



## BLM3620 Digital Signal Processing\*

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\*Based on lecture notes from Ali Can Karaca & Ahmet Elbir

# Lecture #10 – Discrete Fourier Transform and Properties

- Discrete Fourier Transform
- Examples
- Solution using Properties
- MATLAB Applications

# Recap: Discrete Time Fourier Transform

Definition of the **DTFT**:

$$X(e^{j\hat{\omega}}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\hat{\omega}n}$$

- Always periodic with a period of  $2\pi$

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# Discrete Fourier Transform



DFT can be obtained by sampling of DTFT.

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$$\text{Periodic : } X(e^{j(\hat{\omega}+2\pi)}) = X(e^{j\hat{\omega}}) \Rightarrow X[k+N] = X[k]$$

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In DTFT,  $X$  is a continuous function of  $\omega$  whereas in DFT  $X$  is discrete.  $k \rightarrow$  freq. index

**Inverse DFT Transform:**

$$[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j\frac{2\pi}{N}kn}$$

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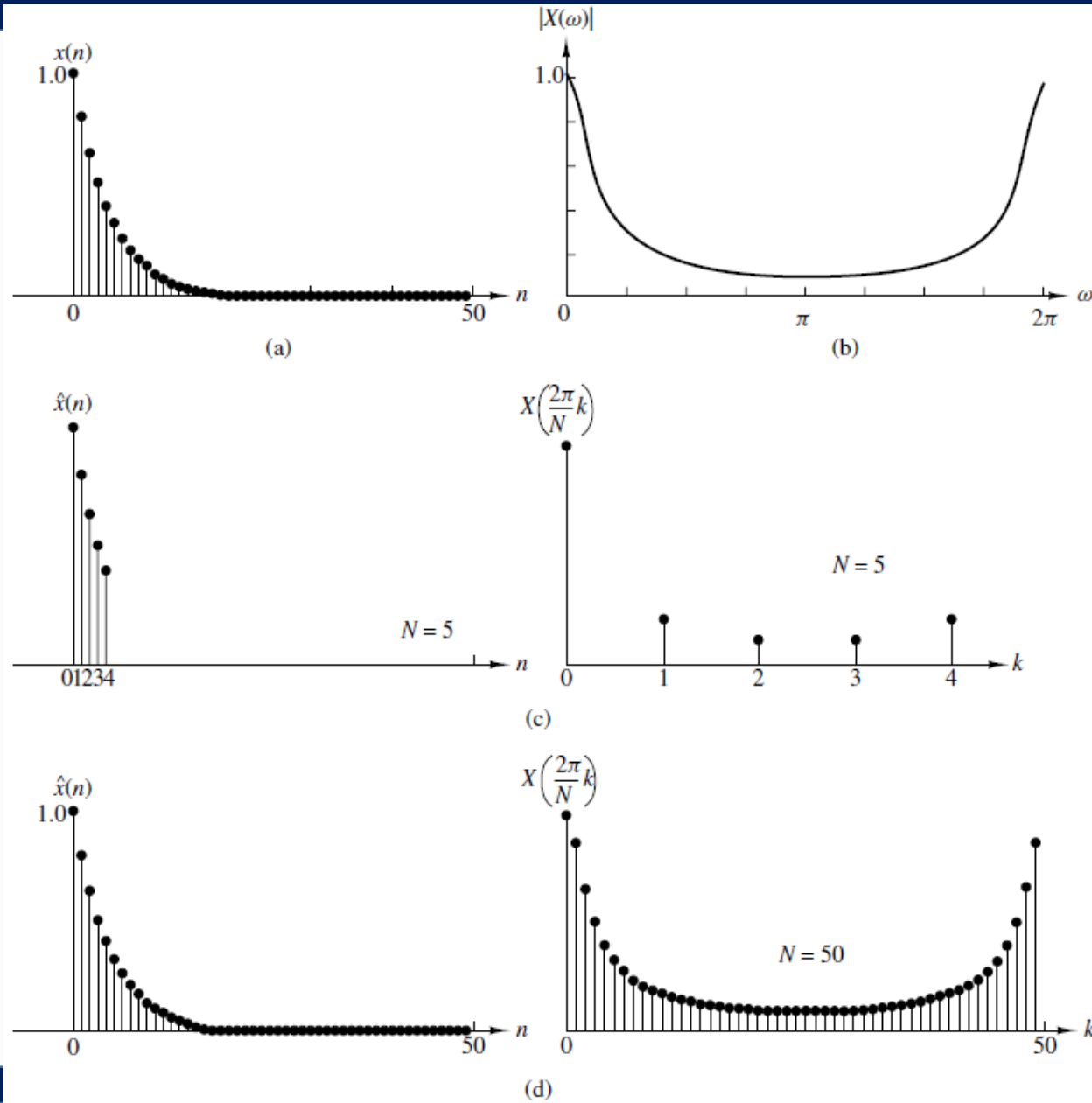
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# Effects of $N$ Value on Result



Should be greater or equal than the number of samples.

# Example from Sarp Erturk's book:

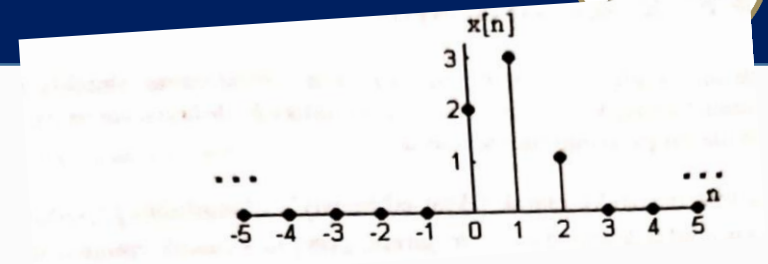


$x[n] = 2\delta[n] + 3\delta[n - 1] + \delta[n - 2]$  find DFT of  $x[n]$ .



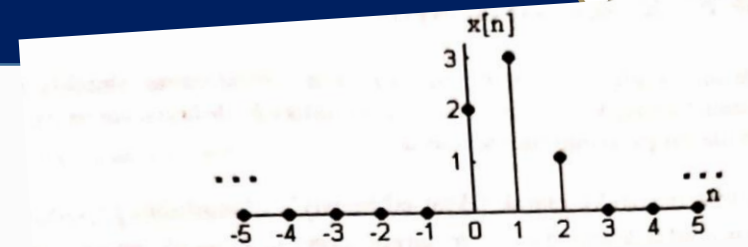
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İşaretin sadece üç değeri sıfırdan farklı olduğu için  $N = 3$  olarak alınabilmektedir. İşaretin ayrık Fourier dönüşümü

$$k = 0 \text{ için} \quad X[0] = \sum_{n=0}^{N-1} x[n]e^{-j(2\pi/N)0n} = \sum_{n=0}^2 x[n] = x[0] + x[1] + x[2] = 6$$

$$k = 1 \text{ için} \quad X[1] = \sum_{n=0}^2 x[n]e^{-j(2\pi/3)n} = x[0] + x[1]e^{-j(2\pi/3)} + x[2]e^{-j(4\pi/3)}$$

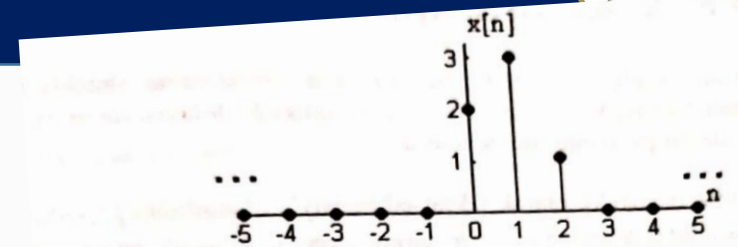
$$X[1] = 2 + 3e^{-j(2\pi/3)} + e^{-j(4\pi/3)} = -j1.7321 = 1.7321e^{-j\pi/2}$$

$$k = 2 \text{ için} \quad X[2] = \sum_{n=0}^2 x[n]e^{-j(2\pi/3)2n} = x[0] + x[1]e^{-j(4\pi/3)} + x[2]e^{-j(8\pi/3)}$$

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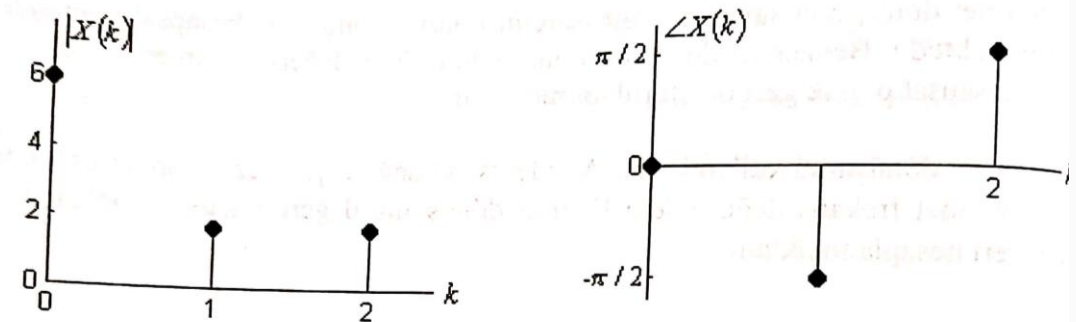
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Şekil 4. 9. Örnek 4.12 için ayrık Fourier dönüşümünün genliği ve fazı.

# 4-pt DFT: Numerical Example

- Take the 4-pt DFT of the following signal

$$x[n] = \delta[n] + \delta[n-1] \quad \{x[n]\} = [1, 1, 0, 0]$$

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$$\begin{aligned} X[3] &= x[0]e^{-j0} + x[1]e^{-j3\pi/2} + x[2]e^{-j3\pi} + x[3]e^{-j9\pi/2} \\ &= 1 + j = \sqrt{2}e^{j\pi/4} \end{aligned}$$



# 4-pt iDFT: Numerical Example

## Example 66-8: Short-Length IDFT

The 4-point DFT in Example 66-7 is  $X[k] = \{2, \sqrt{2}e^{-j\pi/4}, 0, \sqrt{2}e^{j\pi/4}\}$ . If we compute the 4-point IDFT of the sequence  $X[k]$ , we should recover  $x[n]$  when we apply the IDFT summation (66.52) for each value of  $n = 0, 1, 2, 3$ . As before, the exponents in (66.52) will all be integer multiples of  $\pi/2$  when  $N = 4$ .

$$x[0] = \frac{1}{4} \left( X[0]e^{j0} + X[1]e^{j0} + X[2]e^{j0} + X[3]e^{j0} \right)$$

$$= \frac{1}{4} \left( 2 + \sqrt{2}e^{-j\pi/4} + 0 + \sqrt{2}e^{j\pi/4} \right) = 1$$

$$x[1] = \frac{1}{4} \left( X[0]e^{j0} + X[1]e^{j\pi/2} + X[2]e^{j\pi} + X[3]e^{j3\pi/2} \right)$$

$$= \frac{1}{4} \left( 2 + \sqrt{2}e^{j(-\pi/4+\pi/2)} + 0 + \sqrt{2}e^{j(\pi/4+3\pi/2)} \right) = \frac{1}{4}(2 + (1 + j) + (1 - j)) = 1$$

$$x[2] = \frac{1}{4} \left( X[0]e^{j0} + X[1]e^{j\pi} + X[2]e^{j2\pi} + X[3]e^{j3\pi} \right)$$

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$$x[3] = \frac{1}{4} \left( X[0]e^{j0} + X[1]e^{j3\pi/2} + X[2]e^{j3\pi} + X[3]e^{j9\pi/2} \right)$$

$$= \frac{1}{4} \left( 2 + \sqrt{2}e^{j(-\pi/4+3\pi/2)} + 0 + \sqrt{2}e^{j(\pi/4+9\pi/2)} \right) = \frac{1}{4}(2 + (-1 - j) + (-1 + j)) = 0$$

Thus we recover the signal  $x[n] = \{1, 1, 0, 0\}$  from its DFT coefficients,  $X[k] = \{2, \sqrt{2}e^{-j\pi/4}, 0, \sqrt{2}e^{j\pi/4}\}$ .

# DFT Properties



**Table 8-2** Basic discrete Fourier transform properties.

Table of DFT Properties		
Property Name	Time-Domain: $x[n]$	Frequency-Domain: $X[k]$
Periodic	$x[n] = x[n + N]$	$X[k] = X[k + N]$
Linearity	$ax_1[n] + bx_2[n]$	$aX_1[k] + bX_2[k]$
Conjugate Symmetry	$x[n]$ is real	$X[N - k] = X^*[k]$
Conjugation	$x^*[n]$	$X^*[N - k]$
Time-Reversal	$x[((N - n))_N]$	$X[N - k]$
Delay	$x[((n - n_d))_N]$	$e^{-j(2\pi k/N)n_d} X[k]$
Frequency Shift	$x[n]e^{j(2\pi k_0/N)n}$	$X[k - k_0]$
Modulation	$x[n] \cos((2\pi k_0/N)n)$	$\frac{1}{2}X[k - k_0] + \frac{1}{2}X[k + k_0]$
Convolution	$\sum_{m=0}^{N-1} h[m]x[((n - m))_N]$	$H[k]X[k]$
Parseval's Theorem	$\sum_{n=0}^{N-1}  x[n] ^2 = \frac{1}{N} \sum_{k=0}^{N-1}  X[k] ^2$	

# DFT Properties



Özellik Adı	İşaret $x_1[n], x_2[n]$	N noktalı Ayırık Fourier Dönüşümü $X_1[k], X_2[k]$
Periyodiklik	$x_1[n] = x_1[n + N]$	$X_1[k] = X_1[k + N]$
Zamanda Tersleme	$x_1[-n] = x_1[N - n]$	$X_1[-k] = X_1[N - k]$
Lineerlik	$ax_1[n] + bx_2[n]$	$aX_1[k] + bX_2[k]$
Konjuge simetriği	$x[n]$ , reel ise	$X[N - k] = X^*[k]$
Çifteşlik	$X[n]$	$Nx[-k]_{mod\ N}$
Dairesel öteleme	$x_1[n - n_0]_{mod\ N}$ , $n_0$ tamsayı	$e^{-j(2\pi k/N)n_0}X[k]$
Frekansta dairesele öteleme	$e^{j(\frac{2\pi k}{N})l}x_1[n]$ , $l$ tamsayı	$X_1[k - l]_{mod\ N}$
Dairesel Konvolüsyon	$\sum_{m=0}^{N-1} x_1[m]x_2[n - m]_{mod\ N}$	$X_1[k]X_2[k]$
Dairesel Modülasyon	$x_1[n]x_2[n]$	$\frac{1}{N} \sum_{l=0}^{N-1} X_1[l]X_2[k - l]_{mod\ N}$

# DFT periodic in k (frequency domain)

- Since DTFT is periodic in frequency, the DFT must also be periodic in k

$$X[k] = X(e^{j(2\pi/N)k})$$

$$X[k + N] = X(e^{j(2\pi/N)(k+N)}) = X(e^{j(2\pi/N)(k) + (2\pi/N)N}) = X(e^{j(2\pi/N)k})$$

- What about Negative indices and Conjugate Symmetry?

$$X(e^{-j(2\pi/N)k}) = X^*(e^{j(2\pi/N)k})$$

$$\Rightarrow X[-k] = X^*[k]$$

$$X[N - k] = X^*[k]$$

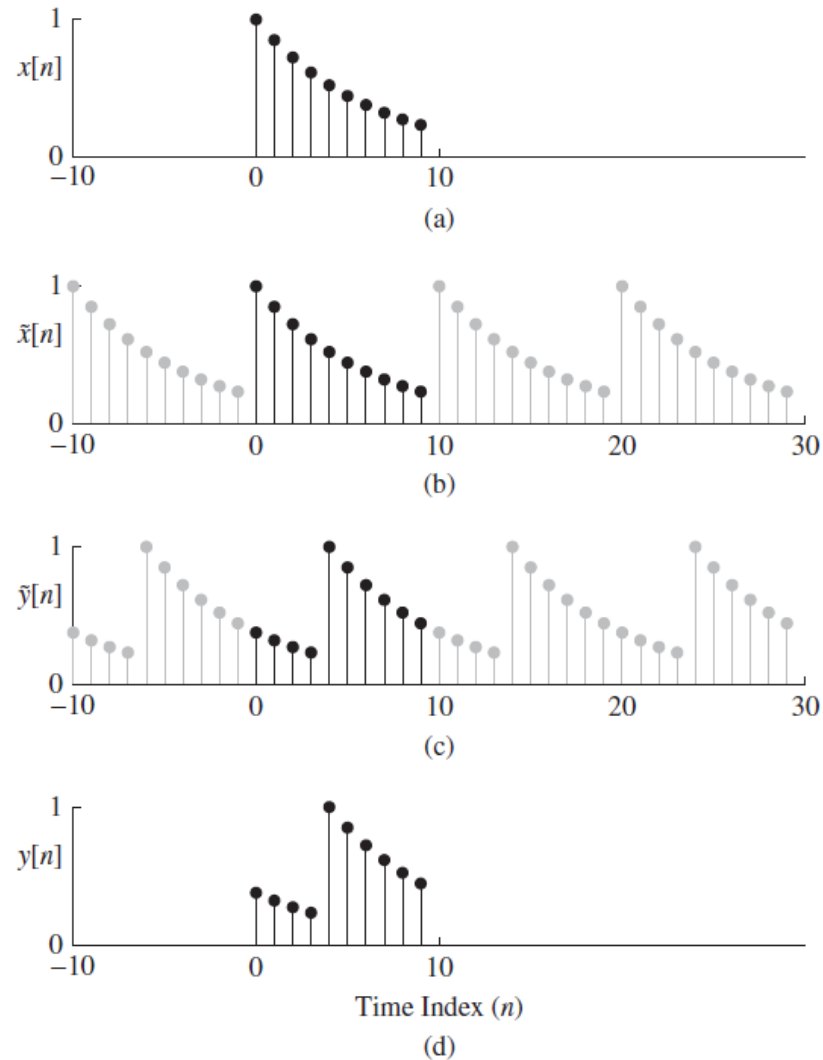
$$N = 32 \Rightarrow$$

$$X[31] = X^*[1]$$

$$X[30] = X^*[2]$$

$$X[29] = X^*[3]$$

# Circular Convolution for Periodic DT Signals



**Figure 8-8** Illustration of the time-shift property of the DFT. (a) A finite-length sequence  $x[n]$  of length 10. (b) The inherent periodic sequence  $\tilde{x}[n]$  for a 10-point DFT representation. (c) Time-shifted periodic sequence  $\tilde{y}[n] = \tilde{x}[n - 4]$  which is also equal to the IDFT of  $Y[k] = e^{-j(2\pi k/10)(4)} X[k]$ . (d) The sequence  $y[n]$  obtained by evaluating the 10-point IDFT of  $Y[k]$  only in the interval  $0 \leq n \leq 9$ .

# Circular Convolution for Periodic DT Signals

$x_1[n] = [1 \ 2 \ 3 \ 4]$  ve  $x_2[n] = [1 \ 1]$  ise bu iki işaretin dairesel konvolüsyonunu bulunuz.

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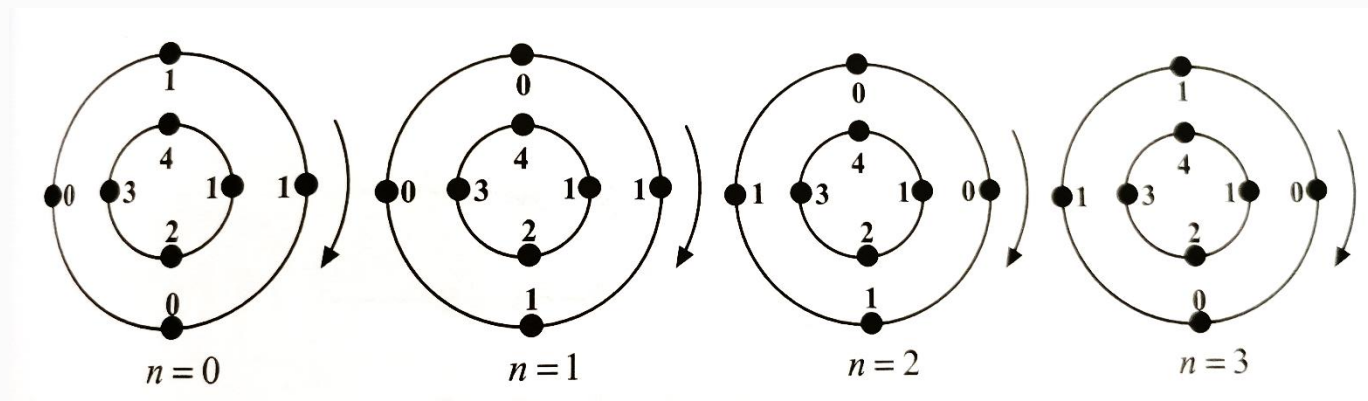
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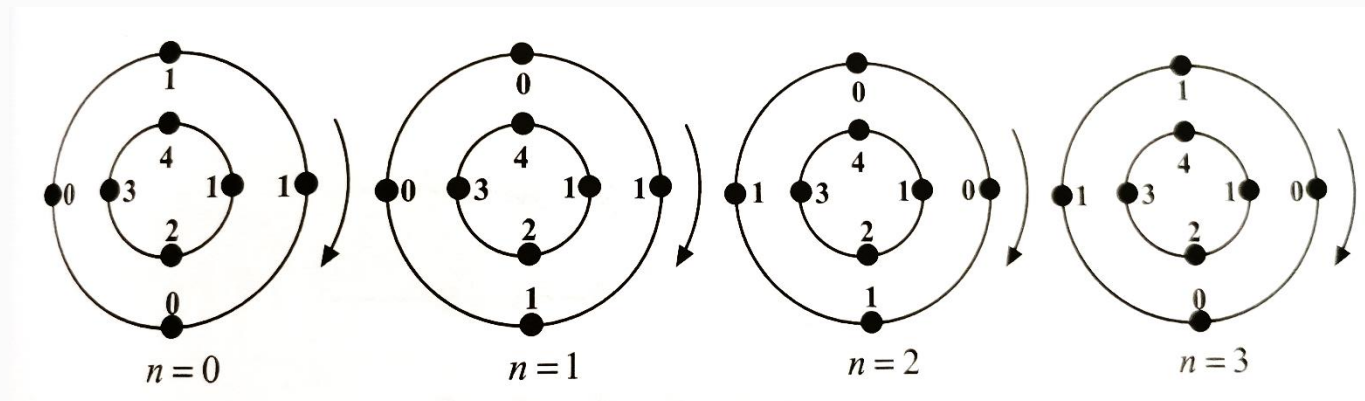
Tamamı hesaplanırsa :  $x_3[n] = [5 \ 3 \ 5 \ 7]$

# Circular Convolution



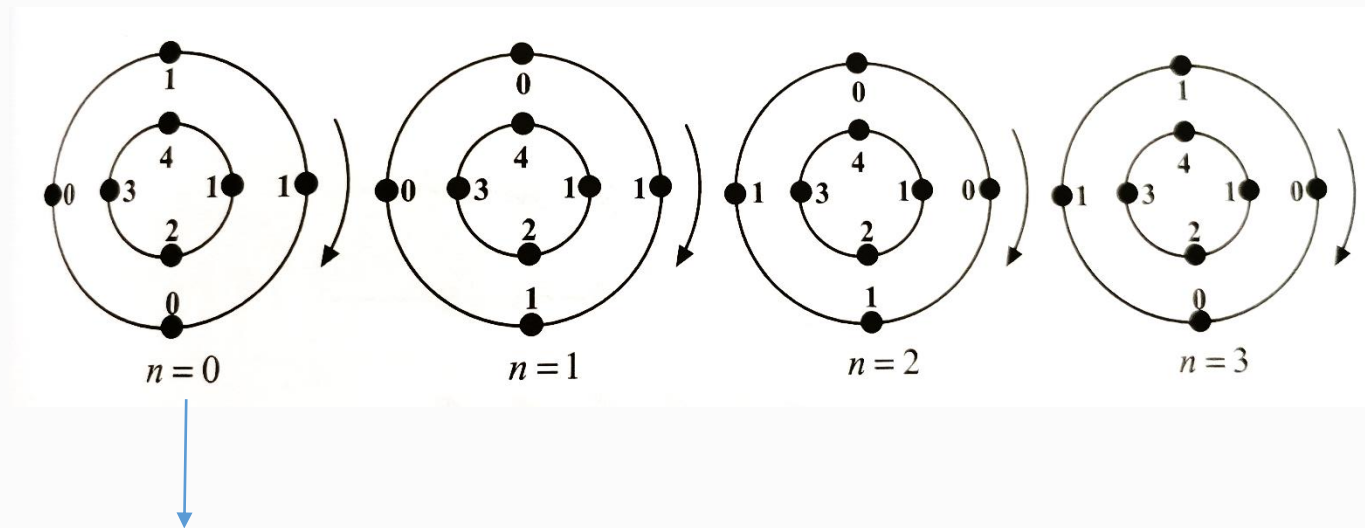
# Circular Convolution

Or...



# Circular Convolution

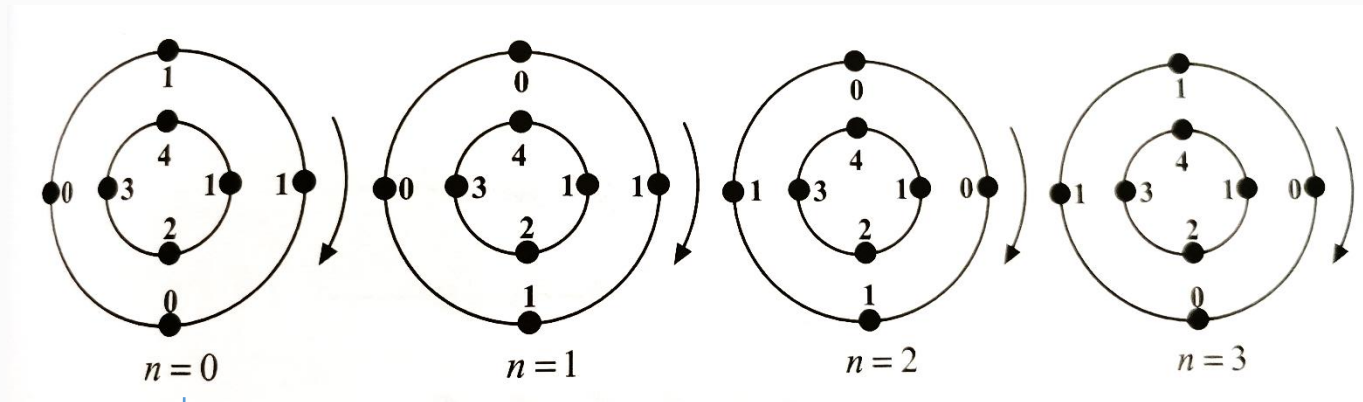
Or...



$$x_3[0] = 1 \times 1 + 4 \times 1 = 5$$

# Circular Convolution

Or...

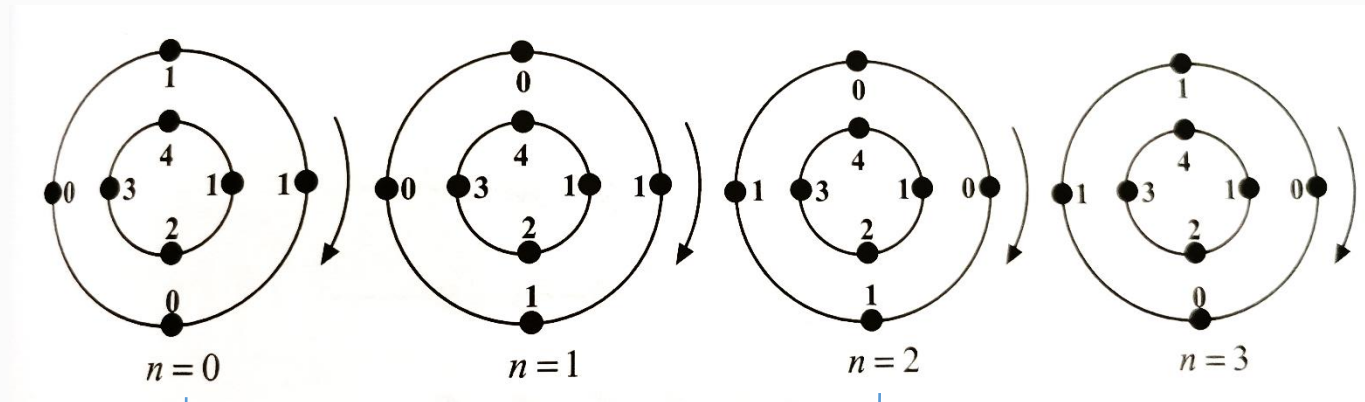


$$x_3[0] = 1 \times 1 + 4 \times 1 = 5$$

$$x_3[1] = 3$$

# Circular Convolution

Or...



$$x_3[0] = 1 \times 1 + 4 \times 1 = 5$$

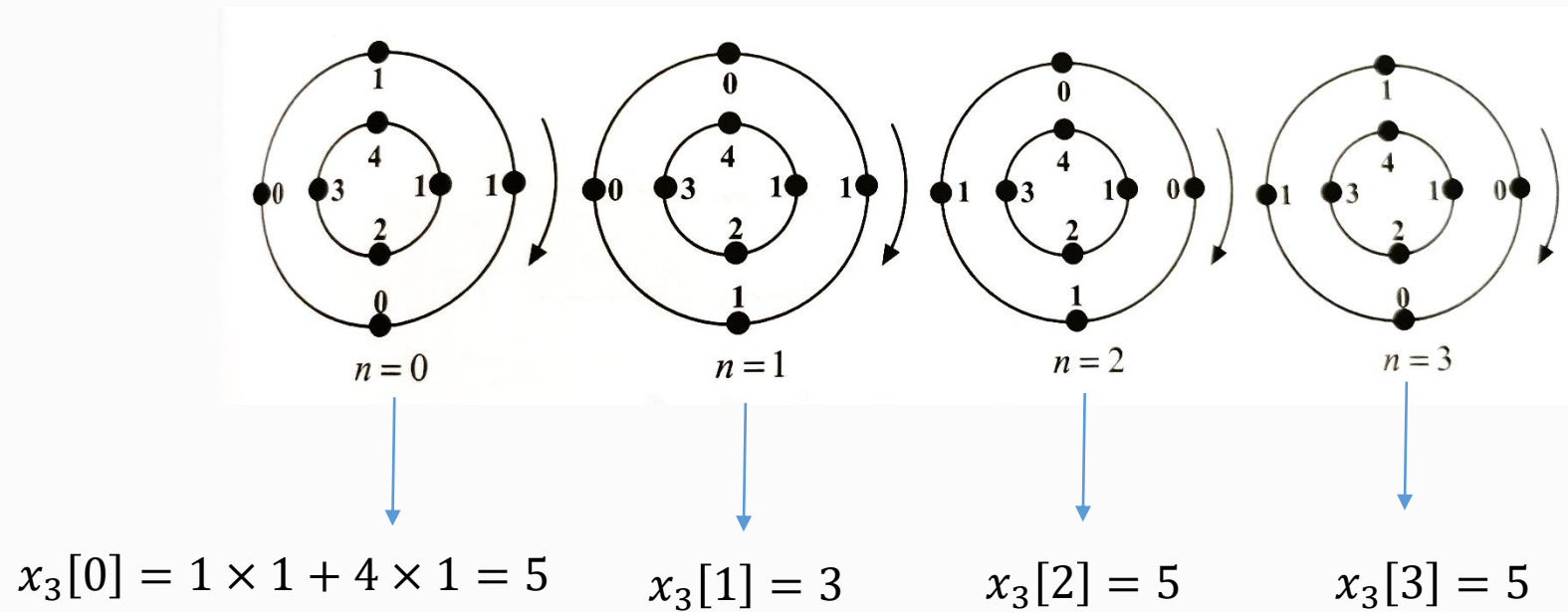
$$x_3[1] = 3$$

$$x_3[2] = 5$$



# Circular Convolution

Or...



$$x[n] \leftrightarrow X[k] \quad \text{ise} \quad X[n] \leftrightarrow Nx[-k]_{\text{mod } N}$$

$x[n] = 2\delta[n] + 3\delta[n-1] + \delta[n-2]$  işaretinin AFDsi  $X[k] = [6 \ -1.7321j \ +1.732j]$  olduğu biliniyor.

Buna göre,  $x[n] = [6 \ -1.7321j \ +1.732j]$  işaretinin Ayırık Fourier Dönüşümü  $X[k]$  nedir?

$$= x[3-n]_{\text{mod } 3} \Rightarrow [2 \ 1 \ 3]$$

Buradan:

$$X[n] \leftrightarrow Nx[-k]_{\text{mod } N} = 3 \times [2 \ 1 \ 3] = [6 \ 3 \ 9]$$

olacaktır.

# Duality (Çifteşlik)



$$x[n] \leftrightarrow X[k] \quad \text{ise} \quad X[n] \leftrightarrow Nx[-k]_{\text{mod } N}$$

$$]3 \ 1 \ 2[ \ x \ 3 - n \ \text{mod } 3 \Rightarrow 3nn \ 3 - n \ x \ 3 - n \ \text{mod } 3 \ mmoodd - x \ 3 - n \ \text{mod } 3 \ xx \ 3 - n \ 3 \ x - k \ \text{mod}$$

Buna göre,  $x[n] = [6 -1.7321j +1.732j]$  işaretinin Ayırık Fourier Dönüşümü  $X[k]$  nedir?

$$x - k - k \ \text{mod } 3 = x[3 - n]_{\text{mod } 3} \Rightarrow [2 \ 1 \ 3]$$

Buradan:

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$$Nx[-k]_{\text{mod } N} = N \times x[-k]_{\text{mod } N} = 3 \times [2 \ 1 \ 3 \ 2 \ 1 \ 3 \ 2 \ 1 \ 3] = [6 \ 3 \ 9 \ 6 \ 3 \ 9 \ 6 \ 3 \ 9]$$

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olacaktır.

# Örnek – Konjuge Simetri:



Konjuge simetriği

$x[n]$ , reel ise

$$X[N - k] = X^*[k]$$

$x[n] = [2 \ 1 \ 2 \ 0 \ 1 \ 1]$  işaretinin Ayırık Fourier Dönüşümünü simetri özelliğinden yararlanarak hesaplayalım.

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$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j 2\pi N kn} \\ = \sum_{n=0}^{N-1} x[n] e^{-j 2\pi N kn} = \sum_{n=0}^{N-1} x^*[n] e^{j 2\pi N kn} = X^*[k]$$

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olmaktadır. Eğer  $X[1]$ ,  $X[2]$  bilinirse  $X[4]$  ve  $X[5]$  i hesaplamadan bulabiliriz.

$x[n] = [2 \ 1 \ 2 \ 0 \ 1 \ 1]$  işaretinin Ayırık Fourier Dönüşümünü simetri özelliğinden yararlanarak hesaplayalım.

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$$X[4] = X^*[2] = 0.5 + j0.86$$

$$X[5] = X^*[1] = 1.5 + j0.86$$

$$k=1 \text{ için } \rightarrow X[5] = X^*[1]$$

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$$X[4] = X^*[2] = 0.5 + j0.86$$

$$X[5] = X^*[1] = 1.5 + j0.86$$

$$k=1 \text{ için } \rightarrow X[5] = X^*[1]$$

$$X[k] = [7 \quad 1.5 - j0.86 \quad 0.5 - j0.86 \quad 3 \quad 0.5 + j0.86 \quad 1.5 + j0.86]$$

$$k=2 \text{ için } \rightarrow X[4] = X^*[2]$$

# MATLAB Code for DFT



```
clc; clear all;  
%%  
x = [1 2 2 1 1 2 3 4];  
N = length(x);  
X = zeros(1,N);
```

```
for k = 0:N-1  
    for n = 0:N-1  
        X(k+1) = X(k+1) + x(n+1)*exp(-j*(2*pi/N)*k*n);  
    end  
end
```

```
X  
fft(x)
```

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(2\pi/N)kn}$$

# Matrix Form for N-pt DFT

- In MATLAB, NxN DFT matrix is **`dfmtx(N)`**
  - Obtain DFT by **`X = dfmtx(N) * x`**
  - Or, more efficiently by **`X = fft(x, N)`**

$$\begin{bmatrix} X[0] \\ X[1] \\ X[2] \\ \vdots \\ X[N-1] \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & e^{-j2\pi/N} & e^{-j4\pi/N} & \dots & e^{-j2(N-1)\pi/N} \\ 1 & e^{-j4\pi/N} & e^{-j8\pi/N} & \dots & e^{-j4(N-1)\pi/N} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & e^{-j2(N-1)\pi/N} & e^{-j4(N-1)\pi/N} & \dots & e^{-j2(N-1)(N-1)\pi/N} \end{bmatrix}}_{\text{DFT matrix}} \begin{bmatrix} x[0] \\ x[1] \\ x[2] \\ \vdots \\ x[N-1] \end{bmatrix}$$

Signal vector

# Understanding DFT Matrix



$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(2\pi/N)kn}$$

# Understanding DFT Matrix

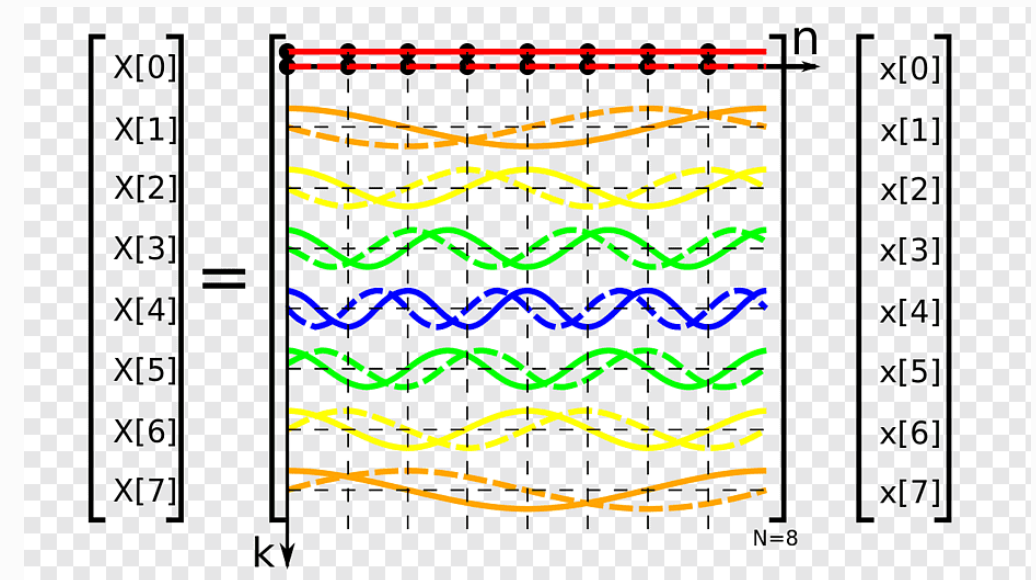
$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j(2\pi/N)kn}$$

$$R = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & e^{-2\pi i/4} & e^{-4\pi i/4} & e^{-6\pi i/4} \\ 1 & e^{-4\pi i/4} & e^{-8\pi i/4} & e^{-12\pi i/4} \\ 1 & e^{-6\pi i/4} & e^{-12\pi i/4} & e^{-18\pi i/4} \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -i & -1 & i \\ 1 & -1 & 1 & -1 \\ 1 & i & -1 & -i \end{pmatrix}$$

# Understanding DFT Matrix

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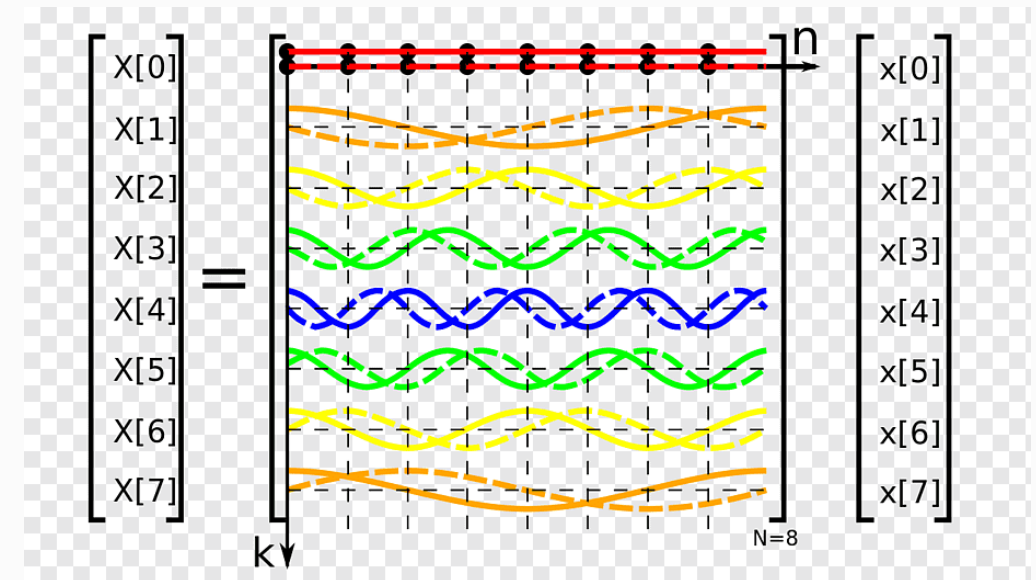
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```
n = 0:0.01:5; x = cos(1.8*pi*n);
N = length(x); X = zeros(1,N);
```

```
DFTM = dftmtx(N);
figure(1);
for i = 1:8
    plot(real(DFTM(i,:))); hold on;
    plot(x, 'r'); hold off; pause;
end

figure(2); plot(abs(fft(x,N)));
```





# Example about Conv. Property



Given  $x = [1 \ 1 \ 0 \ 0]$  and  $h = [0 \ 0 \ 1 \ 1]$ , compute the output by using convolution property.

Convolution	$\sum_{m=0}^{N-1} h[m]x[((n - m))_N]$	$H[k]X[k]$
-------------	---------------------------------------	------------

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-------------	---------------------------------------	------------

- 1- Compute  $H[k]$  using 4-pt DFT,
- 2- Compute  $X[k]$  using 4-pt DFT,
- 3- Product them in freq. Domain,
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$$x = [1 \ 2 \ 3 \ 4];$$

$$h = [1 \ 1 \ 0 \ 0];$$

$$X = \text{myDFT}(x);$$

$$H = \text{myDFT}(h);$$

$$Y = X .* H;$$

$$y = \text{myIDFT}(Y)$$

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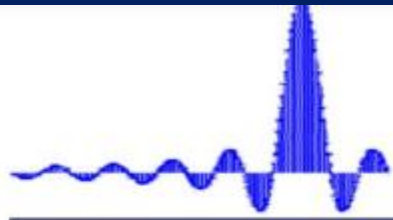
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$$\begin{aligned} x &= [1 \ 2 \ 3 \ 4] ; \\ h &= [1 \ 1 \ 0 \ 0] ; \end{aligned}$$

Circular convolution!

$$\begin{aligned} X &= \text{myDFT}(x) ; \\ H &= \text{myDFT}(h) ; \end{aligned}$$

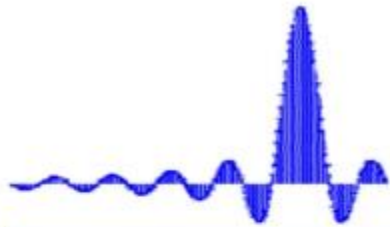
$$\begin{aligned} Y &= X .* H ; \\ y &= \text{myIDFT}(Y) \end{aligned}$$



# LINEAR VS. CIRCULAR CONVOLUTION



- ➔ Note that the results of linear and circular convolution are different. This is a problem! Why?
- ➔ All LTI systems are based on the principle of linear convolution, as the output of an LTI system is the linear convolution of the system impulse response and the input to the system, which is equivalent to the product of the respective DTFTs in the frequency domain.
  - ↳ However, if we use DFT instead of DTFT (so that we can compute it using a computer), then the result appear to be invalid:
    - DTFT is based on linear convolution, and DFT is based on circular convolution, and they are not the same!!!
    - For starters, they are not even of equal length: For two sequences of length  $N$  and  $M$ , the linear convolution is of length  $N+M-1$ , whereas circular convolution of the same two sequences is of length  $\max(N,M)$ , where the shorter sequence is zero padded to make it the same length as the longer one.
    - Is there any relationship between the linear and circular convolutions? Can one be obtained from the other? OR can they be made equivalent?



## ***LINEAR VS. CIRCULAR CONVOLUTION***

⇒ YES!, rather easily, as a matter of fact!

- ✦ **FACT:** If we *zero pad* both sequences  $x[n]$  and  $h[n]$ , so that they are both of length  $N+M-1$ , then linear convolution and circular convolution result in identical sequences
- ✦ **Furthermore:** If the respective DFTs of the zero padded sequences are  $X[k]$  and  $H[k]$ , then the inverse DFT of  $X[k] \cdot H[k]$  is equal to the linear convolution of  $x[n]$  and  $h[n]$
- ✦ Note that, normally, the inverse DFT of  $X[k] \cdot H[k]$  is the circular convolution of  $x[n]$  and  $h[n]$ . If they are zero padded, then the inverse DFT is the linear convolution of the two.

# With Zero Padding

Conv. Length =  $N + M - 1 \rightarrow CL = 2*N-1$ , Zero-Pad signals with  $N+1$

If  $N = 4$ , then

```
x = [1 2 3 4 0 0 0 0 0];
```

```
h = [1 1 0 0 0 0 0 0 0];
```

```
X = myDFT(x);
```

```
H = myDFT(h);
```

```
Y = X.*H;
```

```
y = real(myIDFT(Y))
```

```
conv(x,h)
```

# DFT of an Image



The 2D DFT is

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$



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Fourier basis element  
 $e^{-j2\pi(ux+vy)}$

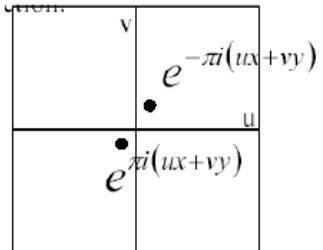
example, real part

$F^{u,v}(x, y)$

$F^{u,v}(x, y) = \text{const. for } (ux+vy) = \text{const.}$

Vector  $(u, v)$

- Magnitude gives frequency
- Direction gives orientation.



Slide credit: S. Thrun

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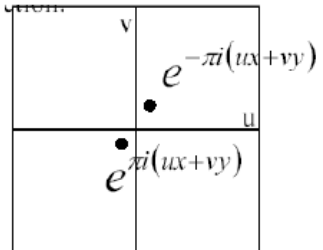
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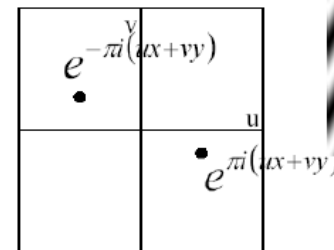
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- Direction gives orientation.



Slide credit: S. Thrun

Here  $u$  and  $v$  are larger than in the previous slide.

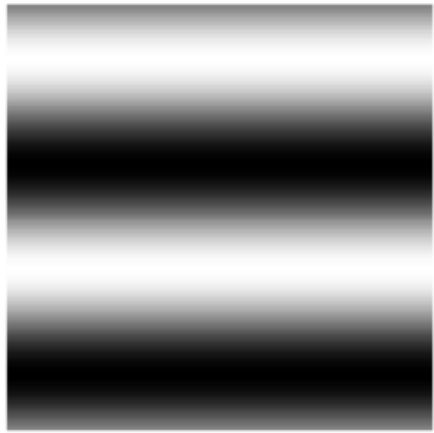


Slide credit: S. Thrun

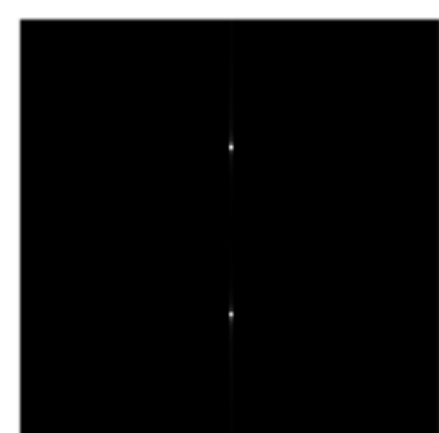
# DFT of an Image

The 2D DFT is

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$



Sinusoid with frequency = 1 and its FFT

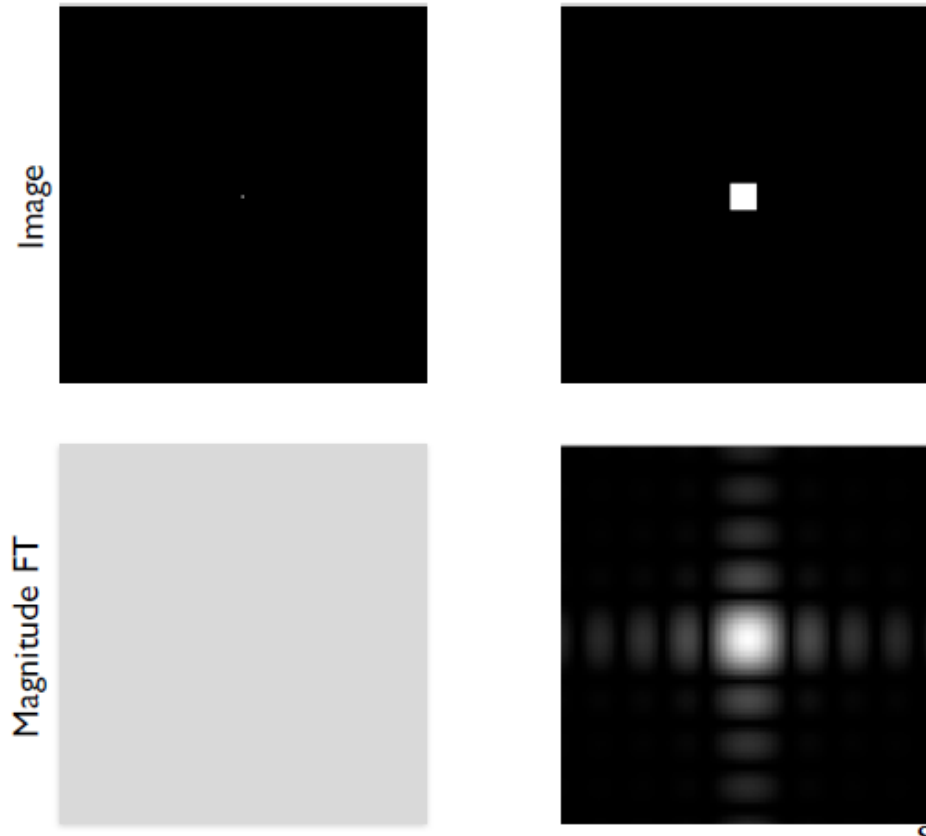


Sinusoid with frequency = 10 and its FFT

# DFT of an Image

The 2D DFT is

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$

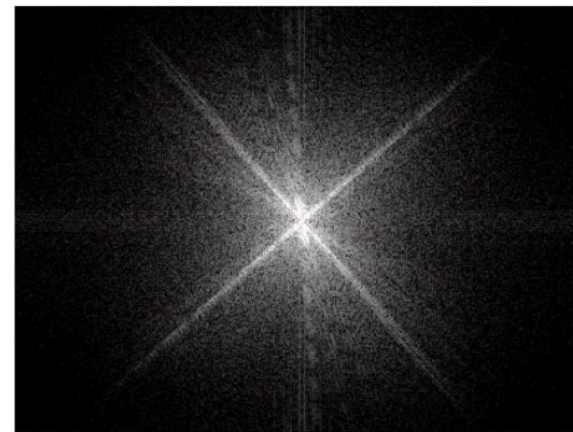
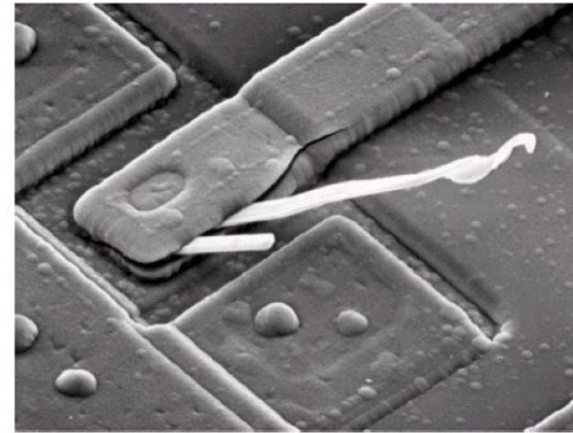


# DFT of an Image

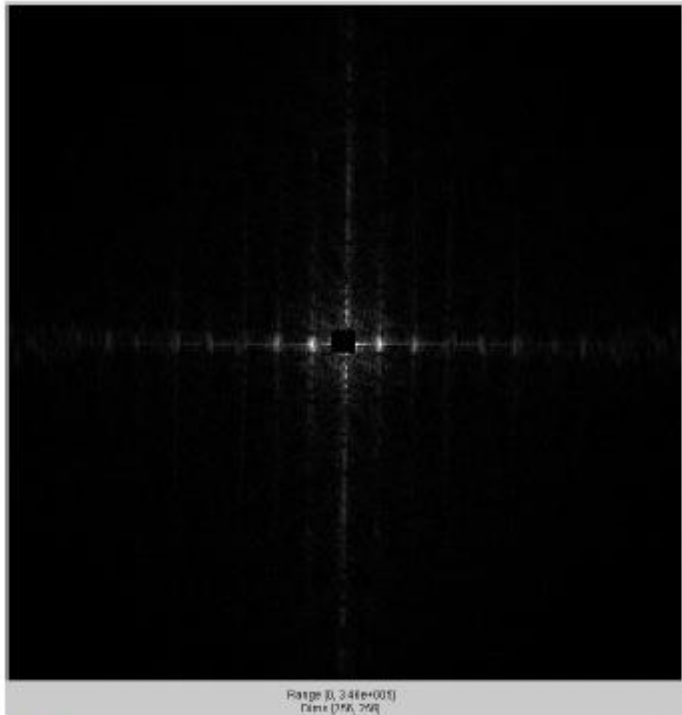
The 2D DFT is

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$

```
I = double(imread('moon.tif'));  
imshow(I, []);  
F = fft2(I);  
figure, imshow(log(abs(fftshift(F))),[]);
```



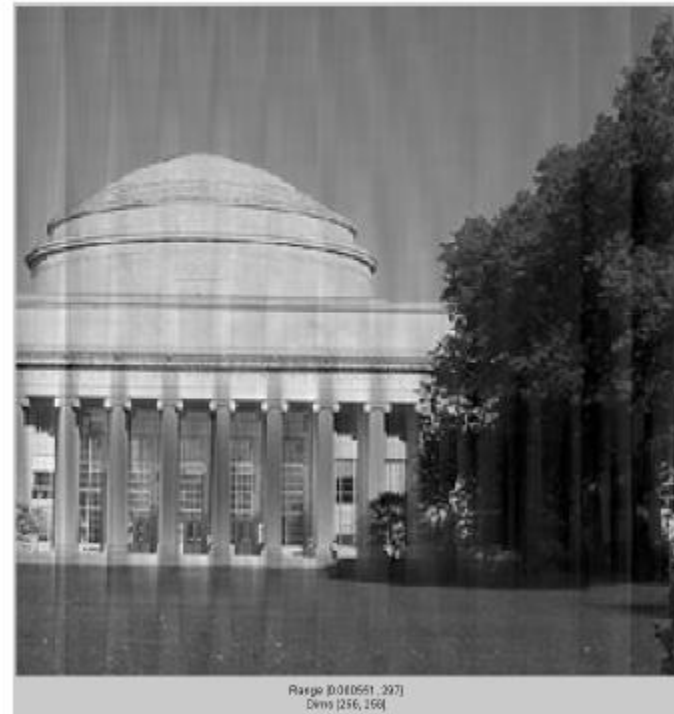
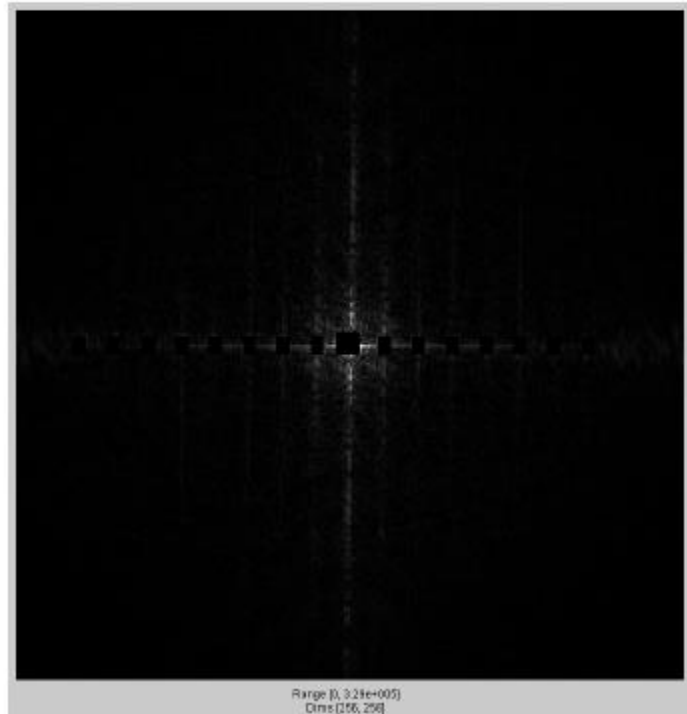
# Pop-up Quiz



What in the image causes the dots?

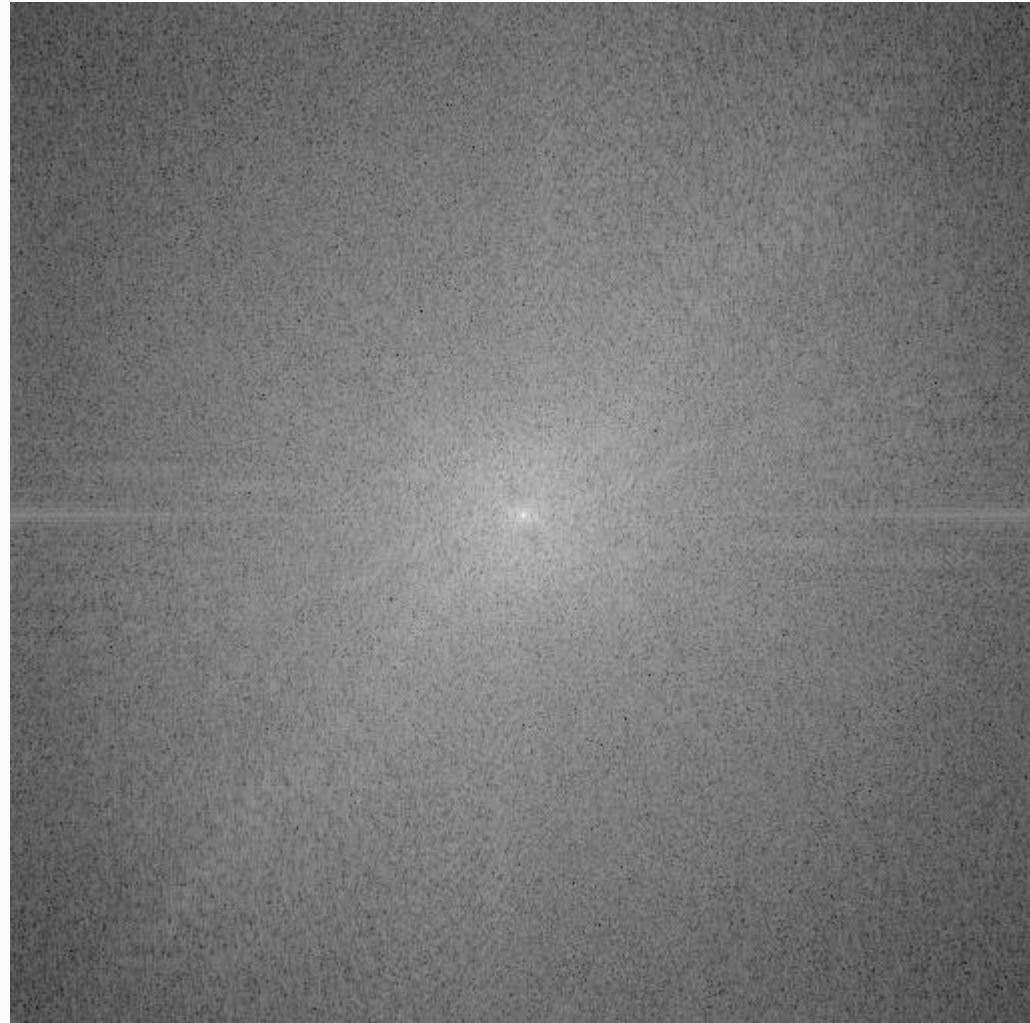
Slide credit: B. Freeman and A. Torralba

## Masking out the fundamental and harmonics from periodic pillars



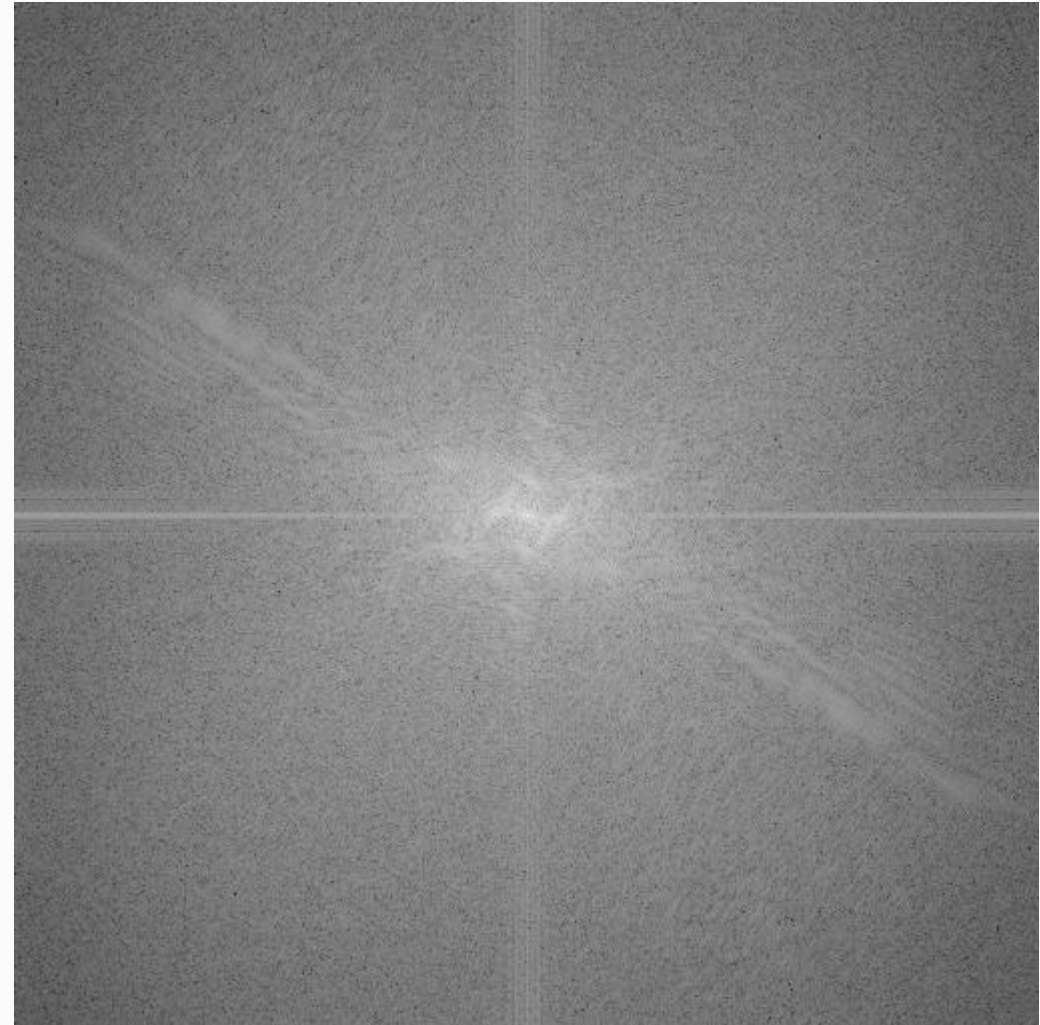


# An Interesting Experiment: Cheetah vs Zebra





# An Interesting Experiment: Cheetah vs Zebra



# Reconstruction with zebra phase, cheetah magnitude

