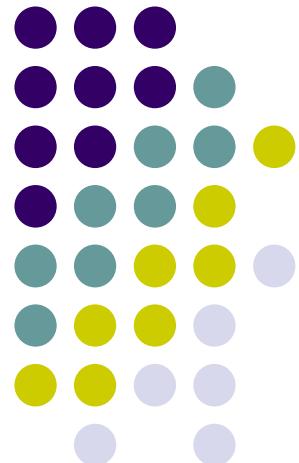


# Digital Signal Processing

*College of Communication & Information Engineering  
Nanjing University of Posts and Telecommunications*

*Fall Semester, 2019*

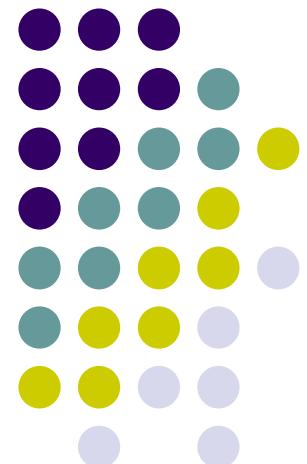
JI Wei



# Chapter 5

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## Finite-Length Discrete Transforms



# Part B: Operations on Finite-Length Sequences and DFT Properties

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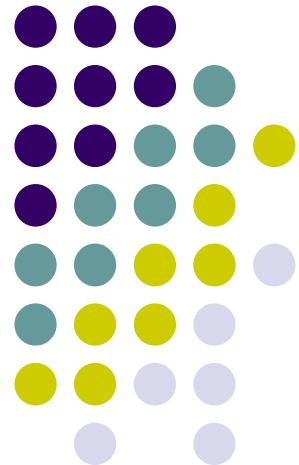


- Circular Time-reversal of a Sequence (Section 2.3.1 )
- Circular Shift of a Sequence (Section 2.3.2 and 5.7 )
- Circular Convolution (Section 5.4 and 5.7 )
- Classification of Finite-Length Sequences ( Section 5.5)
- DFT Symmetry Relations and Theorems (Section 5.6 and 5.7)
- Fourier-Domain Filtering (Section 5.8)
- Computation of the DFT of Real Sequences (Section 5.9)
- Linear Convolution Using the DFT( Section 5.10)

# 1、Circular Time-Reversal Operation

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See in textbook Section 2.3.1





# Modulo Operation(取模运算)

- The time-reversal operation on a finite-length sequence is obtained by the **modulo operation**

$$\langle m \rangle_N = m \text{ modulo } N$$

where  $m$  and  $N$  is any integer, and  $0, 1, \dots, N-1$  be a set of  $N$  positive integers

- $r = \langle m \rangle_N$  is called the **residue(余数)**, which is an integer with a value between 0 and  $N-1$   
 $r = m + lN$ , where  $l$  is a positive or negative integer chosen to make  $m + lN$  an integer between 0 and  $N-1$



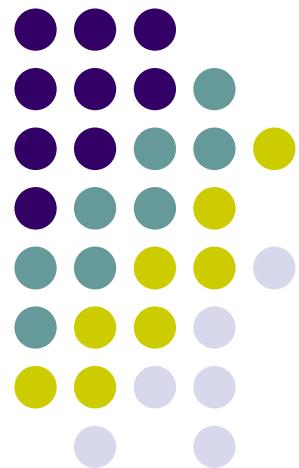
# 1、Circular Time-Reversal Operation

- The **circular time-reversal** version  $\{y[n]\}$  of a length- $N$  sequence  $\{x[n]\}$  defined for  $0 \leq n \leq N-1$  is given by  $\{y[n]\} = \{x[<-n>_N]\}$

## 2. Circular Shift of a Sequence

---

See in Section 2.3.2 and 5.7





## 2、Circular Shift of a Sequence

- The desired shift, called the *circular shift*, is defined using a modulo operation:

$$x_c(n) = x\left(\langle n - n_0 \rangle_N\right)$$

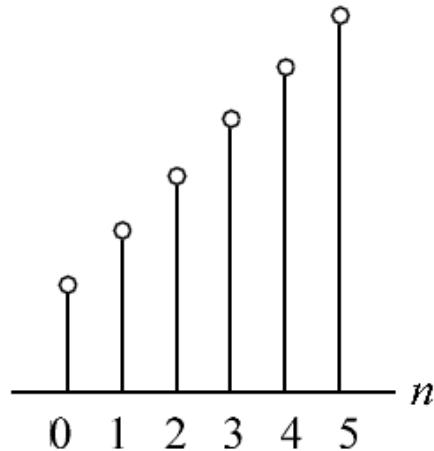
- For  $n_0 > 0$  (*right circular shift*), the above equation implies

$$x_c(n) = \begin{cases} x(n - n_0), & \text{for } n_0 \leq n \leq N - 1 \\ x(N - n_0 + n), & \text{for } 0 \leq n < n_0 \end{cases}$$

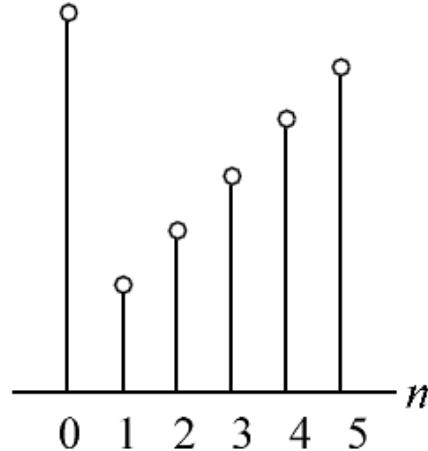


## 2. Circular Shift of a Sequence

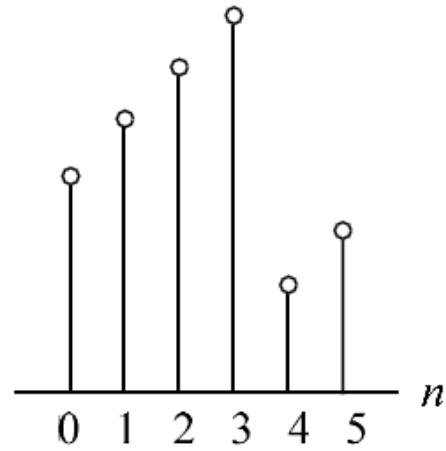
Illustration of the concept of a circular shift



$$x(n)$$

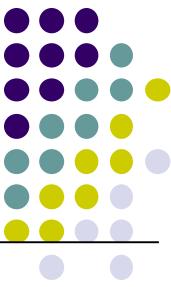


$$x(\langle n-1 \rangle_6) = x(\langle n+5 \rangle_6)$$



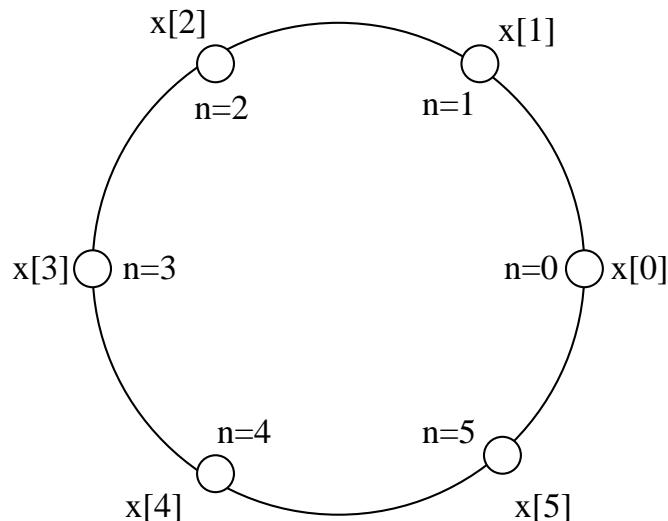
$$x(\langle n-4 \rangle_6) = x(\langle n+2 \rangle_6)$$

- A right circular shift by  $n_0$  is equivalent to a left circular shift by  $N-n_0$  sample periods.

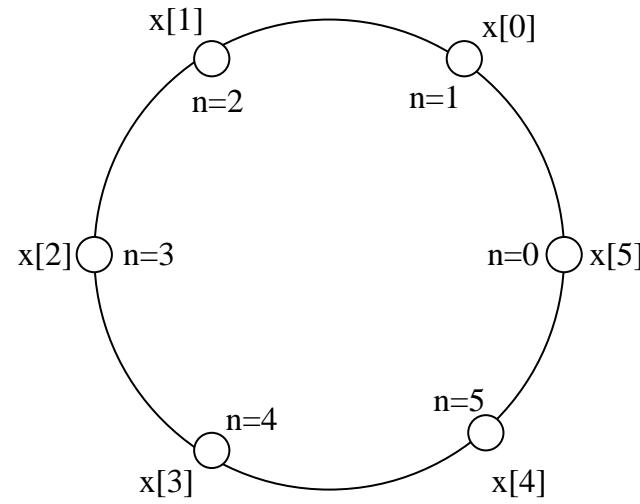


## 2. Circular Shift of a Sequence

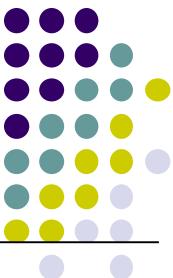
- The length- $N$  sequence is displayed on a circle at  $N$  equally spaced points
- The circular shift operation can be viewed as a **clockwise or anti-clockwise** rotation of the sequence by  $n_0$  sample spacings



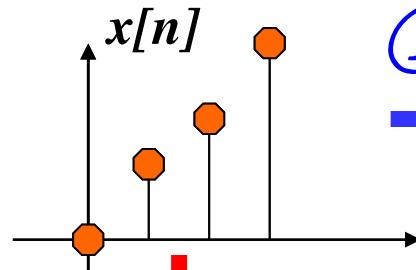
(a)  $x[n]$



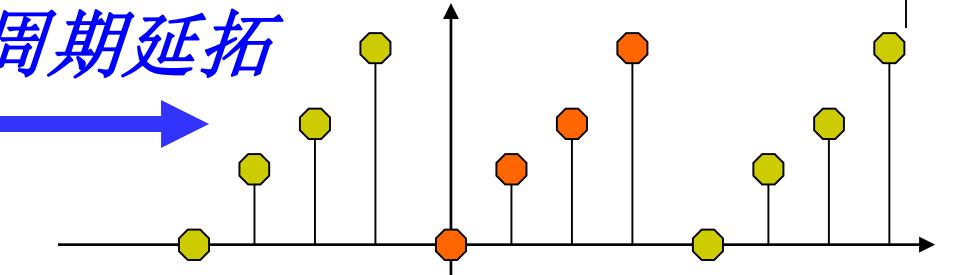
(b)  $x[<n-1>_6] = x[<n+5>_6]$



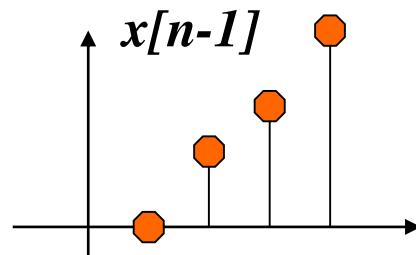
# Compare the shift and circular shift



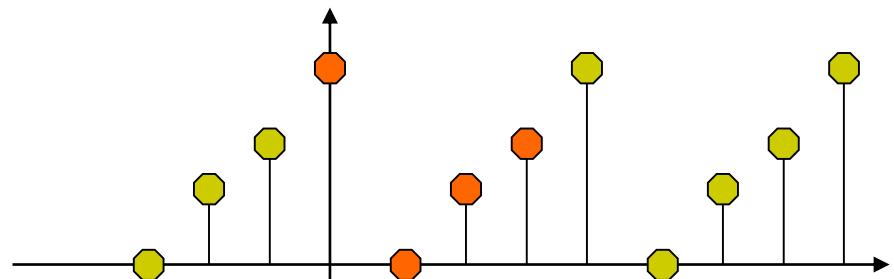
① 周期延拓



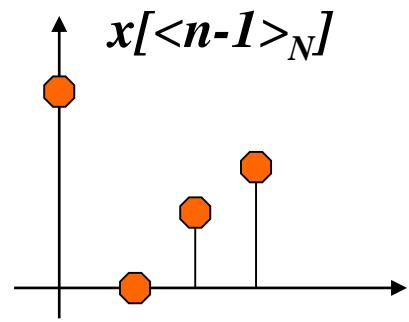
shift



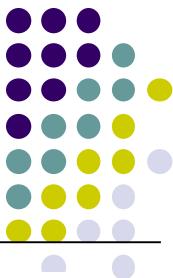
② shift



circular  
shift



③ 取主值区间  $0 \sim N-1$



## 2、Circular Shift of a Sequence

- DFT of the circular shift sequence

$$y(n) = x\left(\langle n+m \rangle_N\right) R_N\left(\langle n+m \rangle_N\right)$$

$$Y(k) = DFT[y(n)]$$

$$= \sum_{n=0}^{N-1} x\left(\langle n+m \rangle_N\right) R_N(n) W_N^{kn}$$

$$= \sum_{n=0}^{N-1} x\left(\langle n+m \rangle_N\right) W_N^{kn}$$



## 2、Circular Shift of a Sequence

- DFT of the circular shift sequence

$$y(n) = x\left(\langle n+m \rangle_N\right) R_N\left(\langle n+m \rangle_N\right)$$

$$Y(k) = DFT[y(n)]$$

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$$= \sum_{n=0}^{N-1} x\left(\langle n+m \rangle_N\right) W_N^{kn}$$



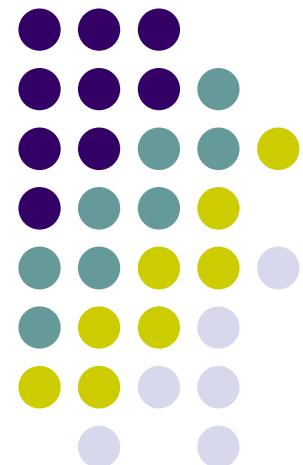
## 2、Circular Shift of a Sequence

$$\begin{aligned} Y(k) &= \sum_{n'=m}^{N-1+m} x(\langle n' \rangle_N) W_N^{k(n'-m)} \\ &= W_N^{-km} \sum_{n'=m}^{N-1+m} x(\langle n' \rangle_N) W_N^{kn'} \\ &= W_N^{-km} \left( \sum_{n'=0}^{N-1} (\cdot) - \underbrace{\sum_{n'=0}^{m-1} (\cdot)}_{\text{cancel}} + \sum_{n'=N}^{N-1+m} (\cdot) \right) \\ &= W_N^{-km} \sum_{n'=0}^{N-1} \underbrace{x(\langle n' \rangle_N) W_N^{kn'}}_{\text{cancel}} \\ &= W_N^{-km} \sum_{n'=0}^{N-1} x(n') W_N^{kn'} = W_N^{-km} X(k) \end{aligned}$$

### 3. Circular Convolution

---

See in Section 5.4 and 5.7





### 3. Circular Convolution

- To develop a convolution-like operation resulting in a length- $N$  sequence  $y_C(n)$ , we need to define a **circular time-reversal**, and then apply **a circular time-shift**.
- Resulting operation, called a ***circular convolution***, is defined by

$$y_C(n) = \sum_{m=0}^{N-1} g(m)h\left(\langle n-m \rangle_N\right), \quad 0 \leq n \leq N-1$$



### 3、Circular Convolution

- Since the operation defined involves two length- $N$  sequences, it is often referred to as an  $N$ -point circular convolution, denoted as

$$y_C(n) = g(n) \circledast h(n)$$

- The circular convolution is commutative, i.e.

$$g(n) \circledast h(n) = h(n) \circledast g(n)$$

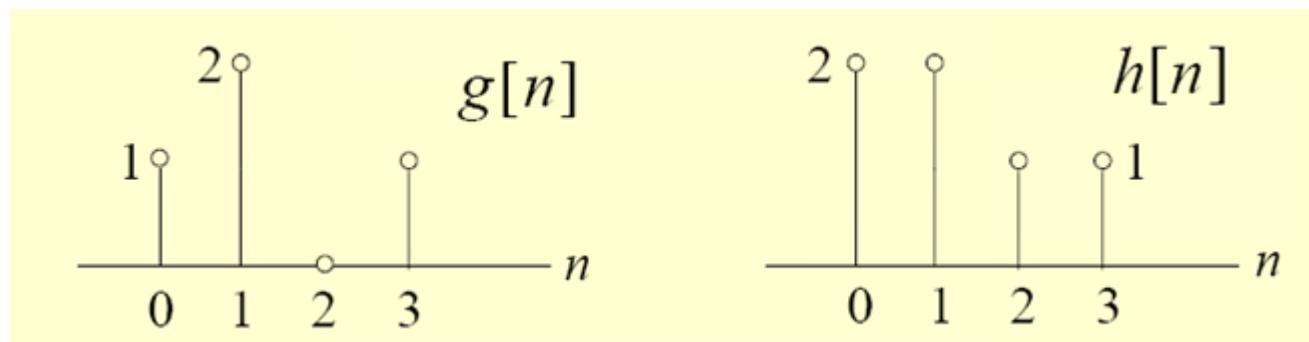


### 3. Circular Convolution

- **Example** Determine the 4-point circular convolution of the two length-4 sequences:

$$g[n] = \{1 \ 2 \ 0 \ 1\}, \quad h[n] = \{2 \ 2 \ 1 \ 1\}$$

as sketched below





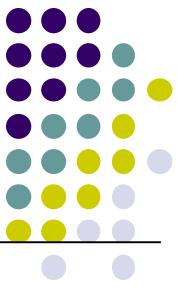
### 3、Circular Convolution

- The result is a length-4 sequence  $y_C[n]$

$$y_C(n) = g(n) \circledast h(n) = \sum_{m=0}^{N-1} g(m)h(\langle n-m \rangle_N), 0 \leq n \leq N-1;$$

- From the above we observe

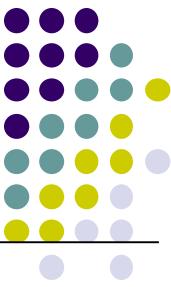
$$\begin{aligned} y_C(0) &= \sum_{m=0}^3 g(m)h(\langle -m \rangle_4) \\ &= g[0]h[0] + g[1]h[3] + g[2]h[2] + g[3]h[1] \\ &= (1 \times 2) + (2 \times 1) + (0 \times 1) + (1 \times 2) = 6 \end{aligned}$$



### 3、Circular Convolution

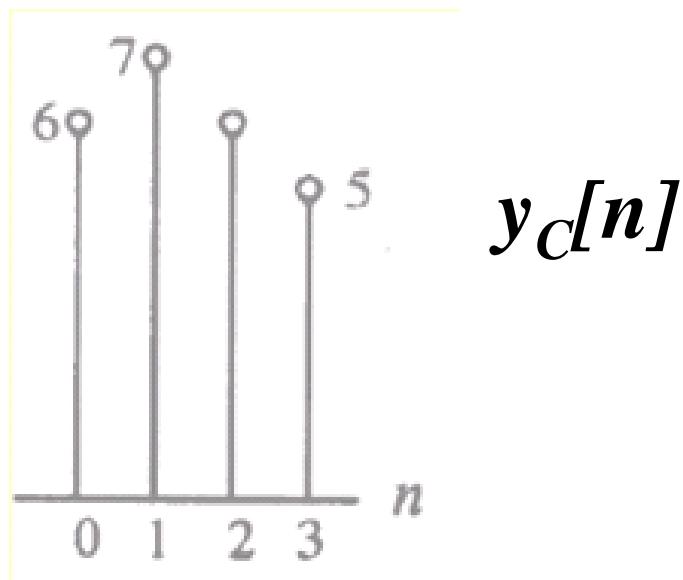
- Likewise,  $y_C(1) = \sum_{m=0}^3 g(m)h(\langle 1-m \rangle_4)$   
 $= g[0]h[1] + g[1]h[0] + g[2]h[3] + g[3]h[2]$   
 $= (1 \times 2) + (2 \times 2) + (0 \times 1) + (1 \times 1) = 7$

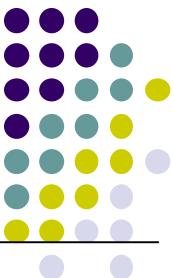
$$\begin{aligned}y_C(2) &= \sum_{m=0}^3 g(m)h(\langle 2-m \rangle_4) \\&= g[0]h[2] + g[1]h[1] + g[2]h[0] + g[3]h[3] \\&= (1 \times 1) + (2 \times 2) + (0 \times 2) + (1 \times 1) = 6\end{aligned}$$



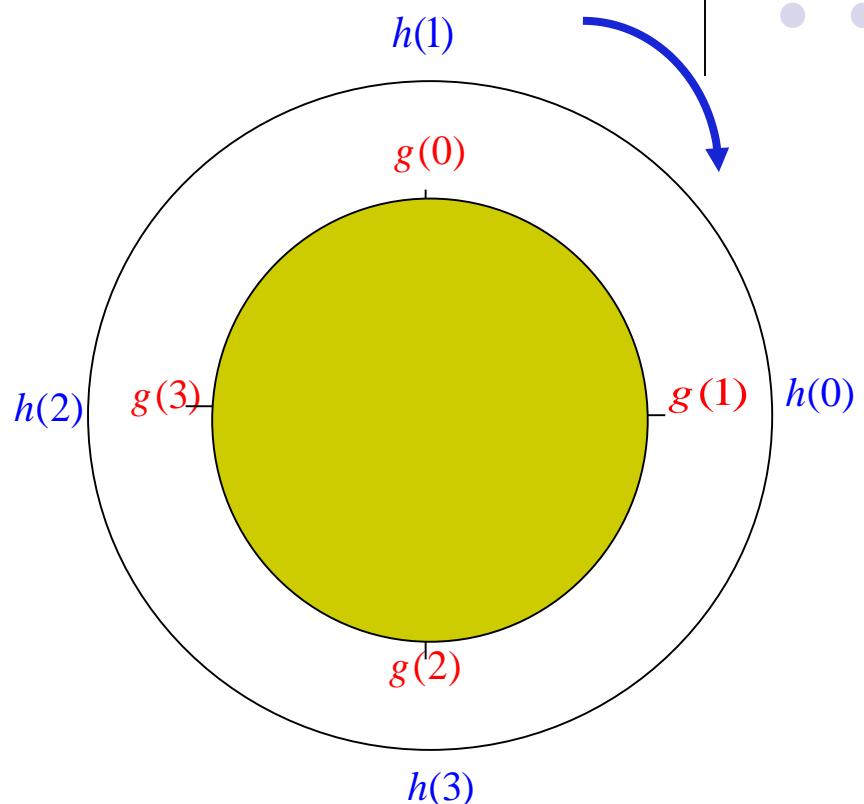
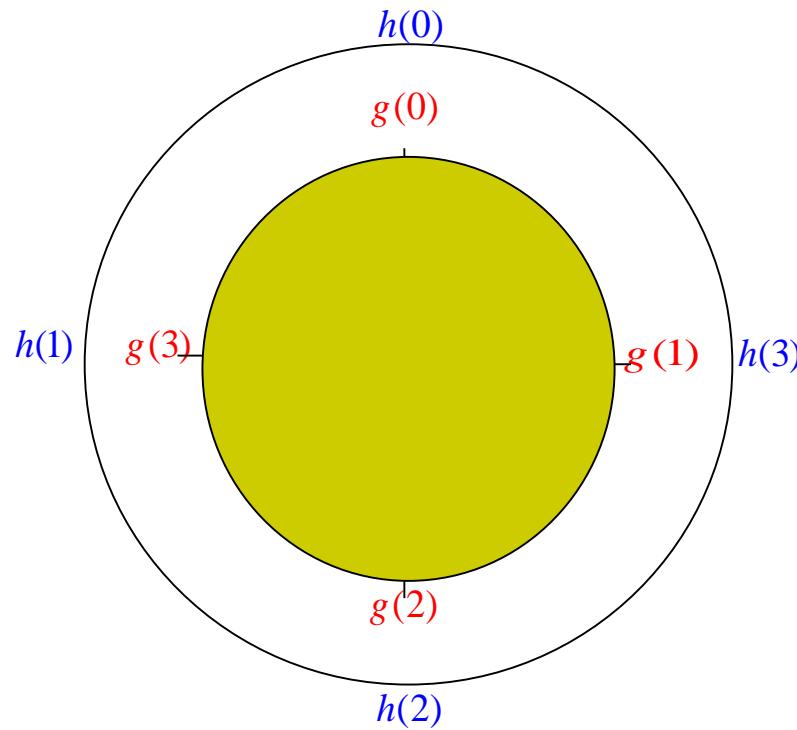
### 3、Circular Convolution

$$\begin{aligned}y_C(3) &= \sum_{m=0}^3 g(m)h\left(\langle 3-m \rangle_4\right) \\&= g[0]h[3] + g[1]h[2] + g[2]h[1] + g[3]h[0] \\&= (1 \times 1) + (2 \times 1) + (0 \times 2) + (1 \times 2) = 5\end{aligned}$$





# 循环卷积过程图解





### 3. Circular Convolution

- The circular convolution can also be computed using a DFT-based approach
- The N-point circular convolution can be written in matrix form as

$$\begin{bmatrix} y_C[0] \\ y_C[1] \\ y_C[2] \\ \vdots \\ y_C[N-1] \end{bmatrix} = \begin{bmatrix} h[0] & h[N-1] & h[N-2] & \cdots & h[1] \\ h[1] & h[0] & h[N-1] & \cdots & h[2] \\ h[2] & h[1] & h[0] & \cdots & h[3] \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ h[N-1] & h[N-2] & h[N-3] & \cdots & h[0] \end{bmatrix} \begin{bmatrix} g[0] \\ g[1] \\ g[2] \\ \vdots \\ g[N-1] \end{bmatrix}$$



### 3、Circular Convolution

$$\begin{bmatrix} y_C[0] \\ y_C[1] \\ y_C[2] \\ \vdots \\ y_C[N-1] \end{bmatrix} = \begin{bmatrix} h[0] & h[N-1] & h[N-2] & \cdots & h[1] \\ h[1] & h[0] & h[N-1] & \cdots & h[2] \\ h[2] & h[1] & h[0] & \cdots & h[3] \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ h[N-1] & h[N-2] & h[N-3] & \cdots & h[0] \end{bmatrix} \begin{bmatrix} g[0] \\ g[1] \\ g[2] \\ \vdots \\ g[N-1] \end{bmatrix}$$

Note: 1、The element in each row of the matrix are obtained by circularly rotating the elements of the previous row to the right by one position. Such a matrix is called a circulant matrix(轮换矩阵、循环行列式矩阵)

2、使用矩阵形式计算循环卷积前，需要通过补零把参与循环卷积的两个输入序列扩充成相同长度，且此长度等于DFT的点数

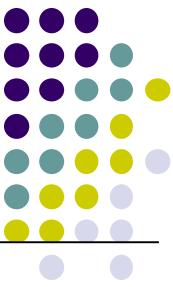


### 3、Circular Convolution

- **Example** Now let us extend the two length-4 sequences to length 7 by appending each with three zero-valued samples, i.e.,

$$g_e[n] = \begin{cases} g[n], & 0 \leq n \leq 3 \\ 0, & 4 \leq n \leq 6 \end{cases}$$

$$h_e[n] = \begin{cases} h[n], & 0 \leq n \leq 3 \\ 0, & 4 \leq n \leq 6 \end{cases}$$



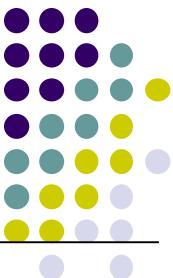
### 3. Circular Convolution

$$g_e(n) = \{1, 2, 0, 1, 0, 0, 0\} \quad 0 \leq n \leq 6$$

$$h_e(n) = \{2, 2, 1, 1, 0, 0, 0\} \quad 0 \leq n \leq 6$$

- We next determine the 7-point circular convolution of  $g_e[n]$  and  $h_e[n]$ :

$$y_C(n) = \sum_{m=0}^{6} g_e(m)h_e(\langle n-m \rangle_7), 0 \leq n \leq 6$$



## Matrix method:

$$Y = \begin{pmatrix} y(0) \\ y(1) \\ y(2) \\ \vdots \\ y(N-1) \end{pmatrix} = H_e G_e$$

$$G_e = \begin{pmatrix} g_e(0) \\ g_e(1) \\ g_e(2) \\ \vdots \\ g_e(N-1) \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}$$

$$H_e = \begin{pmatrix} h_e(0) & h_e(N-1) & h_e(N-2) \cdots h_e(1) \\ h_e(1) & h_e(0) & h_e(N-1) \cdots h_e(2) \\ h_e(2) & h_e(1) & h_e(0) & \cdots h_e(3) \\ \vdots & \vdots & \vdots & \vdots \\ h_e(N-1) & h_e(N-2) & h_e(N-3) & \cdots h_e(0) \end{pmatrix} = \begin{pmatrix} 2 & 0 & 0 & 0 & 1 & 1 & 2 \\ 2 & 2 & 0 & 0 & 0 & 1 & 1 \\ 1 & 2 & 2 & 0 & 0 & 0 & 1 \\ 1 & 1 & 2 & 2 & 0 & 0 & 0 \\ 0 & 1 & 1 & 2 & 2 & 0 & 0 \\ 0 & 0 & 1 & 1 & 2 & 2 & 0 \\ 0 & 0 & 0 & 1 & 1 & 2 & 2 \end{pmatrix}$$

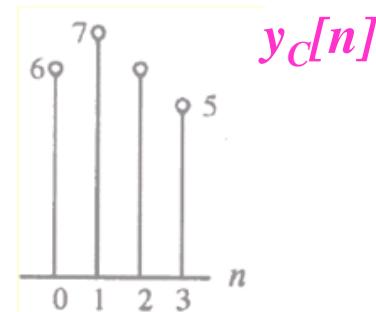
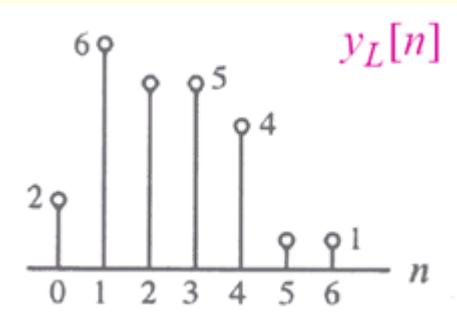


$$\therefore \begin{pmatrix} y(0) \\ y(1) \\ y(2) \\ y(3) \\ y(4) \\ y(5) \\ y(6) \end{pmatrix} = \begin{pmatrix} 2 & 0 & 0 & 0 & 1 & 1 & 2 \\ 2 & 2 & 0 & 0 & 0 & 1 & 1 \\ 1 & 2 & 2 & 0 & 0 & 0 & 1 \\ 1 & 1 & 2 & 2 & 0 & 0 & 0 \\ 0 & 1 & 1 & 2 & 2 & 0 & 0 \\ 0 & 0 & 1 & 1 & 2 & 2 & 0 \\ 0 & 0 & 0 & 1 & 1 & 2 & 2 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 6 \\ 5 \\ 5 \\ 4 \\ 1 \\ 1 \end{pmatrix}$$



### 3. Circular Convolution

- As can be seen from the above that  $y[n]$  is precisely the sequence  $y_L[n]$  obtained by a linear convolution of  $g[n]$  and  $h[n]$



- Try to think:  
What is the relation between the circular convolution and the linear convolution?



# 线性卷积的运算过程

假设 $g(n)$ 和 $h(n)$ 为有限长序列，长度分别为 $L$ 和 $K$ ，那么它们的线性卷积

$$y(n) = g(n) * h(n) = \sum_{m=-\infty}^{\infty} g(m)h(n-m)$$

是一个长度为 $L+K-1$ 的有限长序列。

注意：对卷积序列的长度没有严格要求

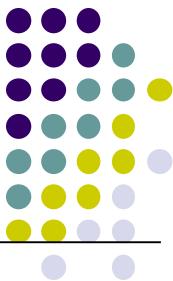
卷积过程：反转平移法



序列  $x(n)=\{1,2,3\}$ ,  $h(n)=\{1,2\}$

求两个序列的线性卷积  $y(n)=x(n)*h(n)$

$x(n)$	1	2	3	
$h(n)$	1	2		
反转 $h(-n)$	2	1		$y(0)=1$
平移 $h(1-n)$	2	1		$y(1)=4$
平移 $h(2-n)$	2	1		$y(2)=7$
平移 $h(3-n)$	2	1		$y(3)=6$



# 周期卷积 periodic convolution

- 周期卷积 (periodic convolution) 定义 (P234)

$$\tilde{y}(n) = \sum_{r=0}^{N-1} \tilde{x}[r] \tilde{h}[n-r]$$

- 注意：三个序列均为周期序列，周期为N
- 卷积过程：反转平移法，过程同线性卷积



序列  $x(n)=\{1,2,3\}$ ,  $h(n)=\{1,2,0\}$  周期  $N=3$

# 求两个序列的周期卷积g(n)

$x'(n)$	1	2	3	1	2	3	1	2	3
$h'(n)$	1	2	0	1	2	0	1	2	0
反转 $h'(-n)$	1	0	2	1	0	2	1	0	2
平移 $h'(1-n)$	2	1	0	2	1	0	2	1	0
平移 $h'(2-n)$	0	2	1	0	2	1	0	2	1



# 借助周期卷积求循环卷积

重新审视循环卷积的过程，所有的运算都是在一个主值区间里完成，根据DFT运算的周期性，可以借助周期卷积来实现循环卷积运算：

1) 由有限长序 $g(n)$ 和 $h(n)$ 构造周期序列  $\tilde{g}(n)$   $\tilde{h}(n)$

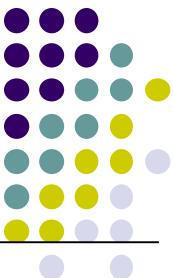
2) 计算周期卷积： 
$$\tilde{y}(n) = \sum_{m=0}^{N-1} \tilde{g}(m)\tilde{h}(n-m)$$

3) 卷积结果取主值：  $y_c(n) = \tilde{y}(n)R_N(n)$



序列  $x(n)=\{1,2,3\}$ ,  $h(n)=\{1,2\}$   $N=3$

求两个序列的循环卷积 $y_1(n)$

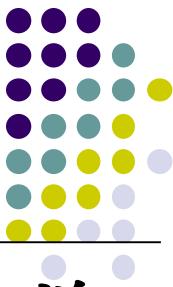


序列  $x(n)=\{1,2,3\}$ ,  $h(n)=\{1,2\}$   $N=4$

求两个序列的循环卷积  $y_2(n)$

	$x'(n)$	2	3	0	1	2	3	0	1	
	$h'(n)$	2	0	0	1	2	0	0	1	
反转	$h'(-n)$	0	0	2	1	0	0	2	1	$y_2(0)=1$
平移	$h'(1-n)$	1	0	0	2	1	0	0	2	$y_2(1)=4$
平移	$h'(2-n)$	2	1	0	0	2	1	0	0	$y_2(2)=7$
平移	$h'(3-n)$	0	2	1	0	0	2	1	0	$y_2(3)=6$

$y_2(n)$  是有限长序列，序列值为  $\{1,4,7,6\}$



## 借助周期卷积求循环卷积

对于重新构造两个有限长序列 $g_e(n)$ 、 $h_e(n)$ 来说，其周期延拓后的序列为：

$$\tilde{g}_e(n) = \sum_{q=-\infty}^{\infty} g_e(n + qN)$$

$$\tilde{h}_e(n) = \sum_{r=-\infty}^{\infty} h_e(n + rN)$$

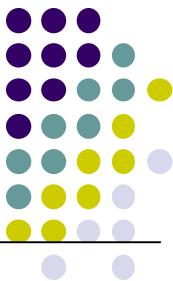


# 借助周期卷积求循环卷积

它们的周期卷积序列为：

$$\begin{aligned}\tilde{y}(n) &= \sum_{m=0}^{N-1} \tilde{g}_e(m) \tilde{h}_e(n-m) = \sum_{m=0}^{N-1} g_e(m) h_e(n-m) \\ &= \sum_{m=0}^{N-1} g_e(m) \sum_{r=-\infty}^{+\infty} h_e(n+rN-m) \\ &= \sum_{r=-\infty}^{\infty} \sum_{m=0}^{N-1} g_e(m) h_e(n+rN-m) \\ &= \sum_{r=-\infty}^{\infty} y(n+rN)\end{aligned}$$

从中可以看出， $g_e(n)$ 、 $h_e(n)$ 周期延拓后的周期卷积是 $g_e(n)$ 、 $h_e(n)$ 线性卷积的周期延拓，周期为 $N$ 。

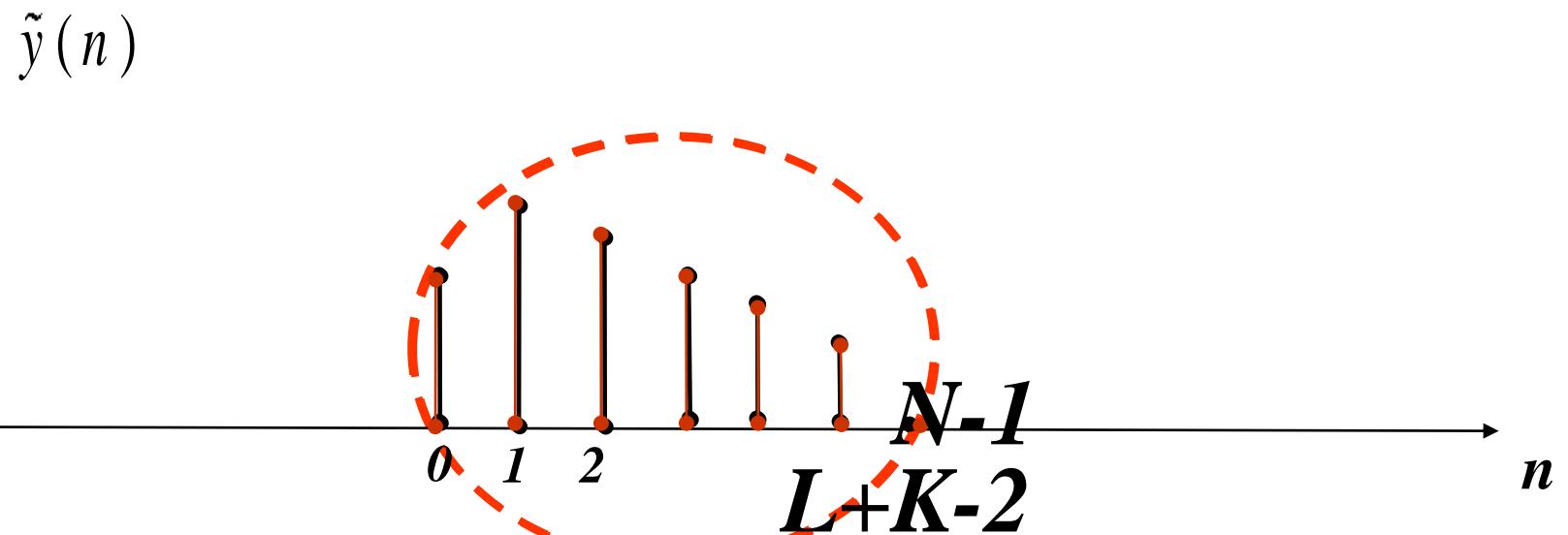
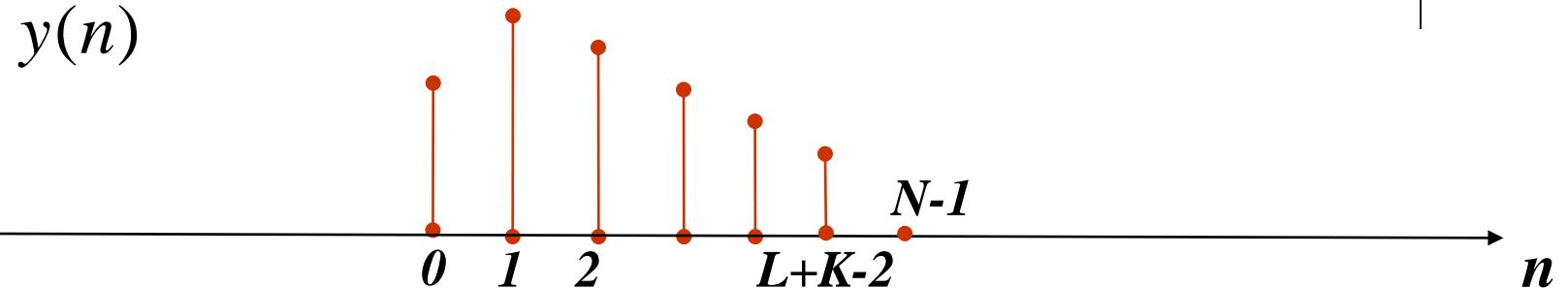


## 借助周期卷积求循环卷积

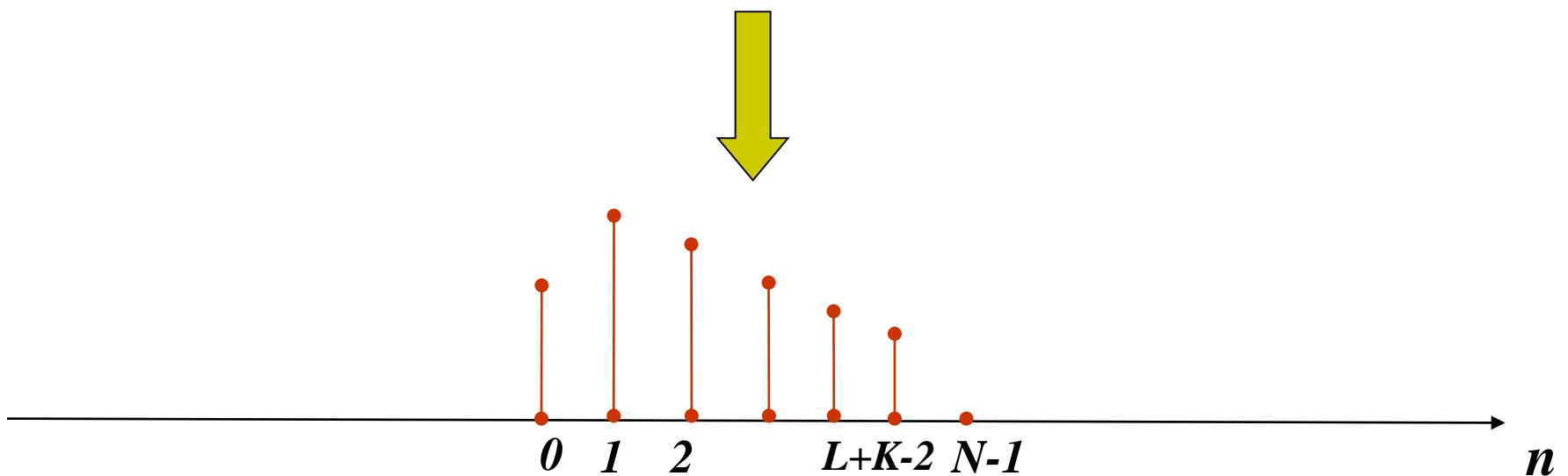
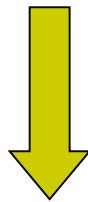
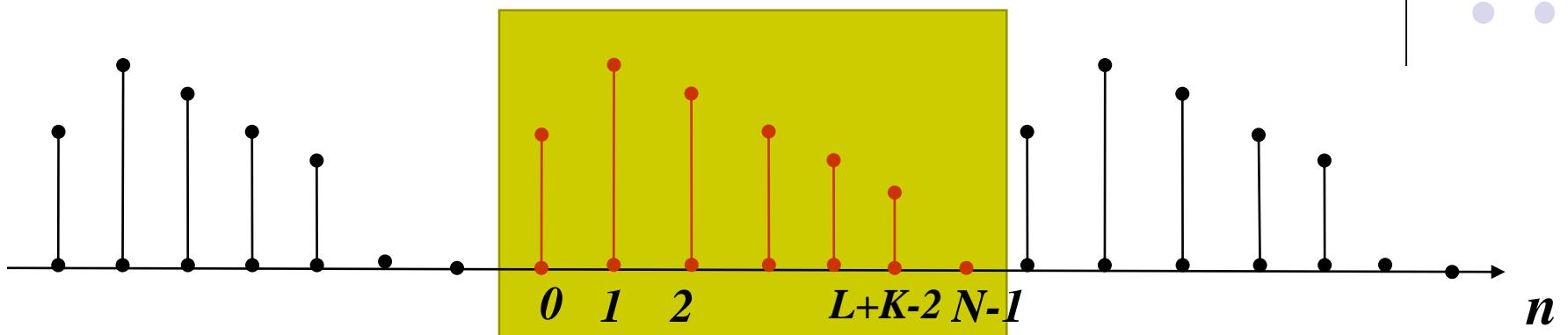
- $y(n)$ 具有 $L+K-1$ 个非零序列值，如果周期卷积的周期 $N < L+K-1$ ，那么， $y(n)$ 周期延拓后，必然有一部分非零序列值要重叠，出现混淆现象。
- 只有 $N \geq L+K-1$ 时，才不会产生交叠，这时 $y(n)$ 的周期延拓中每一个周期 $N$ 内，前 $L+K-1$ 个序列值是 $y(n)$ 的全部非零序列值，而剩下的 $N-(L+K-1)$ 点的序列则是补充的零值。
- 循环卷积正是周期卷积取主值序列：

$$y_c(n) = \tilde{y}(n)R_N(n) = \left[ \sum_{r=-\infty}^{\infty} y(n+rN) \right] R_N(n)$$

$$\tilde{y}(n) = \sum_{r=-\infty}^{\infty} y(n+rN) \quad N \geq L+K-1$$

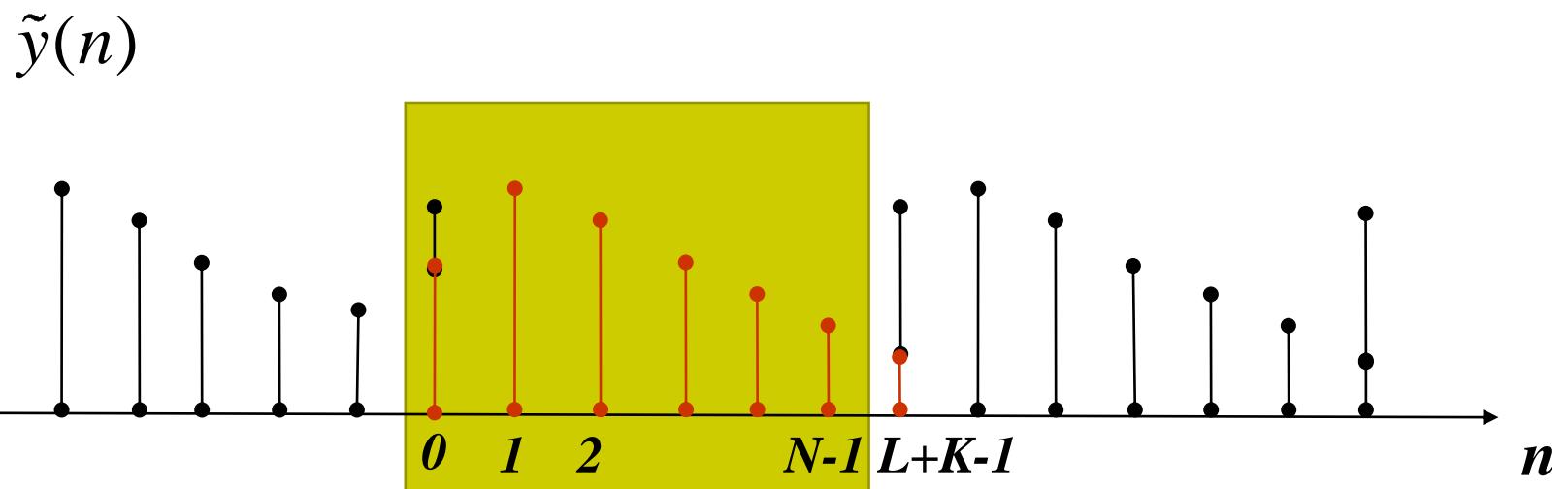
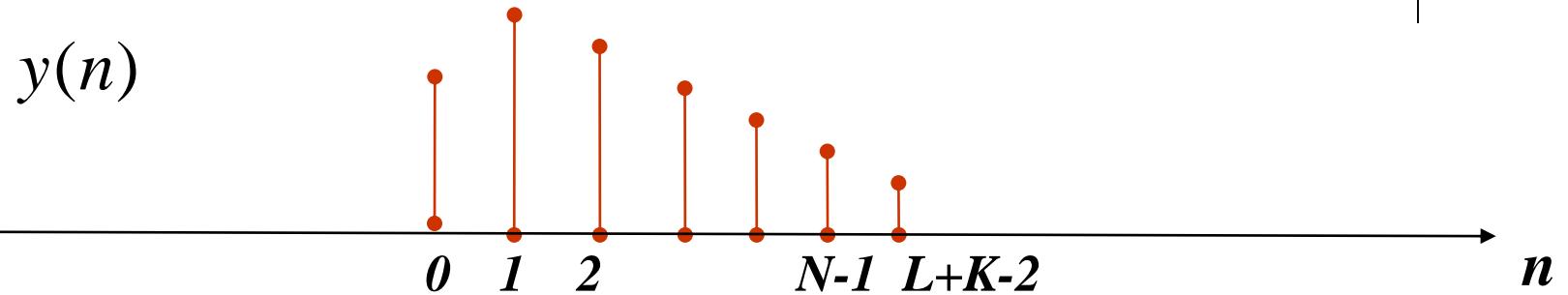
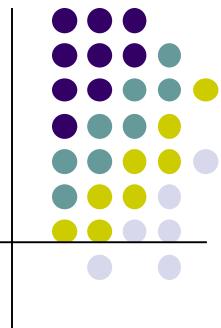


$$y_c(n) = \tilde{y}(n)R_N(n) = \left[ \sum_{r=-\infty}^{\infty} y(n+rN) \right] R_N(n)$$



$$y_c(n) = y(n), \quad 0 \leq n \leq N-1$$

$$\tilde{y}(n) = \sum_{r=-\infty}^{\infty} y(n+rN) \quad N < L + K - 1$$



$$y_c(n) \neq y(n), \quad 0 \leq n \leq N-1$$



# 循环卷积与线性卷积的关系

- 循环卷积等于线性卷积而不产生混淆的必要条件是：

$$N \geq L + K - 1$$



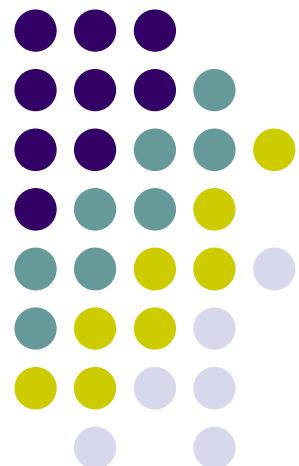
循环卷积的点数

线性卷积结果的长度

循环卷积比起线性卷积，在运算速度上有很大的优越性，可采用快速傅里叶变换（FFT）技术。若能利用循环卷积求线性卷积，会带来很大的方便。

# 4、Classification of Finite-Length Sequences

See in textbook section 5.5





## 4、Classification of Finite-Length Sequences

- **Based on Conjugate Symmetry** (see in Section 5.5.1)

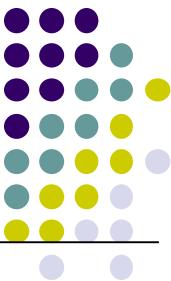
A complex DFT  $X[k]$  can be expressed as a sum of a circular conjugate symmetric part  $X_{cs}[k]$  and a circular conjugate anti-symmetric  $X_{ca}[k]$  part

$$X[k] = X_{cs}[k] + X_{ca}[k], \quad 0 \leq k \leq N-1$$

Where

$$X_{cs}[k] = \frac{1}{2}(X[k] + X^*[-k]) \quad 0 \leq k \leq N-1$$

$$X_{ca}[k] = \frac{1}{2}(X[k] - X^*[-k]) \quad 0 \leq k \leq N-1$$



## 4、Classification of Finite-Length Sequences

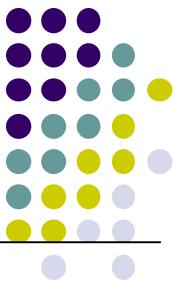
- **Based on Conjugate Symmetry** (see in Section 5.5.1)

- ✓ An  $N$ -point DFT  $X[k]$  is said to be a **circular conjugate-symmetric sequence** if

$$X[k] = X^*[<-k>_N] = X^*[<N-k>_N]$$

- ✓ An  $N$ -point DFT  $X[k]$  is said to be a **circular conjugate-anti-symmetric sequence** if

$$X[k] = -X^*[<-k>_N] = -X^*[<N-k>_N]$$



## 4、Classification of Finite-Length Sequences

- **Based on Geometric Symmetry** (see in Section 5.5.2)

- A length- $N$  *symmetry* sequence  $x(n)$  satisfies the condition

$$x(n) = x(N - 1 - n)$$

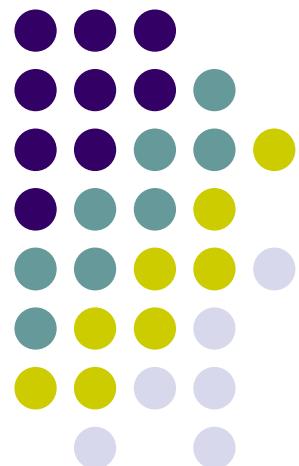
- A length- $N$  *antisymmetry* sequence  $x(n)$  satisfies the condition

$$x(n) = -x(N - 1 - n)$$

# **5、DFT Symmetry Relations and DFT Theorems**

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**See in textbook section 5.6 and 5.7**



# Table 5.1: DFT Properties: Symmetry Relations

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k]$
$x^*[n]$	$X^*[\langle -k \rangle_N]$
$x^*[\langle -n \rangle_N]$	$X^*[k]$
$\text{Re}\{x[n]\}$	$X_{\text{pcs}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] + X^*[\langle -k \rangle_N]\}$
$j \text{Im}\{x[n]\}$	$X_{\text{pca}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] - X^*[\langle -k \rangle_N]\}$
$x_{\text{pcs}}[n]$	$\text{Re}\{X[k]\}$
$x_{\text{pca}}[n]$	$j \text{Im}\{X[k]\}$

Note:  $x_{\text{pcs}}[n]$  and  $x_{\text{pca}}[n]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $x[n]$ , respectively. Likewise,  $X_{\text{pcs}}[k]$  and  $X_{\text{pca}}[k]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $X[k]$ , respectively.

$x[n]$  is a complex sequence



# 用DFT定义证明：

$$DFT(x[n]) = X[k] = \sum_{n=0}^{N-1} x[n] W_N^{nk}, 0 \leq k \leq N-1$$

$$DFT(x^*[n]) = \sum_{n=0}^{N-1} x^*[n] W_N^{nk}$$

$$= \sum_{n=0}^{N-1} (x[n] W_N^{-nk})^* = \left( \sum_{n=0}^{N-1} x[n] W_N^{-nk} \right)^*$$

$$= X^* \left[ \langle -k \rangle_N \right], 0 \leq k \leq N-1$$

# Table 5.1: DFT Properties: Symmetry Relations

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k]$
$x^*[n]$	$X^*[\langle -k \rangle_N]$
$x^*[\langle -n \rangle_N]$	$X^*[k]$
$\text{Re}\{x[n]\}$	$X_{\text{pcs}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] + X^*[\langle -k \rangle_N]\}$
$j \text{Im}\{x[n]\}$	$X_{\text{pca}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] - X^*[\langle -k \rangle_N]\}$
$x_{\text{pcs}}[n]$	$\text{Re}\{X[k]\}$
$x_{\text{pca}}[n]$	$j \text{Im}\{X[k]\}$

Note:  $x_{\text{pcs}}[n]$  and  $x_{\text{pca}}[n]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $x[n]$ , respectively. Likewise,  $X_{\text{pcs}}[k]$  and  $X_{\text{pca}}[k]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $X[k]$ , respectively.

$x[n]$  is a complex sequence



# 用DFT定义证明：

$$\begin{aligned} DFT\left(x^*[-n]_N\right) &= \sum_{n=0}^{N-1} x^*[-n]_N W_N^{nk} \\ &= \sum_{n=0}^{N-1} \left( x[-n]_N W_N^{-nk} \right)^* \\ &= \left( \sum_{n=0}^{N-1} x[N-n]_N W_N^{(N-n)k} \right)^* \\ &= \left( \sum_{m=1}^N x[m]_N W_N^{mk} \right)^* = \left( \sum_{m=0}^{N-1} x[m]_N W_N^{mk} \right)^* = \left( \sum_{m=0}^{N-1} x[m] W_N^{mk} \right)^* \\ &= X^*[k], 0 \leq k \leq N-1 \end{aligned}$$

# Table 5.1: DFT Properties: Symmetry Relations

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k]$
$x^*[n]$	$X^*[\langle -k \rangle_N]$
$x^*[\langle -n \rangle_N]$	$X^*[k]$
$\text{Re}\{x[n]\}$	$X_{\text{pcs}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] + X^*[\langle -k \rangle_N]\}$
$j \text{Im}\{x[n]\}$	$X_{\text{pca}}[k] = \frac{1}{2}\{X[\langle k \rangle_N] - X^*[\langle -k \rangle_N]\}$
$x_{\text{pcs}}[n]$	$\text{Re}\{X[k]\}$
$x_{\text{pca}}[n]$	$j \text{Im}\{X[k]\}$

Note:  $x_{\text{pcs}}[n]$  and  $x_{\text{pca}}[n]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $x[n]$ , respectively. Likewise,  $X_{\text{pcs}}[k]$  and  $X_{\text{pca}}[k]$  are the periodic conjugate-symmetric and periodic conjugate-antisymmetric parts of  $X[k]$ , respectively.

$x[n]$  is a complex sequence



# 用DFT定义证明：

$$\begin{aligned} DFT(\operatorname{Re}\{x[n]\}) &= DFT\left(\frac{1}{2}\{x[n] + x^*[n]\}\right) \\ &= \frac{1}{2}\left\{X[k] + X^*\left[\langle -k \rangle_N\right]\right\} \end{aligned}$$

# Table 5.2: DFT Properties: Symmetry Relations

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k] = \operatorname{Re}\{X[k]\} + j \operatorname{Im}\{X[k]\}$
$x_{pe}[n]$	$\operatorname{Re}\{X[k]\}$
$x_{po}[n]$	$j \operatorname{Im}\{X[k]\}$
Symmetry relations	
	$X[k] = X^*[\langle -k \rangle_N]$
	$\operatorname{Re} X[k] = \operatorname{Re} X[\langle -k \rangle_N]$
	$\operatorname{Im} X[k] = -\operatorname{Im} X[\langle -k \rangle_N]$
	$ X[k]  =  X[\langle -k \rangle_N] $
	$\arg X[k] = -\arg X[\langle -k \rangle_N]$

Note:  $x_{pe}[n]$  and  $x_{po}[n]$  are the periodic even and periodic odd parts of  $x[n]$ , respectively.

$x[n]$  is a real sequence



**5.43** Since  $x[n]$  is a length-11 real sequence, its DFT satisfies  $X[k] = X^*[\langle -k \rangle_{11}]$ . Thus:

$$X[1] = X^*[\langle -1 \rangle_{11}] = X^*[10] = 1.5 + j5.31,$$

$$X[3] = X^*[\langle -3 \rangle_{11}] = X^*[8] = -3.34 - j3.69,$$

$$X[5] = X^*[\langle -5 \rangle_{11}] = X^*[6] = -7.55 - j13.69,$$

$$X[7] = X^*[\langle -7 \rangle_{11}] = X^*[4] = -12.44 - j12.7,$$

$$X[9] = X^*[\langle -9 \rangle_{11}] = X^*[2] = 2.49 + j19.12.$$

# Table 5.2: DFT Properties: Symmetry Relations

Length- $N$ Sequence	$N$ -point DFT
$x[n]$	$X[k] = \operatorname{Re}\{X[k]\} + j \operatorname{Im}\{X[k]\}$
$x_{pe}[n]$	$\operatorname{Re}\{X[k]\}$
$x_{po}[n]$	$j \operatorname{Im}\{X[k]\}$
Symmetry relations	
	$X[k] = X^*[\langle -k \rangle_N]$
	$\operatorname{Re} X[k] = \operatorname{Re} X[\langle -k \rangle_N]$
	$\operatorname{Im} X[k] = -\operatorname{Im} X[\langle -k \rangle_N]$
	$ X[k]  =  X[\langle -k \rangle_N] $
	$\arg X[k] = -\arg X[\langle -k \rangle_N]$

Note:  $x_{pe}[n]$  and  $x_{po}[n]$  are the periodic even and periodic odd parts of  $x[n]$ , respectively.

$x[n]$  is a real sequence



**5.45** Since the DFT  $X[k]$  is real-valued,  $x[n]$  is circularly even:  $x[n] = x[\langle -n \rangle_{10}]$ . Therefore:

$$x[2] = x[\langle -2 \rangle_{10}] = x[8] = 6.26,$$

$$x[6] = x[\langle -6 \rangle_{10}] = x[4] = -3.1,$$

$$x[7] = x[\langle -7 \rangle_{10}] = x[3] = 8.58,$$

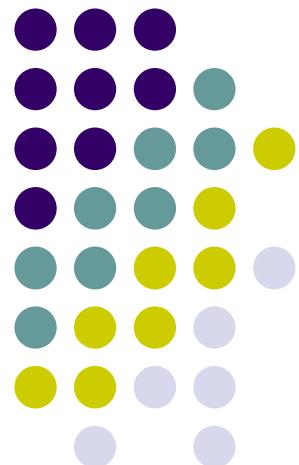
$$x[9] = x[\langle -9 \rangle_{10}] = x[1] = 6.2.$$

# Table 5.3: General Properties of DFT

Type of Property	Length- $N$ Sequence	$N$ -point DFT
	$g[n]$ $h[n]$	$G[k]$ $H[k]$
Linearity	$\alpha g[n] + \beta h[n]$	$\alpha G[k] + \beta H[k]$
Circular time-shifting	$g[\langle n - n_o \rangle_N]$	$W_N^{kn_o} G[k]$
Circular frequency-shifting	$W_N^{-k_o n} g[n]$	$G[\langle k - k_o \rangle_N]$
Duality	$G[n]$	$N g[\langle -k \rangle_N]$
$N$ -point circular convolution	$\sum_{m=0}^{N-1} g[m]h[\langle n - m \rangle_N]$	$G[k]H[k]$
Modulation	$g[n]h[n]$	$\frac{1}{N} \sum_{m=0}^{N-1} G[m]H[\langle k - m \rangle_N]$
Parseval's relation	$\sum_{n=0}^{N-1}  x[n] ^2 = \frac{1}{N} \sum_{k=0}^{N-1}  X[k] ^2$	

# 6、Fourier-Domain Filtering

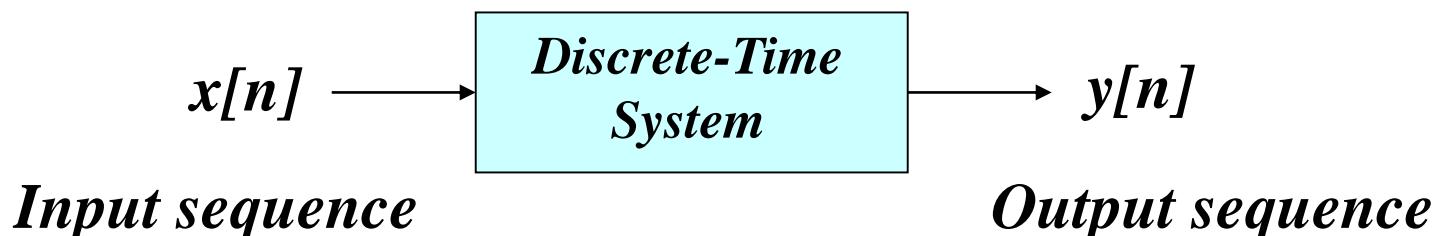
See in textbook section 5.8





## 6. Fourier-Domain Filtering

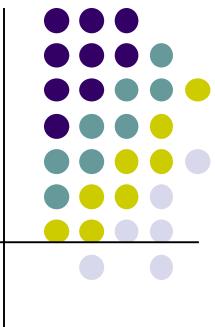
- Often one is interested in removing the components of a finite-length discrete-time signal in one or more frequency bands.



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

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

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



*length L      length K      length L+K-1*

$$x(n) \circledast h(n) = y_L(n) \quad \longleftrightarrow^{FT} \quad Y(e^{j\omega}) = X(e^{j\omega}) \cdot H(e^{j\omega})$$

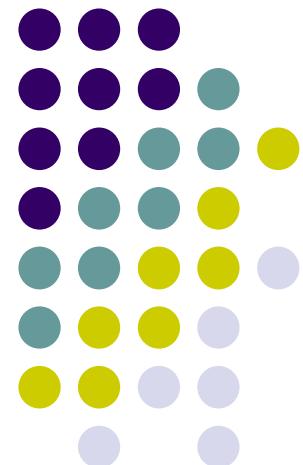
↑   ↑ Sampling ( N samples in a period)

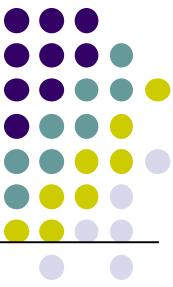
$$x(n) \circledcirc h(n) = y_c(n) \xleftarrow[N \text{ points}]{DFT} Y(k) = X(k) \cdot H(k)$$

*N points      N points      length N*

## 7. Linear Convolution Using the DFT

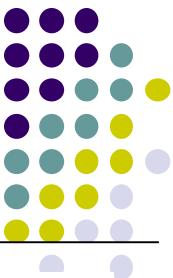
See in textbook section 5.10





# 7. Linear Convolution Using the DFT

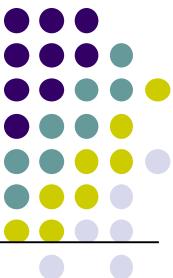
- *Linear convolution* is a key operation in many signal processing applications.
- Since a DFT can be efficiently implemented using FFT algorithms, it is of interest to develop methods for the implementation of linear convolution using the DFT



# Linear Convolution of Two Finite-Length Sequences

- Let  $g(n)$  and  $h(n)$  be two finite-length sequences of length  $N$  and  $M$ , respectively
- Denote  $L=N+M-1$
- Define two length- $L$  sequences

$$g_e(n) = \begin{cases} g(n), & 0 \leq n \leq N-1 \\ 0, & N \leq n \leq L-1 \end{cases}$$
$$h_e(n) = \begin{cases} h(n), & 0 \leq n \leq M-1 \\ 0, & M \leq n \leq L-1 \end{cases}$$

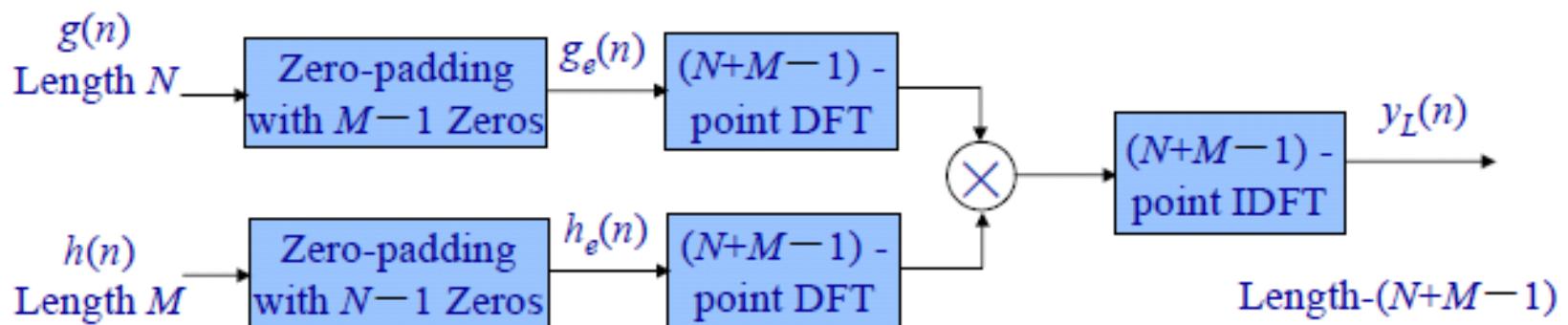


# Linear Convolution of Two Finite-Length Sequences

- Then

$$y_L(n) = g(n) \circledast h(n) = g(n) \textcircled{L} h(n)$$

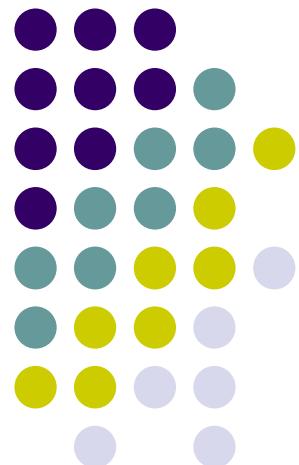
- The corresponding implementation scheme is illustrated below



# Linear Convolution of a Finite Sequence with an Infinite Sequence

**Overlap-Add Method**

**Overlap-Save Method**





# Overlap-Add Method

**Overlap-add method:** When the input  $x$  is infinite or extremely long, divide the long input into contiguous non-overlapping blocks  $x_0, x_1, x_2, \dots$  of manageable length, then filter each block and add the output overlapped blocks

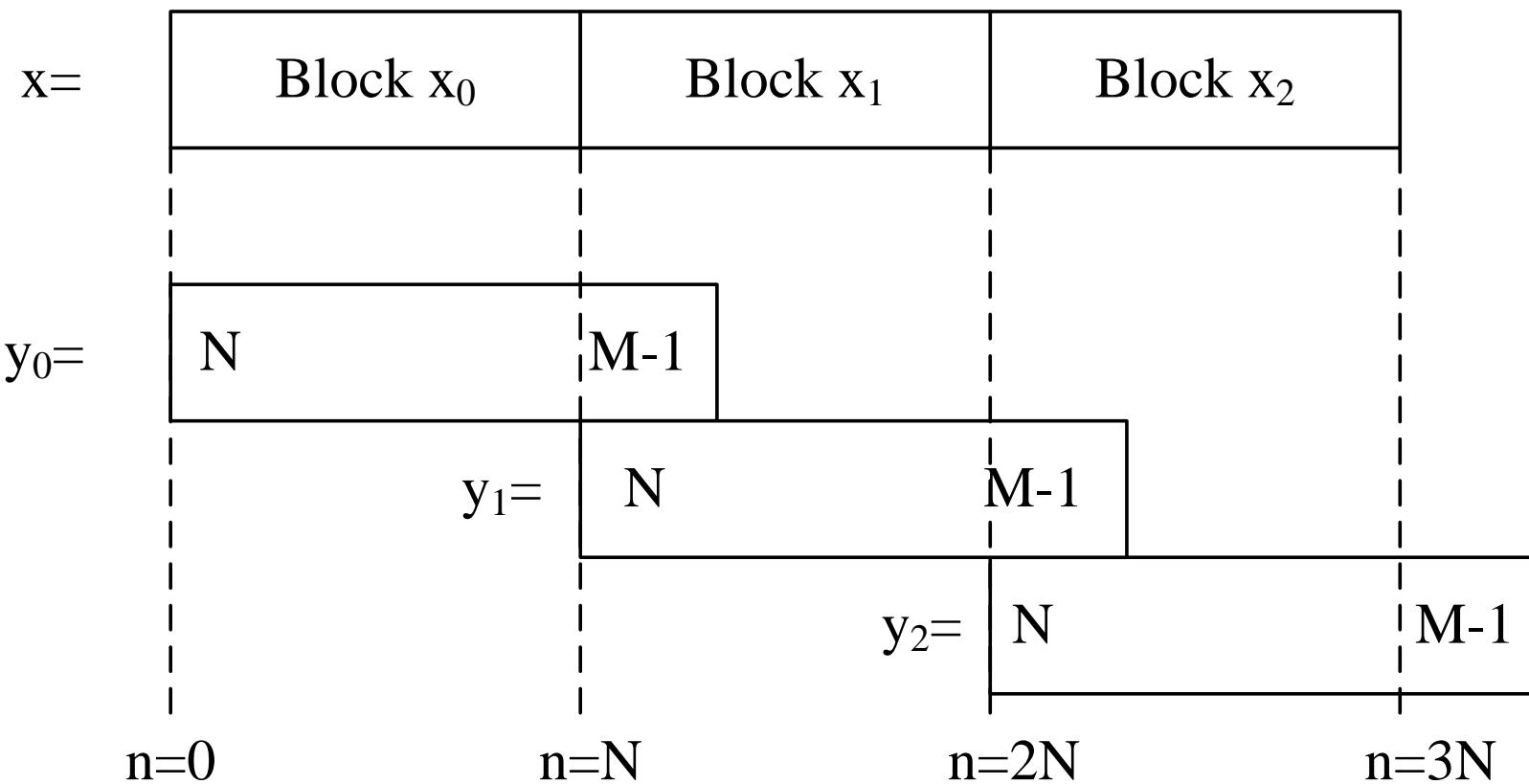
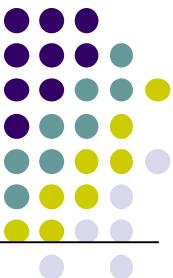
$$y_0 = h * x_0,$$

$$y_1 = h * x_1,$$

$$y_2 = h * x_2,$$

...

to obtain the overall output.





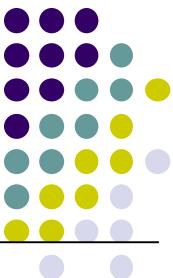
$$\mathbf{X} = [ \underbrace{1, 1, 2}_{\mathbf{X}_0}, \underbrace{1, 2, 2}_{\mathbf{X}_1}, \underbrace{1, 1, 0}_{\mathbf{X}_2} ]$$

$\mathbf{h} \setminus \mathbf{x}$	block 0			block 1			block 2		
	1	1	2	1	2	2	1	1	0
1	1	1	2	1	2	2	1	1	0
2	2	2	4	2	4	4	2	2	0
-1	-1	-1	-2	-1	-2	-2	-1	-1	0
1	1	1	2	1	2	2	1	1	0

$$\mathbf{y}_0 = \mathbf{h} * \mathbf{x}_0 = [1, 3, 3, 4, -1, 2]$$

$$\mathbf{y}_1 = \mathbf{h} * \mathbf{x}_1 = [1, 4, 5, 3, 0, 2]$$

$$\mathbf{y}_2 = \mathbf{h} * \mathbf{x}_2 = [1, 3, 1, 0, 1, 0]$$



# Overlap-add, M=4

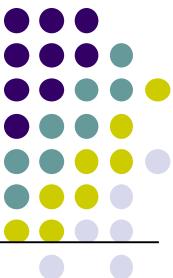
$n$	0	1	2	3	4	5	6	7	8	9	10
$y_0$	1	3	3	4	-1	2					
$y_1$				1	4	5	3	0	2		
$y_2$						1	3	1	0	1	
$y$	1	3	3	5	3	7	4	3	3	0	1



# Overlap-Save Method

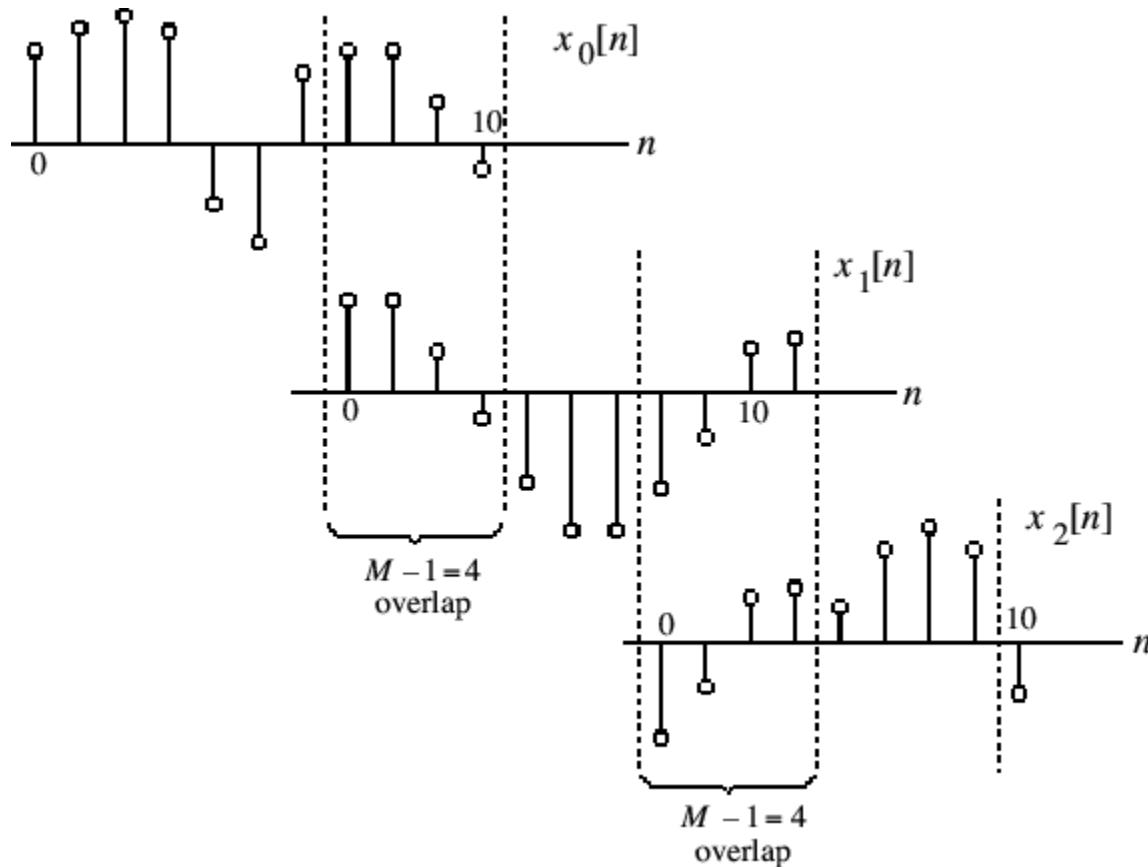
- 基本思想

**segment  $x[n]$  into overlapping blocks , keep the  $y_m[n]$  terms of the circular convolution of  $h[n]$  with  $x_m[n]$  that corresponds to the terms obtained by a linear convolution of  $h[n]$  and  $x_m[n]$ , and throw away the other parts of the circular convolution**



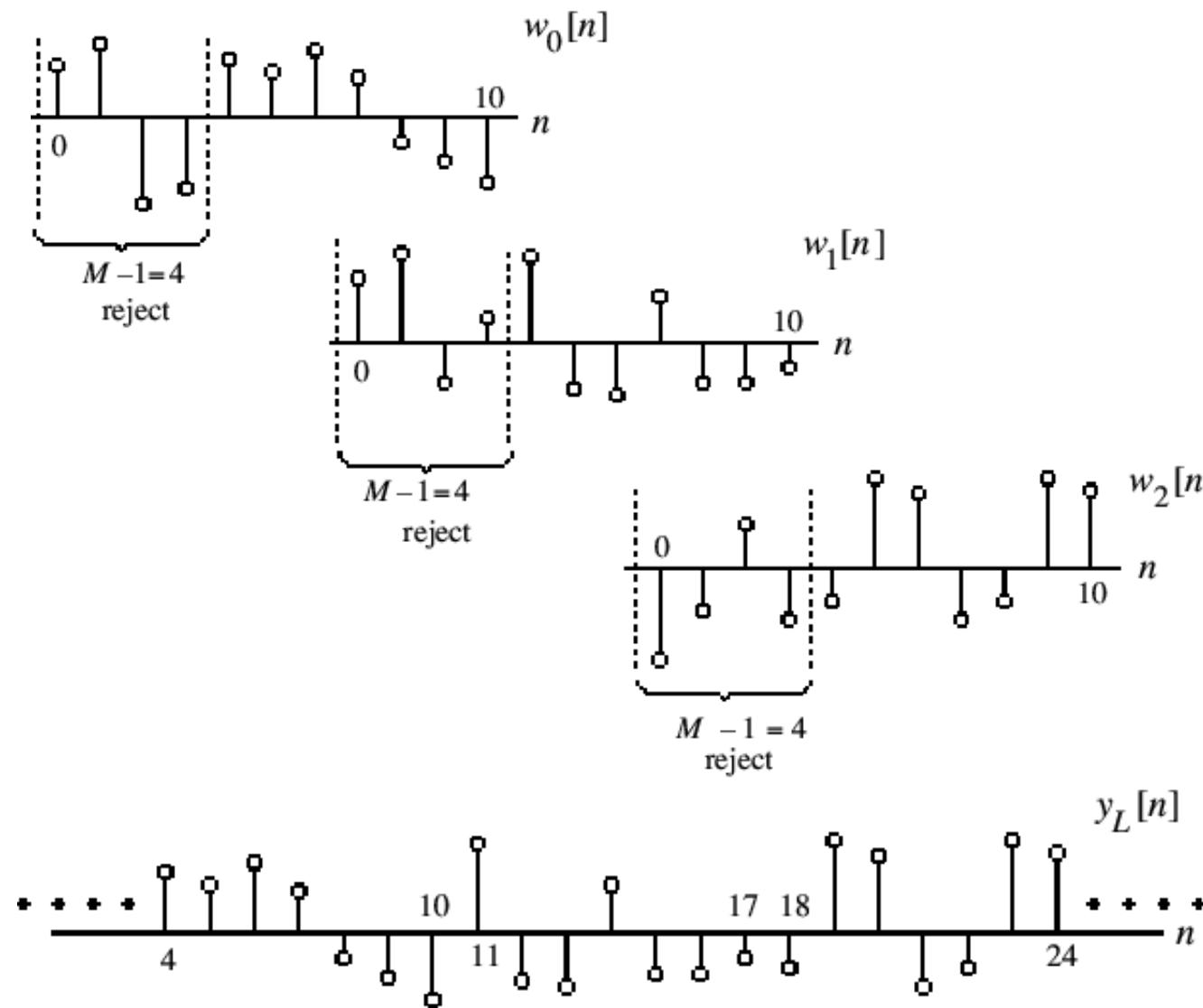
# Overlap-Save Method

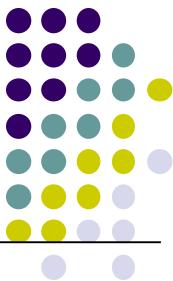
- Process is illustrated next





# Overlap-Save Method





# 本章重点

- DFT的定义、性质及其证明
- 循环卷积、循环卷积与线性卷积的关系
- 实序列的DFT
- 线性卷积的DFT实现
- 重叠相加法、重叠保留法



# Homework

## **Problems:**

**5.2(a), 5.9(a), 5.25, 5.28, 5.43, 5.45, 5.55, 5.68,  
5.76(a,b,c,d)**