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ShanghaiTech University

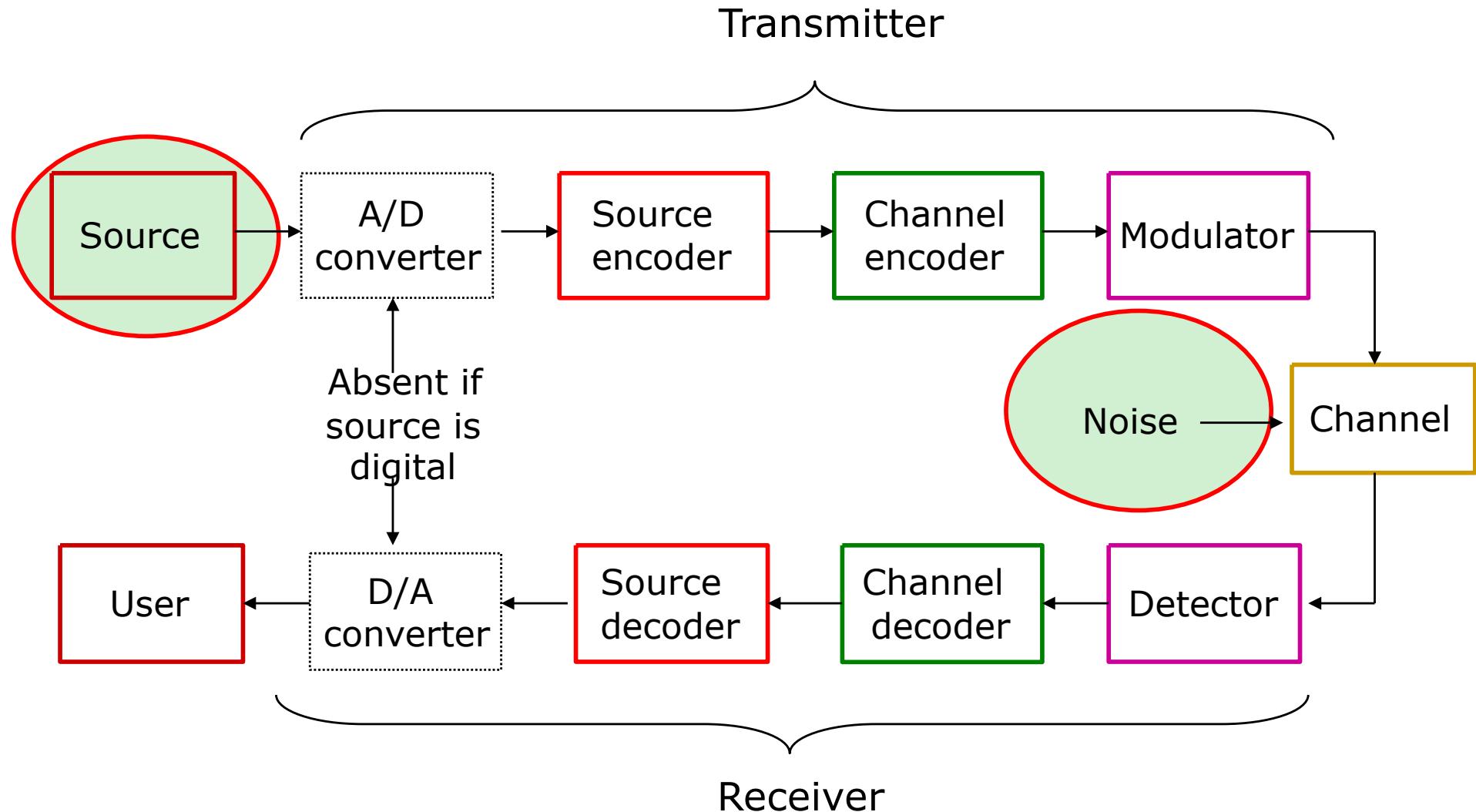
# EE140 Introduction to Communication Systems

## Lecture 2

Instructor: Prof. Lixiang LIAN

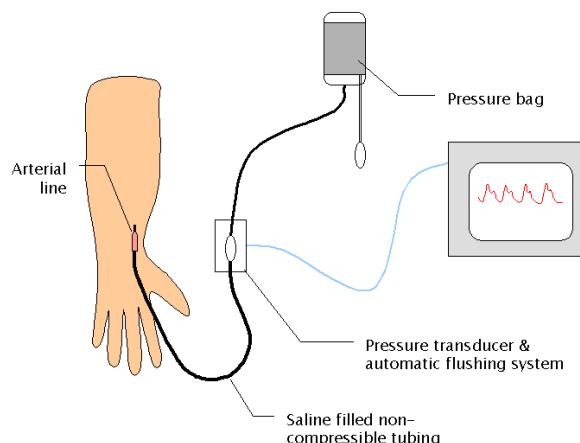
ShanghaiTech University, Fall 2025

# Architecture of a Digital Communication System



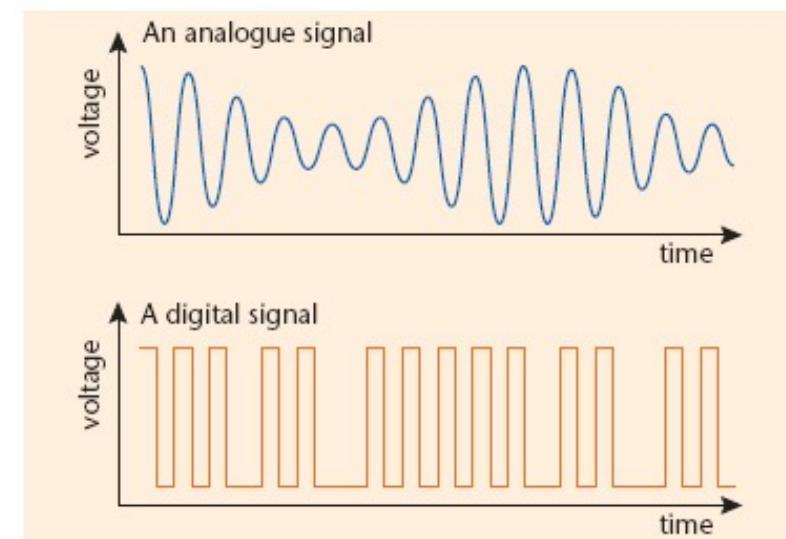
# Source Information

- *Message:* generated by source
- *Information:* the unpredictable part in a message
- *Signal:* a function that conveys information about the behavior or attributes of some phenomenon



**Transducer:**  
convert sensing  
signal to electric  
signal

Analog signal vs.  
digital signal



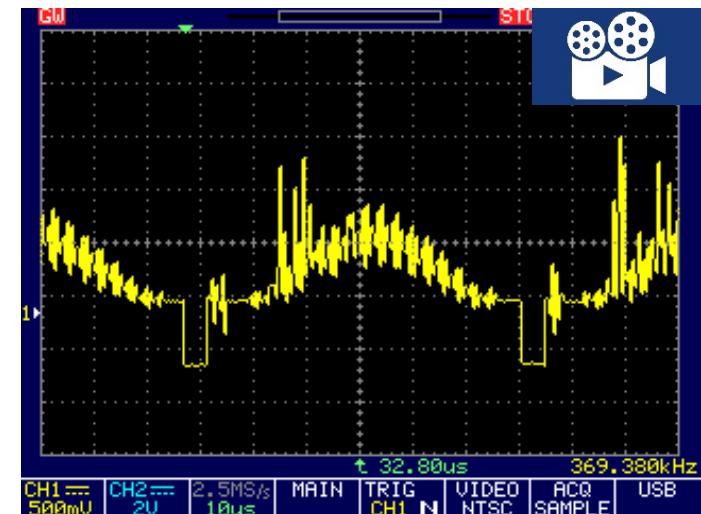
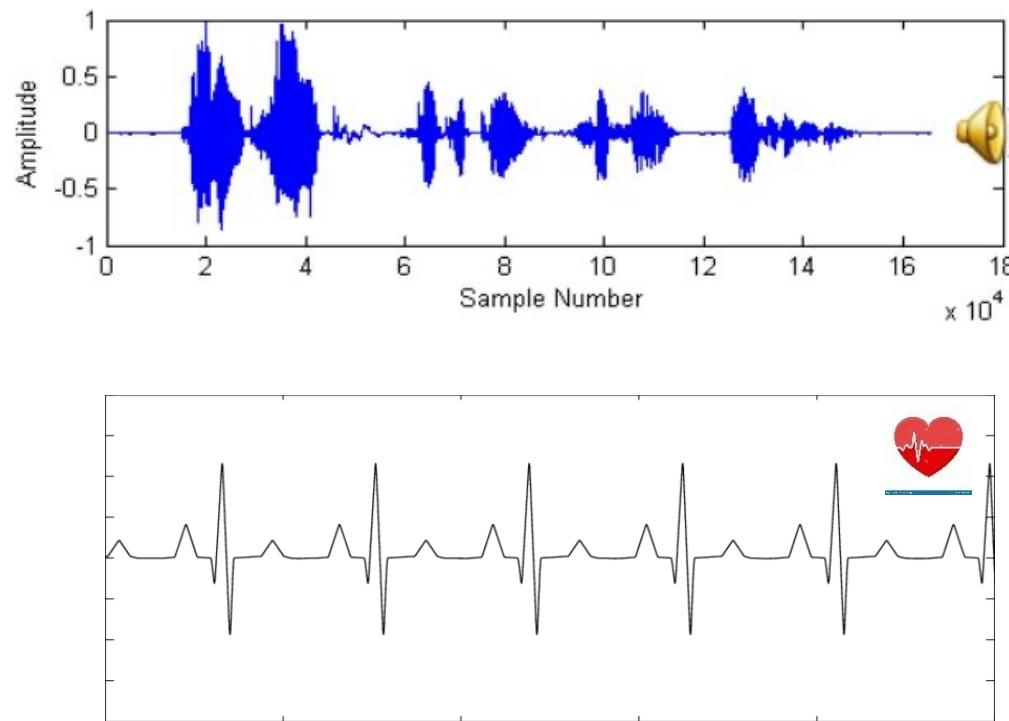
**Fig. 12.4** How analogue and digital signals change with time.

# Contents

- Deterministic signals
  - Classification of signals
  - Review of Fourier Transform
  - Properties of the Fourier Transform
  - Fourier Transform of Periodic Signals
  - Sampling Theory
  - The Hilbert Transform

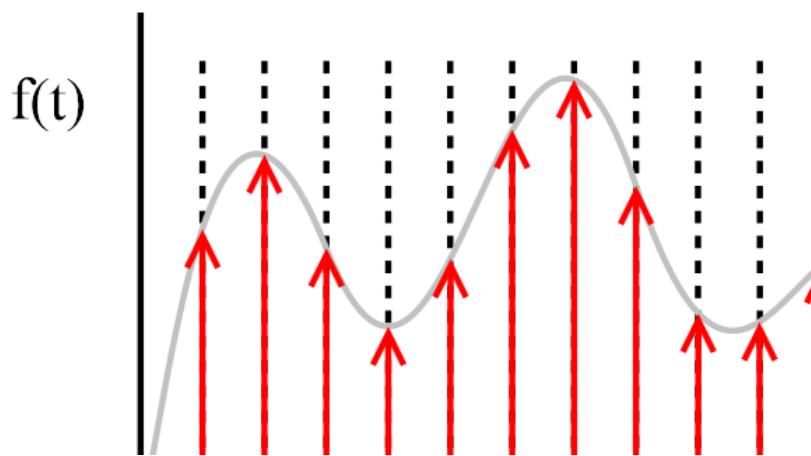
# What is Signal?

- In communication systems, a signal is any function that carries information. Also called information bearing signal.

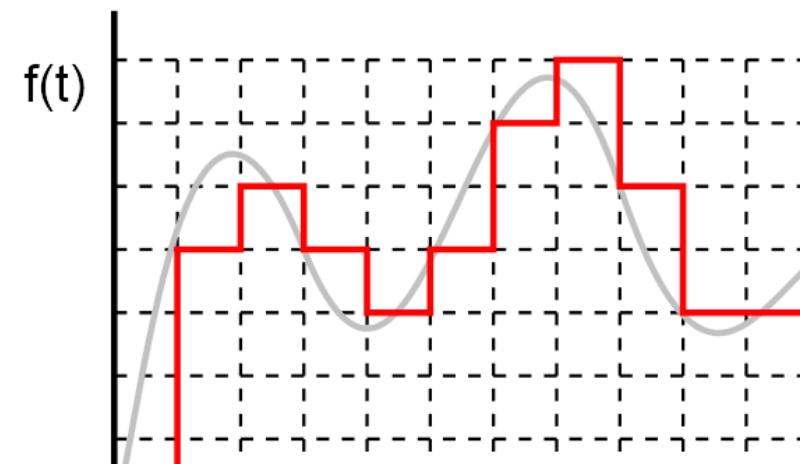


# Classification of Signals

- Analog, discrete-time and digital signals
  - Analog signal: both time and value are continuous
  - Discrete-time signal: discrete time and continuous value
  - Digital signal: both time and value are discrete

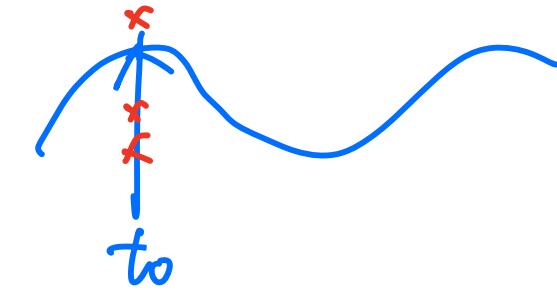


Discrete-time signal



Digital signal

# Classification of Signals



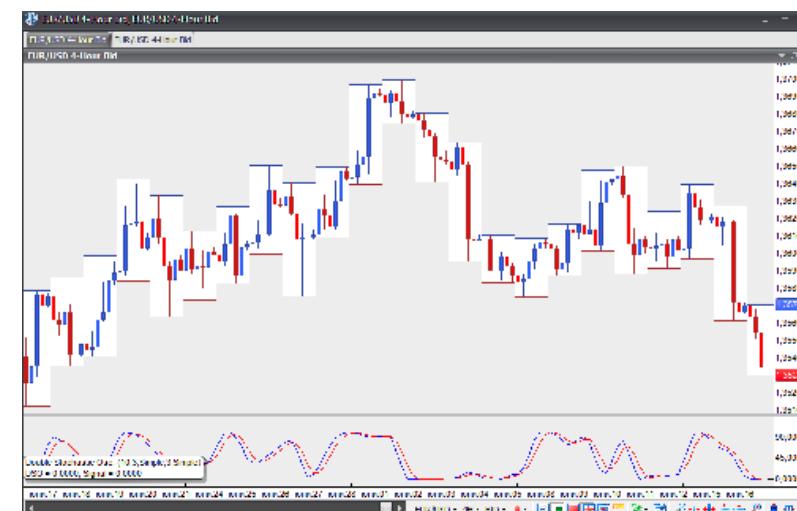
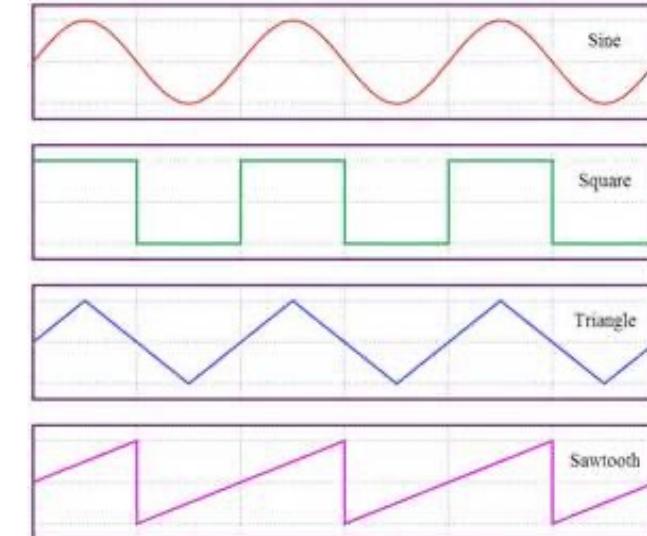
- Periodic and non-periodic signals

$$x(t + T_0) = x(t), \quad -\infty < t < \infty$$

- Random and deterministic

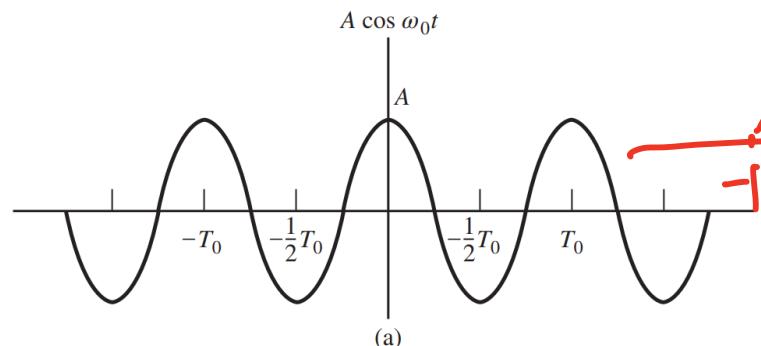
- Deterministic signal: no uncertainty in value. It can be modeled or expressed by an explicit mathematical function of time.

- Random (stochastic) signal: its value is uncertain or unpredictable. Probability distribution MUST be used to model it.

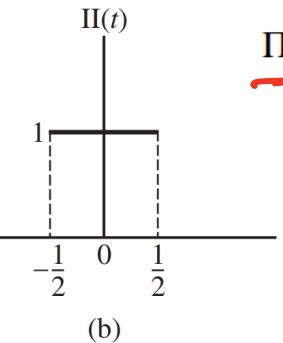


# Classification of Signals

$$x(t) = A \cos(\omega_0 t), \quad -\infty < t < \infty$$



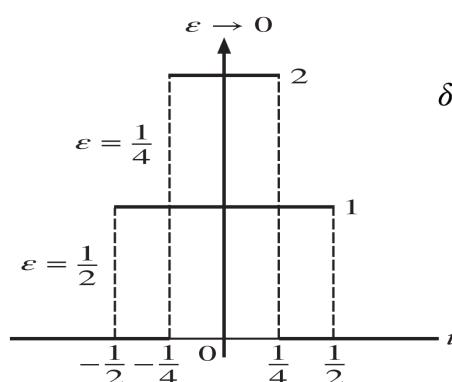
$$\Delta(t) \quad A\Pi\left(\frac{t-t_0}{\tau}\right)$$



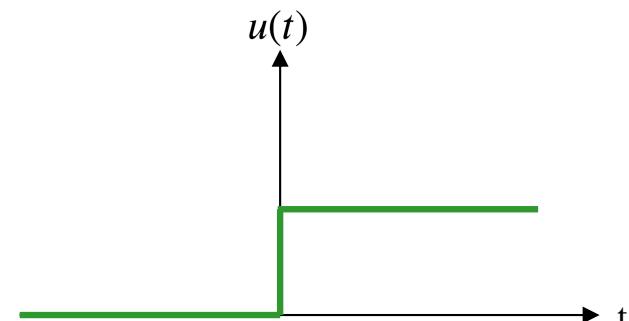
$$\Pi(t) = \begin{cases} 1, & |t| \leq \frac{1}{2} \\ 0, & \text{otherwise} \end{cases}$$

Deterministic (sinusoidal) signal

Unit rectangular pulse signal



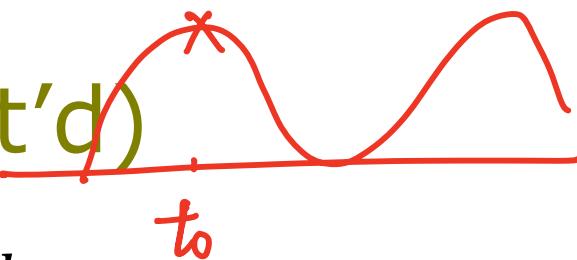
$$\delta_\epsilon(t) = \frac{1}{2\epsilon} \Pi\left(\frac{t}{2\epsilon}\right) = \begin{cases} \frac{1}{2\epsilon}, & |t| < \epsilon \\ 0, & \text{otherwise} \end{cases}$$



Unit impulse function (delta function)

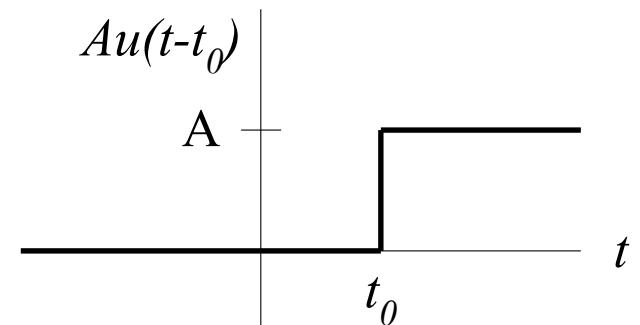
Unit step function

## Dirac Delta Function (cont'd)



- Sifting property:  $x(t_0) = \int_{-\infty}^{\infty} x(t)\delta(t - t_0)dt$   
because  $\int_{-\infty}^{\infty} x(t)\delta(t - t_0)dt = x(t_0) \int_{-\infty}^{\infty} \delta(t - t_0)dt = x(t_0)$ 
  - The impulse function selects a particular value of the function  $x(t)$  in the integration process
- Unit step function

$$u(t - t_0) = \begin{cases} 1 & t > t_0 \\ 0 & t < t_0 \end{cases}$$



- Relationship between  $\delta(t)$  and  $u(t)$

$$\delta(t - t_0) = \frac{d}{dt} u(t - t_0) \quad \leftrightarrow \quad u(t - t_0) = \int_{-\infty}^t \delta(\tau - t_0) d\tau$$

# Classification of Signals

- Energy and power signals

- Total Energy of a signal:  $E = \int_{-\infty}^{\infty} |x(t)|^2 dt$  joules

- Signal  $x(t)$  is an **energy signal** if  $\int_{-\infty}^{\infty} |x(t)|^2 dt < \infty$

- Average Power of a signal:  $P = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} |x(t)|^2 dt$  watts

- Signal  $x(t)$  is a **power signal**  $0 < \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} |x(t)|^2 dt < \infty$

- Remark:

- Energy signal:  $P=0$ ; Power signal:  $E=\infty$ .

- Discrete Signal:  $E = \sum_{n=-\infty}^{\infty} |x[n]|^2$ ;  $P = \lim_{N \rightarrow \infty} \frac{1}{2N+1} \sum_{n=-N}^N |x[n]|^2$

# Exercise: Question

Classify the following signals as energy signals or power signals. Find the normalized energy or normalized power of each.

(1)

$$x(t) = \begin{cases} A \cos(2\pi f_0 t) & \text{for } -T_0/2 \leq t \leq T_0/2, \text{ where } T_0 = 1/f_0 \\ 0 & \text{elsewhere} \end{cases}$$

(2)

$$x(t) = \begin{cases} A \exp(-at) & \text{for } t > 0, a > 0 \\ 0 & \text{elsewhere} \end{cases}$$

# Exercise: Solution

Classify the following signals as energy signals or power signals. Find the normalized energy or normalized power of each.

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## Solution 1

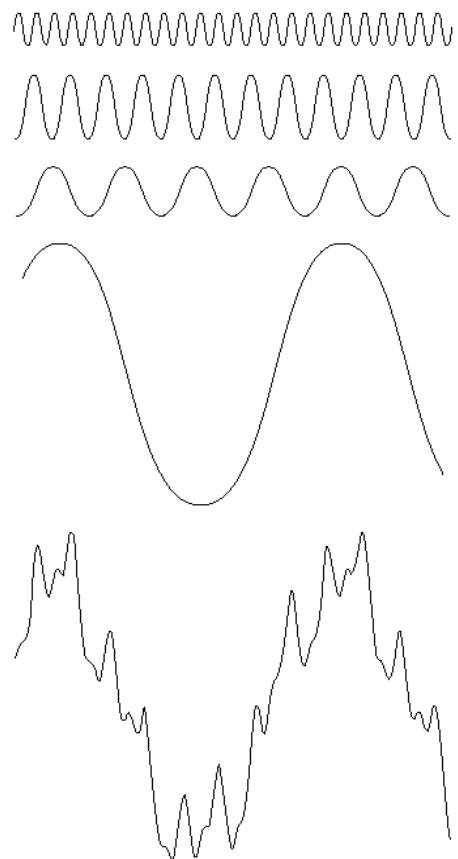
(1) Energy signal.  $E_x = \int_{-\infty}^{\infty} x^2(t) dt = \int_{-T_0/2}^{T_0/2} A^2 \cos^2(2\pi f_0 t) dt = \frac{A^2 T_0}{2}.$

(2) Energy signal.  $E_x = \int_{-\infty}^{\infty} x^2(t) dt = \int_0^{\infty} A^2 \exp(-2at) dt = \frac{A^2 \exp(-2at)}{-2a} \Big|_0^{\infty} = \frac{A^2}{2a}.$

# Contents

- Deterministic signals
  - Classification of signals
  - Review of Fourier Transform
    - Fourier Transform
    - Discrete-time Fourier Transform (skip)
    - Discrete Fourier Series (skip)
    - Discrete Fourier Transform (skip)
    - Fast Fourier Transform (skip)
  - Properties of the Fourier Transform
  - Fourier Transform of Periodic Signals
  - Sampling Theory
  - The Hilbert Transform

# Fourier Series



**FIGURE 4.1** The function at the bottom is the sum of the four functions above it. Fourier's idea in 1807 that periodic functions could be represented as a weighted sum of sines and cosines was met with skepticism.

A periodic function = superposition or linear combination of simple sine and cosine functions



- Jean-Baptiste Joseph Fourier: 1768-1830
- Student of Laplace and Lagrange
- 1807: introduced the Fourier series expansion

# Fourier Transform

*w*  
 $f: \mathbb{H} \ni$

- Fourier Transform of a **continuous-time** signal

$$X(f) = \mathfrak{F}[x(t)] \longleftrightarrow X(f) = \int_{-\infty}^{\infty} x(\lambda) e^{-j2\pi f \lambda} d\lambda$$

if  $x(t)$  is absolutely integrable  $\int_{-\infty}^{\infty} |x(t)| dt < \infty$

- Inverse Fourier Transform

$$x(t) = \mathfrak{F}^{-1}[X(f)] \longleftrightarrow x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi f t} df$$

- Spectrum

- The Fourier transform of a continuous-time signal is a complex signal

$$\underline{X(f)} = |\underline{X(f)}| e^{j \angle \underline{X(f)}}$$

$x(t)$  real signal  
real + even  
real + odd

# Parseval's Theorem

- Parseval's Theorem

$$\int_{-\infty}^{\infty} \underline{x_1(t)x_2^*(t)} dt = \int_{-\infty}^{\infty} \underline{X_1(f)X_2^*(f)} df$$

- When  $x_1(t) = x_2(t)$ , we have

$$E = \int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} |X(f)|^2 df$$

Energy

Energy Spectral Density

- Energy Spectral Density

$$G(f) = |X(f)|^2 \text{ Joules/Hz}$$

$$\begin{aligned}
 \underbrace{\int x_1(t) x_2^*(t) dt} &= \int x_1(t) \int x_2^*(f) e^{-j2\pi f t} \sqrt{f} dt \\
 &= \int x_2^*(f) \boxed{\int x_1(t) e^{-j2\pi f t} dt} \sqrt{f} df \\
 &= \int x_2^*(f) X_1(f) \sqrt{f} df
 \end{aligned}$$

if  $x_2(f) = x_1(f)$

$$E = \boxed{\int |x_1(t)|^2 dt} = \boxed{\int |X_1(f)|^2 df}$$

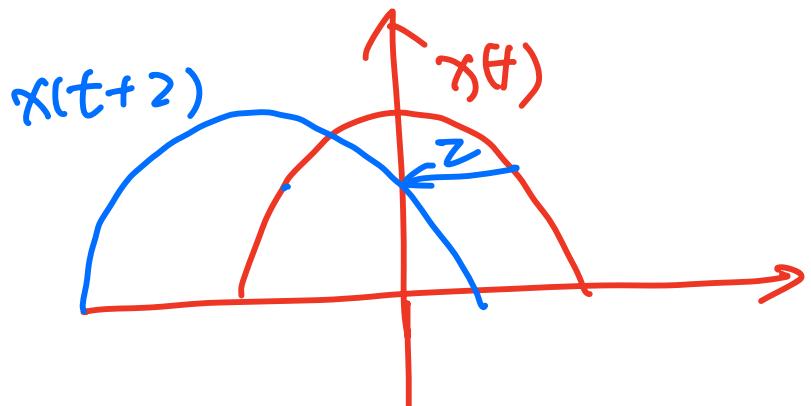
$G(f)$



# Power Spectral Density and Correlation

	<b>Energy Signal</b>	<b>Power Signal</b>
Spectral density	<p>Energy Spectral Density</p> $E = \int_{-\infty}^{\infty}  x(t) ^2 dt = \int_{-\infty}^{\infty}  X(f) ^2 df$ <p><math>G(f) =  X(f) ^2</math></p>	$P = \int_{-\infty}^{\infty} S(f) df = \langle x^2(t) \rangle$
Time-average autocorrelation function	$\phi(\tau) = x(\tau) * x(-\tau) = \int_{-\infty}^{\infty} x(\lambda)x(\lambda + \tau) d\lambda$ $= \lim_{T \rightarrow \infty} \int_{-T}^{T} x(\lambda)x(\lambda + \tau) d\lambda$ (energy signal)	$R(\tau) = \langle x(t)x(t + \tau) \rangle$ $\triangleq \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^{T} x(t)x(t + \tau) dt$
Relationship	$\underline{\phi(t)} = \mathfrak{F}^{-1}(G(f))$ $\underline{G(f)} = \mathfrak{F}(\phi(t))$	$R(\tau) = \mathfrak{F}^{-1}[S(f)] = \int_{-\infty}^{\infty} S(f)e^{j2\pi f\tau} df$ $S(f) = \mathfrak{F}[R(\tau)] = \int_{-\infty}^{\infty} R(\tau) e^{-j2\pi f\tau} d\tau$

$$\phi(z) = x(z) * x(-z) = \int_{-\infty}^{+\infty} \underbrace{x(t)}_{z} x(t+z) dt$$



$$z=0 \quad \phi(0) = E$$

$$z=\infty \quad \phi(\infty) = 0$$

$\phi(z)$  even

$$\phi(-z) = \int x(t) x(t-z) dt$$

$$\downarrow t' = t-z$$

$$= \int x(t'+z) x(t') dt' = \phi(z)$$

$$\mathcal{F}[\phi(z)] = \mathcal{F}[x(z) * x(-z)] = \underline{X(f) X^*(f)}$$

$$= G(f)$$

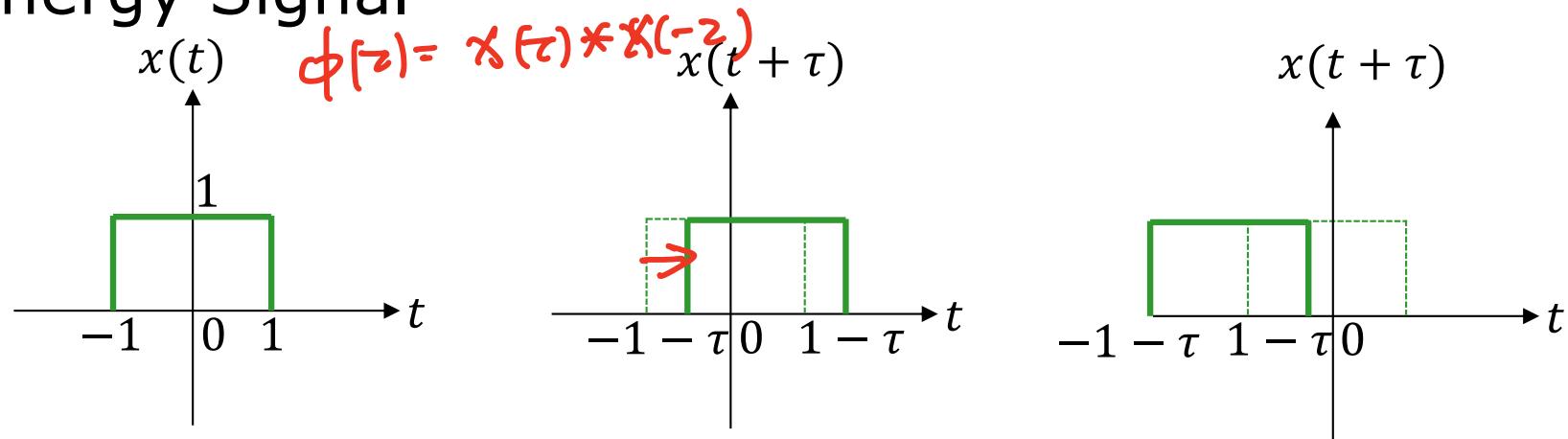
$$\mathcal{F}[x(-t)] = X^*(f)$$

$x(t)$  is real signal



# Example: Autocorrelation Function

- Energy Signal

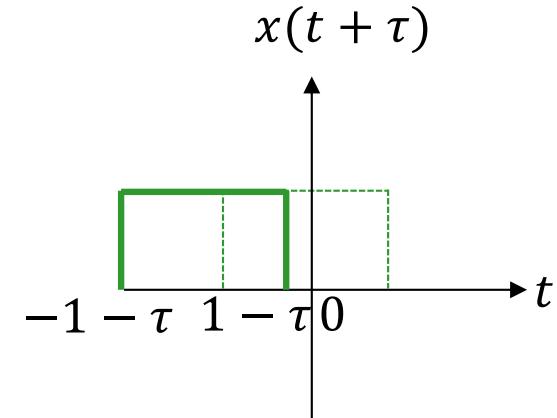


$$\underline{\underline{\tau = 0}}, \phi(0) = \int_{-1}^1 x(t)x(t)dt = 2$$

$$\underline{\underline{-2 < \tau < 0}}, \phi(\tau) = \int_{-1-\tau}^1 x(t)x(t+\tau)dt = 2 + \tau$$

$$0 < \tau < 2, \phi(\tau) = \int_{-1}^{1-\tau} x(t)x(t+\tau)dt = 2 - \tau$$

$$|\tau| \geq 2, \phi(\tau) = 0$$



1.  $\phi(\tau)$  is even.
2.  $\phi(0) \geq |\phi(\tau)|$ .

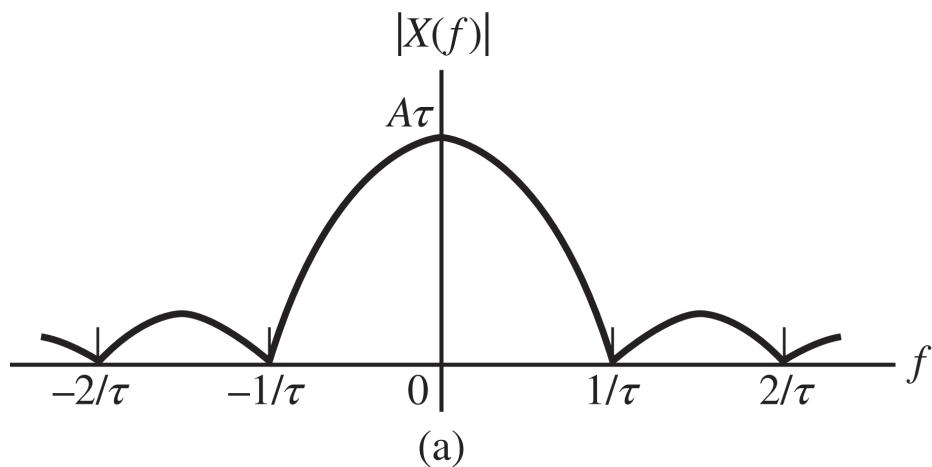
# Example: Rectangular Pulse

$$x(t) = A \Pi \left( \frac{t - t_0}{\tau} \right) \rightarrow X(f) = \int_{-\infty}^{\infty} A \Pi \left( \frac{t - t_0}{\tau} \right) e^{j2\pi f t} dt$$

$$= A \int_{t_0 - \tau/2}^{t_0 + \tau/2} e^{-j2\pi f t} dt = A\tau \operatorname{sinc}(f\tau) e^{-j2\pi f t_0}$$

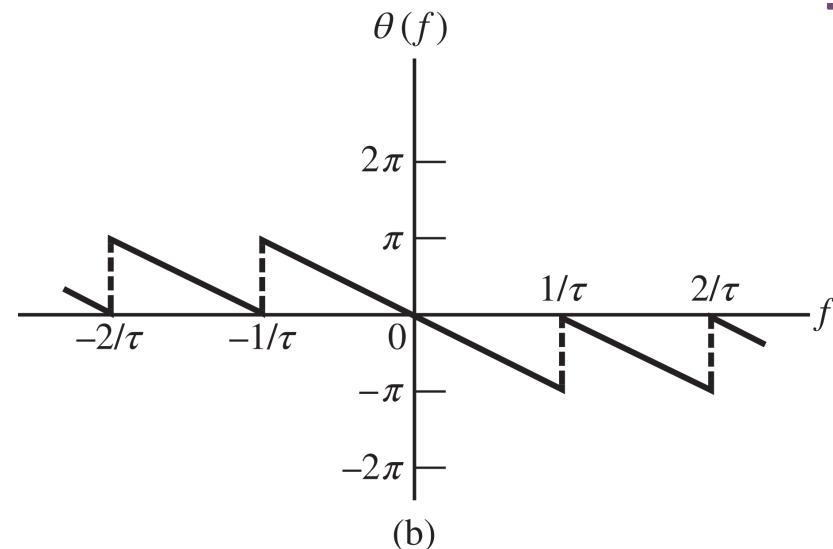
## Amplitude spectrum

$$|X(f)| = A\tau |\operatorname{sinc}(f\tau)|$$



## Phase spectrum

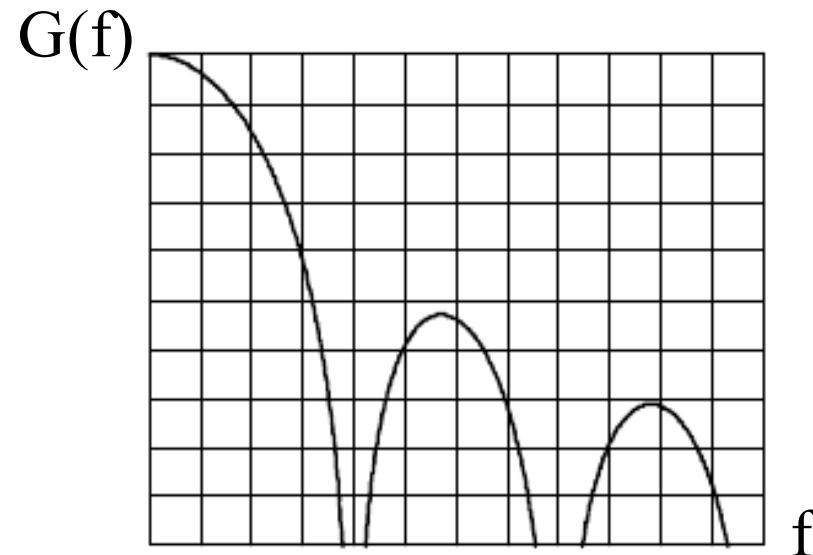
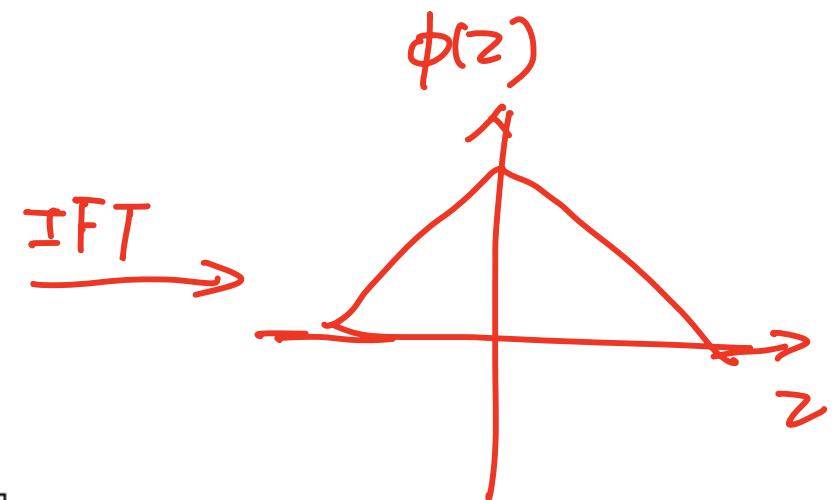
$$\theta(f) = \begin{cases} -2\pi t_0 f & \text{if } \operatorname{sinc}(f\tau) > 0 \\ -2\pi t_0 f \pm \pi & \text{if } \operatorname{sinc}(f\tau) < 0 \end{cases}$$



# Example: Rectangular Pulse

- Energy spectral density

$$G(f) = |X(f)|^2 = A^2 \tau^2 \text{sinc}^2(f\tau)$$

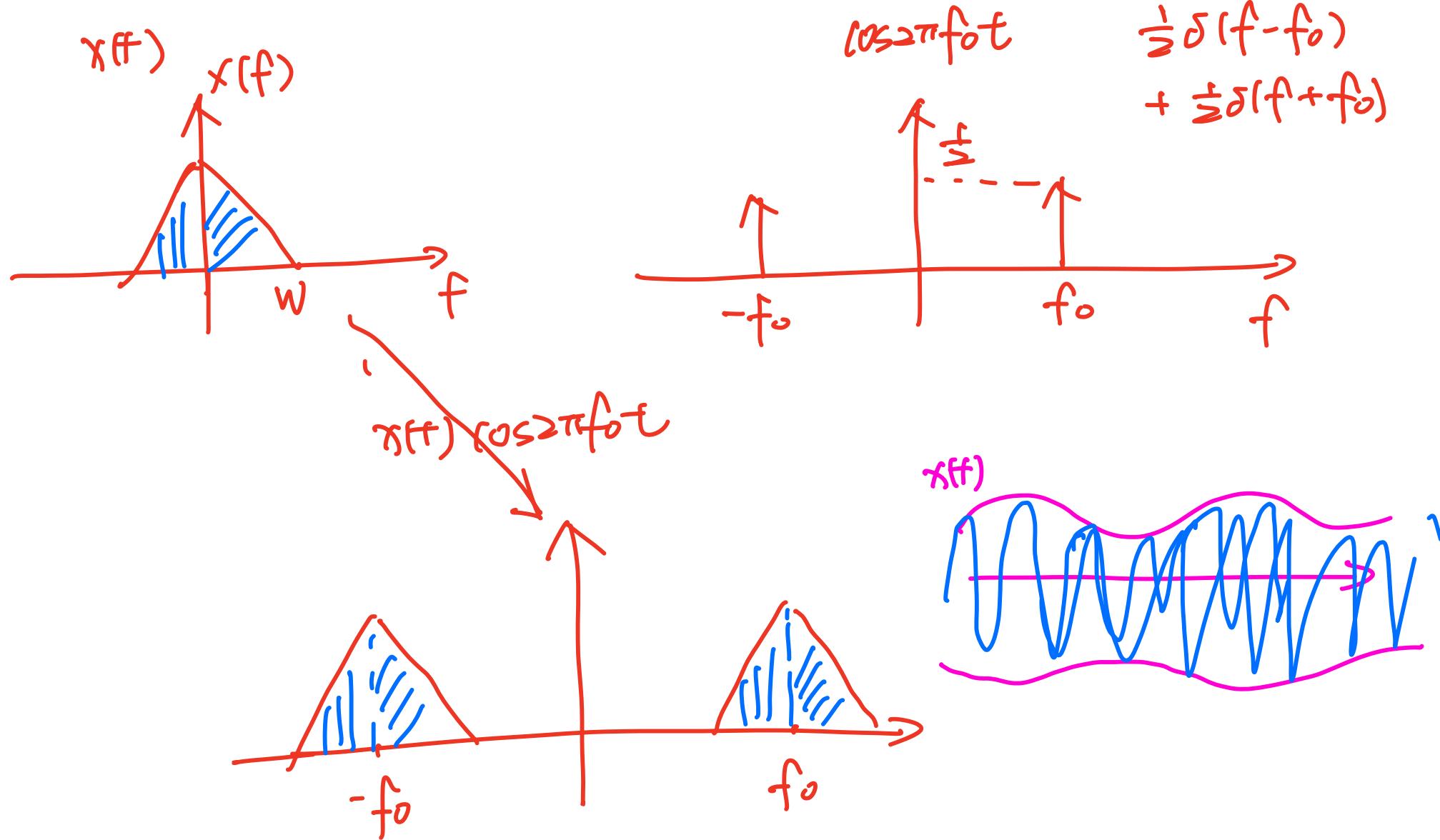


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# Properties of Fourier Transform

Operation	$x(t)$	$X(f)$
1. Superposition	$a_1x_1(t) + a_2x_2(t)$	$a_1X_1(f) + a_2X_2(f)$
2. Scaling	$x(at)$	$\frac{1}{ a }X(\frac{f}{a})$
3. Time shifting	$x(t - t_0)$	$X(f) \exp(-j2\pi f t_0)$
4. Frequency shifting	$x(t) \exp(j2\pi f_0 t)$	$X(f - f_0)$
5. Duality theorem	$X(t)$	$x(-f)$
6. Modulation Theorem <u>—</u>	<u><math>x(t)\cos(2\pi f_0 t)</math></u>	$\frac{1}{2}X(f - f_0) + \frac{1}{2}X(f + f_0)$
7. Time differentiation	$\frac{d^n x}{dt^n}$	$(j2\pi f)^n X(f)$
8. Frequency differentiation	$(-jt)^n x(t)$	$\frac{d^n X}{df^n}$
9. Time integration	$\int_{-\infty}^t x(\tau)d\tau$	$\frac{1}{j2\pi f}X(f) + \frac{1}{2}X(0)\delta(f)$
10. Time convolution	$x_1(t) * x_2(t)$	$X_1(f)X_2(f)$
11. Multiplication	$x_1(t)x_2(t)$	$X_1(f) * X_2(f)$



# Example: Dirac Delta Function

- Dirac delta function

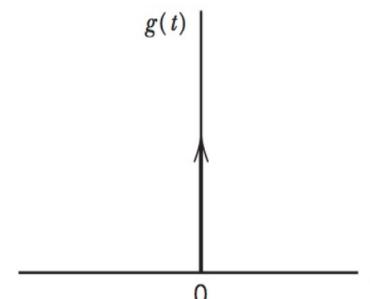
$$\delta(t) = \begin{cases} \infty & t = 0, \\ 0 & t \neq 0, \end{cases}$$

and

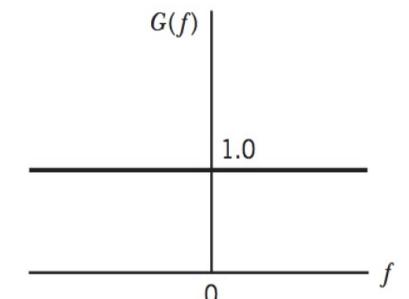
$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$

- Fourier transform

$$\begin{aligned} G(f) &= \mathcal{F}[\delta(t)] = \int_{-\infty}^{\infty} \delta(t) e^{-j2\pi ft} dt \\ &= \int_{-\infty}^{\infty} \delta(t) e^{-j2\pi f_0 t} dt = e^0 = 1 \end{aligned}$$



(a)



(b)

# Example: Dirac Delta Function

- Application of the delta function

1.  $A\delta(t) \longleftrightarrow A$
2.  $\underline{A\delta(t - t_0)} \longleftrightarrow Ae^{-j2\pi f t_0}$
3.  $\underline{A} \longleftrightarrow \underline{A\delta(f)}$
4.  $\underline{Ae^{j2\pi f_0 t}} \longleftrightarrow \underline{A\delta(f - f_0)}$

$\cos(2\pi f_c t) \rightleftharpoons \frac{1}{2} [\delta(f - f_c) + \delta(f + f_c)]$

$\sin(2\pi f_c t) \rightleftharpoons \frac{1}{2j} [\delta(f - f_c) - \delta(f + f_c)]$

Definition of  
delta function

$$\int_{-\infty}^{\infty} \cos(2\pi f t) dt = \delta(f)$$

$$\int_{-\infty}^{\infty} \exp(-j2\pi f t) dt = \delta(f)$$

$$\begin{aligned} & \int \cos 2\pi ft - j \sin 2\pi ft dt \\ &= \delta(f) \end{aligned}$$

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# Line Spectra for Periodic Signal

- Periodic signal  $x(t)$  with period  $T_0$   $f_0 = \frac{1}{T_0}$

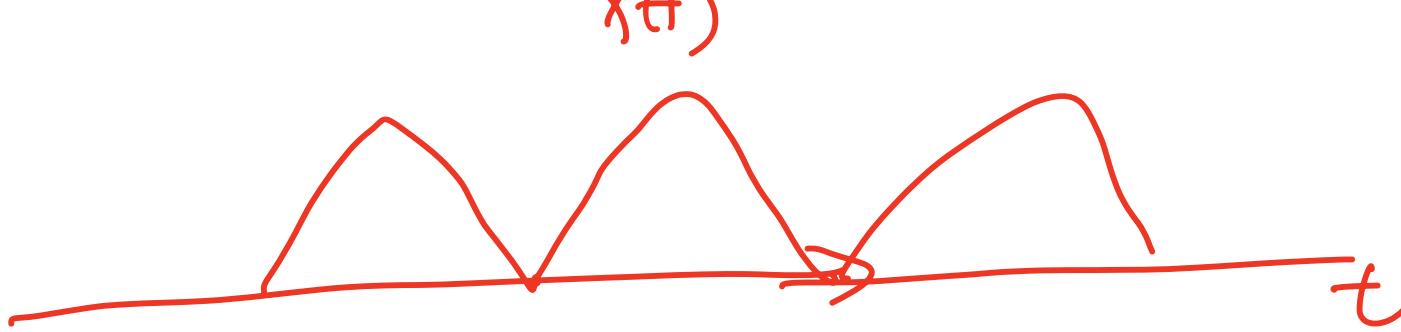
$$\begin{aligned}
 X(f) &= F[x(t)] = \int_{-\infty}^{\infty} x(t) e^{-j2\pi ft} dt \\
 &= \int_{-\infty}^{\infty} \left( \sum_{n=-\infty}^{\infty} X_n e^{j2\pi n f_0 t} \right) e^{-j2\pi ft} dt \\
 &= \sum_{n=-\infty}^{\infty} X_n \int_{-\infty}^{\infty} e^{-j2\pi(f-nf_0)t} dt \\
 &= \sum_{n=-\infty}^{\infty} X_n \delta(f - nf_0)
 \end{aligned}$$

Fourier series

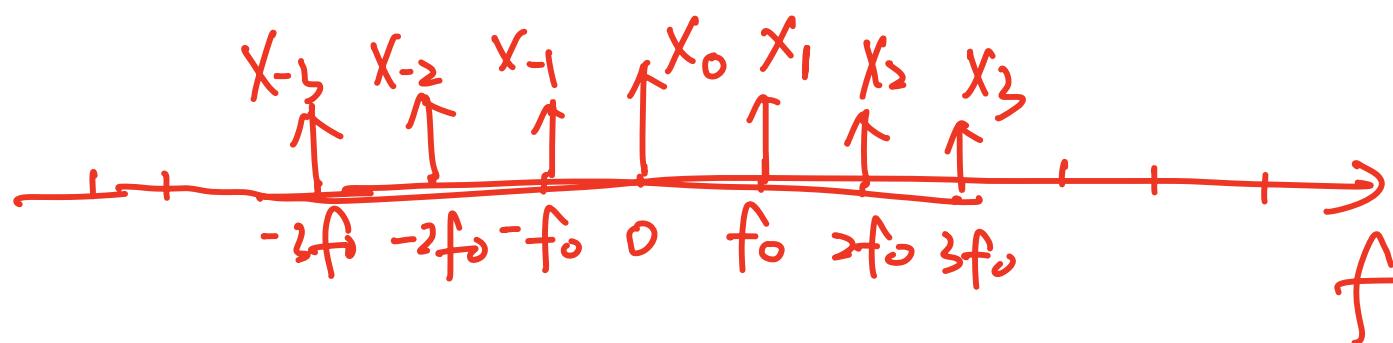
$$X_n = \frac{1}{T_0} \int_{t_0}^{t_0+T_0} x(t) e^{-jn\omega_0 t} dt$$

Continuous Periodic  
FT  
Discrete





↓ FT



# Example: The “Comb” Function

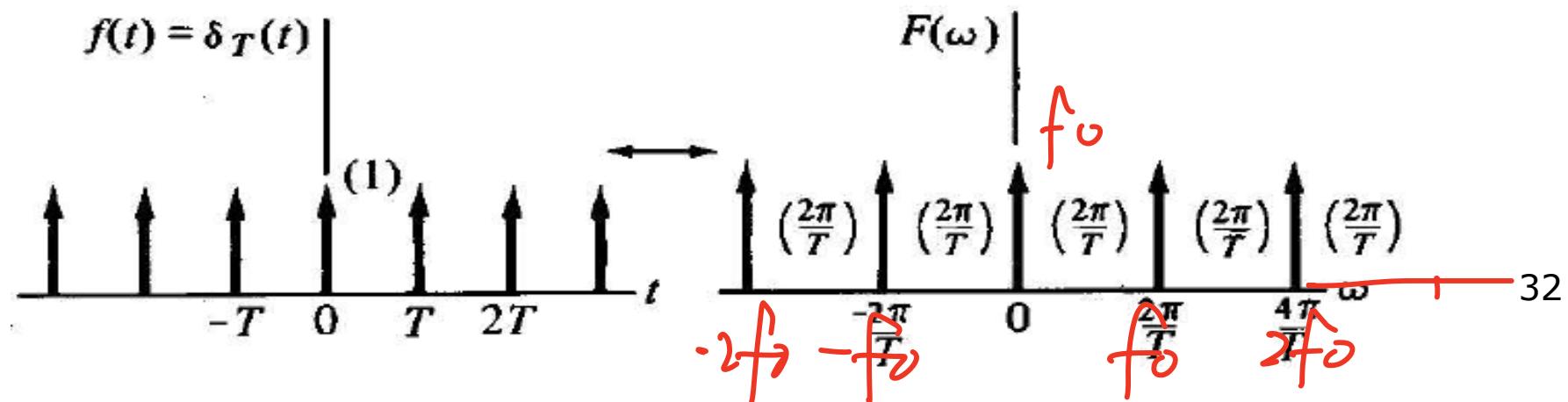
- The “comb” function:

$$\delta_T(t) = \sum_{m=-\infty}^{\infty} \delta(t - mT)$$

- FT 1:  $\delta_T(t) = \sum_{m=-\infty}^{\infty} \delta(t - mT) = \sum_{n=-\infty}^{\infty} X_n e^{jn2\pi f_0 t}, f_0 = \frac{1}{T}$

$$X_n = \frac{1}{T} \int_T \delta(t) e^{-jn2\pi f_0 t} dt = f_0$$

$$\delta_T(t) = \sum_{n=-\infty}^{\infty} f_0 e^{jn2\pi f_0 t} \quad \longleftrightarrow \quad \mathfrak{F}[\delta_T(t)] = \sum_{n=-\infty}^{\infty} f_0 \delta(f - nf_0)$$



# Example: The “Comb” Function

- The “comb” function:

$$\delta_T(t) = \sum_{m=-\infty}^{\infty} \delta(t - mT)$$

- FT 2:

$$\begin{aligned}\mathfrak{F} \left[ \sum_{m=-\infty}^{\infty} \delta(t - mT_s) \right] &= \int_{-\infty}^{\infty} \left[ \sum_{m=-\infty}^{\infty} \delta(t - mT_s) \right] e^{-j2\pi ft} dt \\ &= \sum_{m=-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(t - mT_s) e^{-j2\pi ft} dt \\ &= \sum_{m=-\infty}^{\infty} e^{-j2\pi mT_s f}\end{aligned}$$

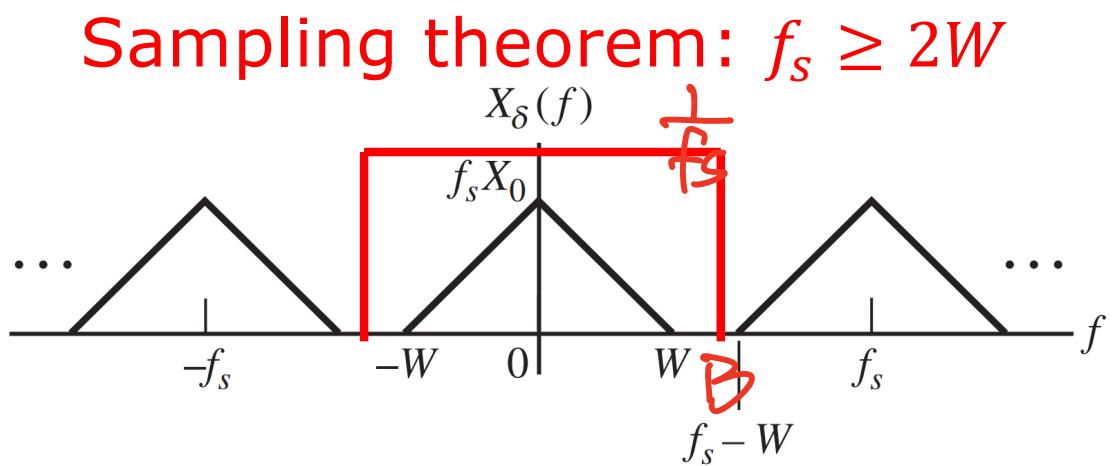
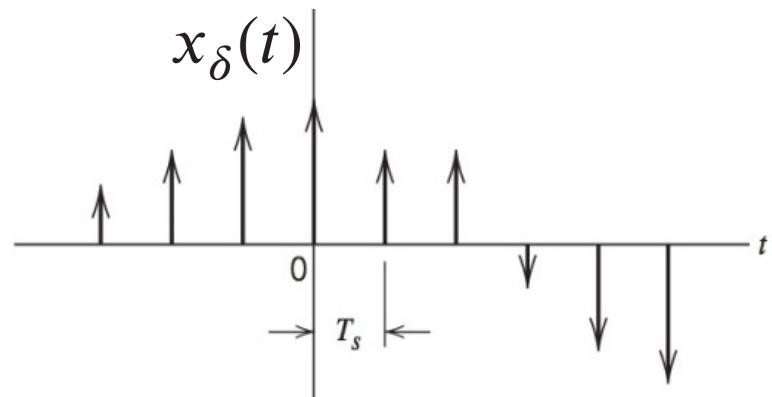
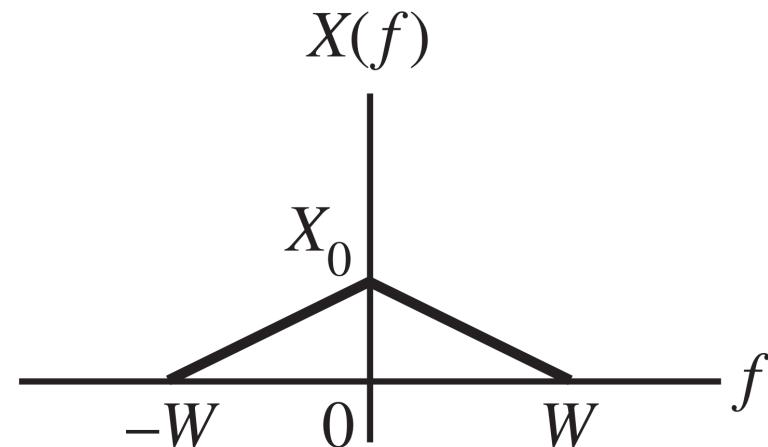
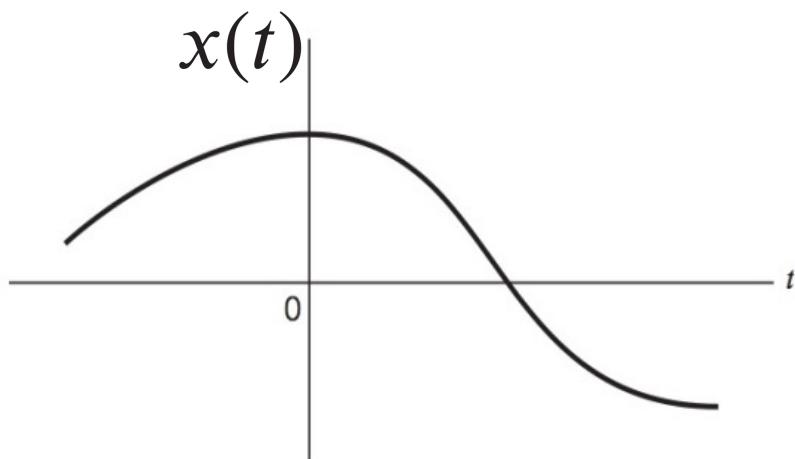
$$\sum_{m=-\infty}^{\infty} e^{j2\pi mTf} = f_0 \sum_{n=-\infty}^{\infty} \delta(f - nf_0)$$

Fourier  
Series

# Contents

- Deterministic signals
  - Classification of signals
  - Review of Fourier Transform
  - Properties of the Fourier Transform
  - Fourier Transform of Periodic Signals
  - Sampling Theory → Digital Com
  - The Hilbert Transform → Analog

# Sampling Theory



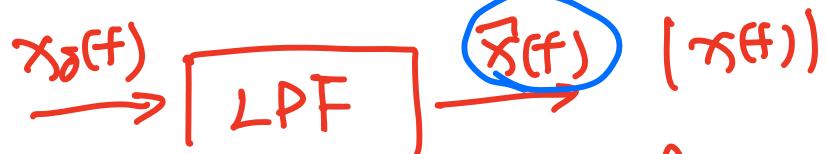
$$\underline{x_\delta(t)} = \sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s)$$

$$X_\delta(f) = f_s \sum_{n=-\infty}^{\infty} X(f - n f_s)$$

$$x_\delta(t) = x(t) \cdot \sum_{n=-\infty}^{+\infty} \delta(t - nT_s) \xrightarrow{F^-}$$

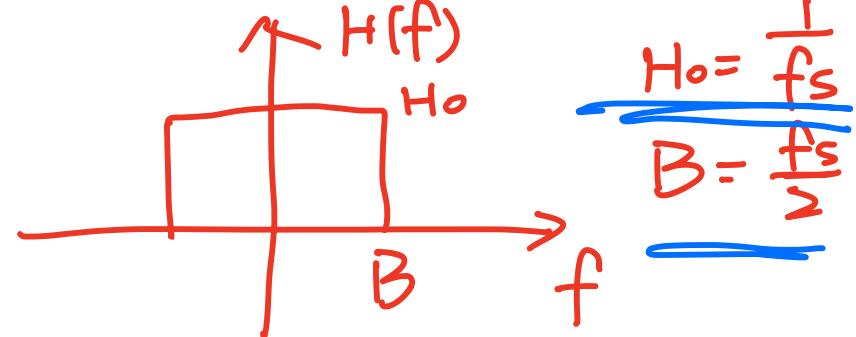
$$X_\delta(f) = X(f) * f_s \sum_{n=-\infty}^{+\infty} \delta(f - \underline{n f_s})$$

$$= \sum_{n=-\infty}^{+\infty} f_s X(f - n f_s)$$



$$H(f) = H_0 \pi\left(\frac{f}{2B}\right)$$

$$h_{ff} = 2H_0B \sin(2Bt)$$



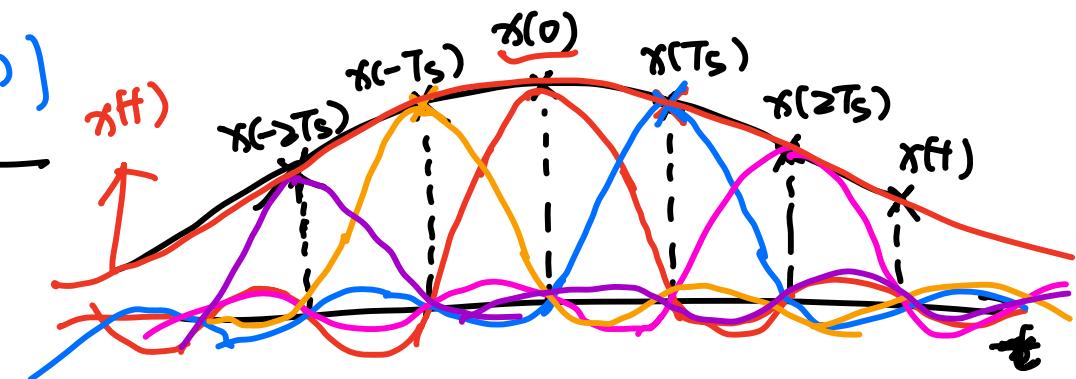
$$\bar{x}(f) = x_\delta(f) * h(f)$$

$$= \sum_{n=-\infty}^{\infty} x(nT_s) \underbrace{\delta(t-nT_s)}_{\sim} * \underbrace{2H_0B \sin(2Bt)}_{\sim}$$

$$= \sum_{n=-\infty}^{+\infty} 2H_0B x(nT_s) \underbrace{\sin(2B(t-nT_s))}_{\sim}$$

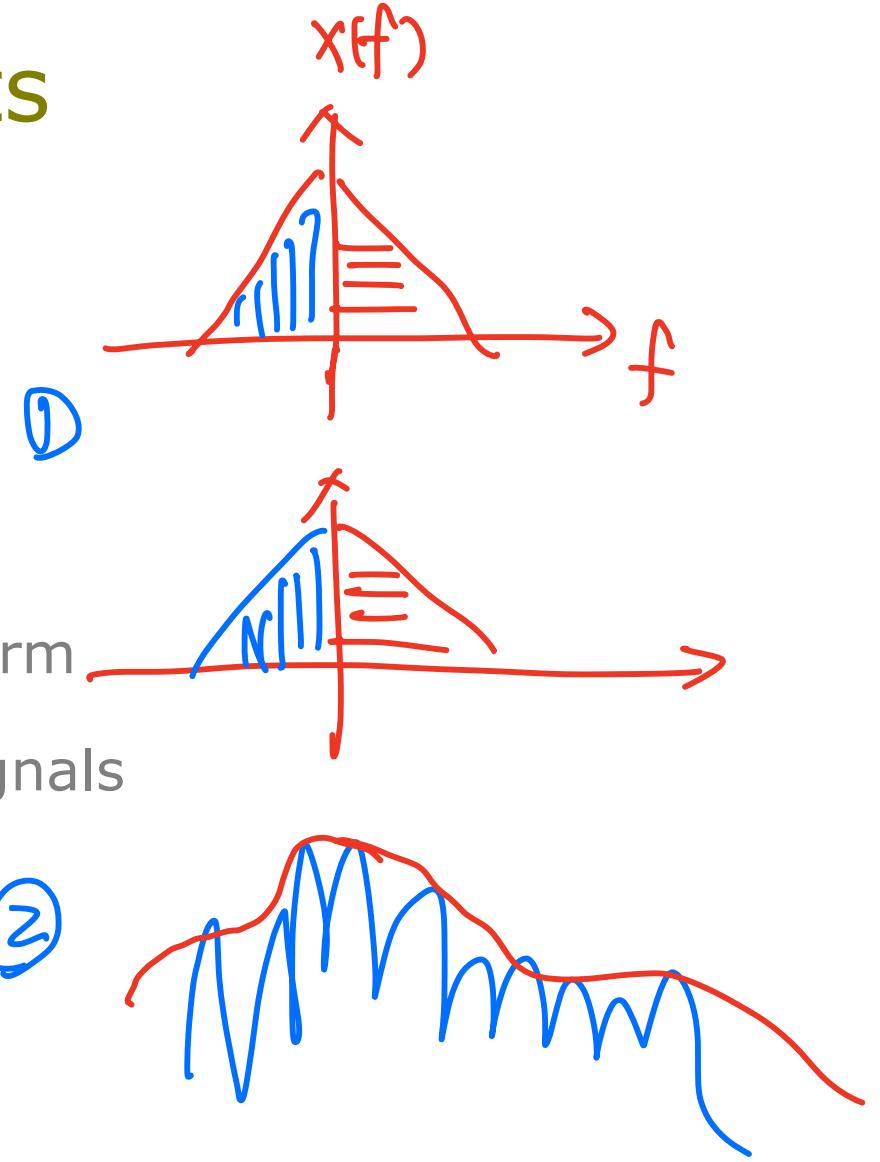
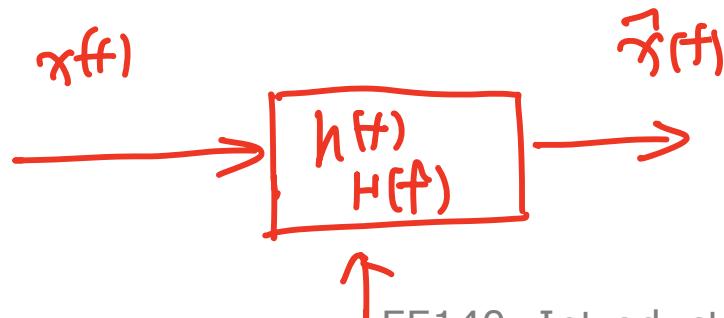
$$= \sum_{n=-\infty}^{+\infty} x(nT_s) \sin(2B(t-nT_s))$$

$$x(t) \Leftrightarrow \{x_1, x_2, \dots, x_n\}$$



# Contents

- Deterministic signals
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  - Review of Fourier Transform
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  - Fourier Transform of Periodic Signals
  - Sampling Theory
  - The Hilbert Transform



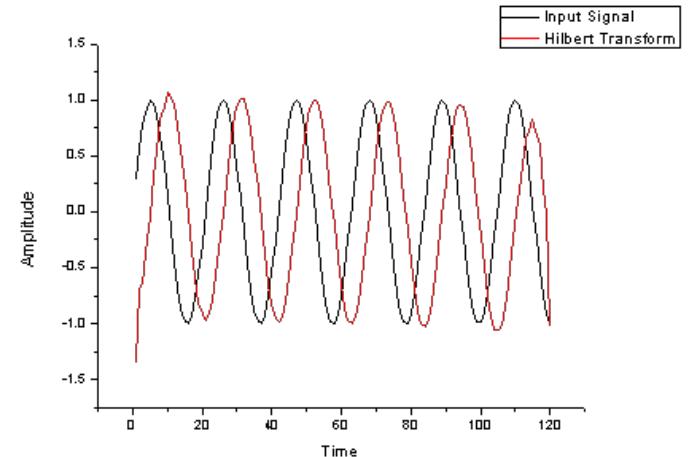
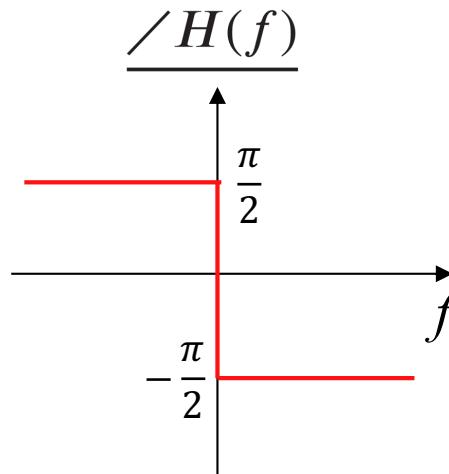
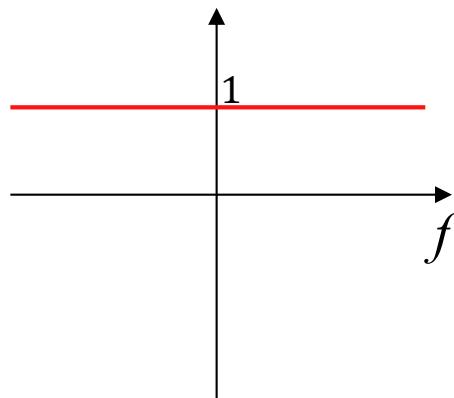
# Hilbert Transform

- Frequency response function

$$H(f) = \boxed{-j \operatorname{sgn} f}$$

$\operatorname{sgn}(f) = \begin{cases} 1, & f > 0 \\ 0, & f = 0 \\ -1, & f < 0 \end{cases}$

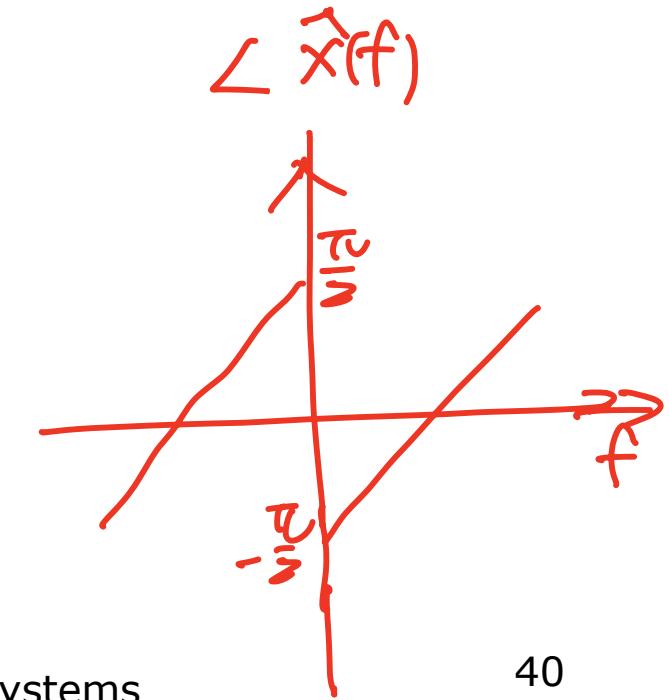
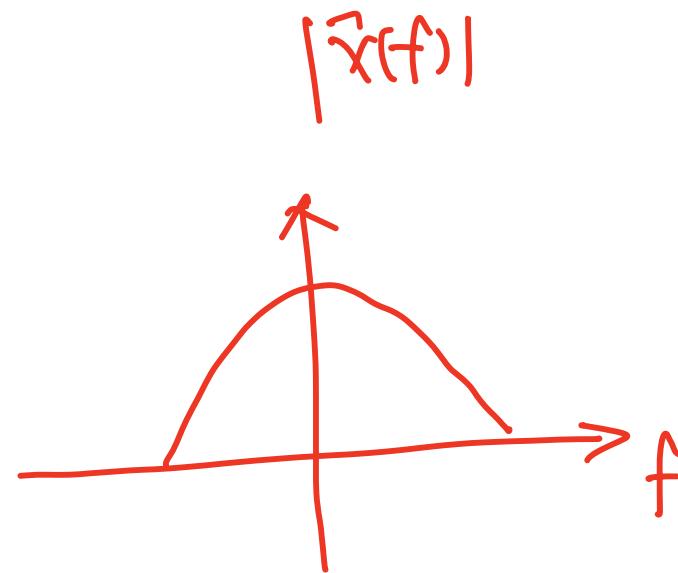
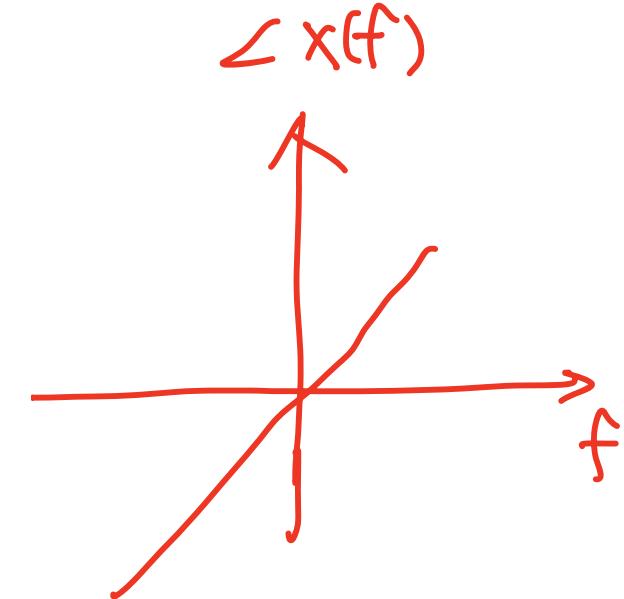
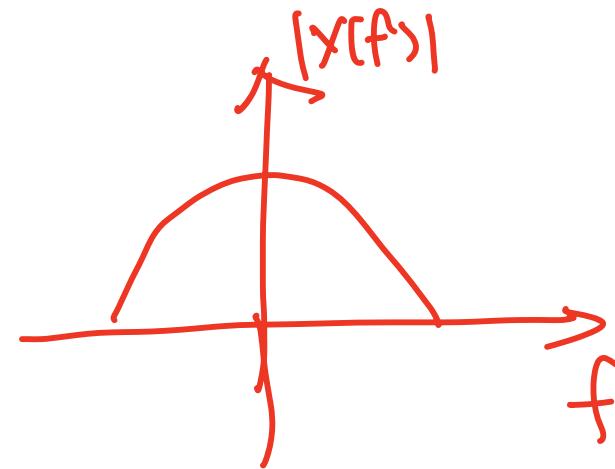
$$= \begin{cases} -j & f > 0 \\ j & f < 0 \end{cases}$$



$$\boxed{\frac{j}{\pi t}} \longleftrightarrow \operatorname{sgn}(f)$$

- Hilbert Transform of  $x(t)$ :
- $$\hat{x}(t) = \mathfrak{F}^{-1}[-j \operatorname{sgn}(f) X(f)]$$
- $$= h(t) * x(t) = \mathcal{T}[t] * \left(\frac{1}{\pi t}\right)$$

$$\hat{x}(f) = \underline{x(f)} H(f) = -j \operatorname{sgn}(f) \underline{x(f)}$$



# Hilbert Transform

- Hilbert Transform of  $x(t)$ :

$$\hat{x}(t) = \mathfrak{I}^{-1}[-j \operatorname{sgn}(f)X(f)]$$

$$= h(t) * x(t)$$

$$\hat{x}(t) = \int_{-\infty}^{\infty} \frac{x(\lambda)}{\pi(t - \lambda)} d\lambda = \int_{-\infty}^{\infty} \frac{x(t - \eta)}{\pi\eta} d\eta$$

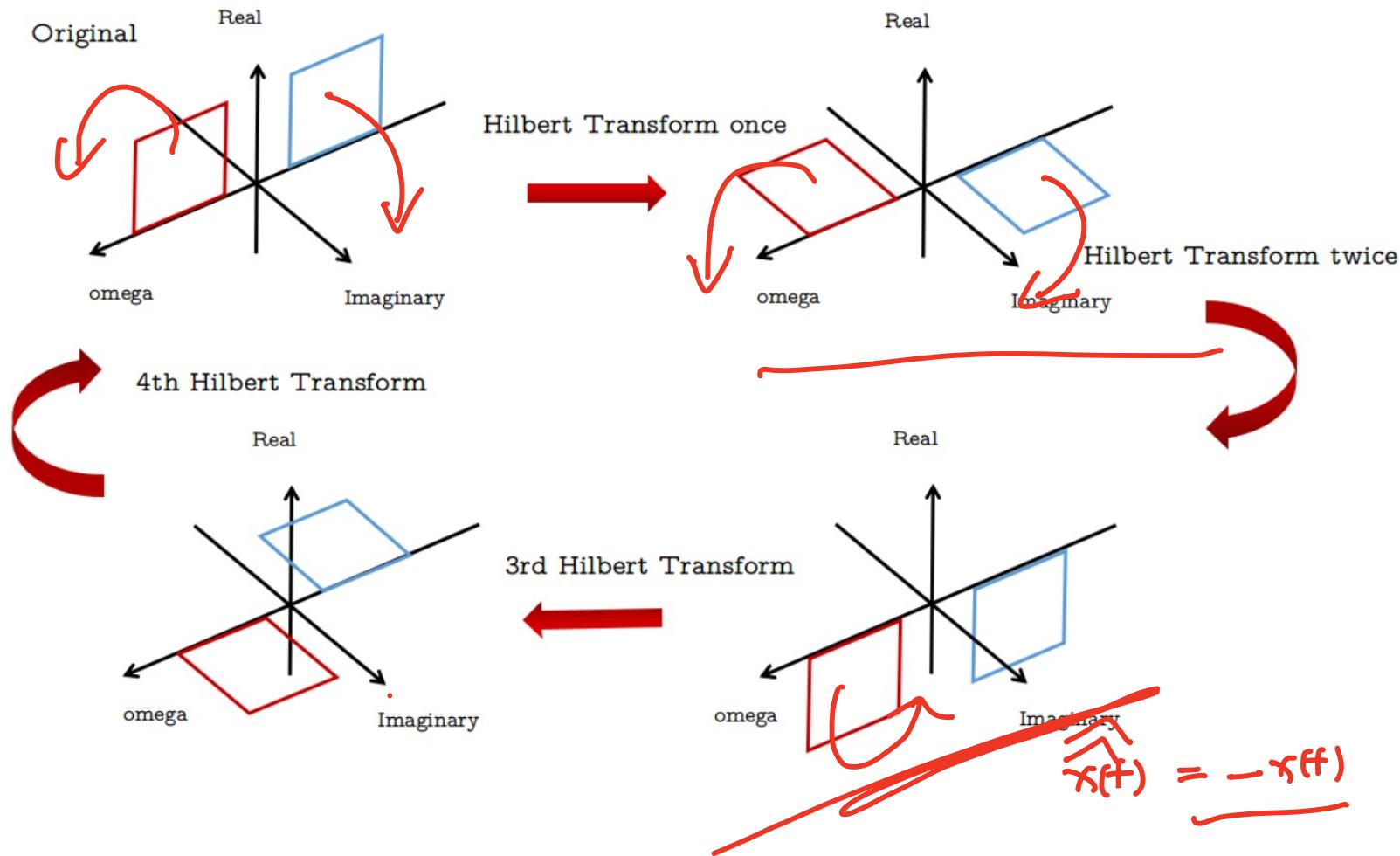
- Hilbert transform of Hilbert transform

$$(-j \operatorname{sgn} f)^2 = -1$$

$$\hat{\hat{x}}(t) = -x(t).$$

# Hilbert Transform

- Hilbert Transform of  $x(t)$ :



# Example

- Calculate the Hilbert transform filter of

$$x(t) = \cos(2\pi f_0 t)$$

- Solution:

$$X(f) = \frac{1}{2}\delta(f - f_0) + \frac{1}{2}\delta(f + f_0)$$

$$\hat{X}(f) = \underbrace{\frac{1}{2}\delta(f - f_0)}_{e^{-j\pi/2}} e^{-j\pi/2} + \underbrace{\frac{1}{2}\delta(f + f_0)}_{e^{j\pi/2}} e^{j\pi/2}$$

$$\hat{x}(t) = \frac{1}{2}e^{j2\pi f_0 t} e^{-j\pi/2} + \frac{1}{2}e^{-j2\pi f_0 t} e^{j\pi/2}$$

$$\widehat{\cos(2\pi f_0 t)} = \sin(2\pi f_0 t)$$

# Properties (P84)

1. Energy in a signal and its Hilbert transform are equal.

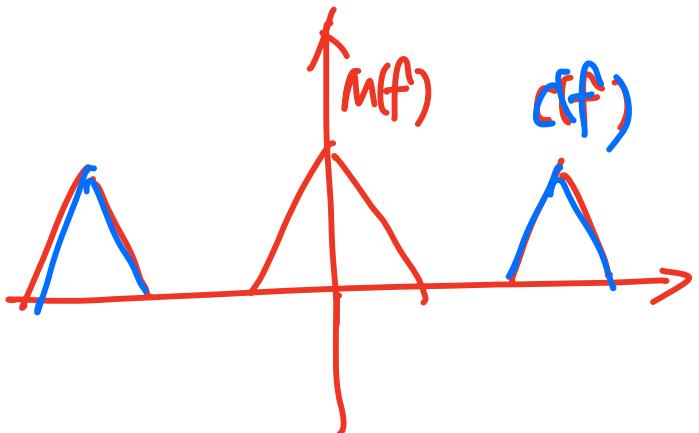
$$|\hat{X}(f)|^2 \triangleq \left| \Im[\hat{x}(t)] \right|^2 = | -j \operatorname{sgn}(f) |^2 |X(f)|^2 = |X(f)|^2$$

2. A signal and its Hilbert transform are orthogonal

$$\int_{-\infty}^{\infty} x(t)\hat{x}(t) dt = 0 \text{ (energy signals)}$$

3. If  $c(t)$  and  $m(t)$  are signals with nonoverlapping spectra, where  $m(t)$  is lowpass and  $c(t)$  is highpass, then

$$\widehat{m(t)c(t)} = \underline{m(t)\hat{c}(t)}$$



$$m(t) = \cos 2\pi f_1 t \quad f_1 = 100 \text{ Hz}$$

$$c(t) = \cos 2\pi f_2 t \quad f_2 = 500 \text{ Hz}$$

$$\overbrace{m(t)c(t)}^{\gamma(t)} = \overbrace{m(t)}^{\cos 2\pi f_1 t} \overbrace{c(t)}^{\cos 2\pi f_2 t}$$

$$= \cos 2\pi f_1 t \sin 2\pi f_2 t$$

# Analytic Signals

- The analytic signal of the real signal  $x(t)$  is

$$\underline{x_p(t)} = \underline{x(t)} + j\underline{\hat{x}(t)}$$

$$X_p(f) = X(f) + j(-j \operatorname{sgn} f) \bar{x}(f)$$

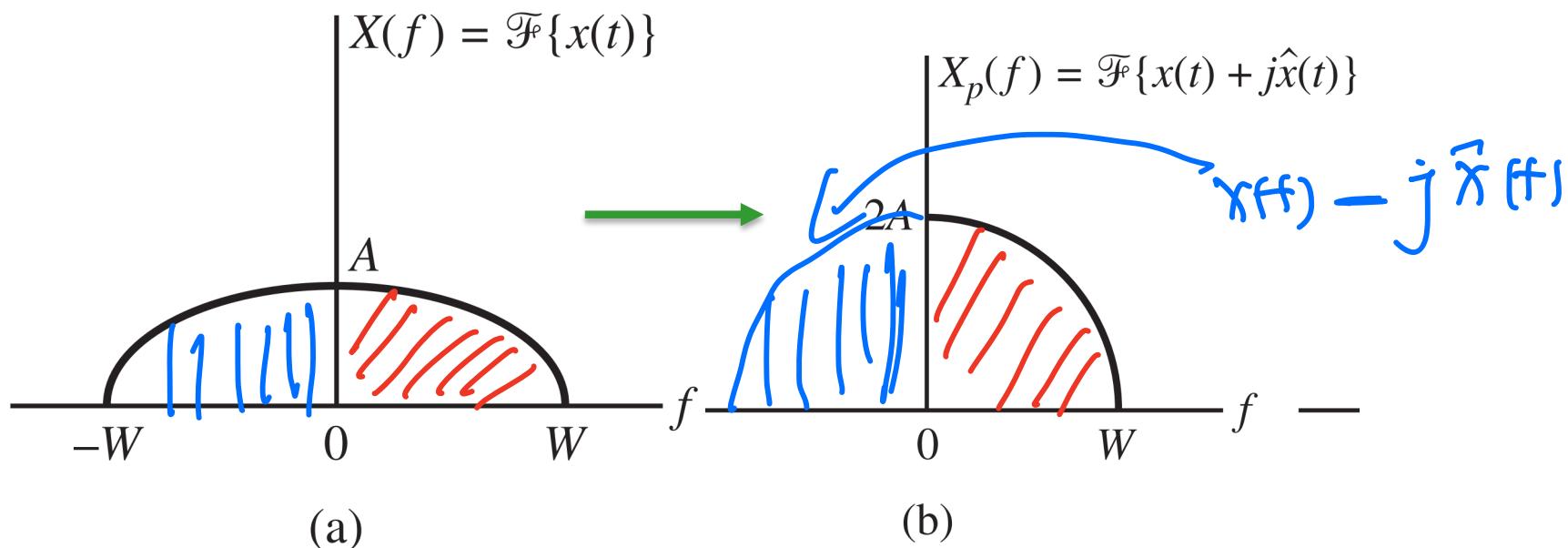
$$= (1 + \operatorname{sgn} f) \bar{x}(f)$$

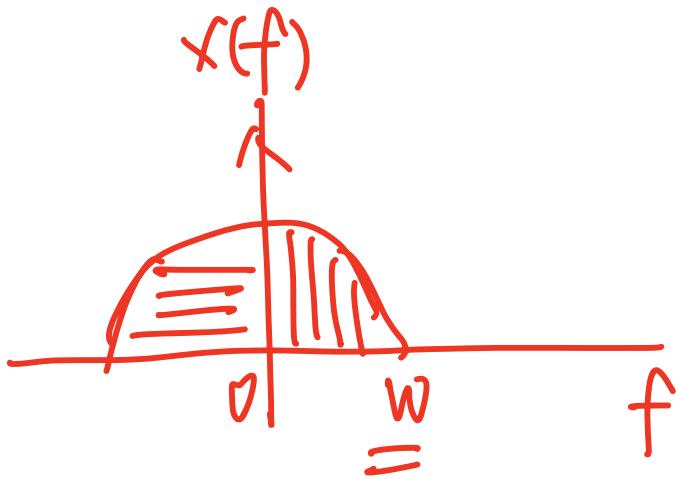
if  $f > 0 \quad X_p(f) = 2x(f)$

- Envelope:  $|x_p(t)| = \sqrt{x(t)^2 + \hat{x}(t)^2}$
- Spectrum:

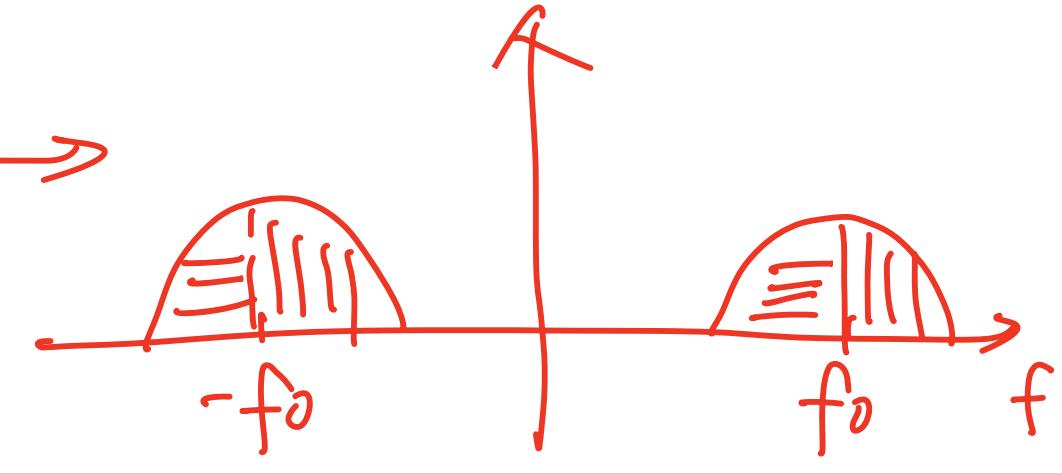
$$X_p(f) = X(f) [1 + \operatorname{sgn} f]$$

: if  $f < 0 \quad X_p(f) = 0$

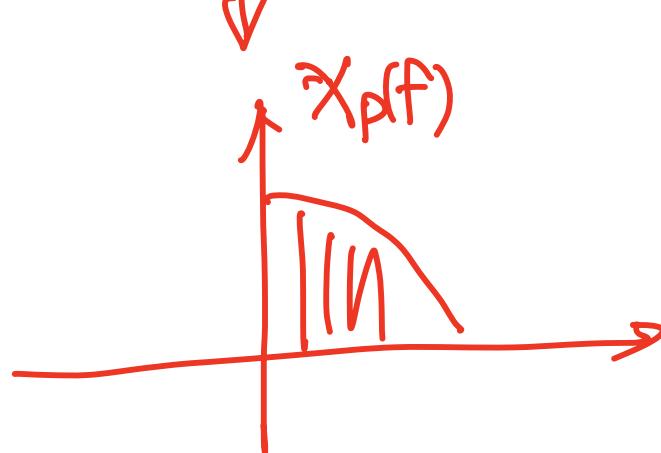




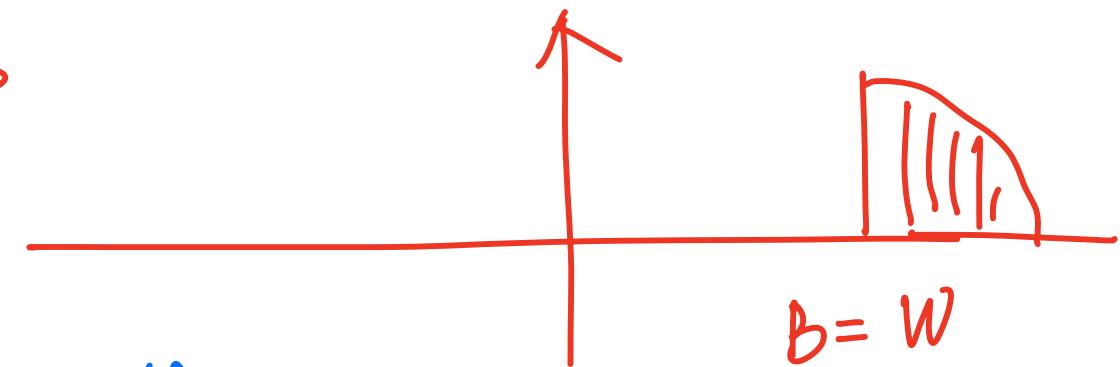
$$x(t) \cos 2\pi f_0 t$$



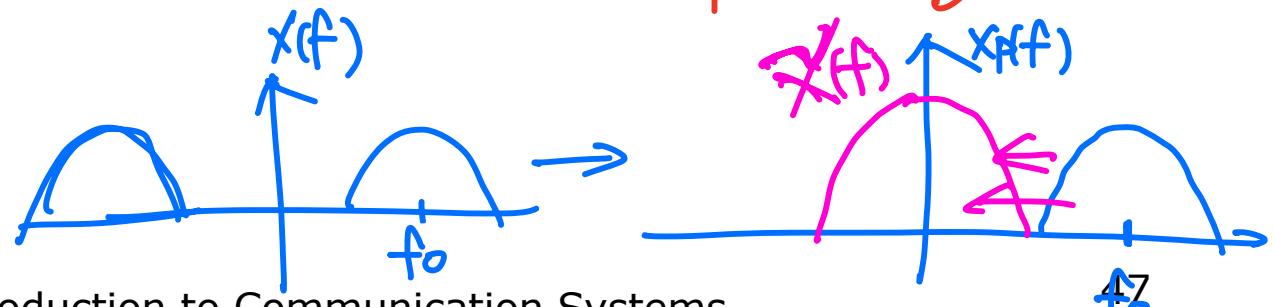
$$B = 2W$$



complex signals



$$B = W$$



$$\tilde{x}(t) = \underline{x_p(t) e^{-j2\pi f_0 t}}$$

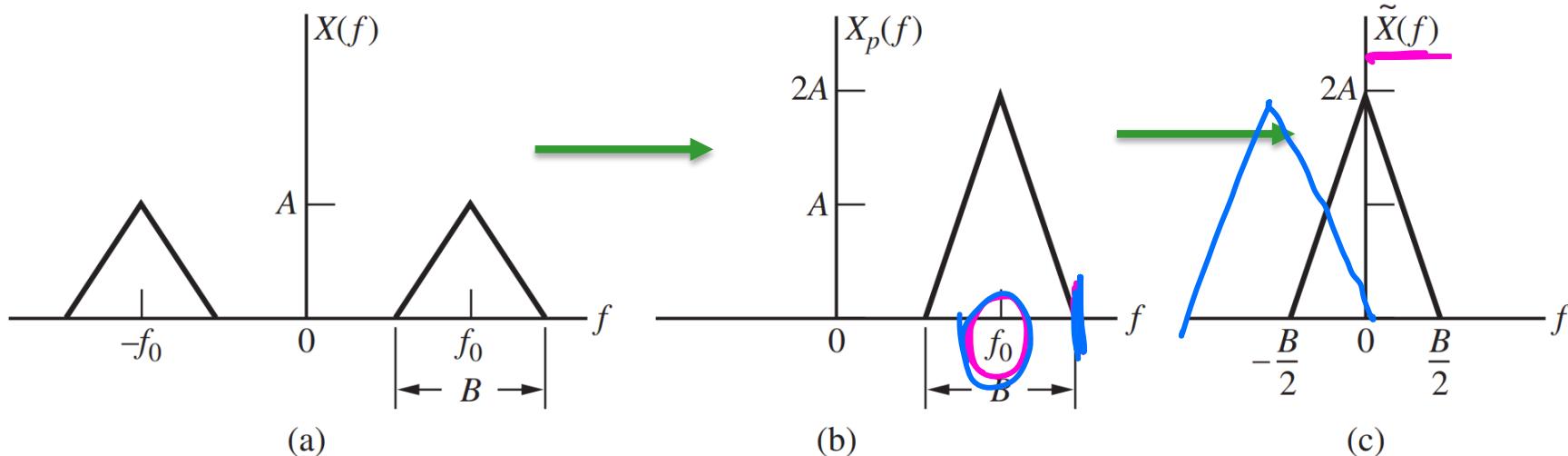
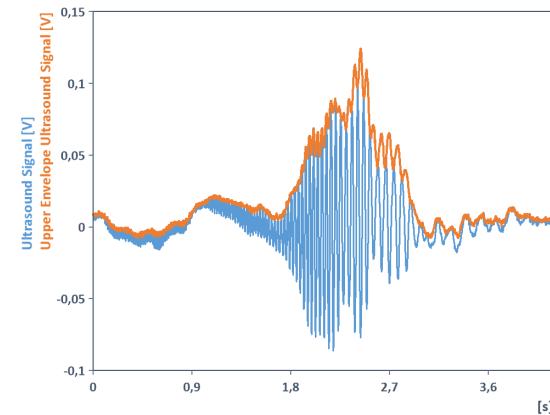
$$\tilde{x}(f) = \underline{x_p(f + f_0)}$$

## Complex Envelope

- The analytic signal can be written as

$$x_p(t) = \underline{\tilde{x}(t)} e^{j2\pi f_0 t}$$

- $\tilde{x}(t)$ : complex envelope of  $x(t)$
- $f_0$ : reference frequency
- Spectrum:



Bandpass signal

lowpass signal

# Inphase and Quadrature Components

- Since

$$\underline{x_p(t)} = \underline{\tilde{x}(t)} e^{j2\pi f_0 t} \triangleq \underline{x(t)} + j \underline{\hat{x}(t)}$$

- Thus

$\text{Re } (\underline{x_p(t)})$

$$\begin{aligned} x(t) &= \underline{\text{Re}} \left[ \tilde{x}(t) e^{j2\pi f_0 t} \right] \\ &= \underline{\text{Re}} \left[ \tilde{x}(t) \right] \cos(2\pi f_0 t) - \underline{\text{Im}} \left[ \tilde{x}(t) \right] \sin(2\pi f_0 t) \\ &= \underline{x_R(t)} \cos(2\pi f_0 t) - \underline{x_I(t)} \sin(2\pi f_0 t) \end{aligned}$$

Inphase  
component of  $x(t)$

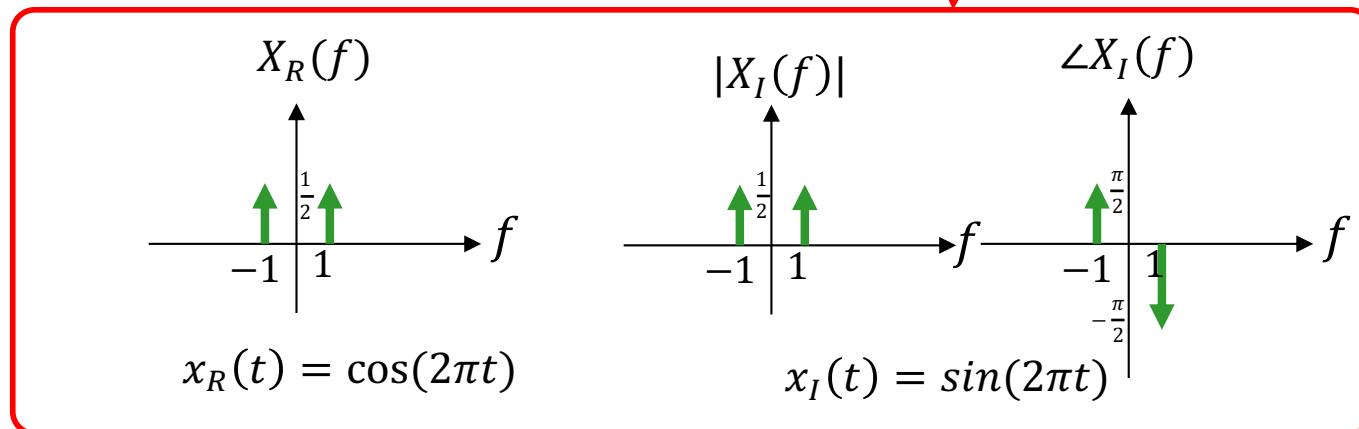
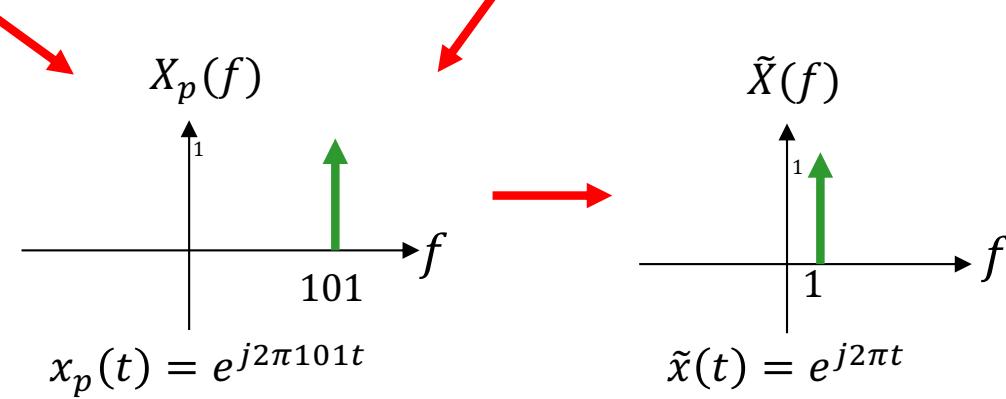
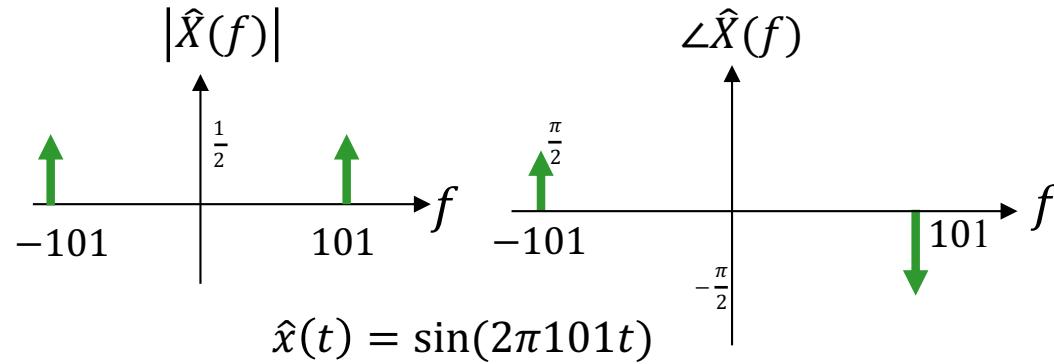
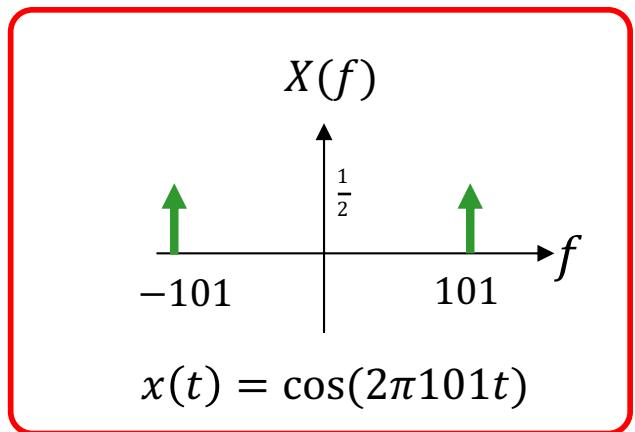
*Lowpass*

Quadrature  
component of  $x(t)$

*Lowpass*

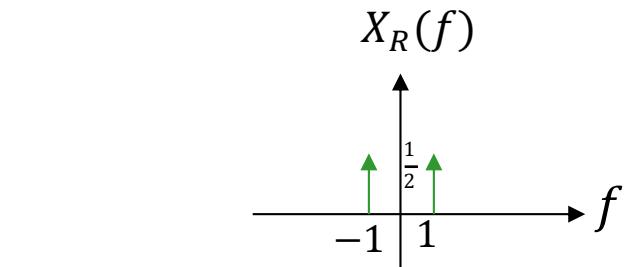
# Example

$$f_0 = 100\text{Hz}$$

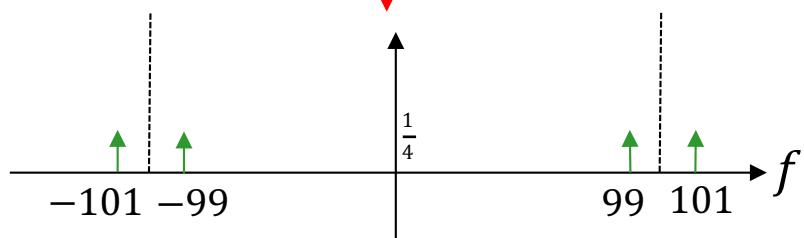


# Example

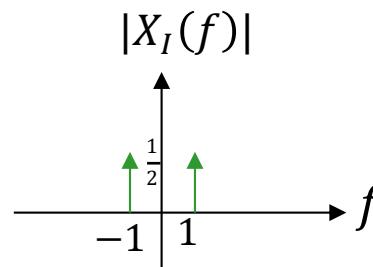
$$f_0 = 100\text{Hz}$$



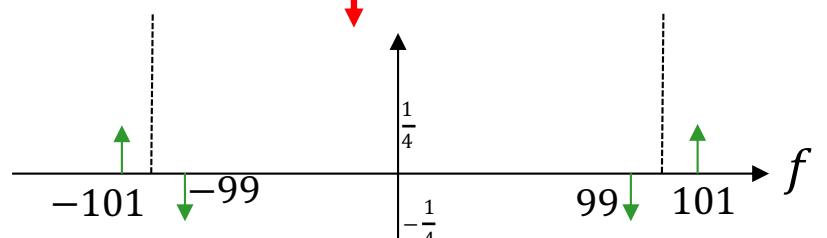
$$x_R(t) = \cos(2\pi t)$$



$$x_R(t)\cos(2\pi 100t)$$



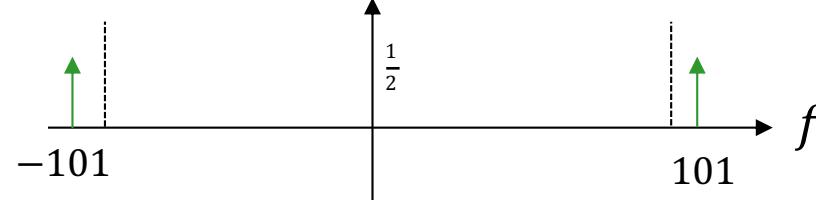
$$x_I(t) = \sin(2\pi t)$$



$$-x_I(t)\sin(2\pi 100t)$$



$$X(f)$$



$$x(t) = \cos(2\pi 101t)$$



上海科技大学  
ShanghaiTech University

Thanks for your kind attention!

Questions?