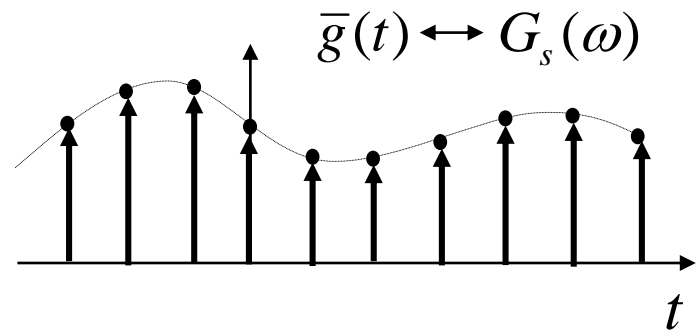
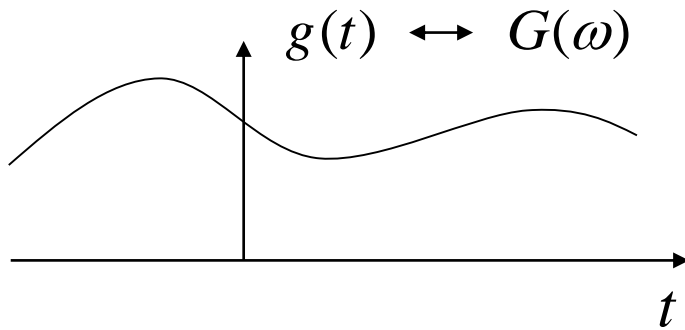
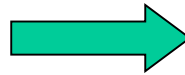
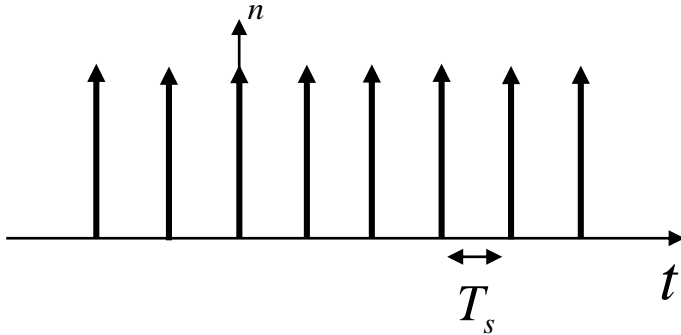


Sampling Theorem

$$\delta_{T_s}(t) = \sum_n \delta(t - nT_s) \leftrightarrow G[n\omega_s]$$



Sampling Frequency

$$f_s = \frac{1}{T_s}$$

Sampling Theorem

Let's $g(t)$ is band limited to B Hz, then samples of $g(t)$ could be written as

$$\bar{g}(t) = g(t)\delta_{T_s}(t) = \sum_n g(nT_s)\delta(t - nT_s)$$

We know that

$$\delta_{T_s}(t) = \frac{1}{T_s} [1 + 2 \cos \omega_s t + 2 \cos 2\omega_s t + 2 \cos 3\omega_s t + \dots]$$

where $\omega_s = \frac{2\pi}{T_s} = 2\pi f_s$

$$\begin{aligned}
\bar{g}(t) &= g(t)\delta_{T_s}(t) \\
&= \frac{1}{T_s} [g(t) + 2g(t)\cos\omega_s t + 2g(t)\cos 2\omega_s t + 2g(t)\cos 3\omega_s t + \dots] \\
&= \frac{g(t)}{T_s} + \frac{2}{T_s} \sum_{n=1}^{\infty} g(t)\cos n\omega_s t
\end{aligned}$$

By taking Fourier Transform on both sides

$$\bar{G}(\omega) = \frac{G(\omega)}{T_s} + \frac{1}{T_s} \sum_{n=1}^{\infty} [G(\omega - n\omega_s) + G(\omega + n\omega_s)]$$

$$= \frac{1}{T_s} \sum_{n=-\infty}^{\infty} G(\omega - n\omega_s)$$

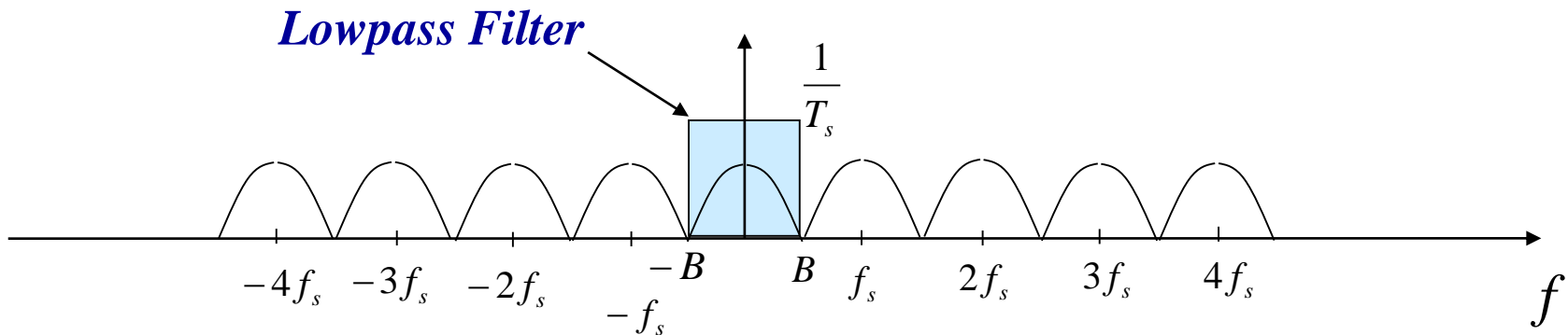
Nyquist Criterion

$$f_s > 2B \quad \text{or} \quad T_s < \frac{1}{2B}$$

Sampling Theorem

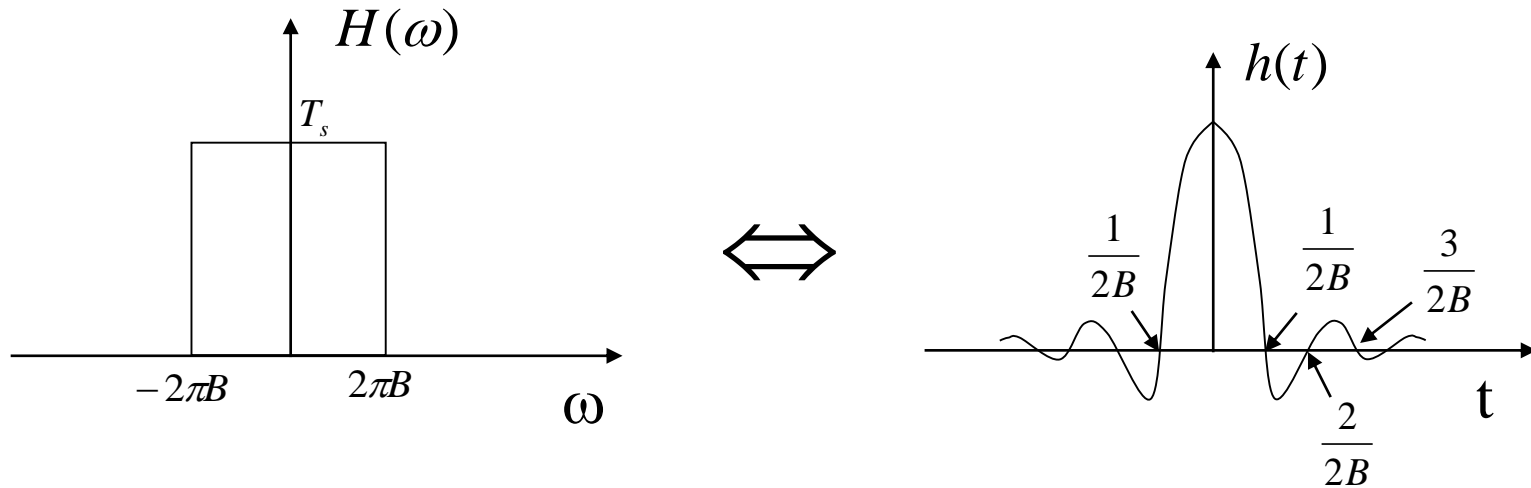
Sampling frequency should be at least equal to or greater than twice the bandwidth of the message signal for successful recovery of the signal from its samples

$$f_s \geq 2B$$



bandwidth of lowpass filter = B Hz

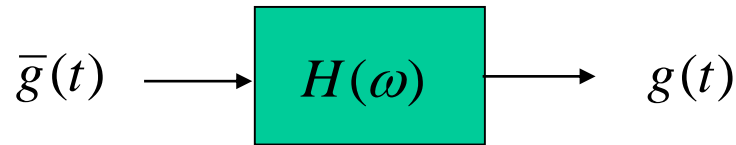
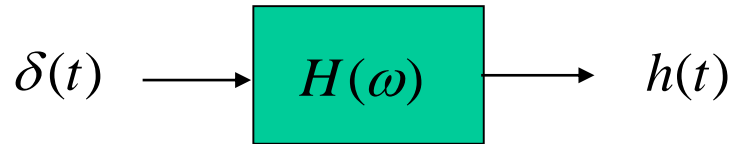
Ideal Reconstruction from Samples



$$H(\omega) = T_s \text{rect}\left(\frac{\omega}{4\pi B}\right)$$

$$h(t) = 2BT_s \text{sinc}(2\pi Bt)$$

$$h(t) = \text{sinc}(2\pi Bt) \quad (2BT_s = 1)$$



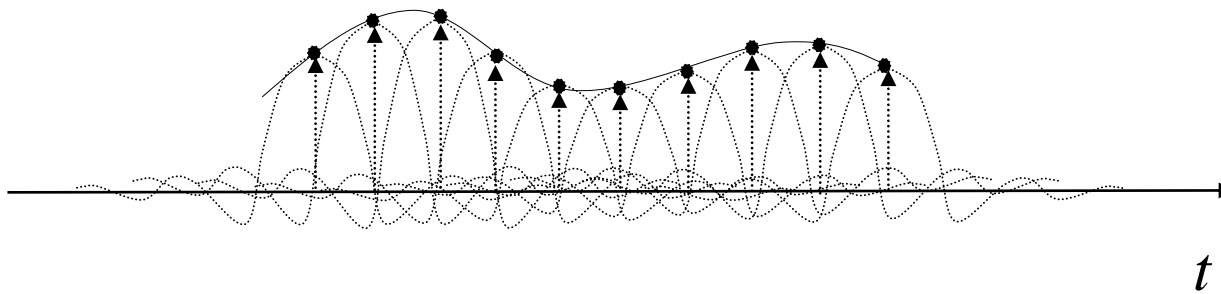
Recall
$$\bar{g}(t) = g(t)\delta_{T_s}(t) = \sum_n g(nT_s)\delta(t - nT_s)$$

$$\Rightarrow g(t) = \sum g(nT_s)h(t - nT_s)$$

$$= \sum g(nT_s) \text{sinc}[2\pi B(t - nT_s)]$$

$$= \sum g(nT_s) \text{sinc}(2\pi Bt - n\pi)$$

Recovered $g(t)$

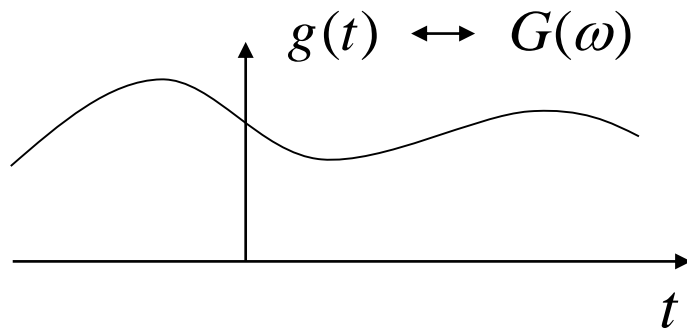
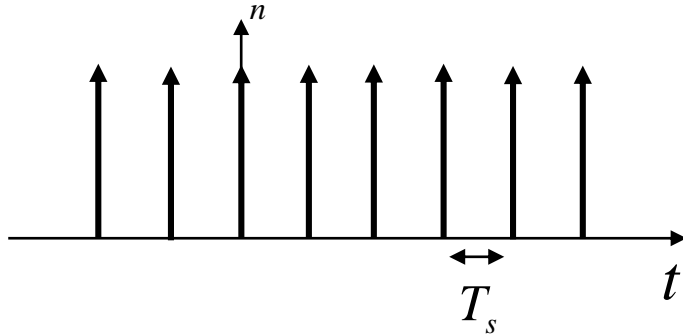


Practical Considerations in Nyquist Sampling

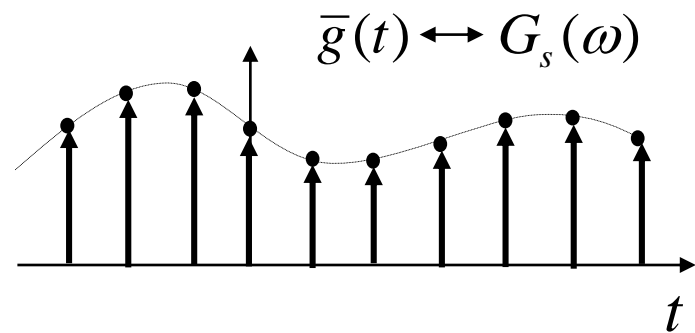
- *Gradual roll-off lowpass filter*
- *Aliasing*
- *Practical Pulses vs. Ideal Impulses*

Sampling Theorem

$$\delta_{T_s}(t) = \sum_n \delta(t - nT_s) \leftrightarrow G[n\omega_s]$$



Not practical to generate!

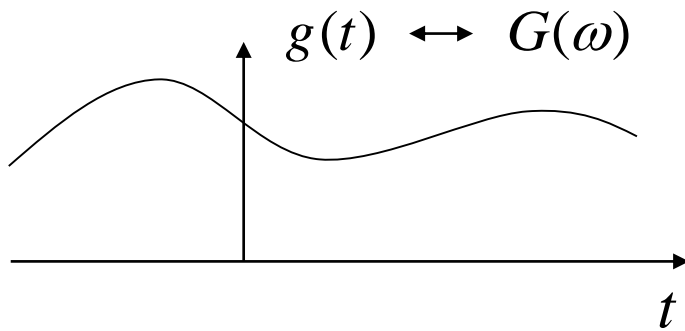
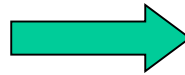
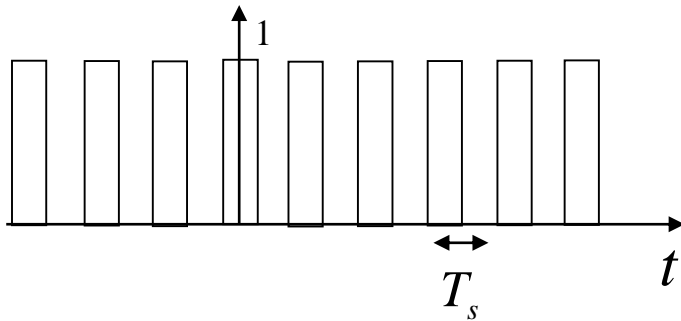


Sampling Frequency

$$f_s = \frac{1}{T_s}$$

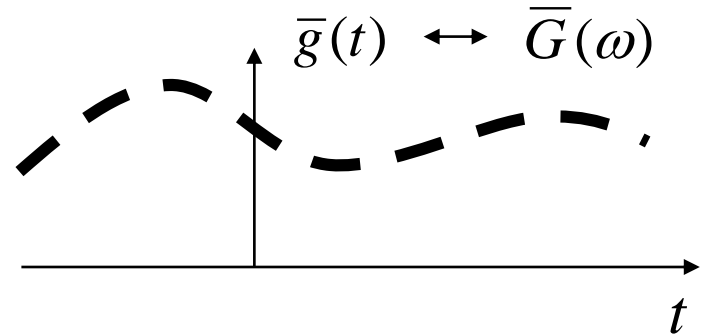
Let's Try Square Wave

$$x(t) \leftrightarrow X[n\omega_s]$$



$$g(t) \leftrightarrow G(\omega)$$

Problem Resolved



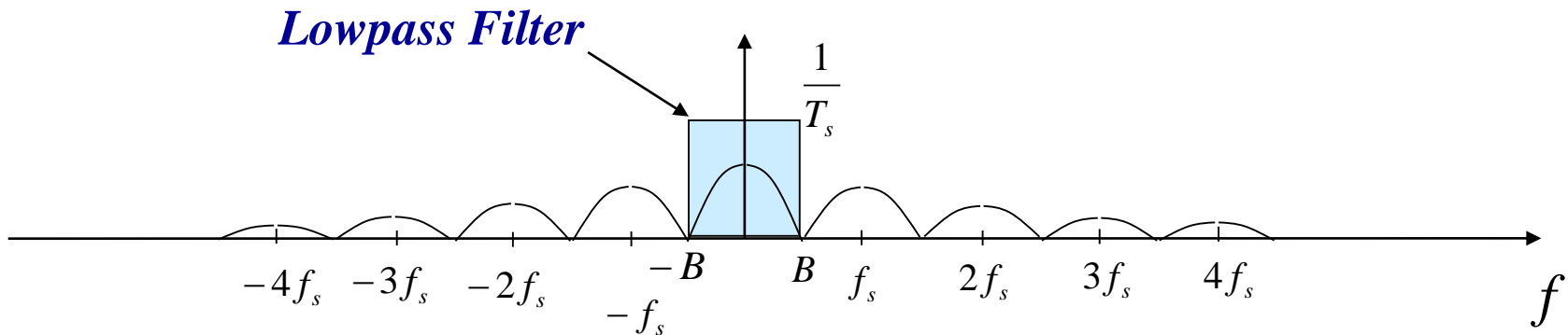
Sampling Frequency

$$f_s = \frac{1}{T_s}$$

Sampling Theorem

Sampling frequency should be at least equal to or greater than twice the bandwidth of the message signal for successful recovery of the signal from its samples

$$f_s \geq 2B$$



bandwidth of lowpass filter = B Hz

Maximum Information Rate

Signal of bandwidth B



$2B$ samples per second



$2B$ Independent pieces of information per second



Two independent pieces of information per second per hertz of bandwidth

Sampling Theorem and Digital Communication

- *Pulse Amplitude Modulation* (M-ary)
- *Pulse Width Modulation* (M-ary)
- *Pulse Position Modulation* (M-ary)
- *Pulse Code Modulation* (Binary)

Pulse Code Modulation (PCM)



Speech Signal from 15Hz to 15kHz

	<i>Maximum Frequency</i>	<i>Sampling Rate</i>	<i>Quantization Levels</i>	<i>Number of Bits/code</i>	<i>Bit Rate</i>
Telephone	3.4 kHz	8 kHz	256	8	64 kb/s
Compact Disk	15 kHz	44.1 kHz	65,536	16	705.6 kb/s

Quantization

Reconstruction of actual samples after sampling

$$g(t) = \sum g(nT_s) \sin c(2\pi Bt - n\pi)$$

Reconstruction of quantized samples after sampling

$$\hat{g}(t) = \sum \hat{g}(nT_s) \sin c(2\pi Bt - n\pi)$$

Distortion in reconstructed signal

$$q(t) = \hat{g}(t) - g(t)$$

$$\Rightarrow q(t) = \sum [\hat{g}(nT_s) - g(nT_s)] \sin c(2\pi Bt - n\pi)$$

$$\Rightarrow q(t) = \sum q(nT_s) \sin c(2\pi Bt - n\pi)$$

where $q(nT_s)$ is the quantization error in n th sample

Power or mean square value of $q(t)$ (quantization noise) is given by

$$\begin{aligned} \overline{q(t)^2} &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} q^2(t) dt \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \left[\sum_n q(nT_s) \operatorname{sinc}(2\pi Bt - n\pi) \right]^2 dt \end{aligned}$$

Now by using the fact that $\operatorname{sinc}()$ are orthogonal functions i.e.,

$$\int_{-\infty}^{\infty} \operatorname{sinc}(2\pi Bt - n\pi) \operatorname{sinc}(2\pi Bt - m\pi) dt = \begin{cases} 0 & m \neq n \\ \frac{1}{2B} & m = n \end{cases}$$

$$\Rightarrow \overline{q(t)^2} = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \sum_n q^2(nT_s) \operatorname{sinc}^2(2\pi Bt - n\pi) dt$$

$$\begin{aligned}
\Rightarrow \overline{q(t)^2} &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \sum_n q^2(nT_s) \sin^2(2\pi Bt - n\pi) dt \\
&= \lim_{T \rightarrow \infty} \frac{1}{T} \sum_n q^2(nT_s) \int_{-T/2}^{T/2} \sin^2(2\pi Bt - n\pi) dt \\
&= \lim_{T \rightarrow \infty} \frac{1}{T} \sum_n q^2(nT_s) \left(\frac{1}{2B}\right) \\
&= \lim_{T \rightarrow \infty} \frac{1}{2BT} \sum_n q^2(nT_s)
\end{aligned}$$

=> quantization noise is average of square of quantization error

Let's calculate this

quantization error lies in the range of $-\Delta v/2$ and $\Delta v/2$

where
$$\Delta v = \frac{2m_p}{L}$$

Assuming quantization error is equally likely in the range of $-\Delta v/2$ and $\Delta v/2$

$$\begin{aligned}\overline{q^2} &= \frac{1}{\Delta v} \int_{-\Delta v/2}^{\Delta v/2} q^2 dq \\ &= \frac{1}{\Delta v} \left| \frac{q^3}{3} \right|_{-\Delta v/2}^{\Delta v/2} \\ &= \frac{1}{\Delta v} \frac{(\Delta v)^3}{12} \\ &= \frac{(\Delta v)^2}{12} = \frac{m_p^2}{3L^2}\end{aligned}$$

$$\Rightarrow N_q = \overline{q^2(t)} = \frac{m_p^2}{3L^2}$$

$$\widehat{g}(t) = \underset{\swarrow}{g(t)} + \underset{\searrow}{q(t)}$$

Desired signal

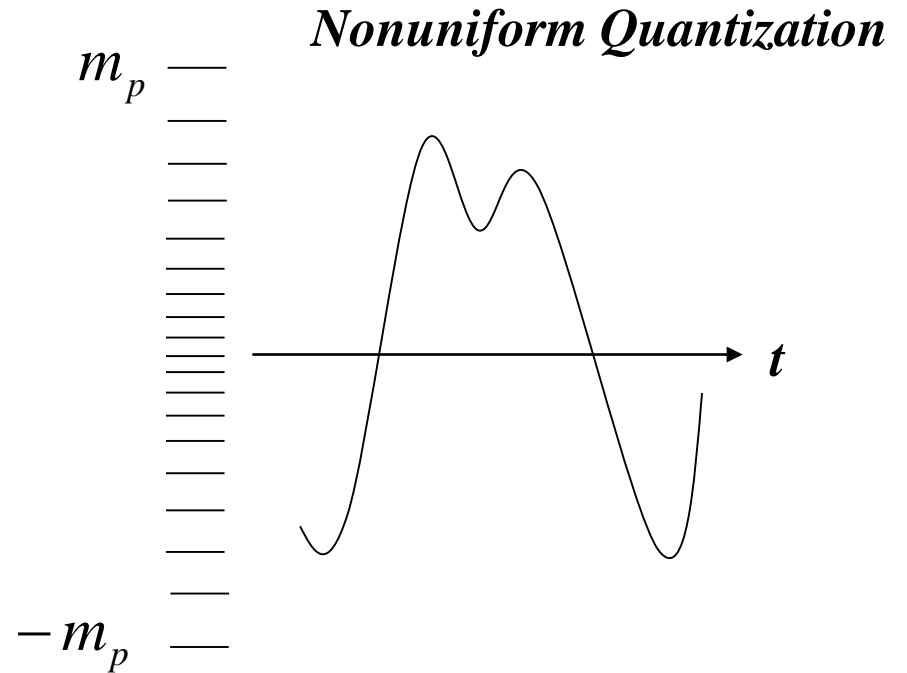
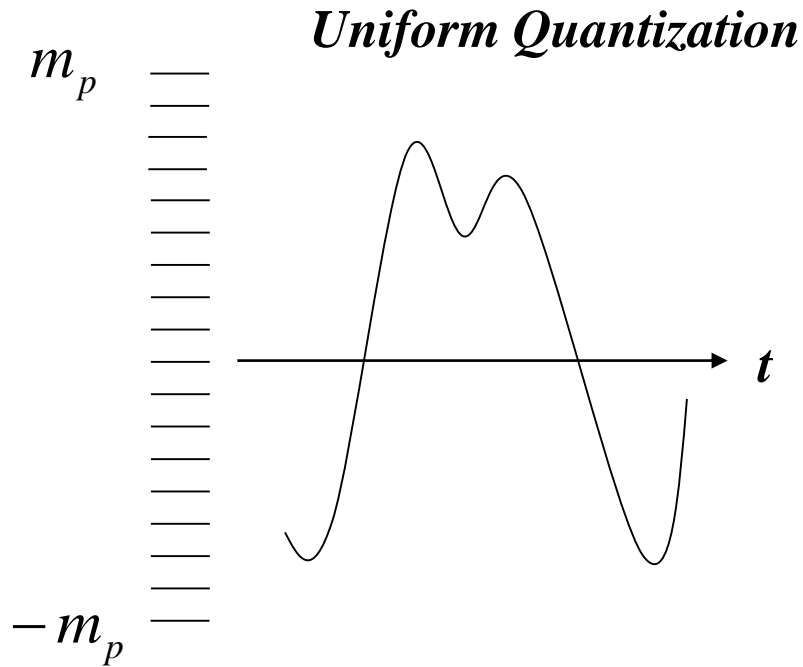
Unwanted noise

$$\Rightarrow S_0 = \overline{g^2(t)}$$

and $N_0 = N_q = \frac{m_p^2}{3L^2}$

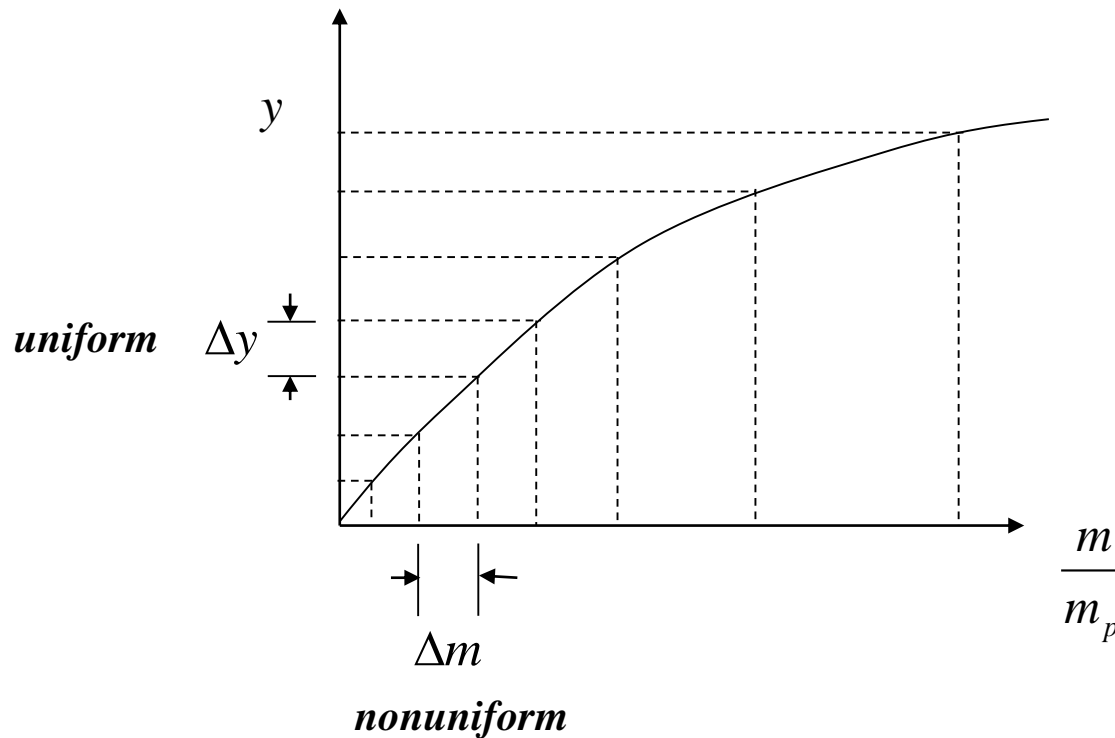
Therefore $SNR = \frac{S_0}{N_0} = 3L^2 \frac{\overline{m^2(t)}}{m_p^2}$

Nonuniform Quantization



Nonuniform Quantization = Compression + Uniform Quantization

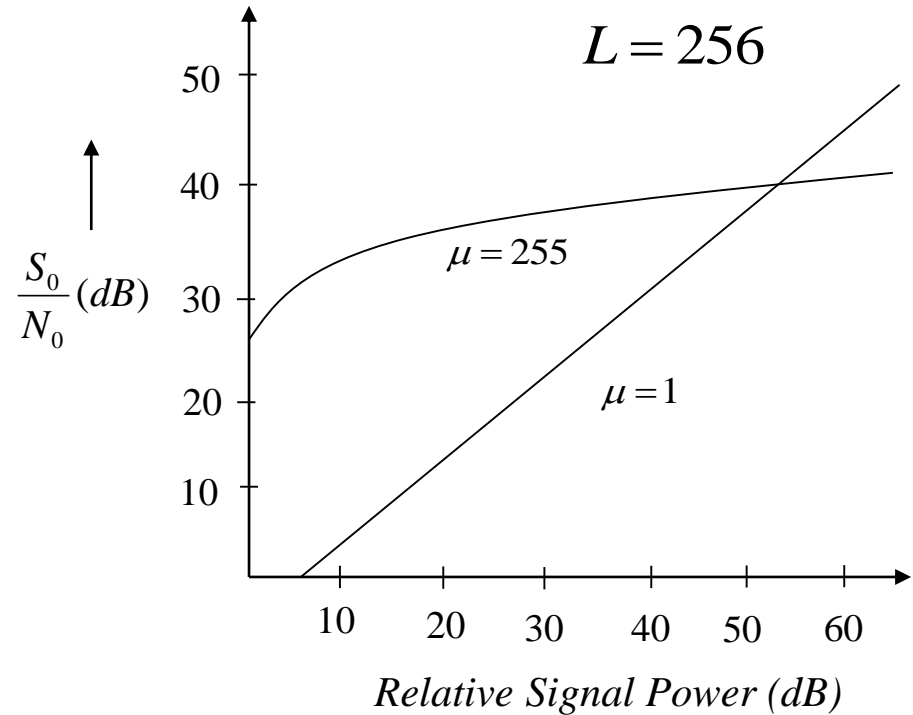
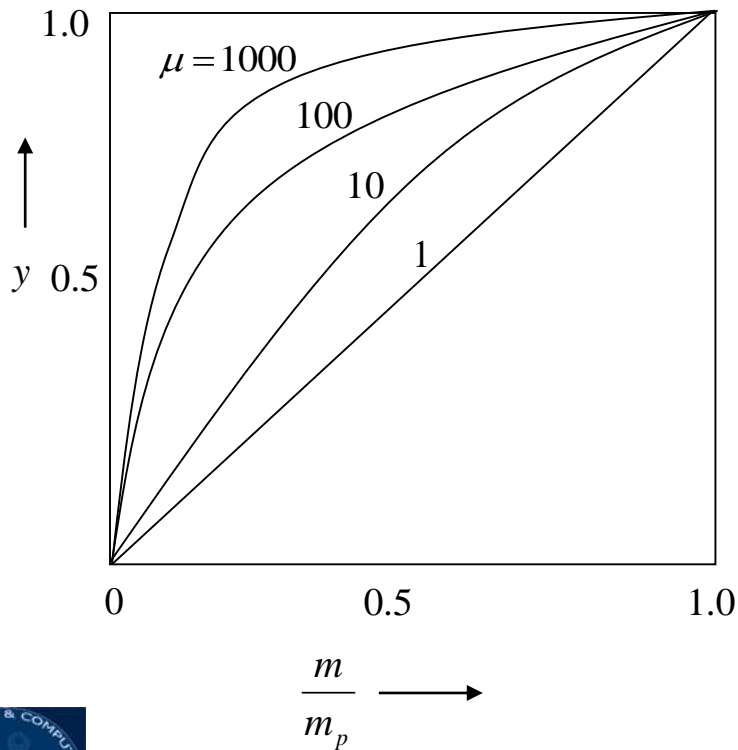
(Usually compression increases the bandwidth but not applicable here)



Example

μ -law

$$y = \frac{1}{\ln(1 + \mu)} \ln\left(1 + \frac{\mu m}{m_p}\right)$$



Transmission Bandwidth for PCM

Signal Bandwidth = B

Sampling Rate = 2B

Quantized levels = $L = 2^n$

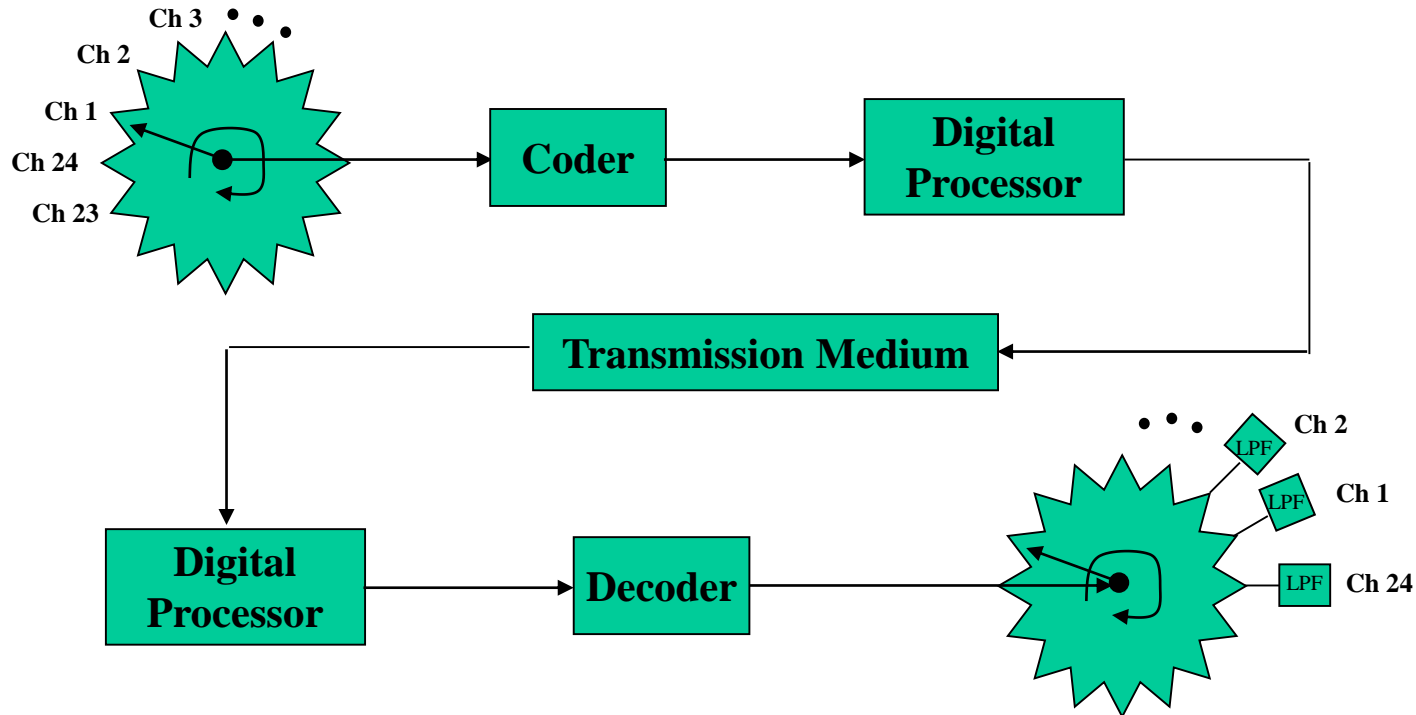
\Rightarrow Number of encoded bits = n

Bit Rate = 2nB bits / sec

\Rightarrow Required Transmission Bandwidth = nB Hz

Digital Communication and Time Division Multiplexing

- *T1 Carrier System (DS1) is an example*
- *1.544Mbits/sec total data rate for 24 channels*
- *Each channel is encoded with 8 bits*



What does Digital Processor do?

- 1. Synchronization (Framing)*
- 2. Signaling*
- 3. Parity Insertion and Check*
- 4. Scrambling*

Energy of a Signal and Parseval's Theorem

$$E_g = \int_{-\infty}^{\infty} g(t)g^*(t)dt$$

$$E_g = \int_{-\infty}^{\infty} g(t) \left[\frac{1}{2\pi} \int_{-\infty}^{\infty} G^*(\omega)e^{-j\omega t} d\omega \right] dt$$

$$E_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} G^*(\omega) \left[\int_{-\infty}^{\infty} g(t)e^{-j\omega t} dt \right] d\omega$$

$$E_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} G^*(\omega)G(\omega)d\omega$$

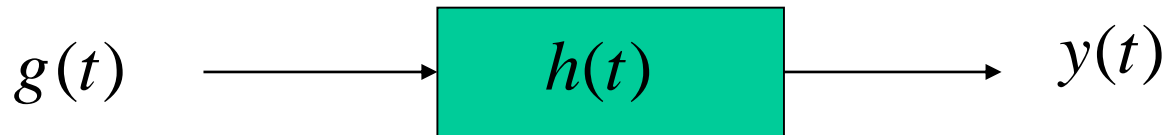
$$E_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} |G(\omega)|^2 d\omega$$

Energy Spectral Density

$$E_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} |G(\omega)|^2 d\omega \quad \Rightarrow \quad \Psi_g(\omega) = |G(\omega)|^2$$

energy per unit bandwidth

ESD of the Input and the Output



$$Y(\omega) = H(\omega)G(\omega)$$

$$|Y(\omega)|^2 = |H(\omega)|^2 |G(\omega)|^2$$

$$\Rightarrow \Psi_y(\omega) = |H(\omega)|^2 \Psi_g(\omega)$$

Autocorrelation function of a signal

$$\psi_g(\tau) = \int_{-\infty}^{\infty} g(t)g(t + \tau)dt$$

let $x = t + \tau$

$$\begin{aligned}\psi_g(\tau) &= \int_{-\infty}^{\infty} g(x - \tau)g(x)dx = \int_{-\infty}^{\infty} g(x)g(x - \tau)dx \\ &= \int_{-\infty}^{\infty} g(t)g(t - \tau)dt\end{aligned}$$

$$\Rightarrow \psi_g(\tau) = \psi_g(-\tau)$$

$$\begin{aligned}
F[\psi_g(\tau)] &= \int_{-\infty}^{\infty} e^{-j\omega\tau} \left[\int_{-\infty}^{\infty} g(t)g(t+\tau)dt \right] d\tau \\
&= \int_{-\infty}^{\infty} g(t) \left[\int_{-\infty}^{\infty} g(t+\tau)e^{-j\omega\tau} d\tau \right] dt \\
&= \int_{-\infty}^{\infty} g(t) \left[\int_{-\infty}^{\infty} g(\tau+t)e^{-j\omega\tau} d\tau \right] dt \\
&= \int_{-\infty}^{\infty} g(t) [G(\omega)e^{j\omega t}] dt \\
&= G(\omega) \int_{-\infty}^{\infty} g(t)e^{j\omega t} dt \\
&= G(\omega)G(-\omega) = |G(\omega)|^2 \\
&\Rightarrow \psi_g(\tau) \Leftrightarrow \Psi_g(\omega)
\end{aligned}$$

Signal Power and Power Spectral Density

$$P_g = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} g^2(t) dt$$

$$g_T(t) = \begin{cases} g(t) & |t| \leq T/2 \\ 0 & |t| \geq T/2 \end{cases}$$

$$P_g = \lim_{T \rightarrow \infty} \frac{\int_{-\infty}^{\infty} g_T^2(t) dt}{T}$$

$$= \lim_{T \rightarrow \infty} \frac{E_{g_T}}{T}$$

$$E_{g_T} = \int_{-\infty}^{\infty} g_T^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |G_T(\omega)|^2 d\omega$$

$$\Rightarrow P_g = \lim_{T \rightarrow \infty} \frac{E_{g_T}}{T} = \lim_{T \rightarrow \infty} \frac{1}{T} \left[\frac{1}{2\pi} \int_{-\infty}^{\infty} |G_T(\omega)|^2 d\omega \right]$$

$$\Rightarrow P_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} \lim_{T \rightarrow \infty} \frac{|G_T(\omega)|^2}{T} d\omega$$

We define PSD as

$$S_g(\omega) = \lim_{T \rightarrow \infty} \frac{|G_T(\omega)|^2}{T}$$

So that

$$P_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_g(\omega) d\omega = \frac{1}{\pi} \int_0^{\infty} S_g(\omega) d\omega$$

Autocorrelation function of a power signal

$$\mathfrak{R}_g(\tau) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} g(t)g(t + \tau)dt$$

Same argument as in energy signals will prove that

$$\mathfrak{R}_g(\tau) = \mathfrak{R}_g(-\tau)$$

and $\mathfrak{R}_g(\tau) \Leftrightarrow S_g(\omega)$

also $S_y(\omega) = |H(\omega)|^2 S_g(\omega)$

if $Y(\omega) = H(\omega)G(\omega)$

Various Digital Line Codes

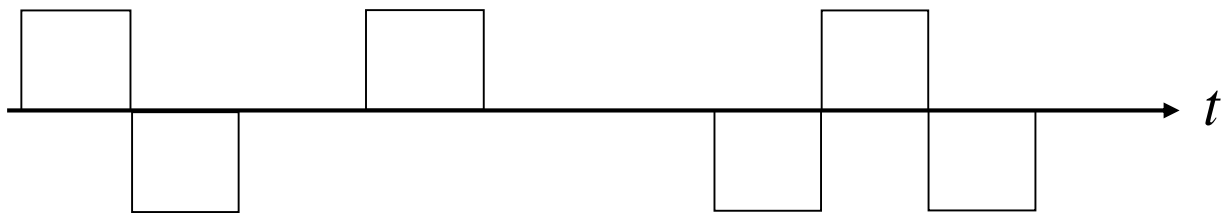
1 1 0 1 0 0 1 1 1 0



On-Off



Polar



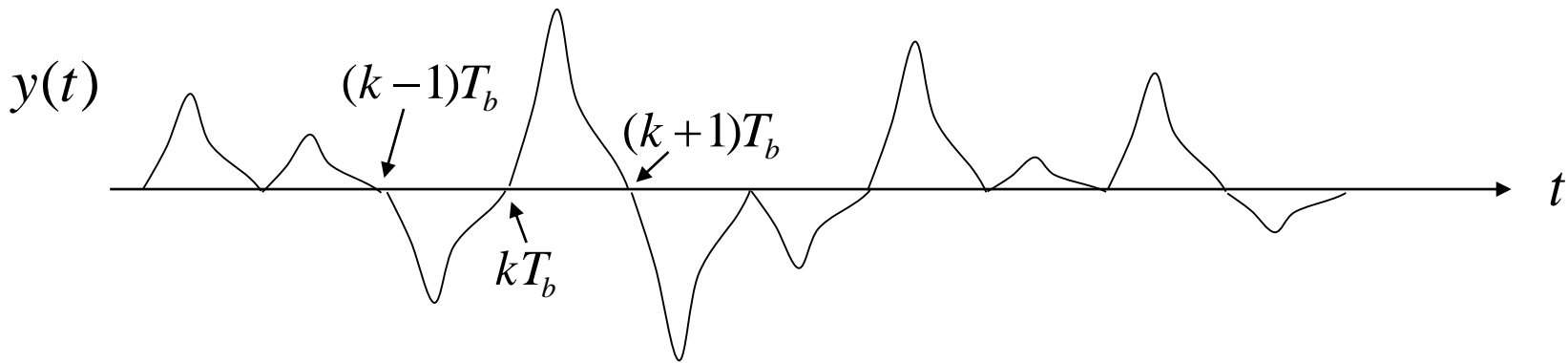
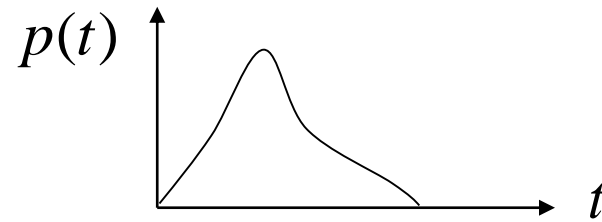
Bi-Polar

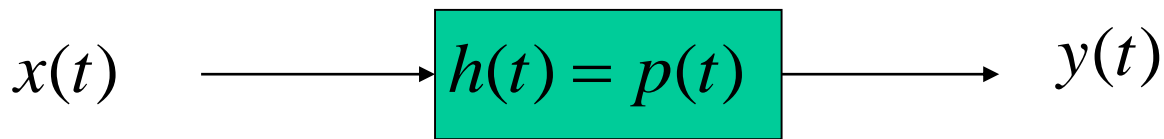
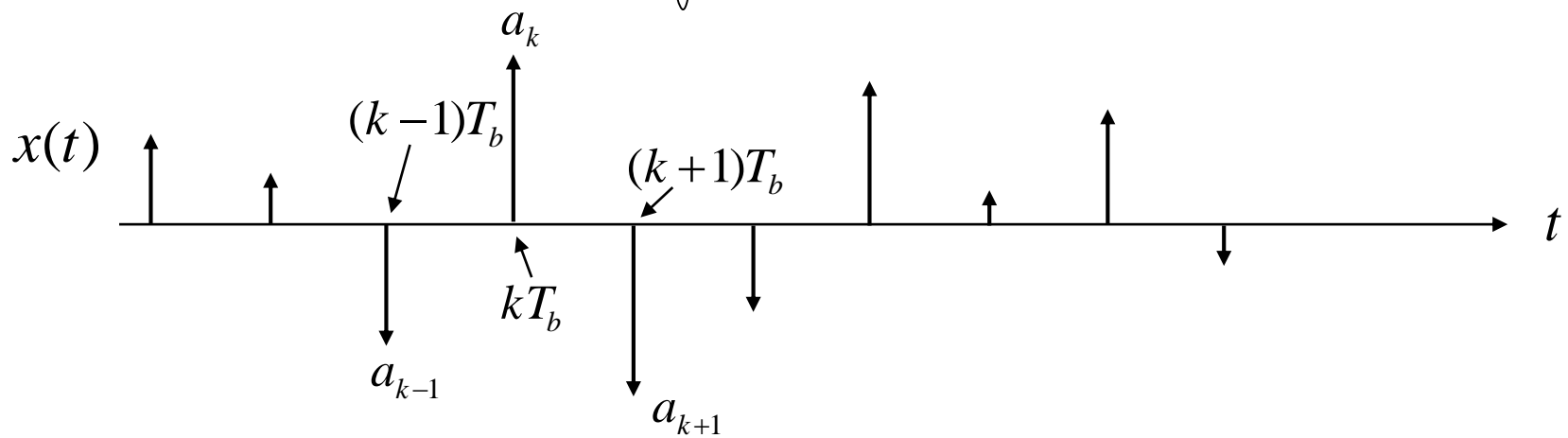
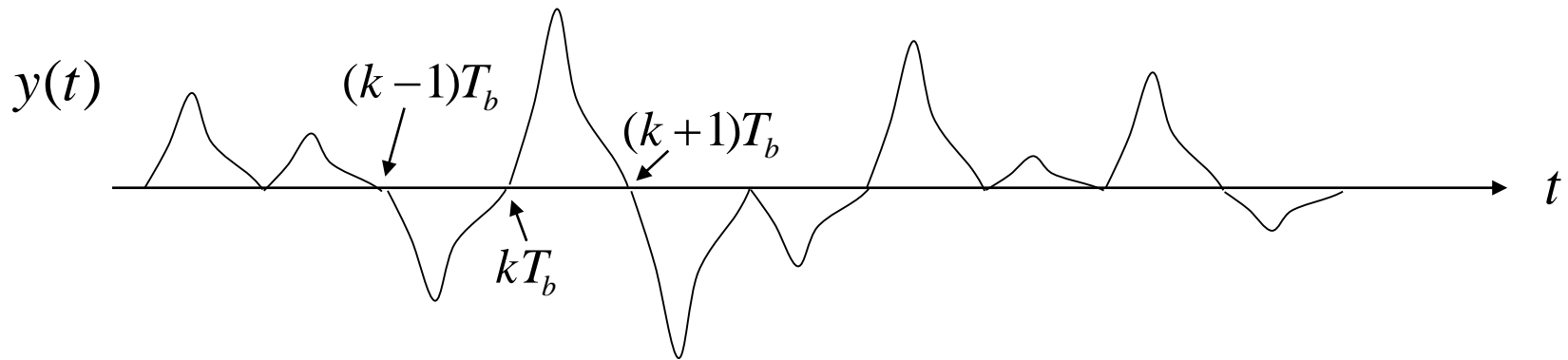
Pulse shape $p(t)$ could be arbitrary

Some Characteristics of Good Line Codes

- 1. Bandwidth Efficient*
- 2. Power Efficient*
- 3. Error Detection or Correction Capability*
- 4. Favorable PSD*
- 5. Adequate Timing Content*
- 6. Transparency*

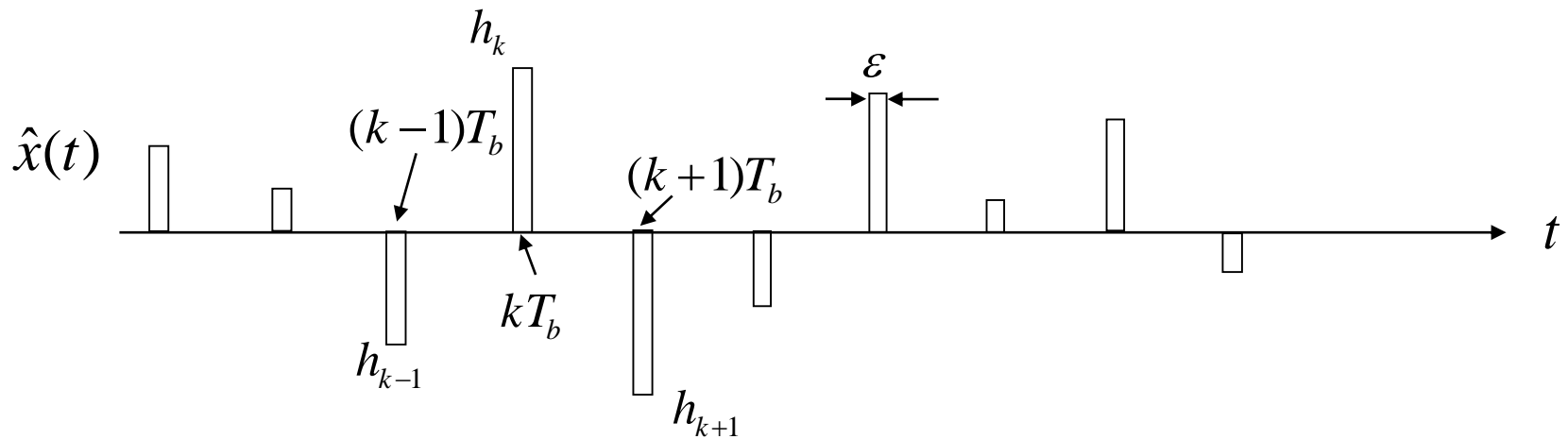
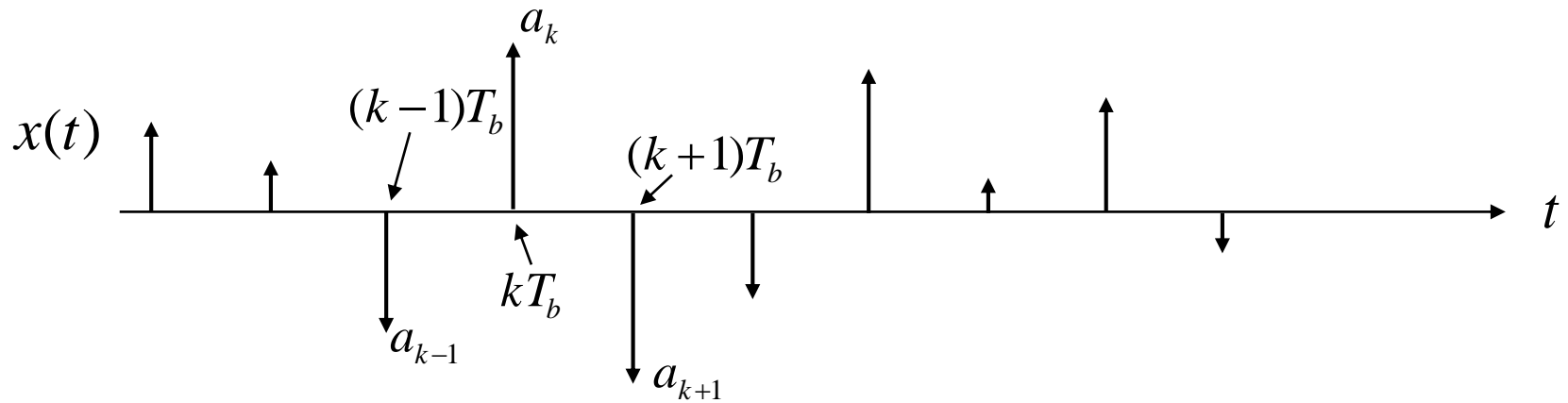
PSD of various Line Codes





$$\Rightarrow S_y(\omega) = |P(\omega)|^2 S_x(\omega)$$

PSD of $x(t)$



$$\epsilon h_k = a_k$$

$$\mathfrak{R}_{\hat{x}}(\tau) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \hat{x}(t) \hat{x}(t + \tau) dt$$

when $\tau < \varepsilon$

$$\begin{aligned} \mathfrak{R}_{\hat{x}}(\tau) &= \lim_{T \rightarrow \infty} \frac{1}{T} \sum_k h_k^2 (\varepsilon - \tau) \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \sum_k a_k^2 \left(\frac{\varepsilon - \tau}{\varepsilon^2} \right) \quad (\varepsilon h_k = a_k) \\ &= \lim_{T \rightarrow \infty} \frac{1}{\varepsilon T} \sum_k a_k^2 \left(1 - \frac{\tau}{\varepsilon} \right) \\ &= \lim_{T \rightarrow \infty} \frac{1}{\varepsilon T} \left(1 - \frac{\tau}{\varepsilon} \right) \frac{T_b}{T_b} \sum_k a_k^2 \\ &= \frac{R_0}{\varepsilon T_b} \left(1 - \frac{\tau}{\varepsilon} \right) \quad \text{where} \quad R_0 = \lim_{T \rightarrow \infty} \frac{T_b}{T} \sum_k a_k^2 \end{aligned}$$

$$\mathfrak{R}_{\hat{x}}(\tau) = \frac{R_0}{\varepsilon T_b} \left(1 - \frac{\tau}{\varepsilon}\right) \quad \text{where} \quad R_0 = \lim_{T \rightarrow \infty} \frac{T_b}{T} \sum_k a_k^2$$

notice $N = \frac{T}{T_b} \quad N \rightarrow \infty \quad \text{when} \quad T \rightarrow \infty$

$$\begin{aligned} \Rightarrow R_0 &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_k a_k^2 \\ &= \overline{a_k^2} \end{aligned}$$

as $\mathfrak{R}_{\hat{x}}(\tau)$ is even function of time

$$\mathfrak{R}_{\hat{x}}(\tau) = \frac{R_0}{\varepsilon T_b} \left(1 - \frac{|\tau|}{\varepsilon}\right) \quad |\tau| < \varepsilon$$

at $\tau = T_b$

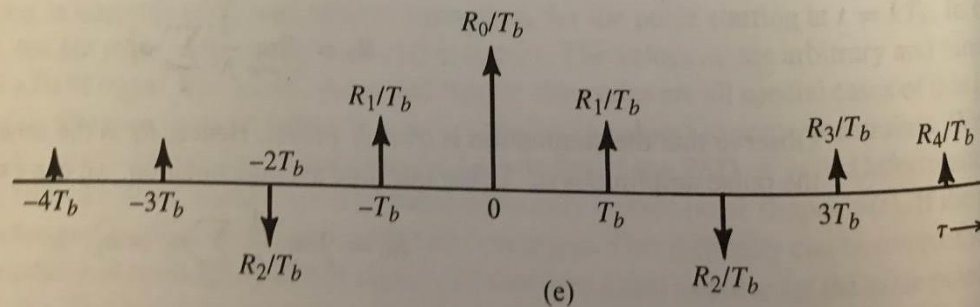
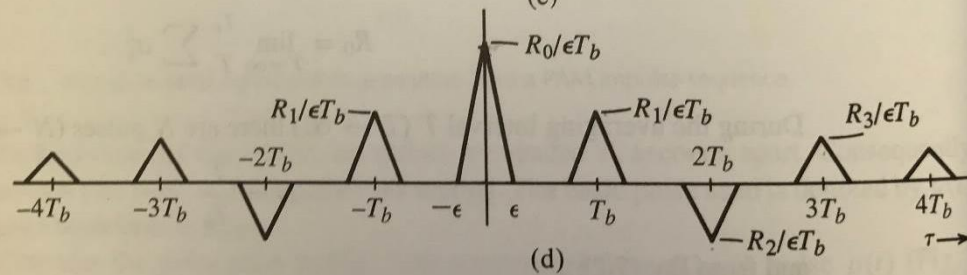
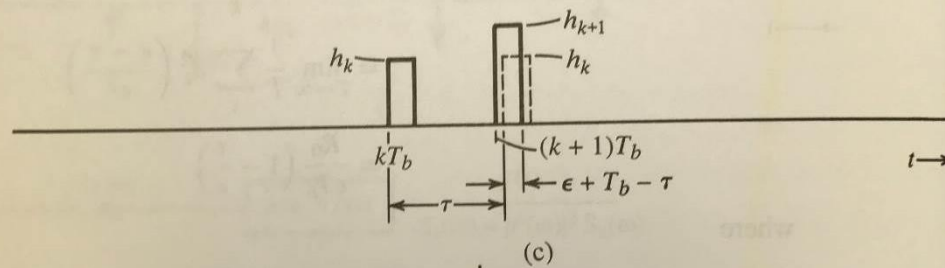
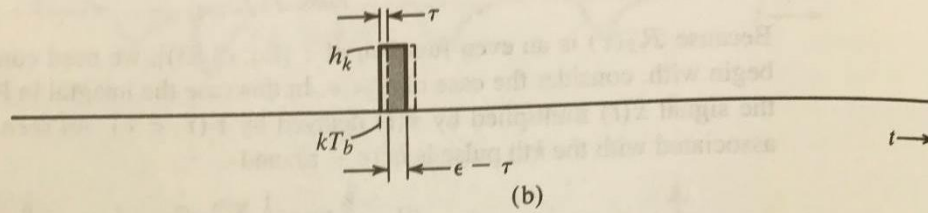
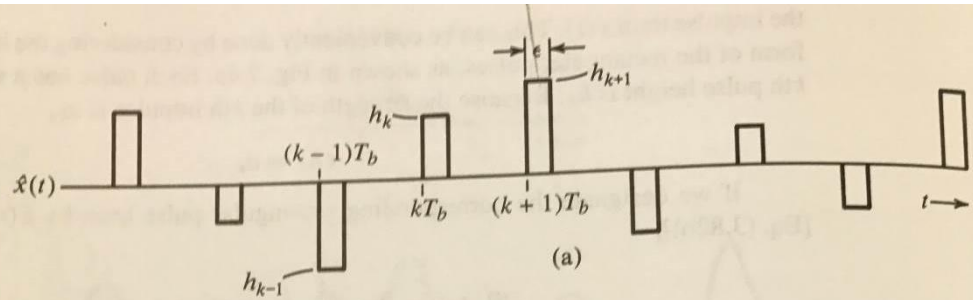
$$\Re_{\hat{x}}(\tau) = \frac{R_1}{\varepsilon T_b} \left(1 - \frac{|\tau|}{\varepsilon}\right)$$

where $R_1 = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_k a_k a_{k+1} = \overline{a_k a_{k+1}}$

Similarly, at $\tau = nT_b$

$$\Re_{\hat{x}}(\tau) = \frac{R_n}{\varepsilon T_b} \left(1 - \frac{|\tau|}{\varepsilon}\right)$$

Where $R_n = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_k a_k a_{k+n} = \overline{a_k a_{k+n}}$



$$\text{at } \tau = nT_b, \quad \mathfrak{R}_{\hat{x}}(\tau) = \frac{R_n}{\varepsilon T_b} \left(1 - \frac{|\tau|}{\varepsilon}\right)$$

$$\mathfrak{R}_x(\tau) = \lim_{\varepsilon \rightarrow 0} \mathfrak{R}_{\hat{x}}(\tau)$$

$$\Rightarrow \mathfrak{R}_x(\tau) = \frac{R_n}{T_b} \delta(\tau - nT_b)$$

For all τ

$$\begin{aligned} \mathfrak{R}_x(\tau) &= \sum_{n=-\infty}^{\infty} \frac{R_n}{T_b} \delta(\tau - nT_b) \\ &= \frac{1}{T_b} \sum_{n=-\infty}^{\infty} R_n \delta(\tau - nT_b) \end{aligned}$$

$$\Rightarrow S_x(\omega) = \frac{1}{T_b} \sum_{n=-\infty}^{\infty} R_n e^{-jn\omega T_b}$$

$$\begin{aligned}\Rightarrow S_x(\omega) &= \frac{1}{T_b} \sum_{n=-\infty}^{\infty} R_n e^{-jn\omega T_b} \\ &= \frac{1}{T_b} (R_0 + 2 \sum_{n=1}^{\infty} R_n \cos n\omega T_b)\end{aligned}$$

We know that

$$S_y(\omega) = |P(\omega)|^2 S_x(\omega)$$

$$\Rightarrow S_y(\omega) = \frac{|P(\omega)|^2}{T_b} (R_0 + 2 \sum_{n=1}^{\infty} R_n \cos n\omega T_b)$$

PSD of Polar Signaling

transmission of 1 with $p(t)$ and 0 with $-p(t)$

$$S_y(\omega) = \frac{|P(\omega)|^2}{T_b} \left(R_0 + 2 \sum_{n=1}^{\infty} R_n \cos n\omega T_b \right)$$

$$\begin{aligned} R_0 &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_k a_k^2 \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} (N) = 1 \end{aligned}$$

$$\begin{aligned} R_1 &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_k a_k a_{k+1} \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} \left[\frac{N}{2} (1) + \frac{N}{2} (-1) \right] = 0 \end{aligned}$$

$$\Rightarrow R_n = 0 \quad \text{if} \quad n > 0$$

$$\Rightarrow S_y(\omega) = \frac{|P(\omega)|^2}{T_b}$$

$$\Rightarrow S_y(\omega) = \frac{|P(\omega)|^2}{T_b}$$

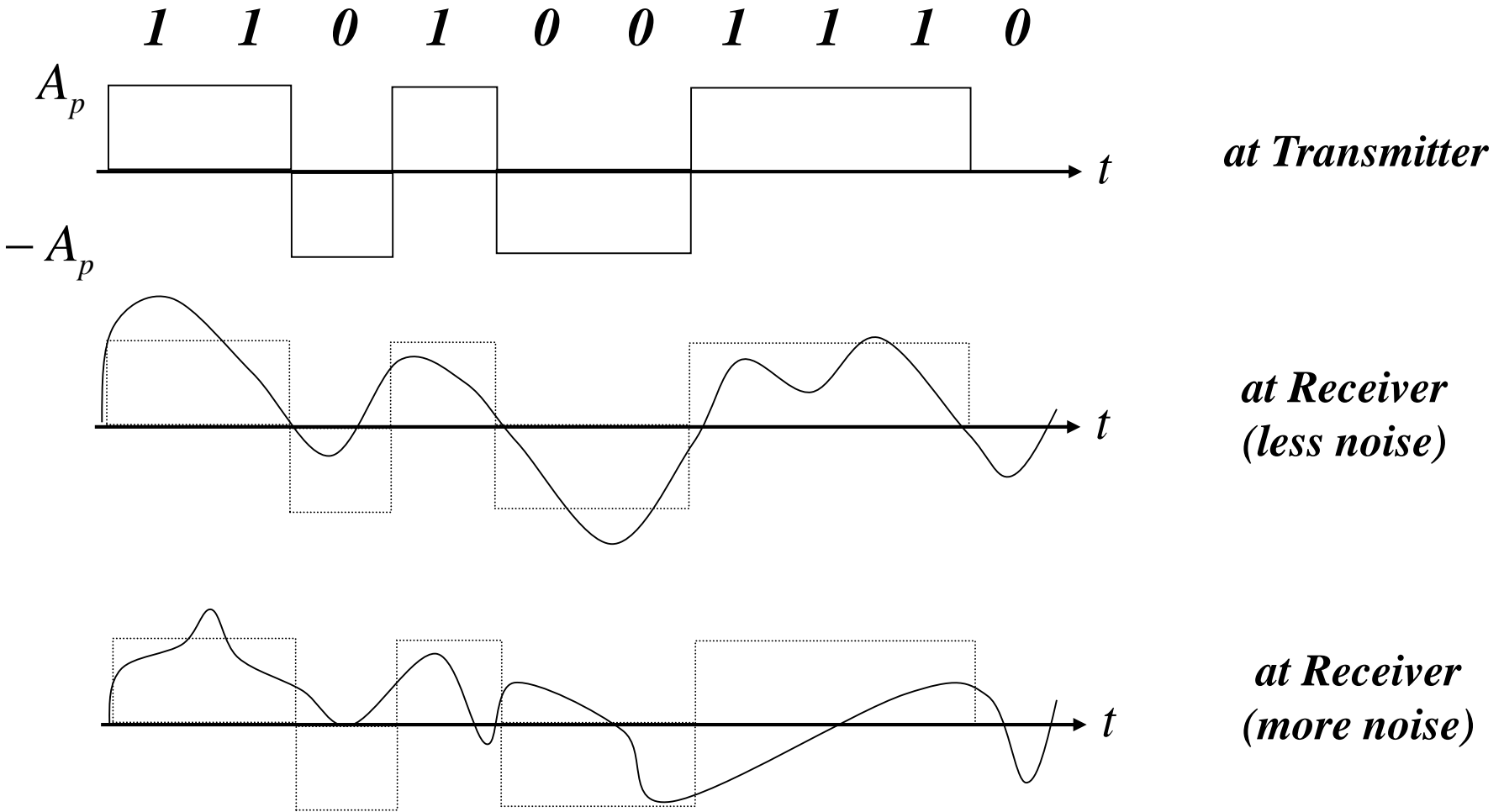
$$\text{if } p(t) = \text{rect}\left(\frac{t}{T_b}\right)$$

$$\text{then } P(\omega) = T_b \text{sinc}\left(\frac{\omega T_b}{2}\right)$$

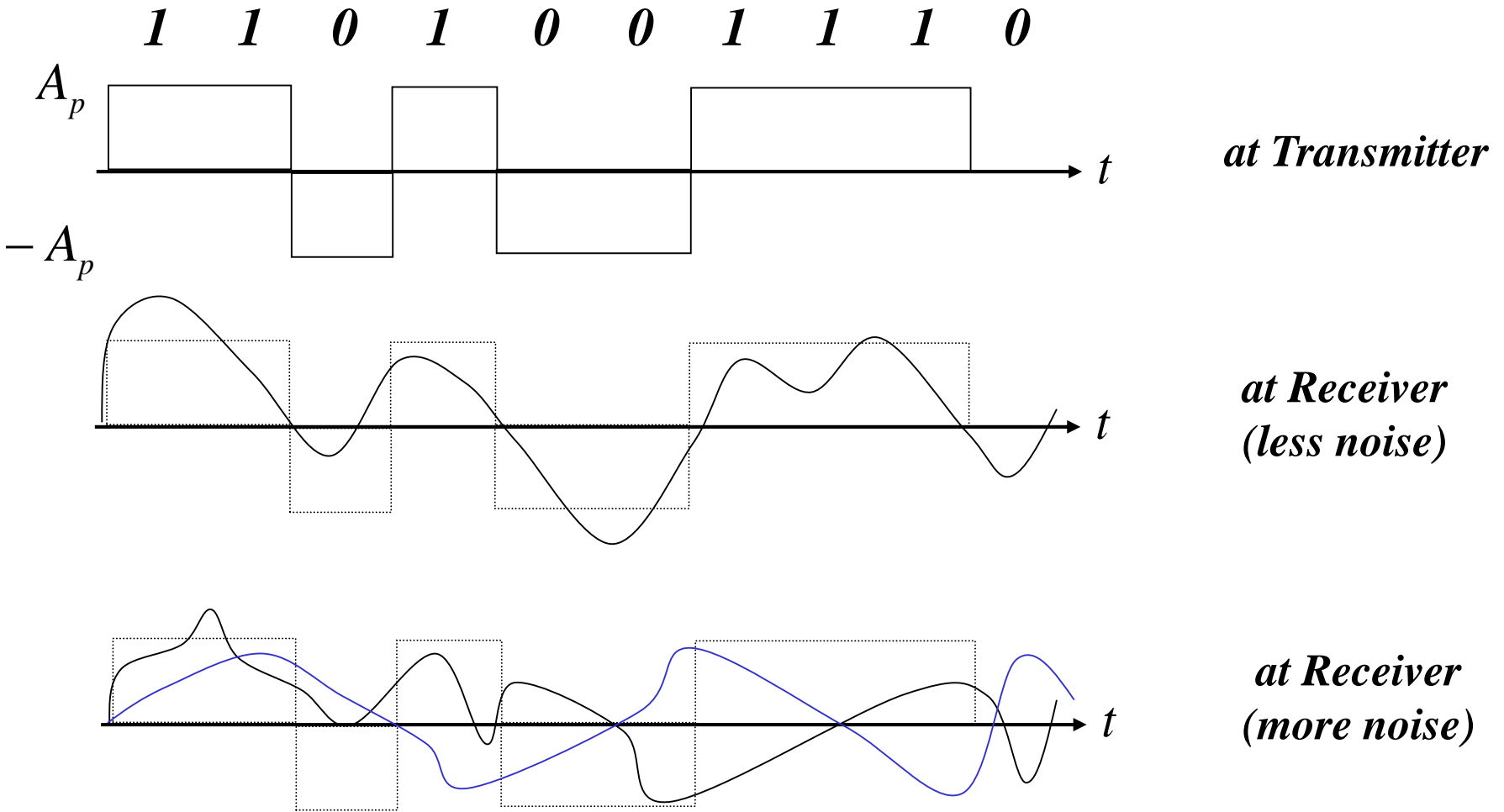
$$\Rightarrow S_y(\omega) = T_b \text{sinc}^2\left(\frac{\omega T_b}{2}\right)$$

****Tweaking PSD with Pulse Shaping
having null at zero frequency***

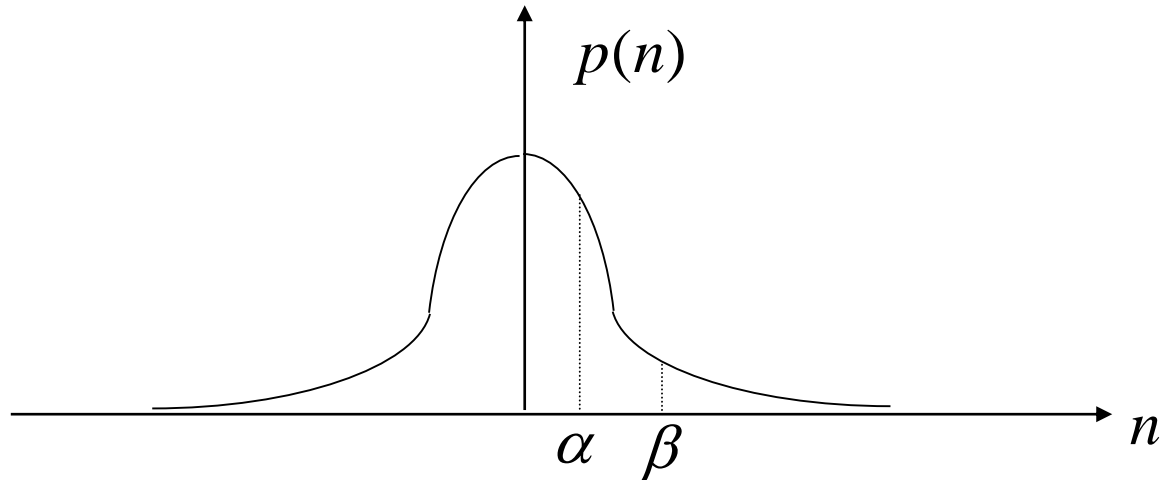
Noise and Digital Communication



Noise and Digital Communication



Gaussian Noise and Probability Density Function



$$p(n) = \frac{1}{\sigma_n \sqrt{2\pi}} e^{\frac{-n^2}{2\sigma_n^2}}$$

$$\text{Prob}(\alpha < n < \beta) = \int_{\alpha}^{\beta} p(n) dn = \int_{\alpha}^{\beta} \frac{1}{\sigma_n \sqrt{2\pi}} e^{\frac{-n^2}{2\sigma_n^2}} dn$$

Error Probability for Polar Signaling

$$\text{Pr ob}(E) = \frac{1}{2} [\text{Pr ob}(E|0) + \text{Pr ob}(E|1)]$$

$\text{Pr ob}(E|0) = \textit{probability that } n > A_p$

$\text{Pr ob}(E|1) = \textit{probability that } n < -A_p$

$$\text{Pr ob}(E|0) = \int_{A_p}^{\infty} p(n) dn = \int_{A_p}^{\infty} \frac{1}{\sigma_n \sqrt{2\pi}} e^{\frac{-n^2}{2\sigma_n^2}} dn$$

$$= \frac{1}{\sigma_n \sqrt{2\pi}} \int_{A_p}^{\infty} e^{\frac{-n^2}{2\sigma_n^2}} dn$$

$$= \frac{1}{\sqrt{2\pi}} \int_{A_p/\sigma_n}^{\infty} e^{\frac{-x^2}{2}} dx \quad \left(x = \frac{n}{\sigma_n}\right)$$

Let's define

$$Q(y) = \frac{1}{\sqrt{2\pi}} \int_y^{\infty} e^{\frac{-x^2}{2}} dx$$

$$\Rightarrow \text{Pr ob}(E|0) = Q\left(\frac{A_p}{\sigma_n}\right)$$

$$\begin{aligned}\text{Pr ob}(E|1) &= \int_{-\infty}^{A_p} p(n) dn = \int_{-\infty}^{A_p} \frac{1}{\sigma_n \sqrt{2\pi}} e^{\frac{-n^2}{2\sigma_n^2}} dn \\ &= Q\left(\frac{A_p}{\sigma_n}\right)\end{aligned}$$

Now

$$\begin{aligned}\text{Pr ob}(E) &= \frac{1}{2} [\text{Pr ob}(E|0) + \text{Pr ob}(E|1)] \\ &= \frac{1}{2} \left[Q\left(\frac{A_p}{\sigma_n}\right) + Q\left(\frac{A_p}{\sigma_n}\right) \right] \\ &= Q\left(\frac{A_p}{\sigma_n}\right)\end{aligned}$$

Error Probability for ON-Off Signaling

$$\text{Pr ob}(E) = \frac{1}{2} [\text{Pr ob}(E|0) + \text{Pr ob}(E|1)]$$

$$\text{Pr ob}(E|0) = \text{probability that } n > \frac{A_p}{2} = Q\left(\frac{A_p}{2\sigma_n}\right)$$

$$\text{Pr ob}(E|1) = \text{probability that } n < \frac{-A_p}{2} = Q\left(\frac{A_p}{2\sigma_n}\right)$$

$$\Rightarrow \text{Pr ob}(E) = \frac{1}{2} \left[Q\left(\frac{A_p}{2\sigma_n}\right) + Q\left(\frac{A_p}{2\sigma_n}\right) \right]$$

$$= Q\left(\frac{A_p}{2\sigma_n}\right)$$

Error Probability for Bi-Polar Signaling

$$\text{Pr ob}(E) = \frac{1}{2} [\text{Pr ob}(E|0) + \text{Pr ob}(E|1)]$$

$$\begin{aligned} \text{Pr ob}(E|0) &= \text{prob}(|n| > \frac{A_p}{2}) \\ &= \text{prob}(n > \frac{A_p}{2}) + \text{prob}(n < -\frac{A_p}{2}) \\ &= 2[\text{prob}(n > \frac{A_p}{2})] \\ &= 2Q\left(\frac{A_p}{2\sigma_n}\right) \end{aligned}$$

$$\text{Pr ob}(E|1) = \text{prob}(n < -\frac{A_p}{2}) \quad \text{when positive pulse is used}$$

$$\text{or } \text{prob}(n > \frac{A_p}{2}) \quad \text{when negative pulse is used}$$

$$= Q\left(\frac{A_p}{2\sigma_n}\right)$$

Now

$$\text{Pr ob}(E) = \frac{1}{2} [\text{Pr ob}(E|0) + \text{Pr ob}(E|1)]$$

$$= \frac{1}{2} \left[2Q\left(\frac{A_p}{2\sigma_n}\right) + Q\left(\frac{A_p}{2\sigma_n}\right) \right]$$

$$= \frac{3}{2} Q\left(\frac{A_p}{2\sigma_n}\right)$$