

# CE 609

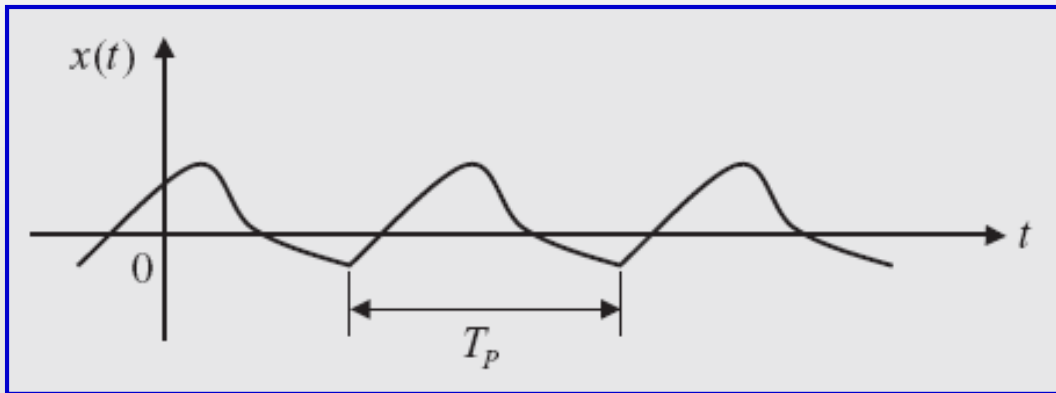
## LECTURE : Fast Fourier transform

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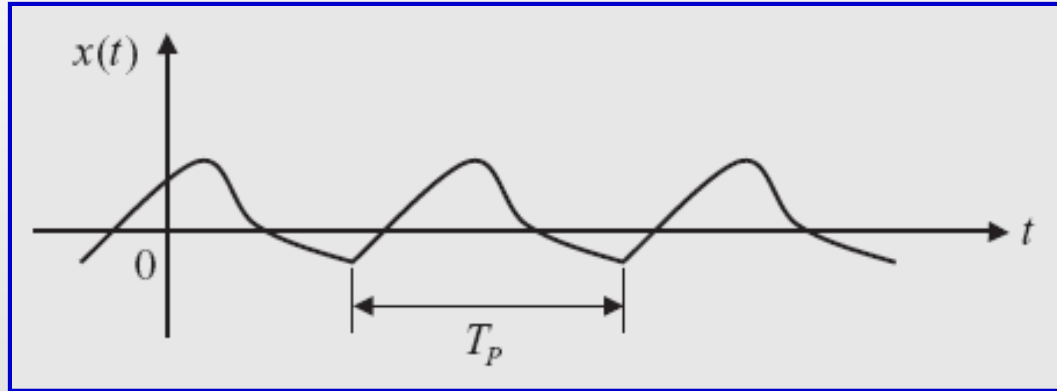
# Fourier series



A period signal with a period  $T_P$

$$x(t) = \frac{a_0}{2} + \sum_{n=1}^{\infty} \left[ a_n \cos \left( \frac{2\pi nt}{T_P} \right) + b_n \sin \left( \frac{2\pi nt}{T_P} \right) \right]$$

# Fourier series



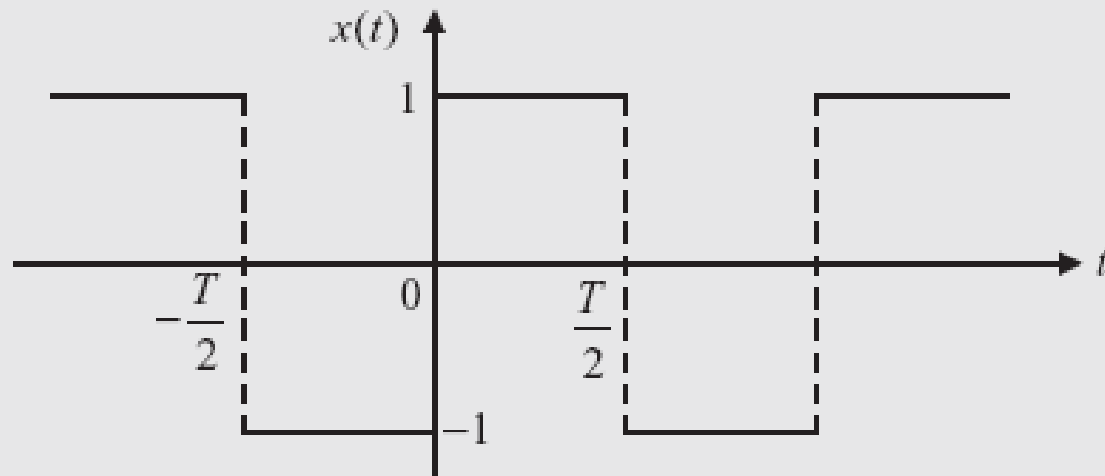
$$\frac{a_0}{2} = \frac{1}{T_P} \int_0^{T_P} x(t) dt = \frac{1}{T_P} \int_{-T_P/2}^{T_P/2} x(t) dt : \quad \text{mean value}$$

$$a_n = \frac{2}{T_P} \int_0^{T_P} x(t) \cos\left(\frac{2\pi nt}{T_P}\right) dt = \frac{2}{T_P} \int_{-T_P/2}^{T_P/2} x(t) \cos\left(\frac{2\pi nt}{T_P}\right) dt \quad n = 1, 2, \dots$$

$$b_n = \frac{2}{T_P} \int_0^{T_P} x(t) \sin\left(\frac{2\pi nt}{T_P}\right) dt = \frac{2}{T_P} \int_{-T_P/2}^{T_P/2} x(t) \sin\left(\frac{2\pi nt}{T_P}\right) dt \quad n = 1, 2, \dots$$

# Fourier series: Square wave

$$x(t) = \begin{cases} -1 & -\frac{T}{2} < t < 0 \\ 1 & 0 < t < \frac{T}{2} \end{cases} \quad \text{and} \quad x(t + nT) = x(t) \quad n = \pm 1, \pm 2, \dots$$



# Fourier series

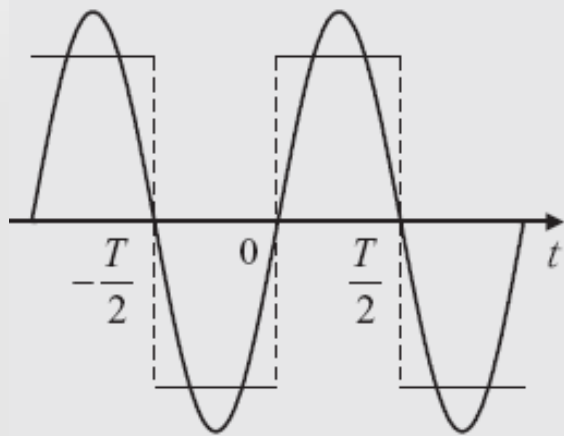
$$a_0 = 0; \quad a_n = 0$$

$$b_n = \frac{2}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t) \sin\left(\frac{2n\pi t}{T}\right) dt = \frac{2}{n\pi} (1 - n\pi)$$

$$x(t) = \frac{4}{\pi} \left[ \sin\left(\frac{2\pi t}{T}\right) + \frac{1}{3} \sin\left(\frac{2\pi 3t}{T}\right) + \frac{1}{5} \sin\left(\frac{2\pi 5t}{T}\right) + \dots \right]$$

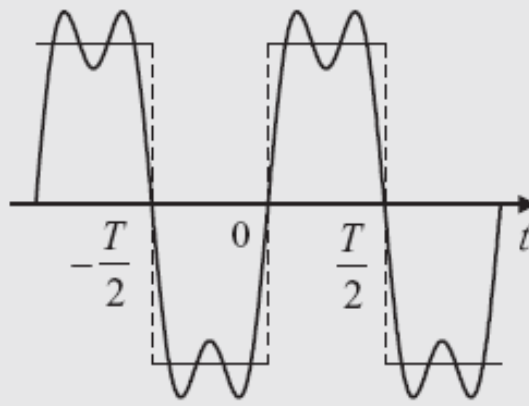
$$x(t) = \frac{4}{\pi} \left[ \sin \omega_1 t + \frac{1}{3} \sin 3\omega_1 t + \frac{1}{5} \sin 5\omega_1 t + \dots \right]$$

# Fourier series



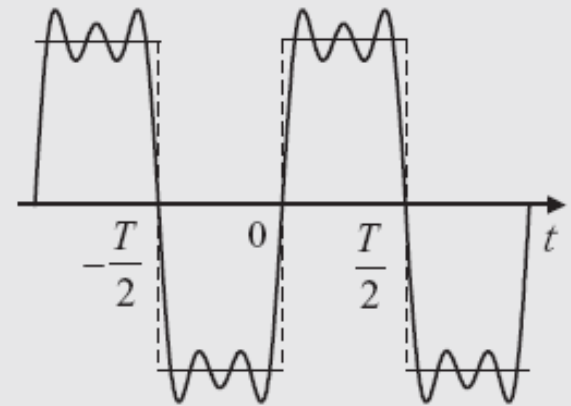
1 Term

$$S_1(t) = \frac{4}{\pi} \sin \omega_1 t$$



2 Terms

$$S_2(t) = \frac{4}{\pi} \left[ \sin \omega_1 t + \frac{1}{3} \sin 3\omega_1 t \right]$$

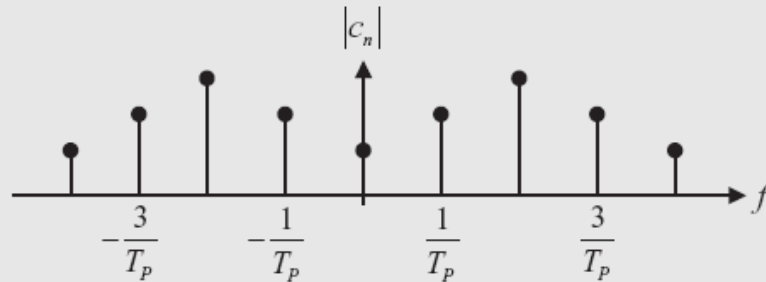


3 Terms

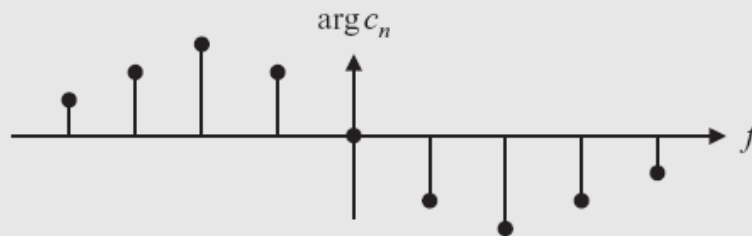
$$S_3(t) = \frac{4}{\pi} \left[ \sin \omega_1 t + \frac{1}{3} \sin 3\omega_1 t + \frac{1}{5} \sin 5\omega_1 t \right]$$

# Fourier series-amplitude spectra

$$x(t) = \sum_{n=-\infty}^{\infty} c_n e^{\frac{j2\pi nt}{T}} \quad c_n = \frac{2}{T} \int_0^T x(t) e^{-\frac{j2\pi nt}{T}} dt$$



Amplitude spectrum of a Fourier series (a *line spectrum* and an *even function*)



Phase spectrum of a Fourier series (a *line spectrum* and an *odd function*)

# MATLAB example

```
clc; close all; clear all
```

```
% Keep hitting enter button if you want to see the term by term  
approximation
```

```
t=[0:0.001:1];
```

```
x=[]; x_tmp=zeros(size(t));
```

```
for n=1:2:39
```

```
x_tmp=x_tmp+4/pi*(1/n*sin(2*pi*n*t));
```

```
x=[x; x_tmp];
```

```
end
```

```
figure,
```

```
for i=1:20
```

```
    drawnow
```

```
        plot(t,x(i,:)) %plot(t,x(i,:),t,x(7,:),t,x(20,:));
```

```
        xlabel('\itt\rm (seconds)'); ylabel('\itx\rm(\itt\rm)')
```

```
        grid on
```

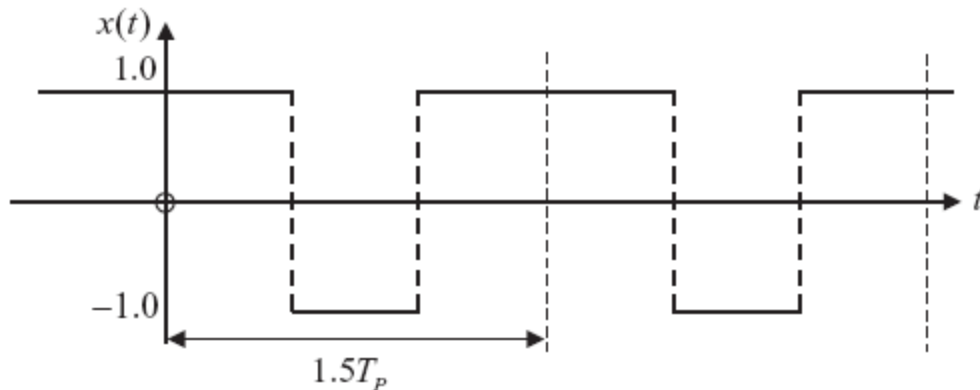
```
    pause
```

```
end
```



# MATLAB exercise: Square wave

$$\left. \begin{aligned} |c_n| &= \frac{2}{n\pi} && \text{for } n = \text{odd} \\ &= 0 && \text{for } n = 0, \text{ even} \end{aligned} \right\}$$



Examine the Fourier coefficients of the square wave for ( $T_p = 1$  s)

- 1)  $r$  an integer number
- 2)  $r$  : a fractional number

$$c_n = \frac{1}{rT_p} \int_0^{rT_p} x(t) e^{-j \frac{2\pi n}{rT_p} t} dt$$

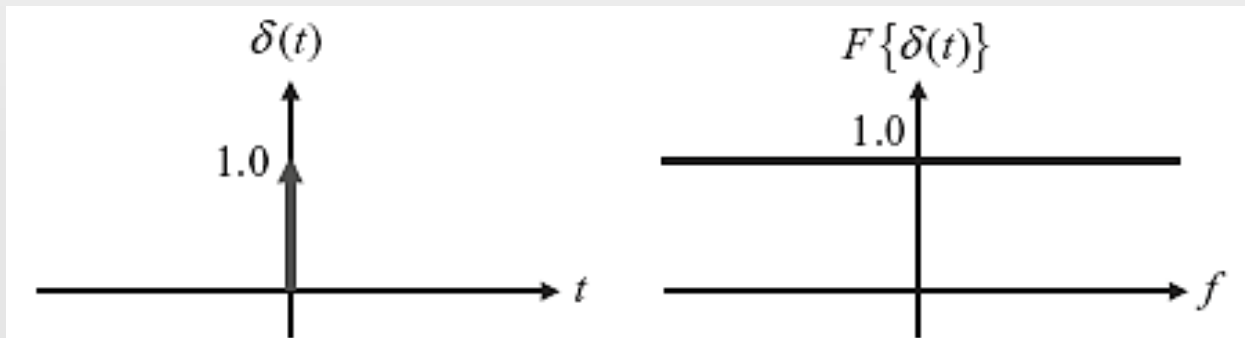
# Fourier series to Fourier transform

- Extension of Fourier analysis to non-periodic phenomena
- Discrete to continuous
- Skipping essential steps, in the limit  $T_p \rightarrow \infty$

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi f t} dt \quad \leftarrow \text{Fourier transform}$$
$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi f t} df \quad \leftarrow \text{Inverse Fourier transform}$$

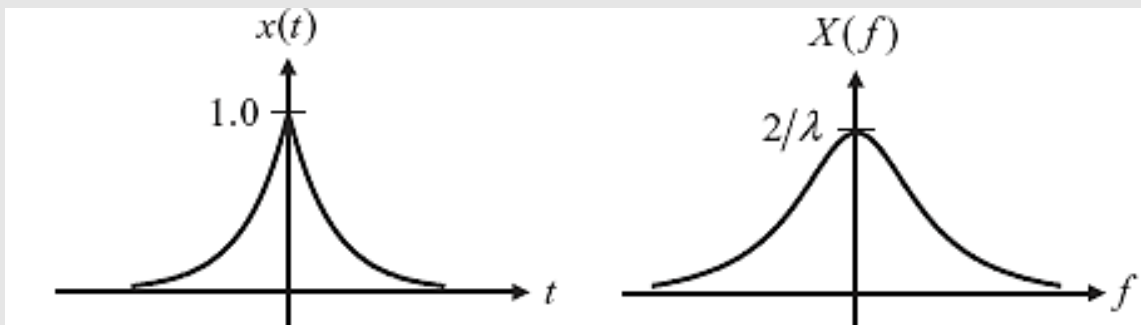
# Some examples: Try to commit some of them to memory, it helps !

## 1) Dirac delta



## 2) Symmetric exponential

$$x(t) = e^{-\lambda|t|}, \quad \lambda > 0$$

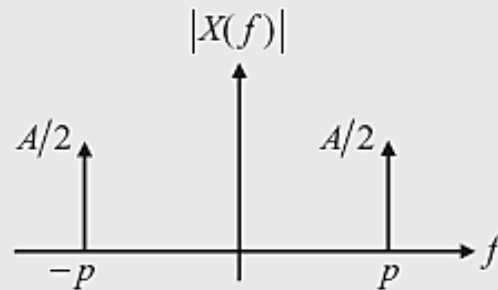
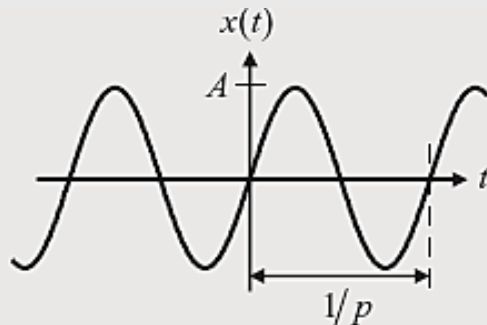


$$X(f) = \frac{2\lambda}{\lambda^2 + 4\pi^2 f^2}$$

# Some examples

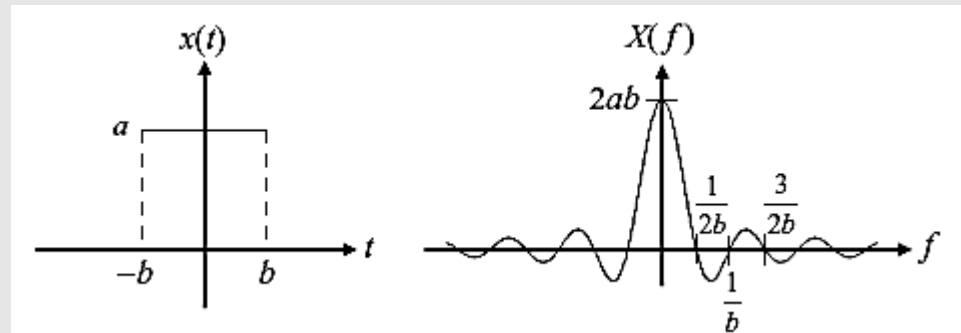
## 3) Sinusoid

$$x(t) = \sin(2\pi f_0 t) \text{ or } \sin(\omega_0 t) \quad X(f) = \frac{1}{2j} [\delta(f - f_0) - \delta(f + f_0)]$$



## 4) Window function

$$\begin{aligned} x(t) &= a & |t| < b \\ &= 0 & |t| > b \end{aligned}$$



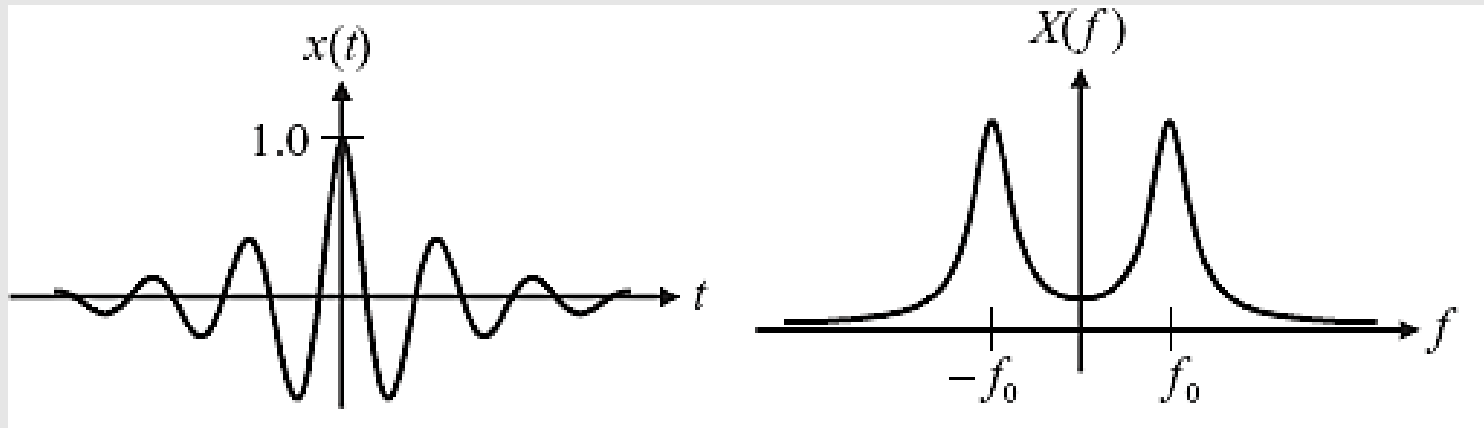
$$X(f) = \frac{2ab \sin(2\pi f b)}{2\pi f b}$$

# Some examples

## 5) Damped symmetrically oscillating function

$$x(t) = e^{-a|t|} \cos 2\pi f_0 t, \quad a > 0$$

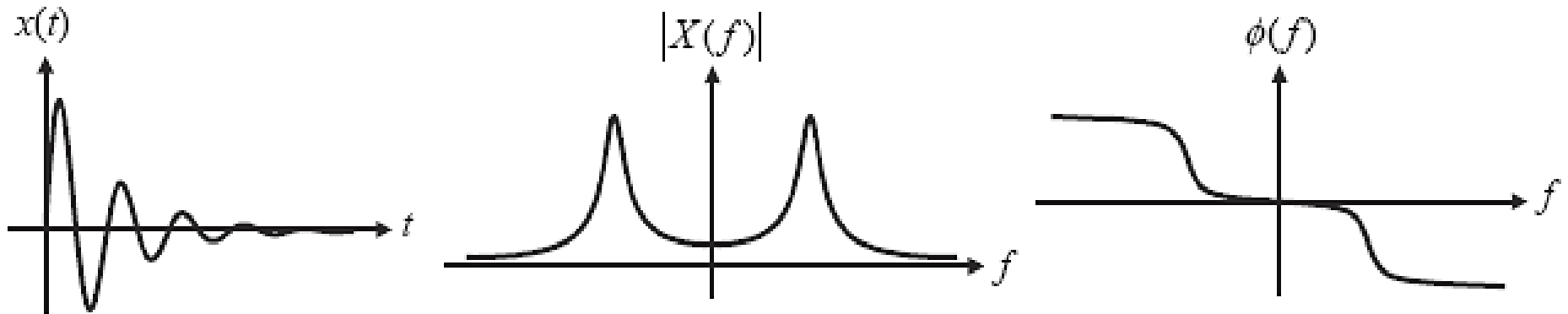
$$X(f) = \frac{a}{a^2 + [2\pi(f - f_0)]^2} + \frac{a}{a^2 + [2\pi(f + f_0)]^2}$$



## 6) Damped oscillating function

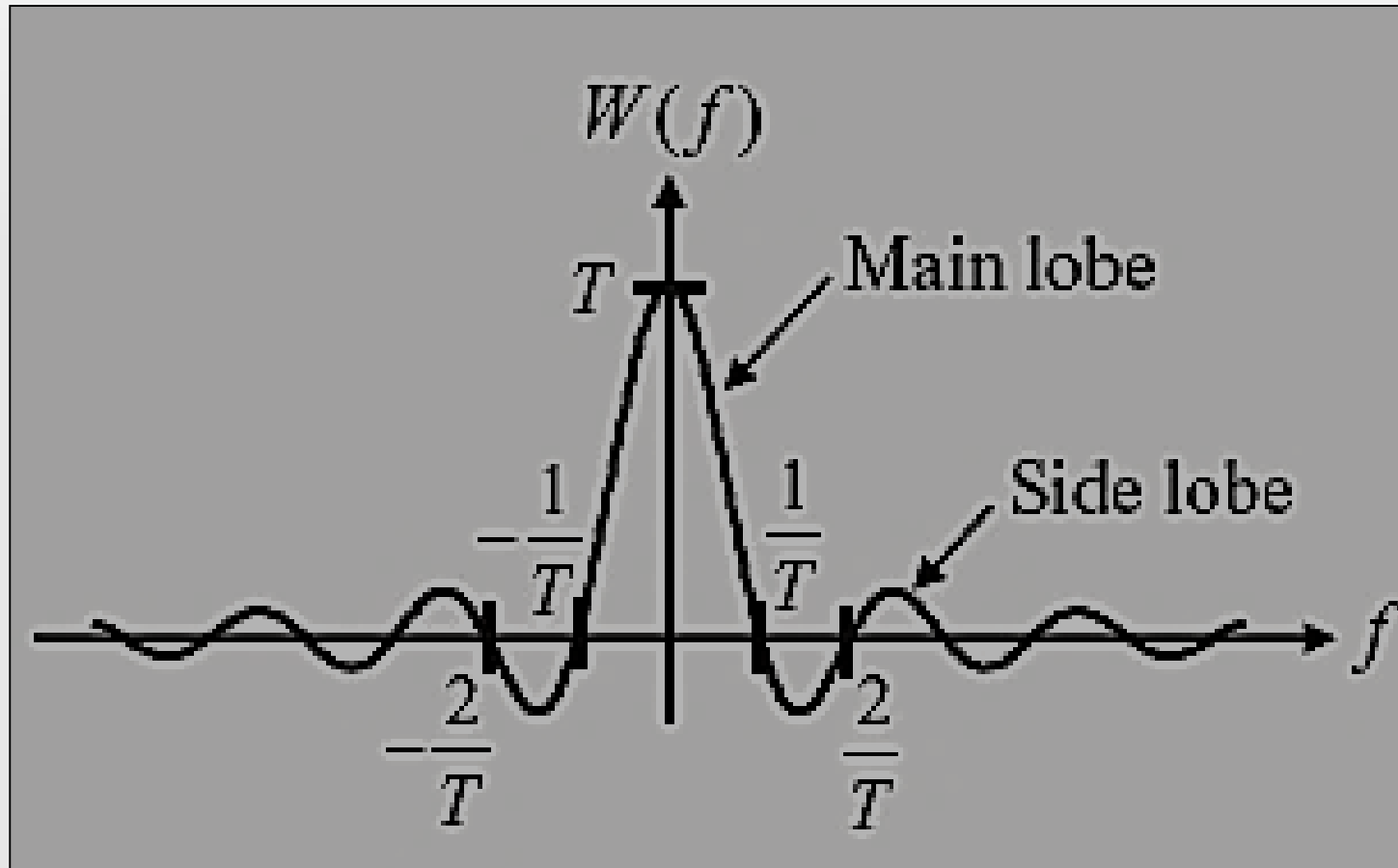
$$x(t) = e^{-at} \sin 2\pi f_0 t, \quad t \geq 0 \text{ and } a > 0$$

$$X(f) = \frac{2\pi f_0}{(2\pi f_0)^2 + (a + j2\pi f)^2}$$



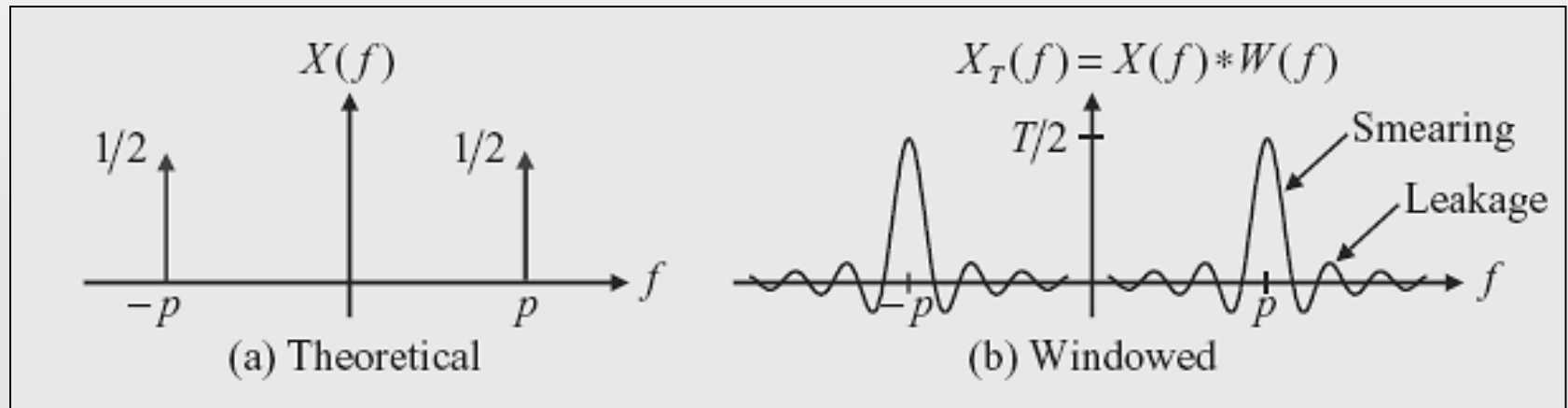
# Windowing

# Fourier transform of the rectangular window



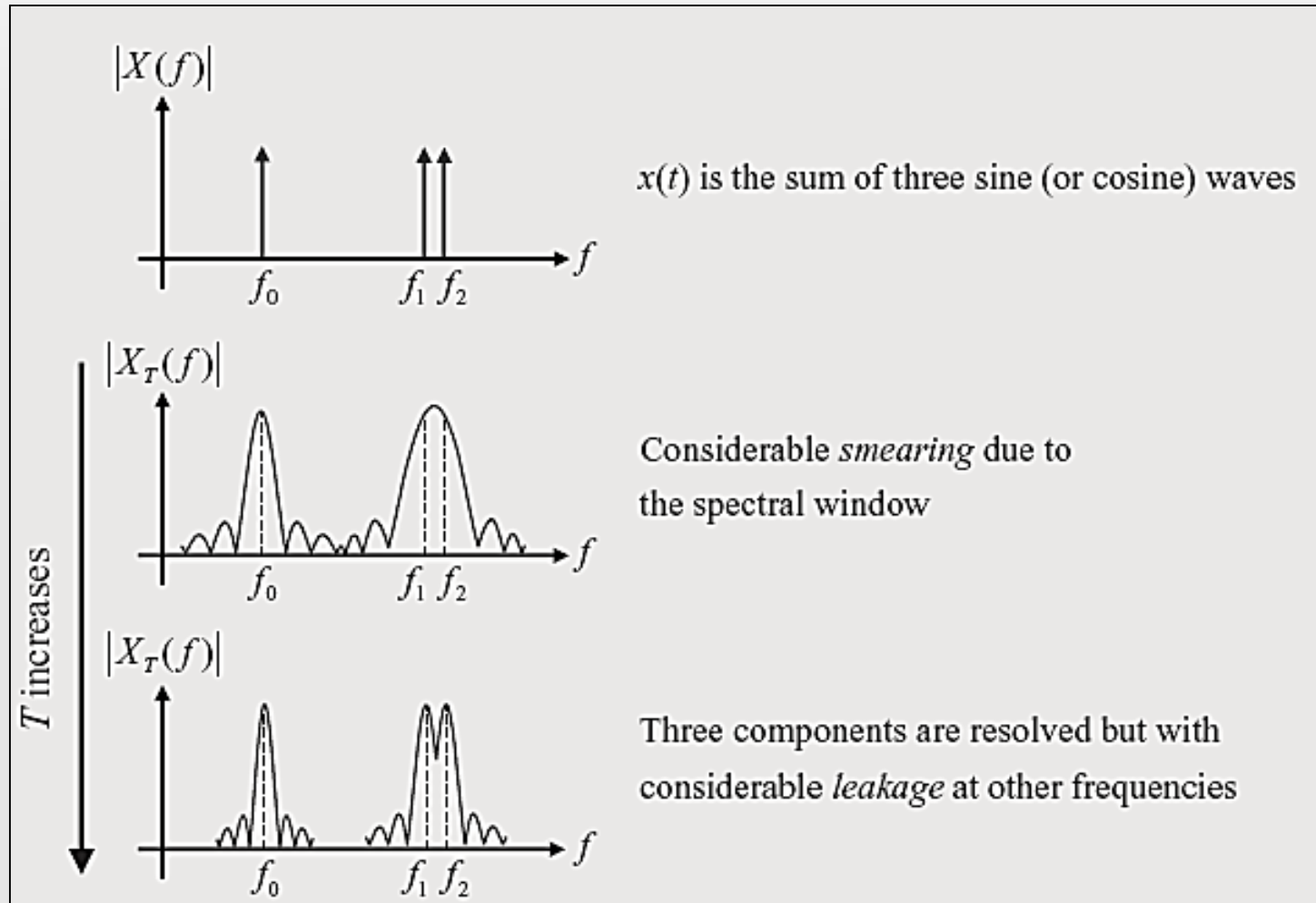


# Windowing: a simple illustration



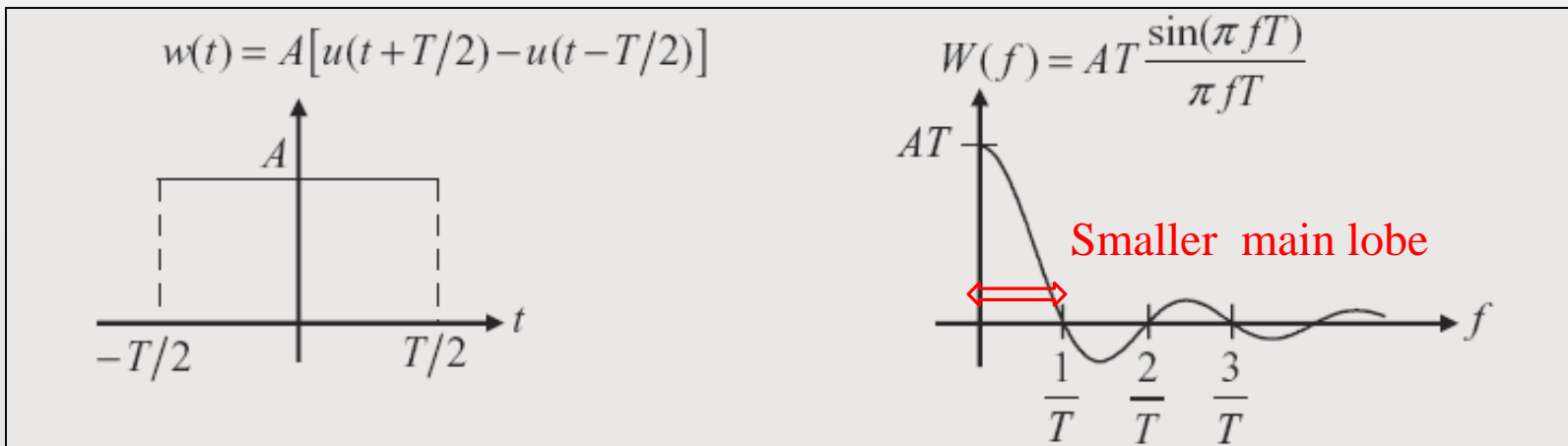
The distortion due to the main lobe is sometimes called *smearing*, and the distortion caused by the side lobes is called *leakage*.

# Effects of windowing

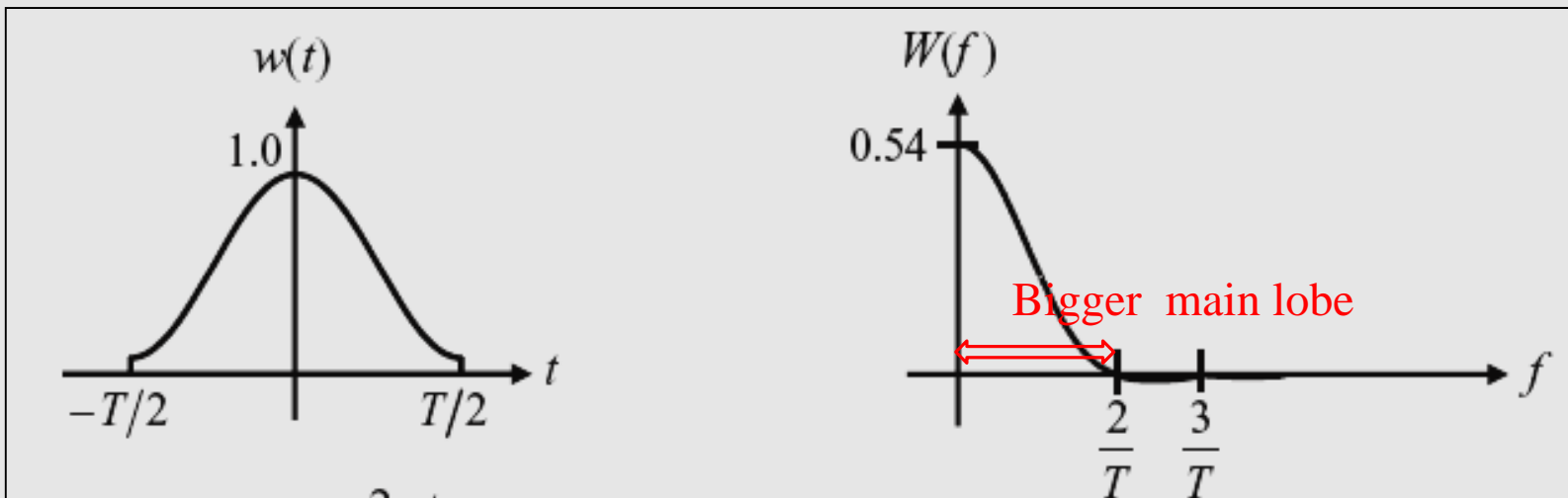


# Common window functions

## Rectangular Window

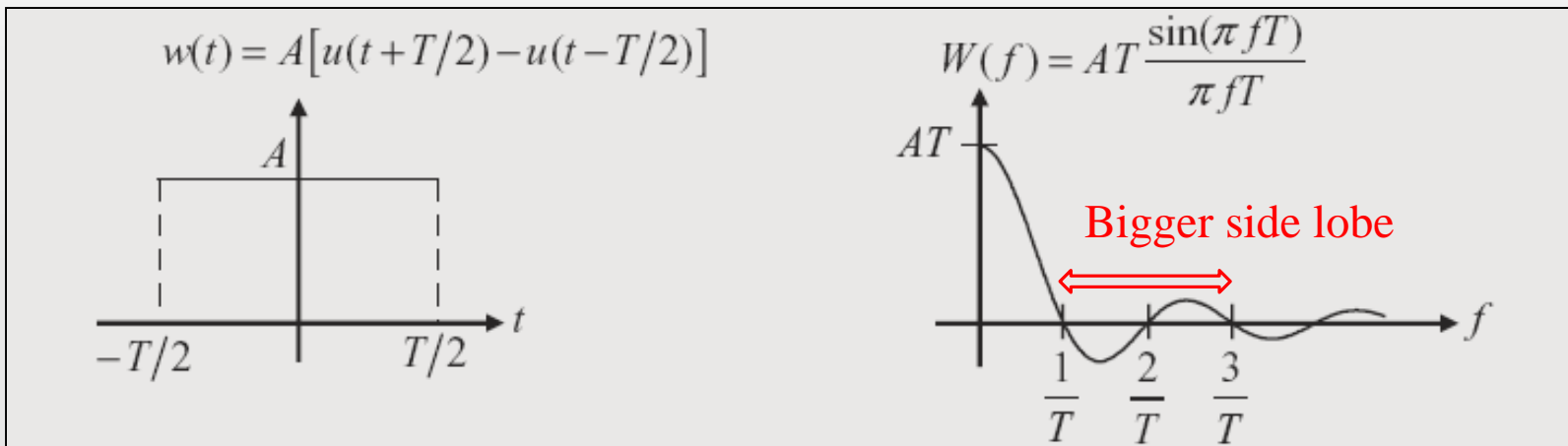


## Hann Window

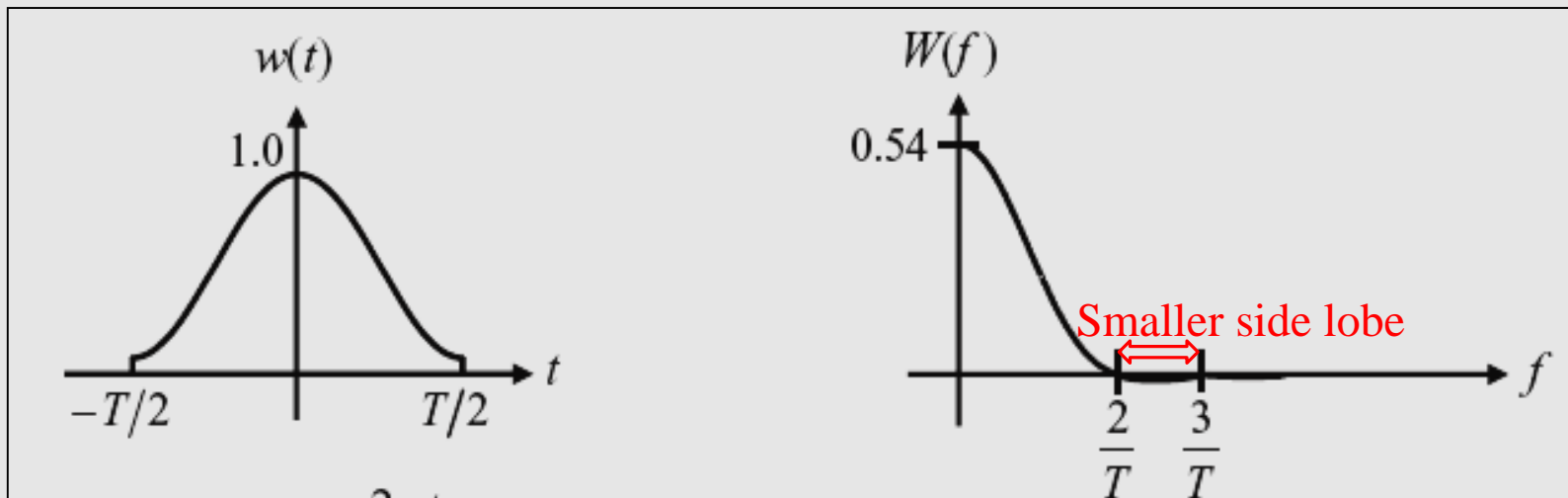


# Common window functions

## Rectangular Window



## Hann Window



# Some comments

- The rectangular window **may be good for separating closely spaced sinusoidal components**, but the leakage is the price to pay.
- The Hann window is a good general purpose window, and has a moderate frequency resolution and a good side lobe roll-off characteristic.

# Discrete Fourier transform

- Consider a sequence  $x(n\Delta)$  at  $n = 0, 1, 2, 3, 4, \dots, N-1$  points. The DFT is defined as :

$$X(e^{j2\pi f \Delta}) = \sum_{n=0}^{N-1} x(n \Delta) e^{-j2\pi f n \Delta}$$

- Note that this is still continuous in frequency

# Discrete Fourier transform

Now let us evaluate this at frequencies:  $f = k/N \Delta$

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

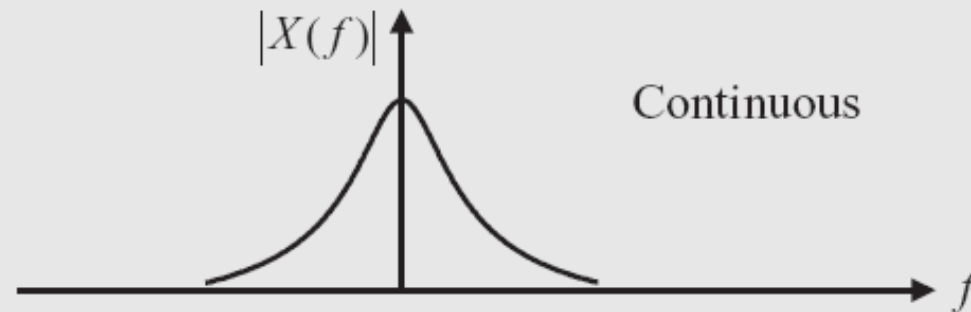
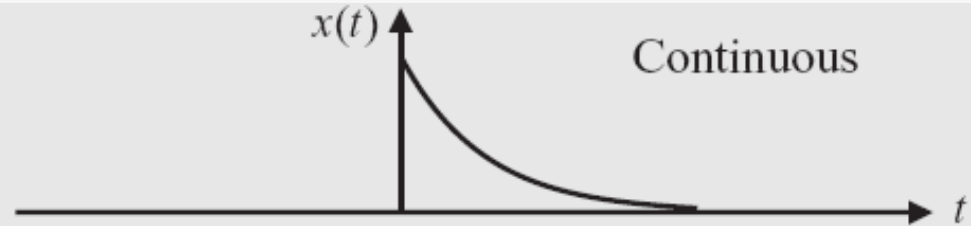
$$X(k) = \left[ X(e^{j2\pi f \Delta}) \text{ evaluated at } f = \frac{k}{N\Delta} \text{ Hz} \right] (k \text{ integer})$$

# Fourier Integral vs DFT

$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi ft} df$$

Fourier Integral

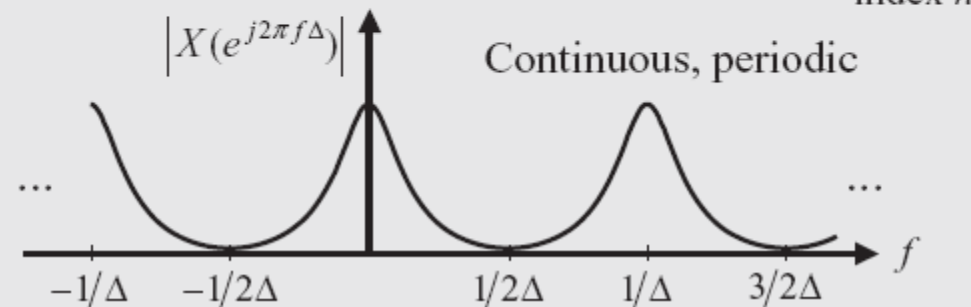
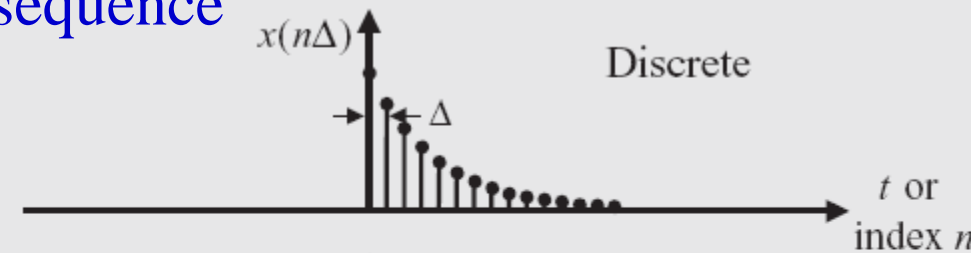
$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi ft} dt$$



Fourier transform of the sampled sequence

$$x(n\Delta) = \Delta \int_{-1/2\Delta}^{1/2\Delta} X(e^{j2\pi f\Delta}) e^{j2\pi fn\Delta} df$$

$$X(e^{j2\pi f\Delta}) = \sum_{n=-\infty}^{\infty} x(n\Delta) e^{-j2\pi fn\Delta}$$

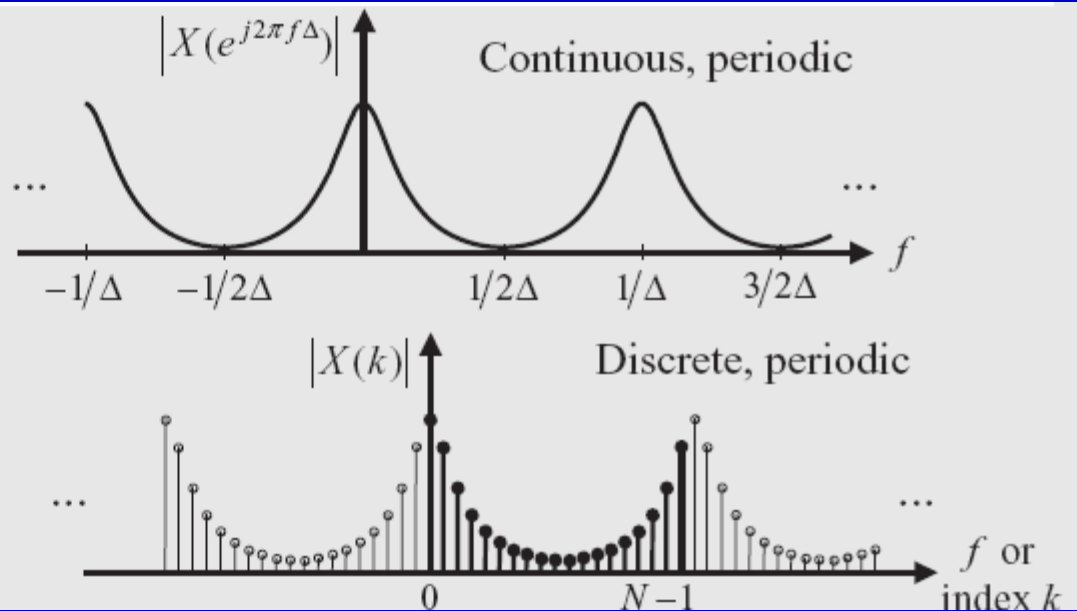




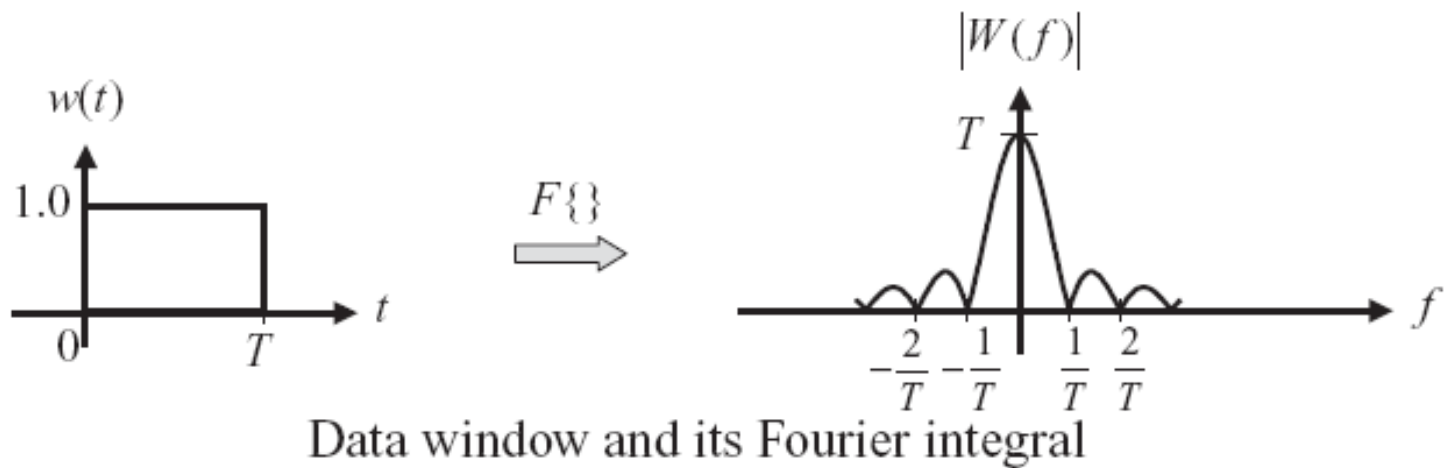
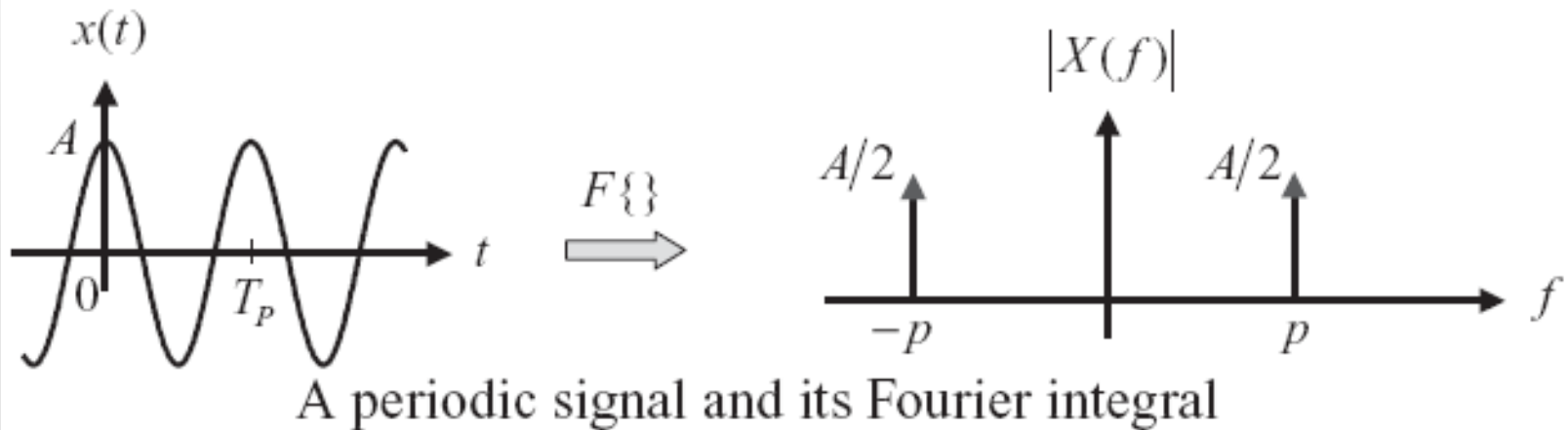
# Fourier Integral vs DFT

$$X(e^{j2\pi f\Delta}) = \sum_{n=-\infty}^{\infty} x(n\Delta)e^{-j2\pi fn\Delta}$$

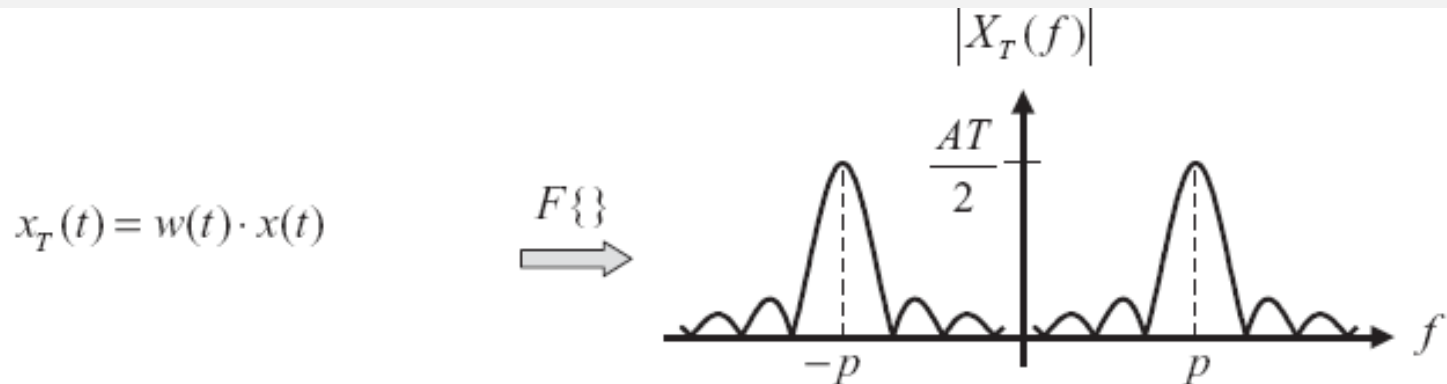
$$X(k) = \sum_{n=0}^{N-1} x(n)e^{-j(2\pi/N)nk}$$



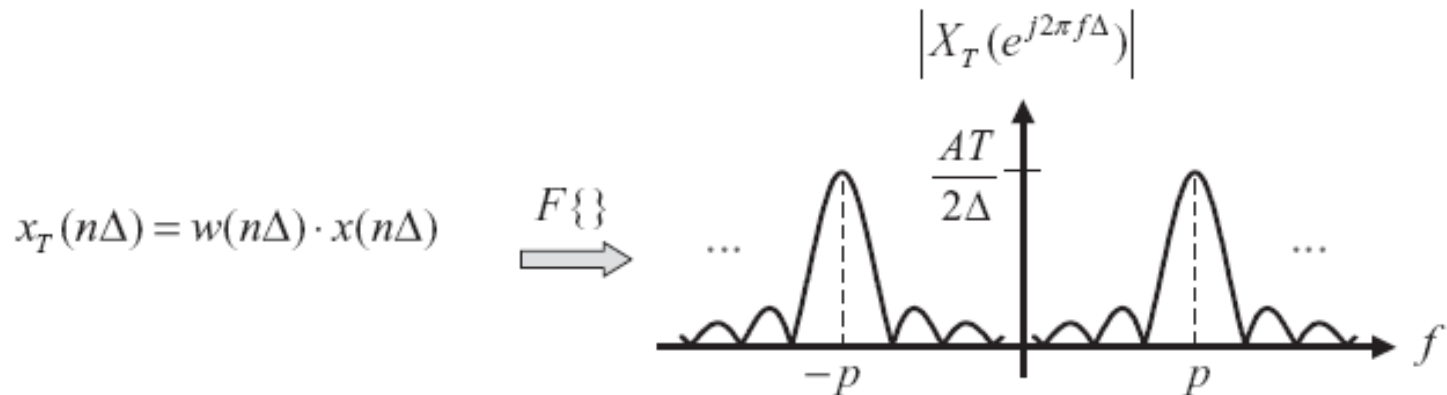
# Fourier Integral vs DFT



# Fourier Integral vs DFT



(c) Truncated signal and its Fourier integral



(d) Truncated and sampled signal and its Fourier transform of a sequence

# FFT algorithm: glimpses

- The DFT provides uniformly spaced samples of the Discrete-Time Fourier Transform (DTFT)

- DFT definition:

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

- Requires  $N^2$  complex multiplications &  $N(N-1)$  complex additions

# FFT algorithm: glimpses

- Take advantage of the symmetry and periodicity of the complex exponential (let  $W_N = e^{-j2\pi / N}$ )
  - symmetry:  $W_N^{k[N-n]} = W_N^{-kn} = (W_N^{kn})^*$
  - periodicity:  $W_N^{kn} = W_N^{k[n+N]} = W_N^{[k+N]n}$
- Note that two **N/2** DFTs take less computation than one length **N** DFT:  $2(N/2)^2 < N^2$