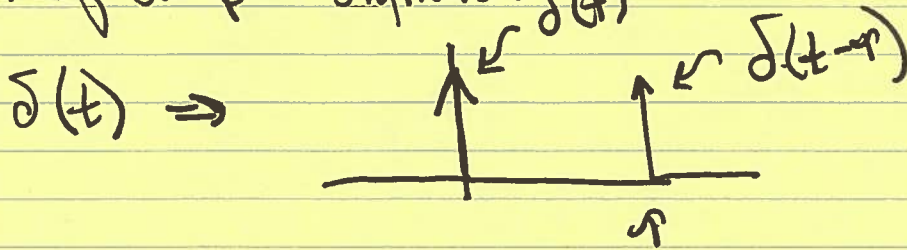


Lecture 2

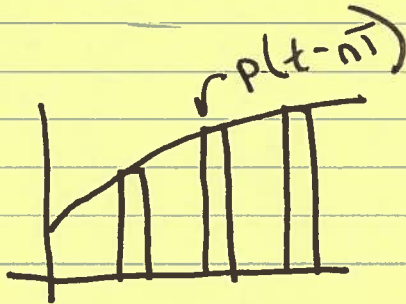
(1)

Spectrum of Sampled Signals: $\delta(t)$



$$\hat{x}(t) = \sum_{n=-\infty}^{\infty} x(nT) \delta(t - nT)$$

in practice:



$$x_{\text{flat}}(t) = \sum_{n=-\infty}^{\infty} x(nT) p(t - nT)$$

Spectrum:

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

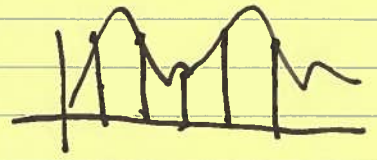
$$= \int_{-\infty}^{\infty} \sum_{n=-\infty}^{\infty} x(nT) \delta(t - nT) e^{-j2\pi f t} dt$$

$$= \sum_{n=-\infty}^{\infty} x(nT) \int_{-\infty}^{\infty} \delta(t - nT) e^{-j2\pi f t} dt$$

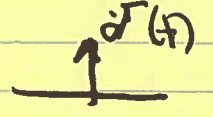
$$= \sum_{n=-\infty}^{\infty} x(nT) e^{-j2\pi f n T} \quad T = \frac{1}{f_s}$$

Discrete-Time Fourier Transform

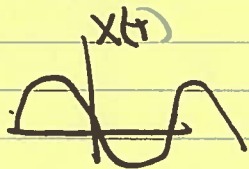
$$\hat{x}(t) = x(t) \underbrace{\sum_{n=-\infty}^{\infty} \delta(t-nT)}_{s(t)} = x(t)s(t)$$



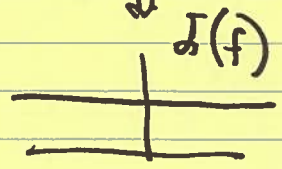
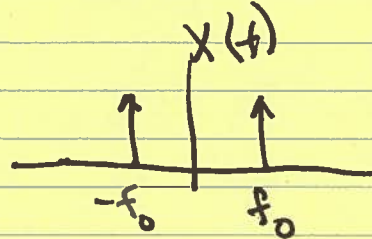
$$s(t) = \sum_{n=-\infty}^{\infty} \delta(t-nT) = \frac{1}{T} \sum_{m=-\infty}^{\infty} e^{j2\pi m f_s t}$$



Sinewave:



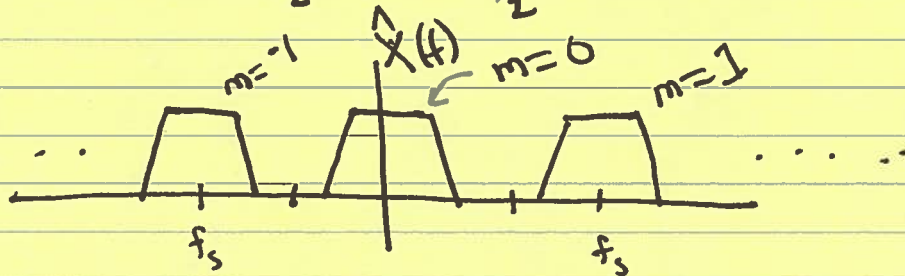
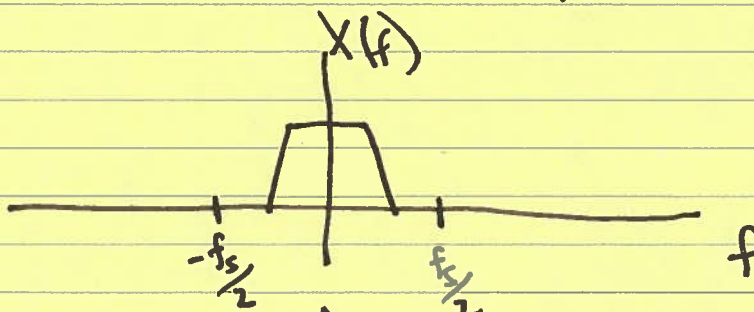
=>



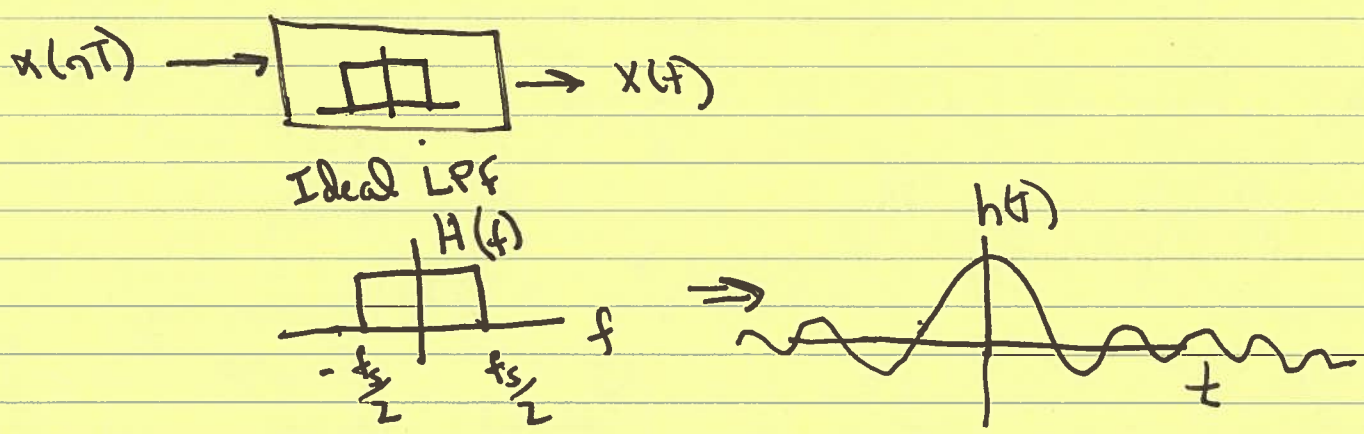
$$\hat{x}(t) = x(t)s(t) = \frac{1}{T} \sum_{m=-\infty}^{\infty} x(t) e^{j2\pi m f_s t}$$

apply modulation theorem

$$\hat{X}(f) = \mathcal{F}\{\hat{x}(t)\} = \frac{1}{T} \sum_{m=-\infty}^{\infty} X(f - m f_s)$$

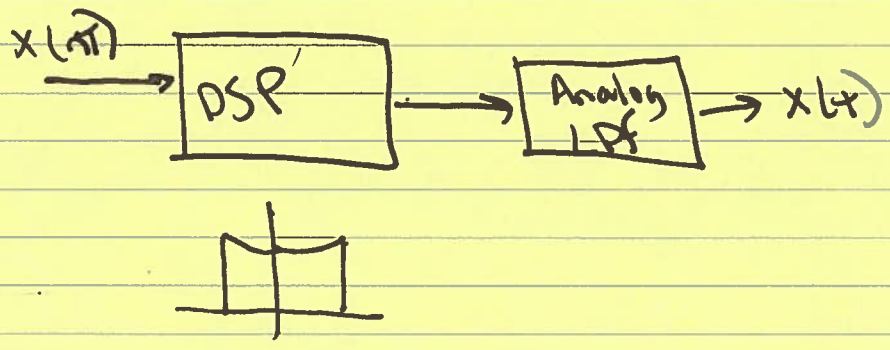
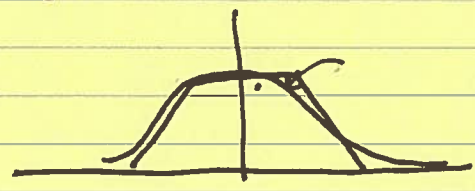


Reconstruction:

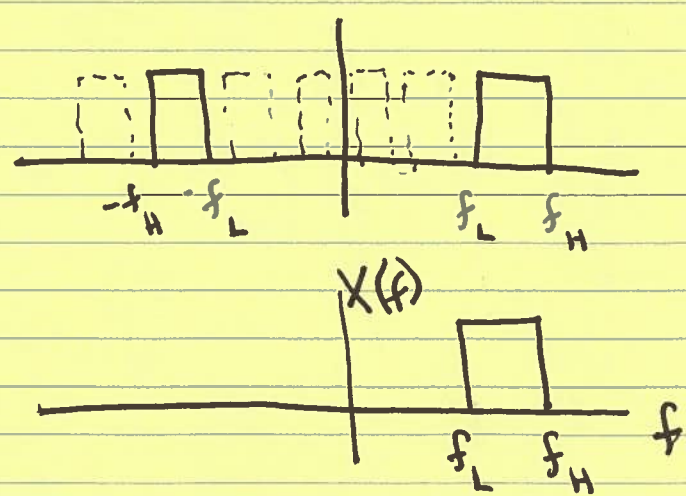


Can we build an ideal LPF? No!

Use an approximation:



Bandlimited Sampling Theorem:



$$f_s \geq 2f_H$$

$$\propto 2(f_H - f_L)$$

Complex!