

# VE216 Recitation Class 7

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*VE216 SU20 TA Group*

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# Overview

- 1 Chapter 6: Filtering
- 2 Chapter 7: Sampling
  - Sampling Theorem
- 3 Conclusion

# Before we start

Uniquely  
characterize  
LTI systems  
 $h[n]$

Decompose signals:  
LTI, Freq. domain

- To me, concept of convolution, FS, FT are “theoretically inspiring”
- Now we turn to applications like

- {
  - filtering (Chap. 6)
  - sampling (Chap. 7)
  - communication (Chap. 8)

which are “practically inspiring”

- What's even more amazing is that all these applications depend on only **two** properties:

- Convolution Property:  $\text{v}[n] \rightarrow [h[n]] \rightarrow x[n]$

F-Hanging:

$$f_1(t) * f_2(t) \xleftrightarrow{\mathcal{F}} F_1(\omega) \cdot F_2(\omega)$$

- Time-domain Multiplication:

Sampling

$$f_1(t) \cdot f_2(t) \xleftrightarrow{\mathcal{F}} \frac{1}{2\pi} F_1(\omega) * F_2(\omega)$$

$x[n] \rightarrow [h[n]] \rightarrow$

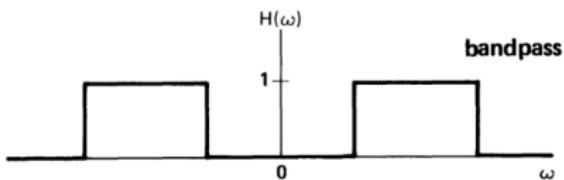
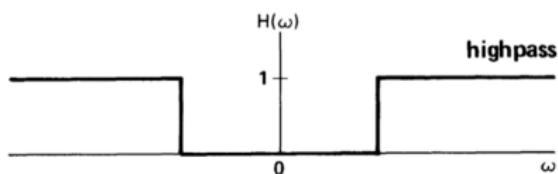
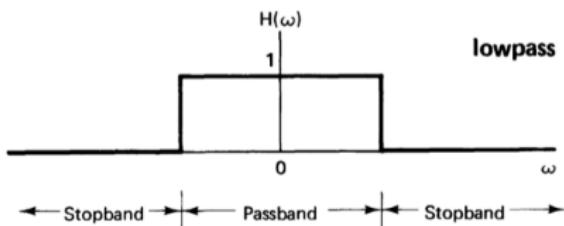
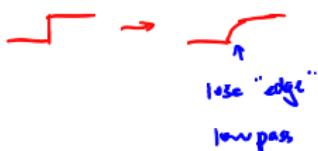
# Filtering

$$x(t) \rightarrow [LTI \text{ LIO}] \rightarrow y(t = h(t) + x(t))$$

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

- Convolution Property:  $h(t) * x(t) \xleftrightarrow{\mathcal{F}} H(\omega)X(\omega)$

**Task 1:**



- LTI systems can be viewed as “filters” in frequency domain

# Exercise - Filtering

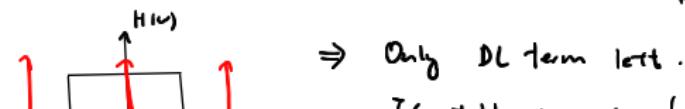


## Example

The signal  $x(t) = \sum_{n=-\infty}^{\infty} \text{rect}(t - 1/2 - 2n)$  is passed through a filter with frequency response  $H(\omega) = 3 \text{ rect}(\omega/\pi)$ . Determine the output signal  $y(t)$ .

(Selected from Midterm Exam 2 of Summer 2014)

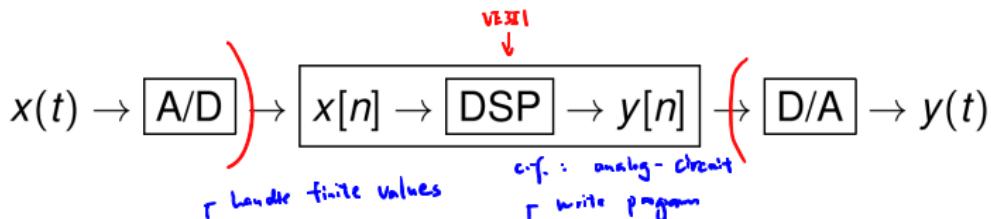
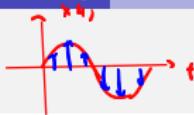
$$\begin{aligned} \text{FS for Periodic Signal: } X(\omega) &= \sum_{k=-\infty}^{\infty} 2\pi \cdot C_k \cdot \delta(\omega - k\omega_0) \\ &= \sum_k 2\pi \cdot C_k \cdot \delta(\omega - k\pi). \end{aligned}$$



$$\text{FS table: } \omega_0 = \frac{f}{2}$$

$$\Rightarrow Y(\omega) = 2\pi \times \frac{1}{2} \times \delta(\omega) \times 3 = 3\pi \delta(\omega) \Leftrightarrow y(t) = \frac{3}{2}$$

# Overview of DSP



- Why DSP? Only way possible, Computers, low price, etc
- Here we only focus on the left and right part. (DSP will be discussed in VE351)
- We want to show that processing the discrete-time signal (the samples) is equivalent to processing the continuous-time signal (the initial signal). - i.e.: we don't need the all the infinite values  $\Rightarrow$  therefore DSP with confidence
- Time-domain Multiplication:

$$f_1(t) \cdot f_2(t) \xleftrightarrow{\mathcal{F}} \frac{1}{2\pi} F_1(\omega) * F_2(\omega)$$



# Sampling Theorem

## Sampling Theorem

- Equally Spaced Samples of  $x(t)$

$$x(nT) \quad n=0, \pm 1, \pm 2, \dots$$

- $x(t)$  Band limited  
 $\Leftrightarrow$  composed of only low freq.

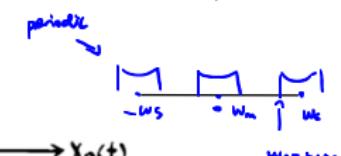
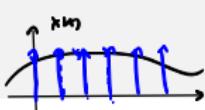
$$\bar{x}(w) = 0 \quad |w| > \omega_m$$

f sample fast enough

$$\text{If } \frac{2\pi}{T} \triangleq \omega_s > 2\omega_m$$

Then  $x(t)$  uniquely  
recoverable  
↑  
recover from samples

Later, when see  $\tilde{x}(t-nT)$ , think of sampling



$$X_p(w) = x(t) \sum_{n=-\infty}^{+\infty} \delta(t-nT) \quad | \quad x(t) \cdot \delta(t-nT) = x(nT) \cdot \delta(t-nT)$$

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

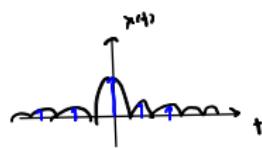
$$\bar{x}_p(w) = \frac{1}{2\pi} [\bar{x}(w) + P(w)] \quad | \quad x(n) \Rightarrow \delta(t-nT) = x(t)$$

∴  $P(w) = \frac{2\pi}{T} \sum_{k=-\infty}^{+\infty} \delta(w - k \frac{2\pi}{T})$  see previous or whole 7.1 notes

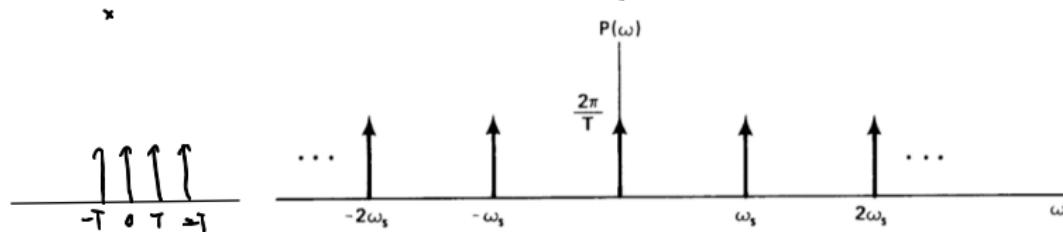
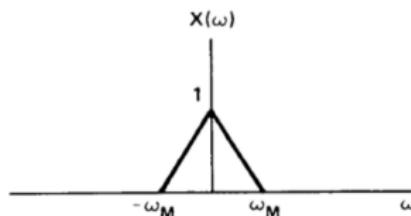
$$\bar{x}_p(w) = \frac{1}{T} \sum_{k=-\infty}^{+\infty} \bar{x}(w - k \frac{2\pi}{T}) \quad | \quad \frac{1}{T} (\bar{x}(w) + x(w - \omega_s) + \dots)$$

## Recoverable

$$xM) = \sin^2(\omega_m)$$

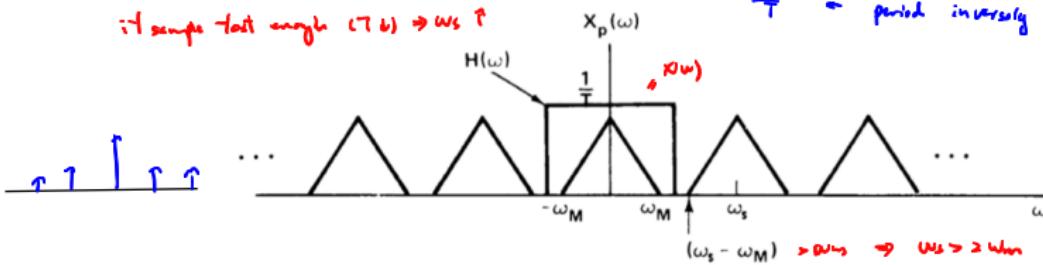


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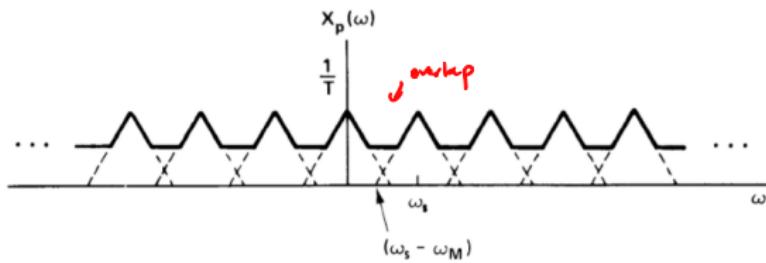
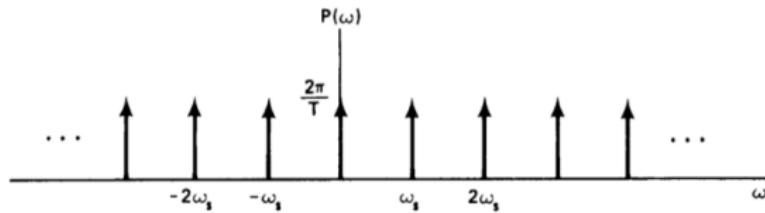
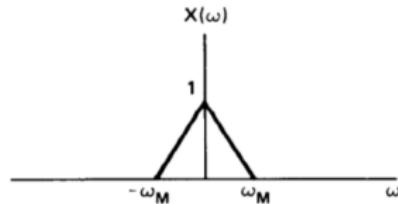


if sample last enough ( $T_b$ )  $\Rightarrow$  ws?

- period inversely related in time to frequency

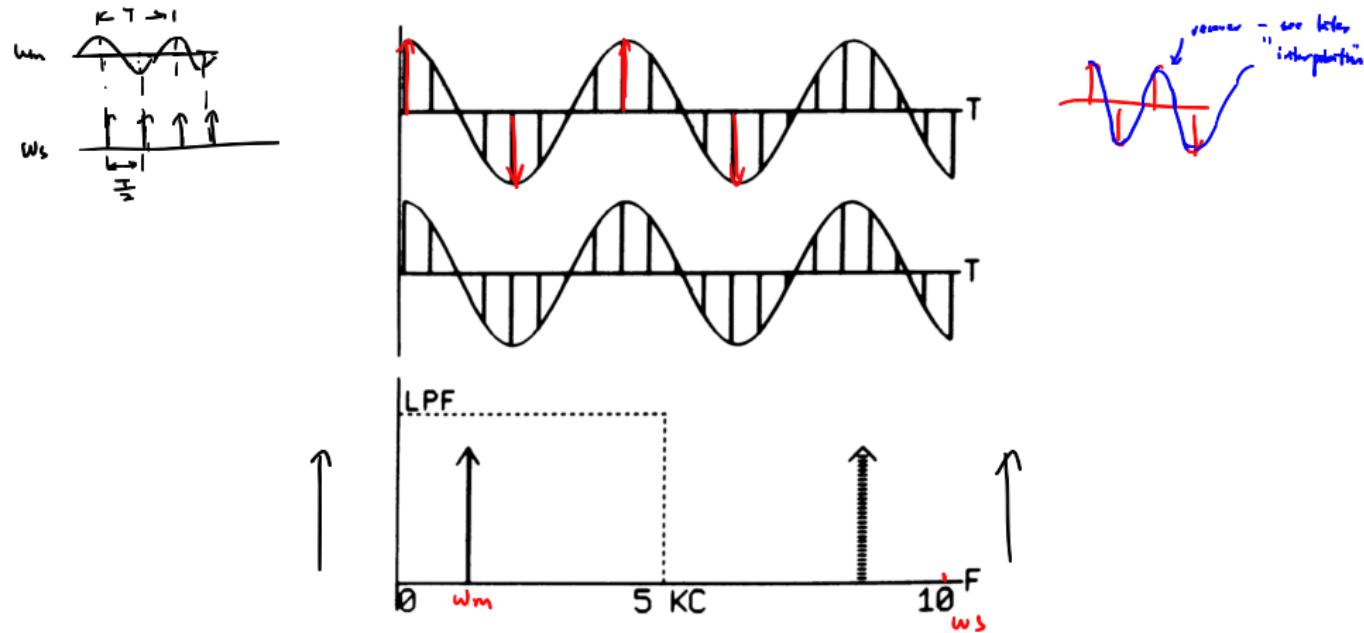


# Aliasing



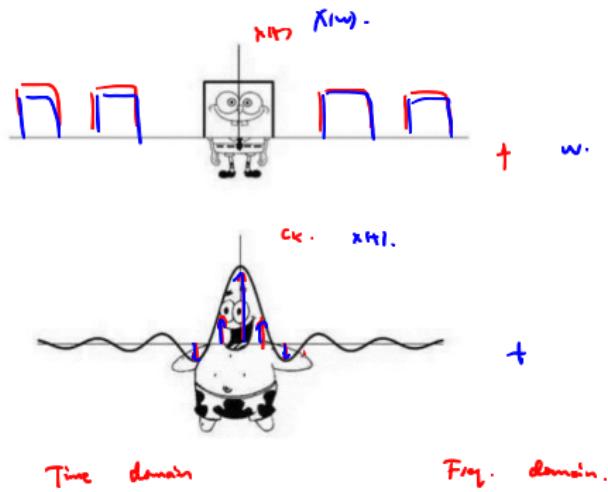
# Nyquist rate - interpretation

$$w_s \geq w_m : X_s(t) = \sum_n X(n \cdot T_s) \cdot \delta(t - nT_s)$$



- Interpretation: at least 2 samples in a period *No, as long as x\_m is band-limited*
- Information lost during sampling? Consider const. signal. *- 2 sample enough*

# Another way to understand Sampling: relation with FS



chap.4.

Patrick Star  $\xleftarrow{FT}$  SpongeBob SquarePants ↔ lowpass filter

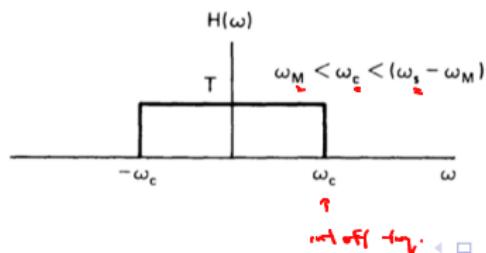
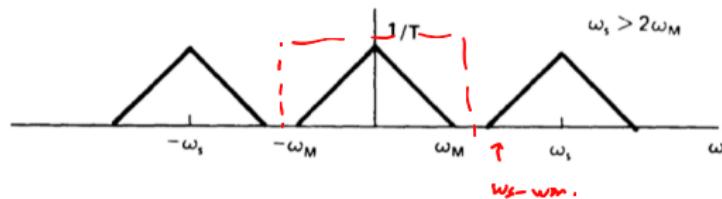
~~chap.3~~ Periodic SpongeBob SquarePants  $\xleftarrow{FS/FT}$  Samples of Patrick Star

~~chap.7~~ Samples of Patrick Star  $\xleftarrow{FT}$  Periodic SpongeBob SquarePants

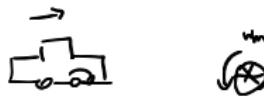
# Sampling & Reconstruction

$p(t) = \sum_{n=-\infty}^{+\infty} \delta(t - nT)$

w<sub>m.</sub> vs w<sub>s</sub> vs w<sub>c</sub>



# Conclusion



$\omega_s \geq 2\omega_n ?$

- The concept of sampling itself is motivating - consider eye (watching the wheels) and ear (ultrasonic)
- Sampling is closely related with reconstruction, which will be the focus of next week.
- For sampling-related problems, I prefer to view them graphically (often in freq. domain).
- Questions in HW5 are very interesting (or, tricky) - not much math involved, rather requires understanding

# The End