

# Ch7.1: Baseband Communications & Noise

Noise!

Information source  
and input transducer

Source Coding

Channel Coding

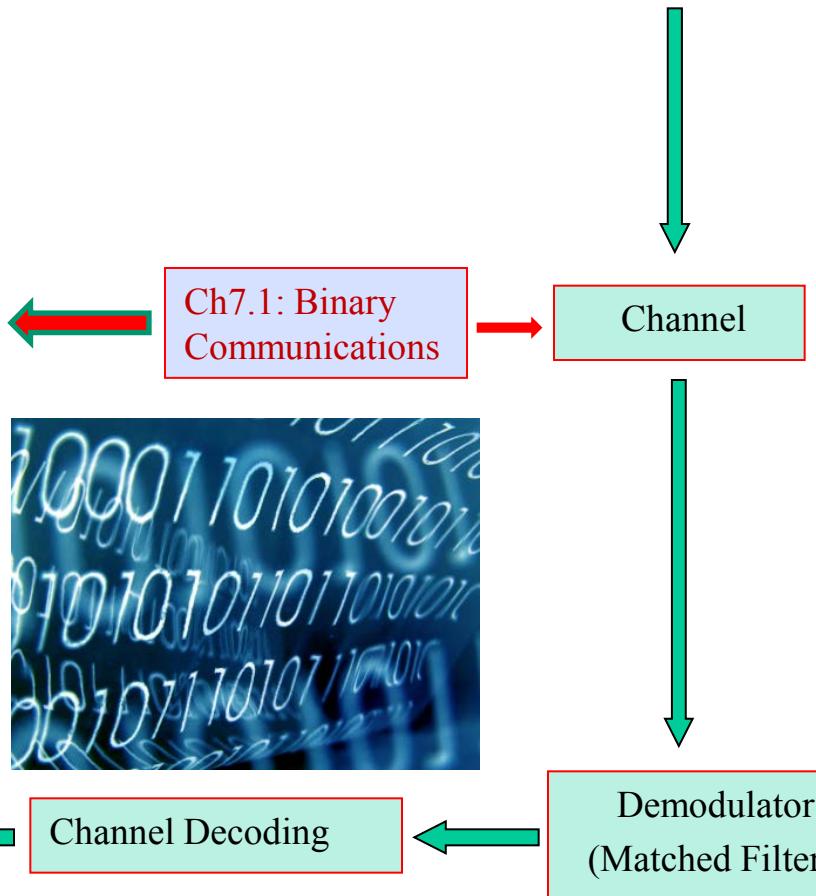
Modulator

- Questions to be answered:

White Gaussian noise!

- System Model:** AWGN Channel
- White Gaussian Noise:** A Random Process
- Suboptimal Receiver:** Integrate-and-Dump
- Performance Evaluation:** How Good is the Receiver?

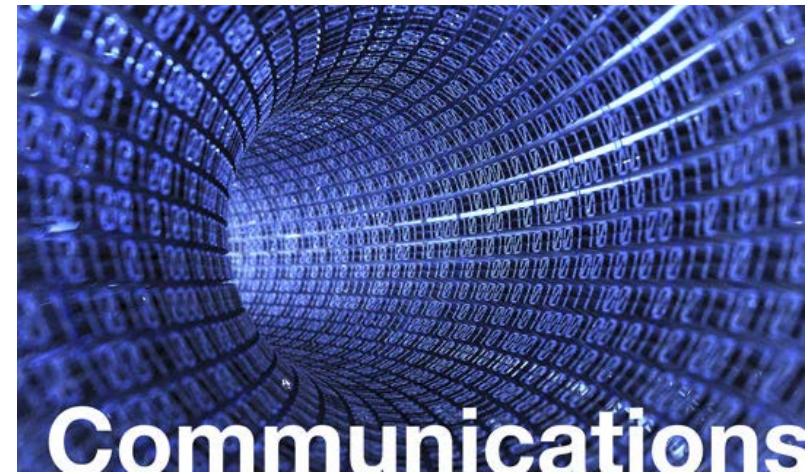
P<sub>e</sub>



Information sink  
and output transducer

# Ch7.1: Baseband Communications & Noise

- System Model**
- White Gaussian Noise
- Suboptimal Receiver
- Performance Evaluation
  - BER
  - A General Case



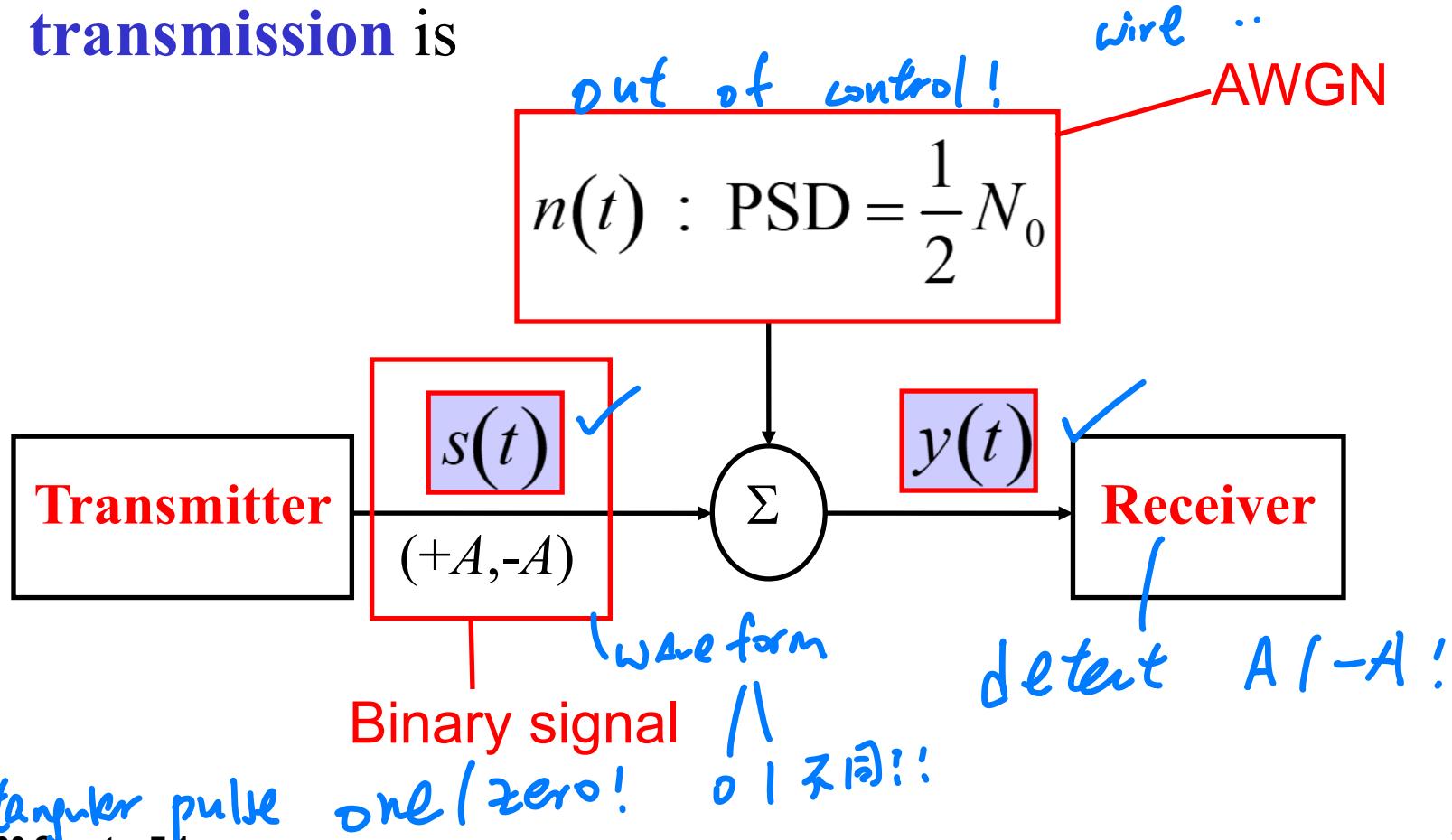
# Digital Communications

- In **digital communication systems**, the **signals take discrete values** to represent binary **data**.
- For example,  $-A$ ,  $A$  can be used to represent logical levels, 0 and 1, respectively.
- The above modulation signaling is called **binary** digital communications since **there are only two bits**.

# Baseband Digital Data Transmission

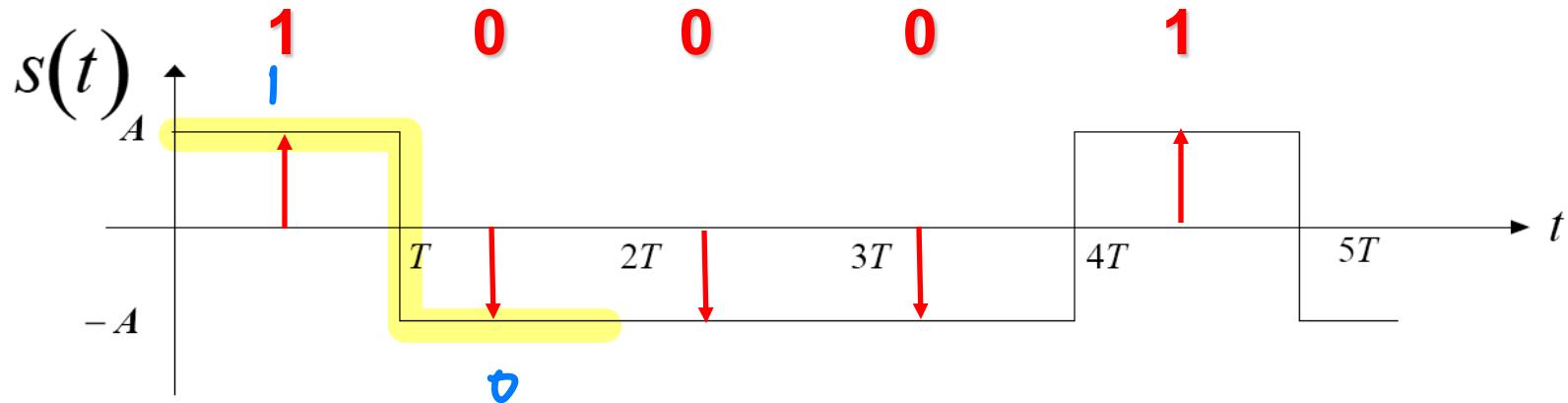
Goal: Minimize  $P_e$

- The system model for **baseband** digital data transmission is

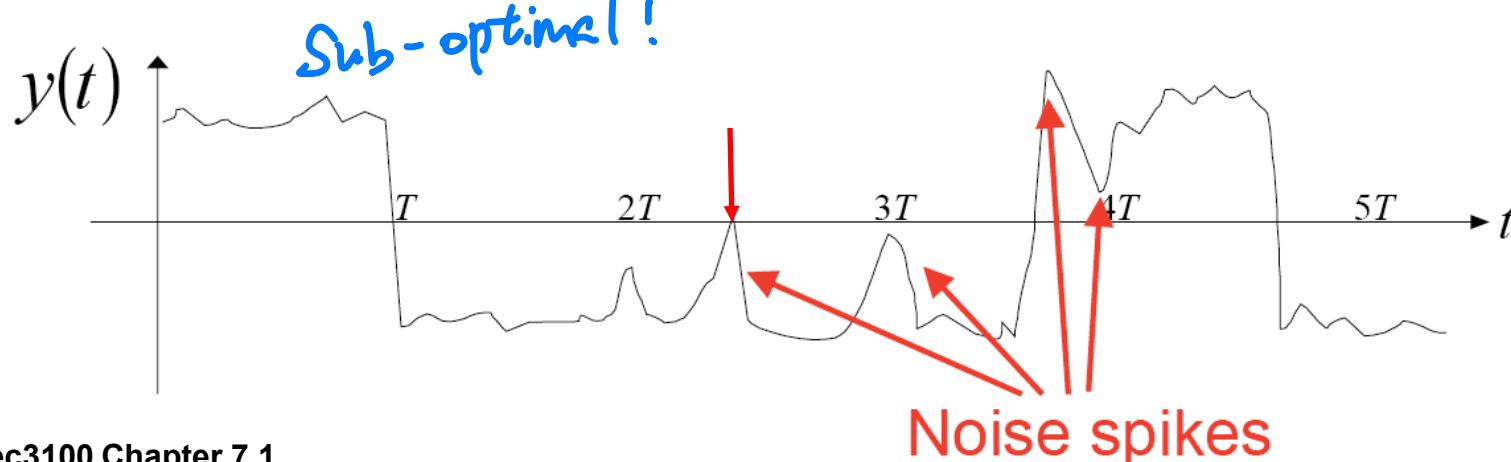


# Baseband Digital Data Transmission

- Example of a digital signal and transmitted waveform is



- Example of a noise-corrupted received signal is



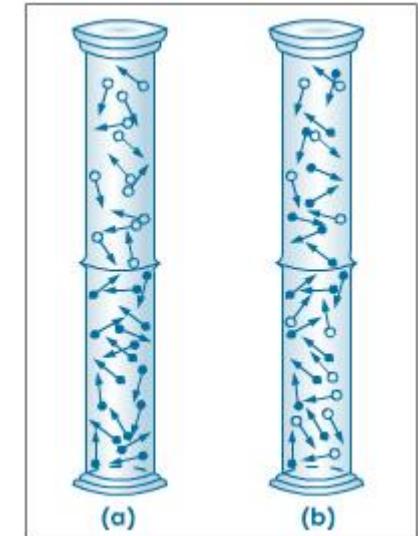
# Ch7.1: Baseband Communications & Noise

- System Model
- **White Gaussian Noise**
- Suboptimal Receiver
- Performance Evaluation
  - BER
  - A General Case



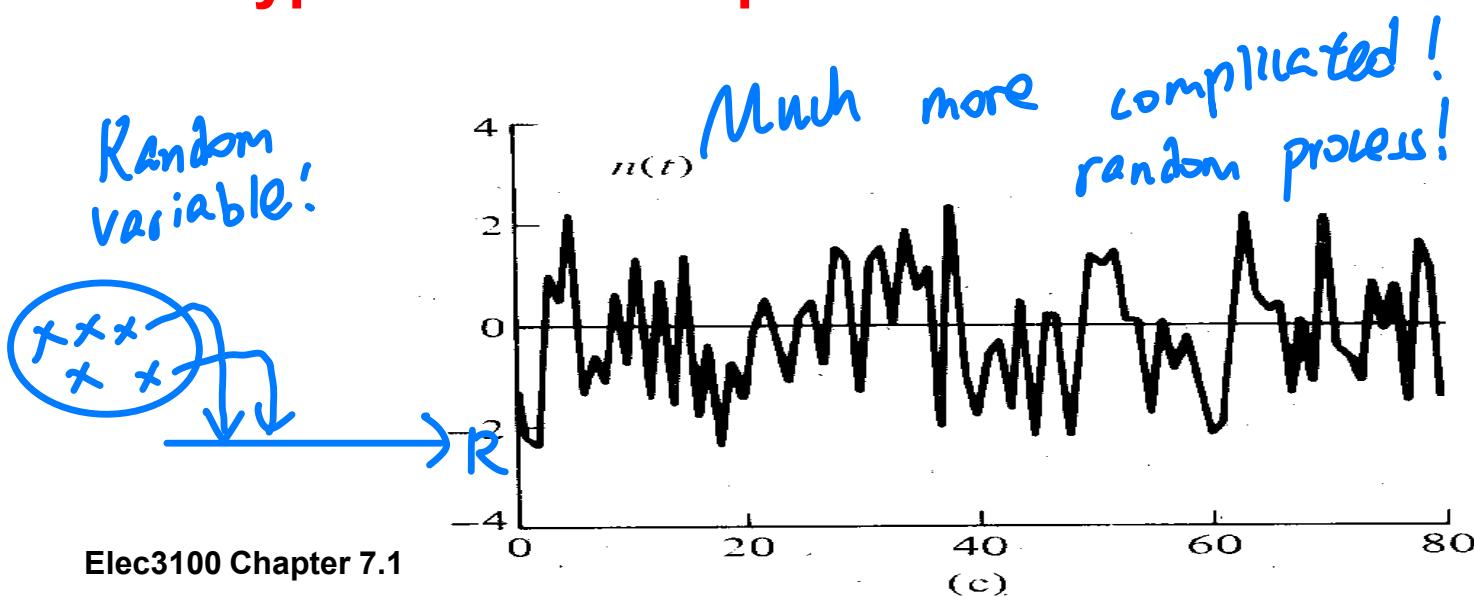
# Types of Noise

- In this course, we assume **internal noise** sources are much more significant than external ones and deal with these only.
- Noise internal to a communications system arises as a result of random motion of charge carriers within the devices (transistors, resistors, diodes, etc) composing the system.
- Internal noise is classified as thermal, shot, flicker or others.

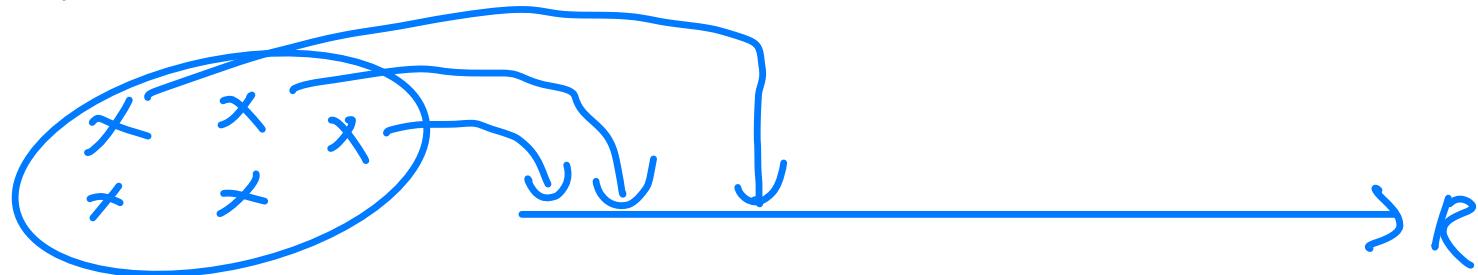


# Thermal Noise

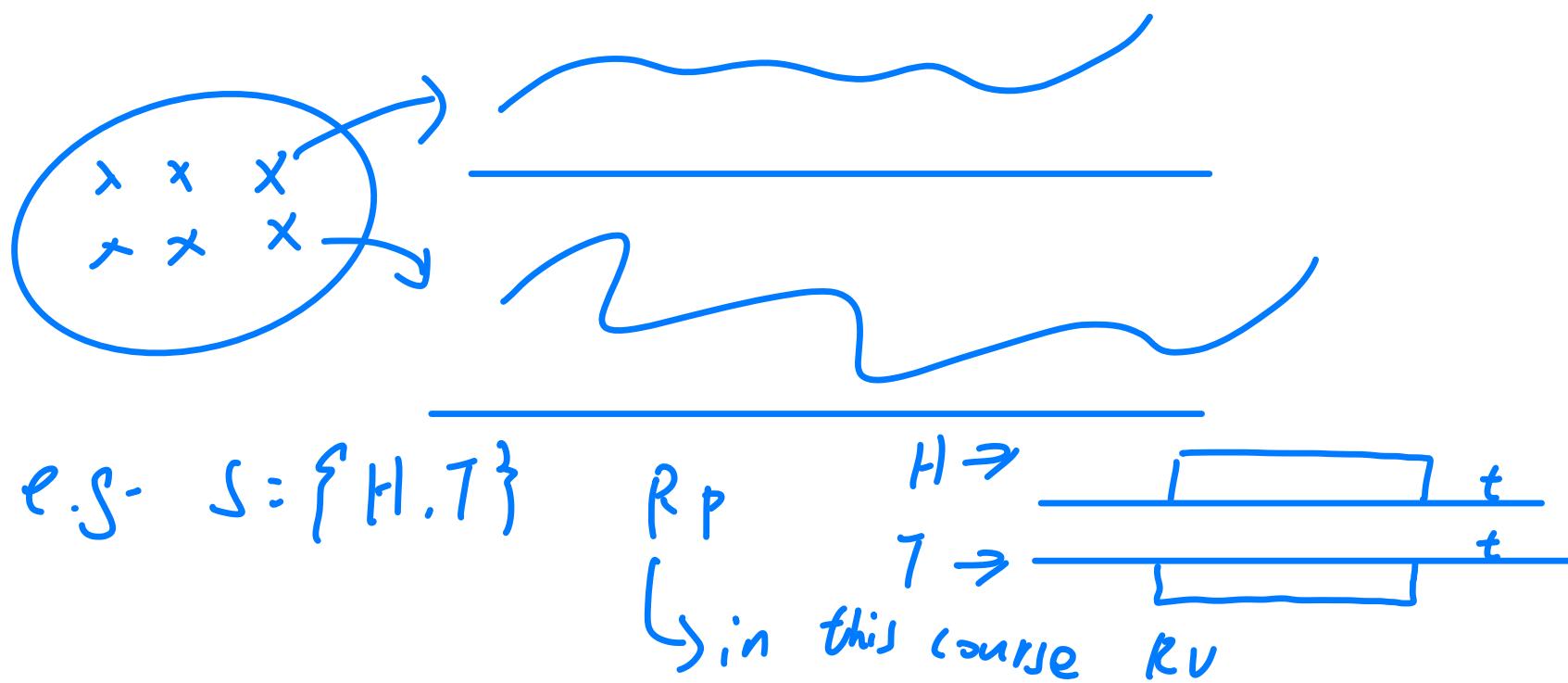
- **Thermal noise** arises from the random motion of charges in conducting medium (such as resistors) and is the most significant noise source we need to consider in ELEC3100.
- Essentially, by connecting a very sensitive oscilloscope across a resistor we can observe thermal noise.
- A **typical noise output** would look like:



Random Variable



Random Process(RP) (noise is waveform)



# Nyquist Theorem

- The signal is **completely random** and **cannot be predicted**. It has an **average value** of zero but has associated with it a certain **mean square voltage**.
- The **mean square voltage** associated with the thermal noise can be found from the **Nyquist Theorem**:

$$\overline{v_n^2} = 4kTBR \quad \begin{matrix} \downarrow \\ \text{Another!} \end{matrix}$$

where

不用的！

$T$  = Temperature of the resistor in Kelvin [K] = [° C] + 273.15

$k$  = Boltzmann's constant  $1.38 \times 10^{-23}$  joule/K

$B$  = Bandwidth of interest

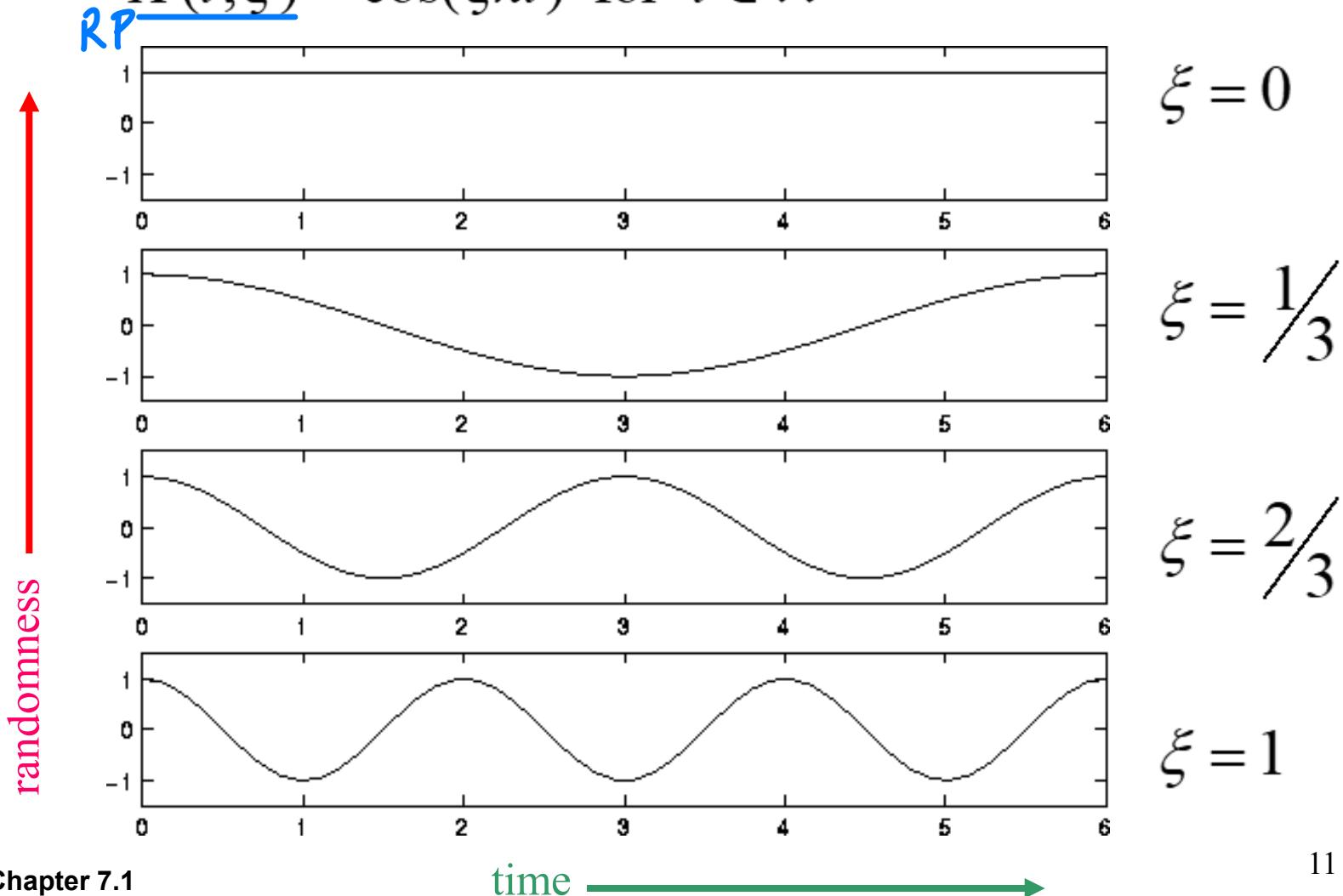
$R$  = Resistance

# Noise can be modeled as a Random Process

- **Definition:** A random process maps a probability space  $S$  to a set of functions,  $X(t, \xi)$ .
- It assigns to every outcome  $\xi \in S$  a time function  $X(t, \xi)$  for  $t \in I$  where  $I$  is a discrete or continuous index set.
- If  $I$  is discrete (e.g. integer valued),  $X(n, \xi)$  is a discrete-time random process.
- If  $I$  is continuous,  $X(t, \xi)$  is a continuous-time random process.
- For a fixed  $t$ ,  $X(t, \xi)$  is a random variable.
- Basically, we can understand a random process as a sequence of random variables.

# Random Process: Example

- Suppose that  $\xi$  is selected *at random* from  $S = [0,1]$  and consider  $X(t, \xi) = \cos(\xi\pi t)$  for  $t \in \mathbb{R}$



# Characterization of A RP: Mean and Variance Functions

- Mean

$$m_X(t) = E[X(t)] = \int xf_{X(t)}(x)dx$$

functions !

- Variance

$$\begin{aligned}\text{Var}[X(t)] &= E\left[\left(X(t) - m_X(t)\right)^2\right] = \int (x - m_X(t))^2 f_{X(t)}(x)dx \\ &= E[X(t)^2] - m_X(t)^2\end{aligned}$$

- Note that the mean and variance may be **functions of time**. However, since the randomness comes from “ $\xi$ ”, we treat “t” as a constant in the above calculations.

# Characterization of A RP: Autocorrelation and Autocovariance

- **Autocorrelation**

$$R_X(t_1, t_2) = E[X(t_1)X(t_2)] = \int \int xyf_{X(t_1), X(t_2)}(x, y)dxdy$$

- **Autocovariance**

$$\begin{aligned} C_X(t_1, t_2) &= E[(X(t_1) - m_X(t_1))(X(t_2) - m_X(t_2))] \\ &= R_X(t_1, t_2) - m_X(t_1)m_X(t_2) \end{aligned}$$

- **Correlation coefficient**

$$\rho_X(t_1, t_2) = \frac{C_X(t_1, t_2)}{\sqrt{C_X(t_1, t_1)}\sqrt{C_X(t_2, t_2)}}$$

# Wide Sense Stationary

- **Definition:** A process  $X(t)$  is wide sense stationary (WSS) if and only if its mean is constant and its autocorrelation function  $R_X(t_1, t_2)$  (or autocovariance function) depends only upon the time difference  $t_1 - t_2$ .  $\triangle$

$$m_x(t) = m \text{ for all } t \quad C_x(t_1, t_2) = C_x(t_1 - t_2) \text{ for all } t_1, t_2$$

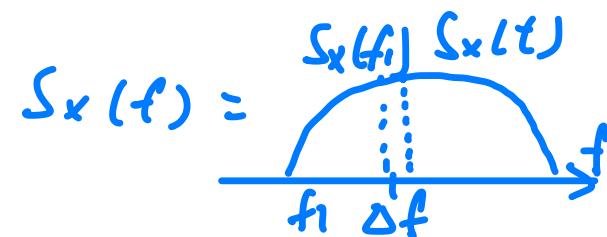
$C_x(\Delta)$  same for time difference!

- Suppose  $X(t)$  is WSS with autocorrelation  $R_X(t)$  where  $t = t_1 - t_2$ .
  - $R_X(0)$  is the average power of the process,  $E[X(t)^2]$ .  $= R_{X(0)}$
  - $R_X(\tau)$  is an even function of  $\underline{\tau}$ .
  - $|R_X(\tau)| \leq R_X(0)$   
(the autocorrelation function is maximum at the origin)

# Wiener-Kinchine Theorem

- Wiener-Kinchine theorem states that the **autocorrelation function** and the **power spectral density** of a stationary random process are Fourier transform pairs.

$$R_X(\tau) \leftrightarrow \underline{\underline{S_X(f)}}$$



- Sample functions  $x(t, \xi_i)$  of stationary random processes are power signals.
- The power spectral density of a stationary random process  $x(t)$  is defined as

$$S_X(f) = \lim_{T \rightarrow \infty} \frac{E[|X_T(f, \xi_i)|^2]}{T}$$

where  $X_T(f, \xi_i) = \int_{-T/2}^{T/2} x(t, \xi_i) e^{-j2\pi ft} dt$

# Gaussian Process

The additive noise in a communications system can be modeled as a **Gaussian process** (by the central limit theorem)

- at a particular time  $t$ , the noise signal amplitude will be Gaussian distributed.
- we further assume that the Gaussian process is stationary and has zero mean:

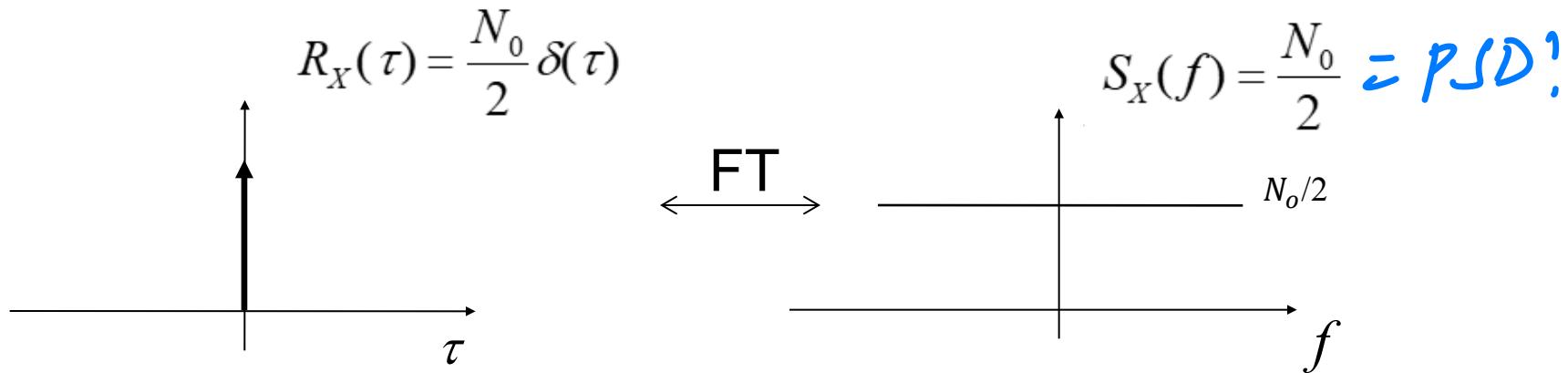
$$\mu(t_i) = E[X(t_i)] = 0 \quad i = 1, 2, \dots, n$$

and its autocorrelation is

$$R_X(\tau) = E[X(t_i)X(t_i + \tau)] = \frac{N_o}{2} \delta(\tau) \quad i = 1, 2, \dots, n$$

# White Noise

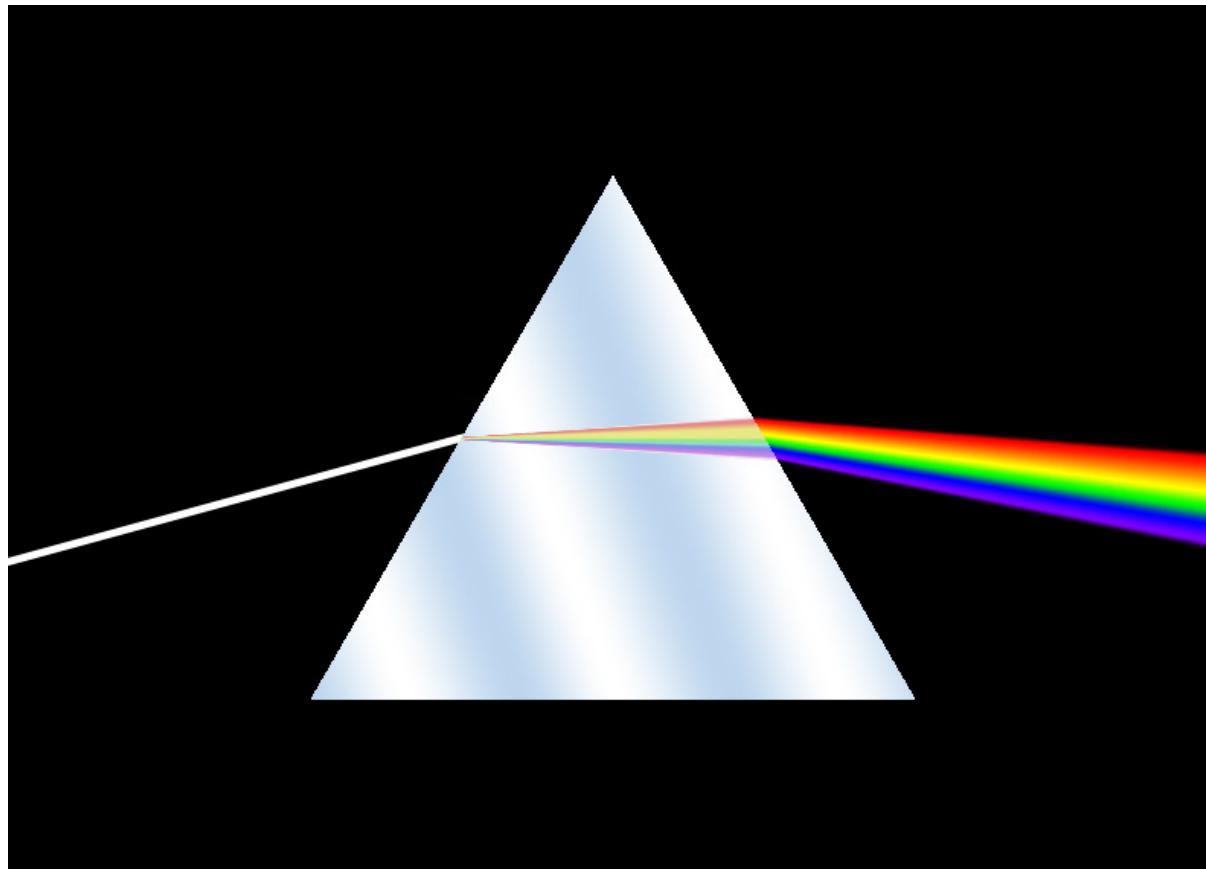
If the power spectrum density is “white”, that is, all frequency components have equal power.



$R_X(\tau)$  is zero except for  $\tau = 0$  implies:

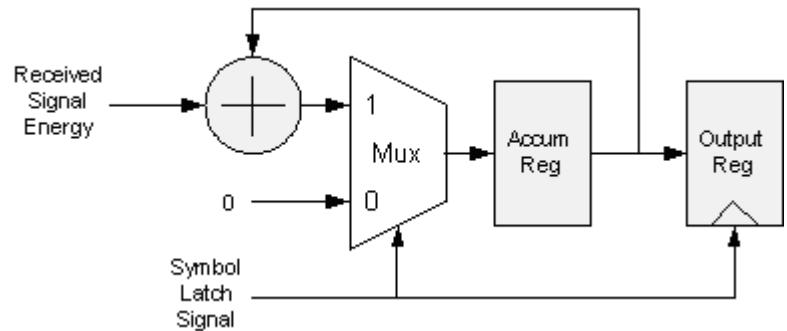
- $R_X(t_i, t_j) = E[X(t_i)X(t_j)] = 0$  for  $i \neq j$ 
  - $X(t_i)$  and  $X(t_j)$  are uncorrelated
- $R_X(0) = \infty = \overline{X^2(t)} = \sigma_X^2$ 
  - Noise has infinite power

# Why White noise?



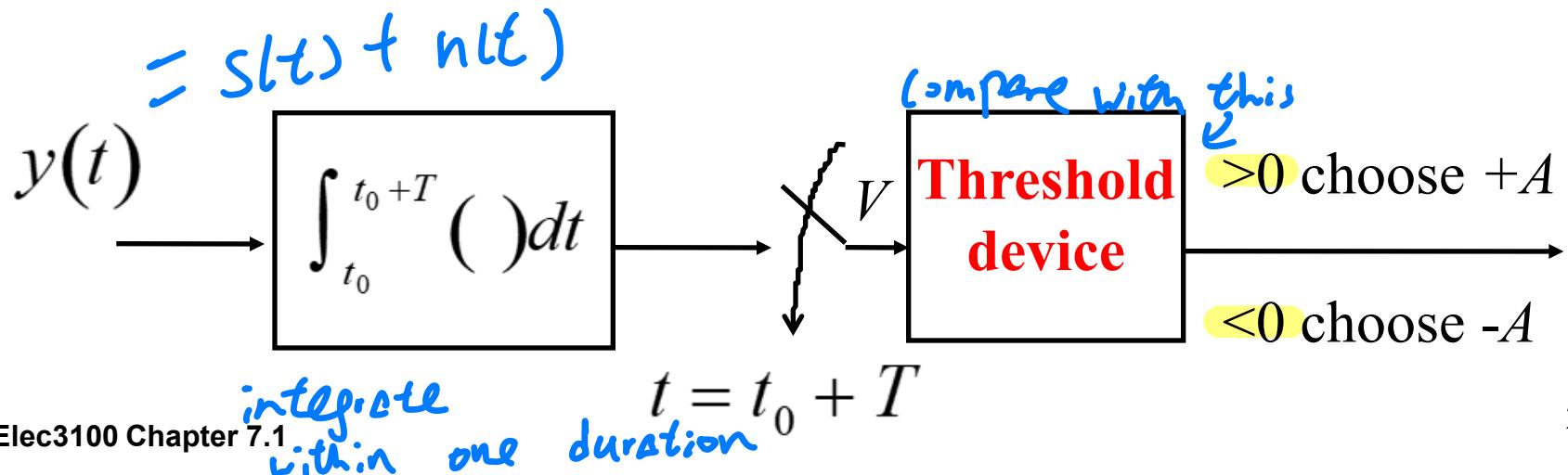
# Ch7.1: Baseband Communications & Noise

- System Model
- White Gaussian Noise
- **Suboptimal Receiver**
- Performance Evaluation
  - BER
  - A General Case



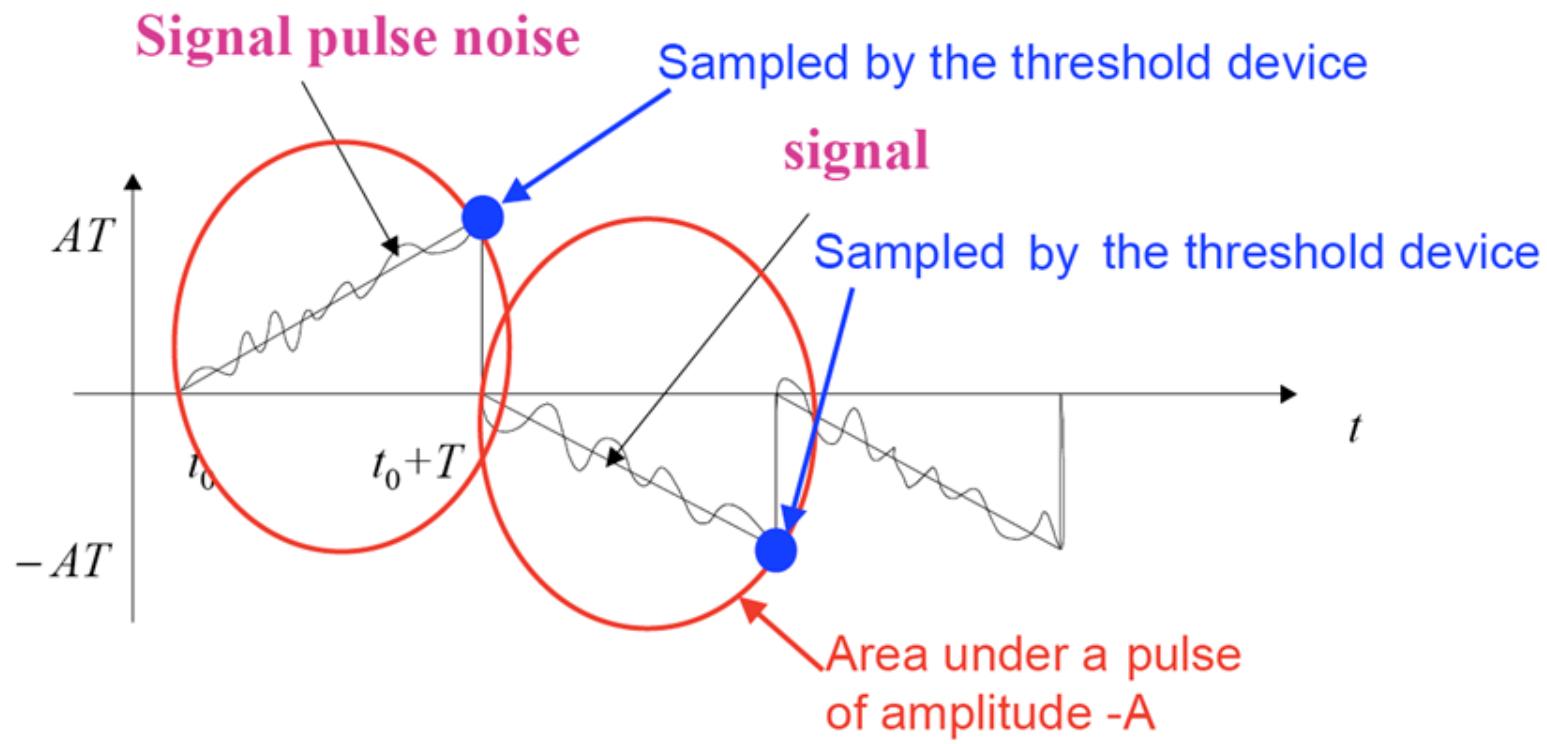
# Receiver Structure

- One of the **significant differences** between analog and digital communications systems is that for digital systems, the **probability of error** is used as a **measure of performance** where as in analog systems SNR is used.
- We have modeled the AWGN. The next question is how to build a receiver to obtain a good performance.
- A **possible receiver structure** (integrate-and-dump) for detecting the digital transmitted signals is shown below



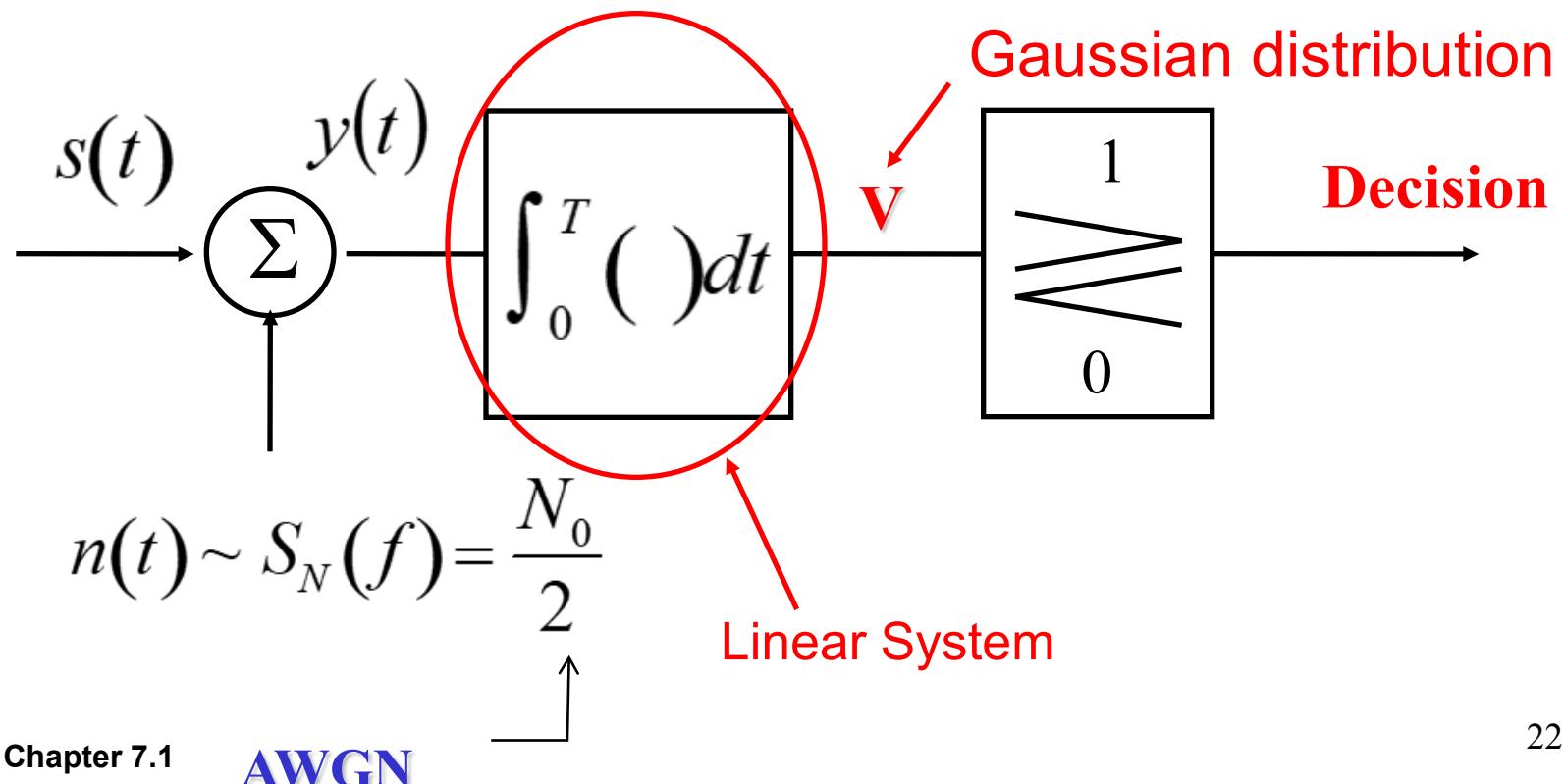
# Integrate-and-Dump

- Not necessarily optimum in all situations.
- The **integrator** averages out the noise received so that the output waveform will look like



# Integrate-and-Dump

- We know that the noise at the input to the receiver is AWGN.
- We can expect the output from the integrator to have a Gaussian noise distribution.
- Putting these ideas into a mathematical framework we get



# Integrate-and-Dump

$$\underline{\underline{s(t)}} = \begin{cases} A & 0 \leq t < T \\ -A & 0 \leq t < T \end{cases}$$

if "1" transmitted  
if "0" transmitted

*signal per t is constant*:

$$V = \int_0^T [s(t) + n(t)] dt$$
$$= \begin{cases} AT + N & \text{Random variable} \\ -AT + N & \text{if "1" is sent} \\ & \text{if "0" is sent} \end{cases}$$

*N is Gaussian distributed*

with

$$N = \int_0^T n(t) dt$$

Linear combination of Gaussian

# Ch7.1: Baseband Communications & Noise

- System Model
- White Gaussian Noise
- Suboptimal Receiver
- **Performance Evaluation**
  - BER
  - A General Case



# Key Figure of Merit

- The **probability of receiving a bit in error** for digital systems is an important **measure of performance**.
- **Digital communications relies heavily on these error calculations - VIP**



# Problem

*total prob. thm!*

Assume the decision threshold is set at 0

$$P_e = P(0 \text{ received} | 1 \text{ sent})P(1 \text{ sent}) + P(1 \text{ received} | 0 \text{ sent})P(0 \text{ sent})$$

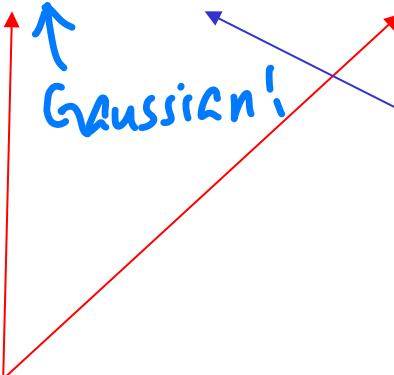
$$P_e = P(V < 0 | 1)P(1) + P(V > 0 | 0)P(0)$$

*bit 1*                   *bit 0*

*V*, is the integrated signal within 7-time interval!

Total probability theorem

$$P_e = P(E|1)P(1) + P(E|0)P(0)$$



Conditional Error Probability

# Error Probability Computation

- To compute  $P_e$ , we need to compute

$$P(E|0) \text{ and } P(E|1).$$

- We have

$$\begin{aligned} V &= \int_0^T [s(t) + n(t)] dt \\ &= \begin{cases} AT + N & \text{if "1" is sent} \\ -AT + N & \text{if "0" is sent} \end{cases} \end{aligned}$$

- $V$  is Gaussian distributed with variance  $\sigma^2$ .

# Conditional Distribution

- The **key** to estimating the error probability is to **find out the distribution of the received signal.**
- **This in turn relies on the distribution of the noise.**
- The **noise mean** can be calculated as

$$E[N] = E \left[ \int_0^T n(t) dt \right] = \int_0^T E[n(t)] dt = 0$$

*mean is 0!*

# AWGN noise! Noise Variance

$E(n) = \text{Just } E(N^2)$

$$\text{Var}[N] = E[N^2] = E\left[\left\{\int_0^T n(t)dt\right\}^2\right]$$

$E(N) = 0$  Autocorrelation

$$= \int_0^T \int_0^T E[n(t)n(v)] dt dv$$

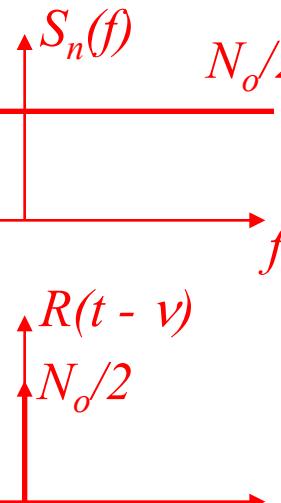
$$\text{AWGN} = \int_0^T \int_0^T R_n(t-v) dt dv$$

AWGN

$$= \int_0^T \int_0^T \frac{N_0}{2} \delta(t-v) dt dv$$

$$= \int_0^T \frac{N_0}{2} dv \quad \frac{N_0 T}{2} \equiv \sigma^2$$

$$= \frac{N_0 T}{2} \equiv \sigma^2$$



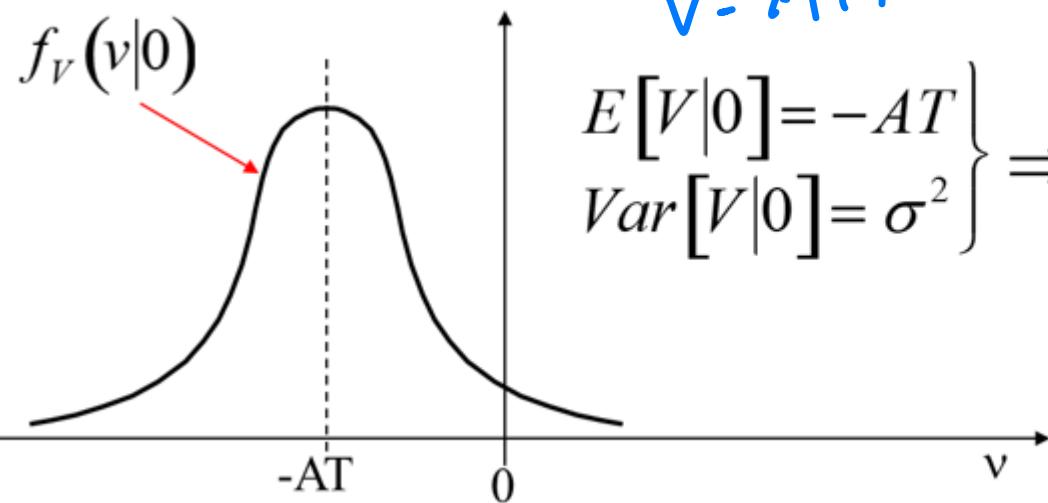
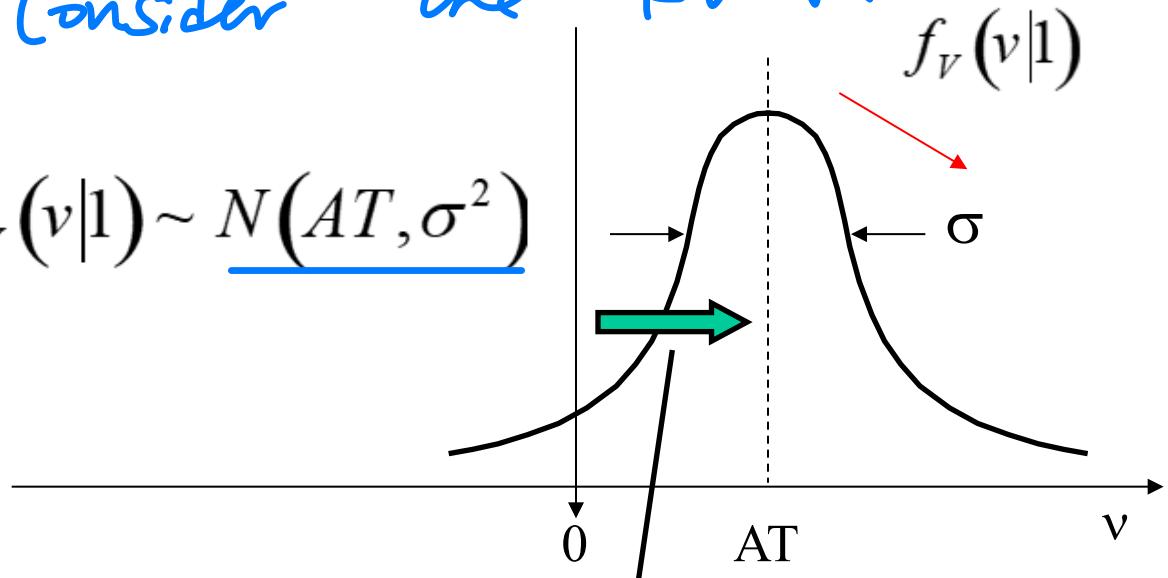
find the distribution  
of  $\sqrt{v}$



# Conditional Probabilities

Consider the RV  $V$ !

$$\left. \begin{array}{l} \text{Mean} = AT \\ E[V|1] = +AT \\ Var[V|1] = \sigma^2 \\ = \frac{\text{Not}}{2} \end{array} \right\} \Rightarrow f_V(v|1) \sim N(AT, \sigma^2)$$



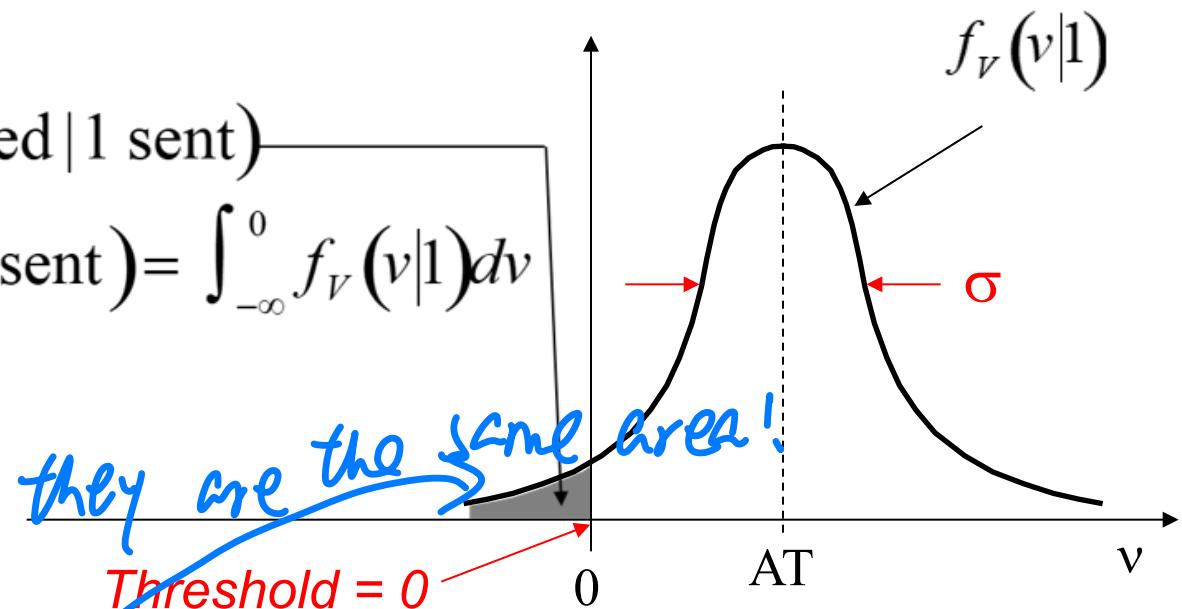
$$\left. \begin{array}{l} V = -AT + N \\ E[V|0] = -AT \\ Var[V|0] = \sigma^2 \end{array} \right\} \Rightarrow f_V(v|0) \sim N(-AT, \sigma^2)$$

Shifted positive due to  
The positive pulse +A

# Conditional Probabilities

$$P(E|1) = P(0 \text{ received} | 1 \text{ sent})$$

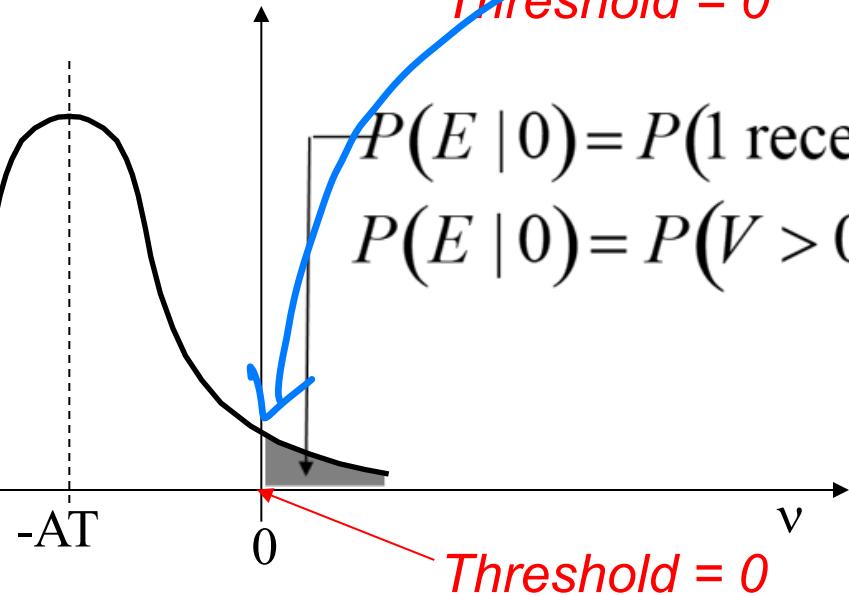
$$P(E|1) = P(V < 0 | 1 \text{ sent}) = \int_{-\infty}^0 f_V(v|1) dv$$



$$f_V(v|0)$$

$$P(E|0) = P(1 \text{ received} | 0 \text{ sent})$$

$$P(E|0) = P(V > 0 | 0 \text{ sent}) = \int_0^{\infty} f_V(v|0) dv$$



# Conditional Error Probability

- Thus, we know that the **output noise from the integrator will have** the following **Gaussian distribution**

$$\therefore N \sim N(0, \sigma^2)$$
$$\Rightarrow f_n(n) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{n^2}{2\sigma^2}\right)$$

- Now,

$$P(E|0) = P(V > 0 | 0 \text{ sent}) = \int_0^\infty f_V(v|0) dv$$

$$P(E|1) = P(V < 0 | 1 \text{ sent}) = \int_{-\infty}^0 f_V(v|1) dv$$

First : write down

$$v : \begin{cases} A\tau + n & \text{if } 1 \text{ is sent} \\ -A\tau + n & \text{if } 0 \text{ is sent} \end{cases}$$

$$n \sim N(0, \sigma^2)$$

$$\sigma^2 = \frac{N_0 T}{2}$$

(white noise)

$$v|1 \sim N(A\tau, \sigma^2)$$

$$v|0 \sim N(-A\tau, \sigma^2)$$

$$P_e = P(E|1)P(1) + P(E|0)P(0)$$

$$= P(v < 0 | 1)P(1) + P(v > 0 | 0)P(0)$$

As  $P(v < 0 | 1) = Q\left(\frac{A\tau}{\sigma}\right)$

# Total Error Probability

- Assumptions:
  - Symmetry,

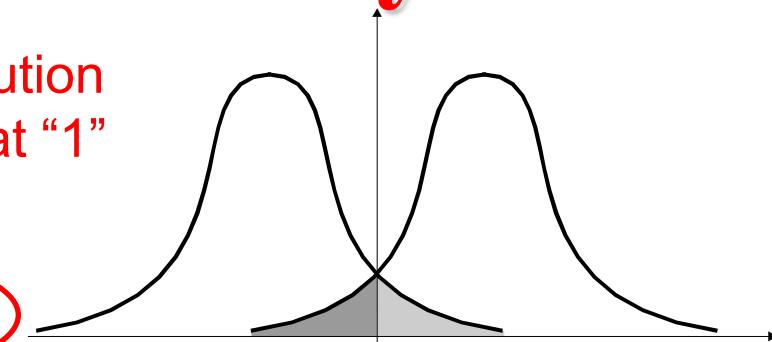
Noise distribution  
is the same at “1”  
and “0”

$$P(E|0) \equiv P(E|1)$$

- “0” and “1” are equally likely, then

~~2<sup>nd</sup> likely~~

$$P(0) \equiv P(1) = \frac{1}{2} \Rightarrow$$



Assume they  
are the same  
(After conditioning!!)

$$P_e = P(E|0) = P(E|1)$$

- We therefore have

$$P_e = P(E|1)P(1) + P(E|0)P(0) = \frac{1}{2} [P(E|0) + P(E|0)] = P(E|0)$$

$$P_e = P(E|0) = \int_0^{\infty} f_V(v|0) dv = \int_0^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(v+AT)^2}{2\sigma^2}} dv$$

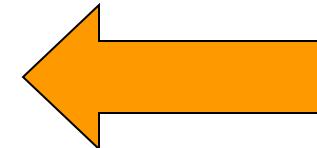
# Evaluation Relies on Q-Function

- Let  $x = \frac{v + AT}{\sigma} \Rightarrow dx = \frac{dv}{\sigma}$  ← VIP Transformation

$$P_e = \int_{\frac{AT}{\sigma}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \triangleq Q\left[\frac{AT}{\sigma}\right]$$

- where

$$Q(x) \triangleq \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-\frac{t^2}{2}} dt$$



Q(.) function

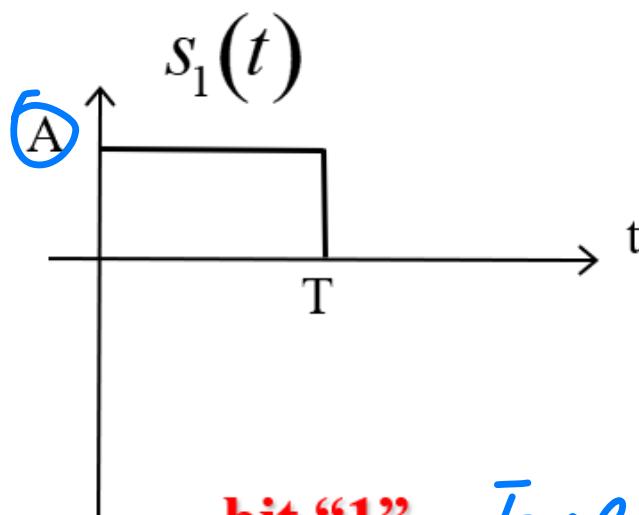
# Ch7.1: Baseband Communications & Noise

- System Model
- White Gaussian Noise
- Suboptimal Receiver
- **Performance Evaluation**
  - BER
  - **A General Case**

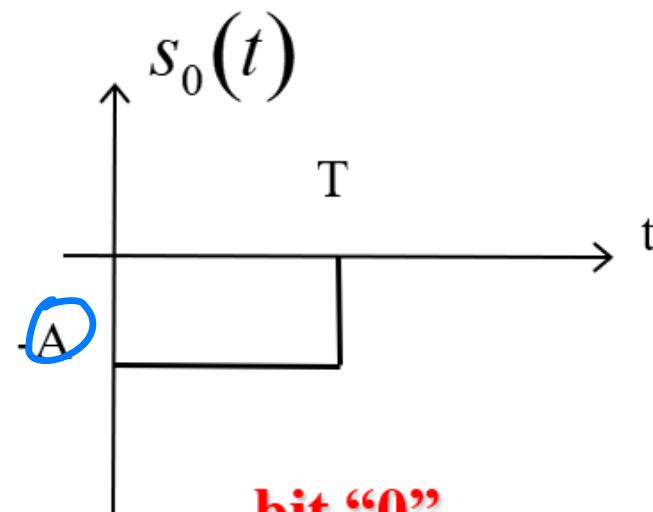


# A General Case

- Next, note that one can represent  $s(t)$  as  $s_0(t)$  ("0" sent) or  $\underline{s_1(t)}$  ("1" sent).



bit "1" Energy !



bit "0"

$$E_1 \triangleq \int_0^T s_1^2(t) dt = A^2 T$$

Energy

$$E_0 \triangleq \int_0^T s_0^2(t) dt = A^2 T$$

In this case,  
they have the  
same Energy !

# A General Result

→  $\therefore E_b \triangleq \text{Energy/bit} = E_0 P(\text{"0" sent}) + E_1 P(\text{"1" sent})$

$$= \frac{1}{2}(E_0 + E_1)$$

But  $E_1 = E_0 = \underline{A^2 T}$ .

- We can use these **energy calculations** to find a more general result for the error probabilities. That is,

$$P_e = Q\left[\frac{AT}{\sigma}\right] = Q\left[\sqrt{\frac{A^2 T^2}{\sigma^2}}\right]$$

where

$$\sigma^2 = \frac{N_0 T}{2}$$

# SNR vs. Eb/No

$$E_b = A^2 T$$

$$\sigma = \frac{N_0 T}{2}$$

- Therefore,

$$P_e = Q\left[\sqrt{\frac{A^2 T^2}{\sigma^2}}\right] = Q\left[\sqrt{\frac{A^2 T}{\sigma^2 / T}}\right]$$

*Energy per bit!*

$$\equiv Q\left[\sqrt{\frac{2E_b}{N_0}}\right]$$

*key quantity*

**Signal Energy to Noise Power Spectral Density Ratio**

Relation with SNR?

$$= \frac{1}{2} \operatorname{erfc}\left[\sqrt{\frac{E_b}{N_0}}\right]$$

where

$$\operatorname{erfc}(u) = 1 - \operatorname{erf}(u)$$

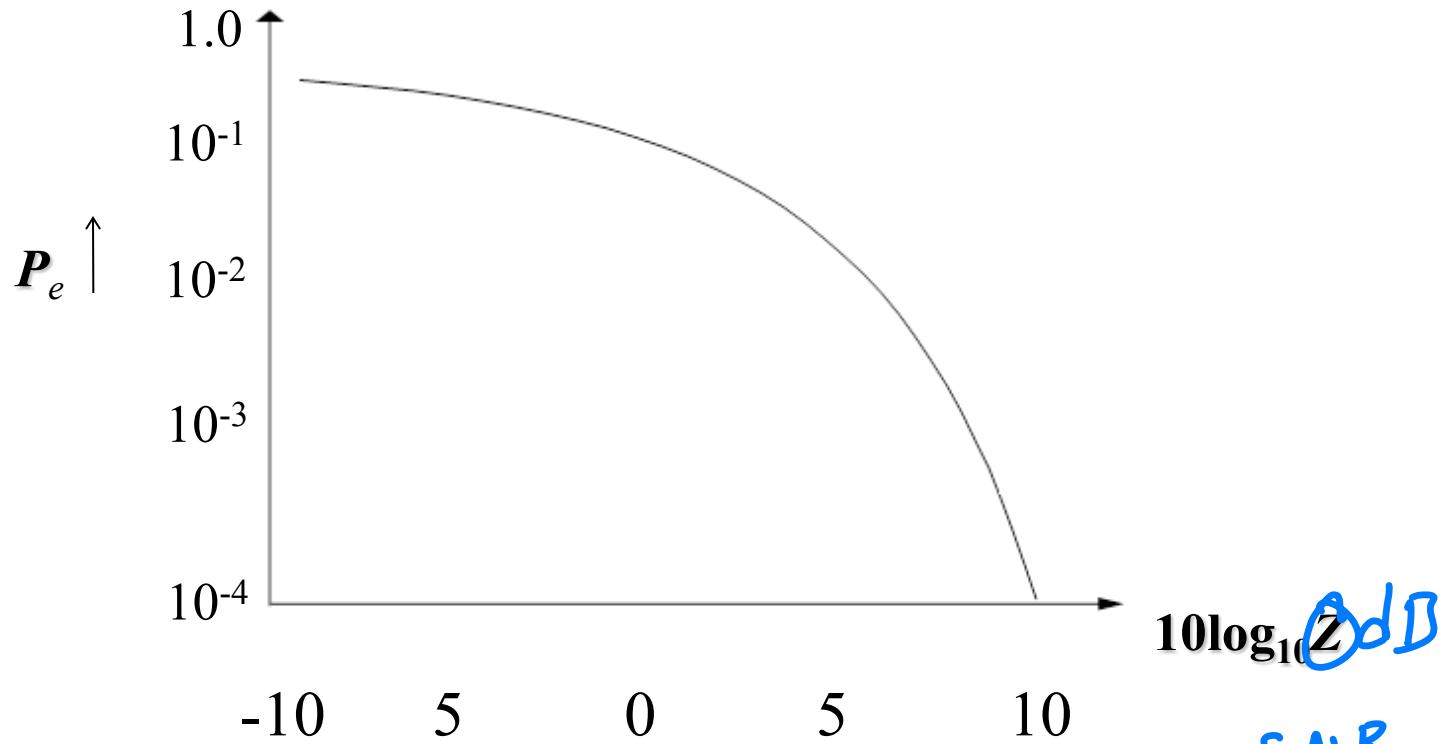
*relationship between erfc and complementary Error Function*

with

$$\operatorname{erf}(u) \triangleq \frac{2}{\sqrt{\pi}} \int_0^u e^{-t^2} dt$$

# Bit Error Rate (BER)

- A graph of  $P_e$  for **baseband signaling** is



where  $P_e = Q\left[\sqrt{2Z}\right] = \frac{1}{2}\operatorname{erfc}\left[\sqrt{Z}\right]$        $Z = \frac{E_b}{N_0}$

design by us!

## Example 7.1

A baseband digital Tx system sends  $\pm A$  valued rectangular pulses through a channel at a rate of 1Mbps with amplitude 1V when the noise PSD is  $10^{-7}$  W/Hz.  $\sim T \quad T = \frac{1}{1M}$   $A = \sqrt{T} = \frac{N_0}{2}$

Answer:  $Q\left(\sqrt{2E_b/N_0}\right) = Q\left(\sqrt{2A^2T/N_0}\right)$

$$T = 1/1000000 = 10^{-6}$$

$$\Rightarrow Q\left(\sqrt{2 \times 10^{-6} / (2 \times 10^{-7})}\right) = Q(\sqrt{10}) = Q(3.16)$$

$$Q(u) \approx \frac{e^{-u^2/2}}{u\sqrt{2\pi}}$$

$$Q(3.16) \approx 0.00085$$

Probability of Error,  $P_e$

## Example 7.2

Digital data is to be transmitted through a baseband system with  $N_0 = 10^{-7} \text{ W/Hz}$  and the received signal amplitude  $A = 20 \text{ mV}$ .

$$P_e = Q\left(\sqrt{2E_b/N_0}\right) = Q(2\sqrt{2})$$

(a) If 1000 bits per second (bps) are transmitted what is the error probability? **Ans.**  $P_e = 2.58 \cdot 10^{-3}$ .

(b) If 10000 bps are transmitted, to what value must  $A$  be adjusted in order to attain the same error probability as in part a)?

**Ans.**  $A = 63.2 \cdot 10^{-3} \text{ V} = 63.2 \text{ mV}$ .

$$\downarrow P_e = Q\sqrt{\frac{2A^2}{N_0/T}}$$
$$2A^2 \left(\frac{1}{10000}\right) / 10^{-7} = 8$$
$$A = 63.2 \text{ mV}$$

Then A need to increase!

# BER vs. Data Rate

- In the last example, for a fixed amplitude, error probability increases as bit rate increases.
- This can be understood by looking at the error probability expression

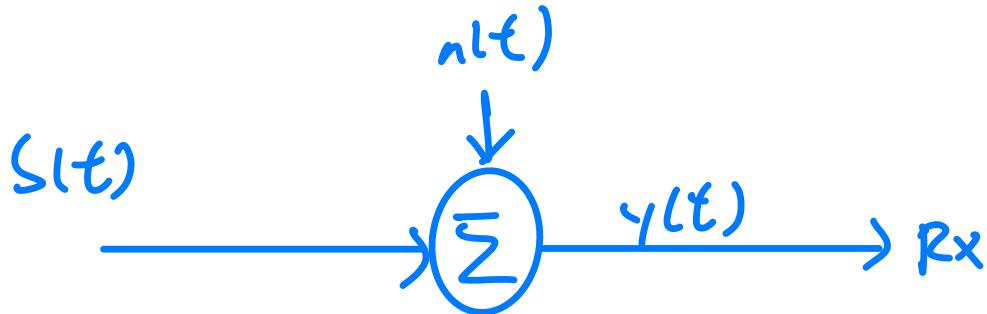
$$P_e = Q\left[\sqrt{\frac{A^2 T^2}{\sigma^2}}\right] = Q\left[\sqrt{\frac{A^2 T}{\sigma^2/T}}\right] = Q\left[\sqrt{\frac{2E_b}{N_o}}\right]$$

- Increase in bit rate → smaller bit period T

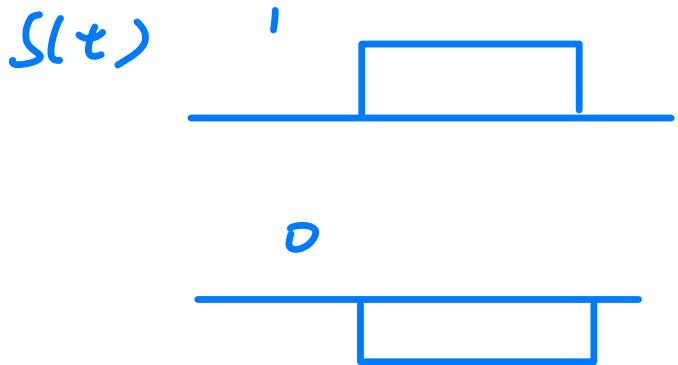
→ higher noise power  
→ lower signal energy

$$P_e = Q\left[\sqrt{\frac{2A^2}{N_o/T}}\right]$$

Are they happening at the same time?



so far



Next : In general

- ① Optimal receiver
- ② Digital modulation
- ③ Channel model
- ④ Multiplexing!

extension of Rx

$n(t)$ : AWGN

Rx : integrate-and-dump

Allow multiple users!