

# Fundamentals of Communications

## Engineering

Department of Communications Engineering, College of Engineering, University of Diyala, 2016-2017

**Class:** Second Year

**Instructor:** Dr. Montadar Abbas Taher

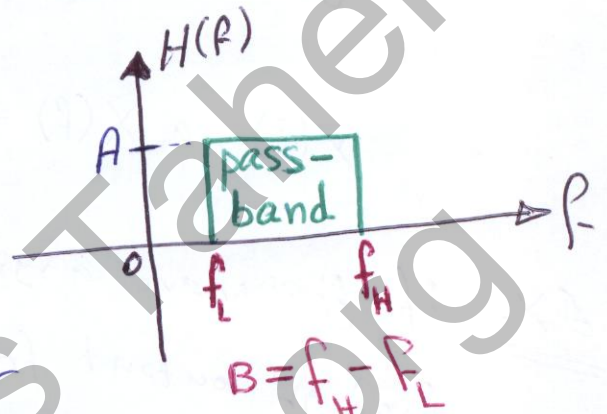
**Room:** Comm-02

**Lecture: 13**

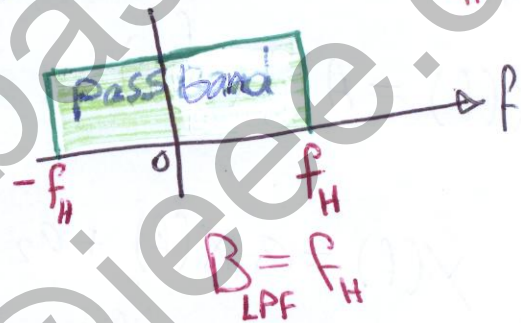
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Filters :- The filter is an electronic circuit, which passes interested frequency band and rejects the unwanted frequency, which is out of the interested band.

\* if  $f_L = 0$  &  $f_H \neq \infty$   
finite

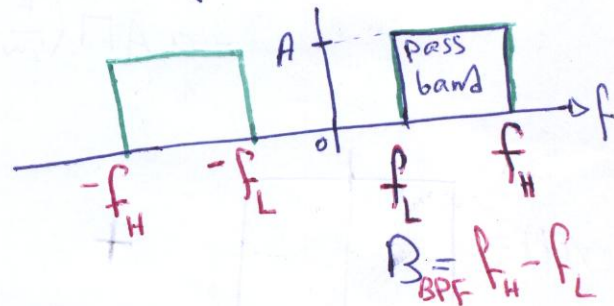


then we have Low-Pass-Filter (LPF)



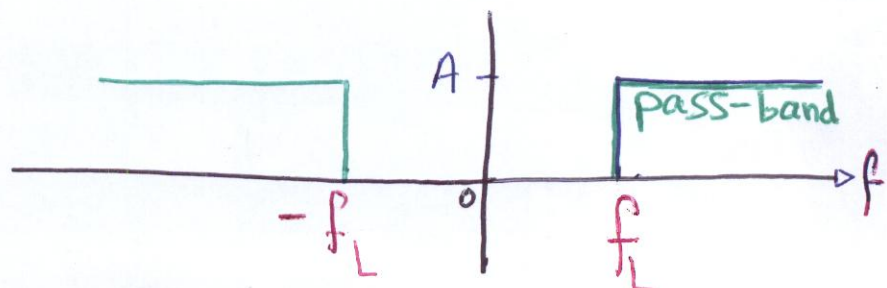
\* if  $f_L > 0$  &  $f_H \neq \infty$   
finite

then we have Band-Pass-Filter (BPF)



\* if  $f_L > 0$  &  $f_H = \infty$

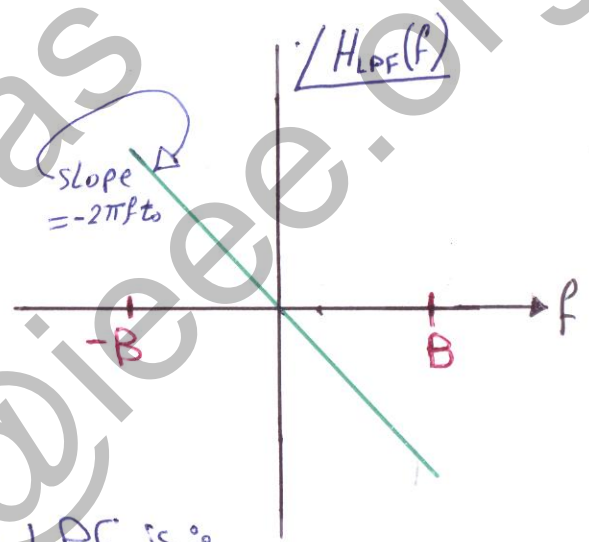
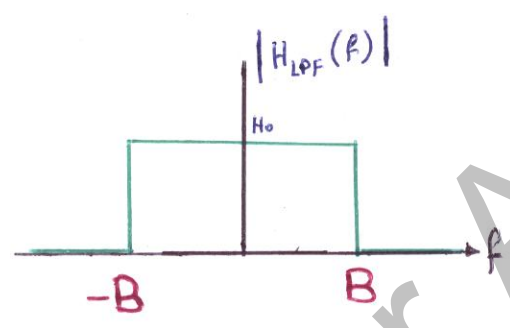
then we have High-pass-Filter (HPF)



# Low Pass Filter (LPF) [ideal]

$$H_{LPF} = H_0 \Pi\left(\frac{f}{2B}\right) e^{-j2\pi f t_0}$$

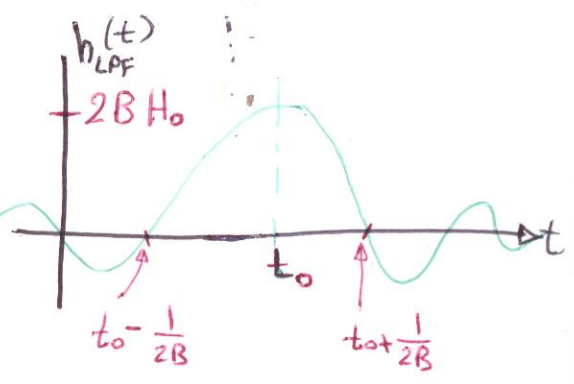
where  $H_0$  is constant,  $H_0$  stands for the passband gain.



\* The impulse response of an ideal LPF is:

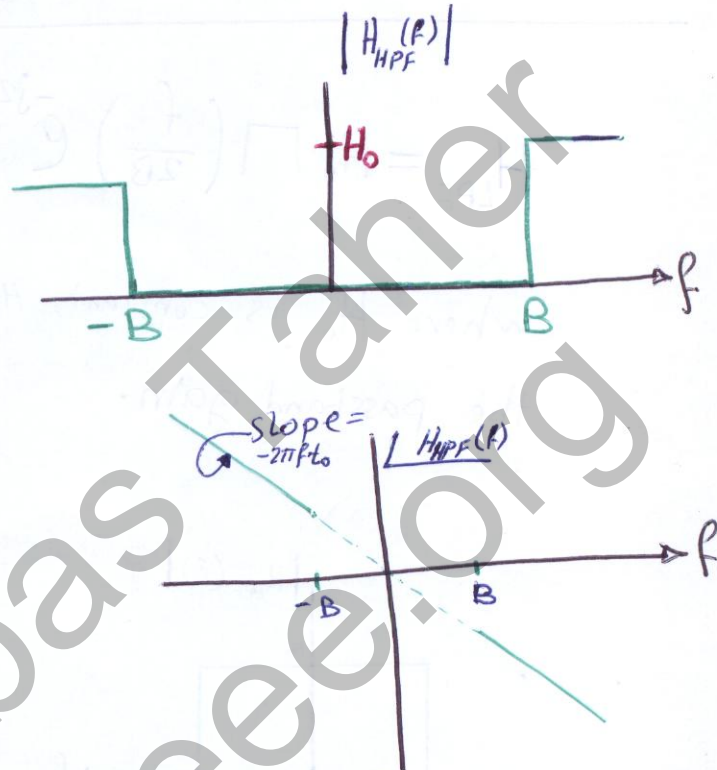
$$h_{LPF}(t) = 2B H_0 \text{sinc}[2B(t - t_0)]$$

therefore, they call the sinc function as the filter function or the interpolation function.



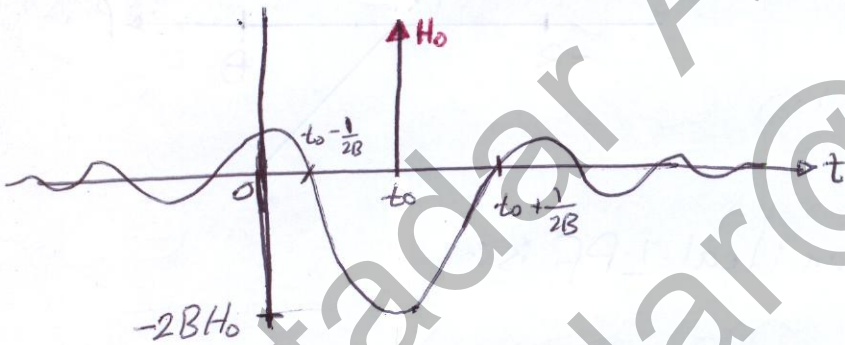
# High Pass Filter (HPF) [ideal]

$$H_{HPF}(f) = H_0 \left[ 1 - \Pi\left(\frac{f}{2B}\right) \right] e^{-j2\pi f t_0}$$



$$h_{HPF}(t) = H_0 \left\{ \delta(t-t_0) - 2B \text{sinc}[2B(t-t_0)] \right\}$$

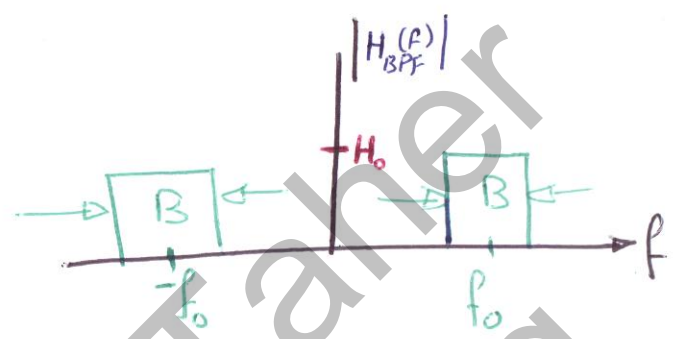
Sketch - Homework



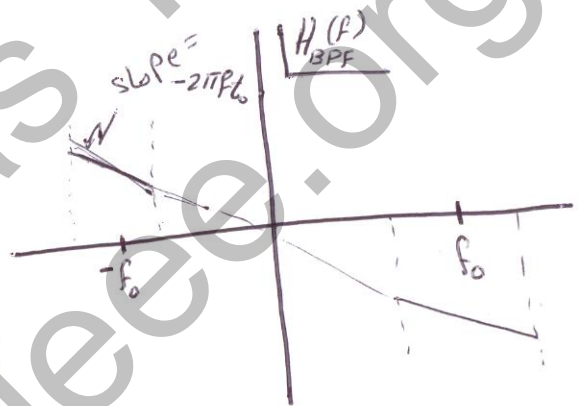
# Band Pass Filter (BPF) [ideal]

$$H_{BPF}(f) = [H_L(f-f_0) + H_L(f+f_0)] e^{-j2\pi f t_0}$$

where  $H_L = H_0 \Pi\left(\frac{f}{B}\right)$

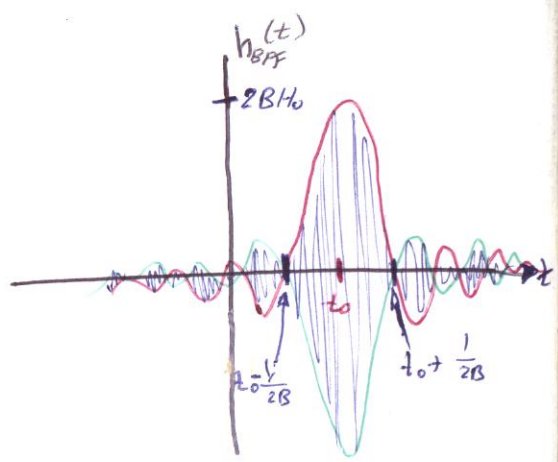


$$h_{BPF}(t) = 2h_L(t-t_0) \cos(2\pi f_0(t-t_0))$$



or

$$h_{BPF}(t) = 2BH_0 \text{sinc}[B(t-t_0)] \cos[2\pi f_0(t-t_0)]$$



# Sampling Theory

The continuous time signal  $x(t)$  can be sampled instantaneously as

$$x_s(t) = \sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s) \quad (1)$$

Where  $T_s$  is the sampling interval.

\* After Sampling

① Is it possible to recover  $x(t)$  from  $x_s(t)$ ?

② How is  $x(t)$  recovered from  $x_s(t)$ ?

\* To answer these two questions, we have to understand the uniform sampling theorem.

Before we go ahead, Lets have some important knowledge about signals and their types.

\* In real world, there are continuous-time signals and discrete-time signals

\* Due to the recent advance development in digital technology:-

- Inexpensive,
- Light weight,
- programmable, and
- easily reproducible

discrete-time systems are available.

→ therefore, the processing of discrete-time signals is more flexible and is also preferable to processing than the continuous-time ones.

\* Thus, we should be able to convert a continuous-time signals into discrete-time signals.

\* Sampling process, is a tool that can be used to convert the continuous-time signal to a discrete-time signal and vice-versa.

## Sampling Theory ∞∞

A band limited signal of bandwidth  $w$  Hz sampled at frequency  $f_s$ , can be reconstructed from the sampled version if  $f_s \geq 2w$ .

## Proving Sampling Theory ∞∞

$$x_s(t) = \sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s) \quad \text{--- (1)}$$

$$T_s = \frac{1}{f_s} = \text{Sampling interval.}$$

Since we know  $\delta(t - nT_s)$  is zero everywhere except at  $t = nT_s$

$$x_s(t) = \sum_{n=-\infty}^{\infty} x(t) \delta(t - nT_s) = x(t) \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \quad \text{--- (2)}$$

$$X_s(f) = X(f) * \left[ f_s \sum_{n=-\infty}^{\infty} \delta(f - nf_s) \right] \quad \text{--- (3)}$$



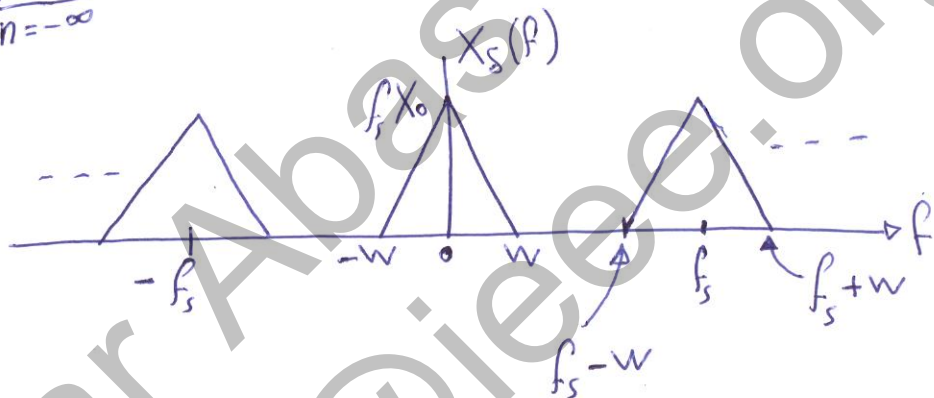
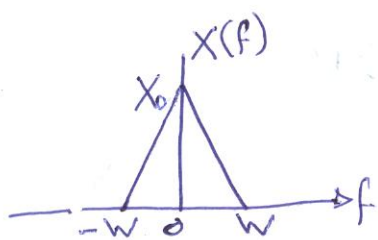


Since  $X(f) * \delta(f - nf_s) = \int_{-\infty}^{\infty} X(u) \delta(f - u - nf_s) du$

$$\therefore X(f) * \delta(f - nf_s) = X(f - nf_s) \quad \text{--- (4)}$$

by the sifting property of the delta function,

$$X_s(f) = f_s \sum_{n=-\infty}^{\infty} X(f - nf_s) \quad \text{--- (5)}$$



Sampling produces a periodic repetition of  $X(f)$  in the frequency domain with a spacing  $f_s$ .

① If  $f_s < 2w$ , the separate terms **overlap**,  $x(t)$  can not be recovered from  $x_s(t)$ . (**aliasing**)

② If  $f_s = 2w$ , the separate terms touch each other,  $x(t)$  can be recovered from  $x_s(t)$  using a sharp LPF.

③ If  $f_s > 2w$ ,  $x(t)$  can be recovered from  $x_s(t)$  using LPF.

End of Prove \_\_\_\_\_

\* The reconstruction filter should have a bandwidth  $B$  in the range

$$W < B < (f_s - W)$$

where  $W$  : Bandwidth of the signal  $x(t)$

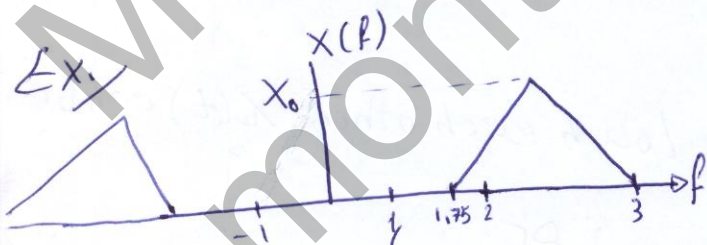
$f_s$  : Sampling frequency

NOTE : Nyquist frequency is the  $2W$  frequency.

### Sampling a Bandpass Signal

if a signal has a spectrum  $W$  Hz and upper frequency  $f_u$  Hz, then a rate  $f_s$  at which the signal can be sampled is  $2f_u/m$  Hz, where

$m$  is the largest integer not exceeding  $f_u/W$ .



$$f_u = 3 \text{ Hz}$$

$$W = 3 - 1.75 = 1.25 \text{ Hz}$$

$$m \leq \left\lceil \frac{f_u}{W} \right\rceil = \frac{3}{1.25} = 2.4$$

$$m = 2$$

$$f_s = \frac{2f_u}{m} = \frac{2 \times 3}{2} = 3 \text{ Hz}$$

note that if the signal is a Lowpass signal then,  $f_s = 6 \text{ Hz}$ .

EX. A continuous time signal  $x(t) = 8 \cos(200\pi t)$ . Determine:-

- ① Minimum sampling rate, (Nyquist rate to avoid aliasing).
- ② If sampling frequency  $f_s = 400$  Hz, what is the discrete-time signal  $x(n)$  or  $x(nT_s)$  obtained after sampling?
- ③ If sampling frequency  $f_s = 150$  Hz, what is the discrete-time signal  $x(n)$  or  $x(nT_s)$  obtained after sampling?
- ④ what is the frequency  $0 < f < \frac{f_s}{2}$  that yields samples identical to those obtained in part ③?

Solution  $\circ \circ$   $x(t) = 8 \cos(2\pi 100t) \rightarrow f = 100$  Hz.

①  $f_s = 2f = 2 \times 100 = 200$  Hz.

②  $f_s = 400$  Hz  $\rightarrow x(nT_s) = x\left(\frac{n}{f_s}\right) = 8 \cos\left(2\pi n \frac{100}{400}\right) = 8 \cos\left(\frac{2\pi n}{4}\right) = 8 \cos\left(\frac{\pi n}{2}\right)$ .

$\circ \circ$   $x(n) = x(nT_s) = 8 \cos\left(\frac{n\pi}{2}\right)$ .

③  $f_s = 150$  Hz  $\rightarrow x(nT_s) = x\left(\frac{n}{f_s}\right) = 8 \cos\left(2\pi n \frac{100}{150}\right) = 8 \cos\left(\frac{4\pi n}{3}\right) = 8 \cos\left(\frac{6\pi n}{3} - \frac{2\pi n}{3}\right)$

$x(n) = 8 \cos\left[\left(2\pi - \frac{2\pi}{3}\right)n\right] = 8 \cos\left(\frac{2\pi n}{3}\right)$

④  $f_s = 150$  Hz  $\rightarrow x(nT_s) = x\left(\frac{n}{f_s}\right) = 8 \cos\left(2\pi f \frac{n}{f_s}\right) = 8 \cos\left(2\pi f n / 150\right)$

From part ③  $\frac{2\pi n}{3} = \frac{2\pi f n}{150} \rightarrow f = \frac{150}{3} = 50$  Hz

$\circ \circ$   $x_4(t) = 8 \cos(100\pi t)$ .

NOTE  $\circ \circ$   $x(t) = 8 \cos(2\pi 100t)$ , if  $f_s = 150$ , then aliis frequency  $f_a = 150 - 100 = 50$  Hz will appears.

Ex. An analog signal is expressed by the equation  $x(t) = 3 \cos(500\pi t) + 10 \sin(3000\pi t) + 10 \sin(3000\pi t) - \cos(1000\pi t)$ . Calculate the Nyquist rate for this signal.

Solu.

$$x(t) = 3 \cos(500\pi t) + 10 \sin(3000\pi t) - \cos(1000\pi t)$$

$$x(t) = 3 \cos(\omega_1 t) + 10 \sin(\omega_2 t) - \cos(\omega_3 t)$$

$$\omega_1 = 2\pi f_1 = 500\pi \rightarrow f_1 = 25 \text{ Hz}$$

$$\omega_2 = 2\pi f_2 = 3000\pi f_2 \rightarrow f_2 = 1500 \text{ Hz}$$

$$\omega_3 = 2\pi f_3 = 1000\pi f_3 \rightarrow f_3 = 500 \text{ Hz}$$

Largest frequency is  $f_2 = 1500 \text{ Hz}$

$$\therefore f_s = 2f_2 = 3000 \text{ Hz}$$

Ex. Find the Nyquist rate and the Nyquist interval for the signal  $x(t) = \frac{1}{2\pi} \cos(4000\pi t) \cos(1000\pi t)$ .

Solu. Since  $2 \cos A \cos B = \cos(A+B) + \cos(A-B)$

$$\therefore x(t) = \frac{1}{2\pi} \cdot \frac{2}{2} \cos(4000\pi t) \cos(1000\pi t)$$

$$= \frac{1}{4\pi} [2 \cos(4000\pi t) \cos(1000\pi t)]$$

$$= \frac{1}{4\pi} [\cos(4000\pi t + 1000\pi t) + \cos(4000\pi t - 1000\pi t)]$$

$$= \frac{1}{4\pi} [\cos(5000\pi t) + \cos(3000\pi t)]$$

$$\therefore \omega_1 = 2\pi f_1 = 5000\pi f_1 \rightarrow f_1 = 2500 \text{ Hz}$$

$$\omega_2 = 2\pi f_2 = 3000\pi f_2 \rightarrow f_2 = 1500 \text{ Hz}$$

$$\therefore f_s = 2f_1 = 5000 \text{ Hz}$$

$$T_s = \frac{1}{f_s} = \frac{1}{5000} = 0.2 \times 10^{-3} \text{ sec.}$$

# Reconstruction Filter (LPF) ∞ ∞

\* reconstruction filter is a low-pass-filter (LPF), which is also known as **interpolation filter**.

\* The process of reconstructing the signal from the discrete-time version is called interpolation.

- Assuming an ideal LPF :

$$H(f) = H_0 \Pi\left(\frac{f}{2B}\right) e^{-j2\pi f t_0}$$

$W \ll B \ll f_s - W$

$$\therefore Y(f) = f_s H_0 X(f) e^{-j2\pi f t_0}$$

and the time-domain becomes :-

$$y(t) = f_s H_0 x(t - t_0)$$

reconstructed signal.

OR 
$$y(t) = \sum_{n=-\infty}^{\infty} x(nT_s) h(t - nT_s)$$

$$y(t) = 2BH_0 \sum_{n=-\infty}^{\infty} x(nT_s) \text{sinc}\left[2B(t - t_0 - nT_s)\right]$$

But  $B = \frac{f_s}{2}$ ,  $H_0 = T_s$ , and let  $t_0 = 0$  for simplicity

$$y(t) = \sum_n x(nT_s) \text{sinc}(f_s t - n)$$

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