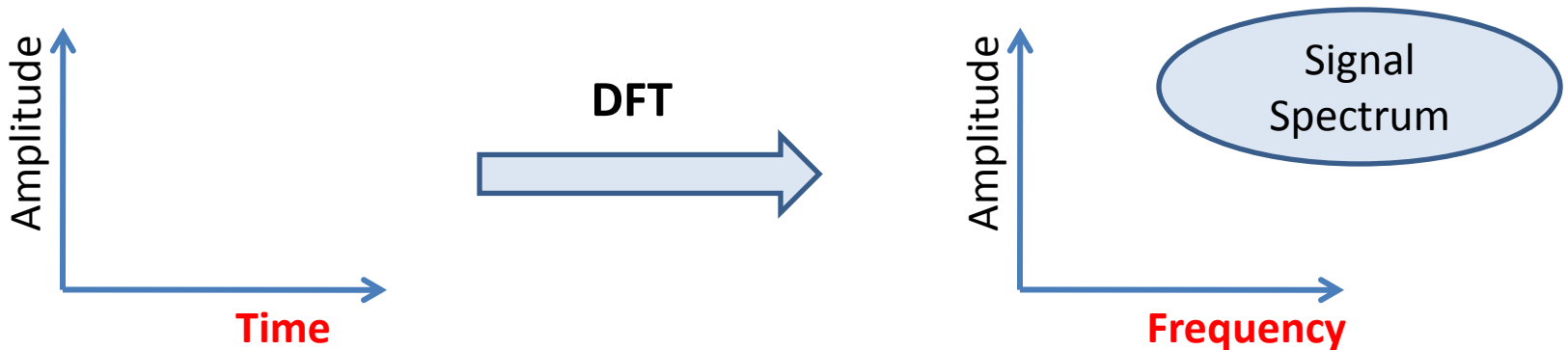




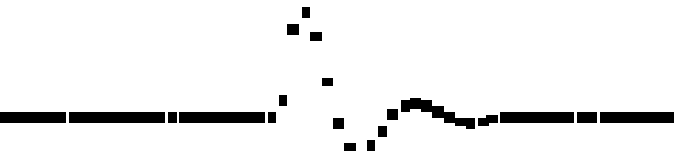
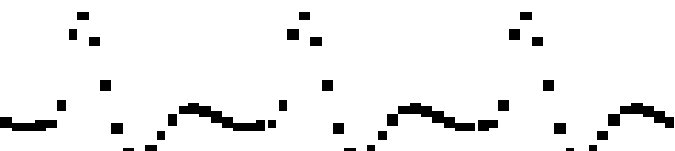
Discrete Fourier Transform (DFT)

DFT transforms the time domain signal samples to the frequency domain components.

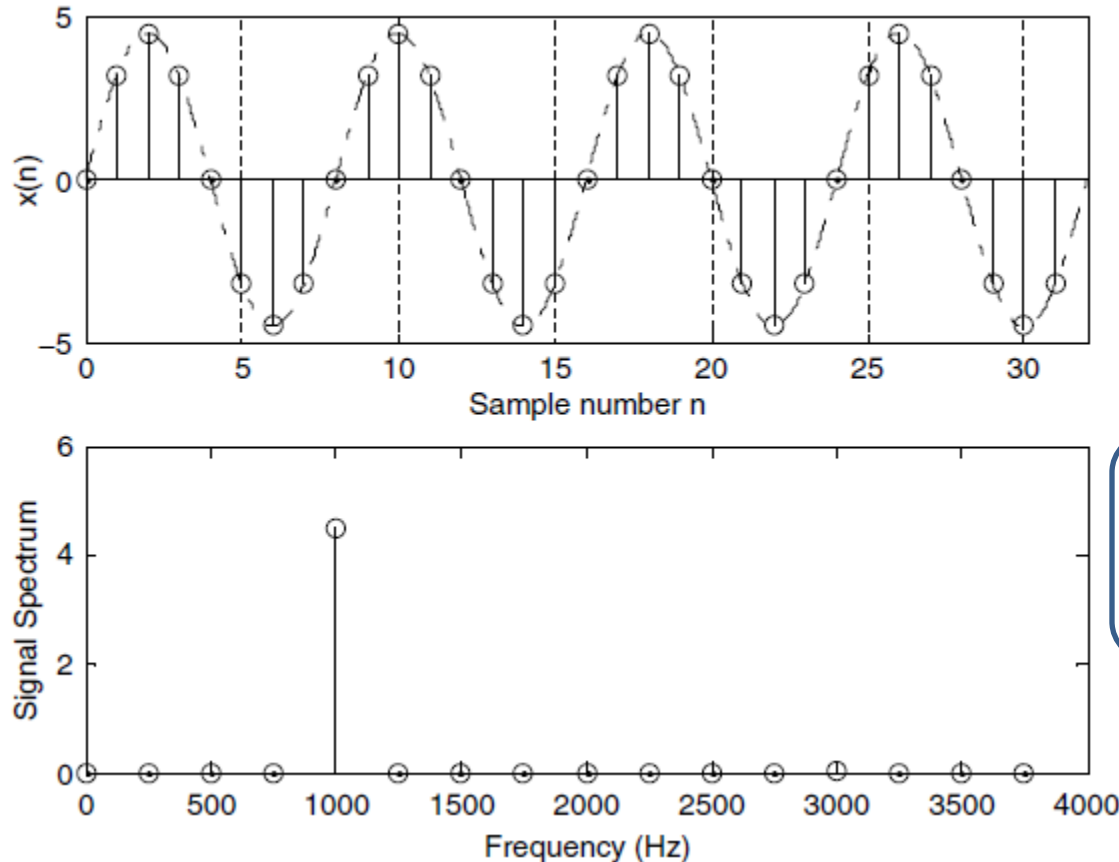


DFT is often used to do frequency analysis of a time domain signal.

Four Types of Fourier Transform

Type of Transform	Example Signal
Fourier Transform <i>signals that are continuous and aperiodic</i>	
Fourier Series <i>signals that are continuous and periodic</i>	
Discrete Time Fourier Transform <i>signals that are discrete and aperiodic</i>	
Discrete Fourier Transform <i>signals that are discrete and periodic</i>	

DFT: Graphical Example



1000 Hz sinusoid with
32 samples at 8000 Hz
sampling rate.

Sampling rate

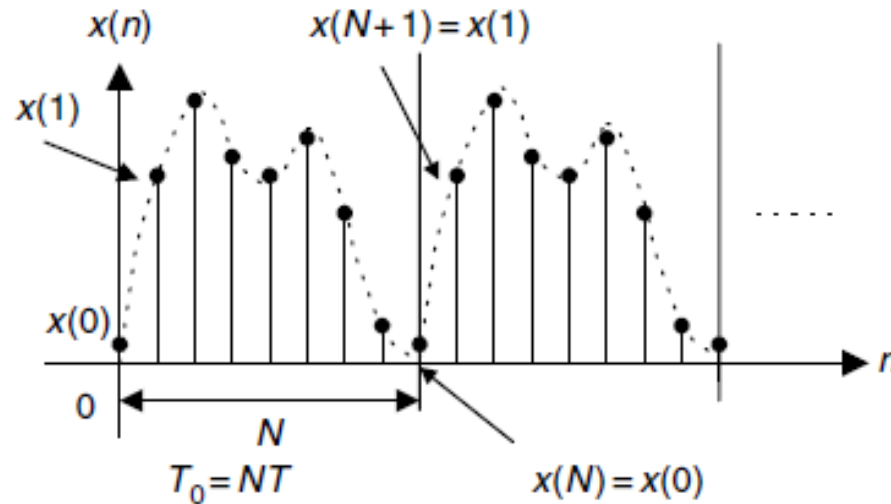
8000 samples = 1 second
32 samples = $32/8000$ sec
= 4 millisecond

Frequency

1 second = 1000 cycles
 $32/8000$ sec =
($1000 * 32/8000 =$) 4 cycles

DFT Coefficients of Periodic Signals

Periodic
Digital
Signal



Equation of DFT coefficients:
$$c_k = \frac{1}{N} \sum_{n=0}^{N-1} x(n) e^{-j \frac{2\pi kn}{N}}, \quad -\infty < k < \infty$$

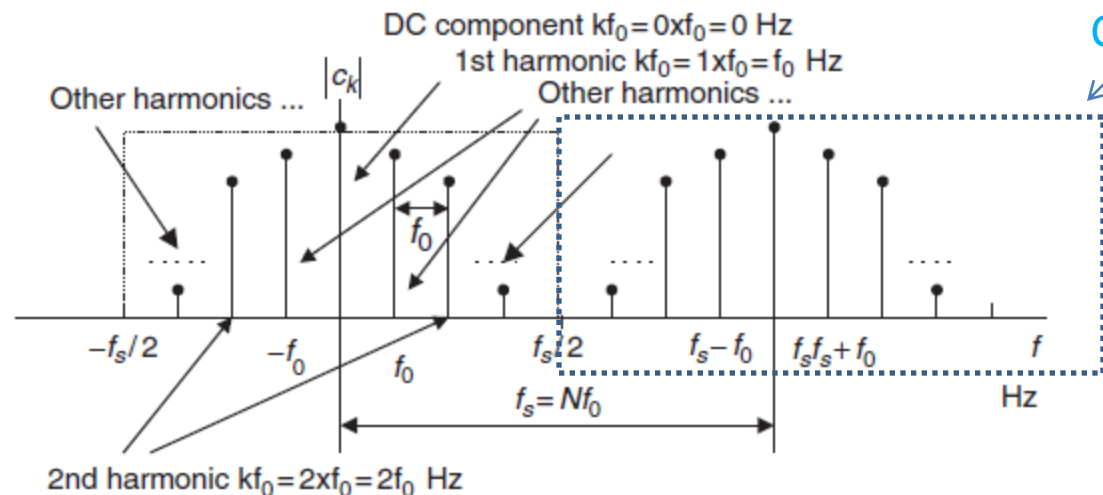
DFT Coefficients of Periodic Signals

Fourier series coefficient c_k is periodic of N

$$c_{k+N} = \frac{1}{N} \sum_{n=0}^{N-1} x(n) e^{-j \frac{2\pi(k+N)n}{N}} = \frac{1}{N} \sum_{n=0}^{N-1} x(n) e^{-j \frac{2\pi kn}{N}} e^{-j 2\pi n}$$

Since $e^{-j 2\pi n} = \cos(2\pi n) - j \sin(2\pi n) = 1$, $\Rightarrow c_{k+N} = c_k$.

Amplitude spectrum of the periodic digital signal



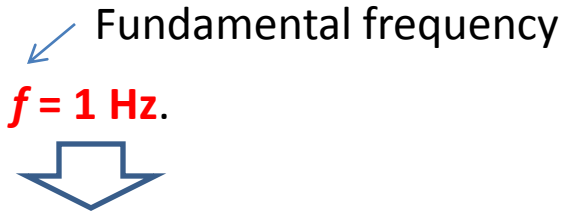
Example 1

The periodic signal: $x(t) = \sin(2\pi t)$ is sampled at $f_s = 4$ Hz

- Compute the spectrum c_k using the samples in one period.
- Plot the two-sided amplitude spectrum $|c_k|$ over the range from -2 to 2 Hz.

Solution:


a. We match $x(t) = \sin(2\pi t)$ with $x(t) = \sin(2\pi f t)$ and get $f = 1$ Hz.

Fundamental frequency


Therefore the signal has 1 cycle or 1 period in 1 second.

Sampling rate $f_s = 4$ Hz  1 second has 4 samples.

Hence, there are 4 samples in 1 period for this particular signal.

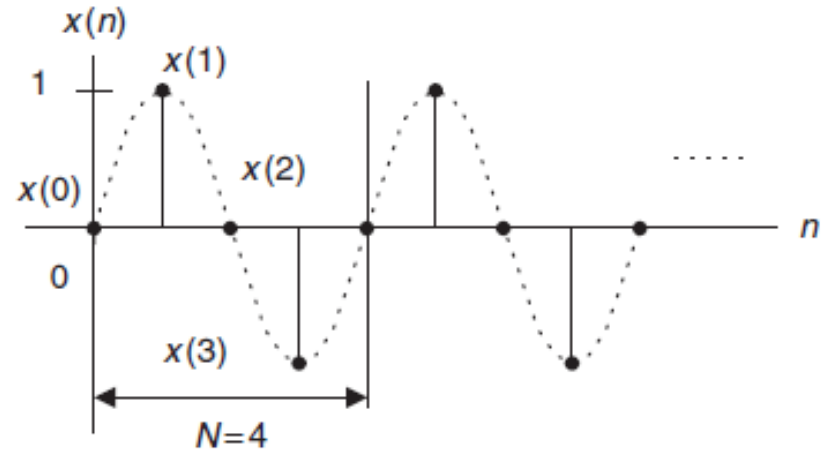
$T = 1/f_s = 0.25$  $x(n) = x(nT) = \sin(2\pi nT) = \sin(0.5\pi n)$

Example 1 - contd. (1)

$$x(0) = 0; x(1) = 1; x(2) = 0; \text{ and } x(3) = -1$$

b.

$$c_k = \frac{1}{N} \sum_{n=0}^{N-1} x(n)e^{-j\frac{2\pi kn}{N}}, \quad -\infty < k < \infty$$



$$c_0 = \frac{1}{4} \sum_{n=0}^3 x(n) = \frac{1}{4} (x(0) + x(1) + x(2) + x(3)) = \frac{1}{4} (0 + 1 + 0 - 1) = 0$$

$$c_1 = \frac{1}{4} \sum_{n=0}^3 x(n)e^{-j2\pi \times 1n/4} = \frac{1}{4} (x(0) + x(1)e^{-j\pi/2} + x(2)e^{-j\pi} + x(3)e^{-j3\pi/2})$$

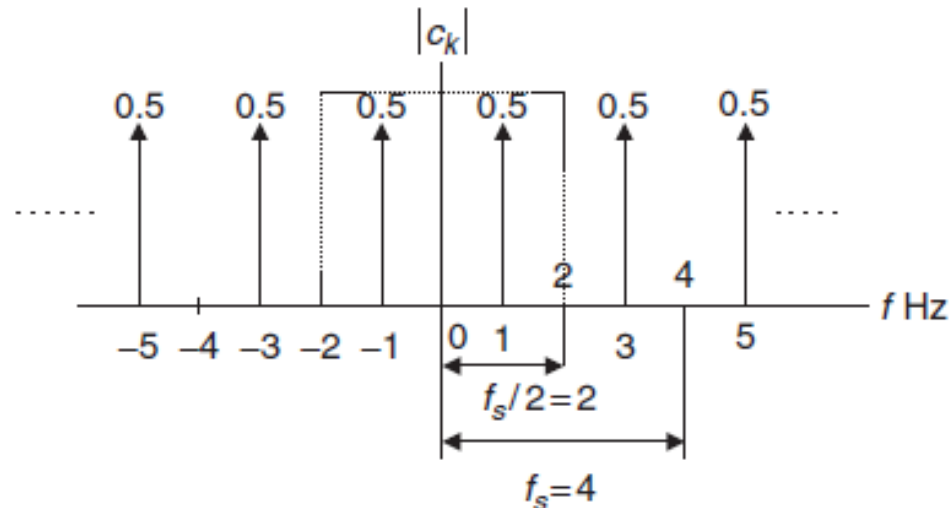
$$= \frac{1}{4} (x(0) - jx(1) - x(2) + jx(3)) = 0 - j(1) - 0 + j(-1) = -j0.5.$$

Example 1 - contd. (2)

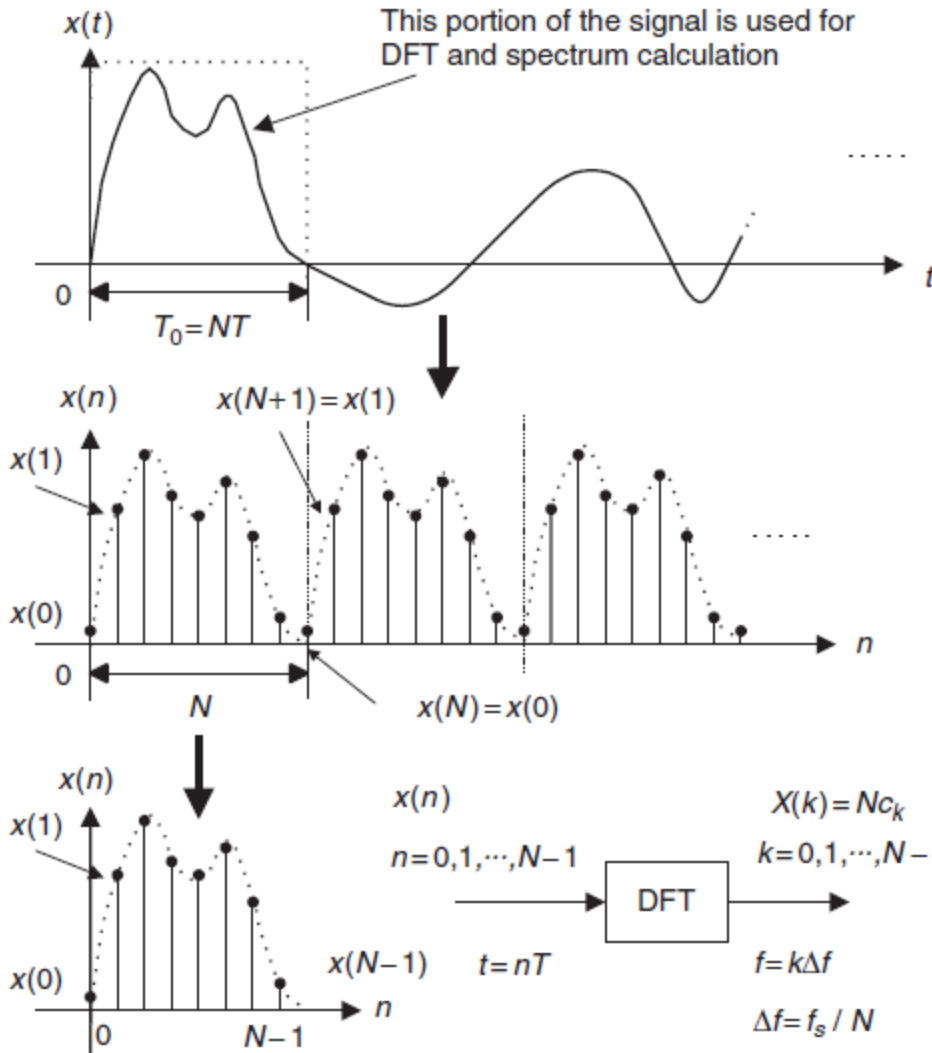
$$c_2 = \frac{1}{4} \sum_{k=0}^3 x(k) e^{-j2\pi \times 2k/4} = 0, \text{ and } c_3 = \frac{1}{4} \sum_{k=0}^3 x(k) e^{-j2\pi \times 3k/4} = j0.5.$$

Using periodicity, it follows that

$$c_{-1} = c_3 = j0.5, \text{ and } c_{-2} = c_2 = 0.$$



On the Way to DFT Formulas



Imagine periodicity of N samples. ←

Take first N samples (index 0 to $N - 1$) as the input to DFT.

DFT Formulas

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi kn/N}$$

$$= \sum_{n=0}^{N-1} x(n) W_N^{kn}, \text{ for } k = 0, 1, \dots, N-1.$$

$$X(k) = x(0) W_N^{k0} + x(1) W_N^{k1} + x(2) W_N^{k2} + \dots + x(N-1) W_N^{k(N-1)}$$

Where, $W_N = e^{-j2\pi/N} = \cos\left(\frac{2\pi}{N}\right) - j \sin\left(\frac{2\pi}{N}\right).$

Inverse DFT:

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j2\pi kn/N} = \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-kn}, \text{ for } n = 0, 1, \dots, N-1.$$

MATLAB Functions

FFT: Fast Fourier Transform

MATLAB FFT functions.

$X = \text{fft}(x)$	% Calculate DFT coefficients
$x = \text{ifft}(X)$	% Inverse DFT
$x = \text{input vector}$	
$X = \text{DFT coefficient vector}$	

Example 2

Given a sequence $x(n)$ for $0 \leq n \leq 3$, where $x(0) = 1$, $x(1) = 2$, $x(2) = 3$, and $x(3) = 4$,

a. Evaluate its DFT $X(k)$.

Solution:

$$N = 4 \text{ and } W_4 = e^{-j\frac{\pi}{2}} \quad \Rightarrow \quad X(k) = \sum_{n=0}^3 x(n)W_4^{kn} = \sum_{n=0}^3 x(n)e^{-j\frac{\pi kn}{2}}$$

Thus, for $k = 0$

$$\begin{aligned} X(0) &= \sum_{n=0}^3 x(n)e^{-j0} = x(0)e^{-j0} + x(1)e^{-j0} + x(2)e^{-j0} + x(3)e^{-j0} \\ &= x(0) + x(1) + x(2) + x(3) \\ &= 1 + 2 + 3 + 4 = 10 \end{aligned}$$

$$\begin{aligned} X(1) &= \sum_{n=0}^3 x(n)e^{-j\frac{\pi n}{2}} = x(0)e^{-j0} + x(1)e^{-j\frac{\pi}{2}} + x(2)e^{-j\pi} + x(3)e^{-j\frac{3\pi}{2}} \\ &= x(0) - jx(1) - x(2) + jx(3) \\ &= 1 - j2 - 3 + j4 = -2 + j2 \end{aligned}$$

Example 2 - contd.

$$\begin{aligned}X(2) &= \sum_{n=0}^3 x(n)e^{-j\pi n} = x(0)e^{-j0} + x(1)e^{-j\pi} + x(2)e^{-j2\pi} + x(3)e^{-j3\pi} \\ &= x(0) - x(1) + x(2) - x(3) \\ &= 1 - 2 + 3 - 4 = -2\end{aligned}$$

$$\begin{aligned}X(3) &= \sum_{n=0}^3 x(n)e^{-j\frac{3\pi n}{2}} = x(0)e^{-j0} + x(1)e^{-j\frac{3\pi}{2}} + x(2)e^{-j3\pi} + x(3)e^{-j\frac{9\pi}{2}} \\ &= x(0) + jx(1) - x(2) - jx(3) \\ &= 1 + j2 - 3 - j4 = -2 - j2\end{aligned}$$

Using MATLAB,

```
» X = fft([1 2 3 4])  
X = 10.0000 - 2.0000 + 2.0000i - 2.0000 - 2.0000 - 2.0000i
```

Example 3

Inverse DFT of the previous example.

$$N = 4 \text{ and } W_4^{-1} = e^{j\frac{\pi}{2}}, \quad \longrightarrow \quad x(n) = \frac{1}{4} \sum_{k=0}^3 X(k)W_4^{-nk} = \frac{1}{4} \sum_{k=0}^3 X(k)e^{j\frac{nk\pi}{2}}.$$

$$\begin{aligned} x(0) &= \frac{1}{4} \sum_{k=0}^3 X(k)e^{j0} = \frac{1}{4} (X(0)e^{j0} + X(1)e^{j0} + X(2)e^{j0} + X(3)e^{j0}) \\ &= \frac{1}{4} (10 + (-2 + j2) - 2 + (-2 - j2)) = 1 \end{aligned}$$

$$\begin{aligned} x(1) &= \frac{1}{4} \sum_{k=0}^3 X(k)e^{j\frac{k\pi}{2}} = \frac{1}{4} (X(0)e^{j0} + X(1)e^{j\frac{\pi}{2}} + X(2)e^{j\pi} + X(3)e^{j\frac{3\pi}{2}}) \\ &= \frac{1}{4} (X(0) + jX(1) - X(2) - jX(3)) \\ &= \frac{1}{4} (10 + j(-2 + j2) - (-2) - j(-2 - j2)) = 2 \end{aligned}$$

Example 3 - contd.

$$\begin{aligned}x(2) &= \frac{1}{4} \sum_{k=0}^3 X(k)e^{jk\pi} = \frac{1}{4} (X(0)e^{j0} + X(1)e^{j\pi} + X(2)e^{j2\pi} + X(3)e^{j3\pi}) \\ &= \frac{1}{4} (X(0) - X(1) + X(2) - X(3)) \\ &= \frac{1}{4} (10 - (-2 + j2) + (-2) - (-2 - j2)) = 3\end{aligned}$$

$$\begin{aligned}x(3) &= \frac{1}{4} \sum_{k=0}^3 X(k)e^{j\frac{k\pi}{2}} = \frac{1}{4} (X(0)e^{j0} + X(1)e^{j\frac{3\pi}{2}} + X(2)e^{j3\pi} + X(3)e^{j\frac{9\pi}{2}}) \\ &= \frac{1}{4} (X(0) - jX(1) - X(2) + jX(3)) \\ &= \frac{1}{4} (10 - j(-2 + j2) - (-2) + j(-2 - j2)) = 4\end{aligned}$$

Using MATLAB,

$$\begin{aligned}\gg x &= \text{ifft}([10 \quad -2 + 2j \quad -2 \quad -2 - 2j]) \\ x &= 1 \quad 2 \quad 3 \quad 4.\end{aligned}$$

Relationship Between Frequency Bin k and Its Associated Frequency in Hz

$$f = \frac{kf_s}{N} \text{ (Hz)}$$

Frequency step or frequency resolution: $\Delta f = \frac{f_s}{N}$ (Hz)

Example 4

In the previous example, if the sampling rate is 10 Hz,

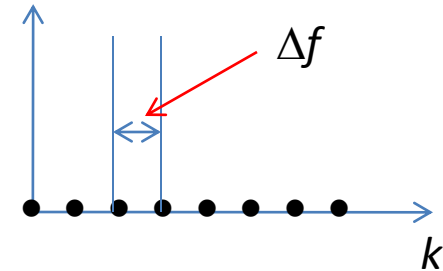
- Determine the sampling period, time index, and sampling time instant for a digital sample $x(3)$ in time domain.
- Determine the frequency resolution, frequency bin number, and mapped frequency for each of the DFT coefficients $X(1)$ and $X(3)$ in frequency domain.

Example 4 - contd.

a. Sampling period: $T = 1/f_s = 1/10 = 0.1$ second

For $x(3)$, time index is $n = 3$, and sampling time instant is $t = nT = 3 \cdot 0.1 = 0.3$ second.

b. Frequency resolution: $\Delta f = \frac{f_s}{N} = \frac{10}{4} = 2.5$ Hz.



Frequency bin number for $X(1)$ is $k = 1$, and its corresponding frequency is $f = \frac{kf_s}{N} = \frac{1 \times 10}{4} = 2.5$ Hz.

Similarly, for $X(3)$ is $k = 3$, and its corresponding frequency is $f = \frac{kf_s}{N} = \frac{3 \times 10}{4} = 7.5$ Hz.

Amplitude and Power Spectrum

Since each calculated DFT coefficient is a complex number, it is not convenient to plot it versus its frequency index

Amplitude Spectrum:

$$A_k = \frac{1}{N} |X(k)| = \frac{1}{N} \sqrt{(\text{Real}[X(k)])^2 + (\text{Imag}[X(k)])^2},$$
$$k = 0, 1, 2, \dots, N - 1.$$

To find one-sided amplitude spectrum, we double the amplitude.

$$\bar{A}_k = \begin{cases} \frac{1}{N} |X(0)|, & k = 0 \\ \frac{2}{N} |X(k)|, & k = 1, \dots, N/2 \end{cases}$$

Amplitude and Power Spectrum -contd.

Power Spectrum:

$$P_k = \frac{1}{N^2} |X(k)|^2 = \frac{1}{N^2} \left\{ (\text{Real}[X(k)])^2 + (\text{Imag}[X(k)])^2 \right\},$$
$$k = 0, 1, 2, \dots, N - 1.$$

For, one-sided power spectrum:

$$\bar{P}_k = \begin{cases} \frac{1}{N^2} |X(0)|^2 & k = 0 \\ \frac{2}{N^2} |X(k)|^2 & k = 1, \dots, N/2 \end{cases}$$

Phase Spectrum:

$$\varphi_k = \tan^{-1} \left(\frac{\text{Imag}[X(k)]}{\text{Real}[X(k)]} \right), \quad k = 0, 1, 2, \dots, N - 1.$$

Example 5

Assuming that $f_s = 100$ Hz,

- a. Compute the amplitude spectrum, phase spectrum, and power spectrum.

Solution:

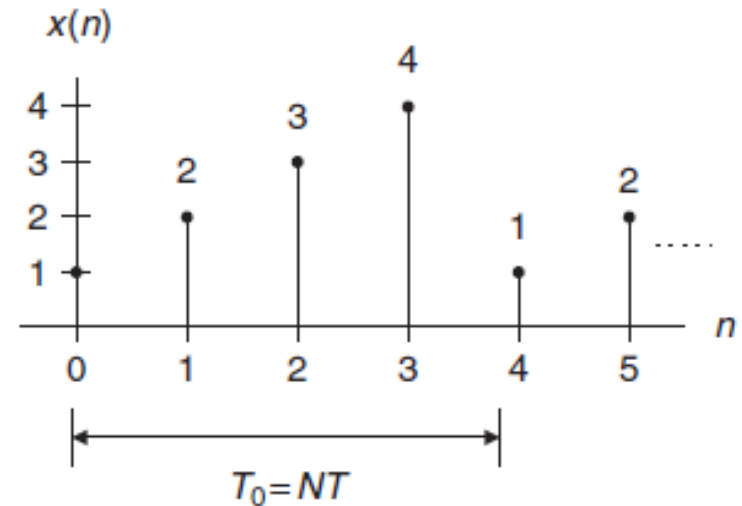
$$X(0) = 10$$

$$X(1) = -2 + j2$$

$$X(2) = -2$$

$$X(3) = -2 - j2.$$

See Example 2.



For $k = 0$, $f = k \cdot f_s / N = 0 \times 100 / 4 = 0$ Hz,

$$A_0 = \frac{1}{4} |X(0)| = 2.5, \quad \varphi_0 = \tan^{-1} \left(\frac{\text{Imag}[X(0)]}{\text{Real}[X(0)]} \right) = 0^\circ,$$

$$P_0 = \frac{1}{4^2} |X(0)|^2 = 6.25.$$

Example 5 - contd. (1)

For $k = 1$, $f = 1 \times 100/4 = 25$ Hz,

$$A_1 = \frac{1}{4} |X(1)| = 0.7071, \varphi_1 = \tan^{-1} \left(\frac{\text{Imag}[X(1)]}{\text{Real}[X(1)]} \right) = 135^\circ,$$

$$P_1 = \frac{1}{4^2} |X(1)|^2 = 0.5000.$$

For $k = 2$, $f = 2 \times 100/4 = 50$ Hz,

$$A_2 = \frac{1}{4} |X(2)| = 0.5, \varphi_2 = \tan^{-1} \left(\frac{\text{Imag}[X(2)]}{\text{Real}[X(2)]} \right) = 180^\circ,$$

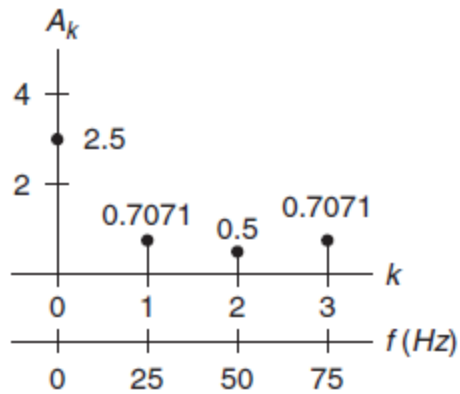
$$P_2 = \frac{1}{4^2} |X(2)|^2 = 0.2500.$$

Similarly, for $k = 3$, $f = 3 \times 100/4 = 75$ Hz,

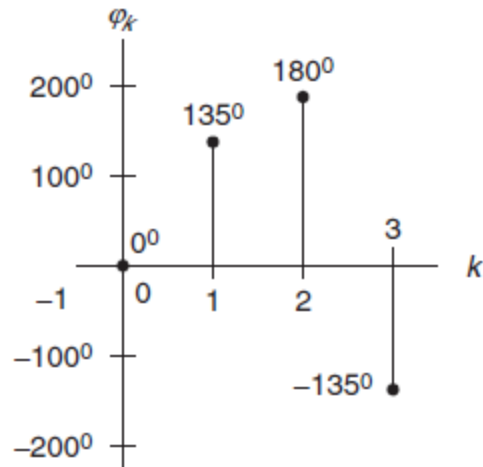
$$A_3 = \frac{1}{4} |X(3)| = 0.7071, \varphi_3 = \tan^{-1} \left(\frac{\text{Imag}[X(3)]}{\text{Real}[X(3)]} \right) = -135^\circ,$$

$$P_3 = \frac{1}{4^2} |X(3)|^2 = 0.5000.$$

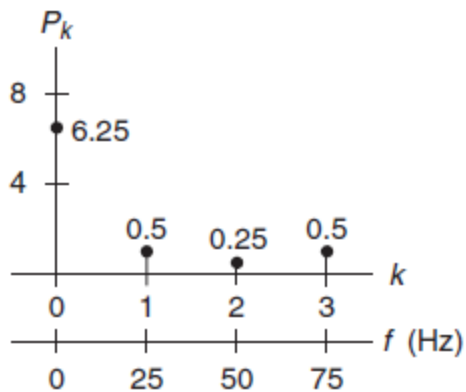
Example 5 - contd. (2)



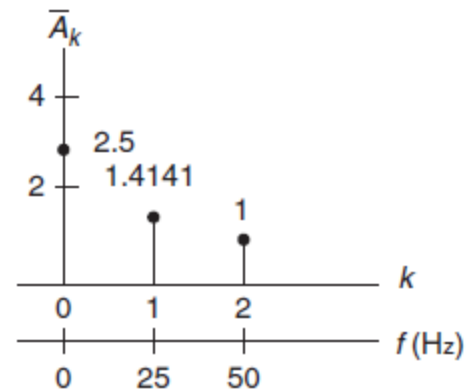
Amplitude Spectrum



Phase Spectrum



Power Spectrum



One sided Amplitude Spectrum

Example 6

Consider a digital sequence sampled at the rate of 10 kHz. If we use a size of 1,024 data points and apply the 1,024-point DFT to compute the spectrum,

- Determine the frequency resolution.
- Determine the highest frequency in the spectrum.

Solution:

a. $\Delta f = \frac{f_s}{N} = \frac{10000}{1024} = 9.776 \text{ Hz.}$

- b. The highest frequency is the folding frequency, given by

$$\begin{aligned} f_{\max} &= \frac{N}{2} \Delta f = \frac{f_s}{2} \\ &= 512 \cdot 9.776 = 5000 \text{ Hz} \end{aligned}$$

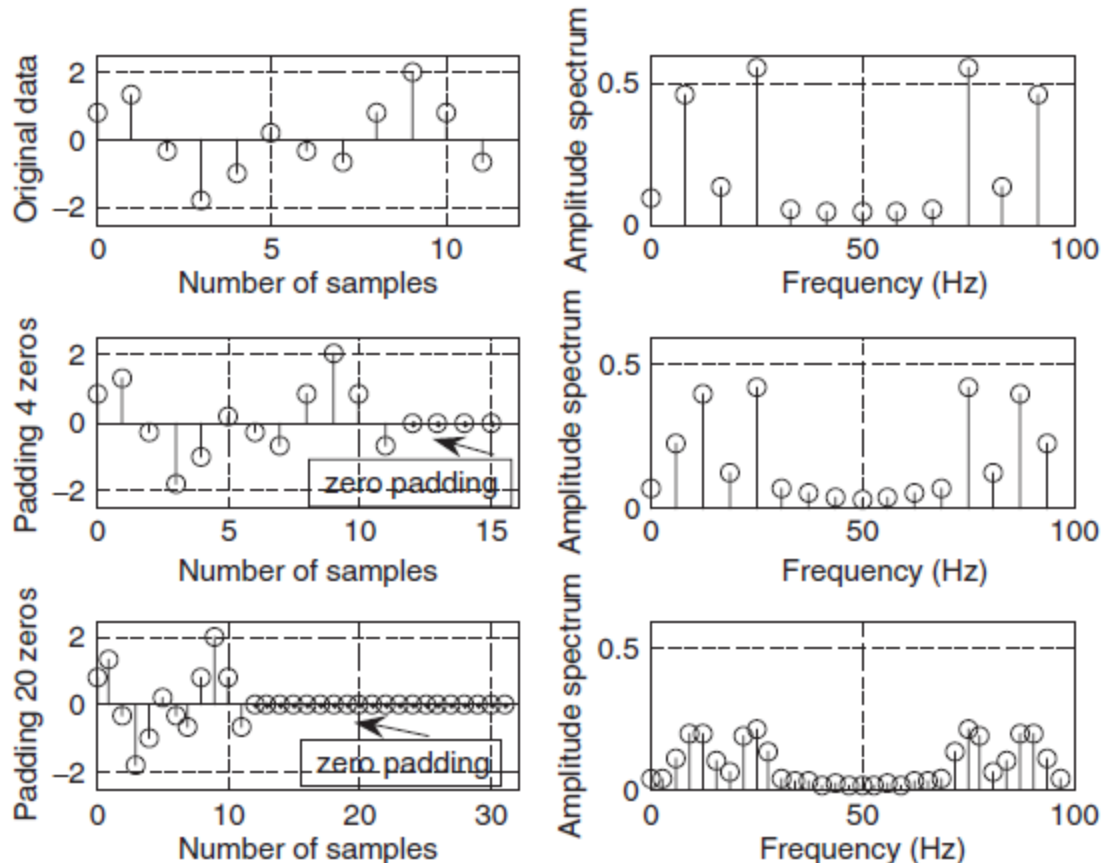
Zero Padding for FFT

FFT: Fast Fourier Transform.

↳ A fast version of DFT; It requires signal length to be power of 2.

Therefore, we need to pad zero at the end of the signal.

However, it does not add any new information.



Example 7

Consider a digital signal has sampling rate = 10 kHz. For amplitude spectrum we need frequency resolution of less than 0.5 Hz. For FFT how many data points are needed?

Solution:

$$\Delta f = 0.5 \text{ Hz}$$

$$N = \frac{f_s}{\Delta f} = \frac{10000}{0.5} = 20000$$

For FFT, we need N to be power of 2.

$$2^{14} = 16384 < 20000$$

And

$$2^{15} = 32768 > 20000$$

Recalculated frequency resolution,

$$\Delta f = \frac{f_s}{N} = \frac{10000}{32768} = 0.31 \text{ Hz.}$$

MATLAB Example - 1

$$x(n) = 2 \cdot \sin\left(2000\pi \frac{n}{8000}\right) \longleftarrow f_s$$

Use the MATLAB DFT to compute the signal spectrum with the frequency resolution to be equal to or less than 8 Hz.

$$N = \frac{f_s}{\Delta f} = \frac{8000}{8} = 1000$$

```
% Generate the sine wave sequence
```

```
fs = 8000;           %Sampling rate
```

```
N = 1000;           % Number of data points
```

```
x = 2 * sin (2000 * pi * [0:1:N - 1] / fs);
```

```
figure(1), plot(x);
```

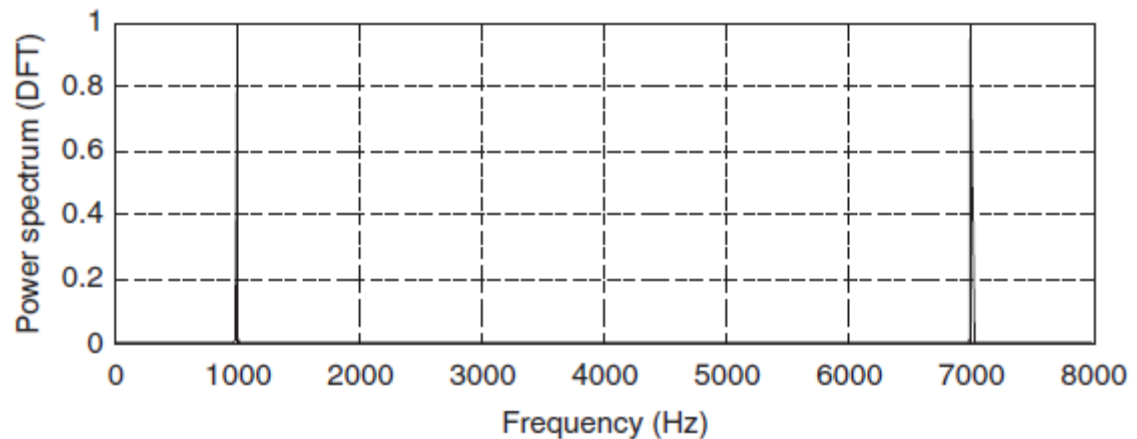
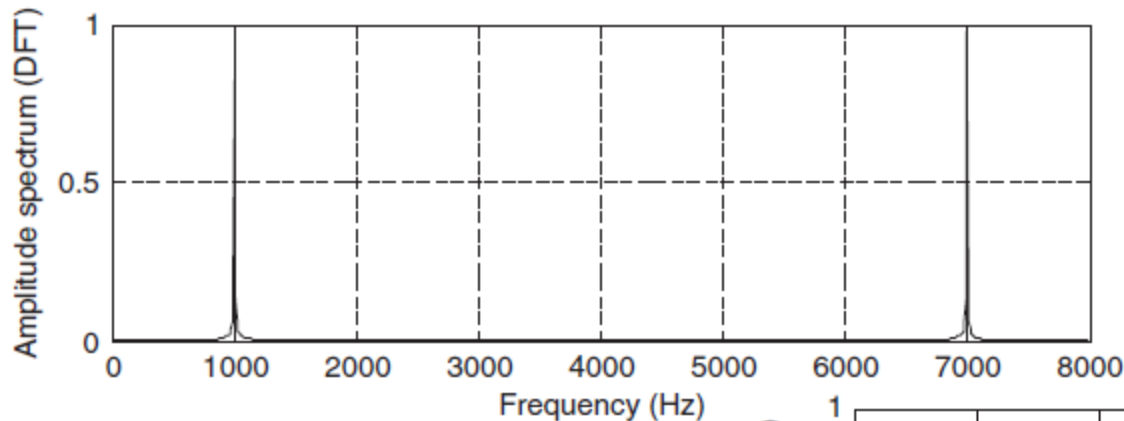
```
xf = abs(fft(x))/N; %Compute the amplitude spectrum
```

```
P = xf.*xf;         %Compute the power spectrum
```

```
f = [0:1:N - 1] * fs / N; %Map the frequency bin to the frequency (Hz)
```

MATLAB Example - contd. (1)

```
subplot(2,1,1); plot(f,xf);grid  
xlabel('Frequency (Hz)'); ylabel('Amplitude spectrum (DFT)');  
subplot(2,1,2); plot(f,P);grid  
xlabel('Frequency (Hz)'); ylabel('Power spectrum (DFT)');
```



MATLAB Example - contd. (2)

```
% Convert it to one-sided spectrum
```

```
xf(2:N) = 2*xf(2:N); % Get the single-sided spectrum
```

```
P = xf.*xf; % Calculate the power spectrum
```

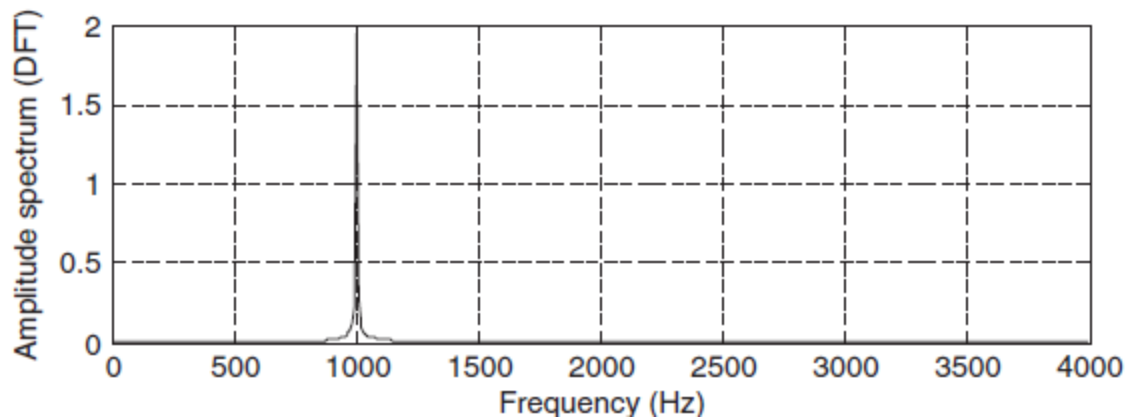
```
f = [0:1:N/2]*fs/N % Frequencies up to the folding frequency
```

```
subplot(2,1,1); plot(f,xf(1:N/2+1)); grid
```

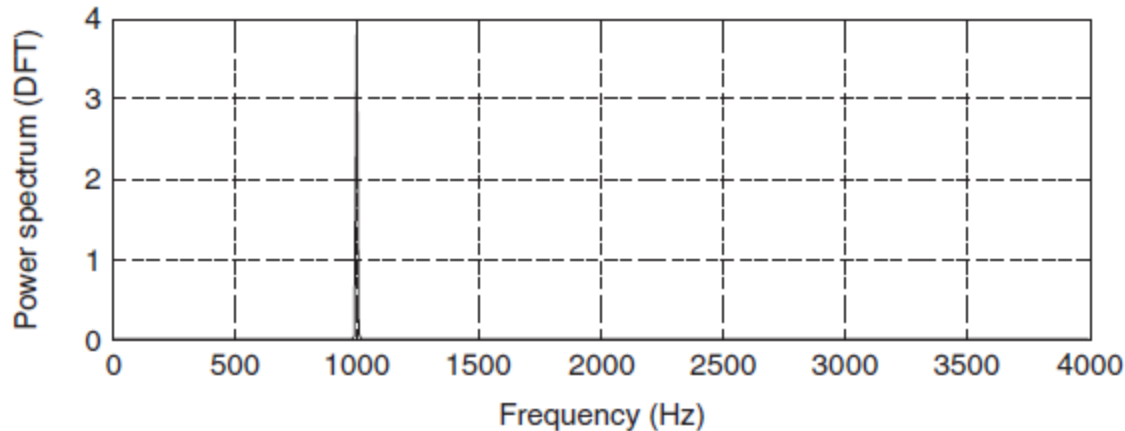
```
xlabel('Frequency (Hz)'); ylabel('Amplitude spectrum (DFT)');
```

```
subplot(2,1,2); plot(f,P(1:N/2+1)); grid
```

```
xlabel('Frequency (Hz)'); ylabel('Power spectrum (DFT)');
```



MATLAB Example - contd. (3)



```
% Zero padding to the length of 1024
```

```
x = [x, zeros(1,24)];
```

```
N = length(x);
```

```
xf = abs(fft(x))/N; %Compute the amplitude spectrum with zero padding
```

```
P = xf.*xf; %Compute the power spectrum
```

```
f = [0:1:N-1]*fs/N; %Map frequency bin to frequency (Hz)
```

```
subplot(2,1,1); plot(f,xf); grid
```

```
xlabel('Frequency (Hz)'); ylabel('Amplitude spectrum (FFT)');
```

```
subplot(2,1,2); plot(f,P); grid
```

```
xlabel('Frequency (Hz)'); ylabel('Power spectrum (FFT)');
```

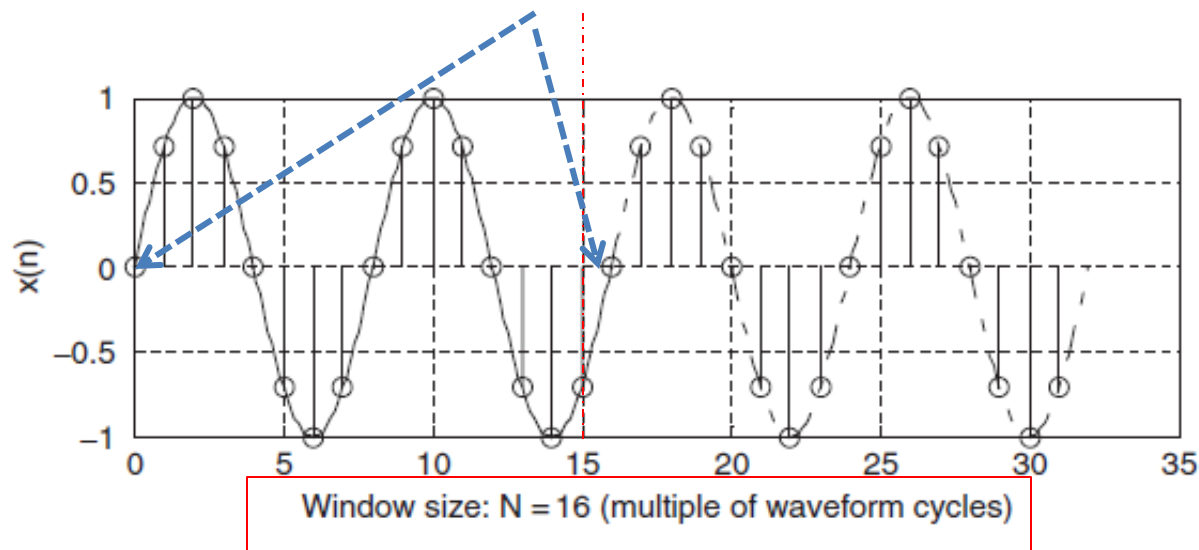
.....

Effect of Window Size

When applying DFT, we assume the following:

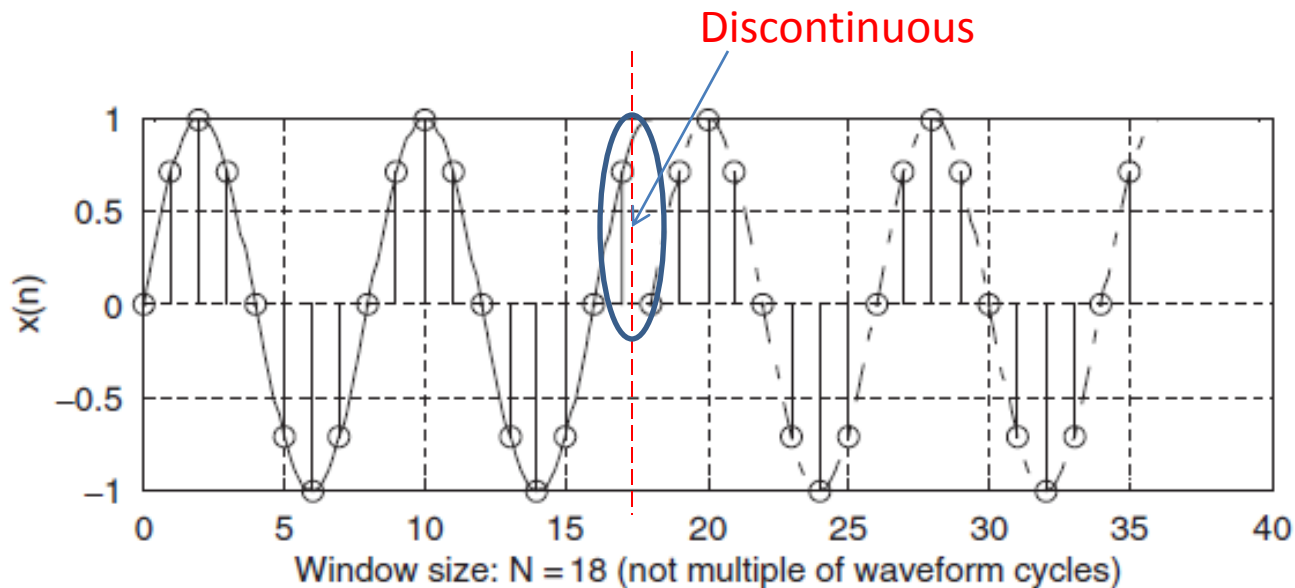
1. Sampled data are periodic to themselves (repeat).
2. Sampled data are continuous to themselves and band limited to the folding frequency.

1 Hz sinusoid,
with 32
samples

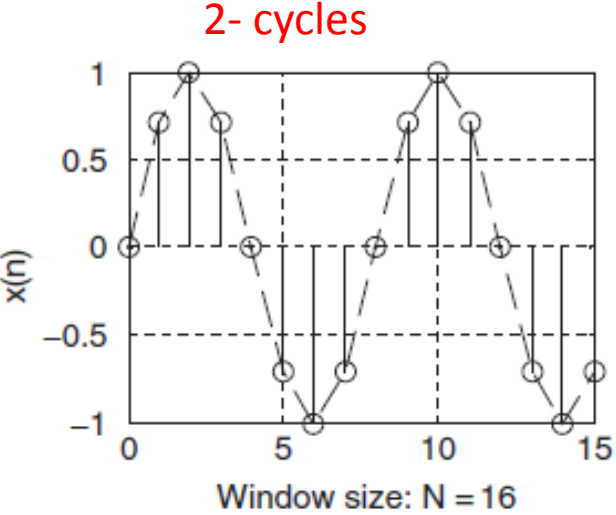


Effect of Window Size -contd. (1)

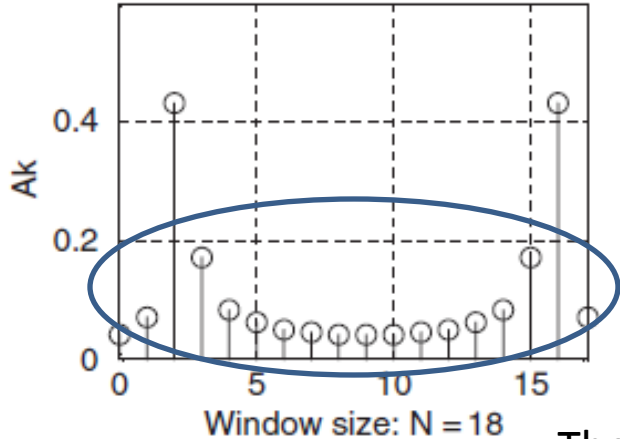
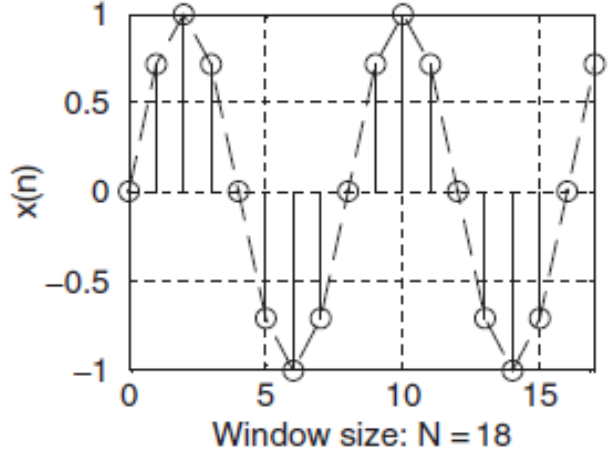
If the window size is not multiple of waveform cycles:



Effect of Window Size -contd. (2)



Produces single frequency



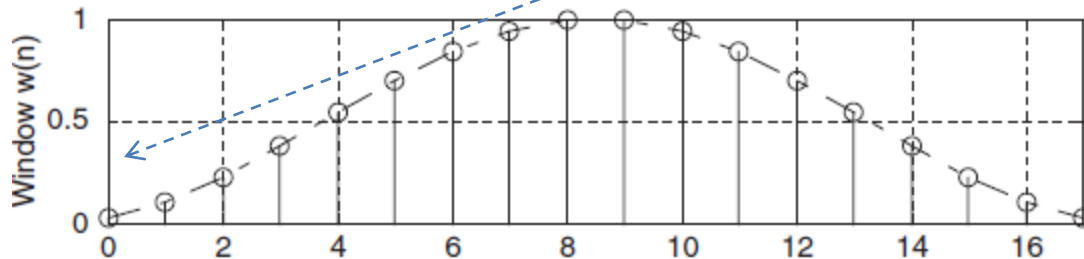
Produces many harmonics as well.

Spectral Leakage

The bigger the discontinuity, the more the leakage

Reducing Leakage Using Window

To reduce the effect of spectral leakage, a window function can be used whose amplitude tapers smoothly and gradually toward zero at both ends.



$$x_w(n) = x(n)w(n), \text{ for } n = 0, 1, \dots, N - 1.$$

Window function, $w(n)$

Data sequence, $x(n)$

Obtained windowed sequence, $x_w(n)$

Example 8

Given,

$$x(2) = 1 \text{ and } w(2) = 0.2265;$$

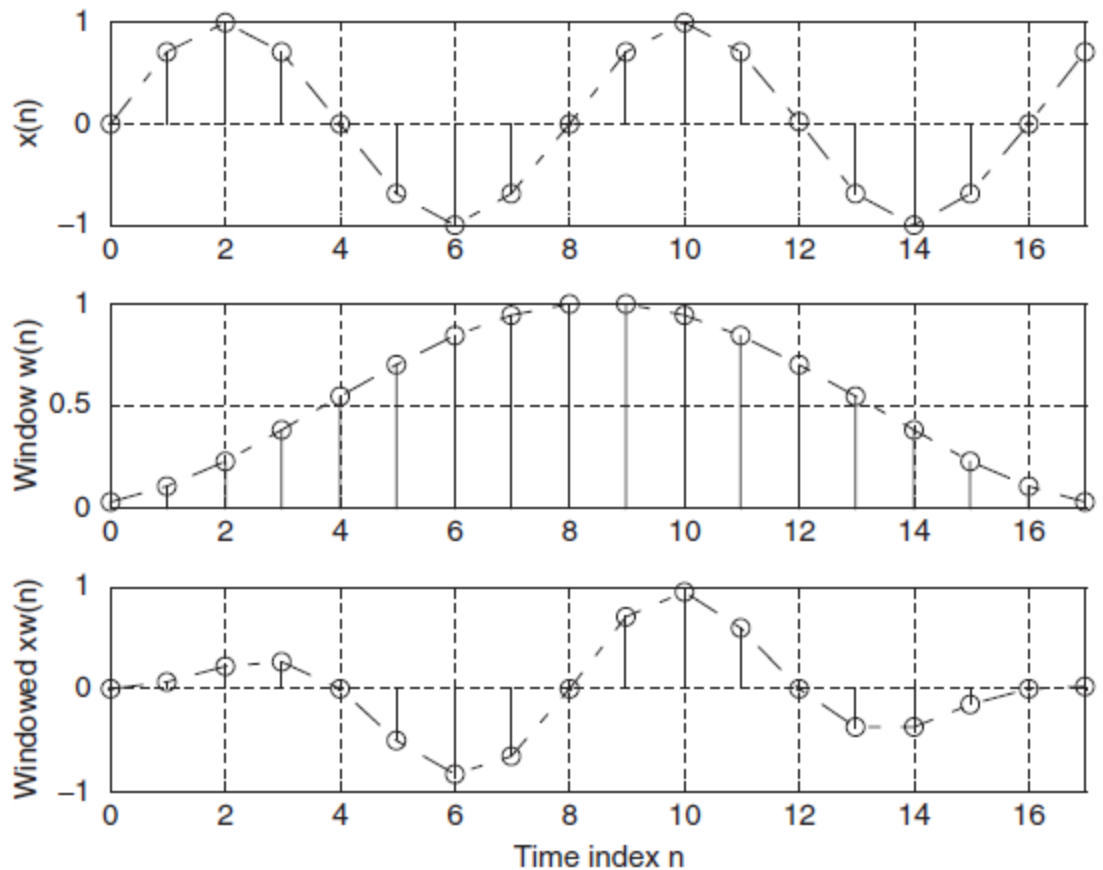
$$x(5) = -0.7071 \text{ and } w(5) = 0.7008,$$

Calculate,

$$x_w(2) \text{ and } x_w(5).$$

$$\begin{aligned} x_w(2) &= x(2) \times w(2) \\ &= 1 \times 0.2265 = 0.2265 \end{aligned}$$

$$\begin{aligned} x_w(5) &= x(5) \times w(5) \\ &= -0.7071 \times 0.7008 = -0.4956 \end{aligned}$$



Different Types of Windows

Rectangular Window (no window): $w_R(n) = 1 \quad 0 \leq n \leq N - 1$

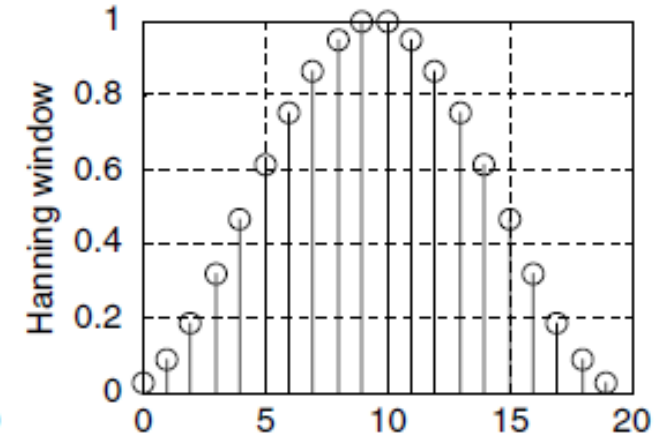
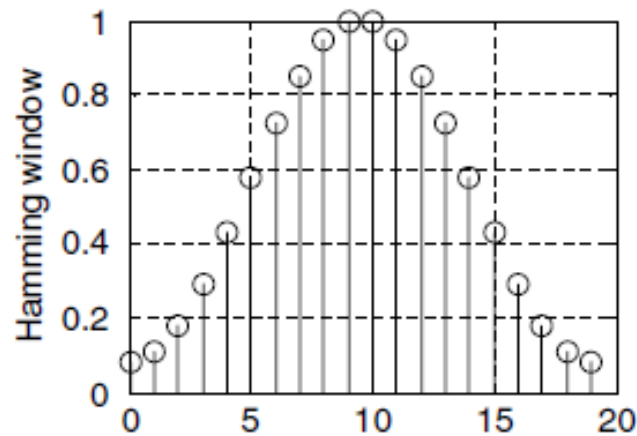
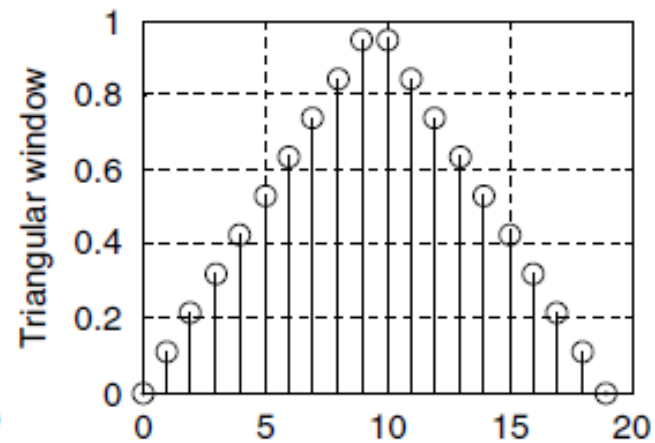
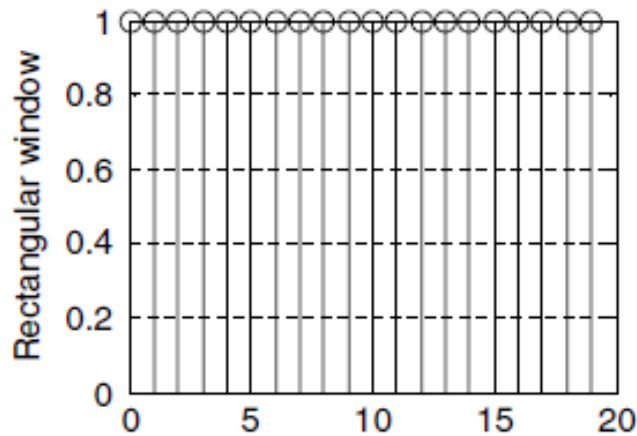
Triangular Window: $w_{tri}(n) = 1 - \frac{|2n - N + 1|}{N - 1}, 0 \leq n \leq N - 1$

Hamming Window: $w_{hm}(n) = 0.54 - 0.46 \cos\left(\frac{2\pi n}{N - 1}\right), 0 \leq n \leq N - 1$

Hanning Window: $w_{hm}(n) = 0.5 - 0.5 \cos\left(\frac{2\pi n}{N - 1}\right), 0 \leq n \leq N - 1$

Different Types of Windows -contd.

Window size of 20 samples



Example 9

Problem:

Considering the sequence $x(0) = 1$, $x(1) = 2$, $x(2) = 3$, and $x(3) = 4$, and given $f_s = 100$ Hz, $T = 0.01$ seconds, compute the amplitude spectrum, phase spectrum, and power spectrum

Using the Hamming window function.

Solution:

Since $N = 4$, Hamming window function can be found as:

$$w_{hm}(0) = 0.54 - 0.46 \cos\left(\frac{2\pi \times 0}{4 - 1}\right) = 0.08$$

$$w_{hm}(1) = 0.54 - 0.46 \cos\left(\frac{2\pi \times 1}{4 - 1}\right) = 0.77.$$

Similarly, $w_{hm}(2) = 0.77$, $w_{hm}(3) = 0.08$.

Example 9 - contd. (1)

Windowed sequence:

$$x_w(0) = x(0) \times w_{hm}(0) = 1 \times 0.08 = 0.08$$

$$x_w(1) = x(1) \times w_{hm}(1) = 2 \times 0.77 = 1.54$$

$$x_w(2) = x(2) \times w_{hm}(2) = 3 \times 0.77 = 2.31$$

$$x_w(3) = x(3) \times w_{hm}(3) = 4 \times 0.08 = 0.32.$$

DFT Sequence:

$$X(k) = x(0) W_N^{k0} + x(1) W_N^{k1} + x(2) W_N^{k2} + \dots + x(N-1) W_N^{k(N-1)}$$

$$\Rightarrow X(k) = x_w(0) W_4^{k \times 0} + x_w(1) W_4^{k \times 1} + x_w(2) W_4^{k \times 2} + x_w(3) W_4^{k \times 3}.$$

$$\Rightarrow \left\{ \begin{array}{l} X(0) = 4.25 \\ X(1) = -2.23 - j1.22 \\ X(2) = 0.53 \\ X(3) = -2.23 + j1.22 \end{array} \right. \quad \Delta f = \frac{1}{NT} = \frac{1}{4 \cdot 0.01} = 25 \text{ Hz}$$

Example 9 - contd. (2)

$$A_0 = \frac{1}{4}|X(0)| = 1.0625, \varphi_0 = \tan^{-1} \left(\frac{0}{4.25} \right) = 0^0,$$

$$P_0 = \frac{1}{4^2}|X(0)|^2 = 1.1289$$

$$A_1 = \frac{1}{4}|X(1)| = 0.6355, \varphi_1 = \tan^{-1} \left(\frac{-1.22}{-2.23} \right) = -151.32^0,$$

$$P_1 = \frac{1}{4^2}|X(1)|^2 = 0.4308$$

$$A_2 = \frac{1}{4}|X(2)| = 0.1325, \varphi_2 = \tan^{-1} \left(\frac{0}{0.53} \right) = 0^0,$$

$$P_2 = \frac{1}{4^2}|X(2)|^2 = 0.0176.$$

$$A_3 = \frac{1}{4}|X(3)| = 0.6355, \varphi_3 = \tan^{-1} \left(\frac{1.22}{-2.23} \right) = 151.32^0,$$

$$P_3 = \frac{1}{4^2}|X(3)|^2 = 0.4308.$$

MATLAB Example - 2

$$x(n) = 2 \cdot \sin\left(2000\pi \frac{n}{8000}\right)$$

Compute the spectrum of a Hamming window function with a window size = 100.

```
% Generate the sine wave sequence
fs = 8000; T = 1/fs;           % Sampling rate and sampling period

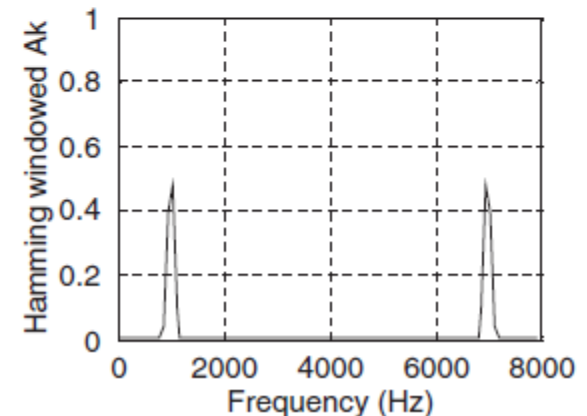
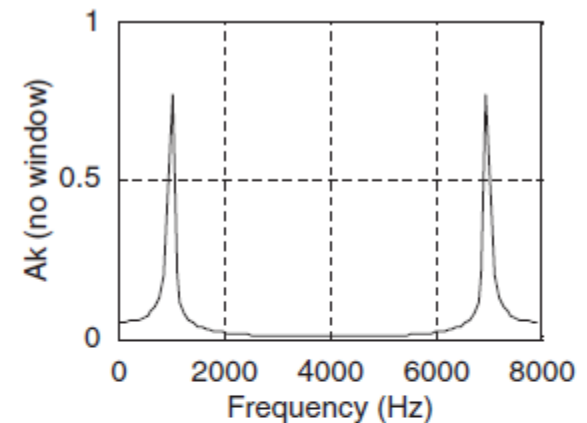
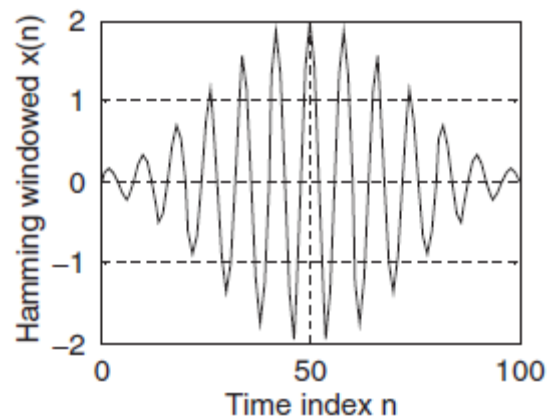
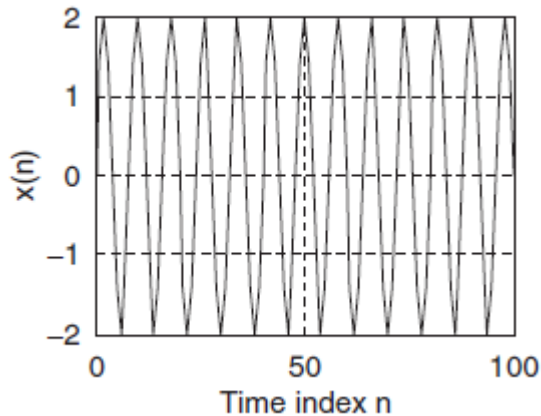
% Generate the sine wave sequence
x = 2* sin (2000*pi*[0:1:100]*T);
% Apply the FFT algorithm
N=length(x);
index_t = [0:1:N-1];
f = [0:1:N-1]*fs/N;
xf = abs (fft (x) ) /N;

%Using the Hamming window
x_hm = x.*hamming (N)';
xf_hm=abs (fft (x_hm) ) /N;

%Apply the Hamming window function
%Calculate the amplitude spectrum
```

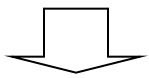

MATLAB Example - 2 contd.

```
subplot(2,2,1);plot(index_t,x);grid
xlabel('Time index n'); ylabel('x(n)');
subplot(2,2,3); plot(index_t,x_hm);grid
xlabel('Time index n'); ylabel('Hamming windowed x(n)');
subplot(2,2,2);plot(f,xf);grid;axis([0 fs 0 1]);
xlabel('Frequency (Hz)'); ylabel('Ak (no window)');
subplot(2,2,4); plot(f,xf_hm);grid;axis([0 fs 0 1]);
xlabel('Frequency (Hz)'); ylabel('Hamming windowed Ak');
```

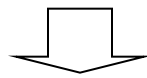


DFT Matrix

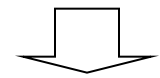
Frequency Spectrum



Multiplication Matrix



Time-Domain samples



$$\begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ \vdots \\ X(N-2) \\ X(N-1) \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & e^{-j\frac{2\pi}{N}} & e^{-j\frac{4\pi}{N}} & \dots & e^{-j\frac{2(N-2)\pi}{N}} \\ 1 & e^{-j\frac{4\pi}{N}} & e^{-j\frac{8\pi}{N}} & \dots & e^{-j\frac{4(N-2)\pi}{N}} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & e^{-j\frac{2(N-2)\pi}{N}} & e^{-j\frac{4(N-2)\pi}{N}} & \dots & e^{-j\frac{2(N-2)^2\pi}{N}} \\ 1 & e^{-j\frac{2(N-1)\pi}{N}} & e^{-j\frac{4(N-1)\pi}{N}} & \dots & e^{-j\frac{2(N-1)(N-2)\pi}{N}} \\ 1 & e^{-j\frac{2(N-1)\pi}{N}} & e^{-j\frac{4(N-1)\pi}{N}} & \dots & e^{-j\frac{(N-1)^2\pi}{N}} \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ x(2) \\ \vdots \\ x(N-2) \\ x(N-1) \end{bmatrix}$$

DFT Matrix

Let, $w_N = e^{-j2\pi/N}$

Then

$$\begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ \vdots \\ X(N-1) \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & w & w^2 & \dots & w^{(N-1)} \\ 1 & w^2 & w^4 & \dots & w^{2(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & w^{(N-1)} & w^{2(N-1)} & \dots & w^{(N-1)^2} \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ x(2) \\ \vdots \\ x(N-1) \end{bmatrix}$$

DFT equation:
$$X(k) = \sum_{m=0}^{N-1} x(m)w_N^{mk} \quad k = 0, \dots, N-1$$

DFT requires N^2 complex multiplications.

FFT

FFT: Fast Fourier Transform

A very efficient algorithm to compute DFT; it requires less multiplication.

The length of input signal, $x(n)$ must be 2^m samples, where m is an integer.

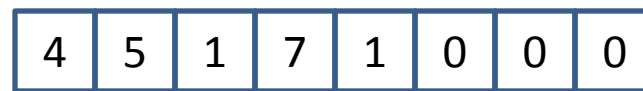


Samples $N = 2, 4, 8, 16$ or so.

If the input length is not 2^m , append (pad) zeros to make it 2^m .



$N = 5$



$N = 8$, power of 2

DFT to FFT: Decimation in Frequency

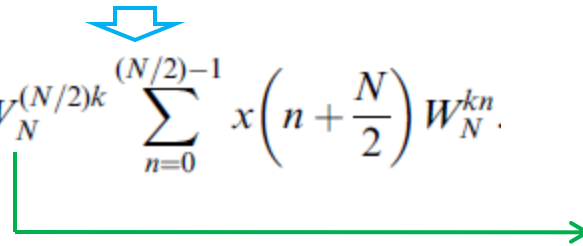
DFT: $X(k) = \sum_{n=0}^{N-1} x(n) W_N^{kn}$ for $k = 0, 1, \dots, N-1$,

$$X(k) = x(0) + x(1)W_N^k + \dots + x(N-1)W_N^{k(N-1)}$$

$$X(k) = \boxed{x(0) + x(1)W_N^k + \dots + x\left(\frac{N}{2} - 1\right)W_N^{k(N/2-1)}} + \boxed{x\left(\frac{N}{2}\right)W_N^{kN/2} + \dots + x(N-1)W_N^{k(N-1)}}$$

$$X(k) = \sum_{n=0}^{(N/2)-1} x(n)W_N^{kn} + \boxed{\sum_{n=N/2}^{N-1} x(n)W_N^{kn}}$$

$$X(k) = \sum_{n=0}^{(N/2)-1} x(n)W_N^{kn} + W_N^{(N/2)k} \sum_{n=0}^{(N/2)-1} x\left(n + \frac{N}{2}\right)W_N^{kn}$$



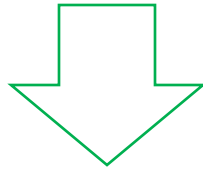
$$W_N^{N/2} = e^{-j\frac{2\pi(N/2)}{N}} = e^{-j\pi} = -1$$

$$X(k) = \sum_{n=0}^{(N/2)-1} \left(x(n) + (-1)^k x\left(n + \frac{N}{2}\right) \right) W_N^{kn}$$

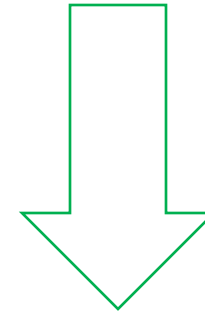
DFT to FFT: Decimation in Frequency

Now decompose into even ($k = 2m$) and odd ($k = 2m+1$) sequences.

$$X(2m) = \sum_{n=0}^{(N/2)-1} \left(x(n) + x\left(n + \frac{N}{2}\right) \right) W_N^{2mn}, \quad X(2m+1) = \sum_{n=0}^{(N/2)-1} \left(x(n) - x\left(n + \frac{N}{2}\right) \right) W_N^n W_N^{2mn}$$



$$W_N^2 = e^{-j\frac{2\pi \times 2}{N}} = e^{-j\frac{2\pi}{(N/2)}} = W_{N/2}$$



$$X(2m) = \sum_{n=0}^{(N/2)-1} a(n) W_{N/2}^{mn} = \text{DFT}\{a(n) \text{ with } (N/2) \text{ points}\}$$

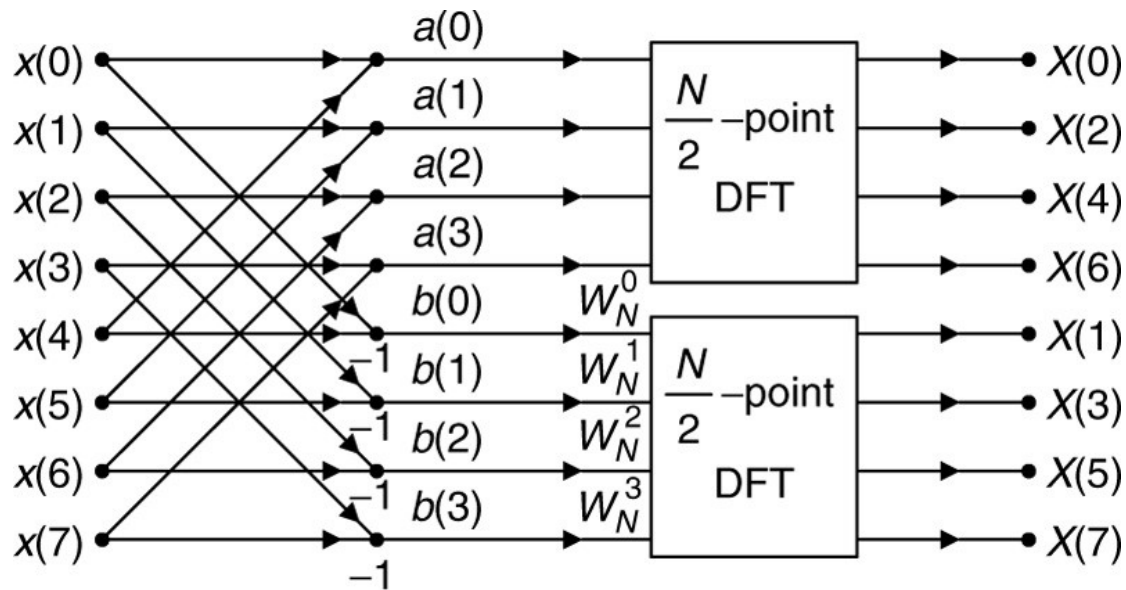
$$X(2m+1) = \sum_{n=0}^{(N/2)-1} b(n) W_N^n W_{N/2}^{mn} = \text{DFT}\{b(n) W_N^n \text{ with } (N/2) \text{ points}\}$$

$$a(n) = x(n) + x\left(n + \frac{N}{2}\right), \text{ for } n = 0, 1, \dots, \frac{N}{2} - 1$$

$$b(n) = x(n) - x\left(n + \frac{N}{2}\right), \text{ for } n = 0, 1, \dots, \frac{N}{2} - 1.$$

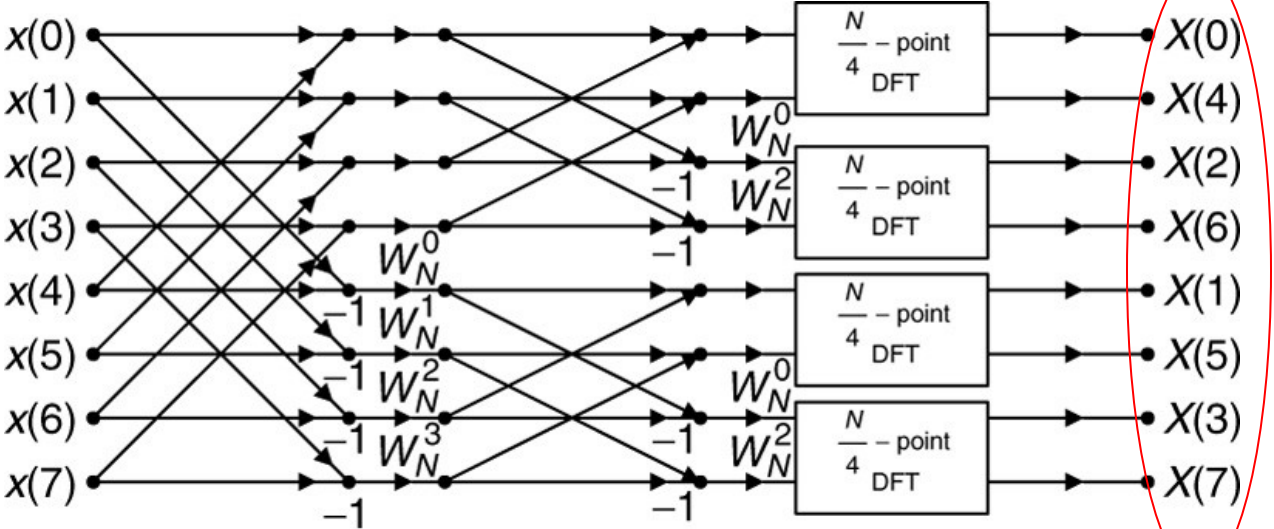
DFT to FFT: Decimation in Frequency

$$DFT\{x(n) \text{ with } N \text{ points}\} = \begin{cases} DFT\{a(n) \text{ with } (N/2) \text{ points}\} \\ DFT\{b(n)W_N^n \text{ with } (N/2) \text{ points}\} \end{cases}$$

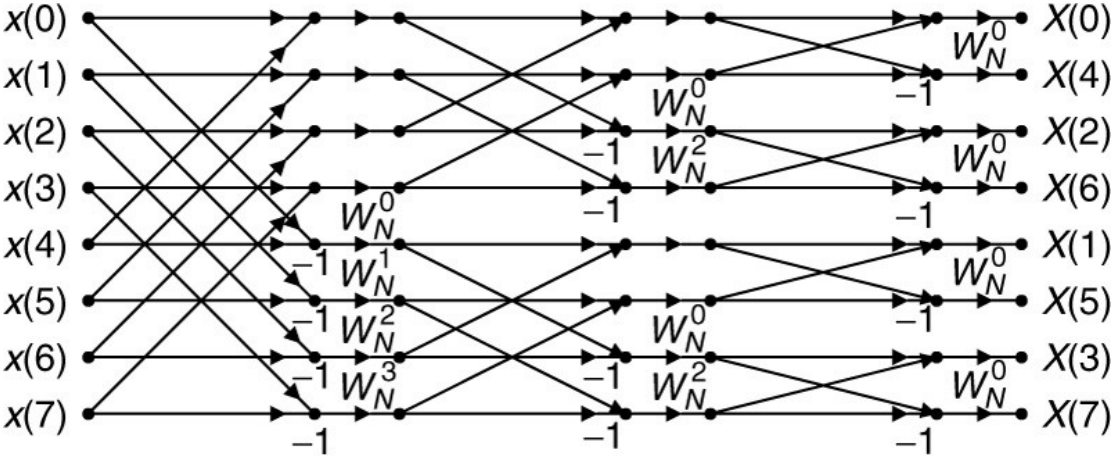


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DFT to FFT: Decimation in Frequency



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12 complex multiplication

DFT to FFT: Decimation in Frequency

Binary	index	1st split	2nd split	3rd split	Bit reversal
000	0	0	0	0	000
001	1	2	4	4	100
010	2	4	2	2	010
011	3	6	6	6	011
100	4	1	1	1	001
101	5	3	5	5	101
110	6	5	3	3	011
111	7	7	7	7	111

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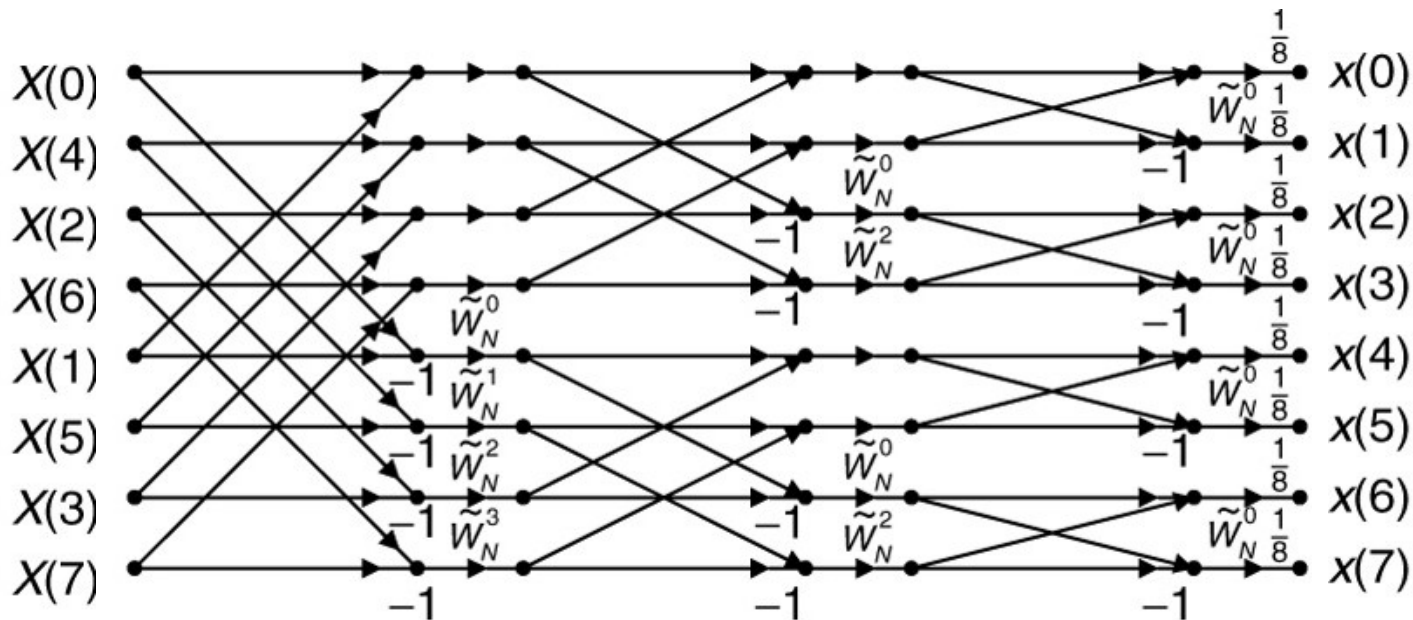
Complex multiplications of DFT = N^2 , and
 Complex multiplications of FFT = $\frac{N}{2} \log_2(N)$



For 1024 samples data sequence, DFT requires $1024 \times 1024 = 1048576$ complex multiplications. FFT requires $(1024/2) \log_2(1024) = 5120$ complex multiplications.

IFFT: Inverse FFT

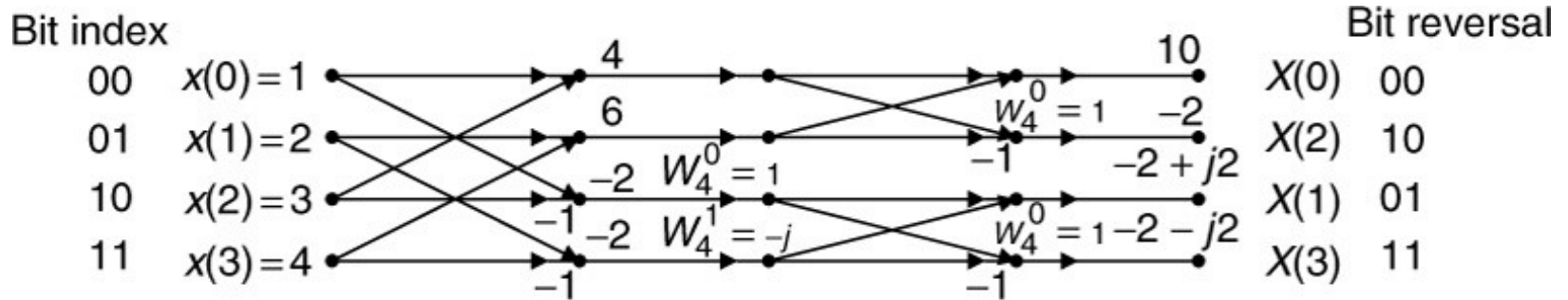
$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-kn} = \frac{1}{N} \sum_{k=0}^{N-1} X(k) \tilde{W}_N^{kn}, \text{ for } k = 0, 1, \dots, N-1.$$



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FFT and IFFT Examples

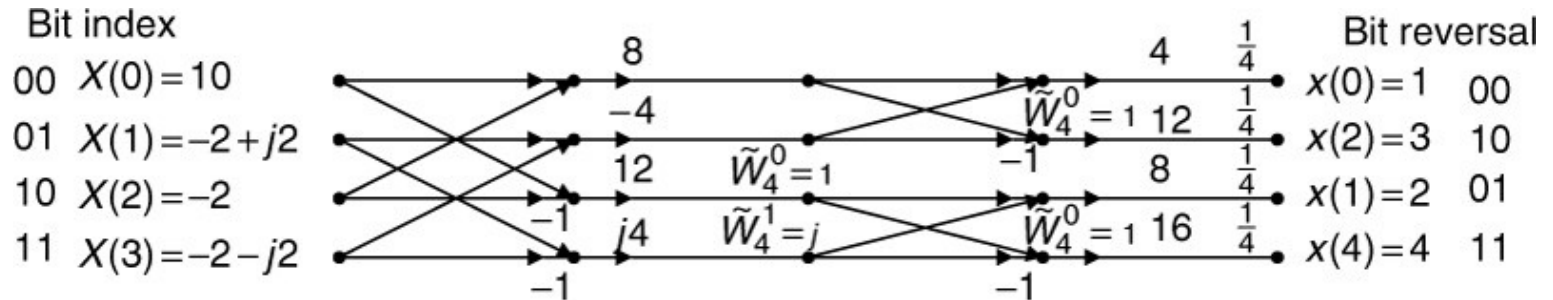
FFT



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$$\text{Number of complex multiplication} = \frac{N}{2} \log_2(N) = \frac{4}{2} \log_2(4) = 4.$$

IFFT



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DFT to FFT: Decimation in Time

Split the input sequence $x(n)$ into the even indexed $x(2m)$ and $x(2m + 1)$, each with $N/2$ data points.

$$X(k) = \sum_{m=0}^{(N/2)-1} x(2m) W_N^{2mk} + \sum_{m=0}^{(N/2)-1} x(2m + 1) W_N^k W_N^{2mk},$$

for $k = 0, 1, \dots, N - 1$.

Using

$$w_N^2 = \left(e^{-j2\pi/N} \right)^2 = e^{-j2\pi/(N/2)} = w_{N/2}$$

$$X(k) = \sum_{m=0}^{(N/2)-1} x(2m) W_{N/2}^{mk} + W_N^k \sum_{m=0}^{(N/2)-1} x(2m + 1) W_{N/2}^{mk},$$

for $k = 0, 1, \dots, N - 1$.

DFT to FFT: Decimation in Time

Define new functions as

$$G(k) = \sum_{m=0}^{(N/2)-1} x(2m)W_{N/2}^{mk} = \text{DFT}\{x(2m) \text{ with } (N/2) \text{ points}\}$$

$$H(k) = \sum_{m=0}^{(N/2)-1} x(2m+1)W_{N/2}^{mk} = \text{DFT}\{x(2m+1) \text{ with } (N/2) \text{ points}\}.$$

As,

$$G(k) = G\left(k + \frac{N}{2}\right), \text{ for } k = 0, 1, \dots, \frac{N}{2} - 1$$

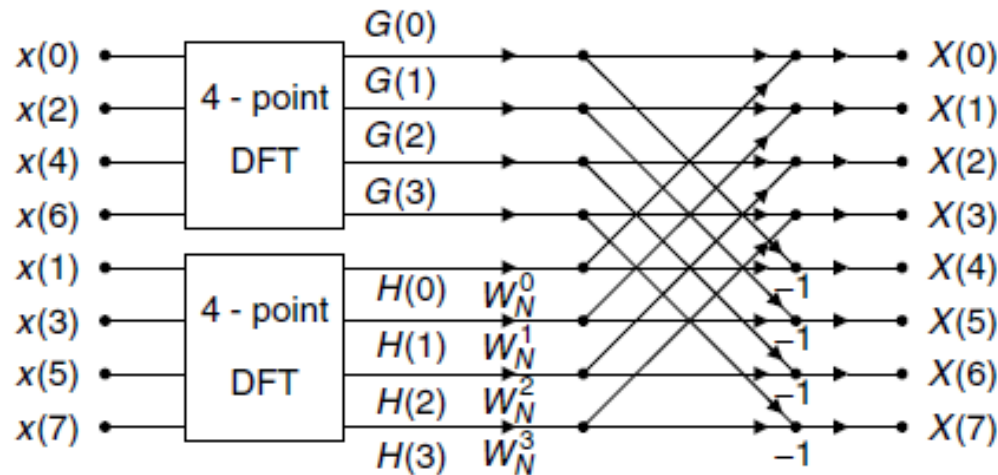
$$H(k) = H\left(k + \frac{N}{2}\right), \text{ for } k = 0, 1, \dots, \frac{N}{2} - 1.$$

$$X(k) = G(k) + W_N^k H(k), \text{ for } k = 0, 1, \dots, \frac{N}{2} - 1.$$

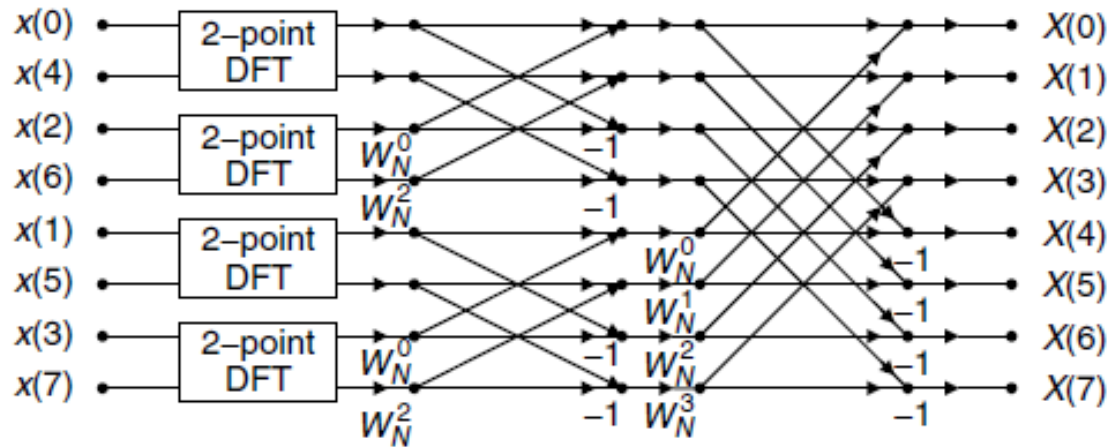
$$X\left(\frac{N}{2} + k\right) = G(k) - W_N^k H(k), \text{ for } k = 0, 1, \dots, \frac{N}{2} - 1. \leftarrow W_N^{(N/2+k)} = -W_N^k.$$

DFT to FFT: Decimation in Time

First iteration:

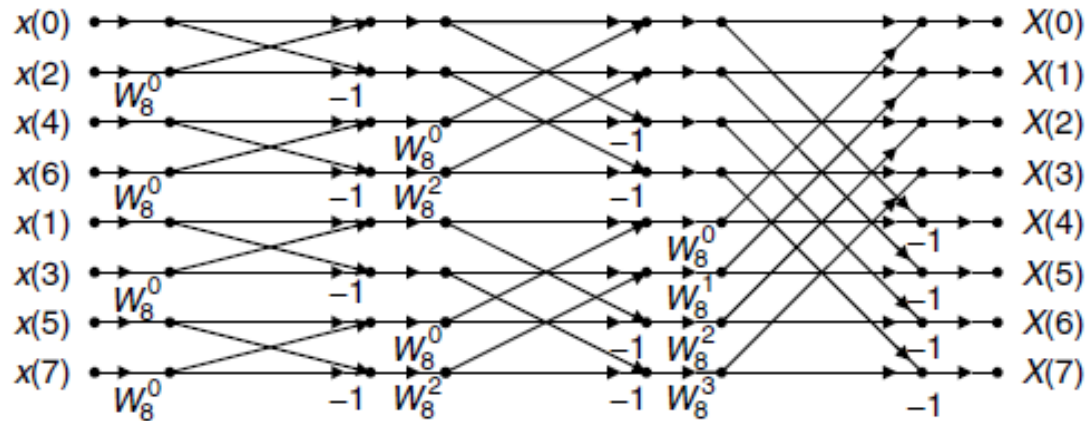


Second iteration:



DFT to FFT: Decimation in Time

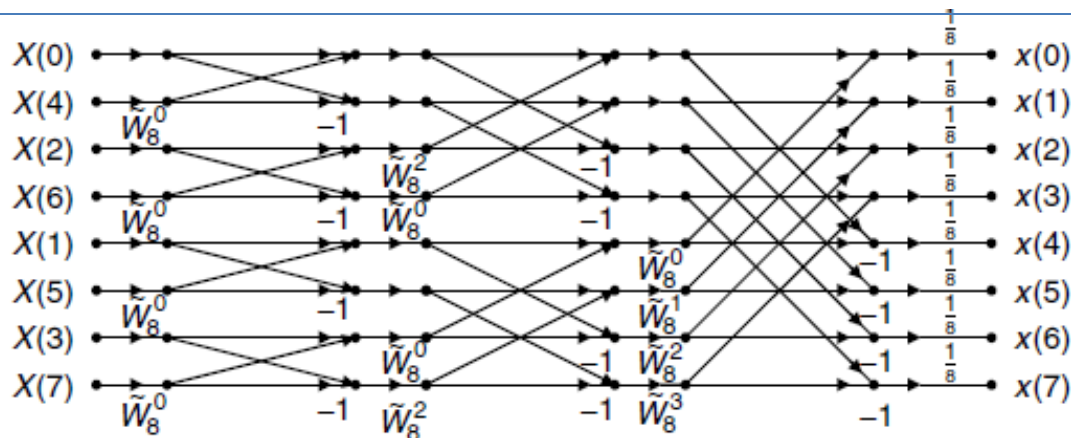
Third iteration:



$$W_N = e^{-\frac{2\pi}{N}} = \cos\left(\frac{2\pi}{N}\right) - j \sin\left(\frac{2\pi}{N}\right)$$

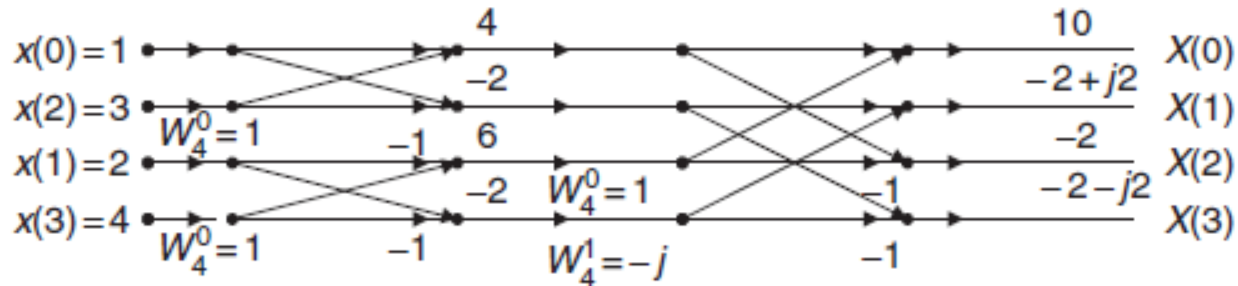
$$W_8^2 = e^{-\frac{2\pi \times 2}{8}} = e^{-\frac{\pi}{2}} = \cos(\pi/2) - j \sin(\pi/2) = -j$$

IFFT

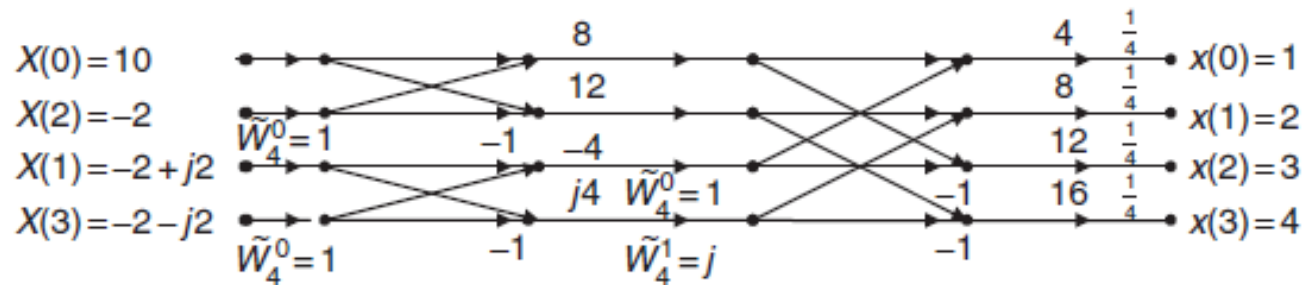


FFT and IFFT Examples

FFT



IFFT



Fourier Transform Properties (1)

Time Domain

Frequency Domain

FT is linear:

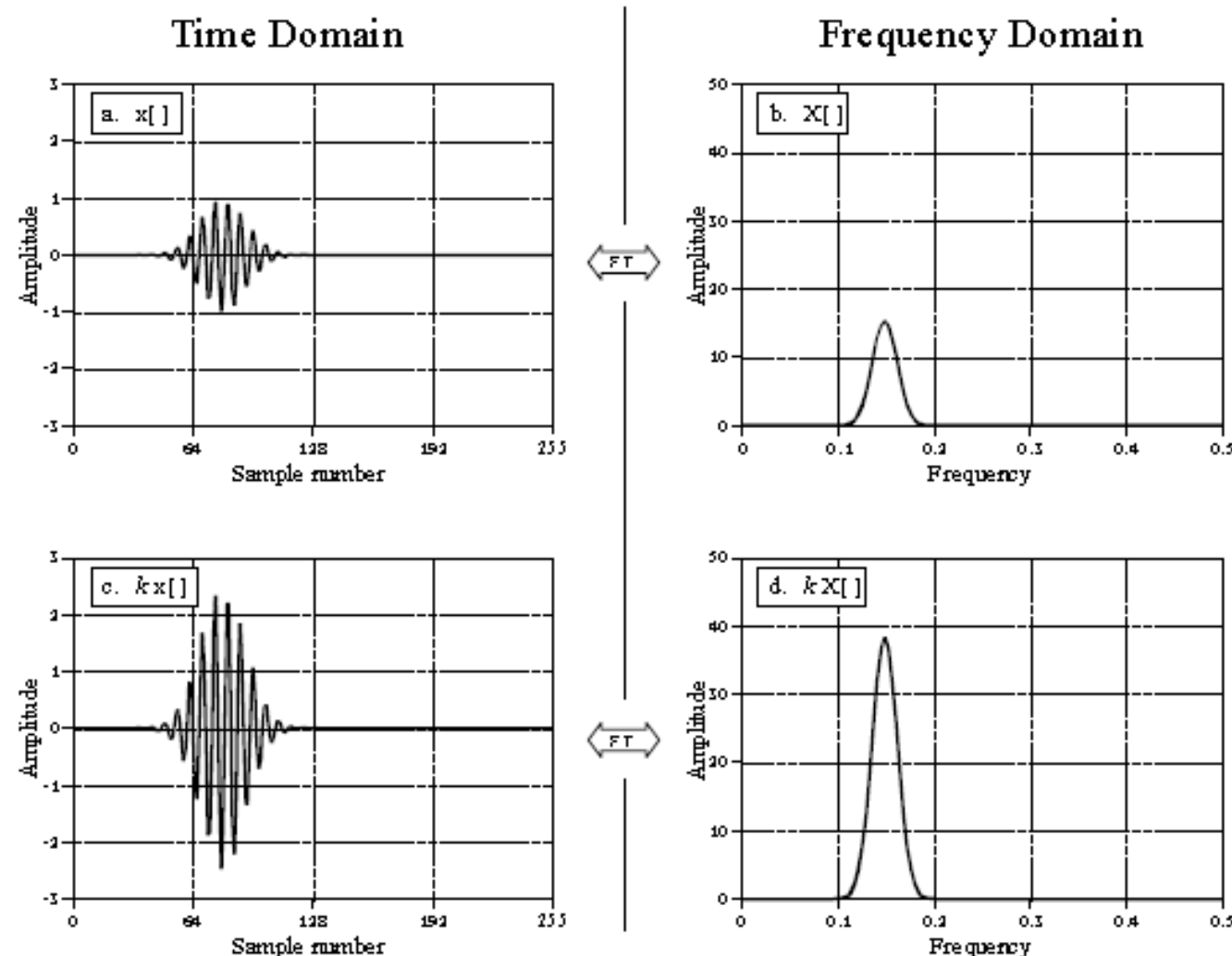
- Homogeneity
- Additivity

Homogeneity:

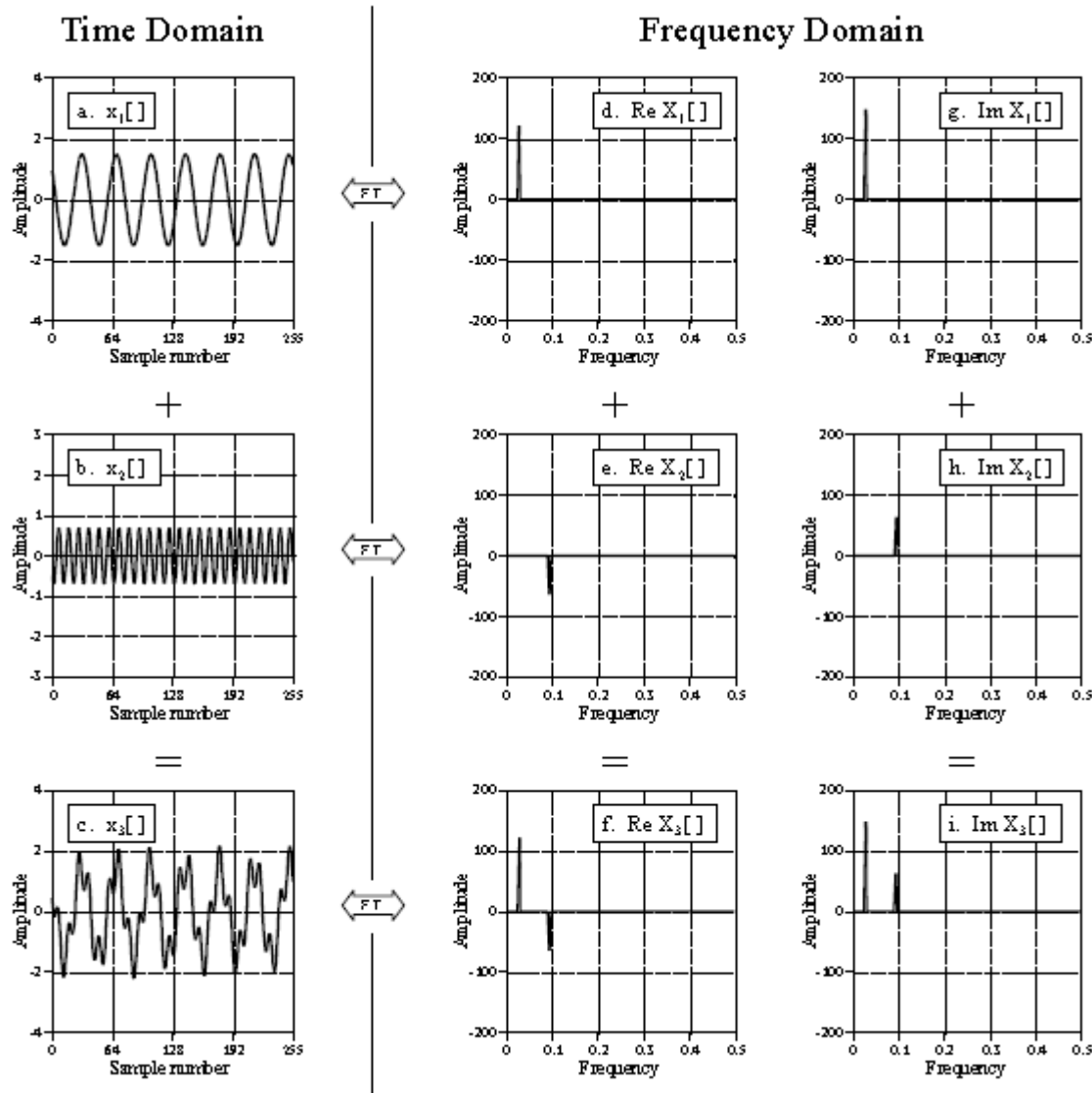
$$x[] \xrightarrow{\text{DFT}} X[]$$

$$kx[] \xrightarrow{\text{DFT}} kX[]$$

Frequency is not changed.



Fourier Transform Properties (2)



Additivity

$$\text{If } : x_1[n] + x_2[n] = x_3[n]$$

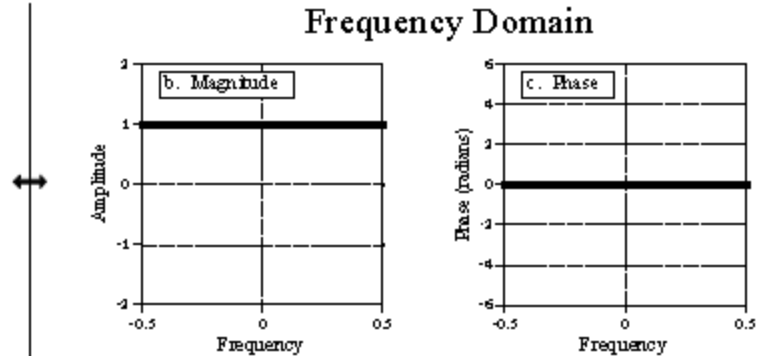
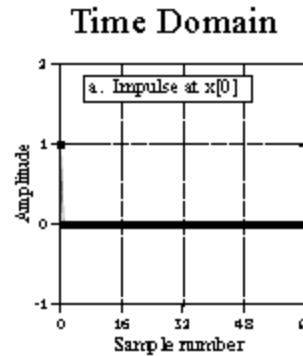
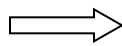
$$\text{Then } : \text{Re } X_1[f] + \text{Re } X_2[f] = \text{Re } X_3[f]$$

$$\text{and } \text{Im } X_1[f] + \text{Im } X_2[f] = \text{Im } X_3[f]$$

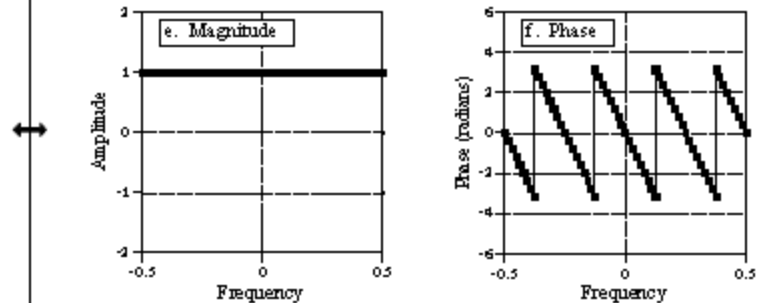
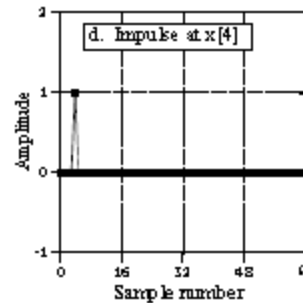
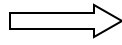
Fourier Transform Pairs

Delta Function Pairs in Polar Form

Delta Function



Shifted Delta Function



Same Magnitude,
Different Phase

Shifted Delta Function

