

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_e

Ch7.1: Binary
Communications

Channel



Demodulator
(Matched Filter)

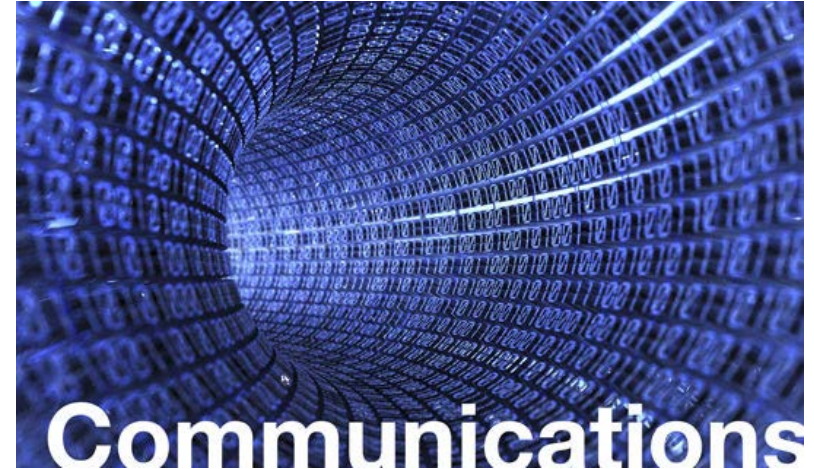
Channel Decoding

Source Decoding

Information sink
and output transducer

Ch7.1: Baseband Communications & Noise

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



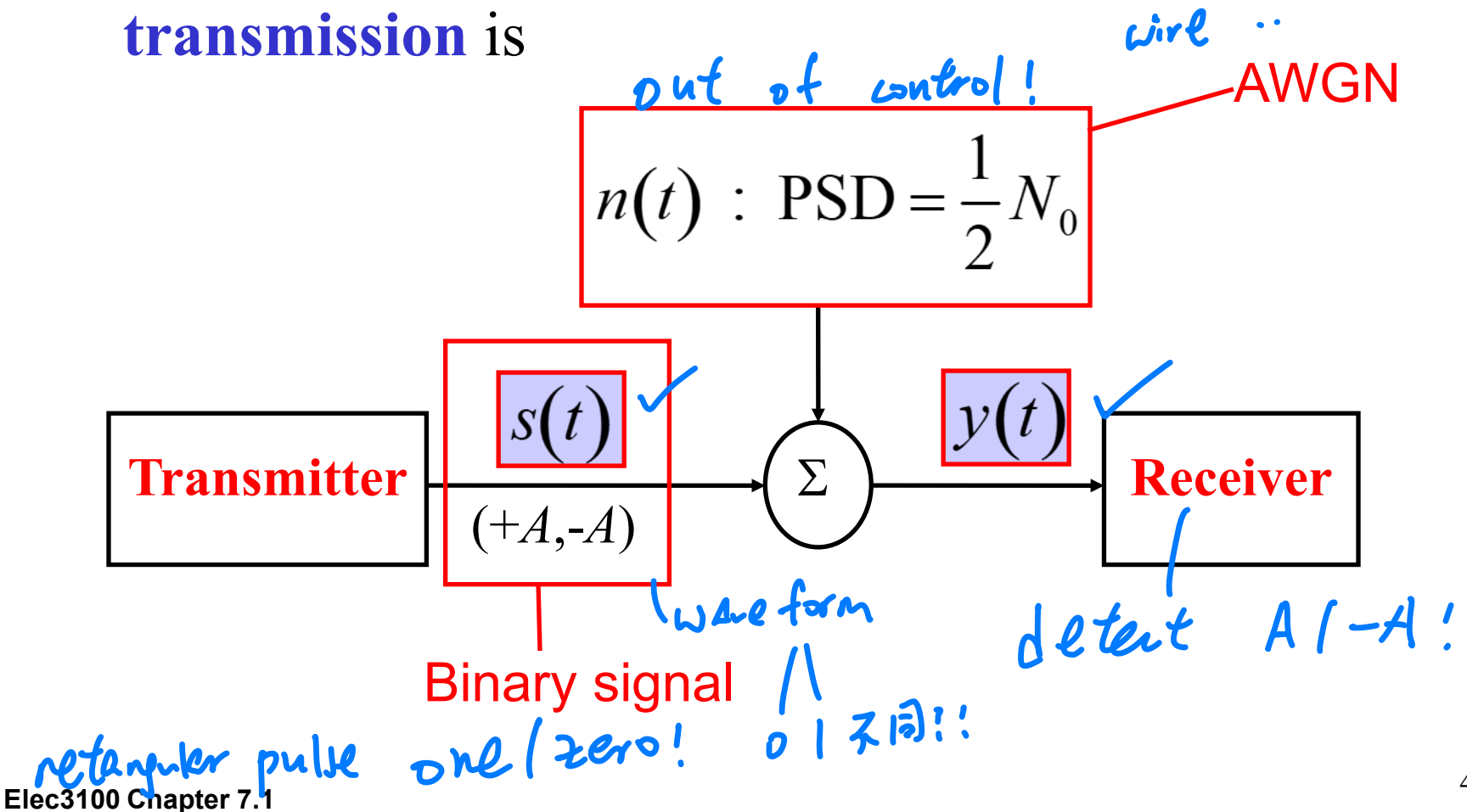
Digital Communications

- 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

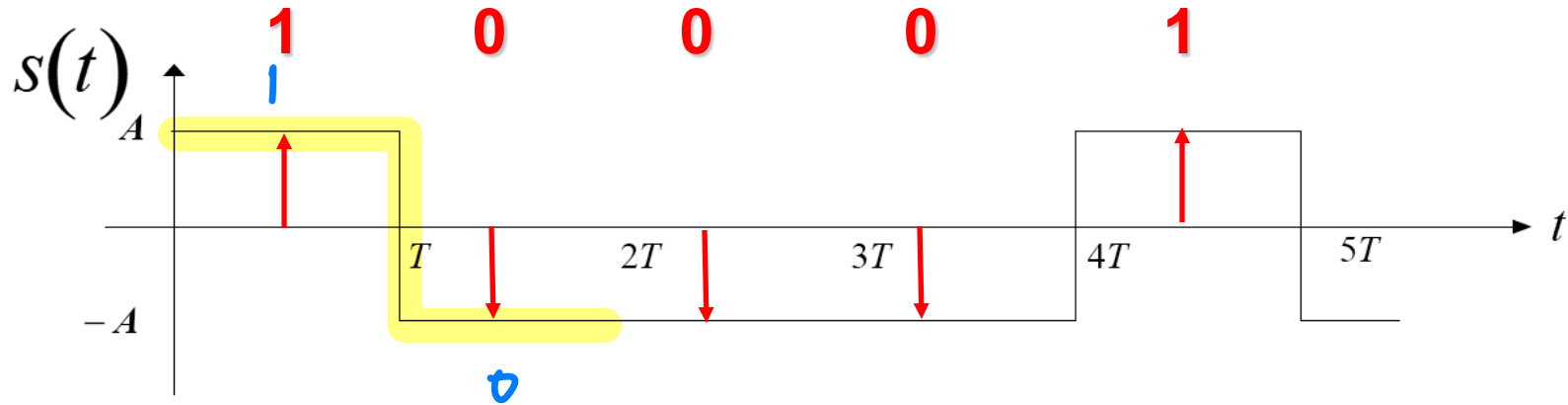
Goal: Minimize P_e

- 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



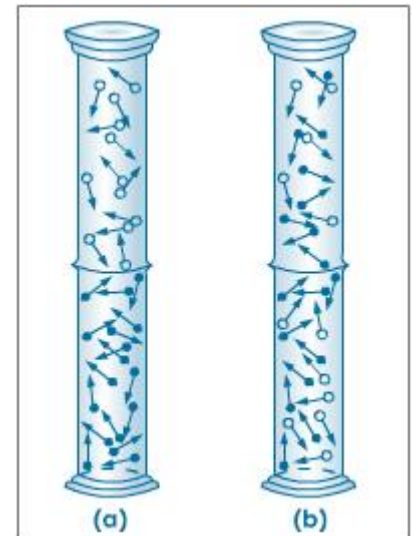
Ch7.1: Baseband Communications & Noise

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



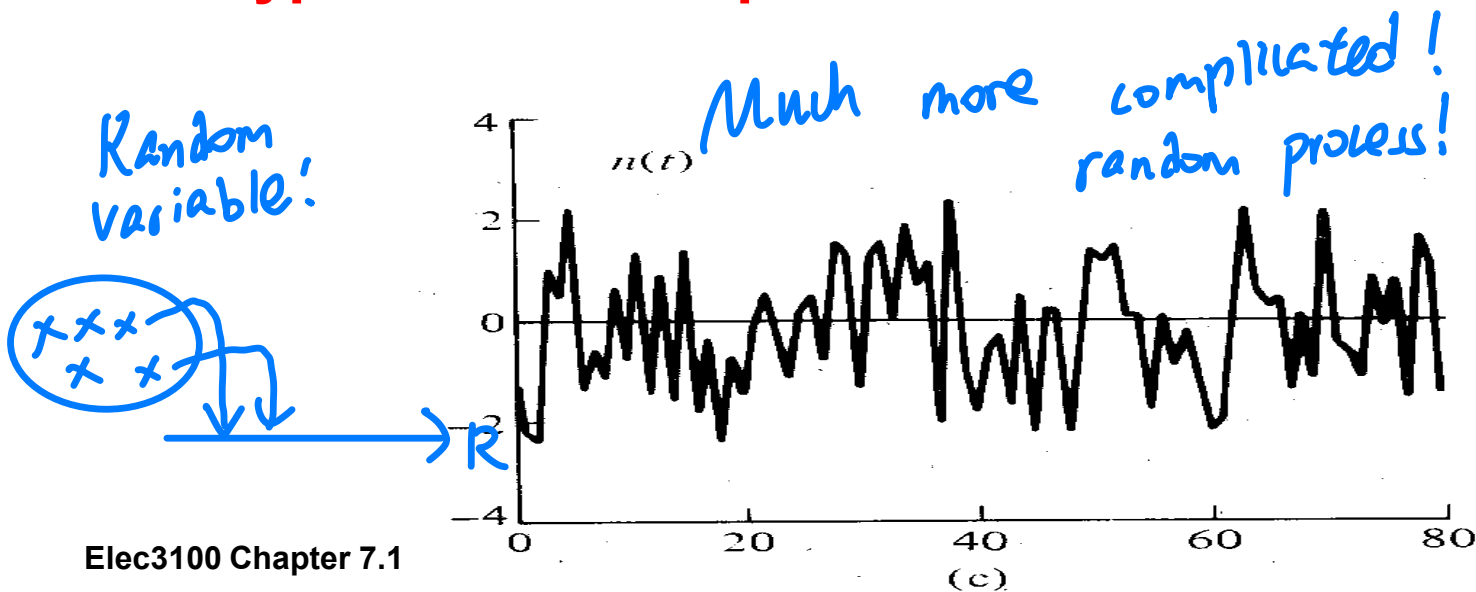
Types of Noise

- 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.

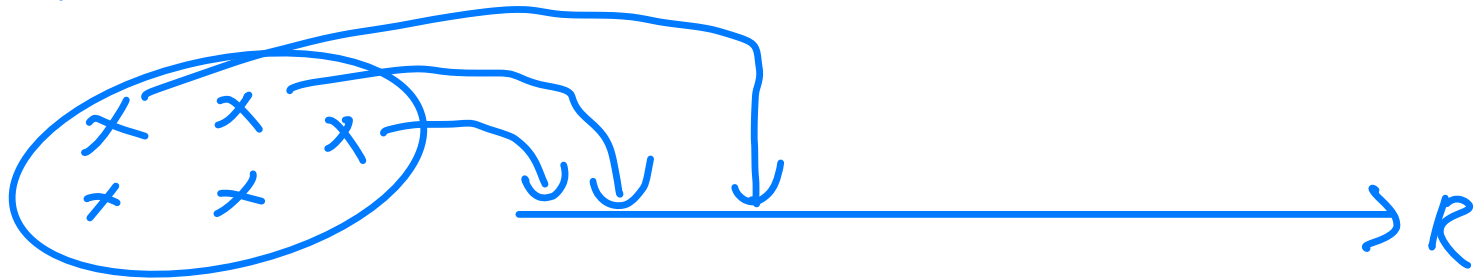


Thermal Noise

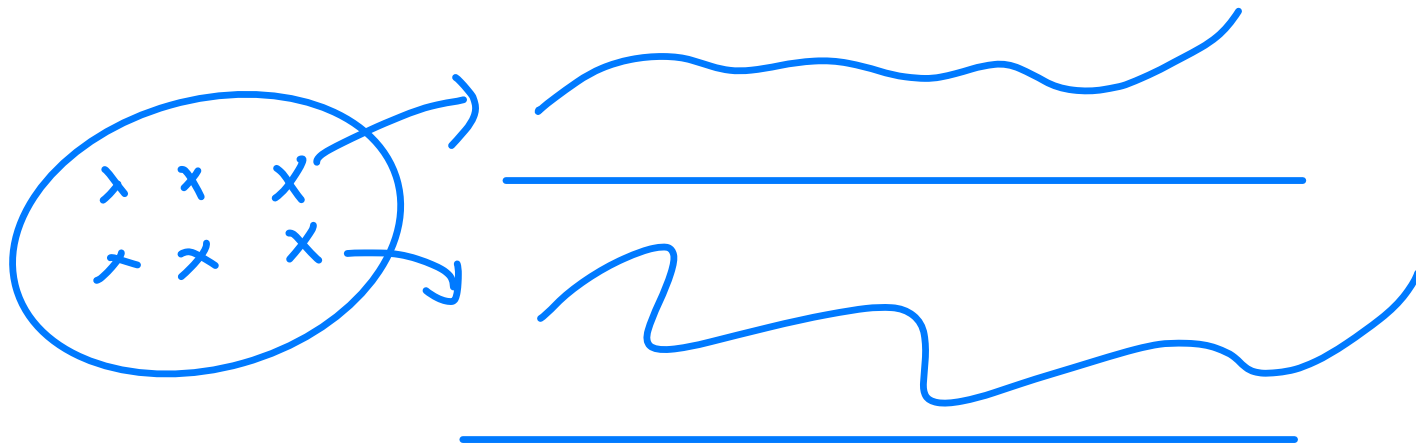
- **Thermal noise** arises from the random motion of charges in conducting medium (such as resistors) and is the most significant noise source we need to consider in ELEC3100.
- Essentially, by connecting a very sensitive oscilloscope across a resistor we can observe thermal noise.
- A **typical noise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



Random Variable



Random Process (RP) (noise is waveform)



e.g. $S = \{H, T\}$

RP

$H \Rightarrow$



$T \Rightarrow$



\hookrightarrow in this course RV

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**:

$$\overline{v_n^2} = 4kTBR$$

↓
Another!

where

T = Temperature of the resistor in Kelvin $[K] = [^{\circ}C] + 273.15$

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

B = Bandwidth of interest

R = Resistance

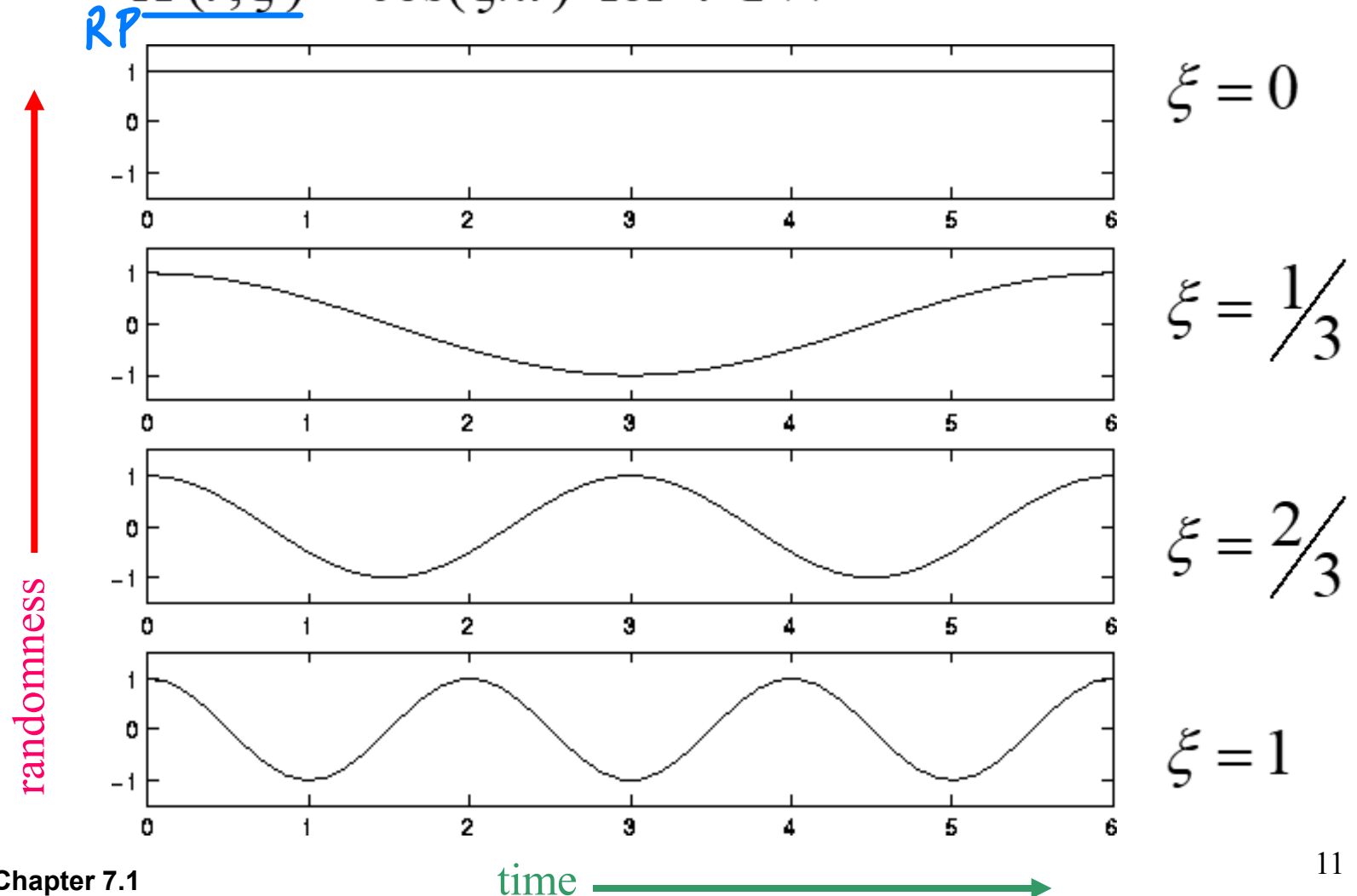
不用的!

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



Characterization of A RP: Mean and Variance Functions

- Mean

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

functions!

- Variance

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

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

Characterization of A RP:

Autocorrelation and Autocovariance

- Autocorrelation

$$R_X(t_1, t_2) = E[X(t_1)X(t_2)] = \int \int xy f_{X(t_1), X(t_2)}(x, y) dx dy$$

- 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 $t_1 - t_2$. Δ

$$m_X(t) = m \quad \text{for all } t$$

$$C_X(t_1, t_2) = C_X(t_1 - t_2) \quad \text{for all } t_1, t_2$$

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

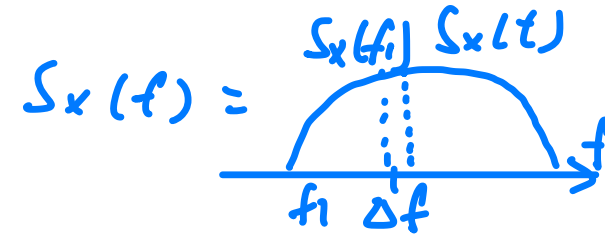
- Suppose $X(t)$ is WSS with autocorrelation $R_X(t)$ where $t = t_1 - t_2$.

- $R_X(0)$ is the average power of the process, $E[X(t)^2]$. $= R_X(0)$
- $R_X(\tau)$ is an even function of τ .
- $|R_X(\tau)| \leq R_X(0)$
(the autocorrelation function is maximum at the origin)

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

$$P([f_i, f_i + \Delta f]) \approx$$

$$S_X(f_i) \Delta f$$

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:

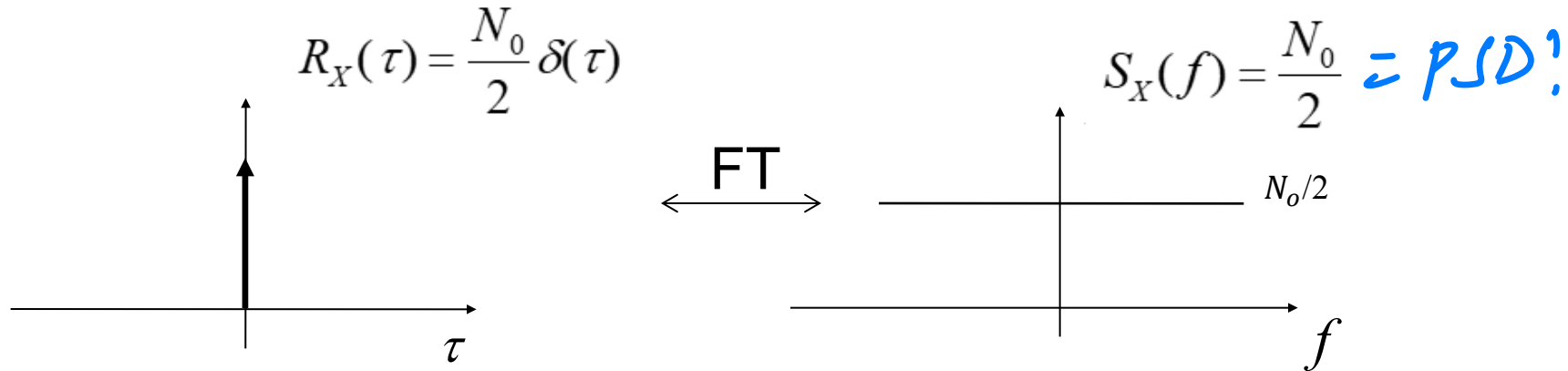
$$\mu(t_i) = E[X(t_i)] = 0 \quad i = 1, 2, \dots, n$$

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

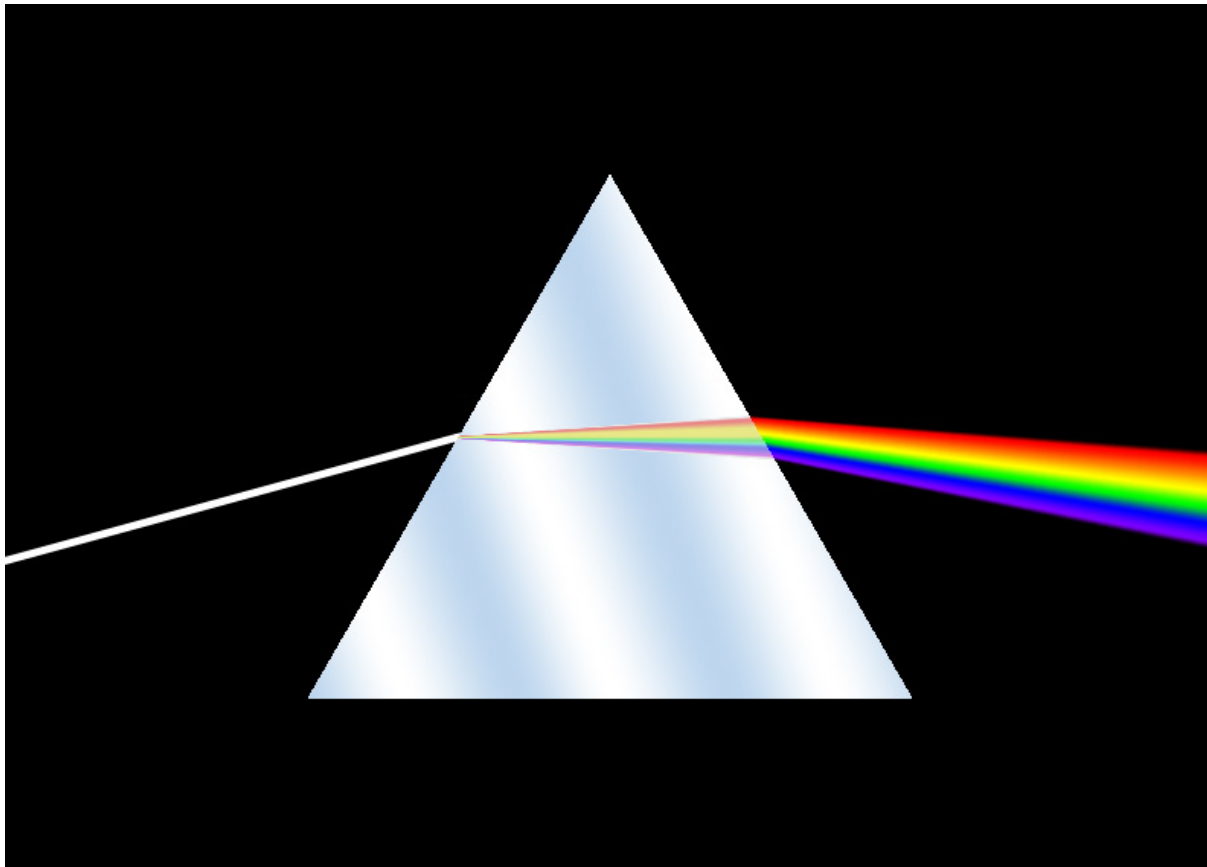
If the power spectrum density is “white”, that is, all frequency components have equal power.



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

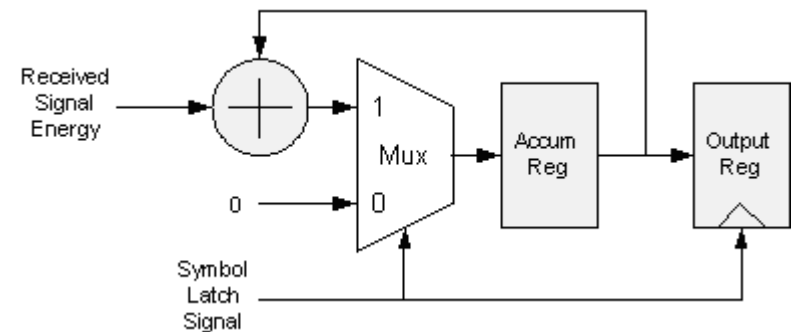
- $R_X(t_i, t_j) = E[X(t_i)X(t_j)] = 0$ for $i \neq j$
 - $X(t_i)$ and $X(t_j)$ are uncorrelated
- $R_X(0) = \infty = \overline{X^2(t)} = \sigma_X^2$
 - Noise has infinite power

Why White noise?



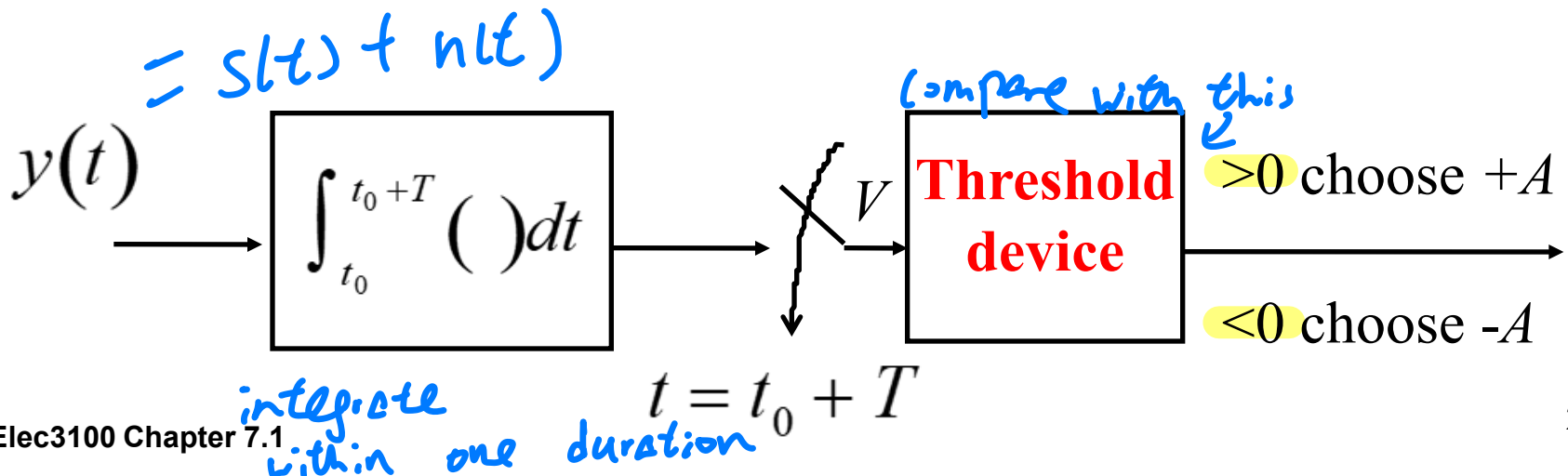
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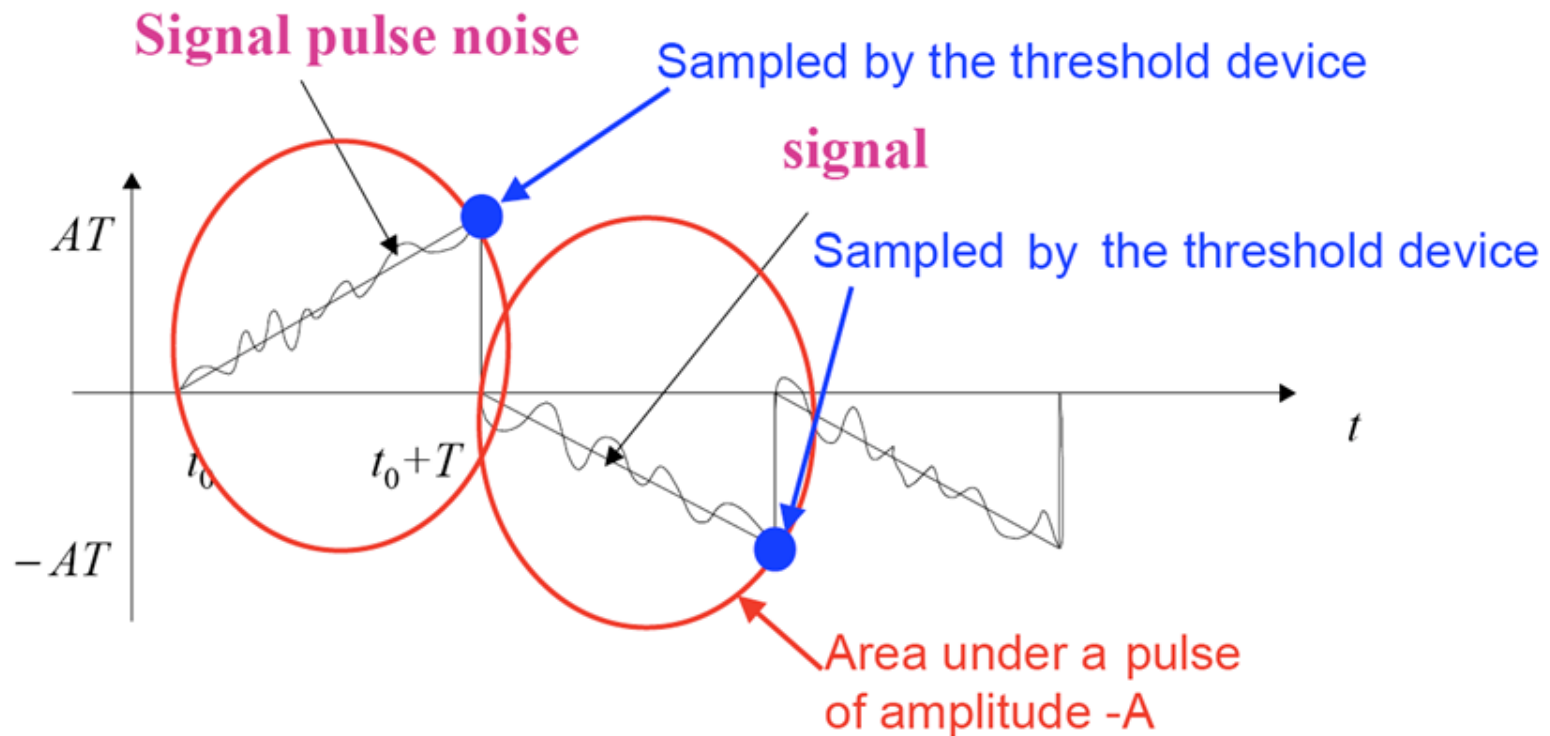
Receiver Structure

- 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



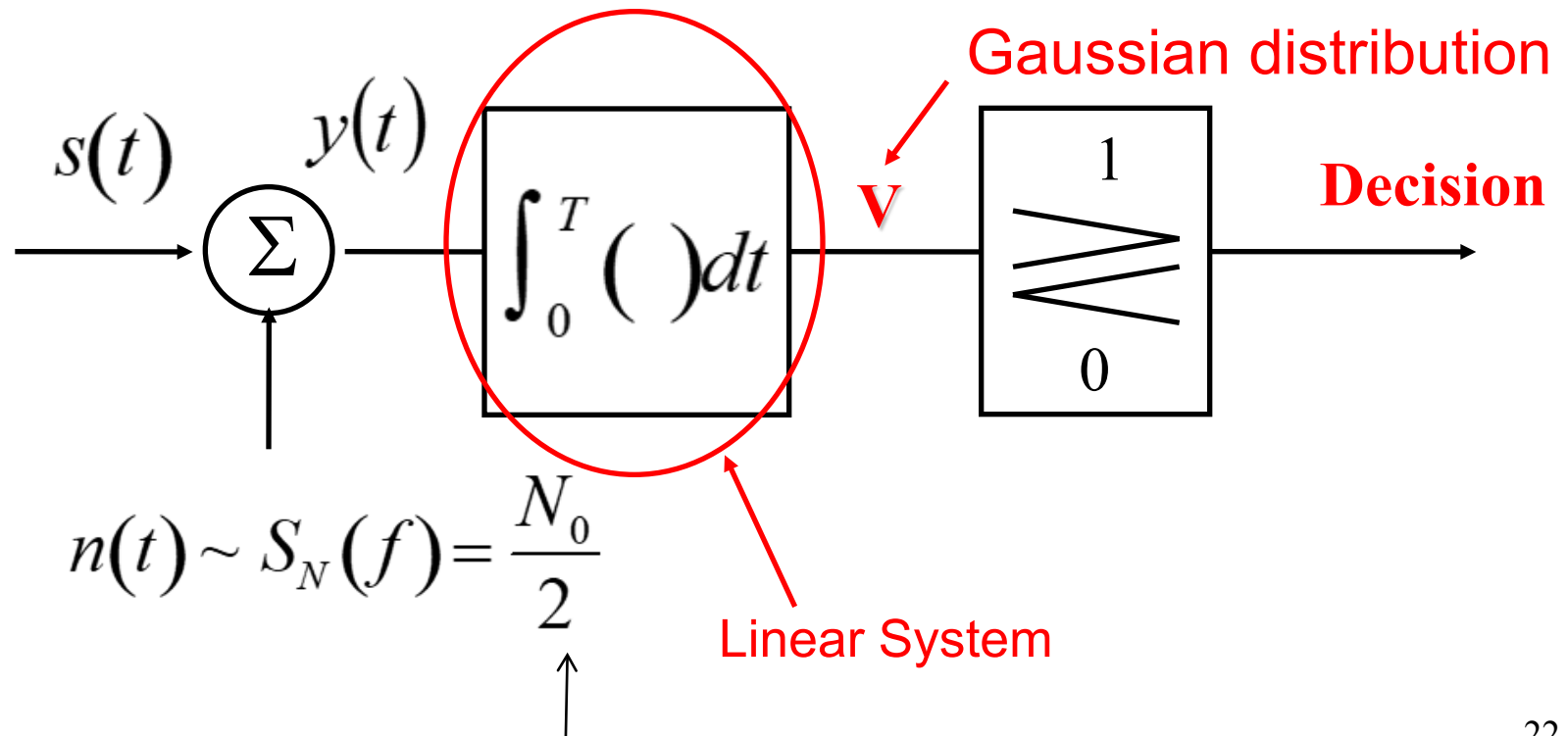
Integrate-and-Dump

- 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

$$\underline{s(t)} = \begin{cases} A & 0 \leq t < T \\ -A & 0 \leq t < T \end{cases} \quad \begin{array}{l} \text{if "1" transmitted} \\ \text{if "0" transmitted} \end{array}$$

$$V = \int_0^T [s(t) + n(t)] dt$$

signal part is constant.

$$= \begin{cases} \underline{AT} + N & \text{if "1" is sent} \\ \underline{-AT} + N & \text{if "0" is sent} \end{cases}$$

Random variable

N is Gaussian distributed

with

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

Linear combination of Gaussian

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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

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

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

bit 1

bit 0

V is the integrated signal within T_b time interval!

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

Total probability theorem

Gaussian!

Conditional Error Probability

Prior 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 σ^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

$E(n) = 0$

Just $E(N^2)$

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

$E(n) = 0$

Autocorrelation

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

AWGN

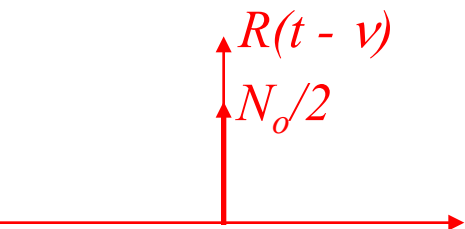
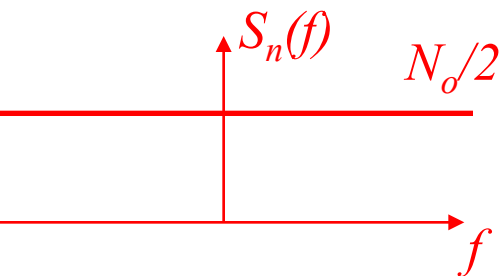
$$= \int_0^T \int_0^T R_n(t-v) dt dv$$

AWGN

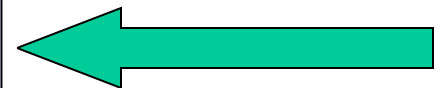
$$= \int_0^T \int_0^T \frac{N_0}{2} \delta(t-v) dt dv$$

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

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



find the distribution of \checkmark



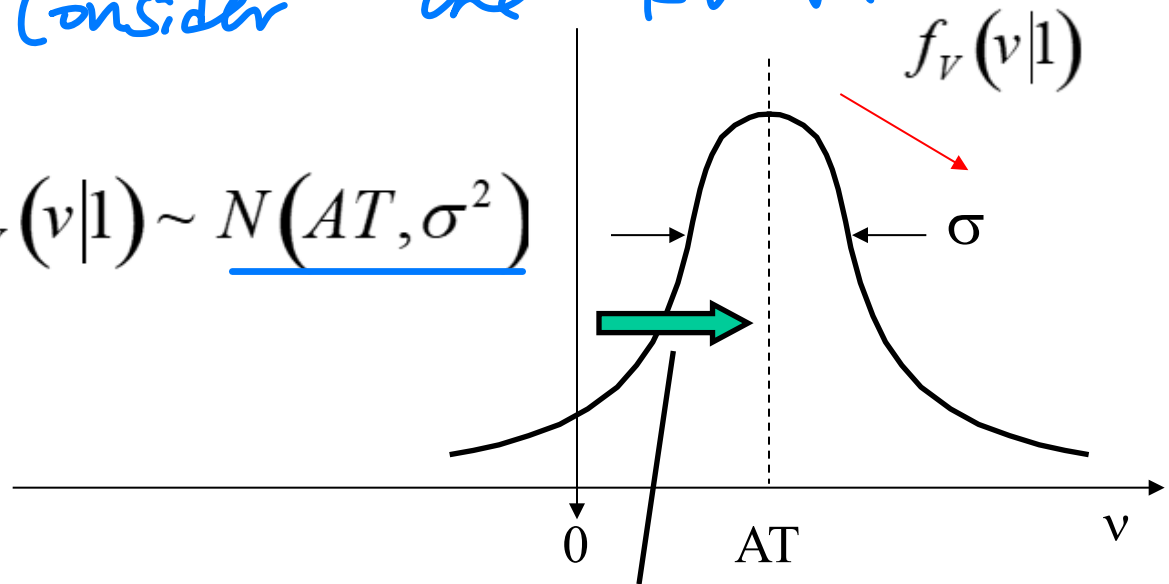
Conditional Probabilities

Consider the RV V !

Mean = AT

$$\left. \begin{array}{l} E[V|1] = +AT \\ Var[V|1] = \sigma^2 \end{array} \right\} \Rightarrow f_V(v|1) \sim \underline{N(AT, \sigma^2)}$$

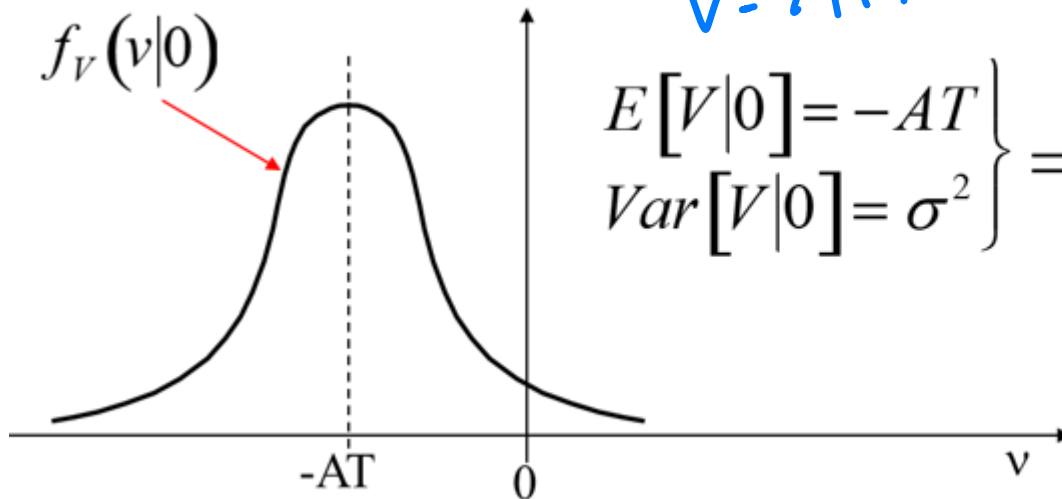
= $\frac{NoT}{2}$



$V = -AT + N$

Shifted positive due to
The positive pulse +A

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



Conditional Probabilities

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

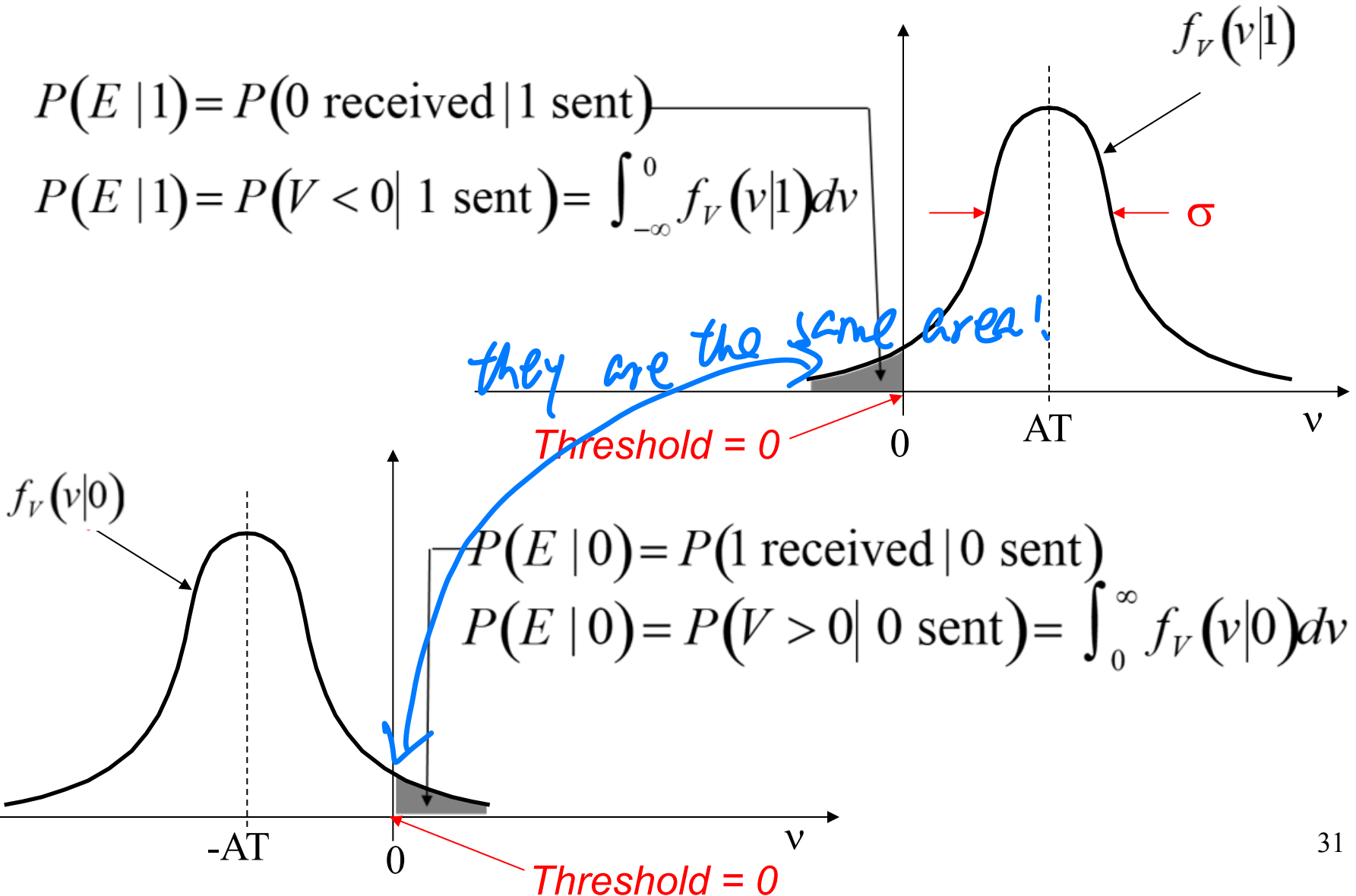
$$P(E | 1) = P(V < 0 | 1 \text{ sent}) = \int_{-\infty}^0 f_V(v|1) dv$$

they are the same area!

Threshold = 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**

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

- 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

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

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

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

(WSS noise)

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

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

$$\begin{aligned} P_e &= P(E|1)P(1) + P(E|0)P(0) \\ &= P(v < 0|1)P(1) + P(v > 0|0)P(0) \end{aligned}$$

As \uparrow

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

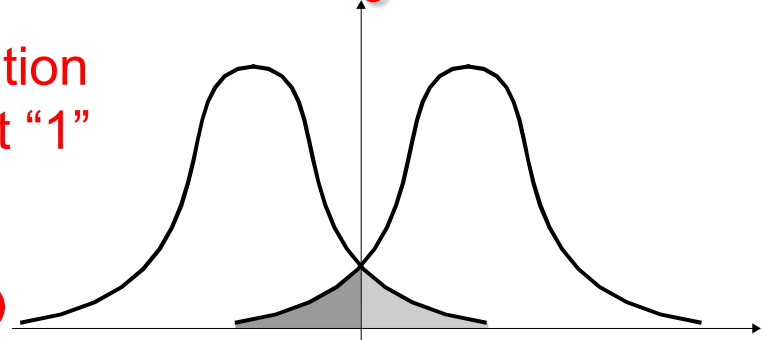
Total Error Probability

- 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

2x 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

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

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

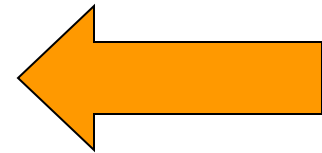
Evaluation Relies on Q-Function

- 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**

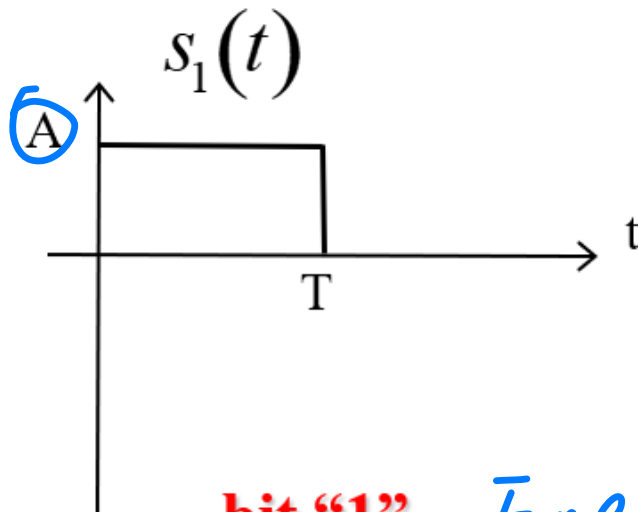
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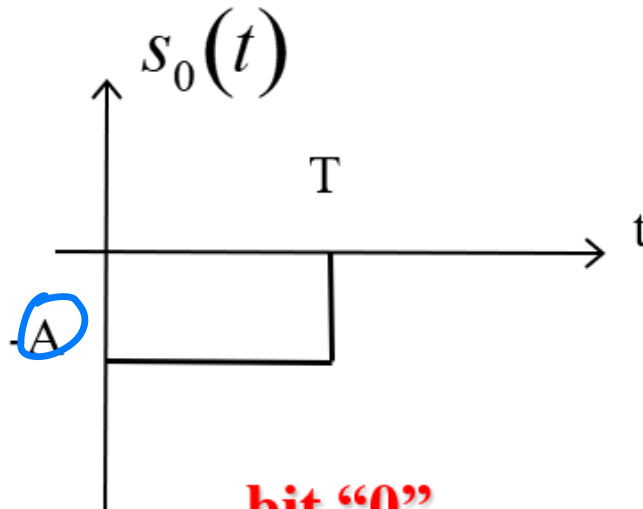


A General Case

- Next, note that one can represent $s(t)$ as $s_0(t)$ (“0” sent) or $s_1(t)$ (“1” sent).



bit "1" Energy!



bit "0"

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

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

Energy

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

$$\bar{E}_b = A^2 T$$

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

• 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

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

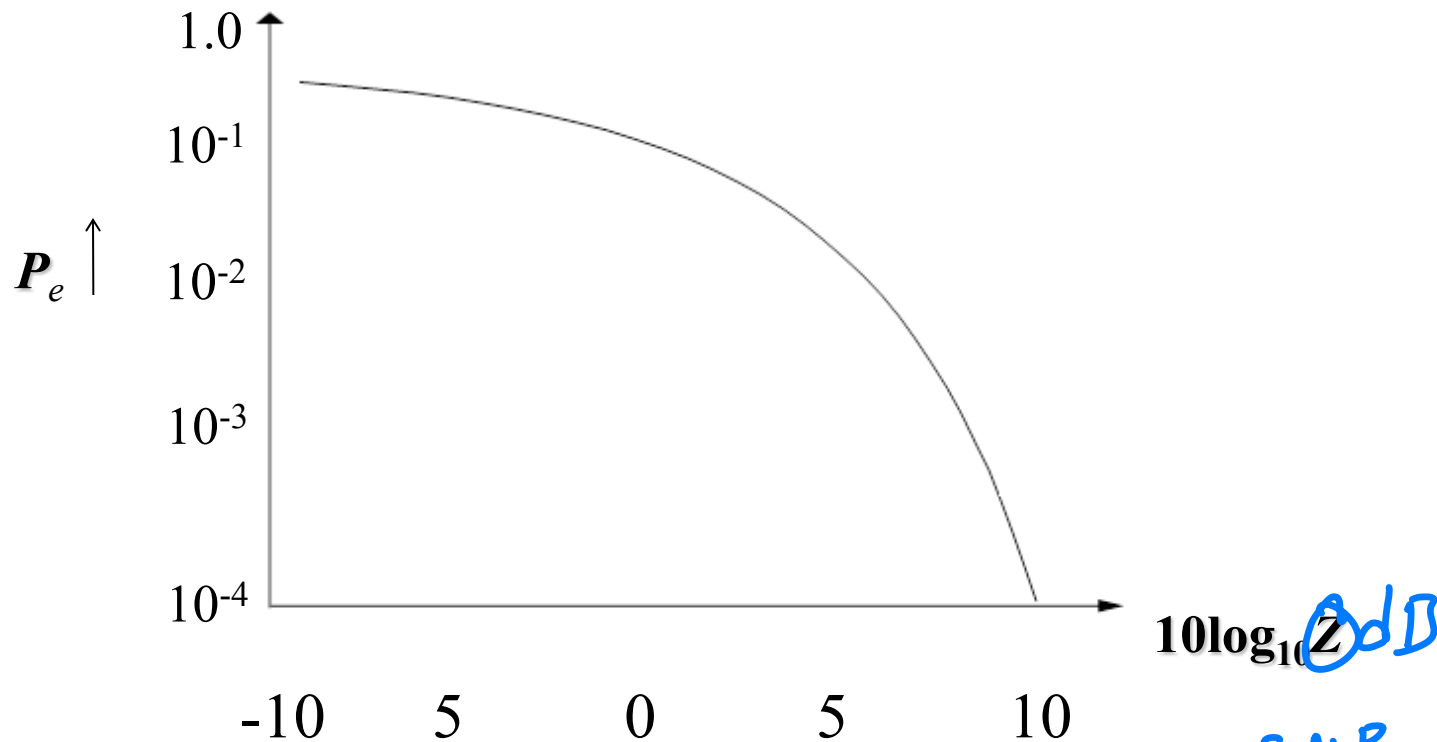
with

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

Error Function
relationship between
erfc and erf function
Complementary
Error Function

Bit Error Rate (BER)

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



where $P_e = Q[\sqrt{2Z}] = \frac{1}{2} \text{erfc}[\sqrt{Z}]$

$$Z = \frac{E_b}{N_0}$$

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.

Handwritten notes:
 $\sim T$
 $T = \frac{1}{1M}$
 $A \sim 1 = \frac{N_0}{2}$
 $Q \sqrt{2 \times 10^{-6} / (2 \times 10^{-7})}$

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

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

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

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

$$Q(3.16) \approx 0.00085$$

Probability of Error, P_e

Example 7.2

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

$$P_e: Q\left(\sqrt{2E_b/N_0}\right) \approx Q(2.028)$$

(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. $A = 63.2 \cdot 10^{-3}$ V = 63.2 mV.

$$\downarrow P_e = Q\left(\sqrt{\frac{2A^2}{N_0/T}}\right) \approx A^2 \left(\frac{1}{1000}\right) / 10^{-7} = 8$$

$A = 63.2$ mV

Then A need to increase!

BER vs. Data Rate

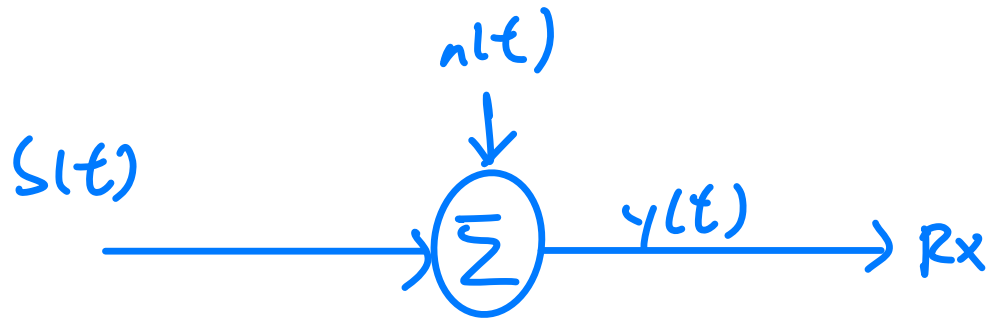
- In the last example, for a fixed amplitude, error probability increases as bit rate increases.
- This can be understood by looking at the error probability expression

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

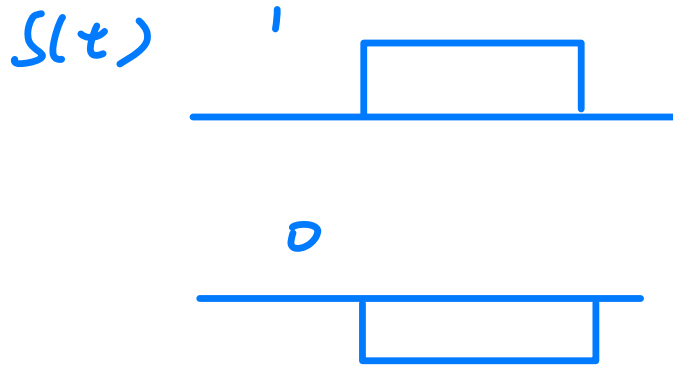
- Increase in bit rate → smaller bit period T
 - higher noise power
 - lower signal energy

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

Are they happening at the same time?



So far



$n(t)$: AWGN

Rx : integrate-and-dump

Next: In general

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

design $s(t)$

extension of Rx

↓
Allow multiple users!