

LTI Systems have an impulse response

$$g(n) \rightarrow h(n)$$

$$g[n] \rightarrow h[n]$$

Time Invariance:  $g(n-k) \rightarrow h(n-k)$

$$g[n-k] \rightarrow h[n-k]$$

"[ ]" = Discrete Time  
"( )" = Continuous Time

Linearity:  $x(k)g(n-k) \rightarrow x(k)h(n-k)$   
 $x[k]g[n-k] \rightarrow x[k]h[n-k]$

Superposition:

$$\sum_{k=-\infty}^{\infty} x(k)g(n-k) \rightarrow \sum_{k=-\infty}^{\infty} x(k)h(n-k)$$

$$\therefore y(n) = \sum_{k=-\infty}^{\infty} x(k)h(n-k) \Rightarrow \sum_{k=-\infty}^{\infty} x(n-k)h(k) \Rightarrow x(n) * h(n)$$

convolution!

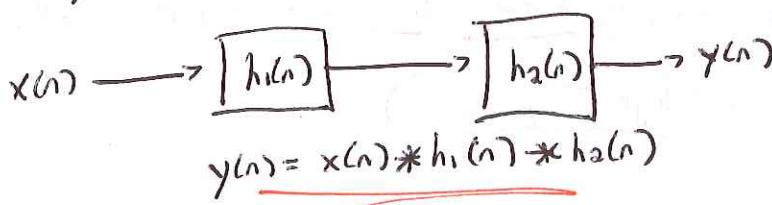
$$h(n) * x(n) = x(n) * h(n)$$

$$y(n) = \sum_{k=-\infty}^{\infty} x(k)h(n-k)$$

$$\Rightarrow \begin{array}{l} h(k) \\ h(-k) \\ h(1-k) \\ h(-1-k) \end{array}$$

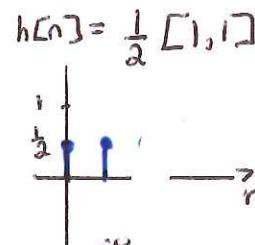
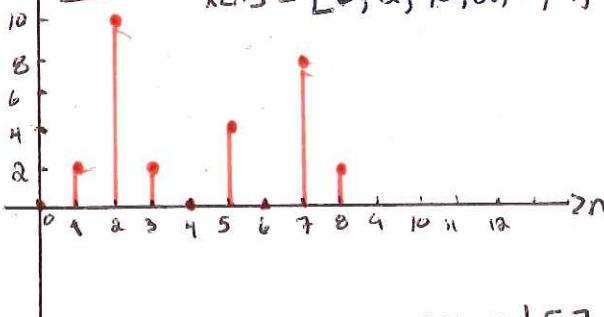
$$\begin{array}{l} y(0) \\ y(1) \\ y(-1) \end{array}$$

Cascaded systems:



Example

$$x[n] = [0, 2, 10, 2, 0, 4, 0, 8, 2]$$



$$y[n] = x[n] * h[n] \rightarrow y[n] = \sum_{k=-\infty}^{\infty} x[n-k] h[k]$$

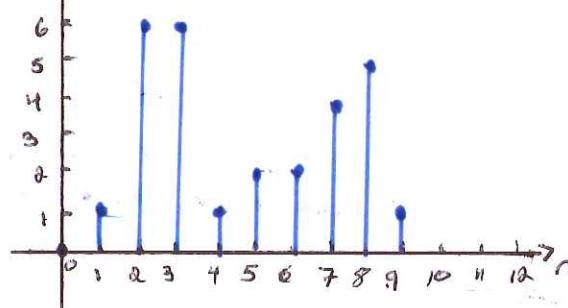
$$n=0 \Rightarrow \sum_{k=-\infty}^{\infty} x[-k] h[k] = y[0] = 0 \\ = x_0 h_0 = 0 \cdot \frac{1}{2} = 0$$

$$n=1 \Rightarrow \sum_{k=-\infty}^{\infty} x[-k] h[k] = y[1] = 1 \\ = x_1 h_0 + x_0 h_1 = 2 \cdot \frac{1}{2} + 0 \cdot \frac{1}{2} = 1$$

$$n=2 \Rightarrow \sum_{k=-\infty}^{\infty} x[-k] h[k] = y[2] = 6 \\ = x_2 h_0 + x_1 h_1 + x_0 h_2 \\ 10 \cdot \frac{1}{2} + 2 \cdot \frac{1}{2} = 6$$

If you work it all the way out, you get

$$y[n] = [0, 1, 6, 6, 1, 2, 2, 4, 5, 1]$$



$h[n] = \frac{1}{2}[1, 1]$  is a smoothing, averaging filter (low-pass filter) that attenuates or suppresses sharp changes (high frequencies)

## Fourier Transform "FT"

### Continuous Time Fourier Transform "CTFT"

For a signal  $x(t)$ , its FT is :  $X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$

The Inverse FT "IFT" of  $X(j\omega)$ :  $x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$

Let's find the FT of a cosine wave with  $x(t) = \cos(2\pi\omega_0 t)$

$$\therefore X(j\omega) = \int_{-\infty}^{\infty} \cos(2\pi\omega_0 t) e^{-j\omega t} dt$$

use Euler's identity  $\cos(2\pi\omega_0 t) = \frac{e^{j2\pi\omega_0 t} + e^{-j2\pi\omega_0 t}}{2}$

$$\therefore X(j\omega) = \int_{-\infty}^{\infty} \left[ \frac{e^{j2\pi\omega_0 t} + e^{-j2\pi\omega_0 t}}{2} \right] e^{-j\omega t} dt$$

$$= \frac{1}{2} \int_{-\infty}^{\infty} \left[ e^{-j\omega_0(\omega - \omega_0)t} + e^{-j\omega_0(\omega + \omega_0)t} \right] dt$$

By definition

$$e^{j\omega_0 t} \xrightarrow{\text{FT}} \delta(\omega - \omega_0)$$

a delta with magnitude  $2\pi$  at  $\omega_0$

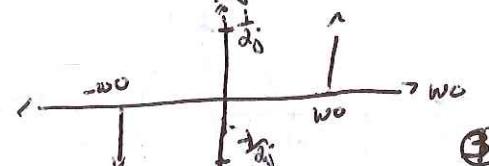
$$= \frac{1}{2} [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$$



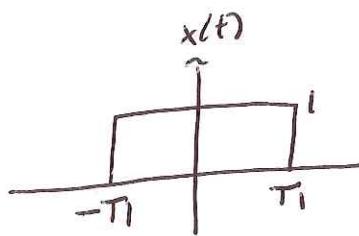
Similar for sine  $\Rightarrow \sin(2\pi\omega_0 t) = \frac{e^{j\omega_0 t} - e^{-j\omega_0 t}}{2j}$

$$\therefore X(j\omega) = \int_{-\infty}^{\infty} \frac{1}{2j} \left[ e^{j2\pi\omega_0 t} - e^{-j2\pi\omega_0 t} \right] e^{-j\omega t} dt$$

$$\rightarrow X(j\omega) = \frac{1}{2j} [f(\omega - \omega_0) - f(\omega + \omega_0)]$$



$$x(t) = \begin{cases} 1, & |t| < T_1 \\ 0, & |t| > T_1 \end{cases}$$



$$X(j\omega) = \int_{-T_1}^{T_1} e^{-j\omega t} dt = \frac{1}{-j\omega} \left[ e^{j\omega t} \right]_{-T_1}^{T_1}$$

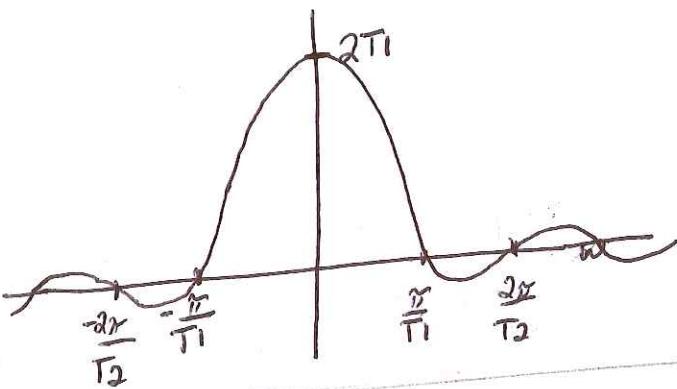
$$= \frac{1}{-j\omega} \left[ e^{-j\omega T_1} - e^{j\omega T_1} \right] \text{ Distribute negative} \Rightarrow \frac{1}{j\omega} \left[ e^{j\omega T_1} - e^{-j\omega T_1} \right]$$

Manipulate with  $\frac{2}{j}$

$$\frac{2}{\omega} \left[ \frac{e^{j\omega T_1} - e^{-j\omega T_1}}{2} \right] = \sin(\omega T_1)$$

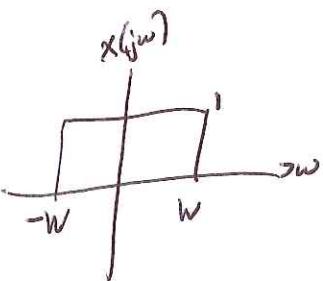
$$\therefore X(j\omega) = \frac{2 \sin(\omega T_1)}{\omega}$$

$$X(j\omega)$$

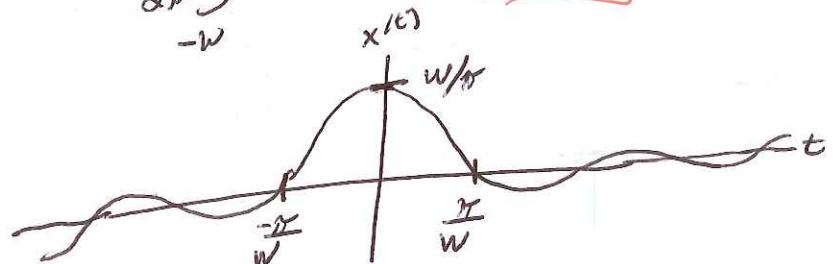


Let's Try an Inverse!

$$x(j\omega) = \begin{cases} 1, & |\omega| < W \\ 0, & |\omega| > W \end{cases}$$

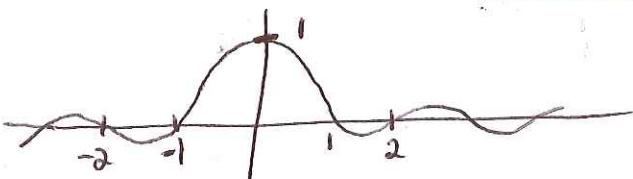


$$x(t) = \frac{1}{2\pi} \int_{-W}^W e^{j\omega t} d\omega = \frac{\sin \omega t}{\pi t}$$



$$\operatorname{sinc} \theta$$

$$\operatorname{sinc} \text{ function : } \operatorname{sinc}(\theta) = \frac{\sin \theta}{\pi \theta}$$



## Parseval's Relation

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

## FT Properties

convolution  $y(t) = h(t) * x(t) \xleftarrow{\mathcal{F}} Y(j\omega) = H(j\omega) X(j\omega)$

linearity  $a x(t) + b y(t) \xrightarrow{\mathcal{F}} a X(j\omega) + b Y(j\omega)$

time shifting  $x(t-t_0) \xrightarrow{\mathcal{F}} e^{-j\omega t_0} X(j\omega)$

frequency shifting  $e^{j\omega_0 t} x(t) \xrightarrow{\mathcal{F}} X(j(\omega - \omega_0))$

Multiplication  $x(t) y(t) \xrightarrow{\mathcal{F}} \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\theta) Y(j(\omega - \theta)) d\theta$

## Basic FT Pairs

$$e^{j\omega_0 t} \xleftarrow{\mathcal{F}} 2\pi g(\omega - \omega_0)$$

$$\cos(\omega_0 t) \xleftarrow{\mathcal{F}} \frac{1}{2} [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$$

$$\sin(\omega_0 t) \xleftarrow{\mathcal{F}} \frac{1}{j} [\delta(\omega - \omega_0) - \delta(\omega + \omega_0)]$$

$$x(t) = 1 \xleftarrow{\mathcal{F}} 2\pi g(\omega)$$

$$x(t) = \begin{cases} 1, & |t| < T \\ 0, & |t| > T \end{cases} \xleftarrow{\mathcal{F}} \frac{2\sin(\omega T)}{\omega}$$

$$\frac{\sin(\omega t)}{\omega t} \xleftarrow{\mathcal{F}} X(j\omega) = \begin{cases} 1, & |\omega| < \omega \\ 0, & |\omega| > \omega \end{cases}$$

$$\delta(t) \xleftarrow{\mathcal{F}} 1$$

$$\delta(t-t_0) \xleftarrow{\mathcal{F}} e^{-j\omega t_0}$$

## Time and Frequency Domain Relations b.p.

CTFT:  $X(j\omega)$

→ magnitude and phase representation :  $X(j\omega) = |X(j\omega)| e^{j \arg X(j\omega)}$

DTFT:  $X(e^{j\omega})$

" :  $X(e^{j\omega}) = |X(e^{j\omega})| e^{j \arg X(e^{j\omega})}$

## Frequency Response of LTI Systems

$$\text{CTFT: } Y(j\omega) = H(j\omega) X(j\omega)$$

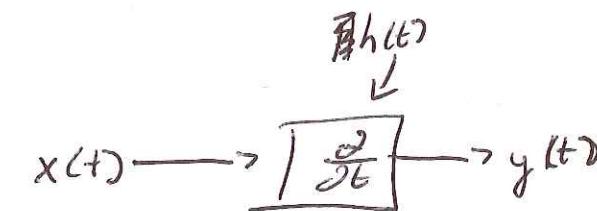
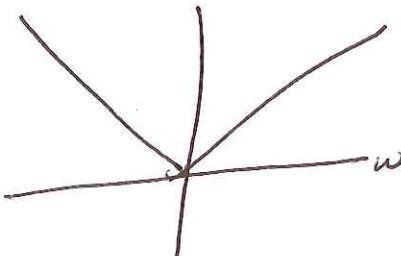
$$\text{DTFT: } Y(e^{j\omega}) = H(e^{j\omega}) X(e^{j\omega})$$

Filters

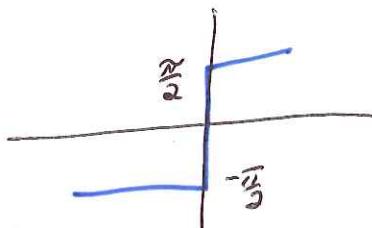
Differentiator

$$y(t) = \frac{d^2}{dt^2} x(t)$$

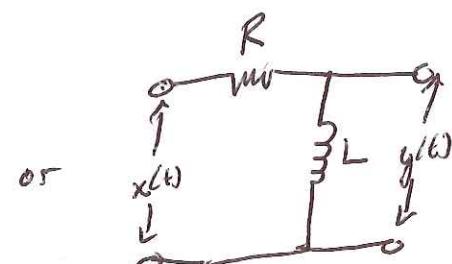
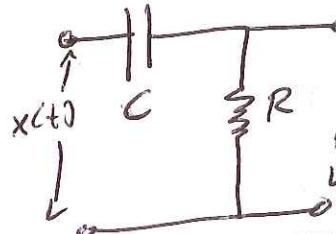
$$|H(j\omega)|$$



$$\neq H(j\omega)$$

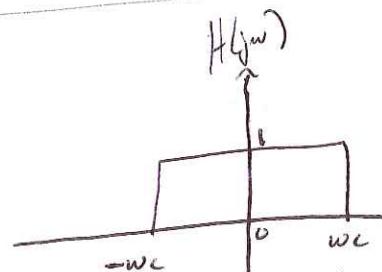


Analog :-



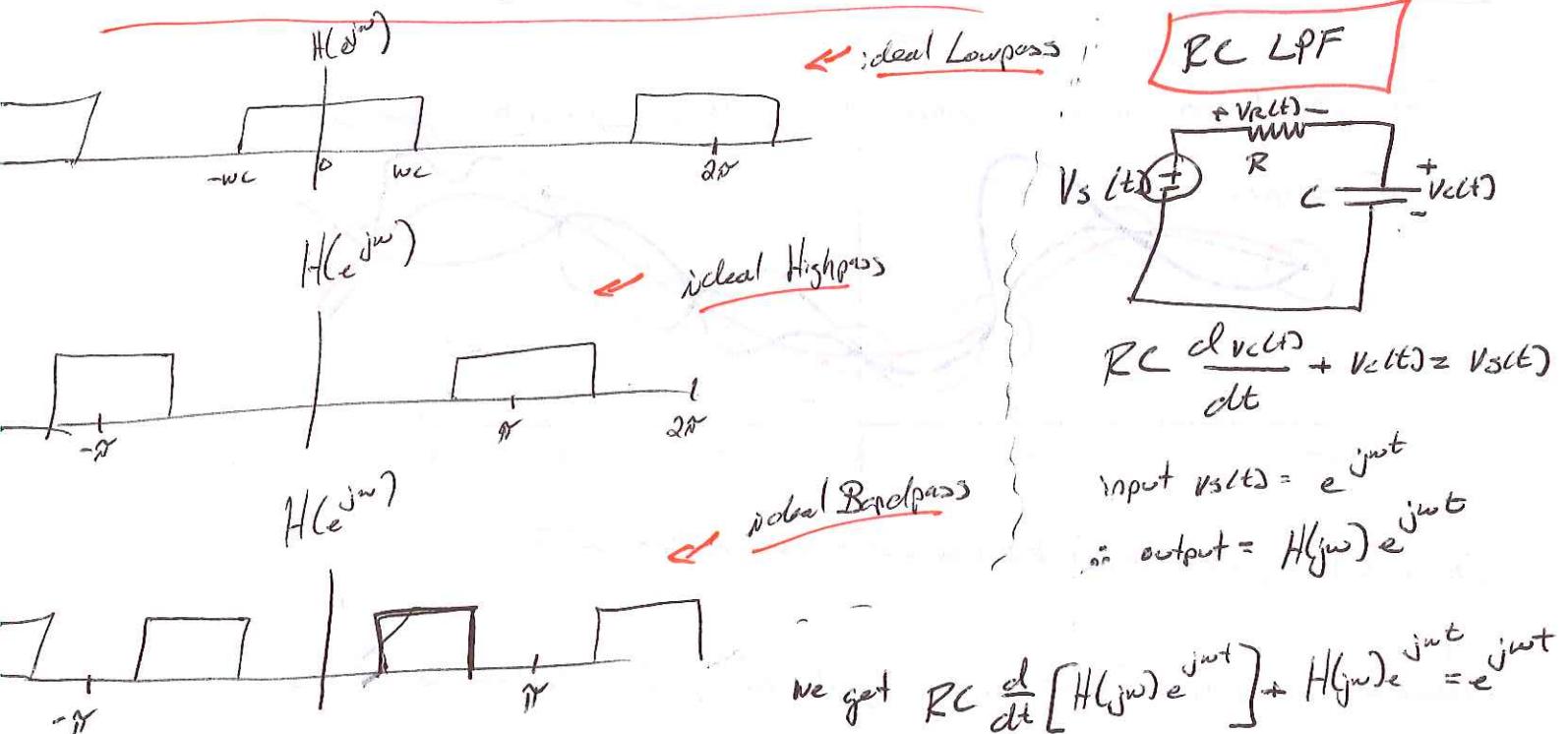
## Frequency Selective Filters

$$H(j\omega) = \begin{cases} 1, & |\omega| \leq \omega_c \\ 0, & |\omega| > \omega_c \end{cases}$$



← Stopband → ← Passband → ← Stopband →

## Discrete time Frequency Selection Filters



$$\text{We get } RC \frac{d}{dt} [H(j\omega) e^{j\omega t}] + H(j\omega) e^{j\omega t} = e^{j\omega t}$$

$$\text{Rearrange } H(j\omega) e^{j\omega t} = \frac{1}{1+RCj\omega} e^{j\omega t}$$

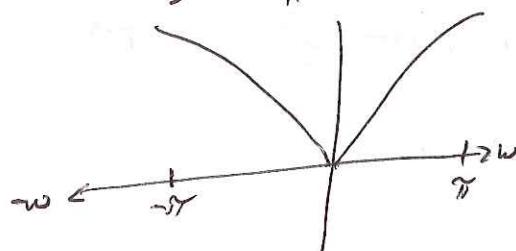
$$\text{or } H(j\omega) = \frac{1}{1+RCj\omega}$$

Filters are usually **NOT** ideal!

$$\therefore h(t) = \frac{1}{RC} e^{-t/RC} u(t)$$

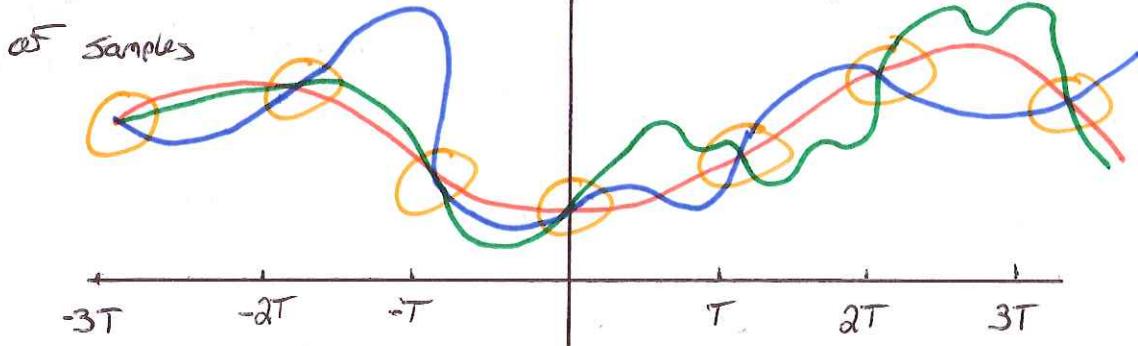
Simple high pass filter  $\rightarrow$

$$|H(e^{j\omega})|$$



## Sampling Theorem as it relates to the frequency domain

In general, an infinite number of signals can generate a given set



Three CT signals with identical values at integer multiples of  $T$

However, we will see that if a signal is band-limited — i.e. if its Fourier Transform is zero outside a finite band of frequencies — and if the samples are taken sufficiently close together in relation to the highest frequency present in the signal, then the samples can uniquely specify the signal, and we can reconstruct it perfectly!

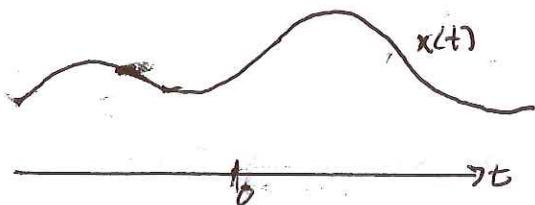
### Impulse Train Sampling

Signal  $x(t)$  is sampled by a periodic pulse train  $p(t)$  "sampling function" with period  $T$  as the sampling period  $p(t)$ ,  $\omega_s = 2\pi/T$  as the sampling frequency

$$\therefore x_p(t) = x(t)p(t) \quad \text{where } p(t) = \sum_{n=-\infty}^{\infty} g(t-nT) \quad \text{and as we know}$$

$$p(t) \quad \text{from before } x(t)g(t-t_0) = x(t_0)g(t-t_0)$$





$$x_p(t) = \sum_{n=-\infty}^{\infty} x(nT) g(t - nT)$$

By the multiplication Property

$$X_p(j\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\theta) P(j(\omega - \theta)) d\theta$$

also, we know the FT of a pulse train

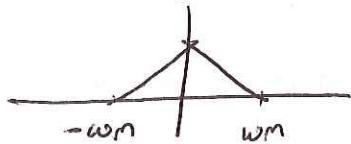
$$P(j\omega) = \frac{2\pi}{T} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s)$$

and since convolution with an impulse just shifts a signal [i.e.  $X(j\omega) * \delta(\omega - \omega_0) = X(j(\omega - \omega_0))$ ]

$$\text{we have } X_p(j\omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\omega - k\omega_s))$$

That is,  $X_p(j\omega)$  is a periodic function of  $\omega$  consisting of a superposition of shifted replicas of  $X(j\omega)$ , scaled by  $\frac{1}{T}$

If  $X(j\omega)$  → where  $\omega_m$  is maximum frequency content and we have

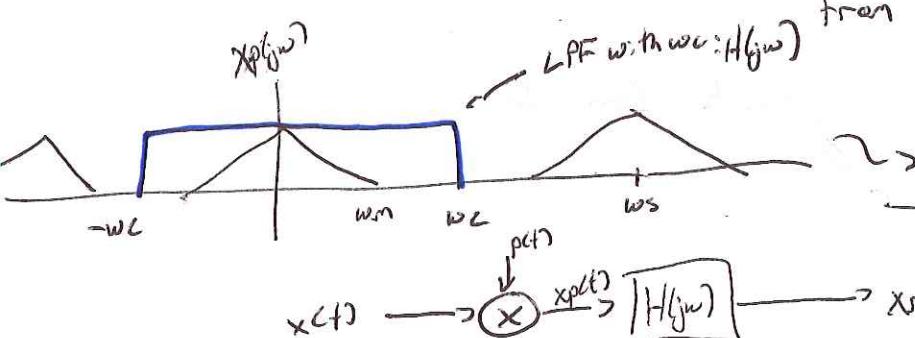
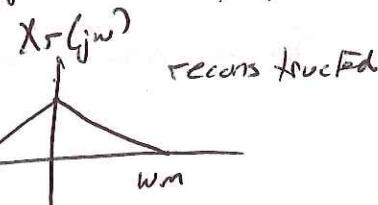
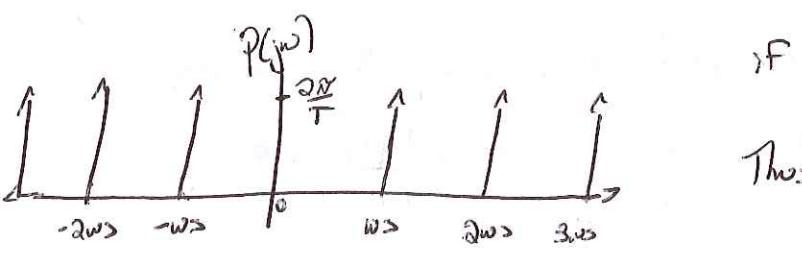


$\omega_m < (\omega_s - \omega_m)$  or equivalently,  $\omega_s > 2\omega_m$ , we have no overlap

If  $\omega_s < 2\omega_m$ , there is overlap.

Thus,  $\omega_s > 2\omega_m$ ,  $x(t)$  can be exactly recovered

from  $x_p(t)$  by means of lowpass filtering

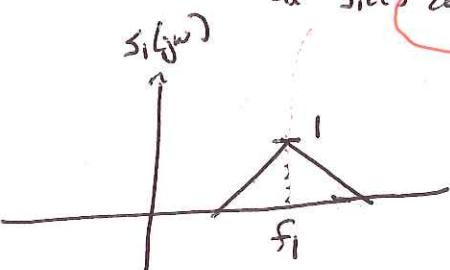


under sampling results in aliasing

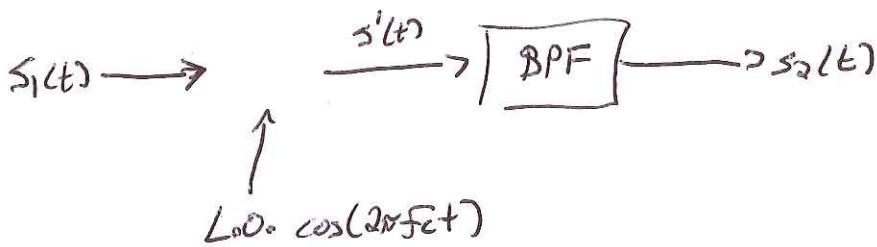
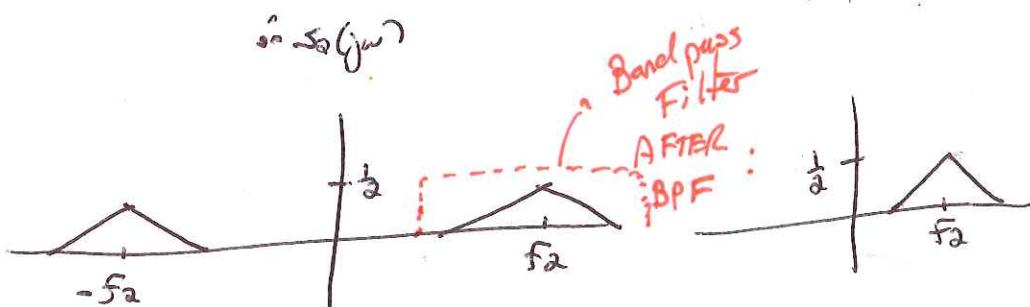
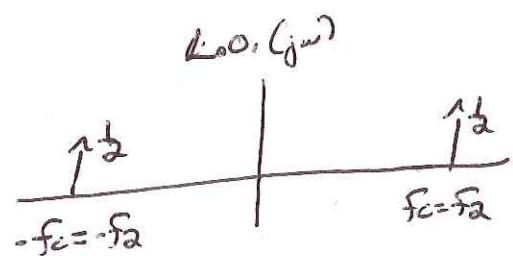
## Frequency Translation

have a signal  $s_1(t)$  with center frequency at  $f_1$ , I want to move that signal to have its center at  $f_2$ , we will translate the frequency or modulate the wave  $s_1$  to  $s_2$  such that

$$s_2 = s_1(t) \cos(2\pi f_c t) \rightarrow \text{LO. Local Oscillator}$$



Multiplication in time  
is convolution in frequency

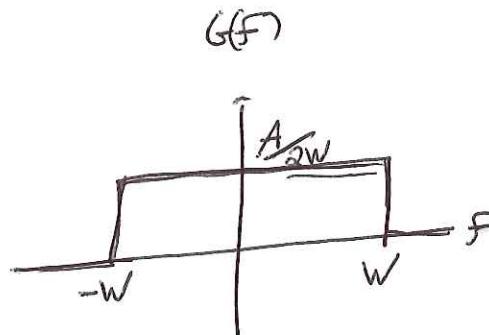
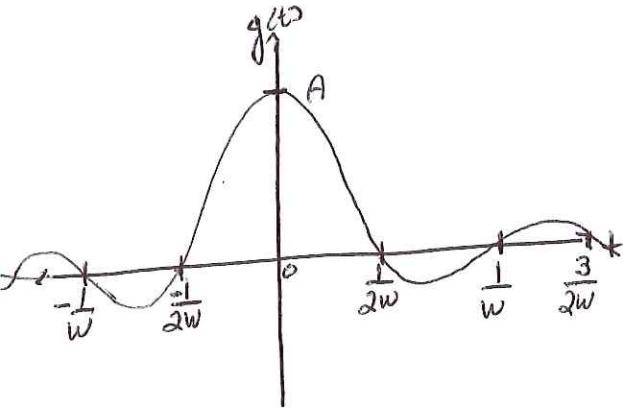


works in a similar way for translating down in Frequency.

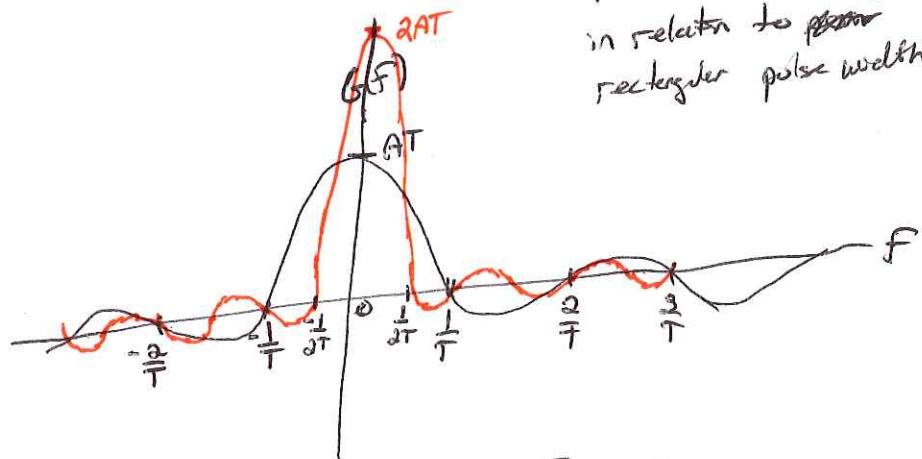
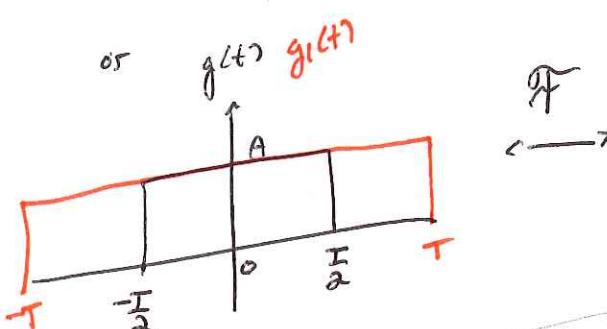
This is how we take low, baseband signals + upconvert them to higher frequencies for other applications like WiFi, Radio, Fiber, Satellite communications, etc.

Source converts up↑ , sink converts down ↓

Sinc in time:  $g(t) = A \text{sinc}(2\omega t)$



First zero crossing  
in relation to power  
rectangular pulse width



Wider in time is narrower in frequency  
wider in frequency is narrower in time



$$m(t) = s_1(t) * s_2(t)$$

Multiplication in time is convolution in frequency

$$s_1(t) = \cos(2\pi 400t)$$

$$s_2(t) = \cos(2\pi 600t)$$

scaled version

$$M(f)$$

$$f_1 + f_2 = 1000 \text{ Hz}$$

$$-f_1 - f_2 = 1000 \text{ Hz}$$

Filters in relation to Decibels. Logarithmic unit for comparing ratio of physical values/quantity usually an intensity or power ratios.

Generally needs a reference level. Let's say whimpers are 30 decibels. + a car's power is 90 decibels, 60dB more is really means  $10^{\frac{60}{20}} \rightarrow 10^6$  times more as loud!

1 million times louder!

