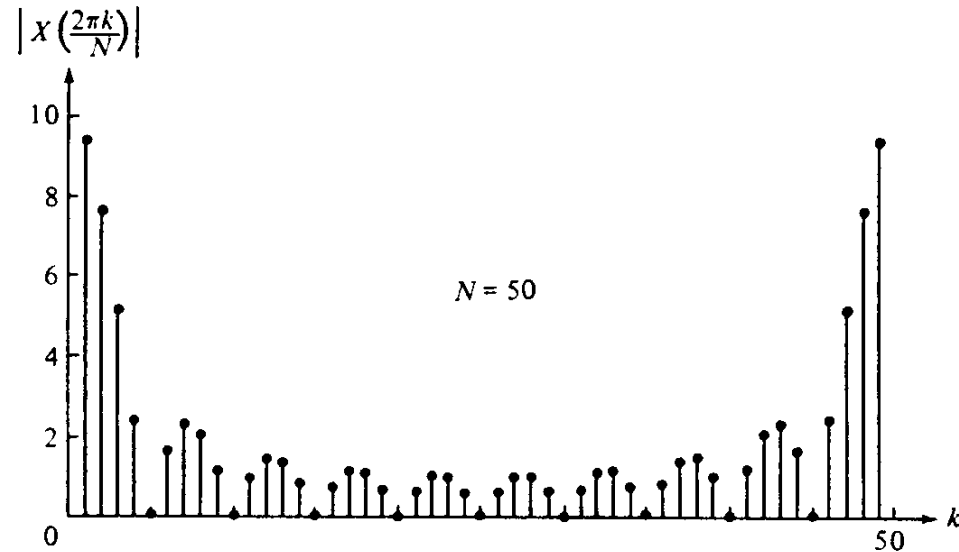




5. Fourier Transform and Spectrum Analysis

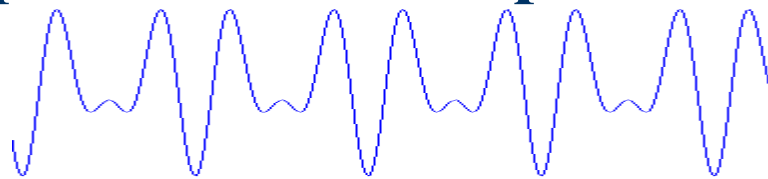


5. Fourier Transform and Spectrum Analysis



Spectrum of Non-periodic Signals

- Fourier series help us to find the spectrum of periodic signals



- Most signals are not periodic

- **Speech, audio, etc.**



- Need another tool to find the spectrum of non-periodic (aperiodic) signals

⇒ **Fourier Transform**



Fourier Transform of Discrete-time Signals

- Let $x(t)$ be an aperiodic continuous-time signal, $x[n]$ is the samples of $x(t)$ such that:

$$x[n] = x(nT_s)$$

- The spectrum of $x[n]$ is given by:

$$X_p(\omega) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega nT_s} \quad \text{or} \quad X_p(\hat{\omega}) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\hat{\omega}n}$$

Radian frequency

$$\hat{\omega} = \omega T_s$$



Aperiodic Signals have Periodic Spectrum

- It is interesting to note that $X_p(\omega)$ is periodic since

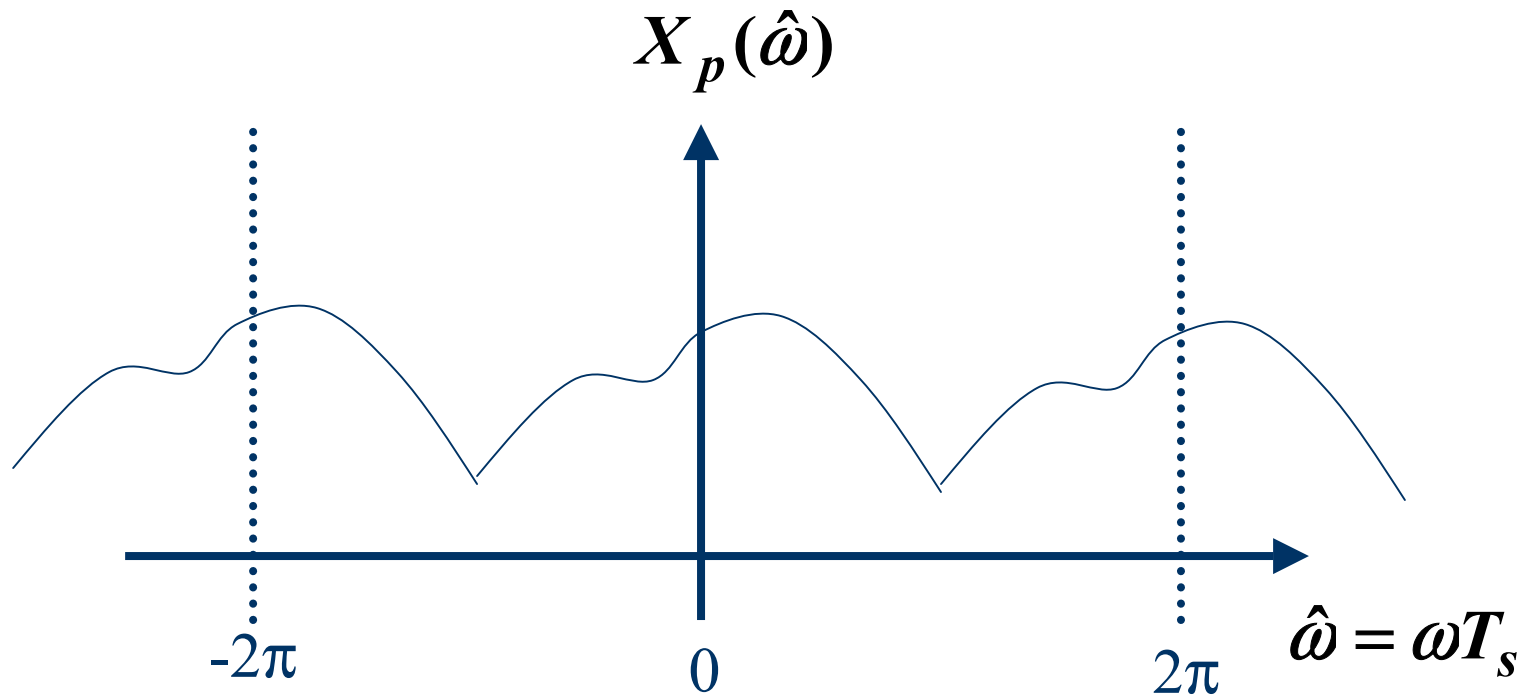
$$\begin{aligned} X_p(\hat{\omega} + 2\pi k) &= \sum_{n=-\infty}^{\infty} x[n] e^{-j(\hat{\omega} + 2\pi k)n} \\ &= \sum_{n=-\infty}^{\infty} x[n] e^{-j\hat{\omega}n} e^{-j2\pi nk} = X_p(\hat{\omega}) \end{aligned}$$

The equation shows the periodicity of the spectrum. A red arrow points from the term $e^{-j2\pi nk}$ to a red '1', indicating that this term is equal to 1, which simplifies the expression to $X_p(\hat{\omega})$.



Signal Processing Fundamentals – Part I
Spectrum Analysis and Filtering

5. Fourier Transform and Spectrum Analysis





- If $x(t)$ has a spectrum of $X(\omega)$ and $x[n] = x(nT_s)$ has a spectrum of $X_p(\omega)$, it can be shown that

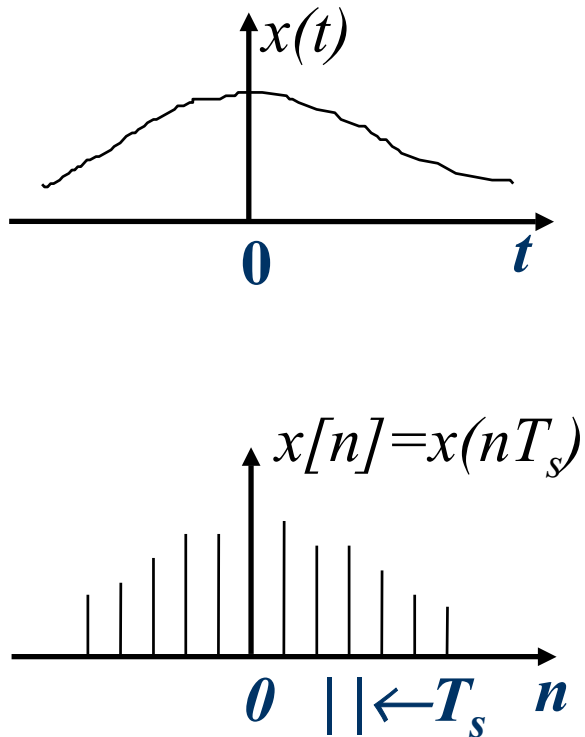
$$\begin{aligned} X_p(\hat{\omega}) &= \frac{1}{T_s} \sum_{k=-\infty}^{\infty} X(\hat{\omega} + 2\pi k) \quad -\infty \leq \omega \leq \infty \\ &= \dots + \frac{1}{T_s} X(\hat{\omega} - 2\pi) \\ &\quad + \frac{1}{T_s} X(\hat{\omega}) \\ &\quad + \frac{1}{T_s} X(\hat{\omega} + 2\pi) + \dots \end{aligned}$$



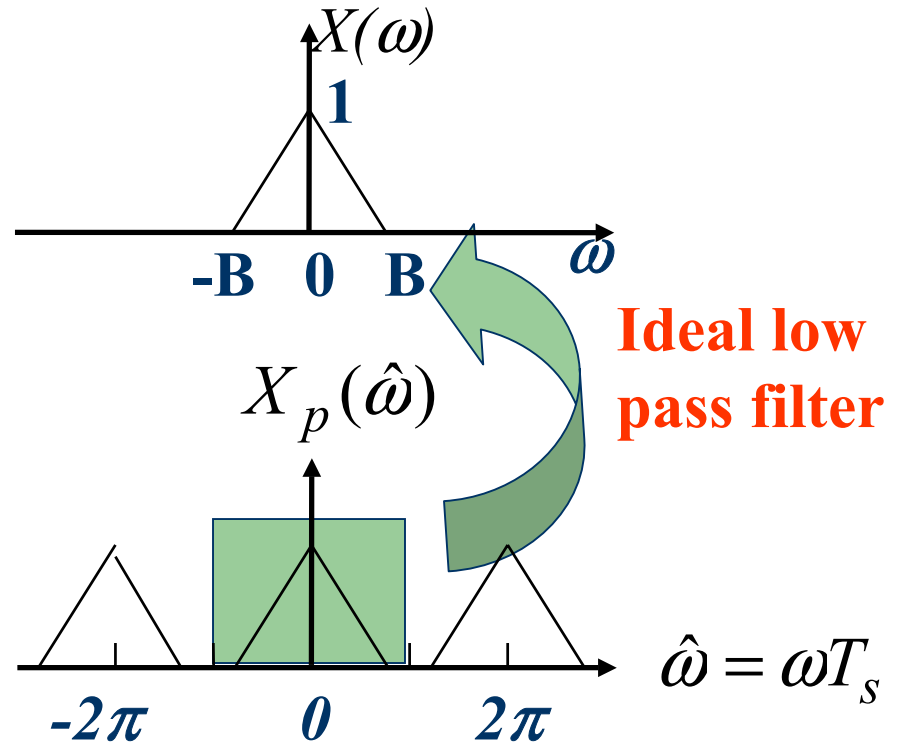
Signal Processing Fundamentals – Part I
Spectrum Analysis and Filtering

5. Fourier Transform and Spectrum Analysis

Time Domain



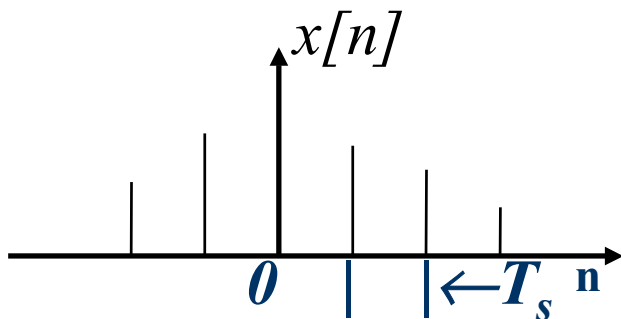
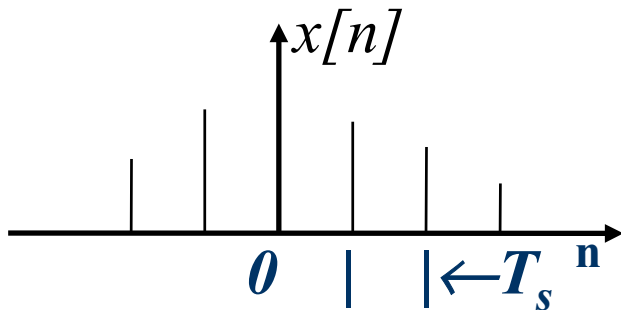
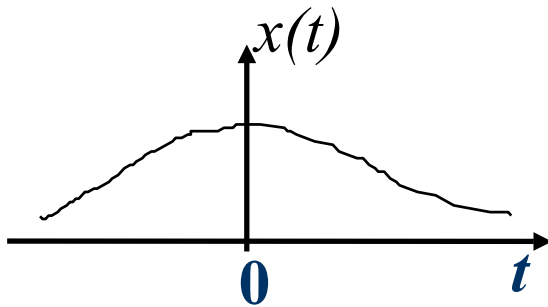
Frequency Domain



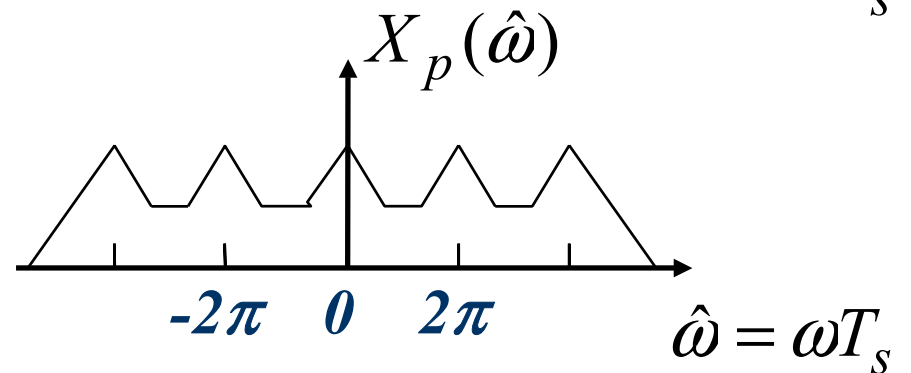
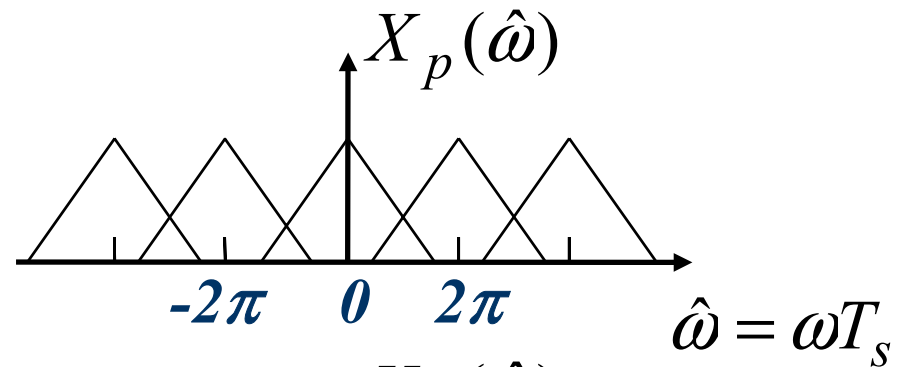
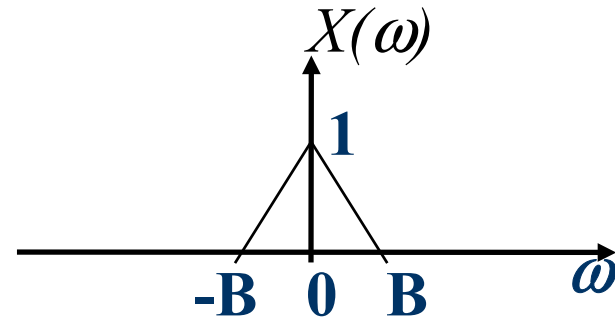


5. Fourier Transform and Spectrum Analysis

Time Domain



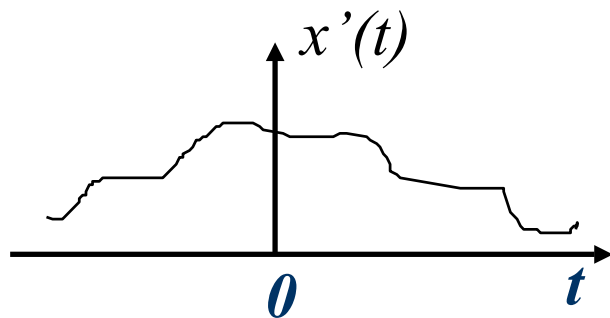
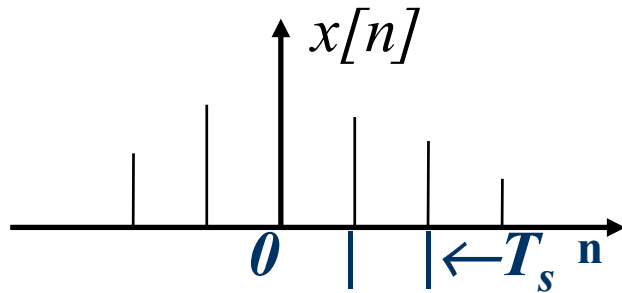
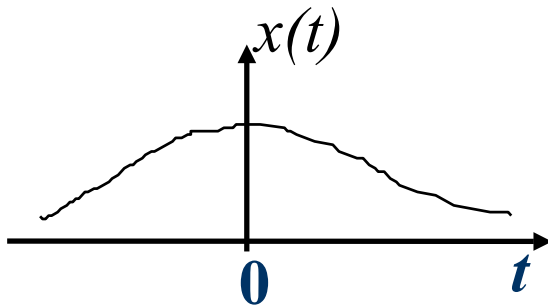
Frequency Domain



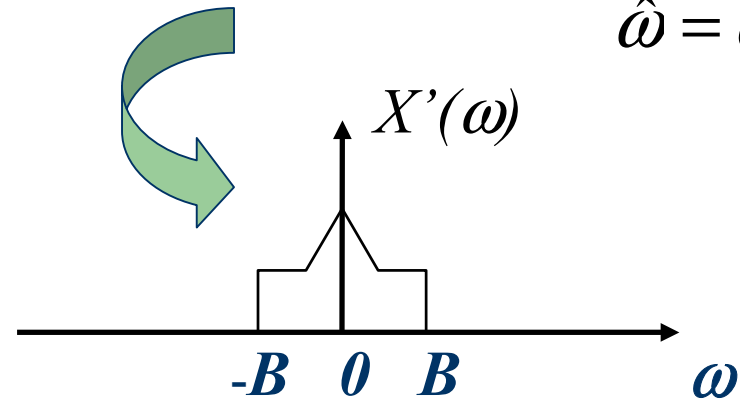
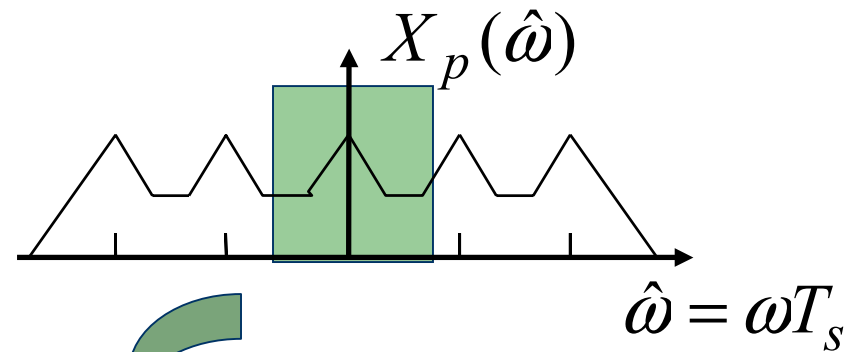
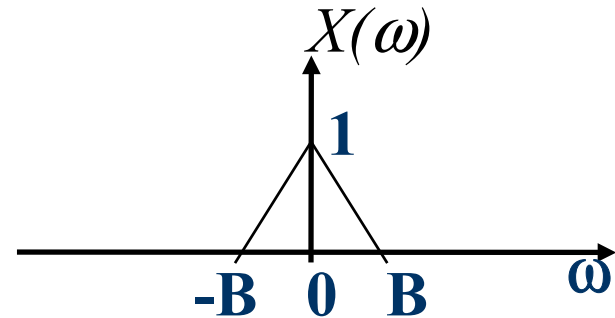


5. Fourier Transform and Spectrum Analysis

Time Domain



Frequency Domain





5. Fourier Transform and Spectrum Analysis

- If the signal has frequency components beyond $|\pi|$, after sampling, these frequency components will affect the other replicas in the spectrum
- Even with an ideal low pass filter, the original signal cannot be reconstructed. This is the so-called **alias effect**
- Restate the **Shannon Sampling Theorem** for general aperiodic signals

$$\begin{aligned}\hat{\omega} \leq |\pi| &\Rightarrow 2\pi f_{\max} T_s \leq |\pi| \\ &\Rightarrow 2\pi f_{\max} \leq |\pi f_s| \quad \text{or} \quad f_s \geq 2f_{\max}\end{aligned}$$



Shannon Sampling Theorem

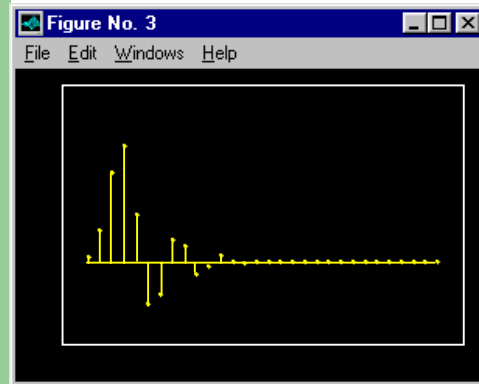
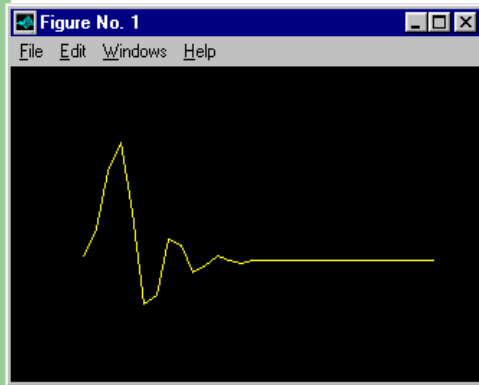
A continuous-time aperiodic signal $x(t)$ with frequencies no higher than f_{max} can be reconstructed exactly from its samples $x[n] = x(nT_s)$ if the samples are taken at a rate $f_s = 1/T_s$ that is greater than $2f_{max}$

Nyquist Frequency

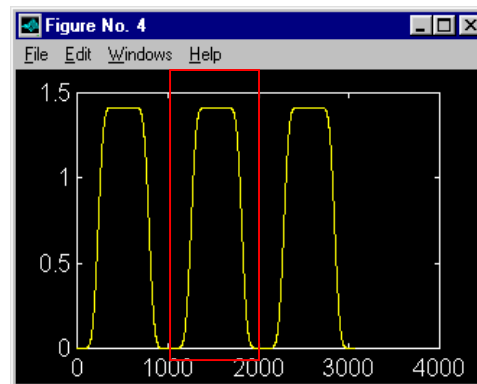
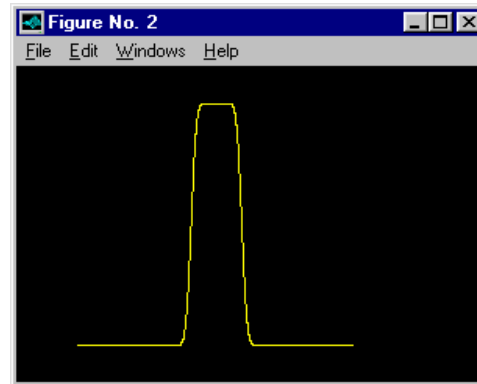


Real Examples

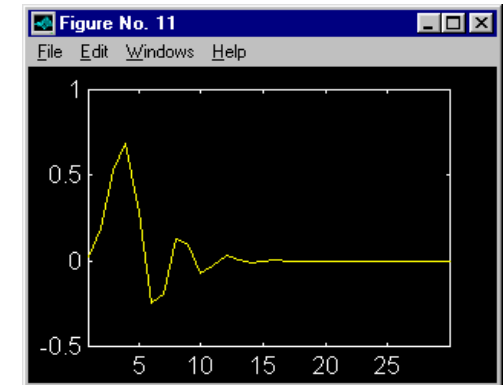
Time Domain



Frequency Domain



Resulted signal



Ideal low pass filter

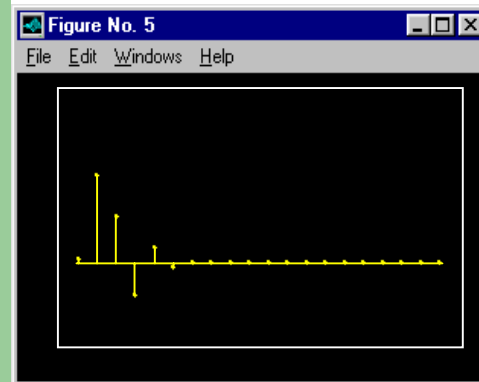
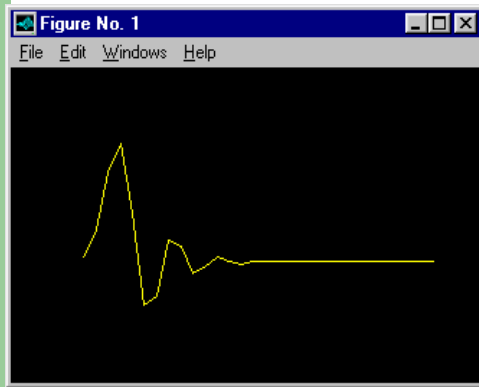


Signal Processing Fundamentals – Part I

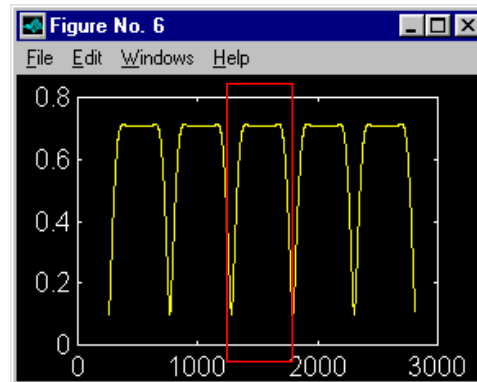
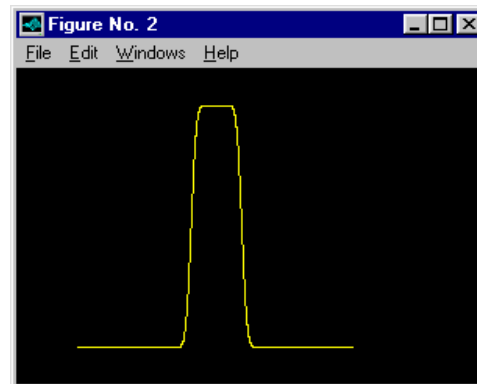
Spectrum Analysis and Filtering

5. Fourier Transform and Spectrum Analysis

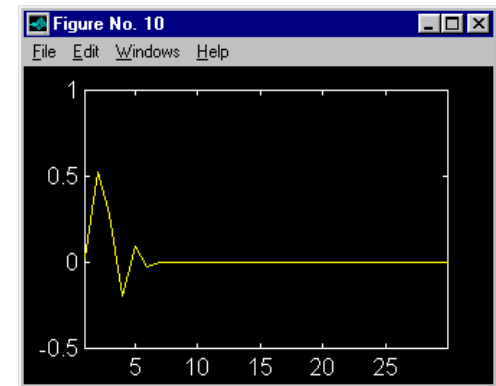
Time Domain



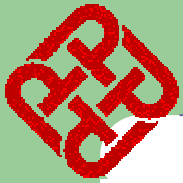
Frequency Domain



Resulted signal



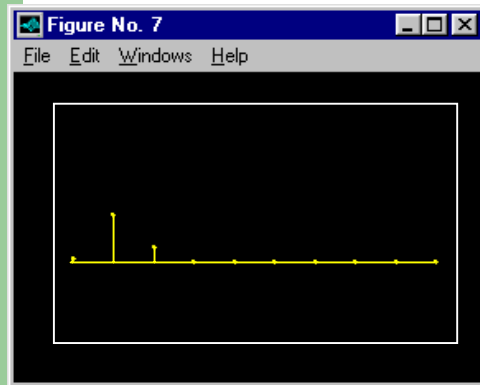
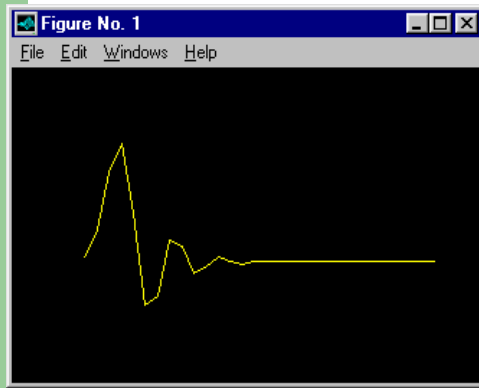
Ideal low pass filter



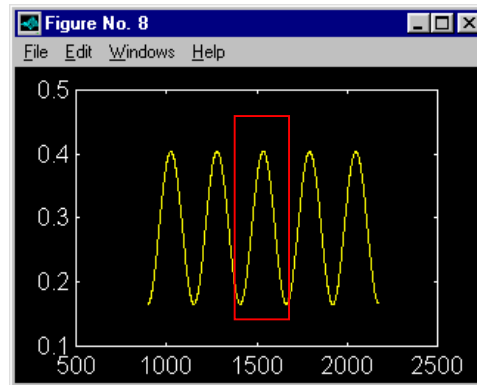
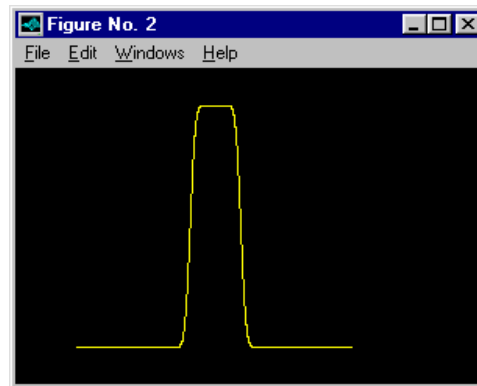
Signal Processing Fundamentals – Part I Spectrum Analysis and Filtering

5. Fourier Transform and Spectrum Analysis

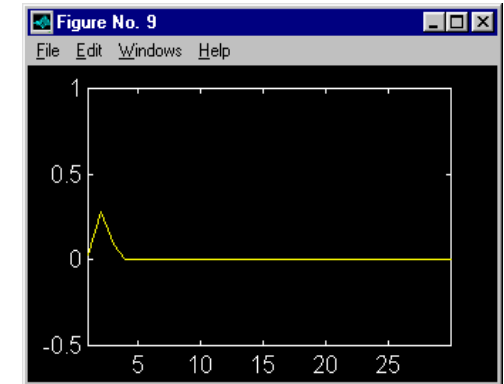
Time Domain



Frequency Domain



Resulted signal



Ideal low pass filter



How to Solve Aliasing Problems?

1. Increase the sampling rate such that $f_s \geq 2f_{max}$
2. Use anti-aliasing filter first

Pre-filter the input signal such that it will never has frequency components beyond $|\pi|$

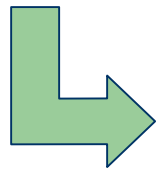
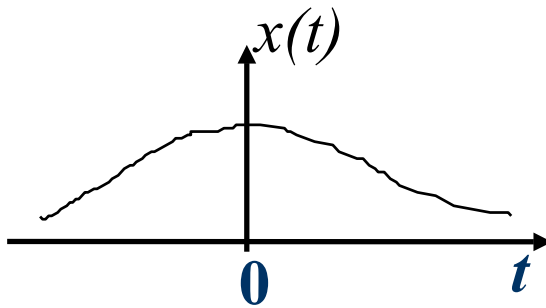




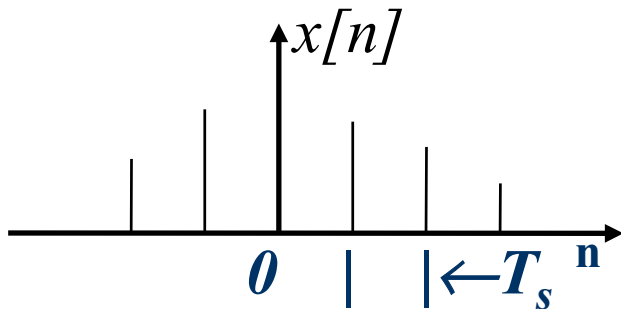
Signal Processing Fundamentals – Part I
Spectrum Analysis and Filtering

5. Fourier Transform and Spectrum Analysis

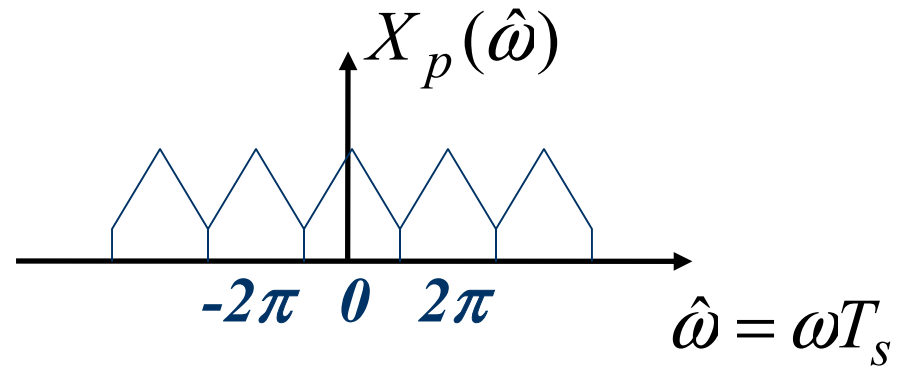
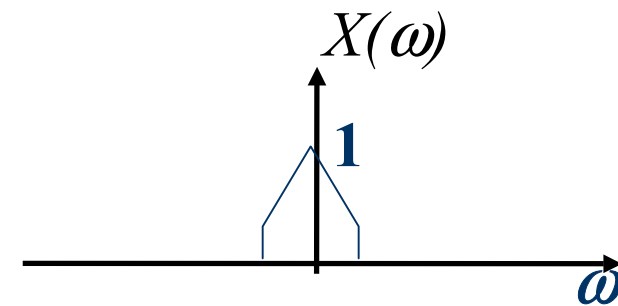
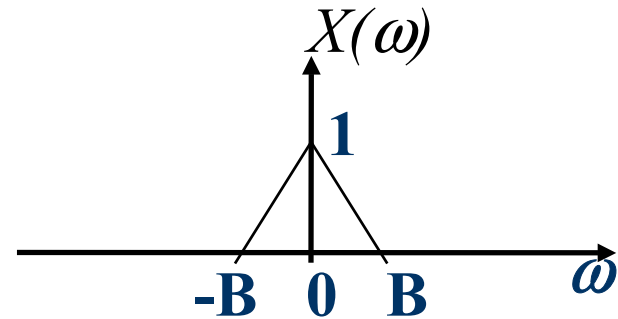
Time Domain

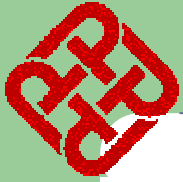


Anti-aliasing
filter



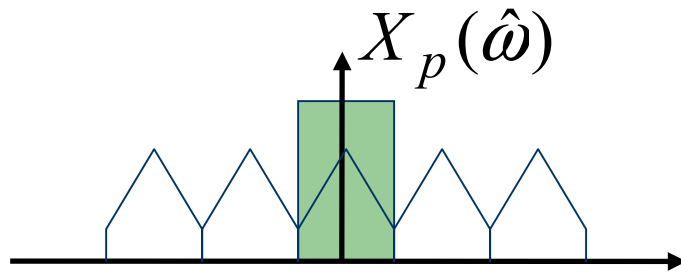
Frequency Domain



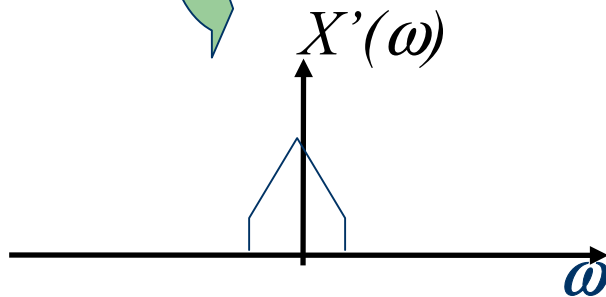


5. Fourier Transform and Spectrum Analysis

With anti-aliasing filter

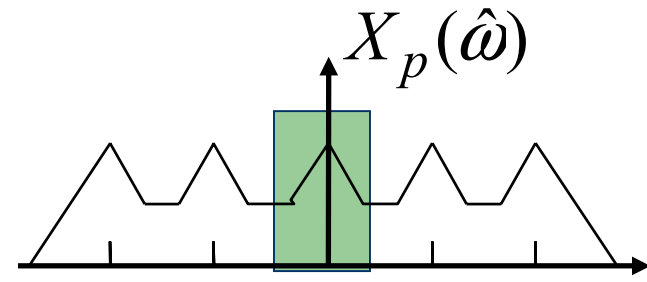


Ideal low pass filter

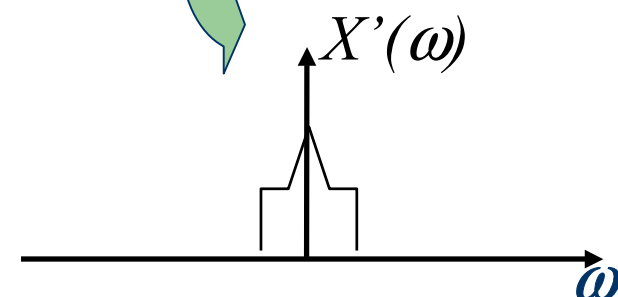


Look better

Without anti-aliasing filter

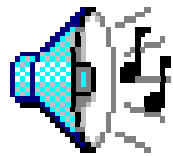


Ideal low pass filter

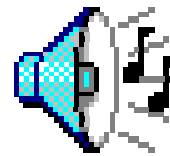




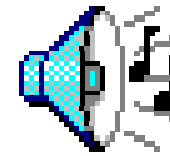
Hear the effect!



Original
44.1kHz
sampling



8kHz
sampling
with
aliasing



8kHz
sampling
with anti-
aliasing filter



Discrete Fourier Transform

- Spectrum of aperiodic discrete-time signals is **periodic and continuous**
- Difficult to be handled by computer
- Since the spectrum is periodic, there's no point to keep all periods – **one period is enough**
- Computer cannot handle continuous data, we can only **keep some samples of the spectrum**
- Interesting enough, such requirements lead to a very simple way to find the spectrum of signals
⇒ **Discrete Fourier Transform**



- Recall the Fourier transform of an aperiodic discrete sequence

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

- Assume $x[n]$ is an aperiodic sequence with N values, i.e. $\{x[n] : n = 0, 1, \dots, N-1\}$

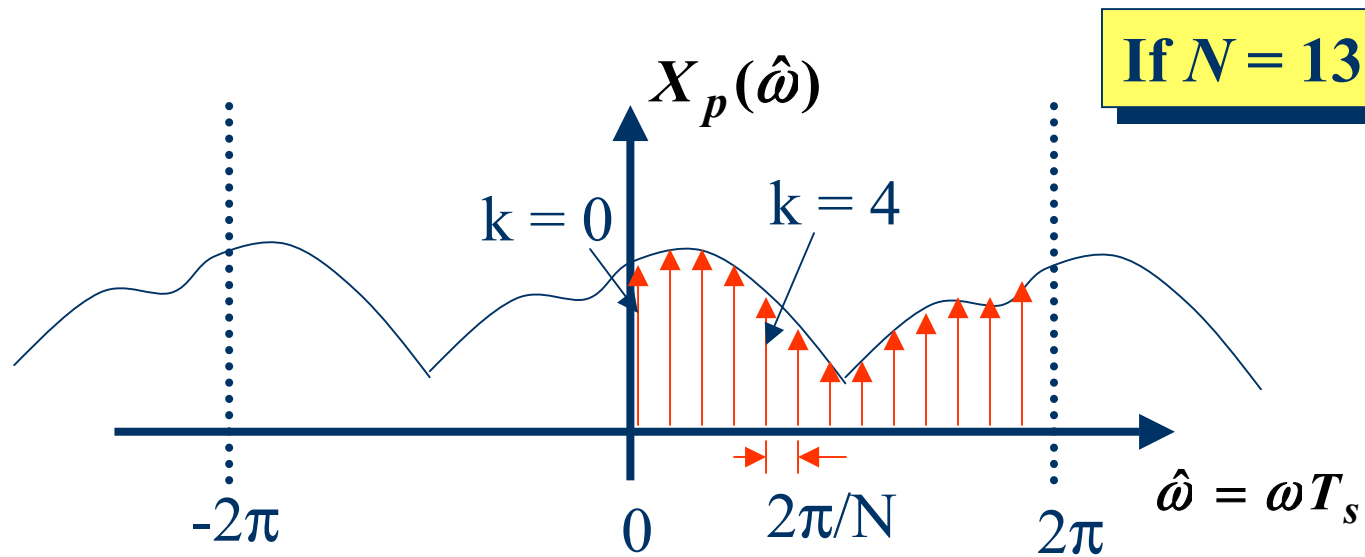
$$X_p(\hat{\omega}) = \sum_{n=0}^{N-1} x[n]e^{-j\hat{\omega}n}$$



5. Fourier Transform and Spectrum Analysis

- If we are now interested only in N equally spaced frequencies of 1 period of the Fourier spectrum, i.e.

$$X[k] = X_p\left(\frac{k \cdot 2\pi}{N}\right) \quad k = 0, 1, \dots, N-1$$





5. Fourier Transform and Spectrum Analysis

- Now if we want to compute the value of these N frequencies,

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j \left(\frac{k \cdot 2\pi}{T_s N} \right) n T_s}$$

$$W_N^{nk} = e^{-j 2\pi n k / N}$$

$$= \sum_{n=0}^{N-1} x[n] e^{-j k \cdot 2\pi n / N} = \sum_{n=0}^{N-1} x[n] W_N^{nk}$$

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

Discrete Fourier Transform



5. Fourier Transform and Spectrum Analysis

- **Discrete Fourier Transform (DFT)** is exactly the output of the Fourier Transform of an aperiodic sequence at some particular frequencies

- Inherently **periodic** since $X[k+N] = X[k]$, although we always only consider one period of $X[k]$

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

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

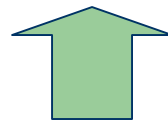
$$= \sum_{n=0}^{N-1} x[n] e^{-j2\pi nk / N} e^{-j2\pi n} = X[k]$$

(Note: A red arrow points from the exponent $-j2\pi n$ to the number 1, indicating that $e^{-j2\pi n} = 1$.)



- If we know $X[k]$, we can reconstruct back the signal $x[n]$ via the **inverse discrete Fourier transform**

$$\begin{aligned}x[n] &= \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j2\pi nk / N} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-nk} \quad \text{for } n = 0, 1, \dots, N-1\end{aligned}$$



Inverse Discrete Fourier Transform



5. Fourier Transform and Spectrum Analysis

- It can be proven as follows:

$$\begin{aligned} \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{jk2\pi n / N} &= \frac{1}{N} \sum_{k=0}^{N-1} \sum_{m=0}^{N-1} x[m] e^{-jk2\pi m / N} e^{jk2\pi n / N} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \sum_{m=0}^{N-1} x[m] e^{jk.2\pi(n-m) / N} \\ &= \frac{1}{N} \sum_{m=0}^{N-1} x[m] \left[\sum_{k=0}^{N-1} e^{jk.2\pi(n-m) / N} \right] \\ &= x[n] \quad \text{for } n = 0, 1, \dots, N-1 \end{aligned}$$

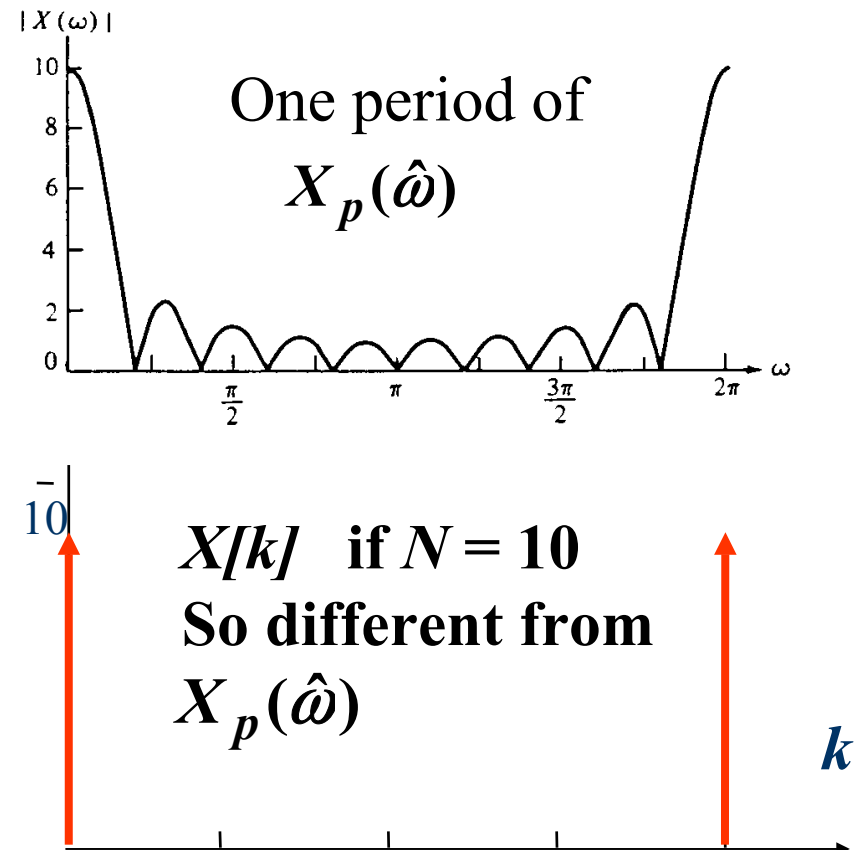
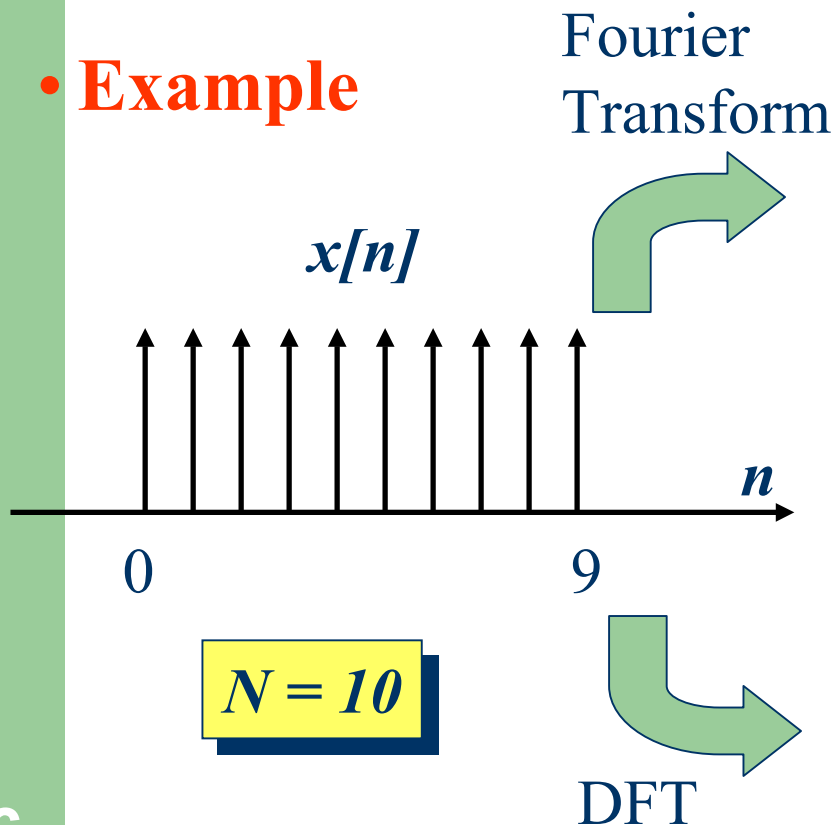
$$\begin{cases} 0 & \text{if } n \neq m \\ N & \text{otherwise} \end{cases}$$



5. Fourier Transform and Spectrum Analysis

- Although DFT gives exact frequency response of a signal, sometimes it may not give the desired spectrum

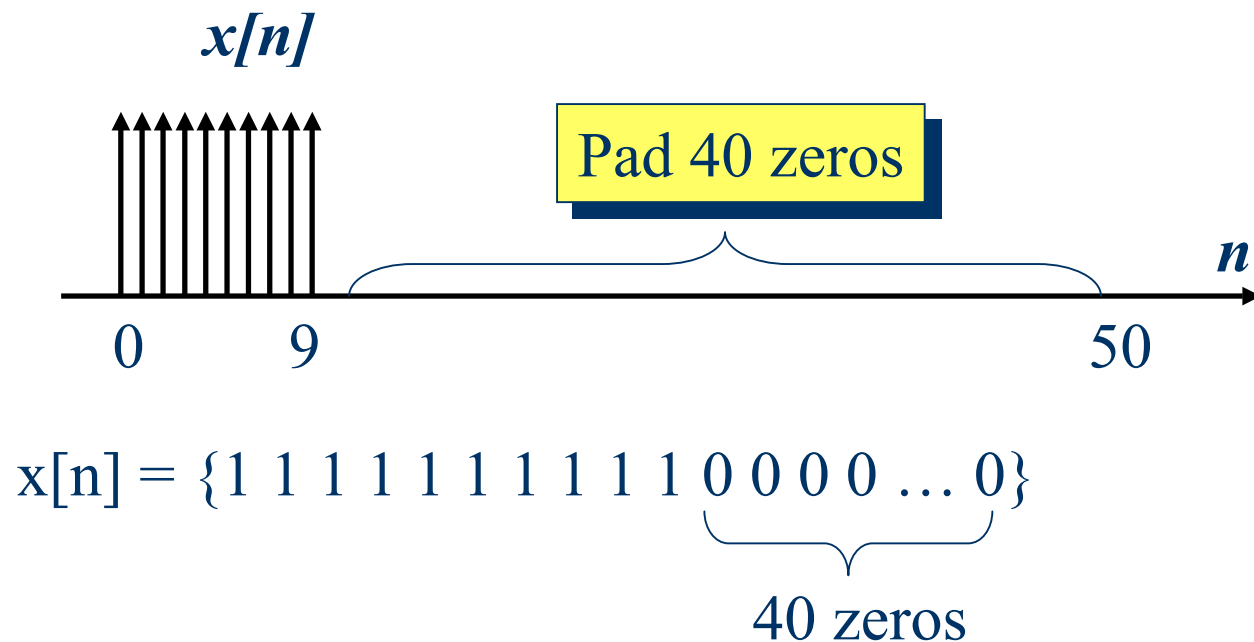
- **Example**

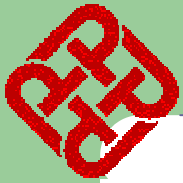




5. Fourier Transform and Spectrum Analysis

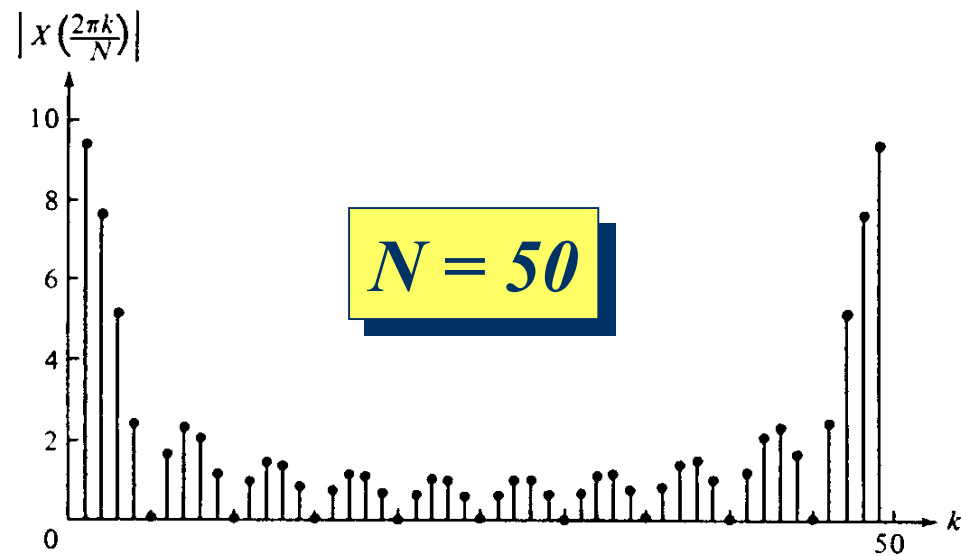
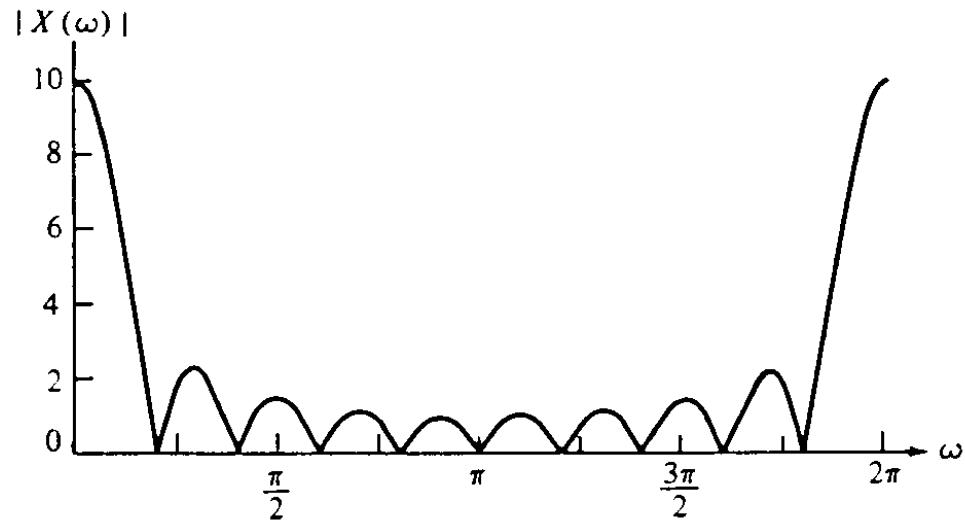
- Need **improved resolution**
- Achieve by **padding zero** to the end of $x[n]$ to make N bigger





Signal Processing Fundamentals – Part I
Spectrum Analysis and Filtering

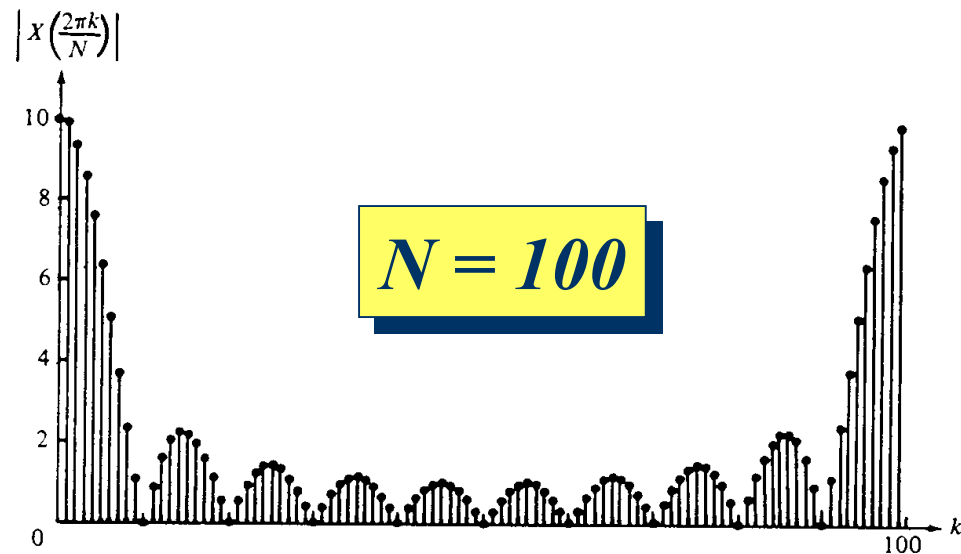
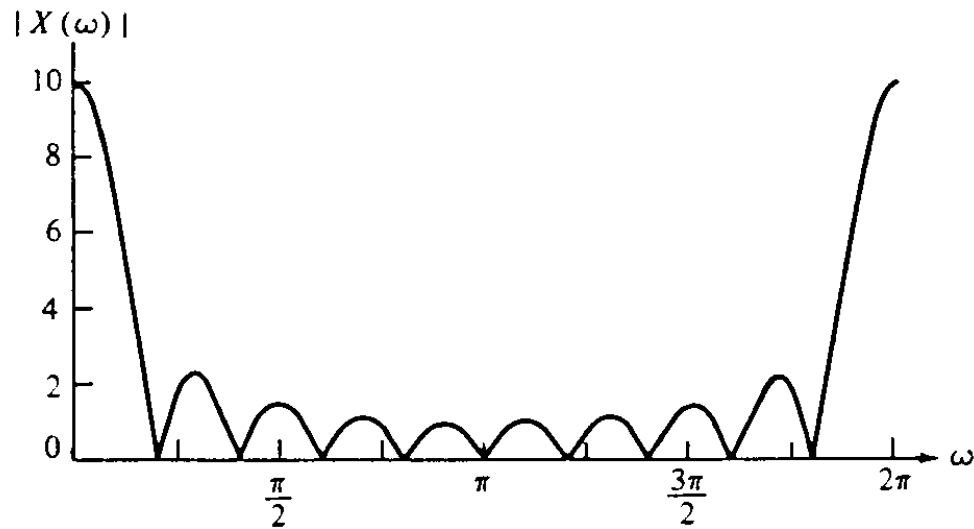
5. Fourier Transform and Spectrum Analysis





Signal Processing Fundamentals – Part I
Spectrum Analysis and Filtering

5. Fourier Transform and Spectrum Analysis





Exercise

Given that $x[n]$ is defined in the following figure, determine its spectrum using DFT with $N = 4$

