

# EE 470 - Signals and Systems

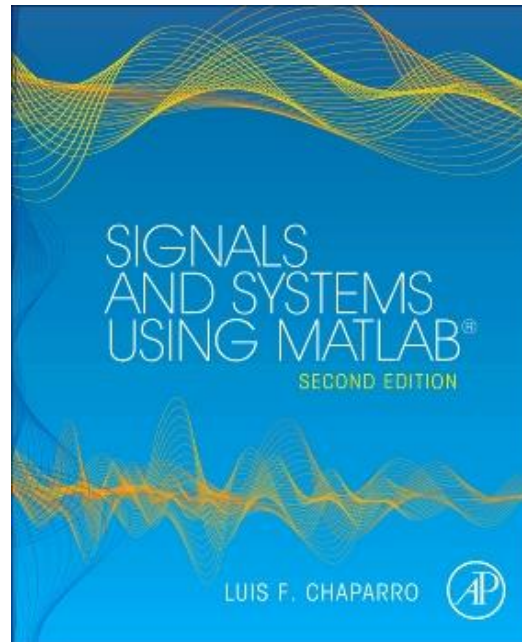
## 8. Fourier Analysis of Discrete-Time Signals and Systems

Prof. Yasser Mostafa Kadah



# Textbook

Luis Chapparo, Signals and Systems Using Matlab, 2<sup>nd</sup> ed., Academic Press, 2015.



# Discrete-Time Fourier Transform

- Sampled time domain signal has a Fourier transform that is periodic
- Periodic time domain signal has a Fourier transform that is sampled
- Periodic and sampled time domain signal has a Fourier transform that is both periodic and sampled

Forward DTFT

$$X(e^{j\omega}) = \sum_n x[n]e^{-j\omega n} \quad -\pi \leq \omega < \pi$$

Inverse DTFT

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega})e^{j\omega n} d\omega$$

# Notes on DTFT

- DTFT is just one period of the continuous Fourier transform of the sampled signal
- Z-transform is related to the DTFT as follows:

$$X_s(e^{j\omega}) = X(z)|_{z=e^{j\omega}}$$

- DTFT is the values on the unit circle in z-Transform
- Unit circle must be included in the ROC
- Problem: the sampled signal is still infinite in length
  - Impossible to obtain using computers
  - Consider cases when the signal has compact support

# Discrete Fourier Transform (DFT)

- Defined as the Fourier transform of a signal that is both discrete and periodic
  - Fourier transform will also be discrete and periodic
  - Can assume periodicity if we have a finite duration of a signal: better than zero assumption since it allows reducing the frequency domain to a sampled function

# DFT Formula

- Given a periodic signal  $x[n]$  of period  $N$ , its DFT is given by,

$$X[k] = \sum_{n=0}^{N-1} x[n]e^{-j2\pi nk/N} \quad 0 \leq k \leq N-1$$

- Its inverse is given by,

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k]e^{j2\pi nk/N} \quad 0 \leq n \leq N-1$$

- Both  $X[k]$  and  $x[n]$  are periodic of the same period  $N$ .

# Computation of DFT Using FFT

- A very efficient computation of the DFT is done by means of the FFT algorithm, which takes advantage of some special characteristics of the DFT
  - DFT requires  $O(N^2)$  computations
  - FFT requires  $O(N \log_2 N)$  computations
  - Both compute the same thing
- It should be understood that the FFT is not another transformation but an algorithm to efficiently compute DFTs.
- Always remember that the signal and its DFT are both periodic

# Improving Resolution in DFT

- When the signal  $x[n]$  is periodic of period  $N$ , the DFT values are normalized Fourier series coefficients of  $x[n]$  that only exist for the  $N$  harmonic frequencies  $\{2\pi k/N\}$ , as no frequency components exist for any other frequencies
  - The number of possible frequencies depend on the length  $L$  chosen to compute its DFT.
  - Frequencies at which we compute the DFT can be seen as frequencies around the unit circle in the  $z$ -plane
- We can improve the frequency resolution of its DFT by increasing the number of samples in the signal without distorting the signal by *padding the signal with zeros*
  - Increase samples of the units circle

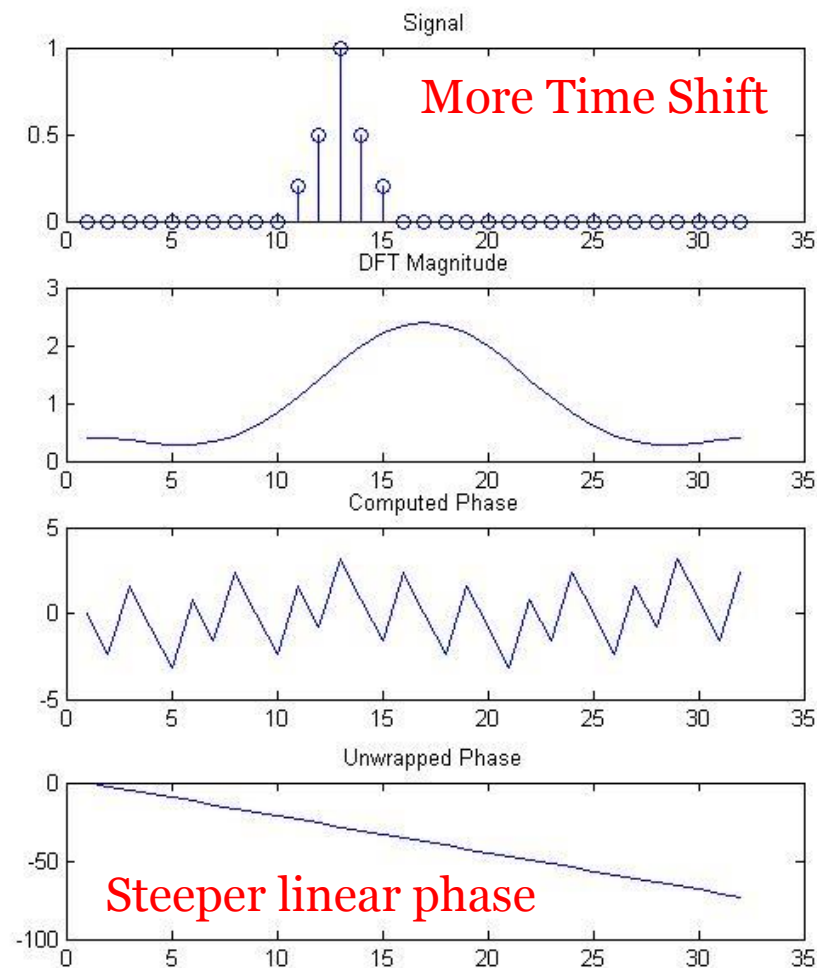
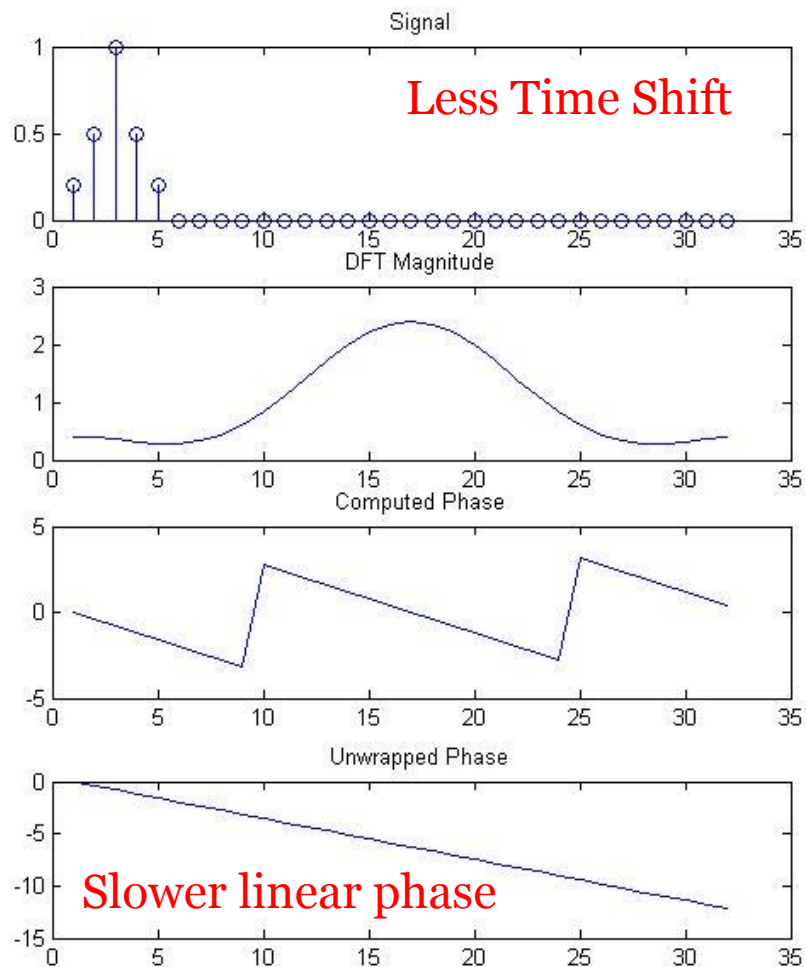


# Phase in DFT

- Output of DFT is complex numbers
- Phase is just as important as magnitude in DFT
- Time shift amounts to linear phase
  - Phase wrapping occurs for large shifts
- Linear phase can be unwrapped to its correct form using the “unwrap” function of Matlab

```
N= 32; x= zeros(N,1); x(11:15)= [0.2 0.5 1 0.5 0.2];  
X= fftshift(fft(x));  
subplot(4,1,1); stem(1:N,x); title('Signal')  
subplot(4,1,2); plot(abs(X)); title('DFT Magnitude')  
subplot(4,1,3); plot(angle(X)); title('Computed Phase')  
subplot(4,1,4); plot(unwrap(angle(X))); title('Unwrapped Phase')
```

# Phase in DFT: Examples

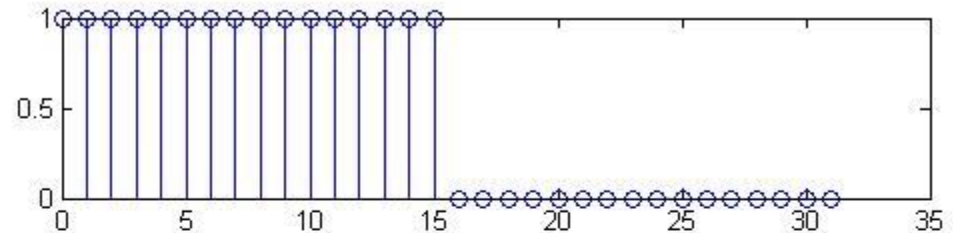


# Zero-Padding Example

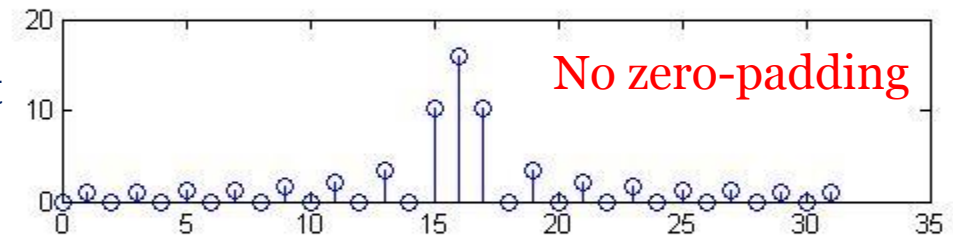
Matlab Code

```
N= 32;  
N2= 128;  
n=0:N-1;  
n2= 0:N2-1;  
x= zeros(N,1);  
x(1:16)= 1;  
X= fftshift(fft(x));  
X2= fftshift(fft(x,N2));  
subplot(3,1,1)  
stem(n,x)  
subplot(3,1,2)  
stem(n,abs(X))  
subplot(3,1,3)  
stem(n2,abs(X2))
```

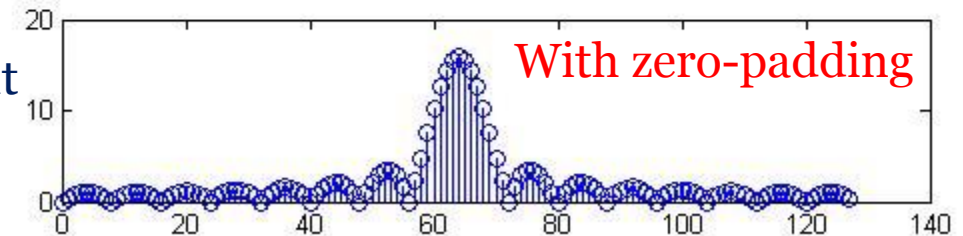
Original  
Signal



32-Point  
FFT



128-Point  
FFT



# Linear vs. Circular Convolution

- Convolution in continuous and discrete time aperiodic signals is called *linear convolution*
- Convolution can be calculated as a multiplication in the frequency domain
  - Convolution property of the Fourier transform
- Problem: when using DFT to compute convolution of aperiodic signals, the periodicity assumption results in errors in computing the convolution
  - Signal outside the N points used is not zero: periodic extension of the N points disturb the computation
  - What we compute is called *circular convolution*

# Linear vs. Circular Convolution

- Given  $x[n]$  and  $h[n]$  of lengths  $M$  and  $K$ , the linear convolution sum  $y[n]$  of length  $N=(M+K-1)$  can be found by following these three steps:
  - Compute DFTs  $X[k]$  and  $H[k]$  of length  $L \geq N$  for  $x[n]$  and  $h[n]$  using zero-padding
  - Multiply them to get  $Y[k]= X[k]H[k]$ .
  - Find the inverse DFT of  $Y[k]$  of length  $L$  to obtain  $y[n]$

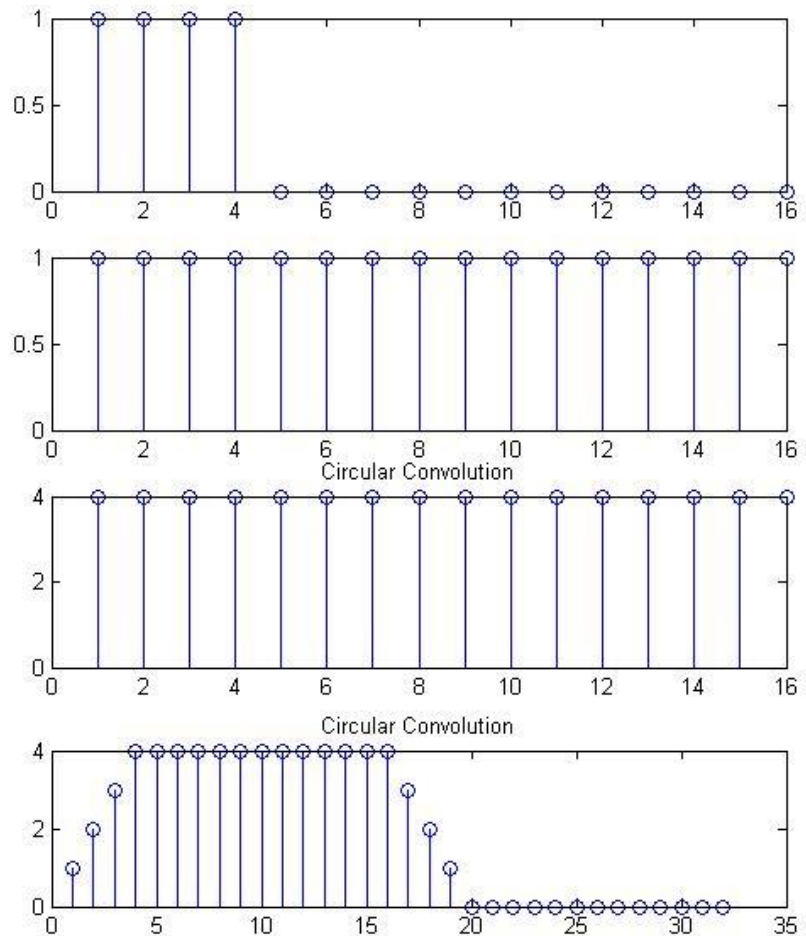
# Linear vs. Circular Convolution: Example

```
N1= 16;  
N2= 16;  
N= 32; % N>N1+N2-1  
x1= zeros(N1,1);  
x2= zeros(N2,1);  
x1(1:4)= 1;  
x2(:)= 1;
```

```
X1= fft(x1,N1);  
X2= fft(x2,N1);  
X_circ= X1.*X2;  
x_circ= ifft(X_circ);
```

```
X1= fft(x1,N);  
X2= fft(x2,N);  
X_lin= X1.*X2;  
x_lin= ifft(X_lin);
```

```
figure(1); subplot(4,1,1); stem(1:N1,x1)  
subplot(4,1,2); stem(1:N2,x2)  
subplot(4,1,3); stem(1:N1, abs(x_circ))  
title('Circular Convolution');  
subplot(4,1,4); stem(1:N,abs(x_lin))  
title('Circular Convolution');
```



# Fourier Transformations Chart

Signal Classification		Frequency	
		Continuous	Discrete
Time	Continuous	Continuous Fourier Transform (CFT)	Discrete Fourier Series (DFS)
	Discrete	Discrete-Time Fourier Transform (DTFT)	<b>Discrete Fourier Transform (DFT)</b>

Most General Form:  
others are special cases



Only one that can be done on a computer



# Covered Material and Assignments

- Chapter 11 of Chaparro's textbook
- Assigned Problem Set #8