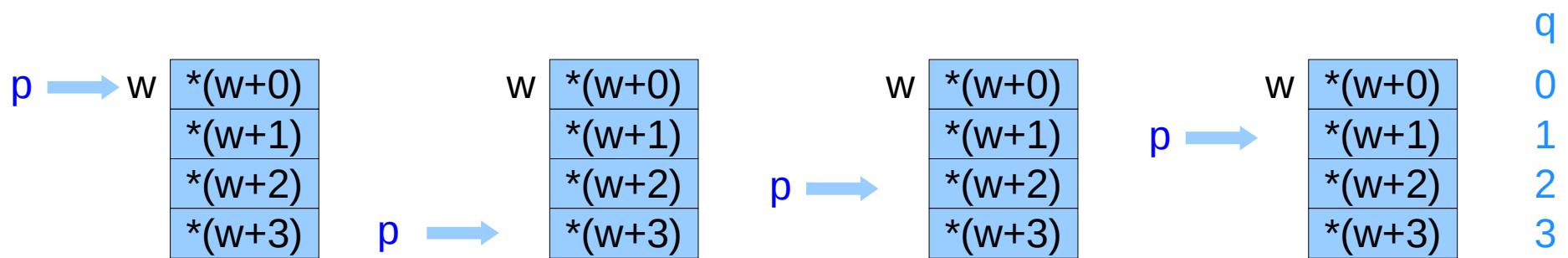
$$0 \leq q \leq M$$



$p[0]$	$* (p+0) = * (w+3+0) = w[3]$
$p[1]$	$* (p+1) = * (w+3+1) = w[0]$
$p[2]$	$* (p+2) = * (w+3+2) = w[1]$
$p[3]$	$* (p+3) = * (w+3+3) = w[2]$

# Accessing the buffer $w[k]$ by $p[i]$

modulo (3+1)



$$p[0] = w[0]$$

$$p[1] = w[1]$$

$$p[2] = w[2]$$

$$p[3] = w[3]$$

when  $q=0$

$$p[0] = w[3]$$

$$p[1] = w[0]$$

$$p[2] = w[1]$$

$$p[3] = w[2]$$

when  $q=3$

$$p[0] = w[2]$$

$$p[1] = w[3]$$

$$p[2] = w[0]$$

$$p[3] = w[1]$$

when  $q=2$

$$p[0] = w[1]$$

$$p[1] = w[2]$$

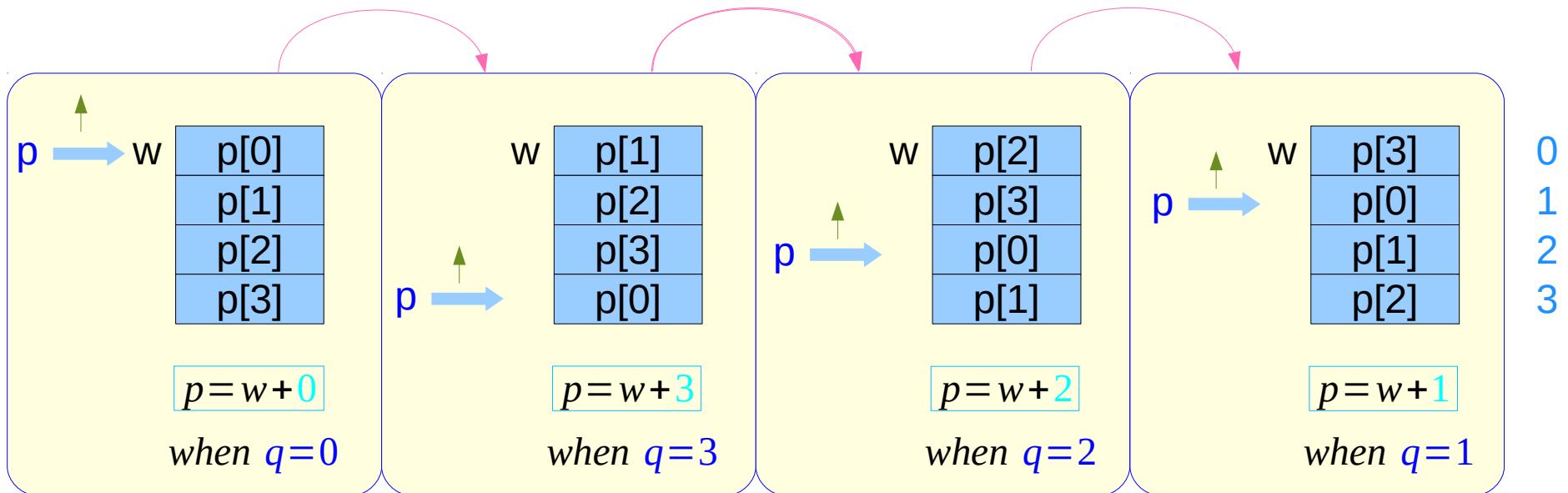
$$p[2] = w[3]$$

$$p[3] = w[0]$$

when  $q=1$

# Address pointer $p$ moves backward

modulo (3+1)



$p[0]$

$p[1]$

$p[2]$

$p[3]$

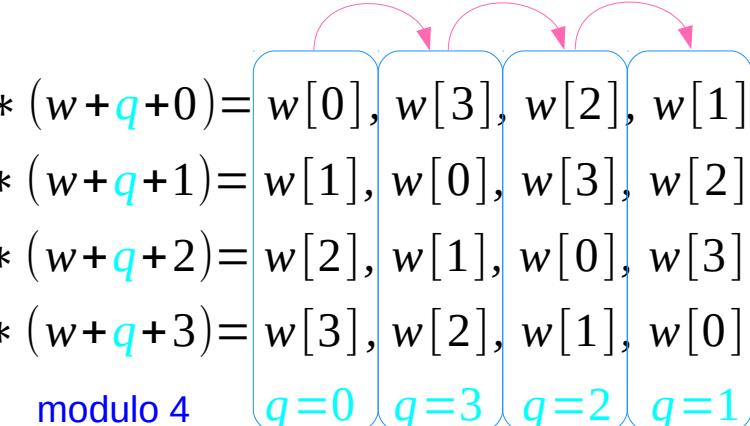
$* (p+0) = * (w+q+0) = w[0], w[3], w[2], w[1]$

$* (p+1) = * (w+q+1) = w[1], w[0], w[3], w[2]$

$* (p+2) = * (w+q+2) = w[2], w[1], w[0], w[3]$

$* (p+3) = * (w+q+3) = w[3], w[2], w[1], w[0]$

modulo 4



# Address pointer p

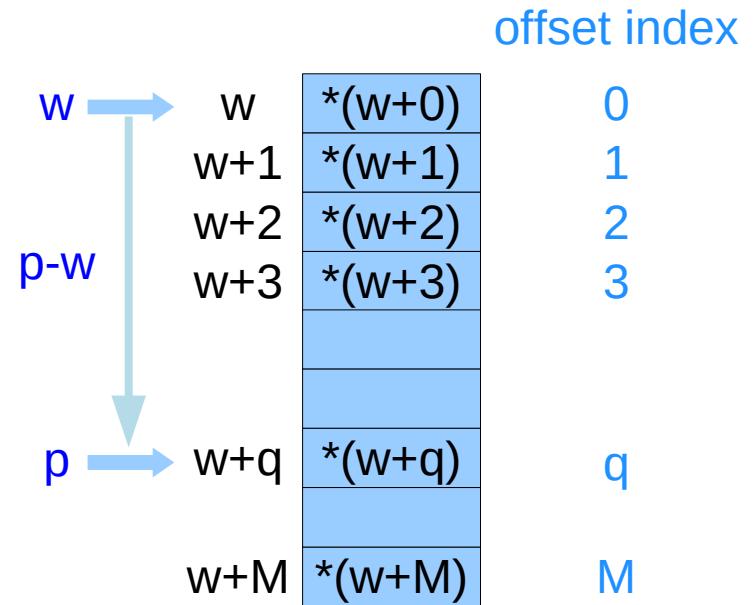
modulo (M+1)

$$w \leq p \leq w+M \quad 0 \leq q \leq M$$

$$\begin{aligned} p = w + q & \quad *p = * (w + q) = w[q] \\ * (p+i) &= * (w + q + i) = w[q+i] \end{aligned}$$

$$p+i = w + (p-w+i)$$

$$\begin{aligned} * (p+i) &= * (w + (p-w+i)) \\ &= * (w + (q+i)) \end{aligned}$$

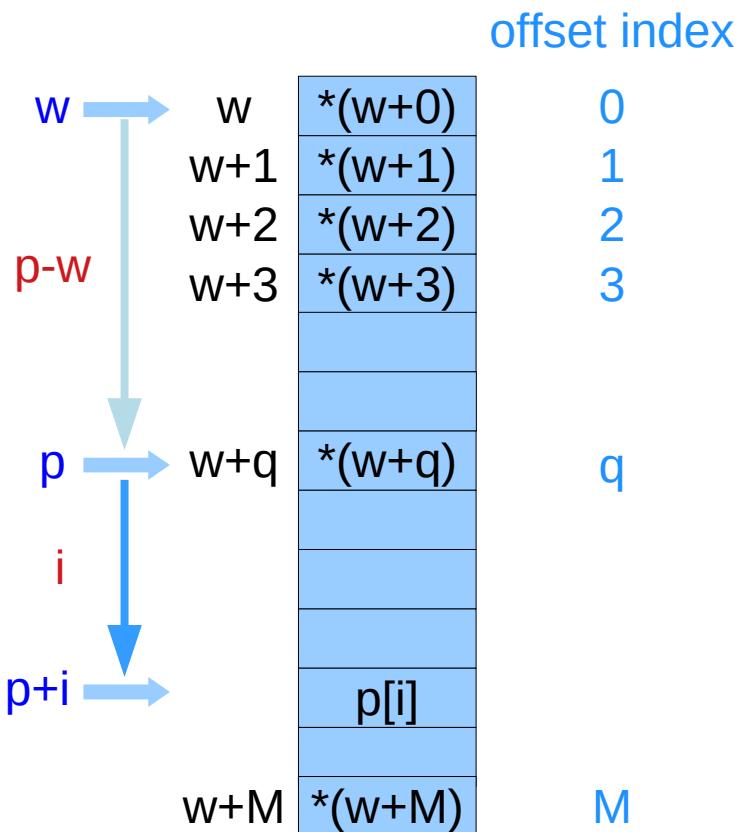


$$w[(p-w+i)\% (M+1)] = w[(q+i)\% (M+1)]$$

# Accessing $w$ buffer by the pointer $p$

modulo  $(M+1)$

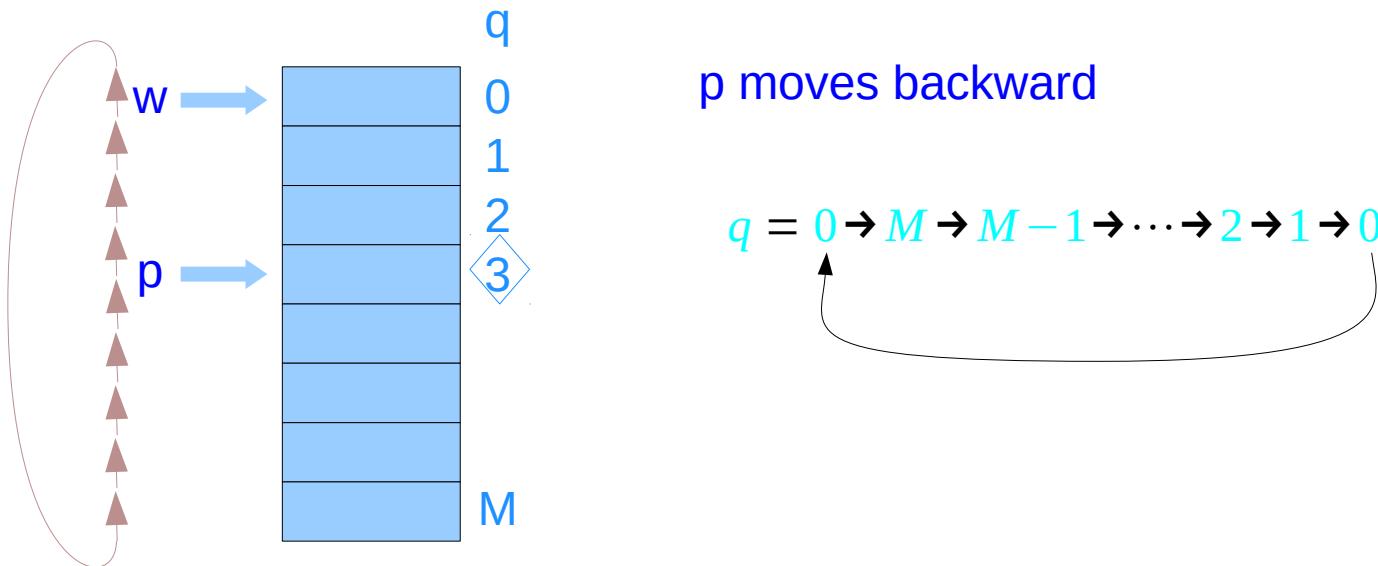
$$w[(p-w+i)\%(M+1)] = w[(q+i)\%(M+1)]$$



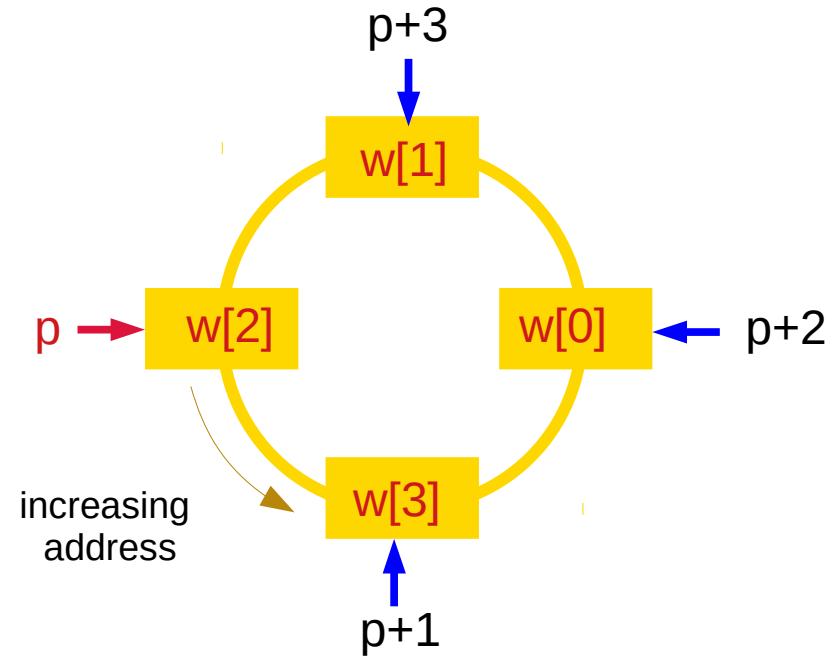
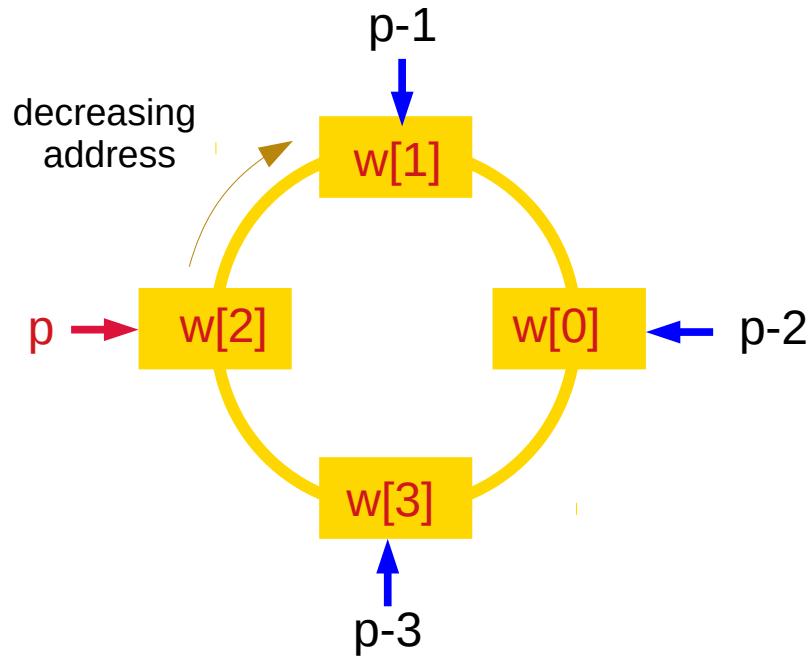
# Address pointer **p** moves backward

modulo ( $M+1$ )

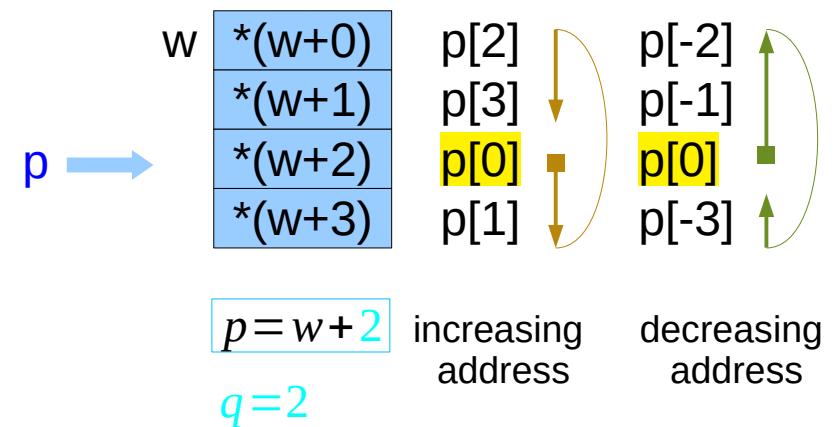
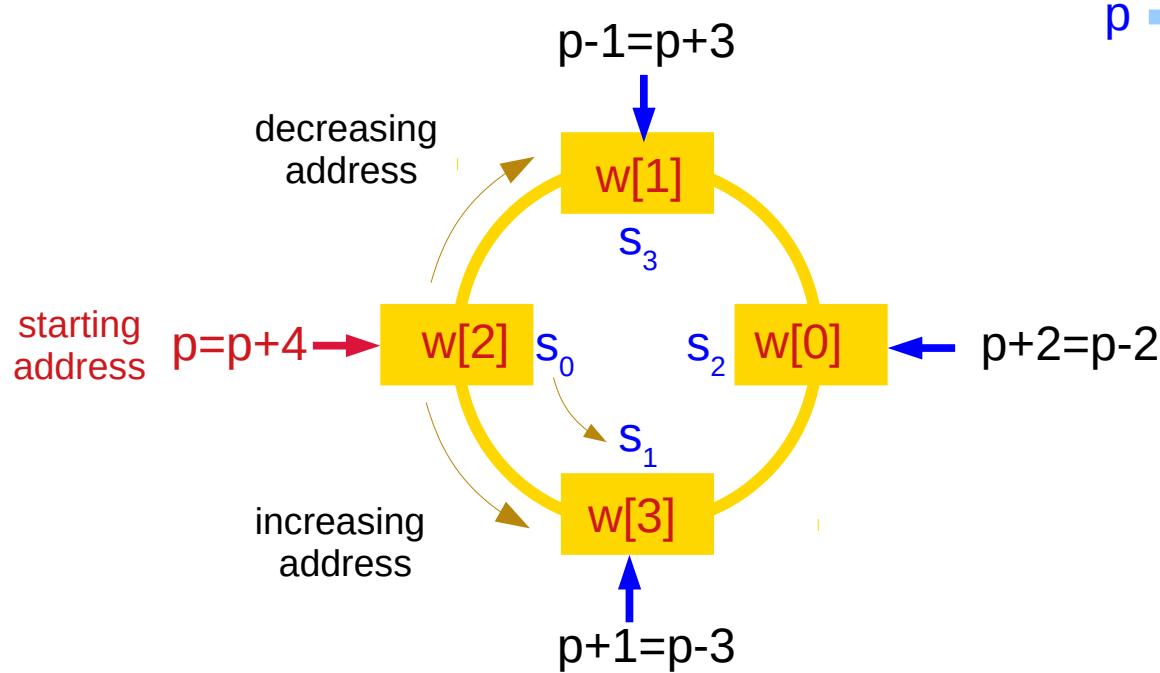
$$w[(p-w+i)\%(M+1)] = w[(q+i)\%(M+1)]$$



# Increasing and decreasing the pointer $p$



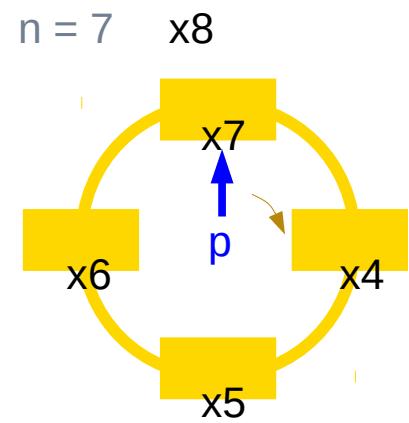
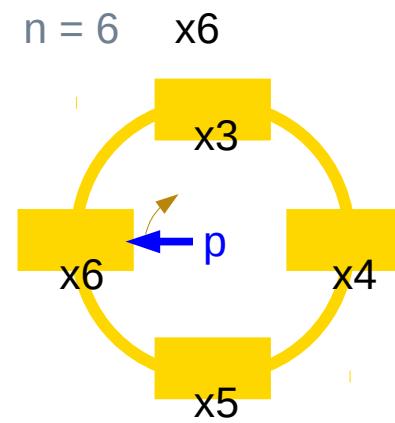
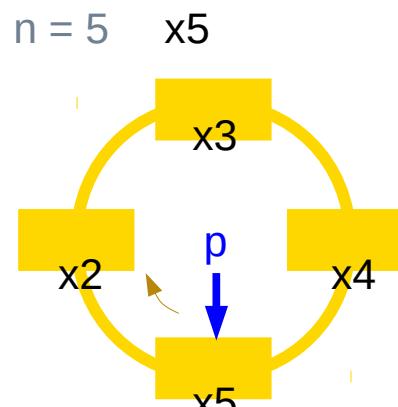
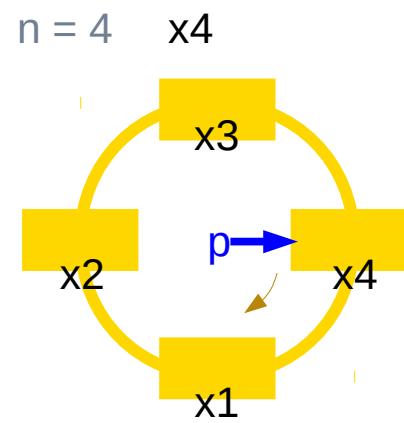
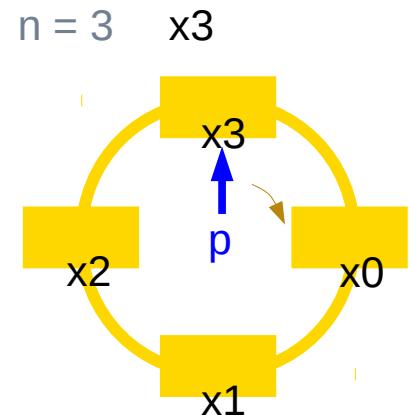
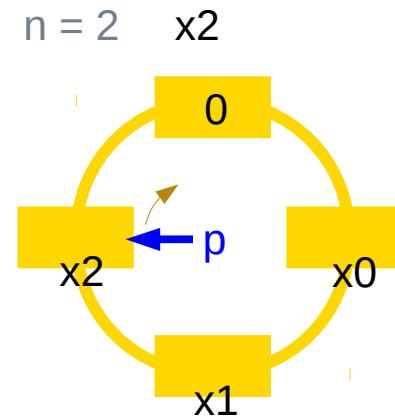
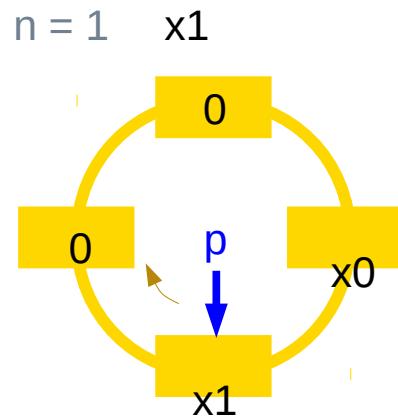
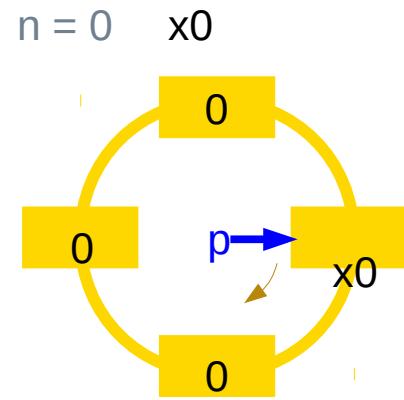
# Internal States



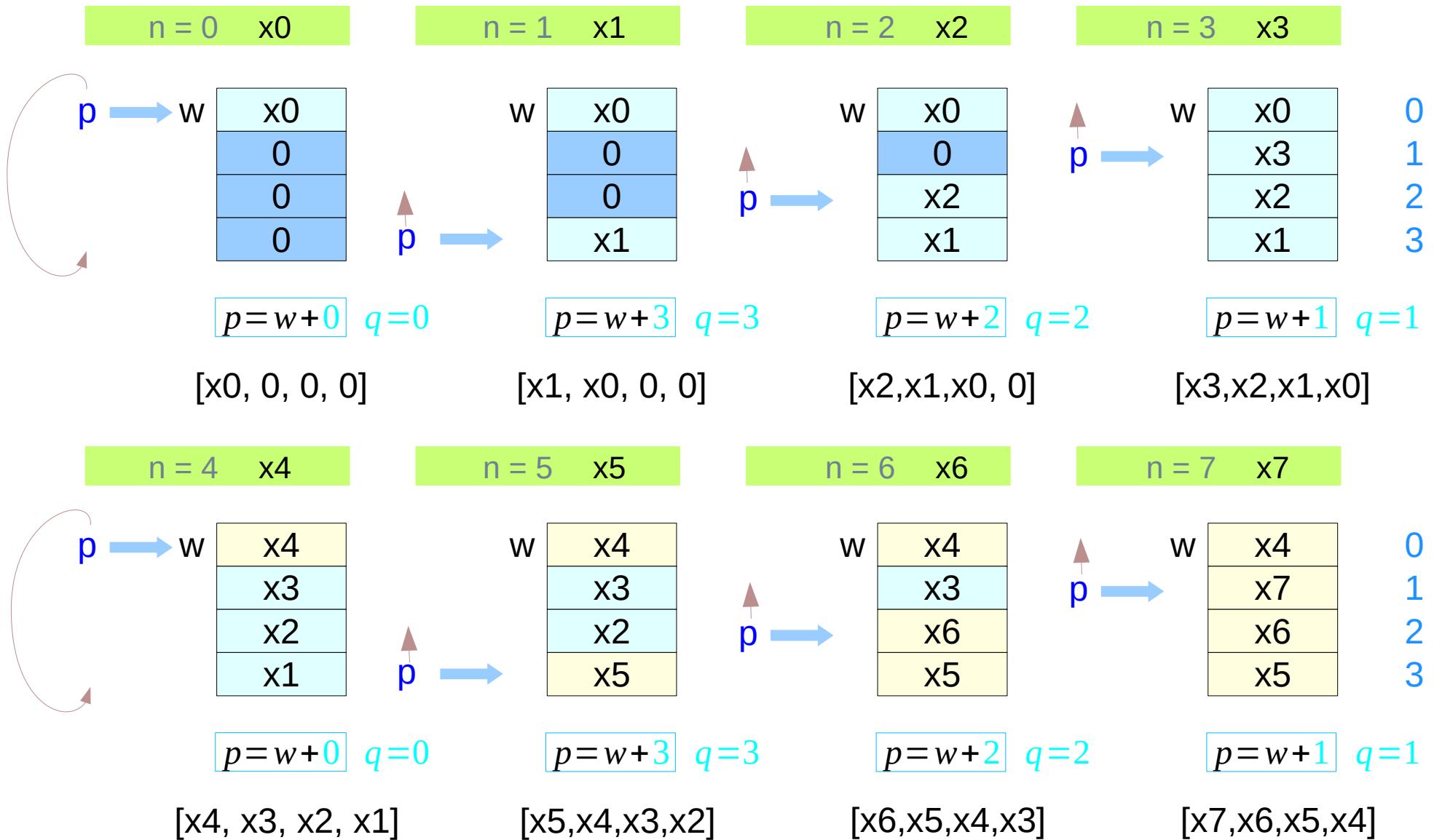
the state vector

$$\begin{pmatrix} s_0 \\ s_1 \\ s_2 \\ s_3 \end{pmatrix} \begin{pmatrix} p[0] \\ p[1] \\ p[2] \\ p[3] \end{pmatrix} \begin{pmatrix} w[2] \\ w[3] \\ w[0] \\ w[1] \end{pmatrix}$$

# Address pointer $p$ and incoming inputs



# Address pointer p and incoming inputs – linear array view



# Internal states over the first 8 steps of n

n	q	[w0]	[w1]	[w2]	[w3]	s0	s1	s2	s3	s0	s1	s2	s3
0	0	x0	0	0	0	[w0]	[w1]	[w2]	[w3]	x0	0	0	0
1	3	x0	0	0	x1	[w3]	[w0]	[w1]	[w2]	x1	x0	0	0
2	2	x0	0	x2	x1	[w2]	[w3]	[w0]	[w1]	x2	x1	x0	0
3	1	x0	x3	x2	x1	[w1]	[w2]	[w3]	[w0]	x3	x2	x1	x0
4	0	x4	x3	x2	x1	[w0]	[w1]	[w2]	[w3]	x4	x3	x2	x1
5	3	x4	x3	x2	x5	[w3]	[w0]	[w1]	[w2]	x5	x4	x3	x2
6	2	x4	x3	x6	x5	[w2]	[w3]	[w0]	[w1]	x6	x5	x4	x3
7	1	x4	x7	x6	x5	[w1]	[w2]	[w3]	[w0]	x7	x6	x5	x4

[w0] value at the buffer w0

[w1] value at the buffer w1

[w2] value at the buffer w2

[w3] value at the buffer w3

$$\begin{pmatrix} s_0 \\ s_1 \\ s_2 \\ s_3 \end{pmatrix} \begin{pmatrix} p[0] \\ p[1] \\ p[2] \\ p[3] \end{pmatrix} \begin{pmatrix} w[2] \\ w[3] \\ w[0] \\ w[1] \end{pmatrix}$$

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