

# Signal Analysis & Communication ECE 355

## Ch. 7-5. Sampling of DT Signals

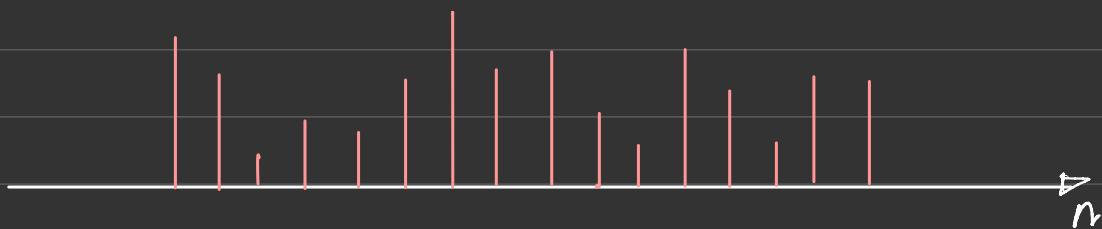
Lecture 31

27-11-2023



## Ch. 7.5. SAMPLING OF DISCRETE TIME SIGNALS

- Consider a DT Sig.  $x_{[n]}$



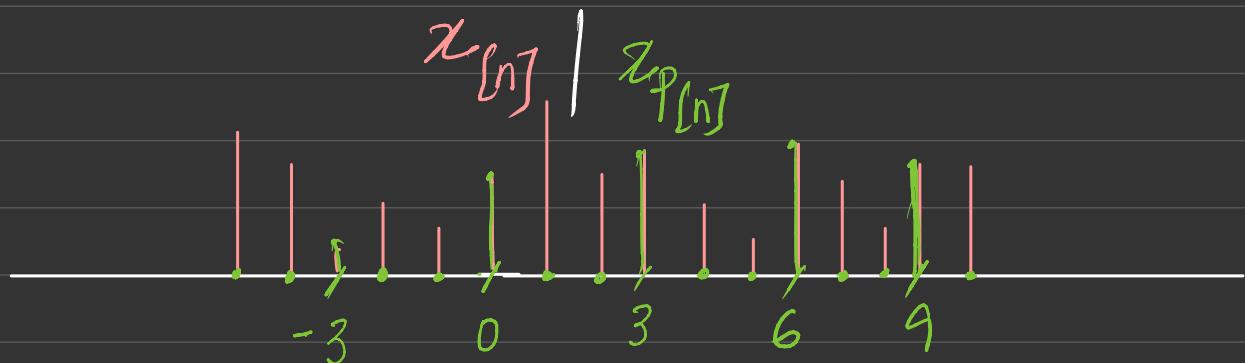
- In analogy with CT Sampling, sampling of DT sig. can be performed as:



$$p_{[n]} = \sum_{k=-\infty}^{\infty} \delta_{[n-kN]}$$

- Sampling period =  $N$

- For example, the above sig. for  $N=3$  would be sampled as:

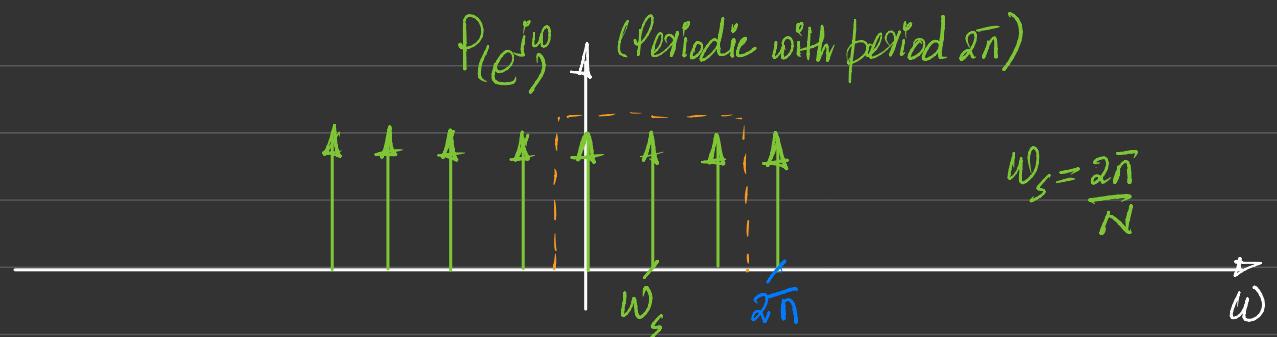


$$x_{p[n]} = \sum_{k=-\infty}^{\infty} x_{[kN]} \delta_{[n-kN]} \quad - \textcircled{1}$$

$$= \begin{cases} x_{[n]}, & \text{if } n=kN \text{ for some } k \in \mathbb{Z} \\ 0, & \text{o/w} \end{cases}$$

- Previously, the DTFT of DT impulse train is (Ref. Sec. 24)

$$P(e^{j\omega}) = \frac{1}{N} \sum_{k=-\infty}^{\infty} \delta(\omega - k\omega_s) \quad \text{periodic with period } 2\pi.$$



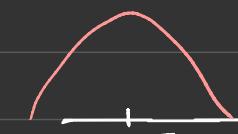
- Accordingly, the spectra of the sampled sig. is given as:  
(DTFT of eqn ① - Multiplication in time domain is convolution in freq.-domain)

$$\begin{aligned} X_P(j\omega) &= \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j(\omega-\theta)}) P(e^{j\theta}) d\theta \\ &= \frac{1}{2\pi} \int_{0}^{2\pi} X(e^{j(\omega-\theta)}) \underbrace{\frac{1}{N} \sum_{k=0}^{N-1} \delta(\theta - k\omega_s)}_{\text{exists at } \theta = k\omega_s} d\theta \end{aligned}$$

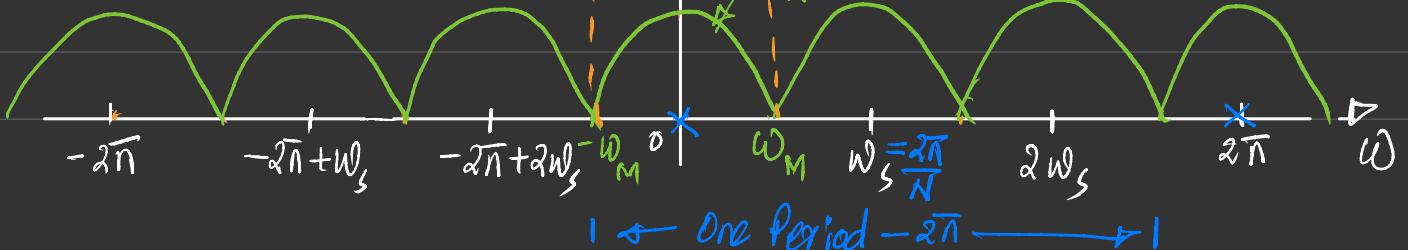
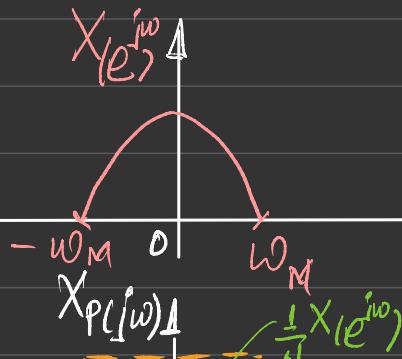
since the integral is  
from 0 to  $2\pi$

$$= \frac{1}{N} \sum_{k=0}^{N-1} X(e^{j(\omega - k\omega_s)})$$

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



$N=3$



- Again, no aliasing if  $\omega_s > 2\omega_M$ .
- Reconstruction by ideal lowpass filtering.

$$H(e^{j\omega}) = \begin{cases} N & , |\omega| < \omega_c, \text{ Periodic} \\ 0 & , \text{ otherwise} \end{cases} \quad (\text{as is discrete in time})$$

&  $\omega_M < \omega_c < \omega_s - \omega_M$

Usually  $\omega_c = \frac{\omega_M}{2}$

- The corresponding time-domain sig. is:

$$h[n] = N \frac{\omega_c}{\pi} \frac{\sin(\omega_c n)}{\omega_c n}$$

- Therefore, the reconstructed sig. after filtering is:

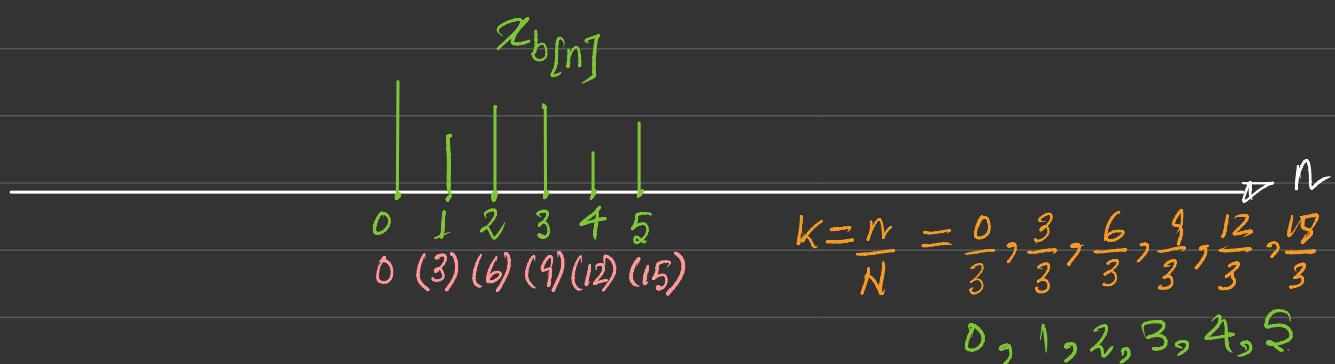
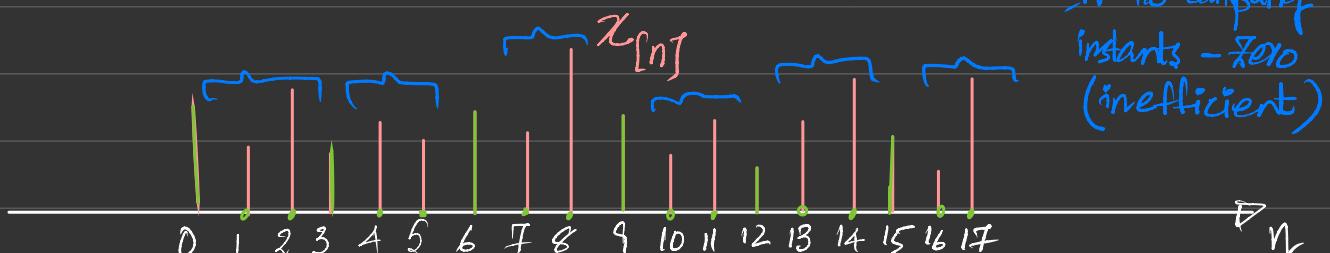
$$\begin{aligned} x_r[n] &= x_p[n] * h[n] \\ &= \sum_{m=-\infty}^{\infty} x_p[m] h[n-m] \quad \left| \begin{array}{l} \text{"Conv. in DT"} \\ \text{=} \end{array} \right. \\ &= \sum_{k=-\infty}^{\infty} x_p[kN] N \frac{\omega_c}{\pi} \frac{\sin(\omega_c(n-kN))}{\omega_c(n-kN)} \end{aligned}$$

## Downsampling ("Decimation")

- In many applications of DT Sampling, such as filter design & implementation or in communication, it is inefficient to represent, transmit, or store the sampled sequence  $x_p[n]$  directly, since in b/w the sampling instants,  $x_p[n]$  is known to be zero.

- Thus, the sampled sequence is typically replaced by a new seq.  $x_{b[n]}$ , which is simply every  $N$ th value of  $x_p[n]$ ; that is,

$$x_{b[n]} = x_{p[nN]} = x_{[nN]}$$



- The operation of extracting every  $N$ th sample is commonly referred to as "decimation".

- To determine the effect of decimation in freq. domain, let's determine the relationship b/w  $X_b(e^{j\omega})$  &  $X_p(e^{j\omega})$ .

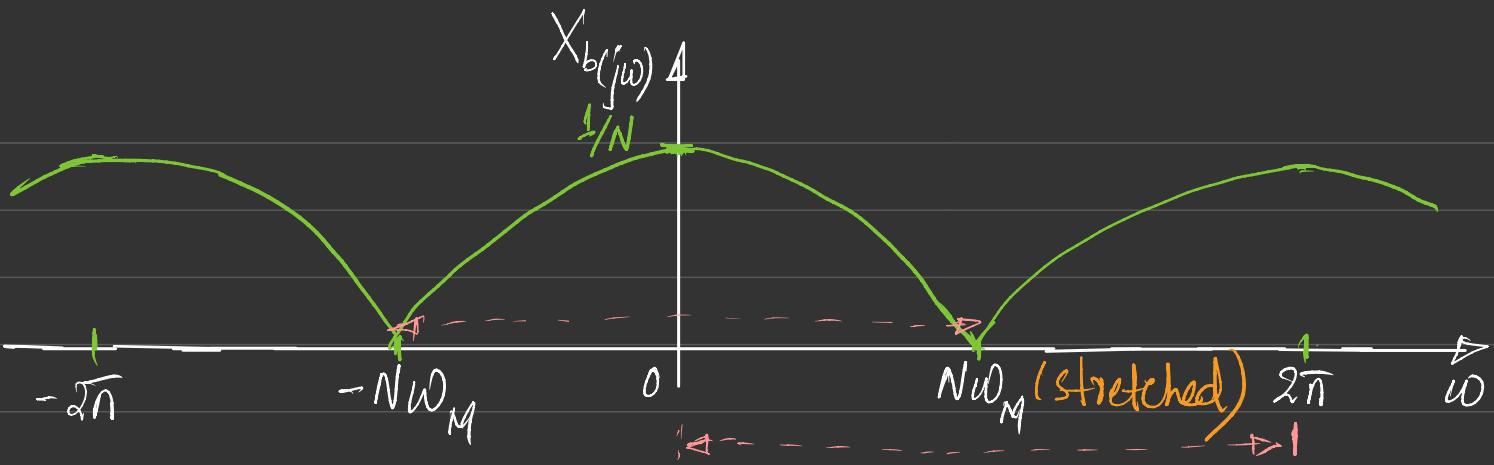
$$X_b(e^{j\omega}) = \sum_{k=-\infty}^{\infty} x_b[k] e^{-jk\omega}$$

$$= \sum_{n=-\infty}^{\infty} x_p[n] e^{-jn\omega \cancel{\frac{1}{N}}}$$

replace  $k$  by  $n$

$$= X_p(e^{-j\omega/N})$$

Stretch  $X_p(j\omega)$  by a factor of 'N'.  
 $\omega_s$  stretched to  $\bar{\omega}$



- If no aliasing in  $X_p(e^{j\omega})$ , then no loss of information in downsampling.

## UPSAMPLING ("Interpolation")

- For some applications, it is useful to convert a sequence to a higher equivalent sampling rate, referred to as "Interpolation".
- Reverse of "Decimation"



- If no aliasing, then  $x_r[n] = x_b[n]$

## RESAMPLING

- Up-sample then downsample
- Example:

- Convert 800x600 image to 1280x960 (1.6-times)

$\frac{8}{5} = 1.6$  ✓  
Not integer.

- Therefore we first upsample by 8, then downsample by 5

$$\frac{8}{5} = 1.6 \quad \checkmark$$