

Signals and Systems

Fall 2003

Lecture #11

9 October 2003

1. DTFT Properties and Examples
2. Duality in FS & FT
3. Magnitude/Phase of Transforms and Frequency Responses

Convolution Property Example

$$h[n] = \alpha^n u[n], \quad x[n] = \beta^n u[n] \quad |\alpha|, |\beta| < 1$$

$$H(e^{j\omega}) = \frac{1}{1 - \alpha e^{-j\omega}}, \quad X(e^{j\omega}) = \frac{1}{1 - \beta e^{-j\omega}}$$

↓

$$y[n] = h[n] * x[n] \longleftrightarrow Y(e^{j\omega}) = \left(\frac{1}{1 - \alpha e^{-j\omega}} \right) \left(\frac{1}{1 - \beta e^{-j\omega}} \right)$$

- ratio of polynomials in $e^{-j\omega}$

$$\beta \neq \alpha : Y(e^{j\omega}) \stackrel{\text{PFE}}{=} \frac{A}{1 - \alpha e^{-j\omega}} + \frac{B}{1 - \beta e^{-j\omega}}$$

A, B - determined by partial fraction expansion

$$y[n] = A\alpha^n u[n] + B\beta^n u[n]$$

$$\beta = \alpha : Y(e^{j\omega}) = \left(\frac{1}{1 - \alpha e^{-j\omega}} \right)^2$$

$$y[n] = \overbrace{\frac{1}{(1 - \alpha e^{-j\omega})^2}}^{nx[n] \leftrightarrow j \frac{dX(e^{j\omega})}{d\omega}} \quad y[n] = (n + 1)\alpha^n u[n]$$

DT LTI System Described by LCCDE's

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^M b_k x[n-k]$$

From time-shifting property: $x[n-k] \longleftrightarrow e^{-j\omega k} X(e^{j\omega})$

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$$\sum_{k=0}^N a_k e^{-jk\omega} Y(e^{j\omega}) = \sum_{k=0}^M b_k e^{-jk\omega} X(e^{j\omega})$$

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$$Y(e^{j\omega}) = \underbrace{\left[\frac{\sum_{k=0}^M b_k e^{-jk\omega}}{\sum_{k=0}^N a_k e^{-jk\omega}} \right]}_{H(e^{j\omega})} X(e^{j\omega})$$

— Rational function of $e^{j\omega}$,
use PFE to get $h[n]$

Example: First-order recursive system

$$y[n] - \alpha y[n - 1] = x[n], \quad |\alpha| < 1$$

with the condition of initial rest \Leftrightarrow *causal*

$$(1 - \alpha e^{-j\omega})Y(e^{j\omega}) = X(e^{j\omega})$$

$$H(e^{j\omega}) = \frac{Y(e^{j\omega})}{X(e^{j\omega})} = \frac{1}{1 - \alpha e^{-j\omega}}$$

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$$h[n] = \alpha^n u[n]$$

DTFT Multiplication Property

$$\begin{aligned}y[n] = x_1[n] \cdot x_2[n] \longleftrightarrow Y(e^{j\omega}) &= \frac{1}{2\pi} \int_{2\pi} X_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta \\ &= \frac{1}{2\pi} X_1(e^{j\omega}) \otimes X_2(e^{j\omega}) \\ &\hookrightarrow \text{Periodic Convolution}\end{aligned}$$

Derivation:

$$\begin{aligned}Y(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} x_1[n] \cdot x_2[n] e^{-j\omega n} \\ &= \sum_{n=-\infty}^{\infty} \left(\frac{1}{2\pi} \int_{2\pi} X_1(e^{j\theta}) e^{j\theta n} d\theta \right) x_2[n] e^{-j\omega n} \\ &= \frac{1}{2\pi} \int_{2\pi} (X_1(e^{j\theta}) \underbrace{\sum_{n=-\infty}^{\infty} x_2[n] e^{-j(\omega-\theta)n}}_{X_2(e^{j(\omega-\theta)})}) d\theta \\ &= \frac{1}{2\pi} \int_{2\pi} X_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta\end{aligned}$$

Calculating Periodic Convolutions

Suppose we integrate from $-\pi$ to π :

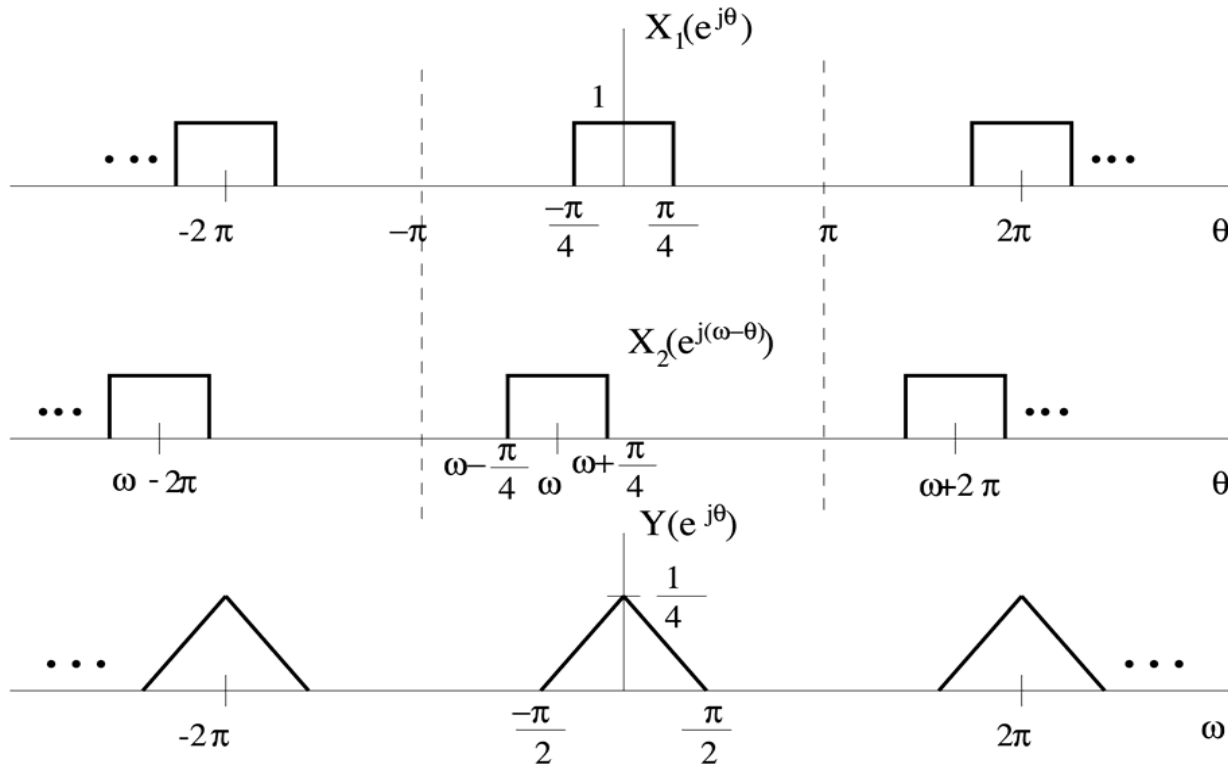
$$\begin{aligned} Y(e^{j\omega}) &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{X}_1(e^{j\theta}) X_2(e^{j(\omega-\theta)}) d\theta \end{aligned}$$

where

$$\hat{X}_1(e^{j\theta}) = \begin{cases} X_1(e^{j\theta}), & |\theta| \leq \pi \\ 0, & \text{otherwise} \end{cases}$$

Example: $y[n] = \left(\frac{\sin(\pi n/4)}{\pi n} \right)^2 = x_1[n] \cdot x_2[n], \quad x_1[n] = x_2[n] = \frac{\sin(\pi n/4)}{\pi n}$

$$Y(e^{j\omega}) = \frac{1}{2\pi} X_1(e^{j\omega}) \otimes X_2(e^{j\omega})$$



Duality in Fourier Analysis

Fourier Transform is highly symmetric

CTFT: *Both* time and frequency are continuous and in general *aperiodic*

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$$

Same except for
these differences

$$X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$$

⇓

Suppose $f(\cdot)$ and $g(\cdot)$ are two functions related by

$$f(r) = \int_{-\infty}^{\infty} g(\tau) e^{-jr\tau} d\tau$$

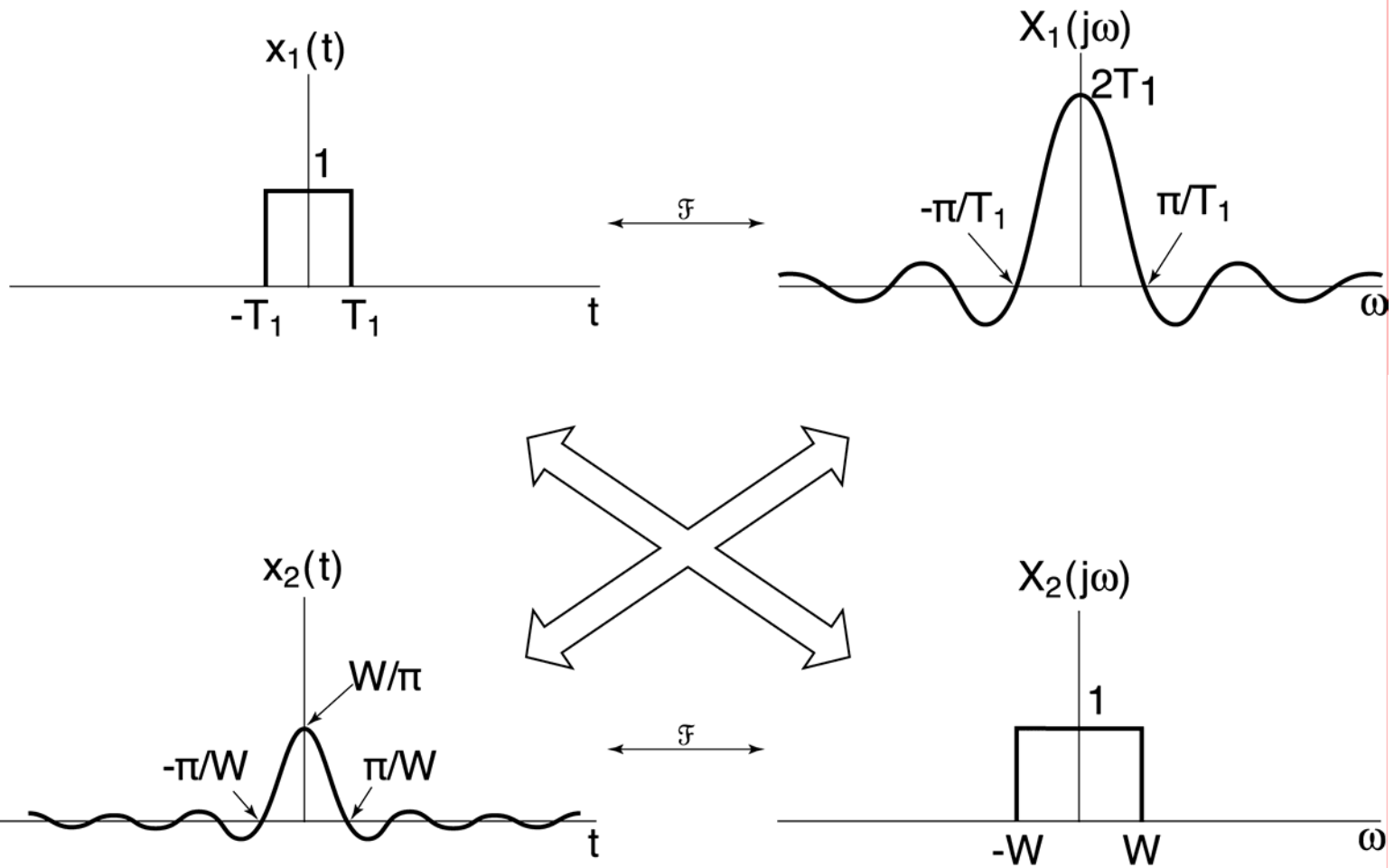
Then

Let $\tau = t$ and $r = \omega$: $x_1(t) = g(t) \longleftrightarrow X_1(j\omega) = f(\omega)$

Let $\tau = -\omega$ and $r = t$: $x_2(t) = f(t) \longleftrightarrow X_2(j\omega) = 2\pi g(-\omega)$

Example of CTFT duality

Square pulse in either time or frequency domain



DTFS

Discrete & periodic in time \longleftrightarrow *Periodic & discrete in frequency*

$$x[n] = \sum_{k=\langle N \rangle} a_k e^{jk\omega_0 n} = x[n + N], \quad \omega_0 = \frac{2\pi}{N}$$

$$a_k = \frac{1}{N} \sum_{k=\langle N \rangle} x[n] e^{-jk\omega_0 n} = a_{k+N}$$

Duality in DTFS

Suppose $f[\cdot]$ and $g[\cdot]$ are two functions related by

$$f[m] = \frac{1}{N} \sum_{r=\langle N \rangle} g[r] e^{-jr\omega_0 m}$$

$$\Rightarrow g[r] = \sum_{m=\langle N \rangle} f[m] e^{jr\omega_0 m}$$

Then

Let $m = n$ and $r = -k$: $x_1[n] = f[n] \longleftrightarrow a_k = \frac{1}{N} g[-k]$

Let $r = n$ and $m = k$: $x_2[n] = g[n] \longleftrightarrow a_k = f[k]$

Duality between CTFS and DTFT

CTFS $x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t} = x(t + T), \quad \omega_0 = \frac{2\pi}{T}$

$$a_k = \frac{1}{T} \int_T x(t) e^{-jk\omega_0 t} dt$$

Periodic in time \leftrightarrow Discrete in frequency

DTFT

$$x[n] = \frac{1}{2\pi} \int_{2\pi} X(e^{j\omega}) e^{j\omega n} d\omega$$

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} = X(e^{j(\omega+2\pi)})$$

Discrete in time \leftrightarrow Periodic in frequency

CTFS-DTFT Duality

Suppose $f(\cdot)$ is a CT signal and $g[\cdot]$ a DT sequence related by

$$f(\tau) = \sum_{m=-\infty}^{+\infty} g[m] e^{jm\tau} = f(\tau + 2\pi)$$

Then

$$x(t) = f(t) \longleftrightarrow a_k = g[k]$$

(periodic with period 2π)

$$x[n] = g[n] \longleftrightarrow X(e^{j\omega}) = f(-\omega)$$

Magnitude and Phase of FT, and Parseval Relation

CT:
$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$$

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

Parseval Relation:
$$\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} \frac{1}{2\pi} \underbrace{|X(j\omega)|^2}_{\text{Energy density in } \omega} d\omega$$

DT:
$$x[n] = \frac{1}{2\pi} \int_{2\pi} X(e^{j\omega}) e^{j\omega n} d\omega$$

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

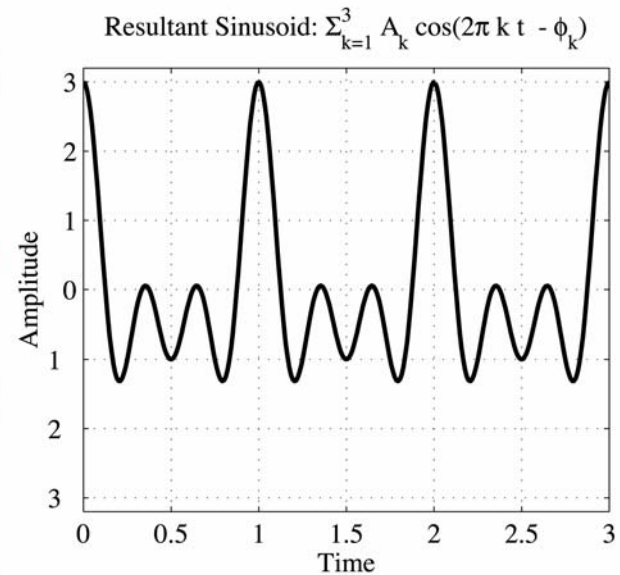
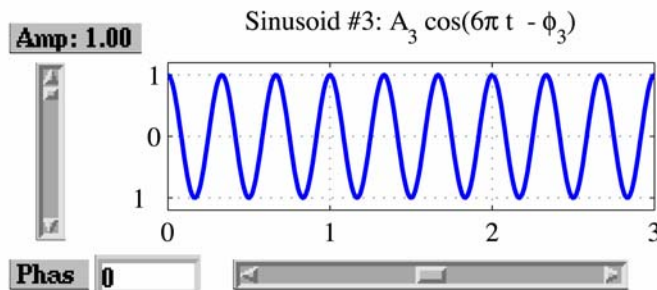
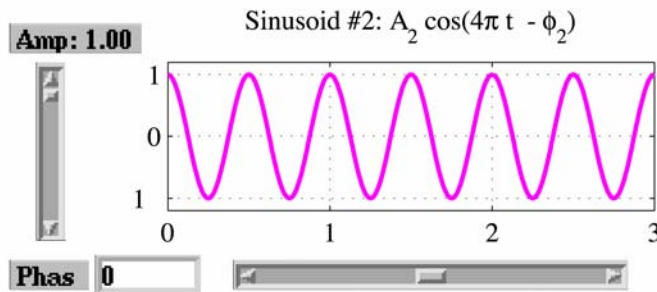
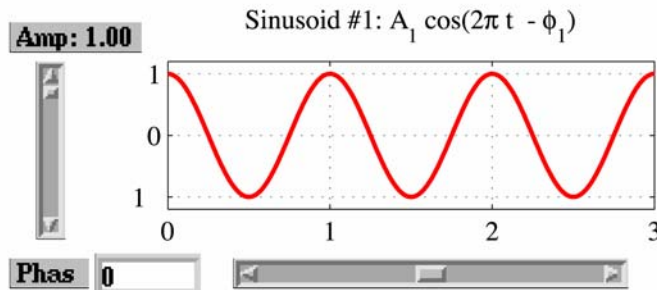
Parseval Relation:
$$\sum_{n=-\infty}^{\infty} |x[n]|^2 = \int_{2\pi} \frac{1}{2\pi} |X(e^{j\omega})|^2 d\omega$$

Effects of Phase

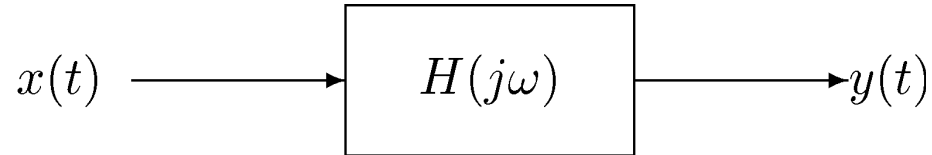
- *Not* on signal energy distribution as a function of frequency
- *Can* have dramatic effect on signal shape/character
 - Constructive/Destructive interference
- Is that important?
 - Depends on the signal and the context

Demo:

- 1) Effect of phase on Fourier Series
- 2) Effect of phase on image processing



Log-Magnitude and Phase

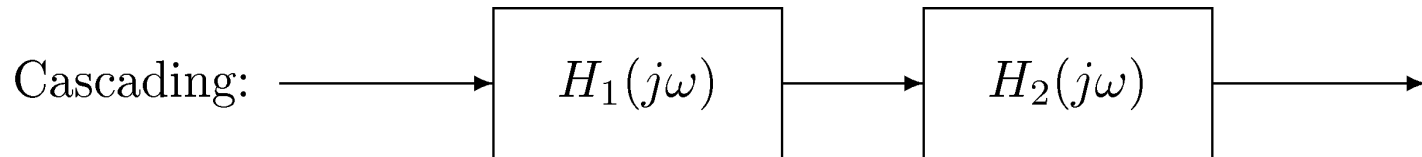


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$$|Y(j\omega)| = |H(j\omega)| \cdot |X(j\omega)|$$

or $\log |Y(j\omega)| = \log |H(j\omega)| + \log |X(j\omega)|$

and $\angle Y(j\omega) = \angle H(j\omega) + \angle X(j\omega)$



$$\left. \begin{aligned} \log |H(j\omega)| &= \log |H_1(j\omega)| + \log |H_2(j\omega)| \\ \angle H(j\omega) &= \angle H_1(j\omega) + \angle H_2(j\omega) \end{aligned} \right\} \text{Easy to add}$$

Plotting Log-Magnitude and Phase

a) For real-valued signals and systems

$$\left. \begin{aligned} |H(-j\omega)| &= |H(j\omega)| \\ \angle H(-j\omega) &= -\angle H(j\omega) \end{aligned} \right\} \Rightarrow \text{Plot for } \omega \geq 0, \text{ often with a } \textit{logarithmic} \text{ scale for frequency in CT}$$

b) In DT, need only plot for $0 \leq \omega \leq \pi$ (with *linear* scale)

c) For historical reasons, log-magnitude is usually plotted in units of *decibels* (dB): $(1 \text{ bel} = 10 \text{ decibels} = \frac{\text{output power}}{\text{input power}} = 10)$

$$10 \log_{10} |H(j\omega)|^2 = 20 \log_{10} |H(j\omega)|$$

↙ power
↙ magnitude

$$|H(j\omega)| = 1 \quad \longrightarrow \quad 0 \text{ dB}$$

$$|H(j\omega)| = \sqrt{2} \quad \longrightarrow \quad \sim 3 \text{ dB}$$

$$|H(j\omega)| = 2 \quad \longrightarrow \quad \sim 6 \text{ dB}$$

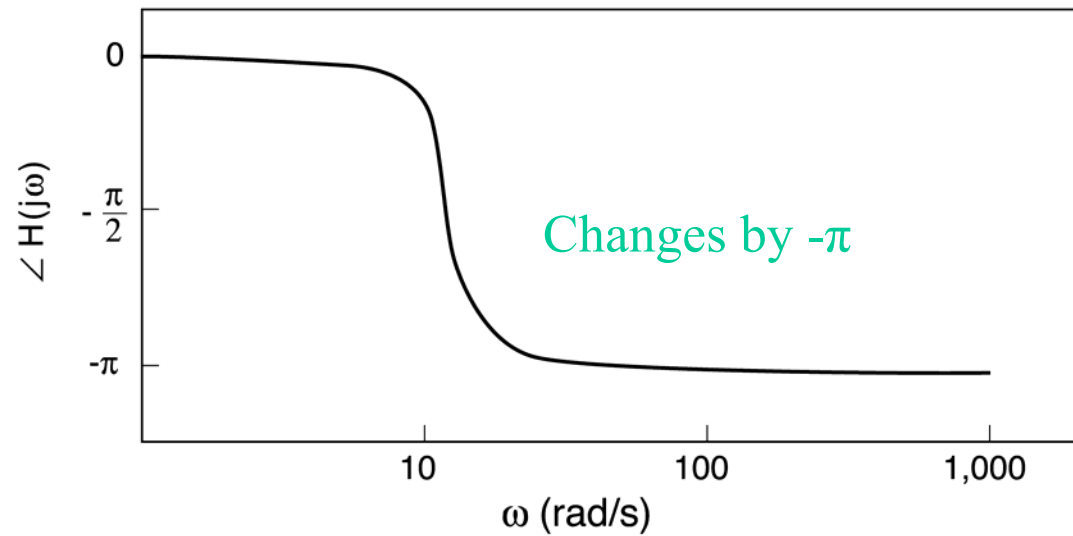
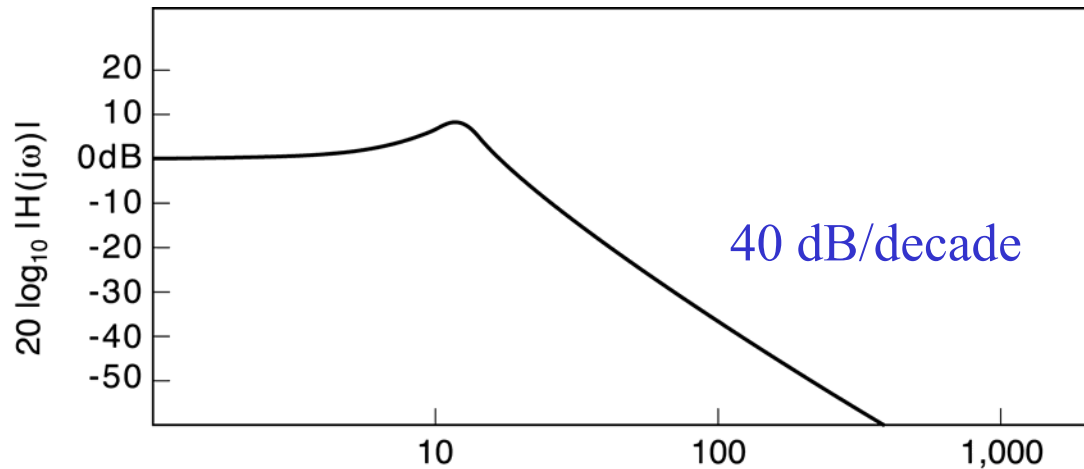
$$|H(j\omega)| = 10 \quad \longrightarrow \quad 20 \text{ dB}$$

$$|H(j\omega)| = 100 \quad \longrightarrow \quad 40 \text{ dB}$$

So... 20 dB or 2 bels:
 = 10 amplitude gain
 = 100 power gain

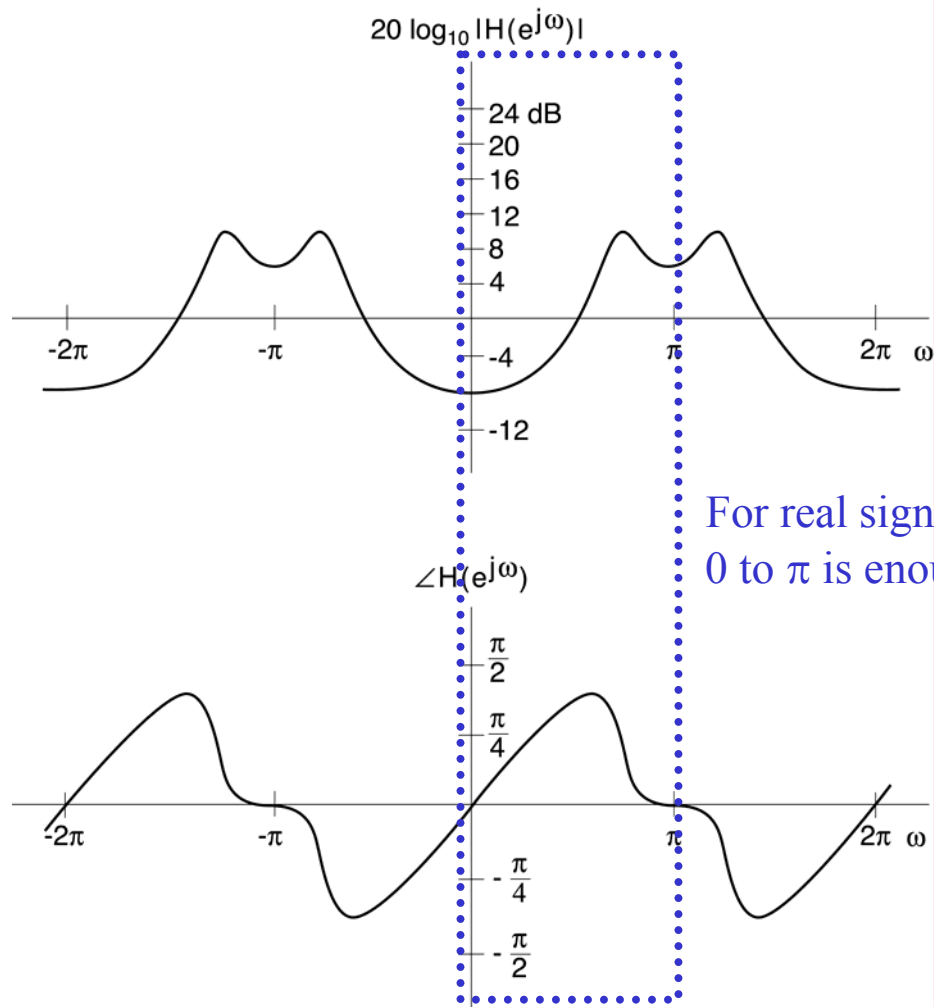
A Typical Bode plot for a second-order CT system

$20 \log|H(j\omega)|$ and $\angle H(j\omega)$ vs. $\log \omega$



A typical plot of the magnitude and phase of a second-order DT frequency response

$20\log|H(e^{j\omega})|$ and $\angle H(e^{j\omega})$ vs. ω



For real signals,
0 to π is enough