

# Homework & Tutorial Problems

- Tutorial
  - ▶ Property of DTFT: 5.24
  - ▶ Difference equation of LTI systems: 5.36
- Homework
  - ▶ 5.23, 5.29, 5.33



# Convolution Property & LTI Systems (1/2)

- Let  $h[n]$  be the **impulse response** of certain LTI system
- The output  $y[n]$  of input  $x[n]$  is given by  $y[n] = x[n] * h[n]$
- For input signal  $x[n] = e^{j\omega n}$ ,

$$y[n] = \sum_{k=-\infty}^{\infty} h[k]e^{j\omega(n-k)} = e^{j\omega n} \underbrace{\sum_{k=-\infty}^{\infty} h[k]e^{-j\omega k}}_{\text{Frequency Response } H(e^{j\omega})}$$

- For periodic input signal  $x[n] = \sum_{k=\langle N \rangle} a_k e^{jk(\frac{2\pi}{N})n}$ :

$$y[n] = \sum_{k=\langle N \rangle} a_k H(e^{jk(\frac{2\pi}{N})}) e^{jk(\frac{2\pi}{N})n}$$

$$b_k = a_k H(e^{jk(\frac{2\pi}{N})})$$

# Convolution Property & LTI Systems (2/2)

## Convolution Property

If  $y[n] = x_1[n] * x_2[n]$ , then

$$Y(e^{j\omega}) = X_1(e^{j\omega})X_2(e^{j\omega})$$

- For general input signal  $x[n]$ :

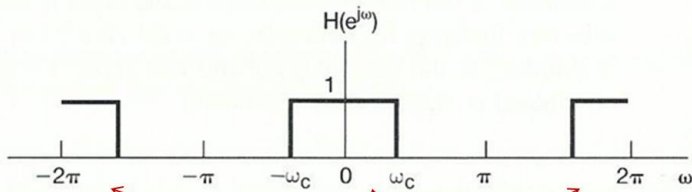
$$y[n] = x[n] * h[n] \longleftrightarrow Y(e^{j\omega}) = X(e^{j\omega})H(e^{j\omega})$$

- **Observation:** It's easier to evaluate LTI systems in frequency domain
- **Drawback:** Not every LTI system has frequency response
  - ▶  $h[n] = a^n u[n]$  ( $a > 1$ )
  - ▶ Stable LTI system has frequency response, because

$$\sum_{n=-\infty}^{\infty} |h[n]| < \infty$$

# Example: Ideal Low-Pass Filter (1/2)

- What's ideal low-pass filter?
- Perfectly maintain the low-frequency component
- Perfectly cancel the high-frequency component

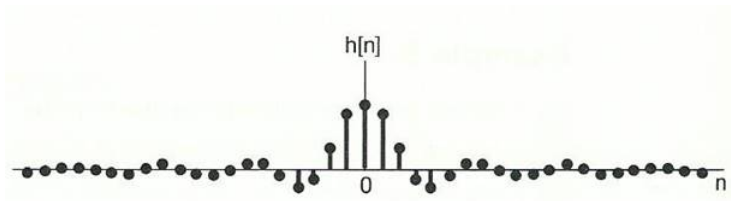


**Repeat with period  $2\pi$**

## Example: Ideal Low-Pass Filter (2/2)

- Impulse response:

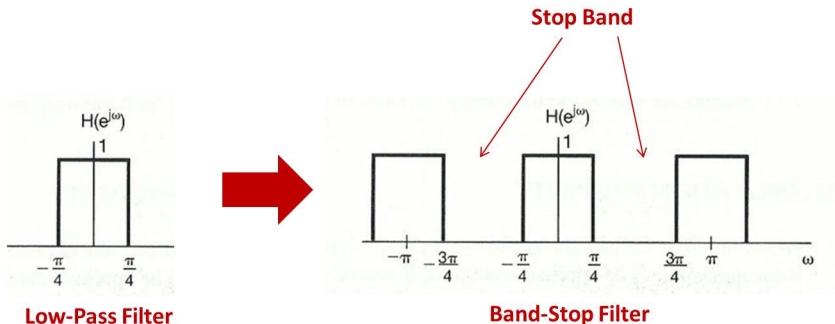
$$h[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} H(e^{j\omega}) e^{j\omega n} d\omega = \frac{1}{2\pi} \int_{-\omega_c}^{\omega_c} e^{j\omega n} d\omega = \frac{\sin \omega_c n}{\pi n}$$



- Pros: no distortion in frequency domain
- Cons: non-causal
- See textbook, Example 5.12

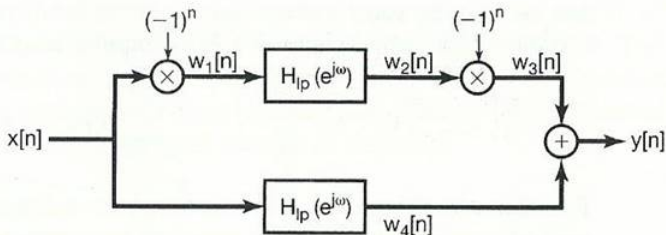
## Example: Band-Stop Filter (1/2)

- Given low-pass filter  $H_{lp}(e^{j\omega})$ , how to generate band-stop effect from low-pass filter?



## Example: Band-Stop Filter (2/2)

- Two branches: low-pass + high-pass



- $(-1)^n = e^{j\pi n} \Rightarrow W_1(e^{j\omega}) = X(e^{j(\omega-\pi)})$

$$\Rightarrow W_2(e^{j\omega}) = H_{lp}(e^{j\omega})X(e^{j(\omega-\pi)})$$

$$\Rightarrow W_3(e^{j\omega}) = H_{lp}(e^{j(\omega-\pi)})X(e^{j(\omega-2\pi)}) = H_{lp}(e^{j(\omega-\pi)})X(e^{j\omega})$$

- See textbook, Example 5.14

# Multiplication Property

## Multiplication Property

Let  $y[n] = x_1[n]x_2[n]$ , then

$$Y(e^{j\omega}) = \frac{1}{2\pi} \int_{2\pi} X_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta,$$

which is *periodic convolution* of  $X_1(e^{j\omega})$  and  $X_2(e^{j\omega})$

- Comparison: Multiplication of periodic signals

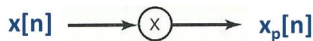
$$x_1[n] \longleftrightarrow a_k \quad \text{and} \quad x_2[n] \longleftrightarrow b_k$$

$$\Rightarrow x_1[n]x_2[n] \longleftrightarrow \sum_{k=\langle N \rangle} a_k b_{n-k} \quad \text{discrete-time periodic convolution}$$



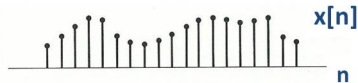
# Example: Sampling on Discrete-Time Signals

- In digital signal processing, we may sample the discrete-time signals
- Example: image, audio, video compression
- Its mathematical model is give below:



$$p[n] = \sum_{k=-\infty}^{+\infty} \delta[n - kN]$$

$$x_p[n] = \sum_{k=-\infty}^{+\infty} x[kN] \delta[n - kN]$$



# Review: Discrete-Time Impulse Chain

- What's the Fourier transform of  $x[n] = \sum_{k=-\infty}^{+\infty} \delta[n - kN]$  ?
- First of all, we calculate the Fourier series:

$$\begin{aligned} a_k &= \frac{1}{N} \sum_{n=\langle N \rangle} x[n] e^{-jk \frac{2\pi}{N} n} \\ &= \frac{1}{N} \sum_{n=\langle N \rangle} \sum_{k=-\infty}^{+\infty} \delta[n - kN] e^{-jk \frac{2\pi}{N} n} \\ &= \frac{1}{N} \sum_{n=\langle N \rangle} \delta[n] e^{-jk \frac{2\pi}{N} n} = \frac{1}{N} \end{aligned}$$

- Then, we have

$$X(e^{j\omega}) = \sum_{l=-\infty}^{\infty} \sum_{k=0}^{N-1} \frac{2\pi}{N} \delta(\omega - k(2\pi/N) - 2\pi l) = \frac{2\pi}{N} \sum_{k=-\infty}^{\infty} \delta(\omega - \frac{2\pi k}{N})$$

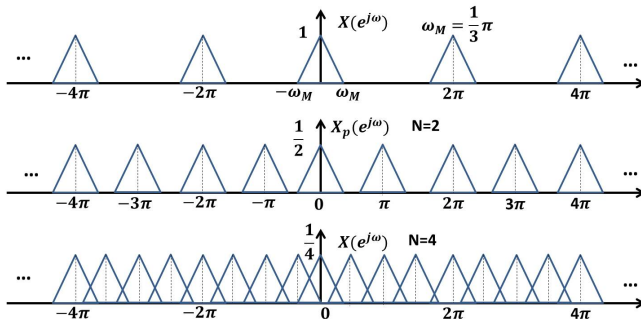
- Time domain period  $\times$  Frequency domain period = ?



# Example: Frequency Analysis

$$P(e^{j\omega}) = \frac{2\pi}{N} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s) \quad \text{where } \omega_s = 2\pi/N$$

$$X_p(e^{j\omega}) = \frac{1}{2\pi} \int_{2\pi} P(e^{j\theta}) X(e^{j(\omega-\theta)}) d\theta = \frac{1}{N} \sum_{k=0}^{N-1} X(e^{j(\omega-k\omega_s)})$$



# Duality in DTFS

- Analysis and synthesis equations of DTFS share a similar form:

$$\begin{aligned}x[n] &= \sum_{k=\langle N \rangle} a_k e^{jk(2\pi/N)n} \\a_k &= \frac{1}{N} \sum_{k=\langle N \rangle} x[n] e^{-jk(2\pi/N)n}\end{aligned}\tag{1}$$

- Every of DTFS has a dual (See Table 3.2)
  - ▶ Time shift v.s. Frequency shift
  - ▶ Time multiplication v.s. Frequency multiplication
- If  $g[n] \longleftrightarrow f[k]$ , then  $f[n] \longleftrightarrow \frac{1}{N}g[-k]$

# Duality Between CTFS and DTFT

- DTFT transform pair is given by

$$\begin{aligned}x[n] &= \frac{1}{2\pi} \int_{2\pi} X(e^{j\omega}) e^{j\omega n} d\omega \\X(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}\end{aligned}$$

- CTFS transform pair is given by

$$\begin{aligned}x(t) &= \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t} \\a_k &= \frac{1}{T} \int_T x(t) e^{-jk\omega_0 t} dt\end{aligned}$$

- Frequency Domain of DTFT  $\Leftrightarrow$  Time Domain of CTFS
- Frequency Domain of CTFS  $\Leftrightarrow$  Time Domain of DTFT
- Discrete-Time Low-pass Filter v.s. Continuous-Time Square Wave

# Summary of Duality

	Continuous time		Discrete time	
	Time domain	Frequency domain	Time domain	Frequency domain
Fourier Series	$x(t) = \sum_{k=-\infty}^{+\infty} a_k e^{jk\omega_0 t}$ <p>continuous time periodic in time</p>	$a_k = \frac{1}{T_0} \int_{T_0} x(t) e^{-jk\omega_0 t} dt$ <p>discrete frequency aperiodic in frequency</p>	$x[n] = \sum_{k=\langle N \rangle} a_k e^{jk(2\pi/N)n}$ <p>discrete time periodic in time</p>	$a_k = \frac{1}{N} \sum_{k=\langle N \rangle} x[n] e^{-jk(2\pi/N)n}$ <p>discrete frequency periodic in frequency</p>
Fourier Transform	$x(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X(j\omega) e^{j\omega t} d\omega$ <p>continuous time aperiodic in time</p>	$X(j\omega) = \int_{-\infty}^{+\infty} x(t) e^{-j\omega t} dt$ <p>continuous frequency aperiodic in frequency</p>	$x[n] = \frac{1}{2\pi} \int_{2\pi} X(e^{j\omega}) e^{j\omega n} d\omega$ <p>discrete time aperiodic in time</p>	$X(e^{j\omega}) = \sum_{n=-\infty}^{+\infty} x[n] e^{-j\omega n}$ <p>continuous frequency periodic in frequency</p>

- See textbook, Table 5.3

# LTI by Difference Equation

- A number of DT LTI systems can be written as the following **linear constant-coefficient difference equation**

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^M b_k x[n-k]$$

What's the frequency response?

- Taking Fourier transform on both side, we have

$$\begin{aligned} \sum_{k=0}^N a_k e^{-jk\omega} Y(e^{j\omega}) &= \sum_{k=0}^M b_k e^{-jk\omega} X(e^{j\omega}) \\ \Rightarrow H(e^{j\omega}) &= \frac{Y(e^{j\omega})}{X(e^{j\omega})} = \frac{\sum_{k=0}^M b_k e^{-jk\omega}}{\sum_{k=0}^N a_k e^{-jk\omega}} \end{aligned}$$

What's the impulse response?

## Example: Difference Equation

Please calculate the frequency and impulse response of the following LTI systems

$$y[n] - \frac{3}{4}y[n-1] + \frac{1}{8}y[n-2] = 2x[n].$$

### Solution

*According to the last slide,*

$$H(e^{j\omega}) = \frac{2}{1 - \frac{3}{4}e^{-j\omega} + \frac{1}{8}e^{-2j\omega}}.$$

*Moreover, by partial fraction expansion*

$$H(e^{j\omega}) = \frac{4}{1 - \frac{1}{2}e^{-j\omega}} - \frac{2}{1 - \frac{1}{4}e^{-j\omega}} \Rightarrow h[n] = 4\left(\frac{1}{2}\right)^n u[n] - 2\left(\frac{1}{4}\right)^n u[n]$$



# Tutorial on Orthogonal Frequency Division Multiplexing (OFDM)



# Discrete Fourier Transform (DFT)

- In practice, there is a huge demand on processing **finite duration signals**
- Given a finite duration signal  $\{x[0], x[1], \dots, x[N-1]\}$ , its Fourier transform is

$$X(e^{j\omega}) = \sum_{n=0}^{N-1} x[n]e^{-j\omega n}$$

- **Drawback:** The spectrum of DTFT is continuous  $\Rightarrow$  Cannot be handle by computer.
- **Discrete Fourier Transfrom (DFT)** is developed for digital processing of finite duration signals

$$\tilde{X}[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n]e^{-jk(2\pi/N)n} \quad k = \langle N \rangle$$

- DFT: frequency sampling of DTFT

$$\tilde{X}[k] = \frac{1}{N} X(e^{j2k\pi/N}) \quad k = \langle N \rangle$$

# Inverse DFT

- Equation system of DFT:

$$\begin{pmatrix} \tilde{X}[0] \\ \tilde{X}[1] \\ \vdots \\ \tilde{X}[N-1] \end{pmatrix} = \underbrace{\frac{1}{N} \begin{pmatrix} e^{-j0} & e^{-j0} & \dots & e^{-j0} \\ e^{-j0} & e^{-j2\pi/N} & \dots & e^{-j2(N-1)\pi/N} \\ \vdots & \vdots & \ddots & \vdots \\ e^{-j0} & e^{-j2(N-1)\pi/N} & \dots & e^{-j2(N-1)(N-1)\pi/N} \end{pmatrix}}_{\mathbf{F}} \begin{pmatrix} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{pmatrix}$$

- Transform matrix  $\mathbf{F}$  is full rank.
- Observation:**  $\{\tilde{X}[k] | k = 0, \dots, N-1\}$  maintains all the information of  $\{x[0], x[1], \dots, x[N-1]\}$
- Inverse DFT is feasible:

$$x[n] = \sum_{k=0}^{N-1} \tilde{X}[k] e^{jk(2\pi/N)n} \quad n = 0, 1, \dots, N-1$$

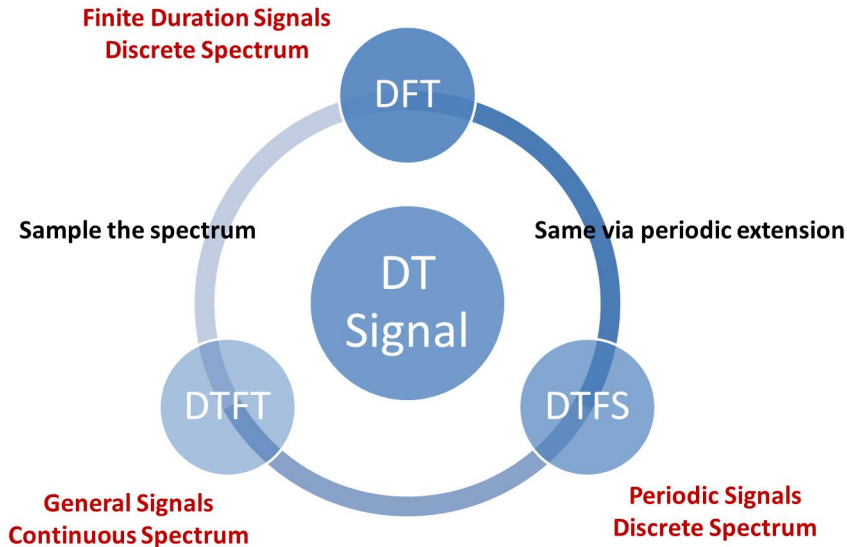
# DFT and DTFS

- DFT is for finite duration signals; DTFS is for periodic signals
- Define  $\tilde{x}[n]$  as the periodic extension of  $x[n]$ : Repeat  $x[n]$  with period  $N$
- Fourier series of  $\tilde{x}[n]$ :

$$\frac{1}{N} \sum_{n=0}^{N-1} \tilde{x}[n] e^{-jk(2\pi/N)n} = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jk(2\pi/N)n} \Rightarrow \text{DFT of } x[n]$$

- DFT of a finite-duration signal = DTFS of its periodic extension
- Reference on DFT:
  - ▶ Textbook: Problem 5.53, 5.54
  - ▶ [http://en.wikipedia.org/wiki/Discrete\\_Fourier\\_transform](http://en.wikipedia.org/wiki/Discrete_Fourier_transform)

# Comparison



# Periodic Convolution of Finite Duration Signals

## Periodic Convolution

Let  $x$  and  $y$  be two finite duration signals with duration  $N$ ,  $\tilde{x}$  and  $\tilde{y}$  be the associated periodic extension, then the periodic convolution of finite duration signals is defined as

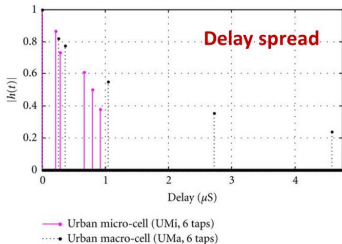
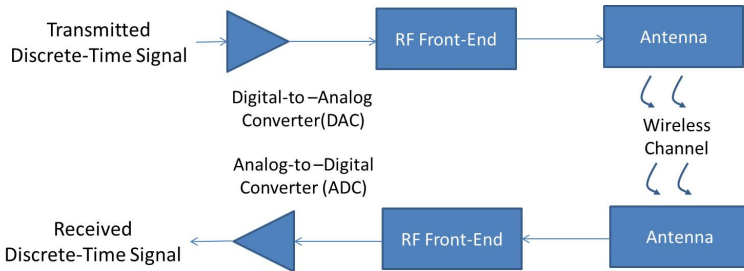
$$x[n] \circledast y[n] := \tilde{x}[n] \circledast \tilde{y}[n] = \sum_{k=\langle N \rangle} \tilde{x}[k] \tilde{y}[n-k] \quad n = 0, 1, 2, \dots, N-1$$

## Convolution Property of DFT

Time domain periodic convolution is equivalent to frequency domain multiplication, thus,

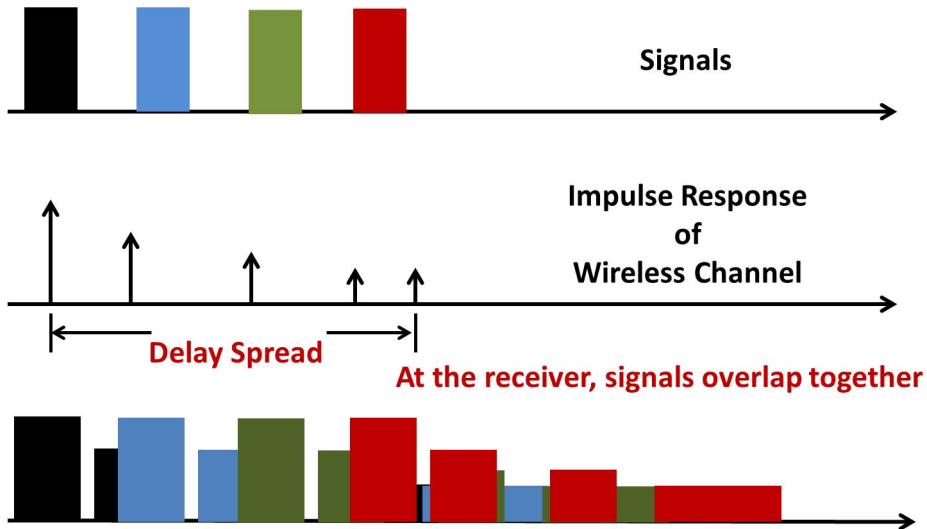
- $x[n] \circledast y[n] \longleftrightarrow N\tilde{X}[k]\tilde{Y}[k]$

# Wireless Systems



**Delay spread causes inter-symbol interference**

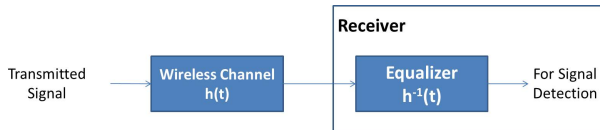
# Inter-Symbol Interference (1/2)





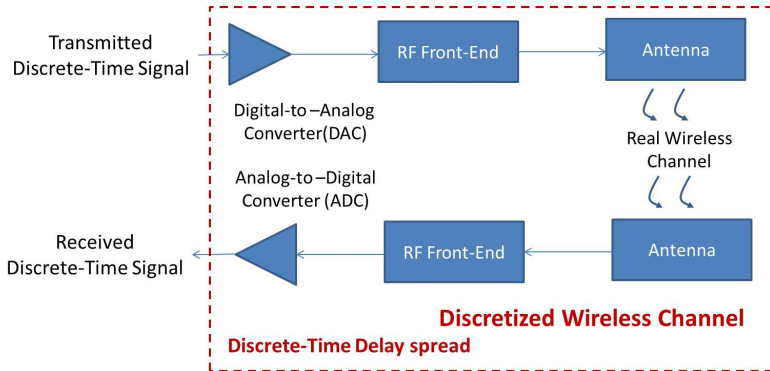
# Inter-Symbol Interference (2/2)

- How to deliver signals without inter-symbol interference?
  - ▶ Suppose the duration of delay spread is  $\Delta H$  seconds
- Approach 1: Send data on every  $\Delta H$  seconds
- Approach 2: Channel equalizer



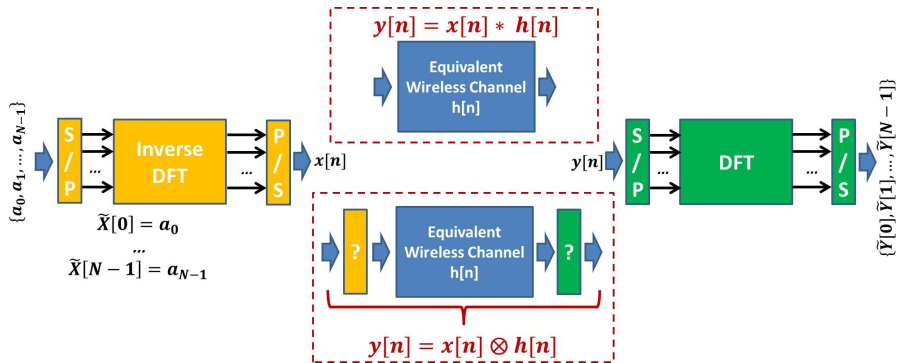
- Approach 3: Orthogonal Frequency Division Multiplexing
  - ▶ Pre-processing at the transmitter + post-processing at the receiver

# Discretetize Wireless Channel



- Tx front-end + wireless channel + Rx front-end: approximately discrete-time LTI system
- Denote the impulse response as  $h[n]$

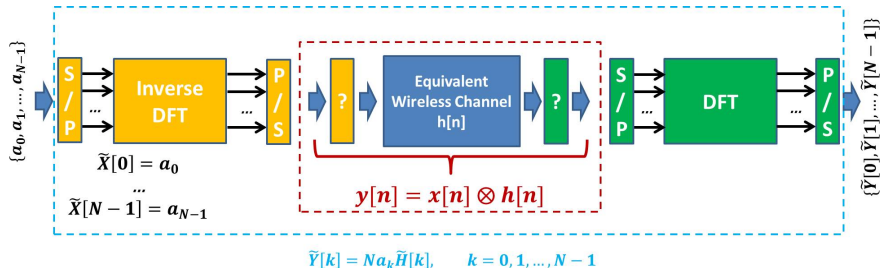
# OFDM at First Glance



- Signals are loaded in frequency domain
- Some mechanism is necessary to generate the effect of periodic convolution



So, ...



- How to detect  $\{a_k | \forall k\}$  from  $\{\tilde{Y}[k] | \forall k\}$ ?
- How to design the blocks "?" ?

# Channel Estimation & Signal Detection

- **Channel estimation:** Estimate the channel gain  $\tilde{H}[k]$  ( $k = 0, 1, \dots, N - 1$ )
  - ▶ The transmitter sends the common information to the receiver  $\{a_0^c, a_1^c, \dots, a_{N-1}^c\}$ :

$$\tilde{Y}^c[k] = Na_k^c \tilde{H}[k] \quad k = 0, 1, \dots, N - 1$$

- ▶ Evaluate channel via

$$\tilde{H}[k] = \tilde{Y}^c[k] / (Na_k^c) \quad k = 0, 1, \dots, N - 1$$

- **Signal detection:** Detect the transmitter's signal via the knowledge of  $\tilde{H}[k]$  ( $k = 0, 1, \dots, N - 1$ ):
  - ▶ The transmitter sends the information to the receiver  $\{a_0, a_1, \dots, a_{N-1}\}$ :

$$\tilde{Y}[k] = Na_k \tilde{H}[k] \quad k = 0, 1, \dots, N - 1$$

- ▶ Detect signal via

$$a_k = \tilde{Y}[k] / (N\tilde{H}[k]) \quad k = 0, 1, \dots, N - 1$$

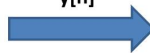
# Convolution vs. Periodic Convolution

Transmission Signal  
 $x[n]$

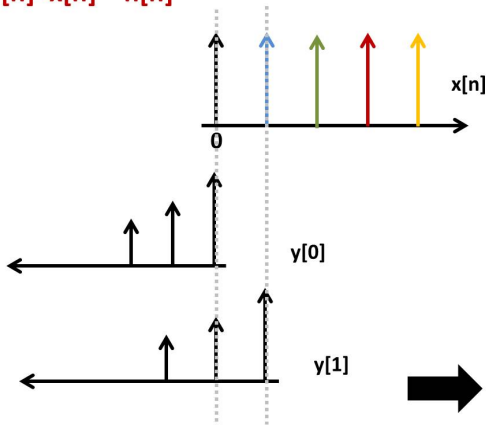
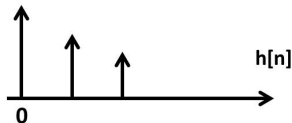


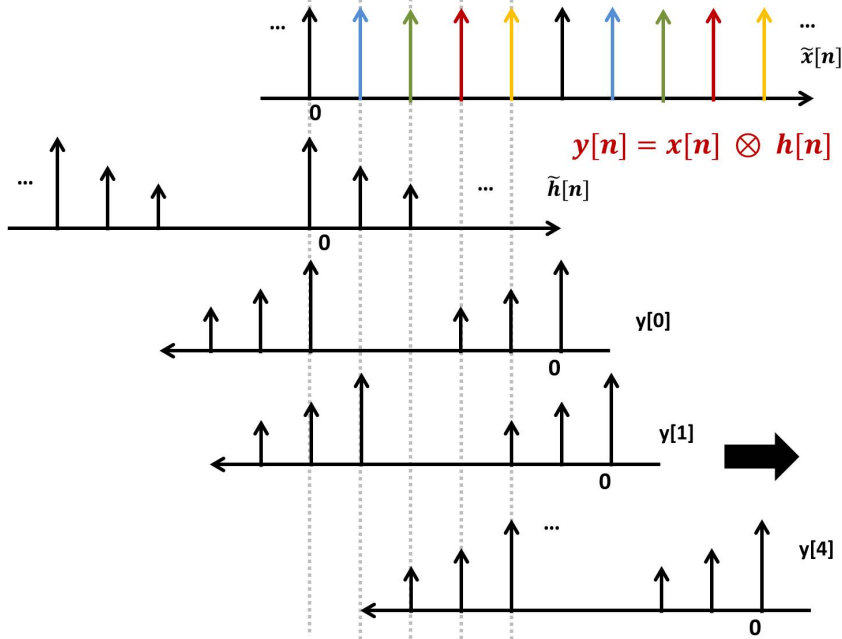
Equivalent Wireless Channel  
DT LTI System  $h[n]$

Received Signal  
 $y[n]$



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



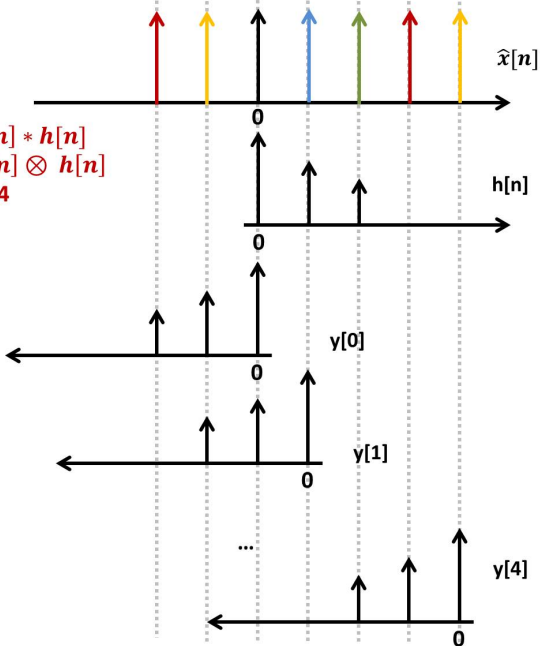




$$y[n] = \hat{x}[n] * h[n]$$

$$= x[n] \otimes h[n]$$

$$n=0,1,2,3,4$$



# Observations

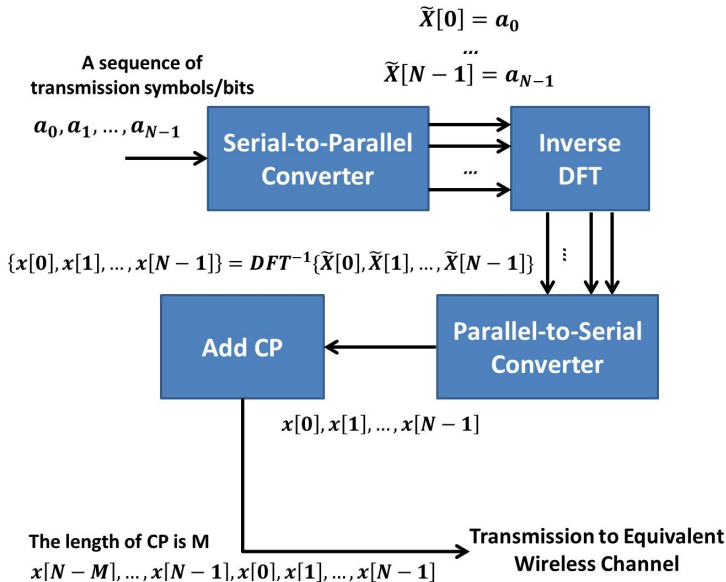
- Periodic convolution can be different from convolution
- In order to guarantee that

$$y[n] = x[n] \circledast h[n], \text{ for } n = 0, 1, 2, \dots, N - 1,$$

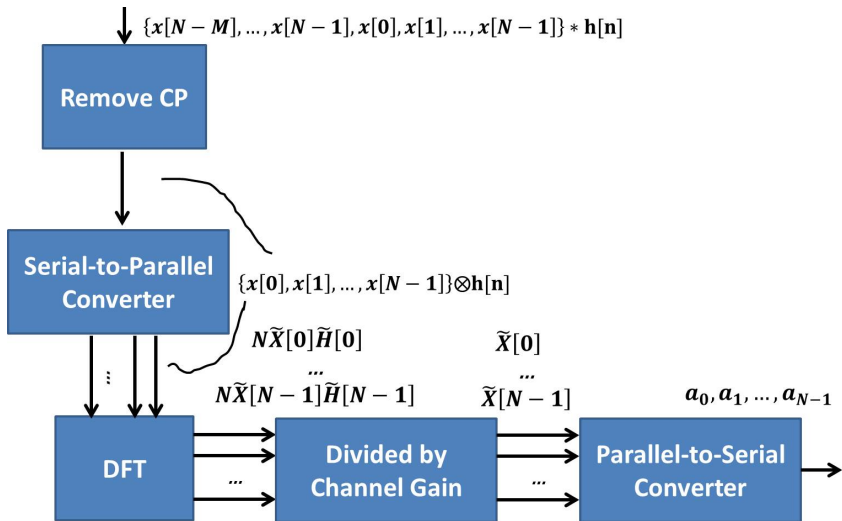
we should copy the last few samples of  $x[n]$  to the beginning, which is named as **cyclic prefix**

- ▶ How many samples should we copy?

# Transmitter Structure of OFDM



# Receiver Structure of OFDM



# Reference

- Reference

- ▶ [www.gaussianwaves.com/2011/05/introduction-to-ofdm-orthogonal-frequency-division-multiplexing-2/](http://www.gaussianwaves.com/2011/05/introduction-to-ofdm-orthogonal-frequency-division-multiplexing-2/)
- ▶ [www.wirelesscommunication.nl/reference/chaptr05/ofdm/ofdmmath.htm](http://www.wirelesscommunication.nl/reference/chaptr05/ofdm/ofdmmath.htm)

