

Signal Analysis & Communication ECE 355

Ch. 7-5. Sampling of DT Signals

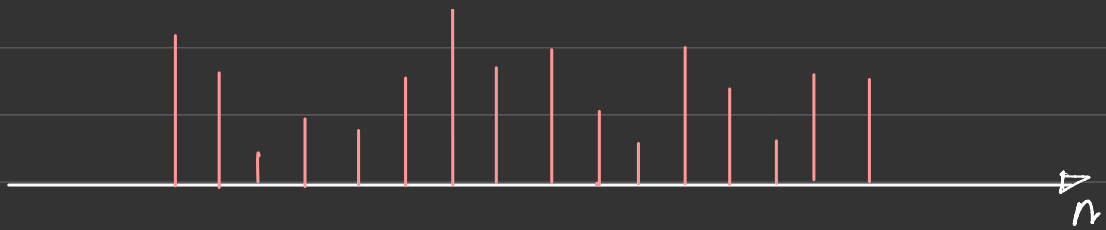
Lecture 31

27-11-2023

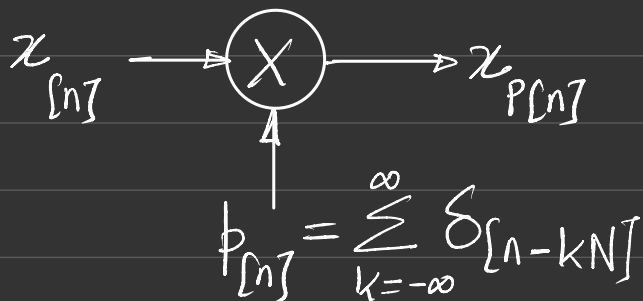


Ch. 7.5. SAMPLING OF DISCRETE TIME SIGNALS

- Consider a DT sig. $x[n]$

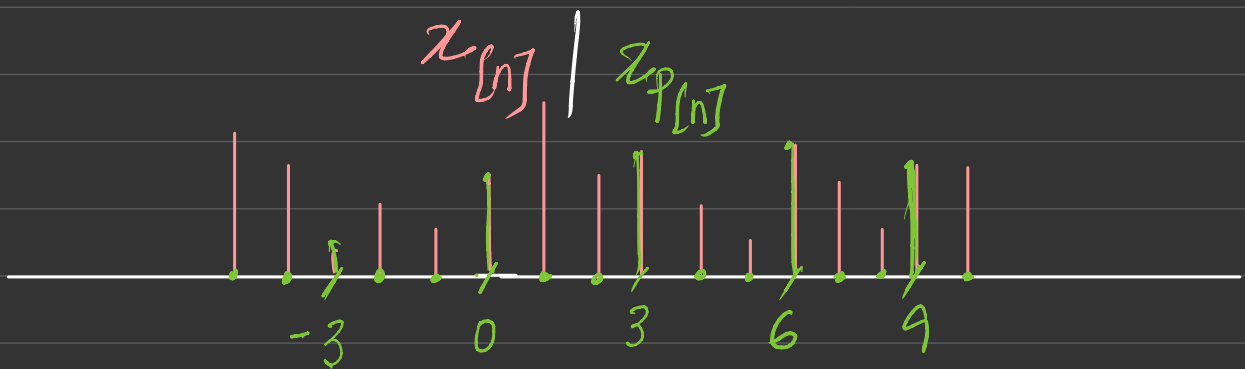


- In analogy with CT Sampling, sampling of DT sig. can be performed as:



- Sampling period = N

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



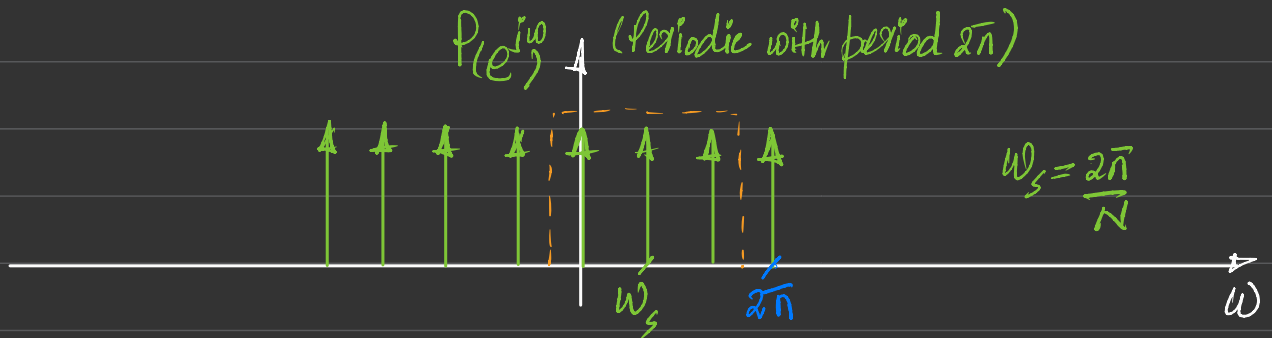
$$x_p[n] = \sum_{k=-\infty}^{\infty} x[kN] \delta[n - kN] \quad - (1)$$

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

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

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

periodic with period 2π .



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

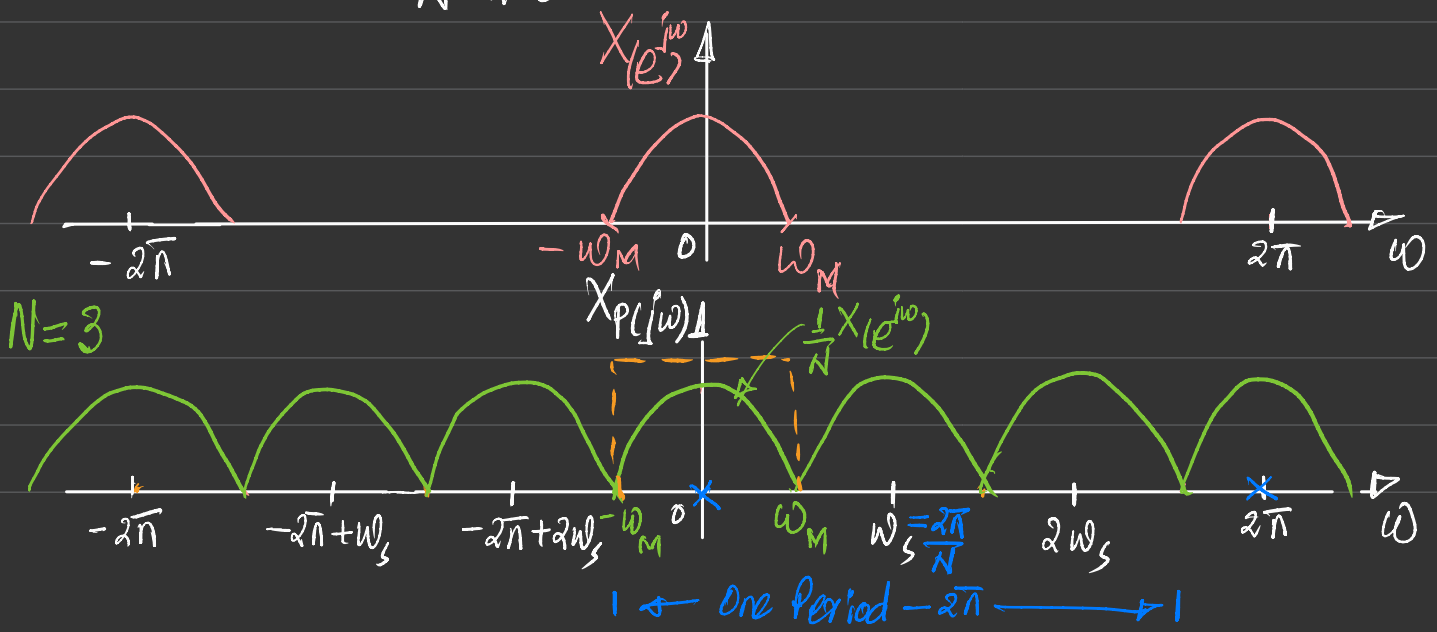
$$X_P(j\omega) = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j(\omega-\theta)}) P(e^{j\theta}) d\theta$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j(\omega-\theta)}) \frac{2\pi}{N} \sum_{k=0}^{N-1} \delta(\theta - k\omega_s) d\theta$$

since the integral is from 0 to 2π

exists at $\theta = k\omega_s$.

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



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

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

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

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

- The corresponding time-domain sig. is:

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

- Therefore, the reconstructed sig. after filtering is:

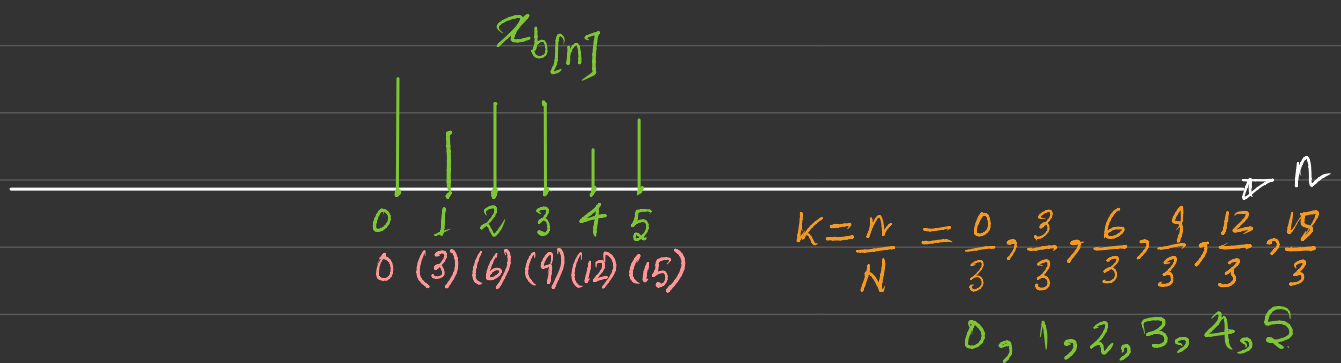
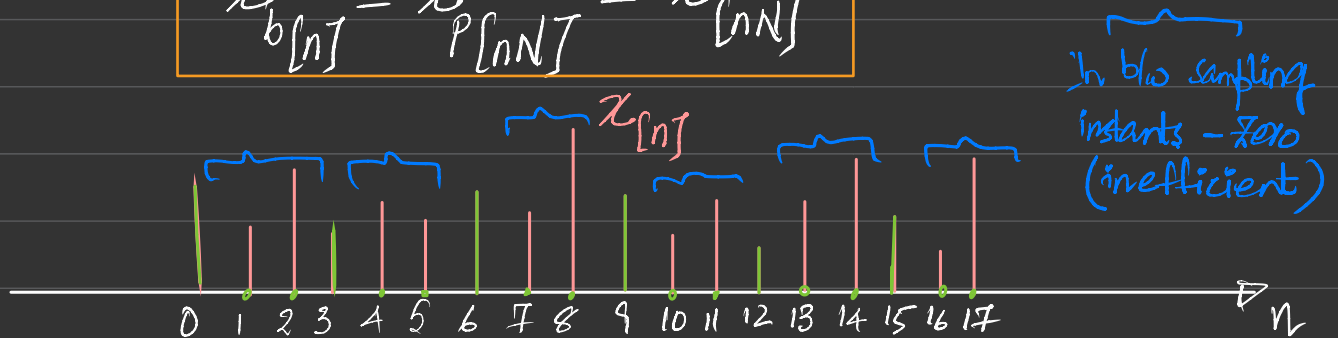
$$\begin{aligned} x_r[n] &= x_p[n] * h_r[n] \\ &= \sum_{m=-\infty}^{\infty} x_p[m] h_r[n-m] \quad \left| \text{ 'conv. in DT' } \right. \\ &= \sum_{k=-\infty}^{\infty} x_p[kN] N \frac{\omega_c}{\pi} \frac{\sin(\omega_c (n-kN))}{\omega_c (n-kN)} \end{aligned}$$

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 $X_b(e^{j\omega})$ & $X(e^{j\omega})$.

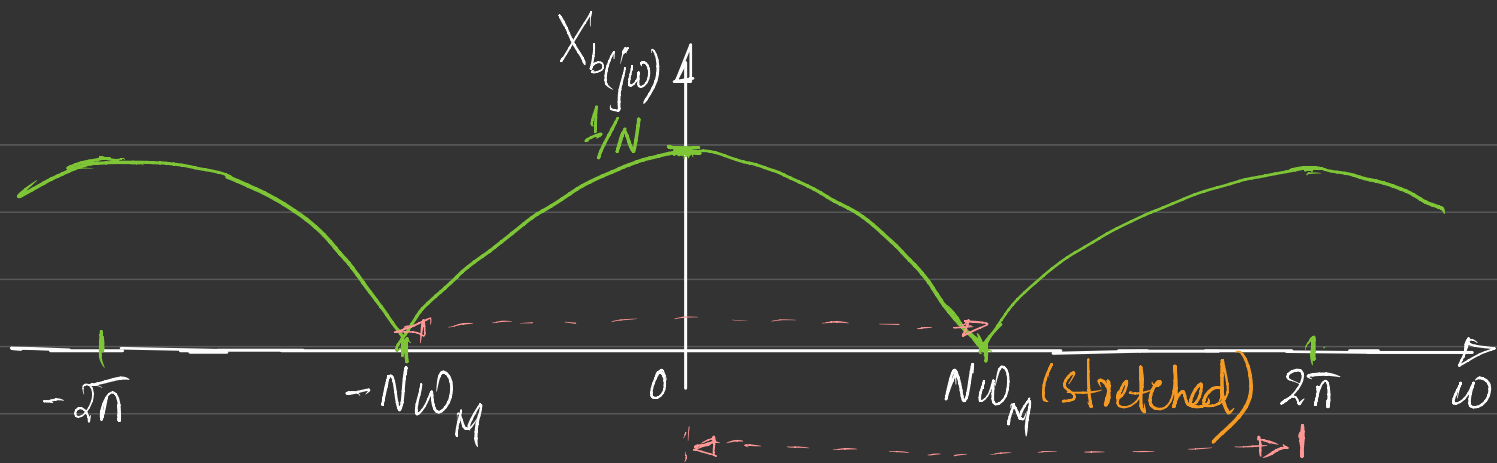
$$X_b(e^{j\omega}) = \sum_{k=-\infty}^{\infty} x_b[k] e^{-j k \omega}$$

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

replace k by n

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

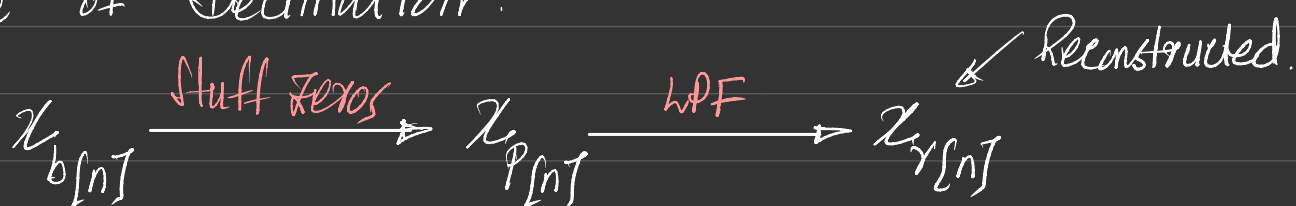
Stretch $X_p(j\omega)$ by a factor of N .
 ω_s stretched to 2π



- 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 $x_r[n] = x[n]$

RESAMPLING

- Upsample then downsample.
- Example:

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

↑
Not integer.

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

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