

Electrical Engineering
IIT Madras



EE 3101 Digital Signal Processing

BS Electronic Systems programme

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September – December 2024

Session #9

October 7, 2024



EE3101 Digital Signal ProcessingEE3101

Session 8

03-01-2018

Session 9

Outline

Last session

- Sampling
- Reconstruction
- Oversampling

✓✓ Week 1-2

Introduction to sampling - Review of Signals and Systems; Basic operations on signals
 ✓ - Discrete time complex exponentials and other basic signals – system properties (linearity, time-invariance, memory, causality, BIBO stability) -- impulse response and convolution

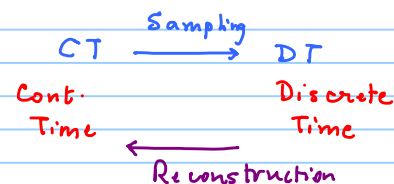
→ Week 3-4 ✓

Sampling: Impulse train sampling—relationship between impulse train sampled continuous-time signal spectrum and the DFT of its discrete-time counterpart—scaling of the frequency axis—relationship between true frequency and digital frequency—reconstruction through sinc interpolation—aliasing—effect of sampling at a discontinuous point—relationship between analog and digital sinc—effects of oversampling—discrete-time processing of continuous-time signals

Quantization ✓Reading Assignment

O&S Chapter 4: Sampling of CT signals

O&S ch 2 Sec 7: Representation of Sequences by Fourier Transforms



Week 5-6

D&S ch2 Section 7-9

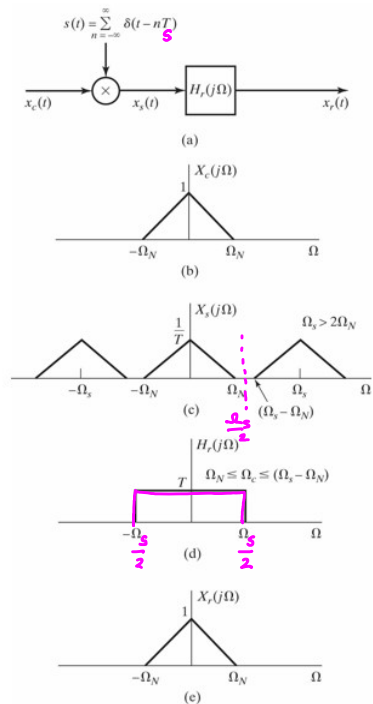
Discrete-Time Fourier Transform (DTFT): Complex exponentials as eigensignals of LTI systems—DTFT definition—inversion formula—properties—relationship to continuous-time Fourier series (CTFS)

D&S ch2 Sec 7 Representation of Sequences by Fourier Transforms

ch2 Sec 8 Symmetry Properties of Fourier Transform

ch2 Sec 9 Fourier Transform Theorems

Properties
&
Applications



Nyquist Sampling Theorem

Let $x_c(t)$ be a band-limited signal with $X_c(j\Omega) = 0$ $|\Omega| \geq \Omega_N$ rads/sec
 $x_c(t)$ is uniquely determined by its samples $x[n] = x_c(nT_s)$ $n = 0, \pm 1, \pm 2, \dots$

$$\text{if } \Omega_s = \frac{2\pi}{T_s} \geq 2\Omega_N$$

$$\text{Nyquist rate} = 2\Omega_N \quad \Omega_s = \text{rads/sec}$$

$$s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \xleftrightarrow{F} S(j\Omega) = \boxed{\frac{2\pi}{T_s}} \sum_{m=-\infty}^{\infty} \delta(\Omega - m\Omega_s) \quad \boxed{\Omega_s = 2\pi F_s = \frac{2\pi}{T_s}}$$

$$x_s(t) = x_c(t) \cdot s(t) \xleftrightarrow{F} X_s(j\Omega) = \frac{1}{T_s} \sum_{m=-\infty}^{\infty} X_c(j(\Omega - m\Omega_s))$$

Do not want copies to overlap @ $m=0, m=1$ Nyquist Theorem

$$\Omega_N \leq \Omega_s - \Omega_N \quad \boxed{2\Omega_N \leq \Omega_s} \quad \boxed{\Omega_s \geq 2\Omega_N}$$

Ideal Reconstruction Filter

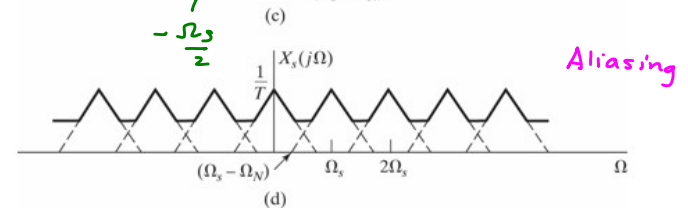
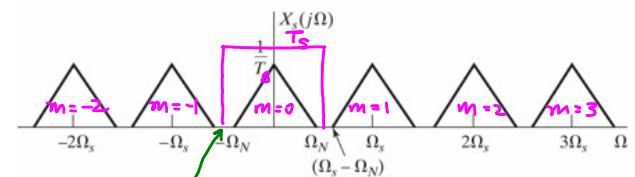
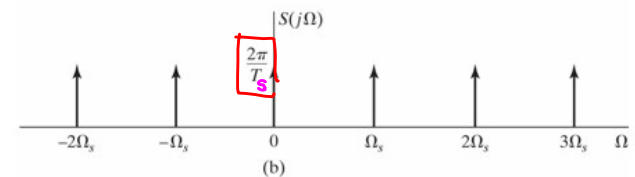
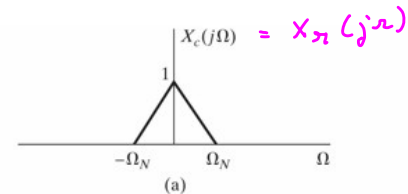
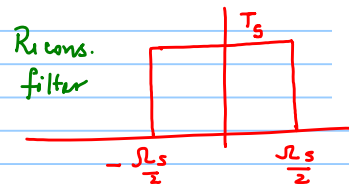
\equiv Lowpass filter with cutoff $\Omega_c = \frac{\Omega_s}{2}$

Reconstructed Signal $x_r(t)$ ← $x_s(j\Omega)$ Spectrum of the sampled signal
 $x_c(t) \rightarrow x_s(t) \rightarrow x[n]$

- ① Remove unwanted copies $m = \pm 1, \pm 2$
- ② Scale factor T_s

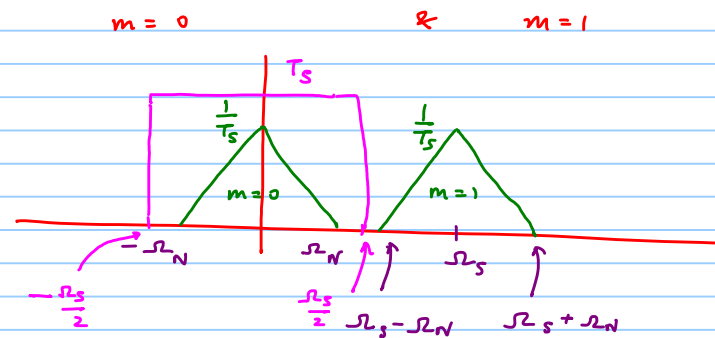
CT signal
non-zero for
all t

CT signal
non-zero only
@ nT_s



$$X_S(j\omega) = \underbrace{\frac{1}{T_S}}_{\text{scale factor}} \sum_{m=-\infty}^{\infty} \underbrace{X_C(j(\omega - m\omega_s))}_{\text{shifted by multiples of } \omega_s}$$

many copies



No overlap
of spectra

$$\omega_s - \omega_N > \omega_N$$

$$\boxed{\omega_s > 2\omega_N}$$

Special case

Nyquist rate $\omega_s = 2\omega_N$

oversampled

$$\omega_s > 2\omega_N$$

undersampled

$$\omega_s < \omega_N \leftrightarrow \text{aliasing}$$

$$h_n(t) = \text{sinc}\left(\frac{t}{T_S}\right)$$

$$\left\{ \begin{array}{l} = 1 \quad @ \quad t=0 \\ = 0 \quad @ \quad t = nT_S \quad n = \pm 1, \pm 2, \dots \end{array} \right.$$

→ important property for reconstruction

$$h_n(t) \longleftrightarrow H_n(j\omega)$$

Is $h_n(t)$ is BL? Yes.



Can apply Nyquist sampling

$$h_n[n] = h_n(t) \Big|_{t = n \frac{T_s}{8}} = \frac{\sin \frac{\pi t}{T_s}}{\frac{\pi t}{T_s}} \Big|_{t = n \frac{T_s}{8}} = \frac{\sin \frac{\pi n}{8}}{\frac{\pi n}{8}}$$

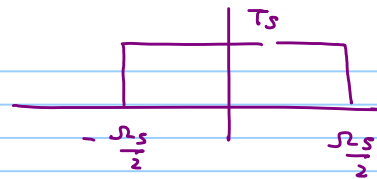
Oversampled
system
Factor of 8

$$h_n[n] = \frac{\sin \frac{\pi n}{8}}{\frac{\pi n}{8}} \quad -\infty < n < \infty$$

non-causal.
 ∞ duration

Is $h_n[n]$ is BIBO stable?

$$\sum_{n=-\infty}^{\infty} |h_n[n]| < \infty \text{ (finite)}$$



$$h_n[n] = \frac{\sin\left(\frac{\pi n}{8}\right)}{\frac{\pi n}{8}}$$

$$= 1 \quad n=0$$

$$h_n[0] \quad \text{L'Hospital's rule}$$

$$\lim_{n \rightarrow 0} \frac{\sin \frac{\pi n}{8}}{\frac{\pi n}{8}} = \lim_{n \rightarrow 0} \frac{\frac{\pi}{8} \cos \frac{\pi n}{8}}{\frac{\pi}{8}} = 1$$

$$\sum_{n=-\infty}^{\infty} |h_n[n]| = 1 + 2 \sum_{n=1}^{\infty} \left| \frac{\sin \frac{\pi n}{8}}{\frac{\pi n}{8}} \right| \leq 1 + 2 \sum_{n=1}^{\infty} \left| \frac{1}{\frac{\pi n}{8}} \right|$$

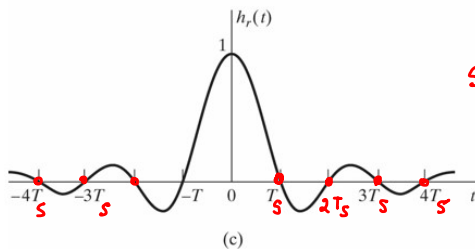
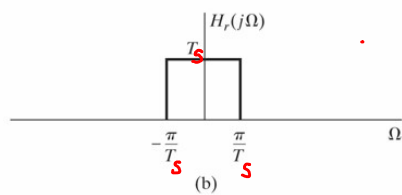
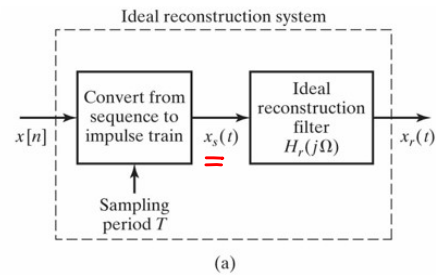
$$\left| \sin \frac{\pi n}{8} \right| \leq 1 \leq 1 + 2 \cdot \frac{8}{\pi} \sum_{n=1}^{\infty} \left(\frac{1}{n} \right)$$

Harmonic series $\sum_{n=1}^{\infty} \frac{1}{n}$ does not converge

$$\sum_{n=1}^{\infty} \frac{1}{n} = 1 + \frac{1}{2} + \underbrace{\frac{1}{3} + \frac{1}{4}}_{> \frac{1}{2}} + \underbrace{\frac{1}{5} + \frac{1}{6} + \frac{1}{7} + \frac{1}{8}}_{> \frac{1}{2}} + \dots > 1 + \frac{1}{2} + \frac{1}{2} + \frac{1}{2} + \dots$$

$\frac{1}{4} + \frac{1}{4} = \frac{1}{2}$
 $\frac{1}{8} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8} = \frac{1}{2}$

does not converge



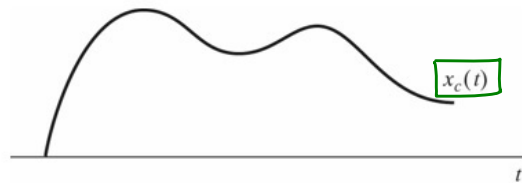
Reconstructed
signal

impulse response of reconstruction filter

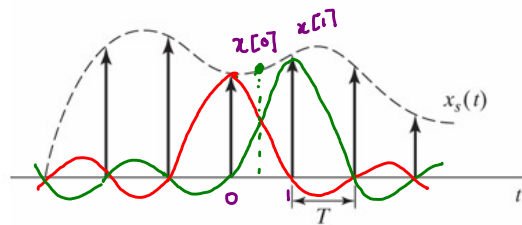
$$h_r(t) = \frac{1}{2\pi} \int_{-\frac{\pi}{T_s}}^{\frac{\pi}{T_s}} T_s e^{j\Omega t} d\Omega = \frac{\sin \frac{\pi t}{T_s}}{\frac{\pi t}{T_s}}$$

$$x_r(t) = x_s(t) * h_r(t) \longleftrightarrow X_s(j\Omega) H_r(j\Omega)$$

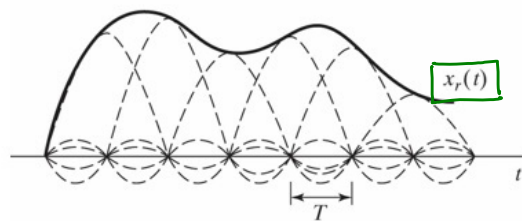
$\text{sinc}\left(\frac{t}{T_s}\right)$



(a)



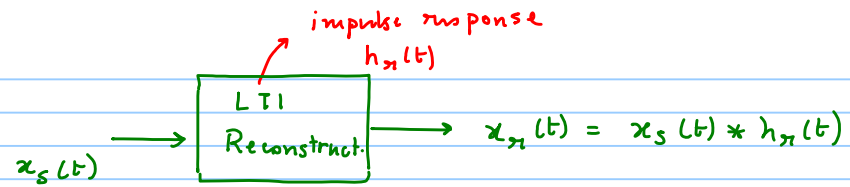
(b)



(c)

Dirac Delta functions

Reconstruction



$$h_n(t) = \frac{\sin \frac{\pi t}{T_s}}{\frac{\pi t}{T_s}} = \text{sinc}\left(\frac{t}{T_s}\right)$$

$$x_s(t) = \sum_{n=-\infty}^{\infty} x[n] \delta(t - nT_s) \quad \text{scaled Dirac Delta func}$$

$$x_n(t) = x_s(t) * h_n(t) = \left[\sum_{n=-\infty}^{\infty} x[n] \delta(t - nT_s) \right] * h_n(t)$$

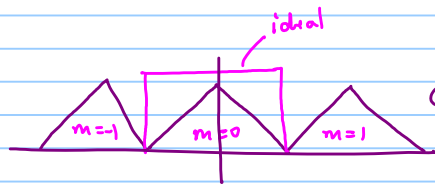
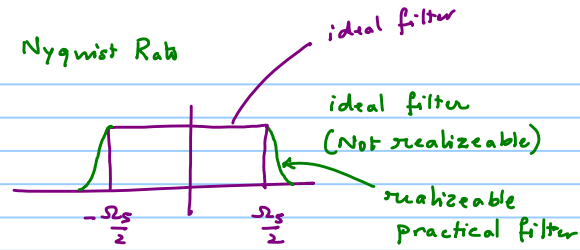
$x_s(t)$

$$x_n(t) = \sum_{n=-\infty}^{\infty} x[n] h_n(t - nT_s)$$

Oversampling

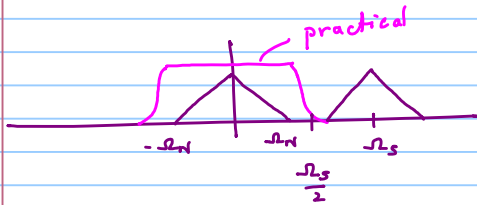
$\Omega_s > \text{Nyquist Rate}$

$\Omega_s = \text{Nyquist Rate}$



@ Nyquist rate

need ideal reconstruction filter

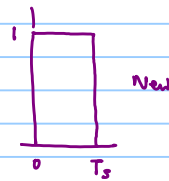
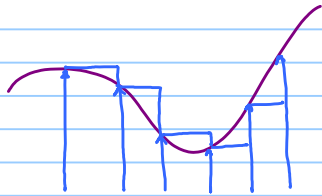


> Nyquist rate

$$\frac{\Omega_s}{2} > \Omega_N \Rightarrow \Omega_s > 2\Omega_N$$

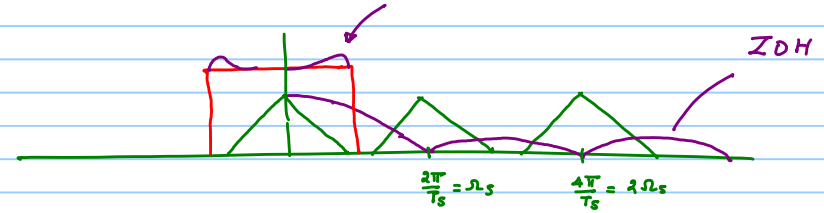
$\Omega_s > \text{Nyquist Rate}$

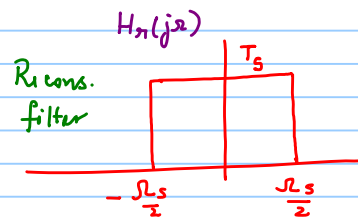
Practical Reconstruction



$$h_0(t) \longleftrightarrow H_0(j\omega) = e^{-j\omega \frac{T_s}{2}} \frac{2 \sin \frac{\omega T_s}{2}}{\omega}$$

Compensation in Reconstruction filter





$$h_n(t) = \text{sinc}\left(\frac{t}{T_s}\right)$$

$$H_0(j\Omega) = e^{-j\Omega \frac{T_s}{2}} \frac{2 \sin \frac{\Omega T_s}{2}}{\Omega} \quad \text{Zero Order Hold}$$

Modified Reconstruction Filter

$$H'_n(j\Omega) = \frac{H_n(j\Omega)}{H_0(j\Omega)} \quad \begin{array}{l} \rightarrow \text{original} \\ \rightarrow \text{ZOH} \end{array}$$

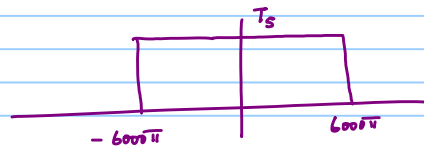
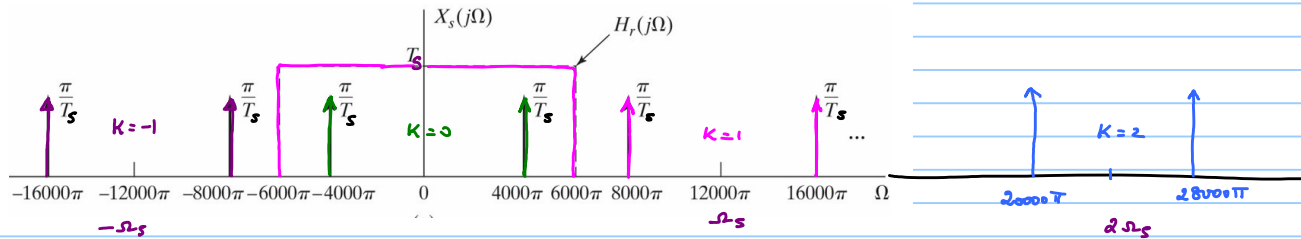
OKS
4.1

$$x_c(t) = \cos 4000\pi t \quad \text{Sample} \quad \boxed{\Omega_s = 12000\pi} \Rightarrow T_s = \frac{1}{6000}$$

$$\delta(\Omega - 4000\pi - k\Omega_s)$$

$$\delta(\Omega + 4000\pi - k\Omega_s)$$

$k=0$	4000π	-4000π
$k=1$	16000π	8000π
$k=2$	28000π	20000π
$k=-1$	-8000π	-16000π



$$\begin{aligned} x[n] &= x_c(t) \Big|_{t=nT_s} \\ &= \cos 4000\pi n T_s \Big|_{T_s = \frac{1}{6000}} \\ &= \cos \frac{4000\pi n}{6000} \\ \boxed{x[n] &= \cos \frac{2\pi}{3} n} \end{aligned}$$

Part (b)

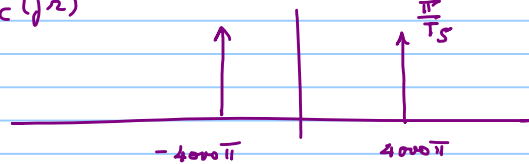
$$x_c(t) = \cos 4000\pi t$$

$$\max \text{ freq} = 2000 \text{ Hz}$$

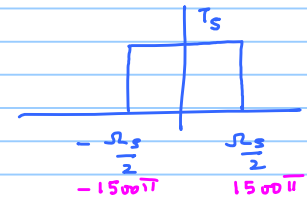
$$\text{Nyquist rate} = 8000 \pi$$

$$\Omega_s = 12000\pi \text{ (part (a))} \rightarrow \text{oversampled}$$

$X_c(j\omega)$



$$\Omega_s = 3000\pi$$



$$\delta(\Omega - 4000\pi - k\Omega_s)$$

$$k=0$$

$$4000\pi$$

$$k=1$$

$$7000\pi$$

$$k=2$$

$$10000\pi$$

$$k=-1$$

$$1000\pi$$

$$k=-2$$

$$-2000\pi$$

$$\delta(\Omega + 4000\pi - k\Omega_s)$$

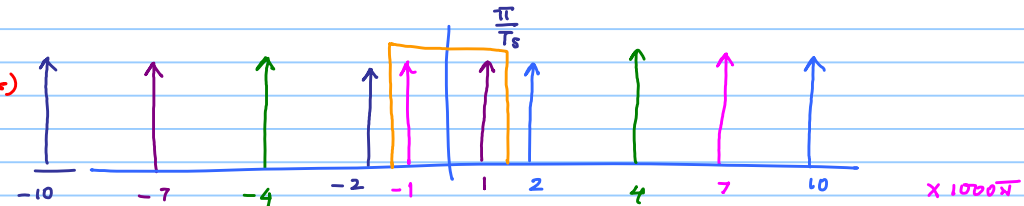
$$-4000\pi$$

$$-1000\pi$$

$$2000\pi$$

$$-7000\pi$$

$$-10000\pi$$



$$x'_n(t) = \cos 1000\pi t$$

$$x_c(t) = \cos 1000\pi t \Big|_{t=nT_s} \quad T_s = \frac{1}{1500}$$

$$x[n] = \cos \frac{1000\pi n}{1500} = \cos \frac{2\pi n}{3}$$

① Below Nyquist rate \rightarrow aliasing

② w. aliasing \rightarrow will not get the original signal via reconstruction

③ In reconstruction process, T_s plays a very important role

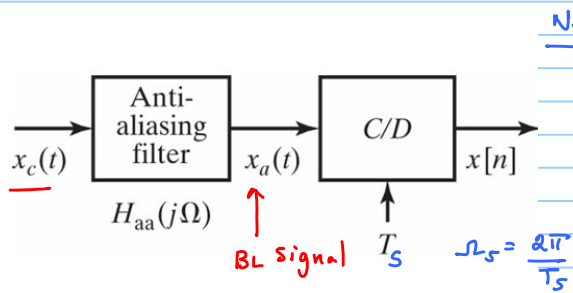
④ Same DT signal $x[n] = \cos \frac{2\pi}{3}n$ \rightarrow $\cos 4000\pi t$ $T_s = \frac{1}{6000}$
 \rightarrow $\cos 1000\pi t$ $T_s = \frac{1}{1500}$

Quantization

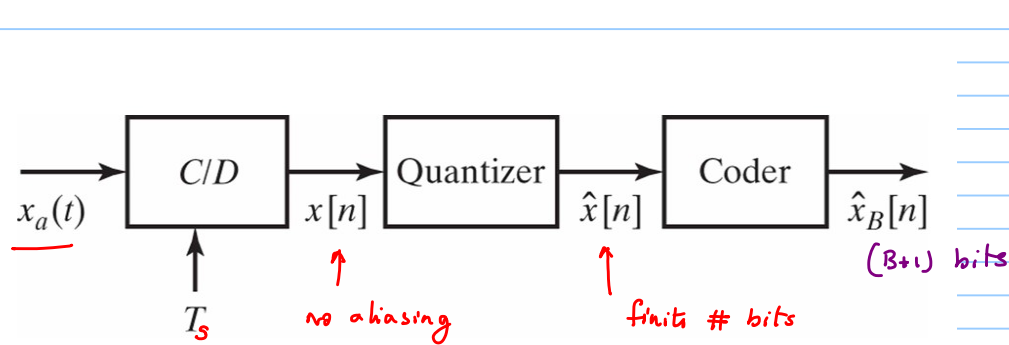
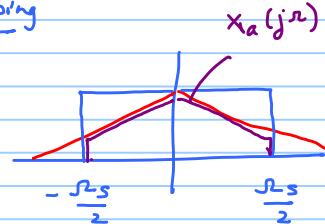
$$x_c(t) \xrightarrow{\text{Samp.}} x[n] \xrightarrow{\text{Quant}} \hat{x}[n]$$

Disc. in time
Cont. in amplitude
(∞ bits)

Disc. in time
Disc. in amplitude (finite # bits)



No aliasing



B+1 bit representation
fractional
sign

2's complement representation

(B+1) bits

X-axis input DT signal $x[n]$
 max value $[-X_m, X_m]$
 Total range $2X_m$

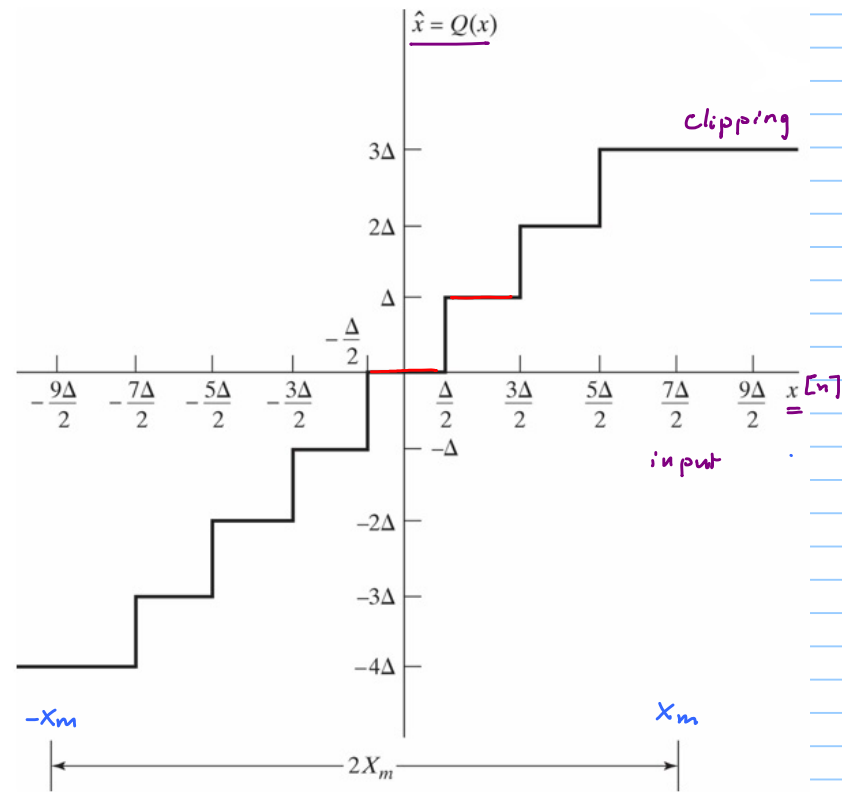
Uniform Quantizer via rounding

3 Bits $a_0 | a_1 a_2$
 sign fraction

value
 $-a_0 2^0 + a_1 2^{-1} + a_2 2^{-2} + \dots$

Output
 $-\frac{\Delta}{2} \leq x[n] < \frac{\Delta}{2}$ 000

$\frac{\Delta}{2} \leq x[n] < \frac{3\Delta}{2}$ 001

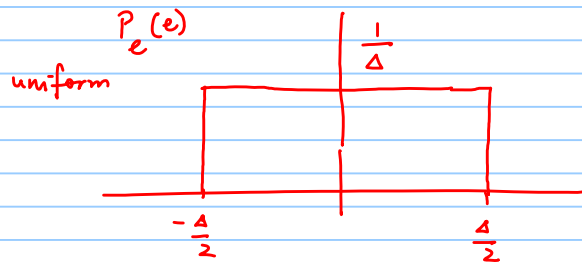
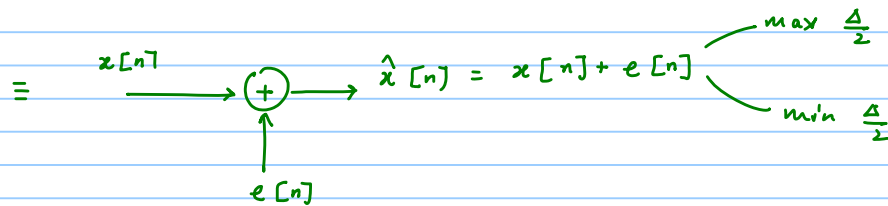


$B+1=3$

Two's-complement code

Two's-complement code	Value
011	$3/4$
010	$1/2$
001	$1/4$
000	0
111	$-1/4$
110	$-1/2$
101	$-3/4$
100	-1

$$\text{Stepsize } \Delta = \frac{2^{B+1} X_m}{2^{B+1}} = \frac{X_m}{2^B}$$



Characteristics of $e[n]$

- ① $e[n]$ is a sample seq. of a Wide Sense Stationary (WSS) Random Process
- ② $e[n]$ error seq. is uncorrelated with $x[n]$
- ③ $e[n]$ is a white noise random noise
- ④ PDF of $e[n]$ $P_e(e)$ is uniform

$$\begin{aligned} \text{① } \mu_e &= \int_{-\frac{\Delta}{2}}^{\frac{\Delta}{2}} e P_e(e) de \\ &= \frac{1}{\Delta} \int_{-\frac{\Delta}{2}}^{\frac{\Delta}{2}} e de = \frac{1}{\Delta} \left. \frac{e^2}{2} \right|_{-\frac{\Delta}{2}}^{\frac{\Delta}{2}} = 0 \end{aligned}$$

$$\textcircled{2} \quad \sigma_e^2 = \int_{-\frac{\Delta}{2}}^{\frac{\Delta}{2}} e^2 P_e(e) de = \frac{1}{\Delta} \int_{-\frac{\Delta}{2}}^{\frac{\Delta}{2}} e^2 de = \frac{1}{\Delta} \left. \frac{e^3}{3} \right|_{-\frac{\Delta}{2}}^{\frac{\Delta}{2}} = \frac{\Delta^2}{12}$$

$$\Delta = \frac{X_m}{2^B} = 2^{-B} X_m$$

$$\sigma_e^2 = 2^{-2B} \frac{X_m^2}{12} \quad \textcircled{1}$$

$$\text{SQNR} = 10 \log_{10} \frac{\sigma_x^2}{\sigma_e^2} \quad \textcircled{2}$$

Signal to Quantiz. Noise Ratio

Substituting $\textcircled{1}$ in $\textcircled{2}$ $\text{SQNR} = 10 \log_{10} 12 \cdot 2^{2B} \left(\frac{\sigma_x^2}{X_m^2} \right)$

σ_x^2 = energy of input signal

X_m = Range of input signal

Typically $\sigma_x^2 = \frac{X_m^2}{4} \quad \textcircled{3}$

$$\text{SQNR} = 10 \log_{10} (2^{2B}) + 10 \log_{10} (12) - 20 \log_{10} \left(\frac{X_m}{\sigma_x} \right) \quad \textcircled{4}$$

Substitute $\textcircled{3}$ in $\textcircled{4}$

$$\text{SQNR} = [6.02B - 1.25] \text{ dB}$$

$B = 16 \text{ bits} \quad \text{SQNR} = 96 - 1.25$
 $= 94.75 \text{ dB}$

-6dB/bit Quantization
 Rule-of-thumb for Quantiz

Discrete Time Fourier Transform

$$\text{DT seq } x[n] \xrightarrow[\text{DTFT}]{f} X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}$$

↑
continuous function of ω
 ω is periodic function with period $= 2\pi$
 $X(e^{j\omega})$ is also periodic function

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{j\omega n} d\omega \xleftarrow[\text{Inv DTFT}]{f^{-1}}$$

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

$x[n]$ can be real valued or complex-valued

$X(e^{j\omega})$ is complex valued (in general)

Fourier Transform
or
Fourier Spectrum
or
DTFT

$$X(e^{j\omega}) = X_R(e^{j\omega}) + j X_I(e^{j\omega})$$

$$= \underbrace{|X(e^{j\omega})|}_{\text{Magnitude Spectrum}} e^{j \underbrace{\arg\{X(e^{j\omega})\}}_{\text{phase spectrum}}}$$

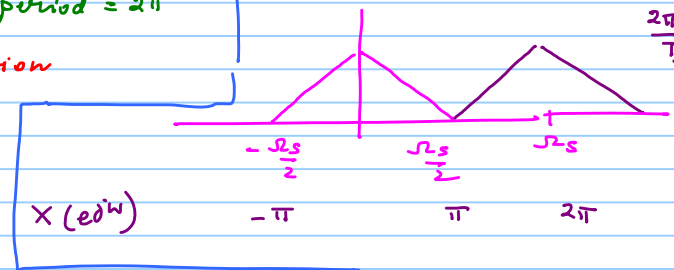
$$x_c(t) \longrightarrow x_s(t)$$

$X_s(j\omega)$ periodic
period $= \Omega_s$

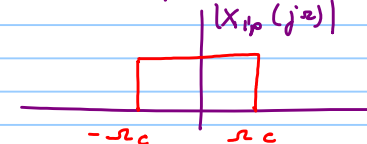
$$x[n] \longrightarrow X(e^{j\omega})$$

$$\Omega_s T_s = \omega_s$$

$$\frac{2\pi}{T_s} \cdot T_s = \omega_s$$



Low pass filter



Magnitude Spectrum

LTI system \longleftrightarrow Impulse Response $h[n]$ \longleftrightarrow DTFT $H(e^{j\omega})$

example

$$h[n] = \delta[n-2]$$

$$\longleftrightarrow H(e^{j\omega}) = \sum_{n=-\infty}^{\infty} h[n] e^{-j\omega n}$$

$$y[n] = x[n] * h[n] = x[n-2]$$

$$H(e^{j\omega}) = e^{-j2\omega}$$

$$|H(e^{j\omega})| = 1 \quad \forall \omega \quad \left| \arg(H(e^{j\omega})) = -2\omega \right.$$

Proof of Inv. DTFT

Given $X(e^{j\omega}) = \sum_{k=-\infty}^{\infty} x[k] e^{-j\omega k}$

Suppose $\hat{x}[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} \underbrace{X(e^{j\omega})}_{\text{from given}} e^{j\omega n} d\omega$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} \left(\sum_{k=-\infty}^{\infty} x[k] e^{-j\omega k} \right) e^{j\omega n} d\omega = \sum_{k=-\infty}^{\infty} x[k] \underbrace{\frac{1}{2\pi} \int_{-\pi}^{\pi} e^{j\omega(n-k)} d\omega}_{\substack{=1 \quad k=n \\ 0 \quad k \neq n}}$$

$$\hat{x}[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k] = x[n]$$

$$\boxed{\hat{x}[n] = x[n]}$$

DTFT pair

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} \quad \text{DTFT (forward)}$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{j\omega n} d\omega \quad \text{Inv DTFT (Inverse)}$$