

Ch7.2: Optimal Receiver

Information source
and input transducer



Source Coding



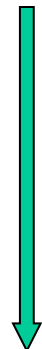
Channel Coding



Modulator



Channel



Demodulator
(Matched Filter)



Channel Decoding



Source Decoding



Information sink
and output transducer

*If $p(t) \neq p(t)$
waveform may
not be square pulses!*

Ch7.2: Matched
Filter



• Questions to be answered:

- ☐ **BER for General Signals & Receivers:**
Performance evaluation for general signals
 - ☐ Optimum Threshold
- ☐ **The Matched Filter:** The optimal receiver structure
 - ☐ Optimal Receiver
 - ☐ Optimal Signals

Summary/Outline

Target:
minimize P_e

- We have introduced very important concepts
 - Transmission of bits
 - Baseband binary communications
 - Computation of error probability P_e as a function of signal energy per bit to noise density E_b/N_o . *Optimal!*
- The above discussion were based on some assumptions (e.g. integrate-dump receiver)
- We will now talk about the *optimal receiver structure*, i.e. a receiver that can minimize the error probability P_e for a given E_b/N_o

Ch7.2: Optimal Receiver

❑ Error Probability for General Signals & Receivers

❑ Optimum Threshold

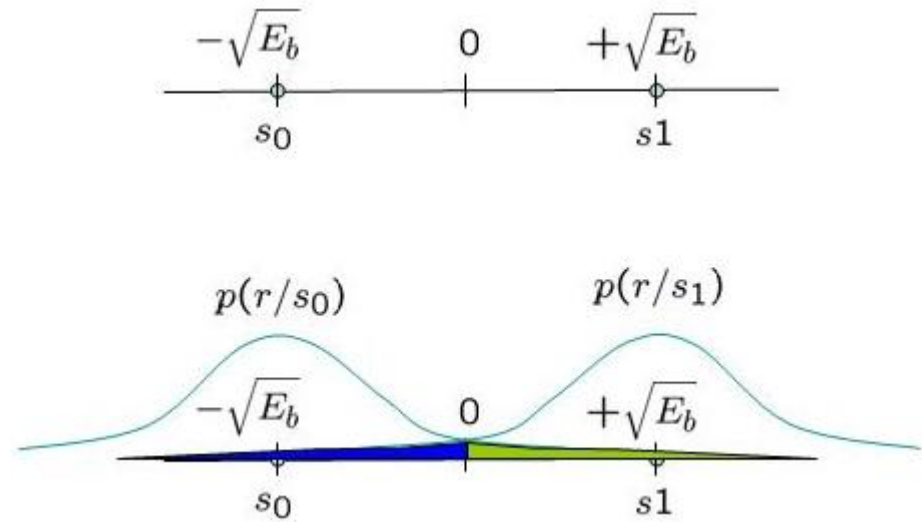
❑ Input-Output Relation

❑ The Matched Filter

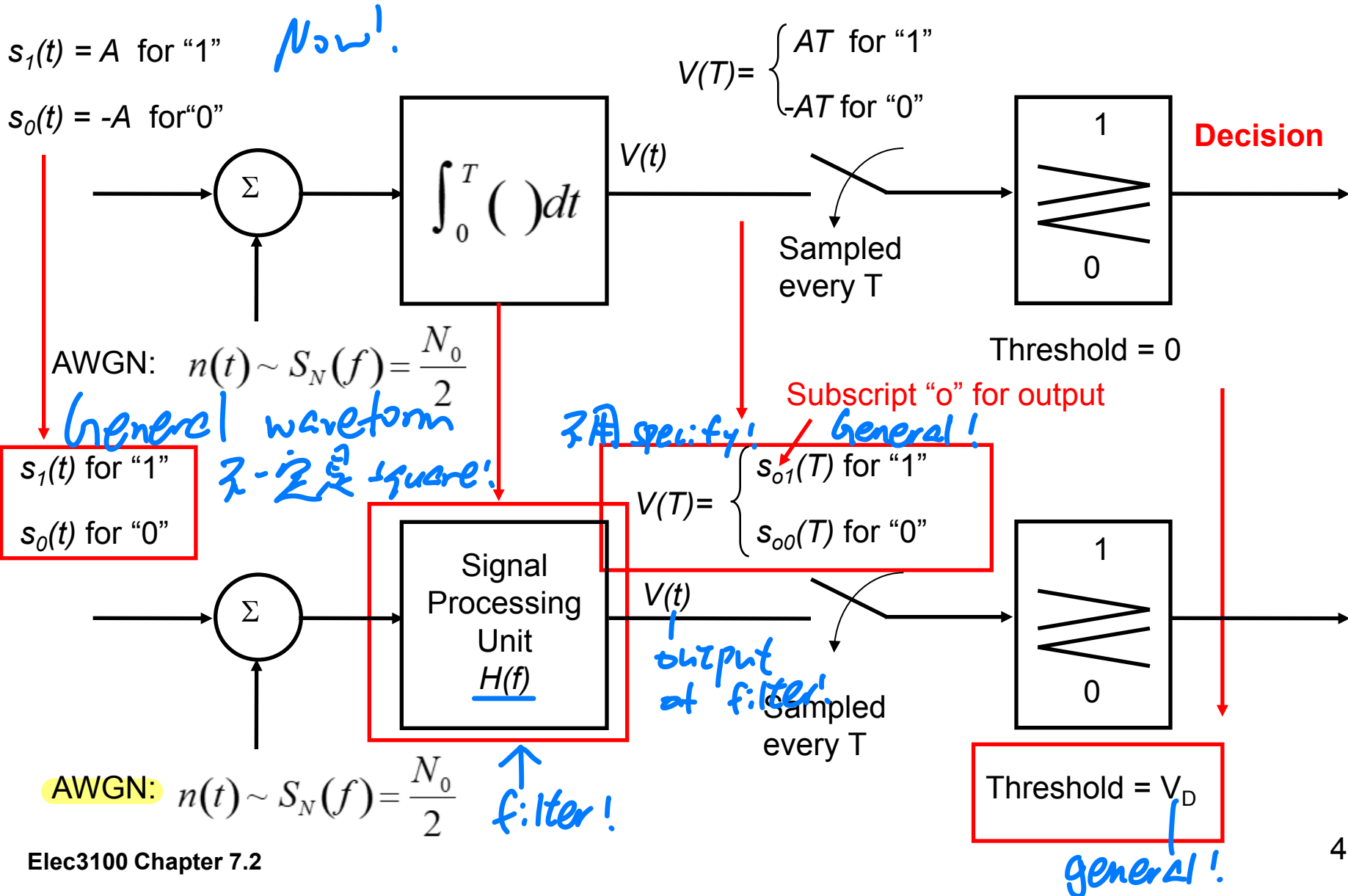
❑ Optimal Receiver

❑ Examples

❑ Optimal Signals



General Receiver Structures



Generalization

1. Input Signal	$s_1(t) = A$ for “1” $s_0(t) = -A$ for “0”	$s_1(t)$ for “1” $s_0(t)$ for “0”
2. Signal Processing Unit	Integration and dump	Generalized Linear Filter $H(f)$
3. Output Signal	AT for “1” $-AT$ for “0”	$s_{o1}(t)$ for “1” $s_{o0}(t)$ for “0”
4. Decision threshold	<i>only they⁰ are the same!</i>	
5. Input Noise	AWGN: $S_N(f) = \frac{N_0}{2}$	AWGN: $S_N(f) = \frac{N_0}{2}$
6. Sampling Period	T	T

Performance Evaluation

- Let's write the input signal as: (the duration of the signal is T)

$$s(t) = \begin{cases} s_0(t) & \text{"0" sent} & 0 \leq t < T \\ s_1(t) & \text{"1" sent} & 0 \leq t < T \end{cases}$$

- The sampled signal and noise at the output of the linear filter $H(f)$ is (subscript "o" stands for output)

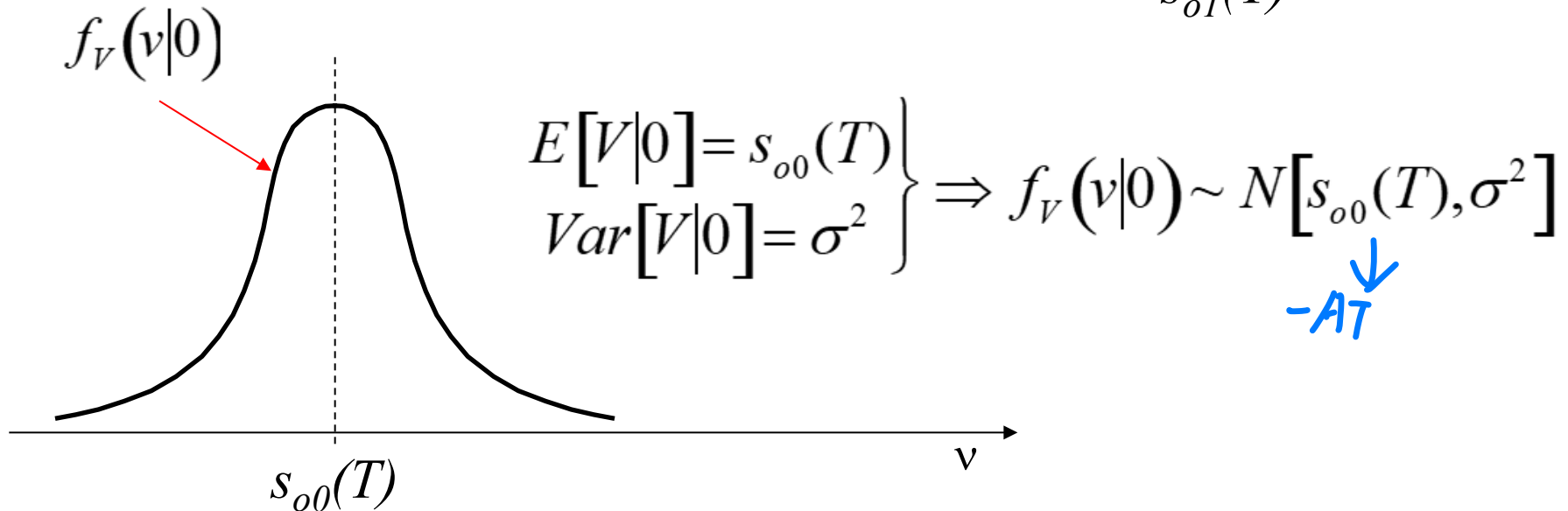
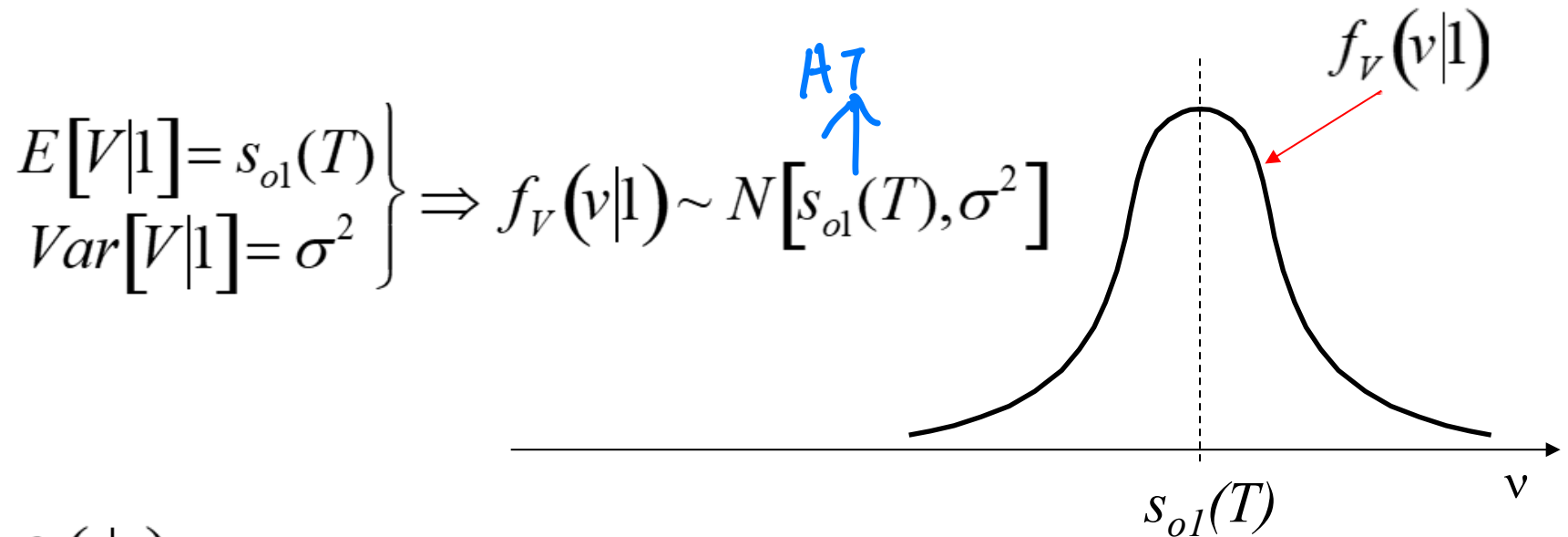
Sampled every bit period T
 $T = \text{bit period!}$

$$V(T) = \begin{cases} s_{o1}(T) + N & \text{"1" sent} \\ s_{o0}(T) + N & \text{"0" sent} \end{cases}$$

Handwritten notes:
 - Above the top case: $h(f)AT(1-Af)$
 - To the right: Gaussian (mean = 0)
 - Red arrows point from the noise N in both cases to the boxed note below.

- The noise component of the output N should be Gaussian distributed with zero mean and variance of σ^2 (not known at this stage).

Performance Evaluation

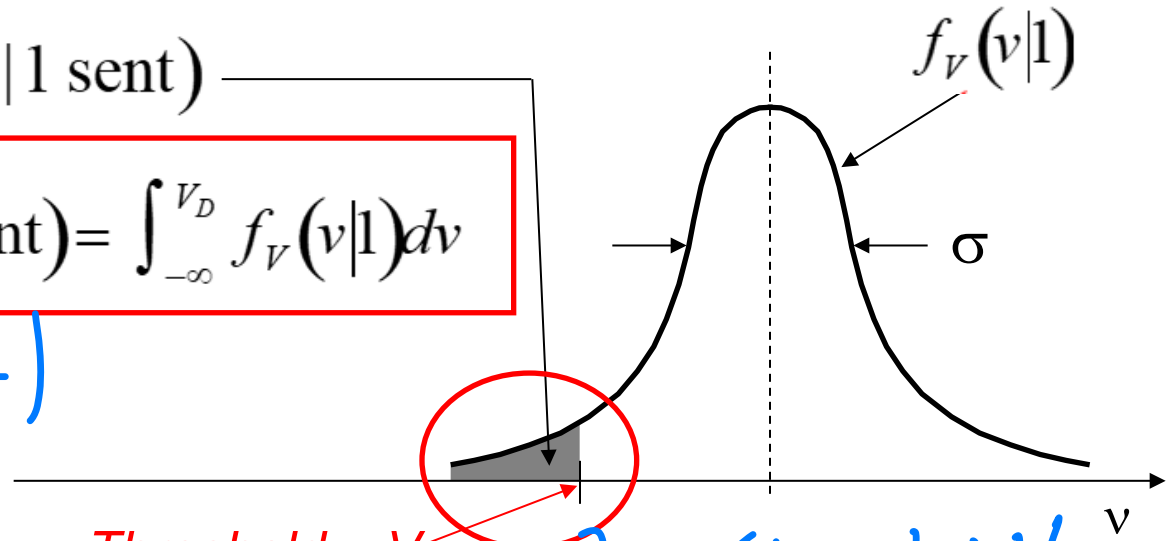


Performance Evaluation

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

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

$$\Phi\left(\frac{V_D - \mu_1(\gamma)}{\sigma}\right)$$

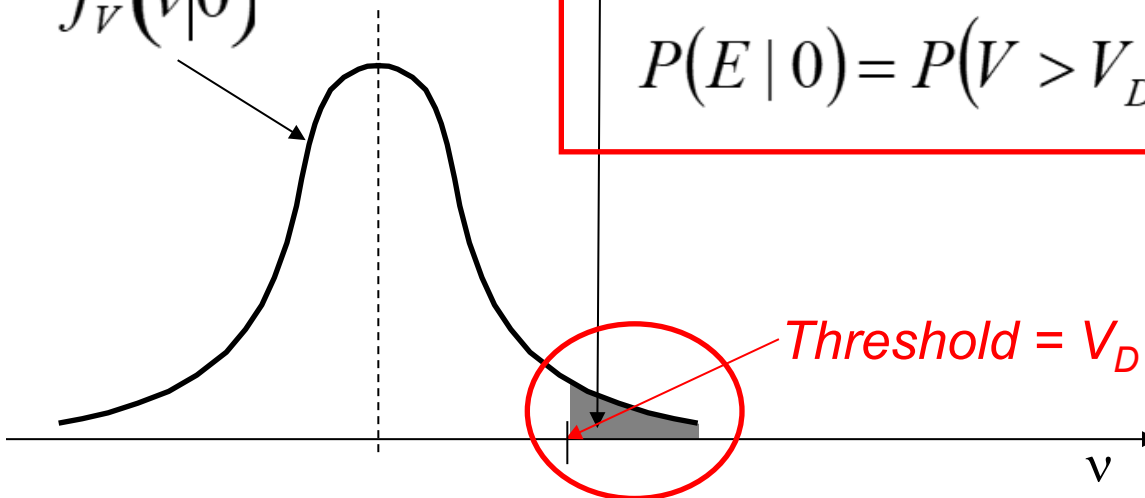


Threshold = V_D 不是0 threshold!

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

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

$$f_V(v|0)$$

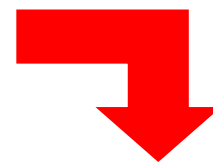


Threshold = V_D

Performance Evaluation

$$P(E|0) = \int_{V_D}^{\infty} f_V(v|0) dv = \int_{V_D}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(v-s_{o0})^2}{2\sigma^2}} dv$$

$$\text{Transformation: } x = \frac{v - s_{o0}}{\sigma} \Rightarrow dx = \frac{dv}{\sigma}$$



$$P(E | 0) = \int_{\frac{V_D - s_{o0}}{\sigma}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = Q\left[\frac{V_D - s_{o0}(T)}{\sigma}\right]$$

$$P(E | 1) = Q\left[\frac{s_{o1}(T) - V_D}{\sigma}\right]$$

Can be obtained through similar steps

One more step after the transformation:
 $x' = -x$

$$V|1 \sim N(\mu_0, \sigma^2)$$

$$V|0 \sim N(\mu_1, \sigma^2)$$

$$P(E|1) = P(V < V_D | 1) = P\left(\frac{V - \mu_0}{\sigma} < \frac{V_D - \mu_0}{\sigma} \mid 1\right)$$

$\underbrace{\hspace{1.5cm}}_{N(0,1)}$

$$= \Phi\left(\frac{V_D - \mu_0}{\sigma}\right) = Q\left(\frac{\mu_0 - V_D}{\sigma}\right)$$

$\underbrace{\hspace{1.5cm}}_{N(0,1)}$

$$P(\bar{E}|0) = P(V > V_D | 0) = P\left(\frac{V - \mu_1}{\sigma} > \frac{V_D - \mu_1}{\sigma} \mid 0\right)$$

$\underbrace{\hspace{1.5cm}}_{N(0,1)}$

$$= Q\left(\frac{V_D - \mu_1}{\sigma}\right) = \Phi\left(\frac{\mu_1 - V_D}{\sigma}\right)$$

Performance Evaluation

Error probability

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

$$P_e = Q\left[\frac{s_{o1}(T) - V_D}{\sigma}\right]P(1) + Q\left[\frac{V_D - s_{o0}(T)}{\sigma}\right]P(0)$$

VIP (Very Important assumption)

$$1. P(1) = P(0) = 1/2$$

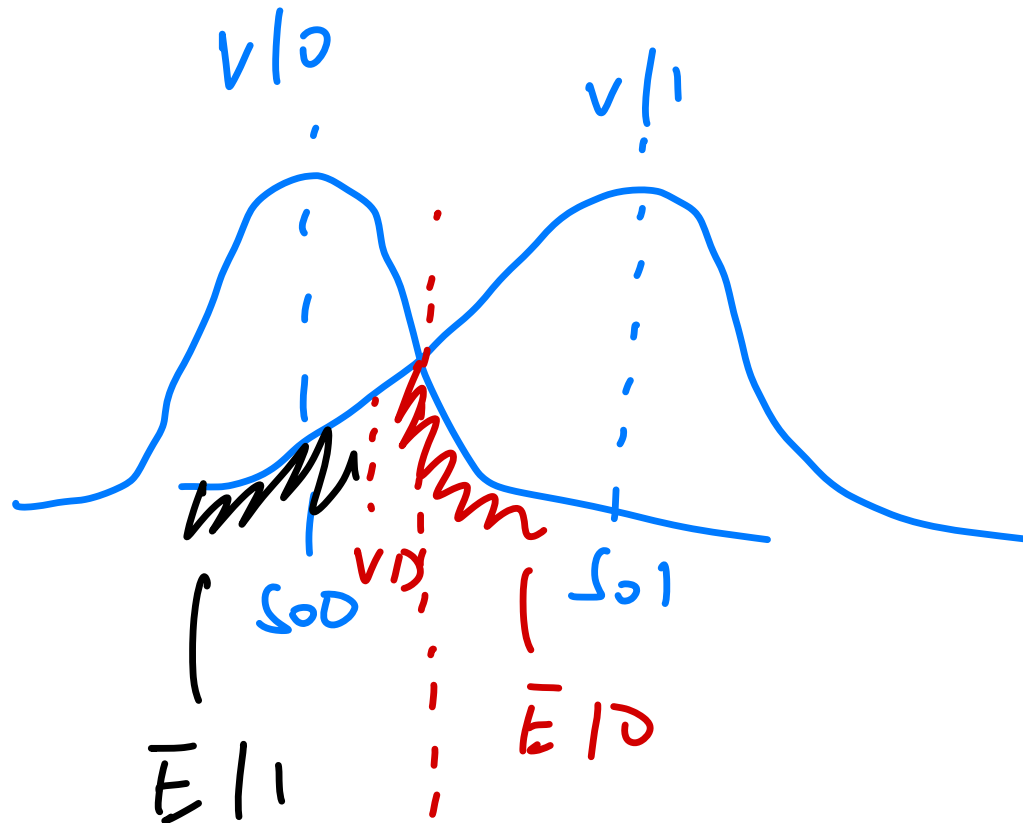
Assume they are same!

$$2. V_{D,opt} = \frac{s_{o1}(T) + s_{o0}(T)}{2}$$

$$= \left[Q\left(\frac{s_{o1} - V_D}{\sigma}\right) + Q\left(\frac{V_D - s_{o0}}{\sigma}\right) \right]^{\frac{1}{2}}$$

⇒ select V_D to

Optimum decision threshold → best P_e minimize P_e



= Summation of two parts!

Optimal v_D :

$$= \frac{S_{00} + S_{01}}{2} \text{ (mid-point)}$$

Optimum Threshold

not depends on sigma!

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

First optimum: Optimum Threshold

mid-point!

$$V_{D,opt} = \frac{s_{o0}(T) + s_{o1}(T)}{2}$$

Thus, when $V_D = V_{D,opt}$

Physical meaning?

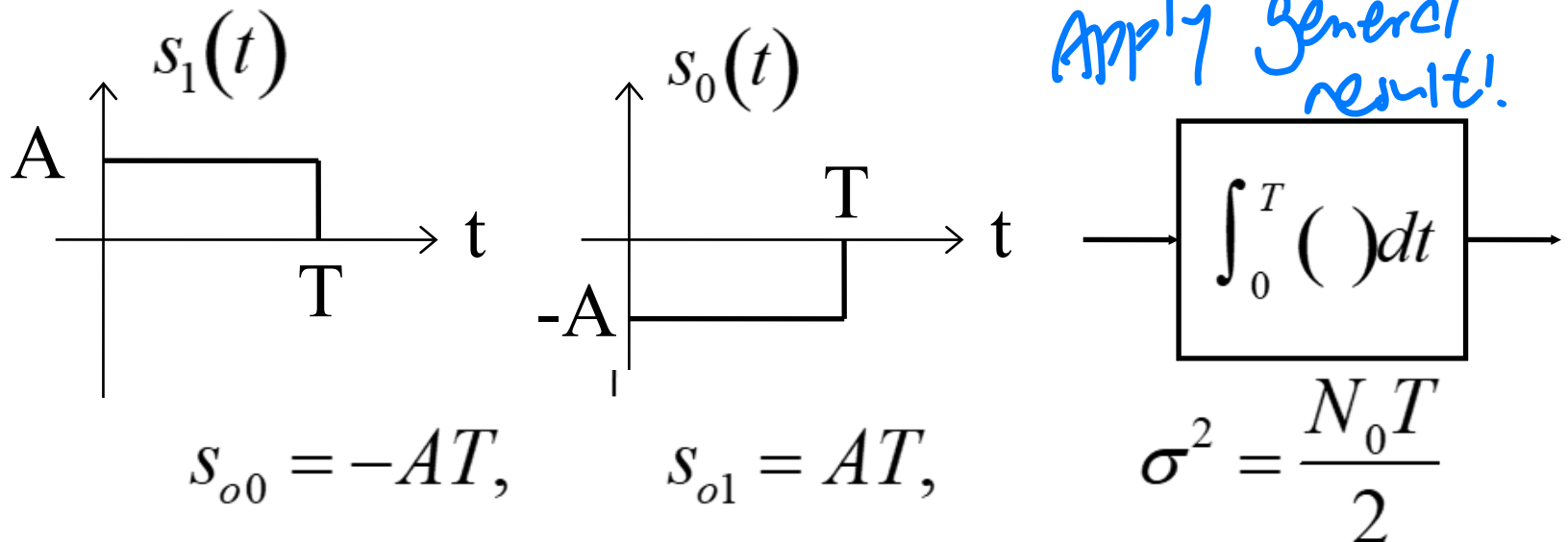
two points larger! => difference larger

$$P_e = Q \left[\sqrt{\frac{(s_{o1}(T) - s_{o0}(T))^2}{4\sigma^2}} \right]$$

pe ↓

Performance Evaluation: Example

- Example:** Assume signal processing \sim *Integrate-and-dump* and that $s_k(t)$ are as follows, then



- So**
$$\frac{(s_{o1} - s_{o0})^2}{4\sigma^2} = \frac{4A^2T^2}{4\sigma^2} = \frac{2A^2T}{N_0} \triangleq \frac{2E_b}{N_0}$$

$$\therefore P_e = Q[Z] = Q\left[\sqrt{\frac{2E_b}{N_0}}\right]$$

Optimal Receiver

- We have obtained error probability for arbitrary linear filter

$$\underline{h(t) \leftrightarrow H(f),}$$

$$P_e = Q \left[\sqrt{\frac{(s_{o1} - s_{o0})^2}{4\sigma^2}} \right] = Q \left[\sqrt{\frac{\zeta^2}{4}} \right]$$

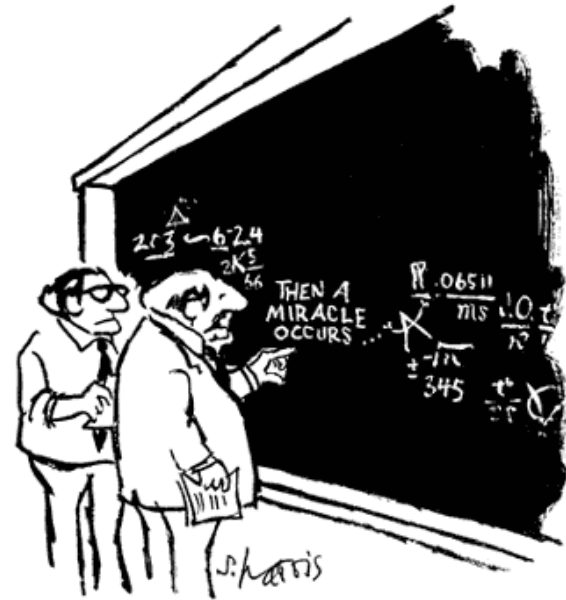
$$\frac{s_{o1} - s_{o0}}{\sigma} = \zeta$$

- Recall that $Q[z]$ gets smaller as z increases.
- We have yet to put any specification on $H(f)$.
- An optimal filter/receiver $H(f)$ can be found such that ζ is maximized and therefore P_e is minimized
- We call such filter a Matched Filter. \leftarrow

details next lecture!

Ch7.2: Optimal Receiver

- ❑ Error Probability for General Signals & Receivers
 - ❑ Optimum Threshold
- ❑ **Input-Output Relation**
- ❑ The Matched Filter
 - ❑ Optimal Receiver
 - ❑ Examples
 - ❑ Optimal Signals

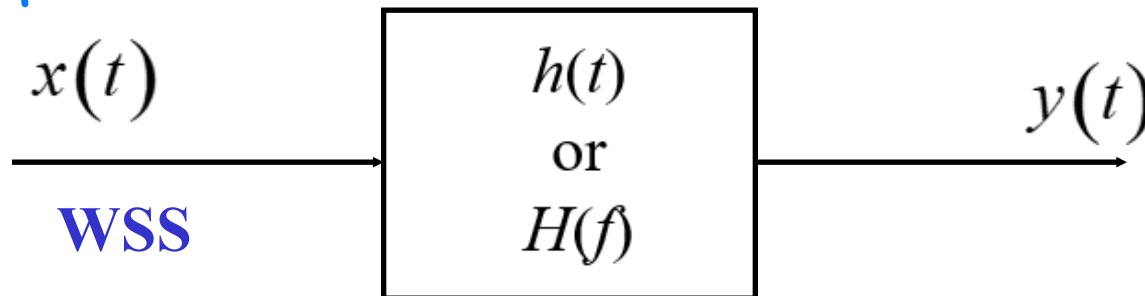


"I think you should be more explicit here in step two."

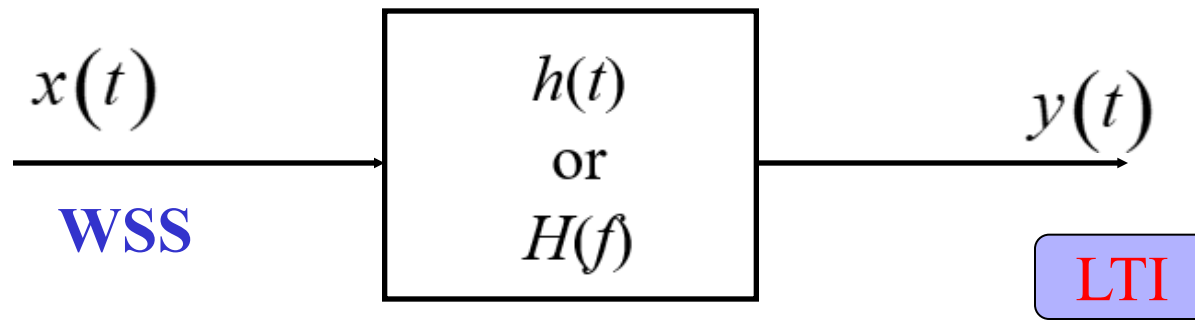
Input and Output Relationships

- In order to design the optimal receiver, we need to understand how a linear filter will affect the signal and noise, respectively.
- The signal has a fixed waveform. Thus, the input and output relationships are easy to obtain.
- However, the noise is a random process. Then, what will be the input and output relationships if a wide sense stationary (WSS) random process (AWGN) goes through a linear filter?

input with noise!



Input and Output Relationships for PSD



$$y(t) = \int_{-\infty}^{\infty} h(s)x(t-s)ds$$

$$R_{xy}(\tau) = E[x(t)y(t+\tau)] \quad \text{cross correlation}$$

$$= E\left[x(t) \int_{-\infty}^{\infty} h(s)x(t+\tau-s)ds\right]$$

$$= \int_{-\infty}^{\infty} h(s)E[x(t)x(t+\tau-s)]ds$$

$$= \int_{-\infty}^{\infty} h(s)R_x(\tau-s)ds \quad \text{WSS}$$

Cross Correlation between $x(t)$ and $y(t)$

- Thus,

$$\begin{aligned} R_{xy}(\tau) &= h(\tau) * R_x(\tau) \\ S_{XY}(f) &= H(f) S_X(f) \end{aligned}$$

Power Spectral Density = FT of Correlation Function

- Likewise, can show that

$$R_{yx}(\tau) = h(-\tau) * R_x(\tau)$$

$$S_{YX}(f) = \underline{H^*(f) S_X(f)}$$

**Time reversal
properties of F.T.
applied for $H(f)$**

Input and Output Relationships

Autocorrelation

$$\underline{R_y(\tau)} = E[y(t)y(t+\tau)] = \underline{h(\tau)} * \underline{R_{yx}(\tau)}$$

derive by yourself!

→ $R_y(\tau) = h(\tau) * h(-\tau) * R_x(\tau)$ Expression for $R_{yx}(\tau)$ from last page

FT
PSD

$$S_Y(f) = H(f)H^*(f)S_X(f)$$

FT on both side

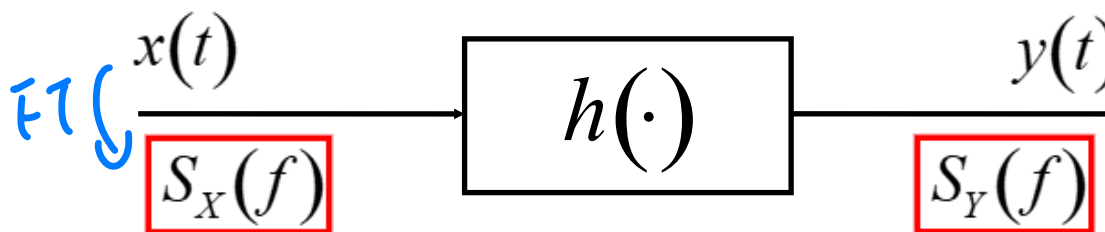
simple relationship!

know noise

the power!

$$\underline{S_Y(f) = |H(f)|^2 S_X(f)}$$

Key Result: PSDs of the input and output random processes



Analysis the noise

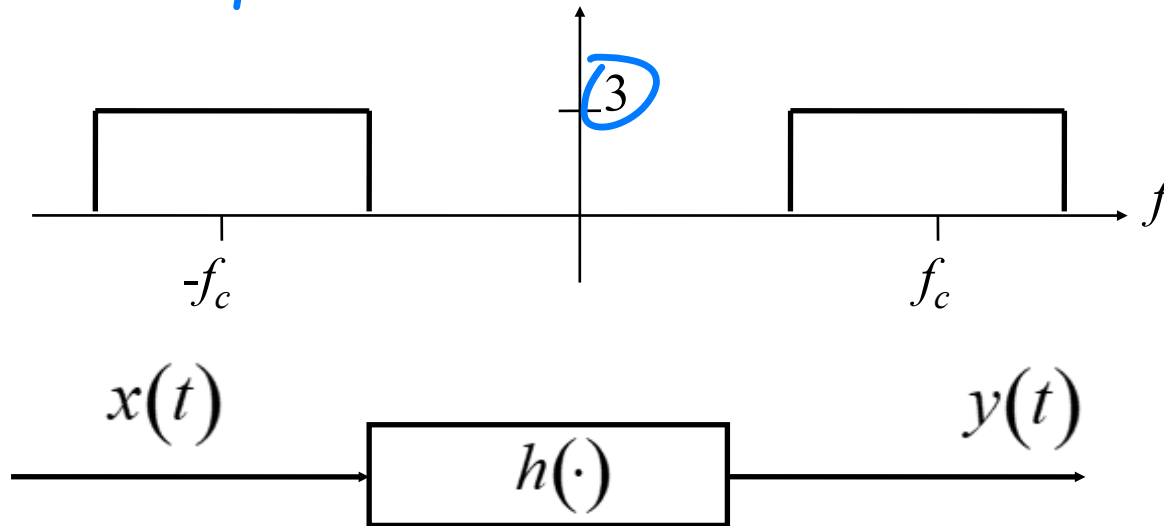
Output PSD

$$S_Y(f) = |H(f)|^2 S_X(f)$$

Input PSD

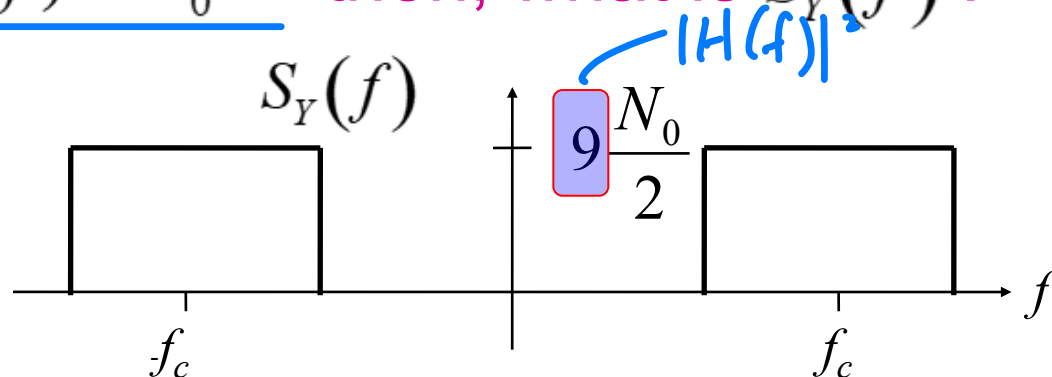
Input and Output Relationships: Example

↓
 AWGN → Apply filter → $H(f)$ filter particular spectrum



AWGN noise!

- If $S_X(f) = N_0/2$ then, what is $S_Y(f)$?



Ch7.2: Optimal Receiver

- ❑ Error Probability for General Signals & Receivers
 - ❑ Optimum Threshold

- ❑ Input-Output Relation

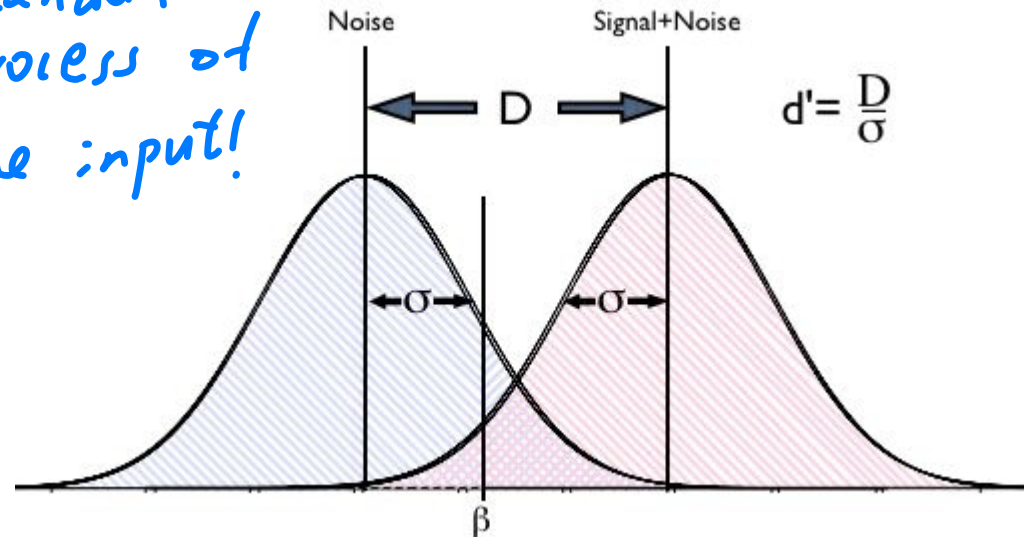
- ❑ **The Matched Filter**

- ❑ **Optimal Receiver**

- ❑ Examples

- ❑ Optimal Signals

Random process of the input!



Optimal Receiver

- We want to find a receiver structure that can minimize the BER

$$P_e = Q \left[\sqrt{\frac{(s_{o1}(T) - s_{o0}(T))^2}{4\sigma^2}} \right]$$

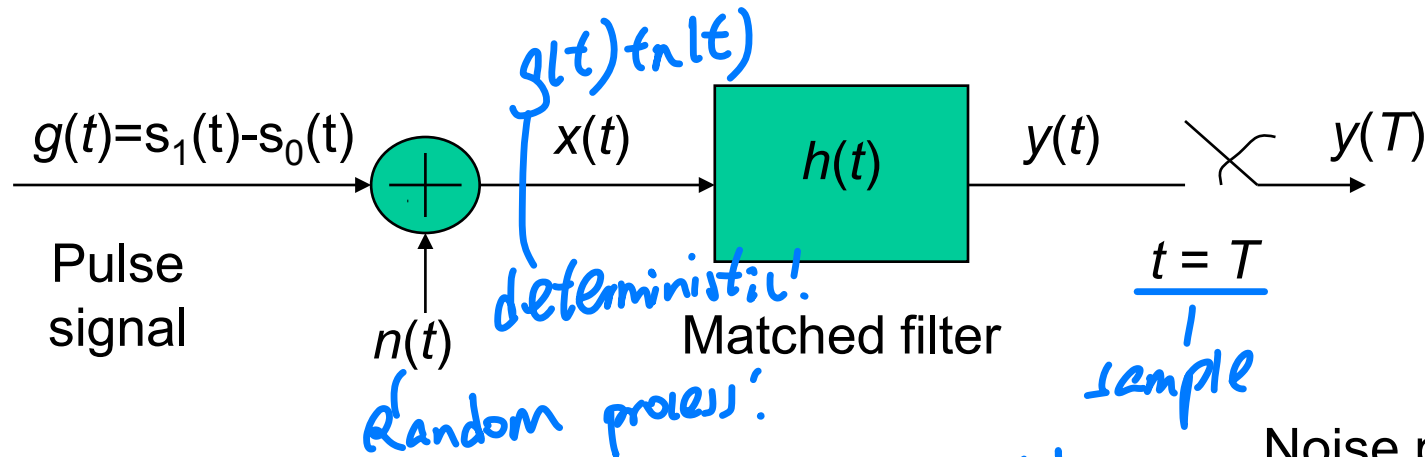
Q function
反轉!

- This is equivalent to maximize the term

$$\zeta^2 = \frac{(s_{o1}(T) - s_{o0}(T))^2}{\sigma^2} = \frac{|g_{o1}(T)|^2}{\sigma^2}$$

where $s_{o1}(T) - s_{o0}(T)$ can be regarded as the output of the filter to input signal $g(t) = s_1(t) - s_0(t)$ at time T.

Matched Filter Derivation



- **Noise** $w(t) = \underline{n(t) * h(t)}$

output!
PSD

$$S_W(f) = S_N(f) S_H(f) = \frac{N_0}{2} |H(f)|^2$$

AWGN

Filter

input!
PSD
= noise $S_N(f)$
plug in!

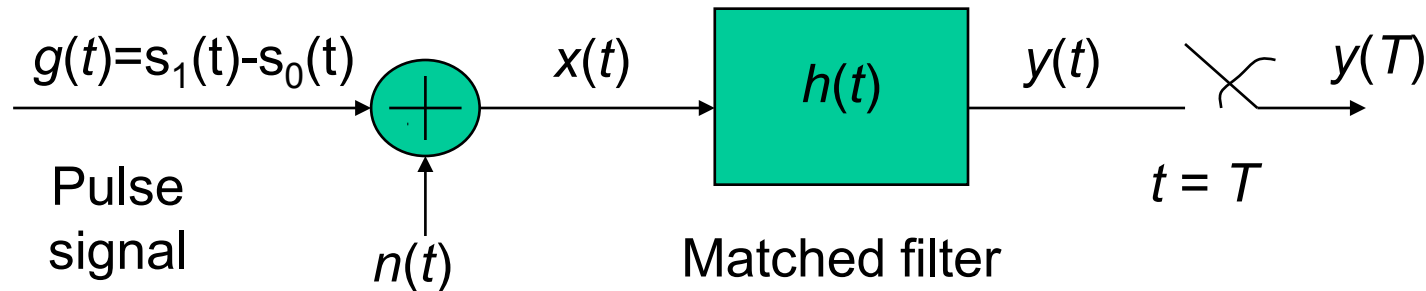
Noise power spectrum $S_N(f)$

$$\frac{N_0}{2f}$$

$$\underline{\underline{\sigma^2 = E\{w^2(T)\}}} = \int_{-\infty}^{\infty} S_W(f) df = \frac{N_0}{2} \int_{-\infty}^{\infty} |H(f)|^2 df$$

noise power!

Matched Filter Derivation



- Signal** $\underline{g_o(t)} = \overset{\text{input!}}{\underline{g(t)}} * h(t) \quad \underline{G_o(f)} = \underline{H(f)G(f)}$

$$g_o(t) = \int_{-\infty}^{\infty} \overset{\text{inverse FT}}{\boxed{H(f)G(f)}} e^{j 2 \pi f t} df$$

$$|g_o(T)|^2 = \left| \int_{-\infty}^{\infty} H(f) G(f) e^{j 2 \pi f T} df \right|^2$$

Matched Filter Derivation

- Find $h(t)$ that maximizes pulse peak SNR

$$\zeta^2 = \frac{\left| \int_{-\infty}^{\infty} \underbrace{H(f)}_{x(f)} \underbrace{G(f)}_{y(f)} e^{j 2 \pi f T} df \right|^2}{\frac{N_0}{2} \int_{-\infty}^{\infty} |H(f)|^2 df}$$

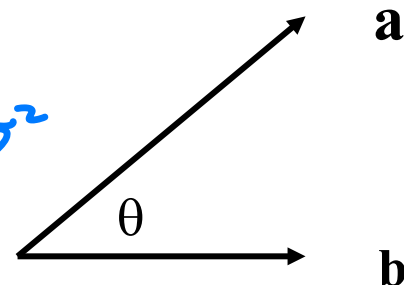
Max this!

signal power!
 $= |g_0(t)|^2$

$$\vec{a} = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$$

$$\vec{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$$

useful



- Schwartz's inequality

same dimension, inner product

For vectors: $|\underline{a^T b}| \leq ||a|| ||b|| \leftrightarrow \cos(\theta) = \frac{a^T b}{||a|| ||b||} \leq 1$



For functions: $|\int_{-\infty}^{\infty} x(t) y^*(t) dt|^2 \leq \int_{-\infty}^{\infty} |x(t)|^2 dt \int_{-\infty}^{\infty} |y(t)|^2 dt$,
 (and the lower bound reached iff $x(t) = k y(t), \forall k \in \mathbb{C}$.)

complex functions

Apply this to maximize!

For functions

$$\left| \underbrace{\int_{-\infty}^{\infty} x(t) y^*(t) dt}_{\text{inner product}} \right|^2 \leq \frac{\int_{-\infty}^{\infty} |x(t)|^2 dt}{\text{norm}^2} \frac{\int_{-\infty}^{\infty} |y(t)|^2 dt}{\text{norm}^2}$$

Matched Filter Derivation

- Let $x(f) = H(f)$ and $y(f) = G^*(f)e^{-j2\pi fT}$ |exp| = 1
- Thus, $|\int_{-\infty}^{\infty} H(f)G(f)e^{j2\pi fT}df|^2 \leq \int_{-\infty}^{\infty} |H(f)|^2 df \int_{-\infty}^{\infty} |G(f)|^2 df$,
which further gives |∫ x(f)y*(f)df| ≤ ∫...∫...

$$\zeta^2 = \frac{|\int_{-\infty}^{\infty} H(f)G(f)e^{j2\pi fT}df|^2}{\frac{N_0}{2} \int_{-\infty}^{\infty} |H(f)|^2 df} \leq \frac{2}{N_0} \int_{-\infty}^{\infty} |G(f)|^2 df$$

independent on H(f)!

- The maximum $\zeta_{\max}^2 = \frac{2}{N_0} \int_{-\infty}^{\infty} |G(f)|^2 df$ occurs when

? x(f) y(f)
 $H_{\text{opt}}(f) = kG^*(f)e^{-j2\pi fT}$ by Schwartz's inequality.
constant!

- Hence, $h_{\text{opt}}(t) = kg(T - t)$. matched filter
typically same as signal!

Matched to what?

- Let $g(t) = [s_1(t) - s_0(t)]$ *apply for any waveform!*

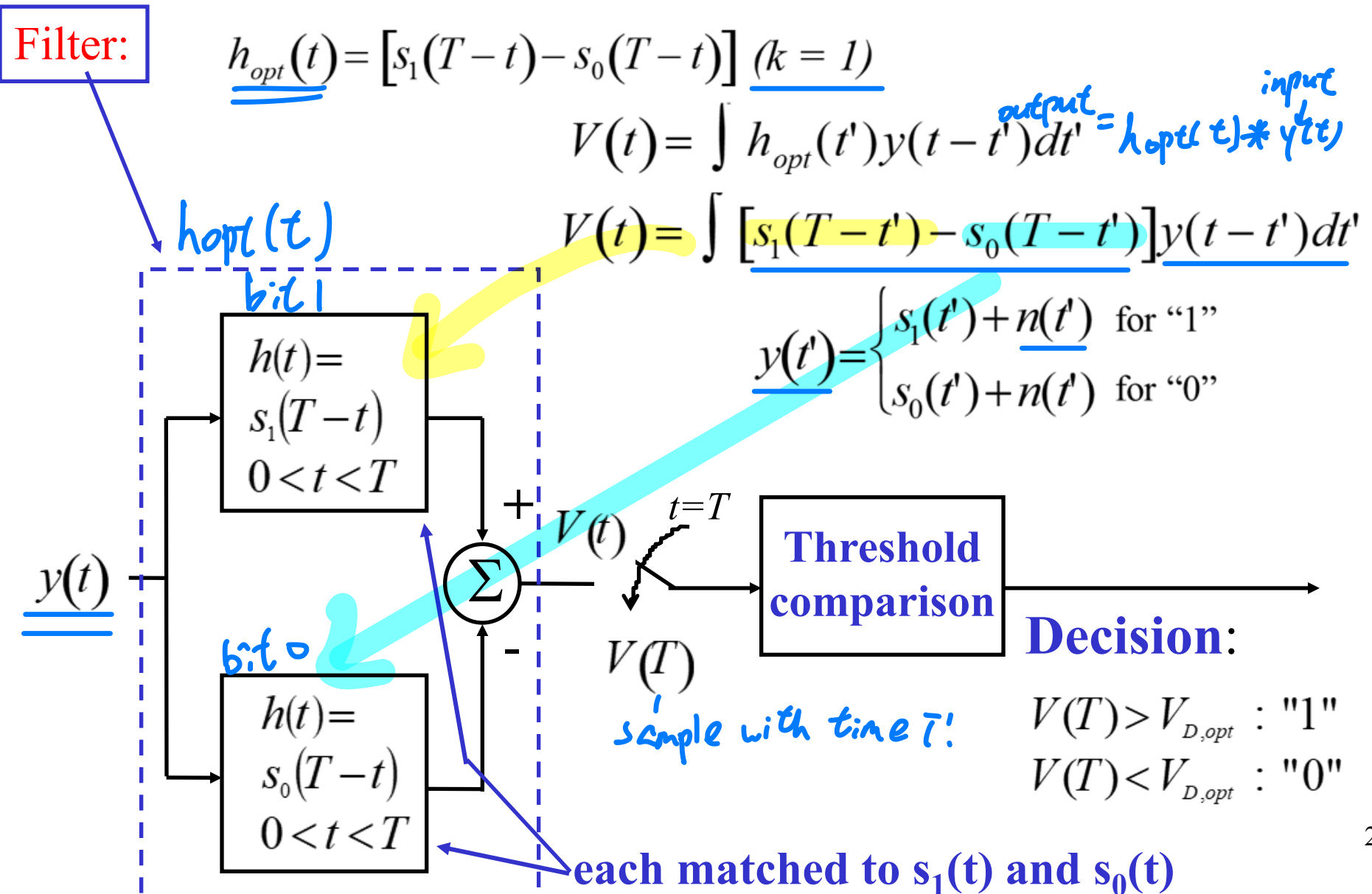
Impulse Response
of Matched Filter

$$h_{opt}(t) = s_1(T - t) - s_0(T - t)$$

optimal threshold!
Second optimum: Optimal Receiver

- The matched filter matches to the difference between the two pulses utilized to transmit bit “1” and bit “0”, respectively.
- What can we see in frequency domain?

Receiver Block Diagram



Ch7.2: Optimal Receiver

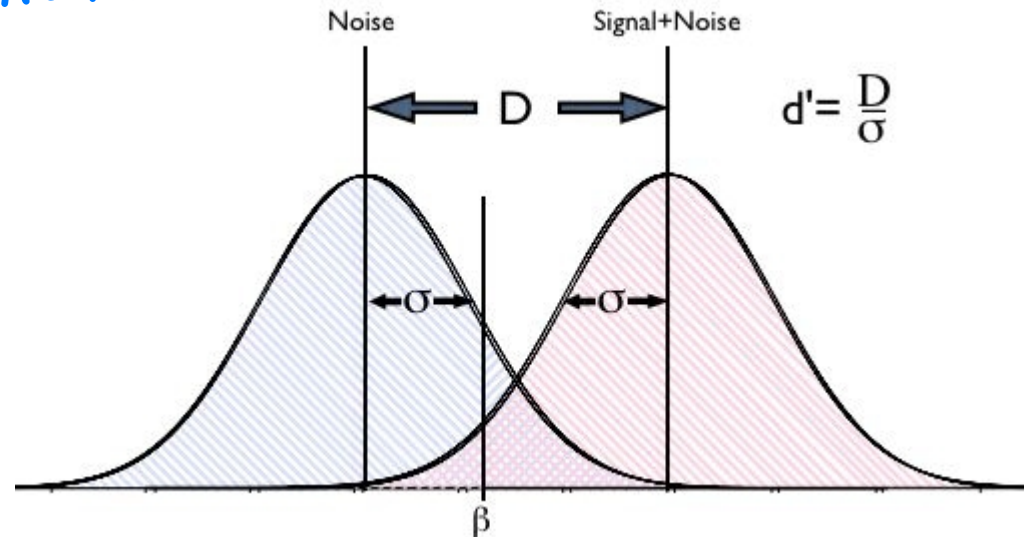
- ❑ Error Probability for General Signals & Receivers
 - ❑ Optimum Threshold

- ❑ Input-Output Relation

- ❑ **The Matched Filter**

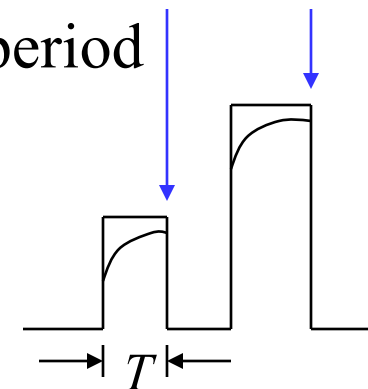
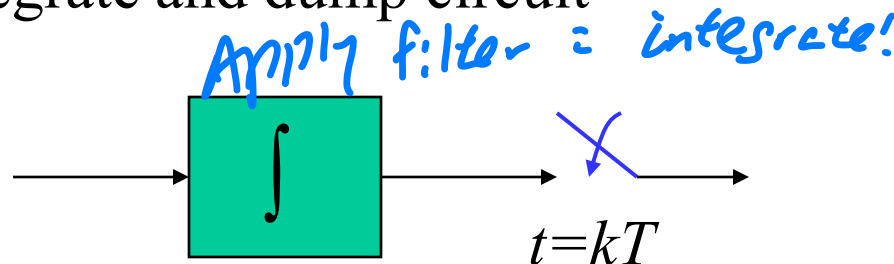
- ❑ Optimal Receiver
- ❑ **Examples**
- ❑ Optimal Signals

another filter!



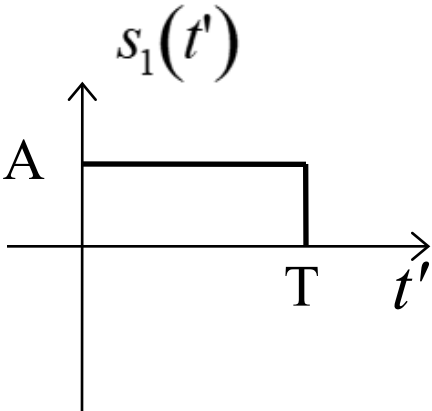
Matched Filter for Rectangular Pulse

- Matched filter for causal rectangular pulse has an impulse response that is a causal rectangular pulse.
- Convolve input with rectangular pulse of duration T sec and sample result at T sec is same as to:
 - First, integrate for T sec \rightarrow in 7.1
 - Second, sample at symbol period T sec
 - Third, reset integration for next time period
- Integrate and dump circuit



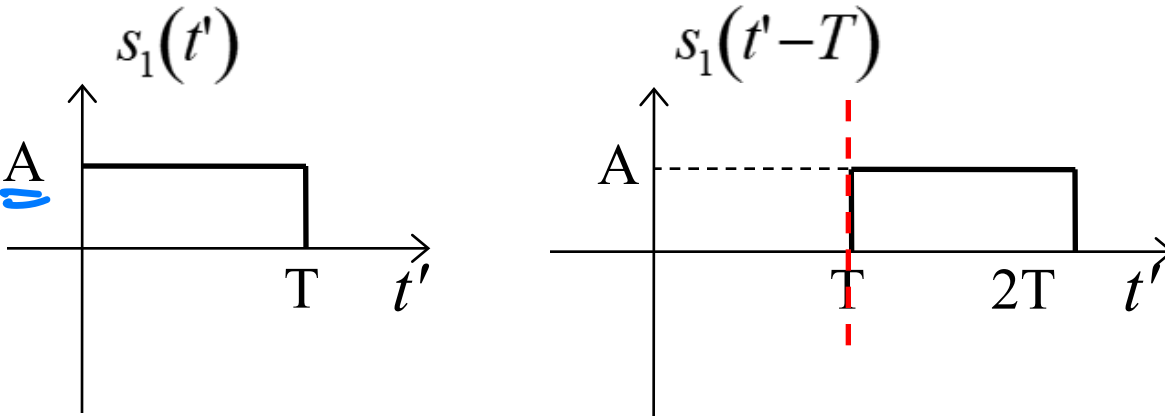
Matched Filter: Example

Antipodal Baseband Signal



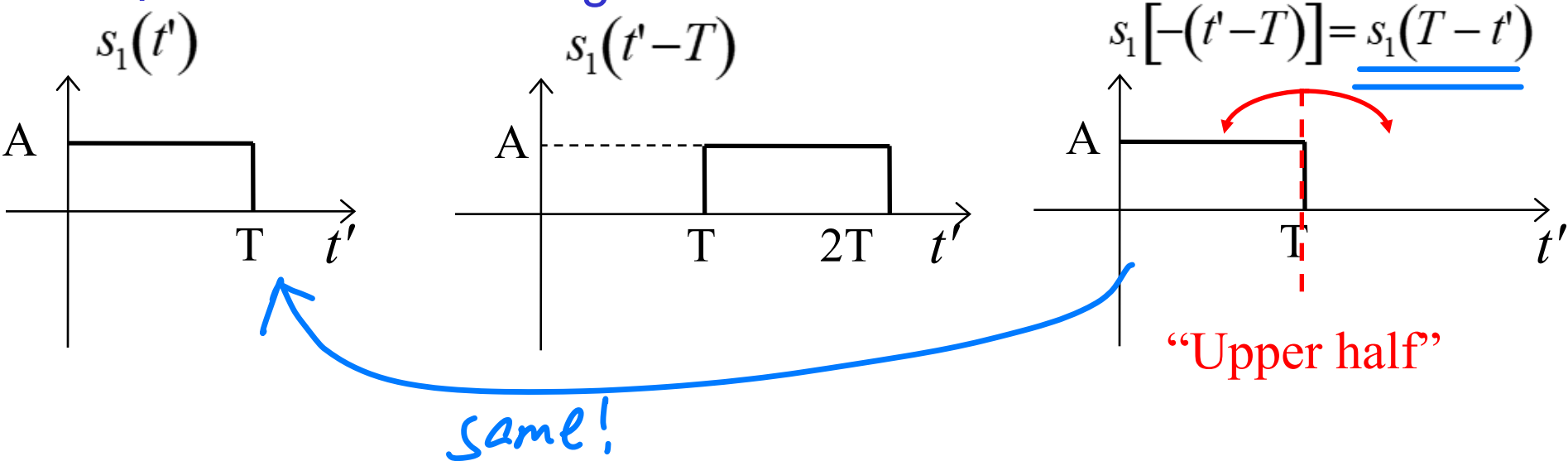
Matched Filter: Example

Antipodal Baseband Signal



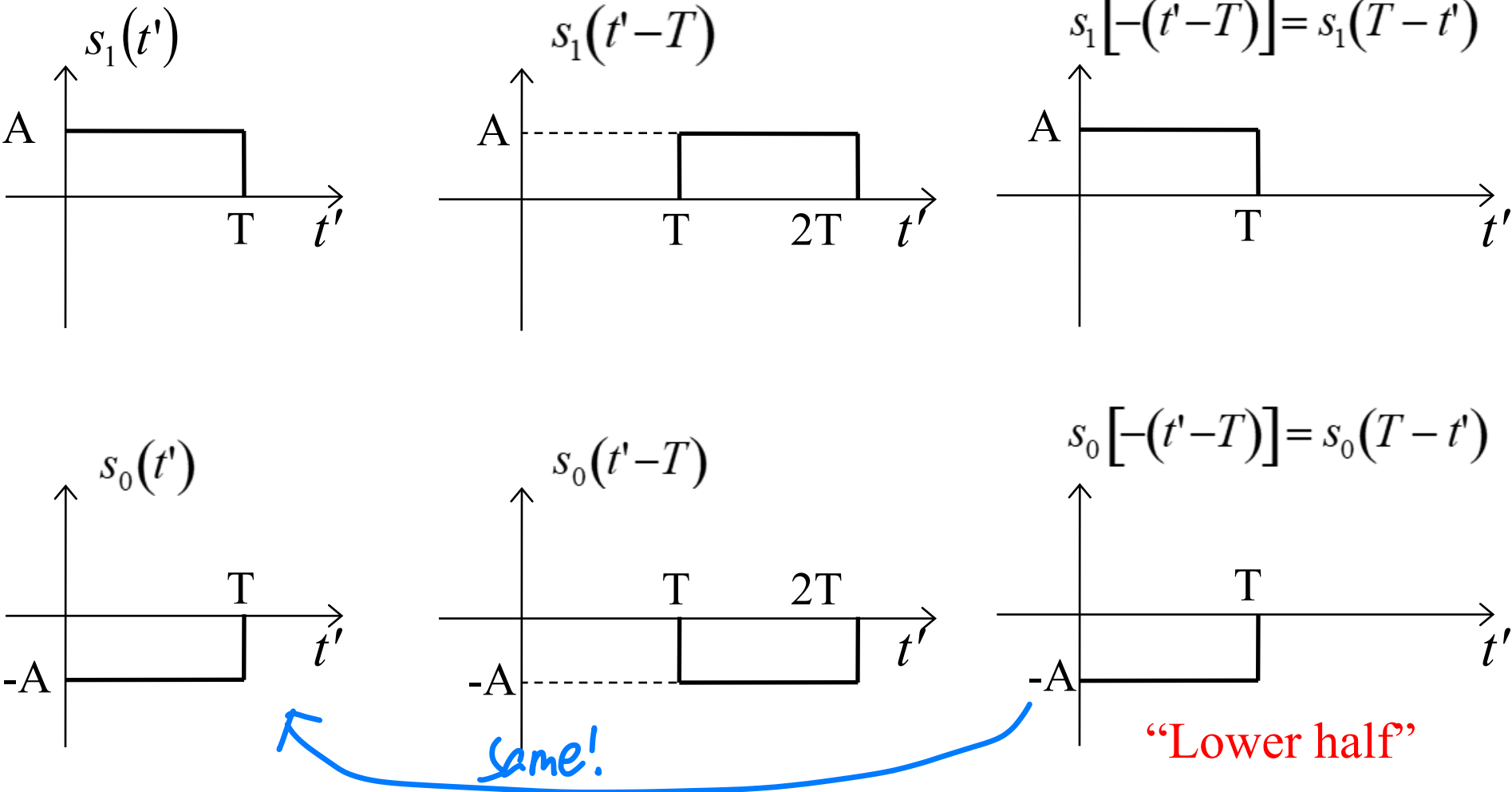
Matched Filter: Example

Antipodal Baseband Signal



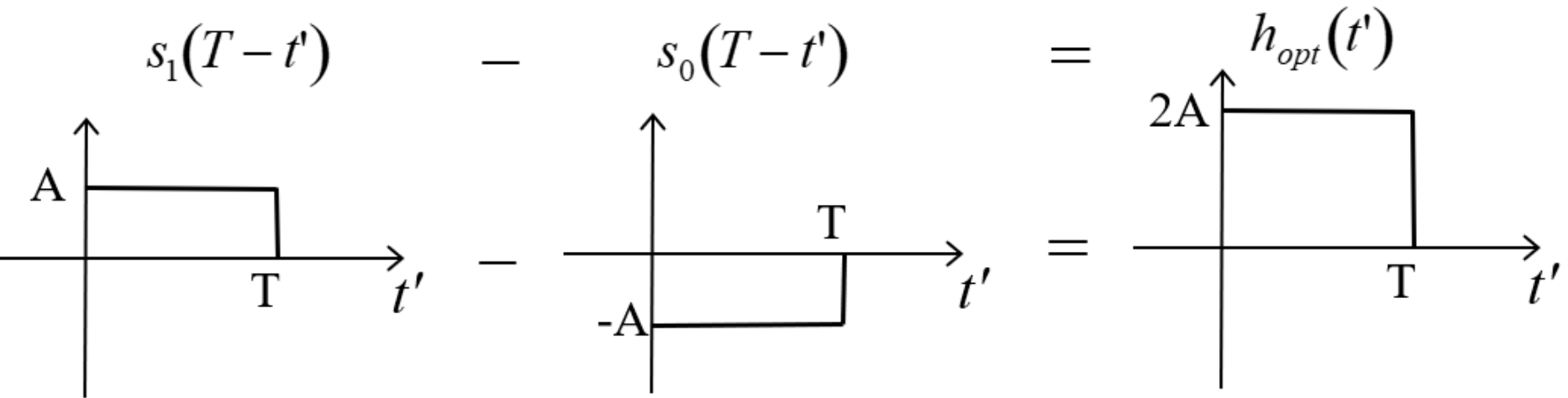
Matched Filter: Example

Antipodal Baseband Signal



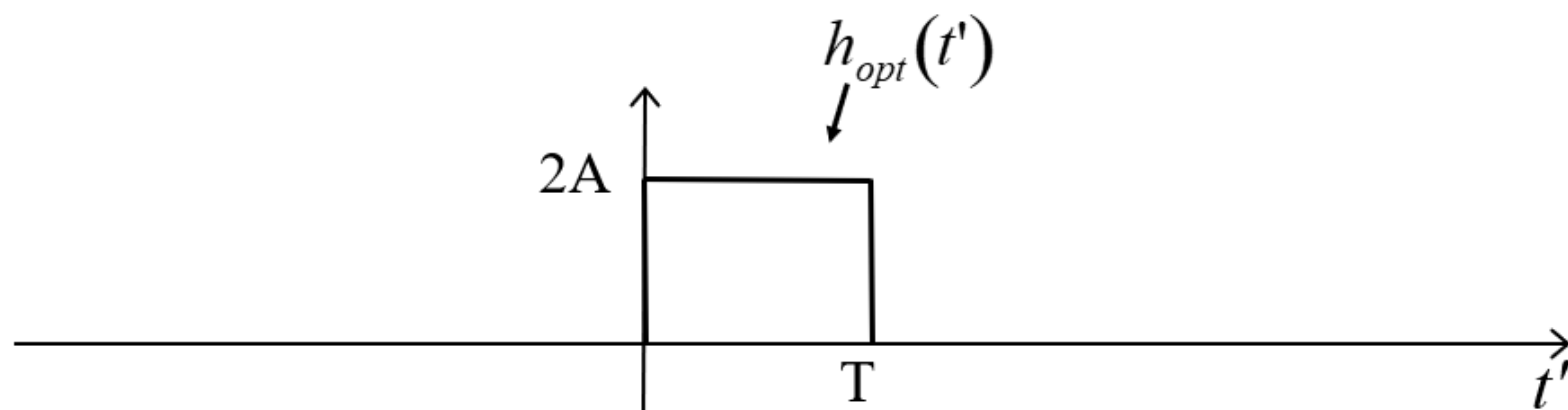
$$h_{opt}(t) = [s_1(T - t') - s_0(T - t')]$$

overall = difference!



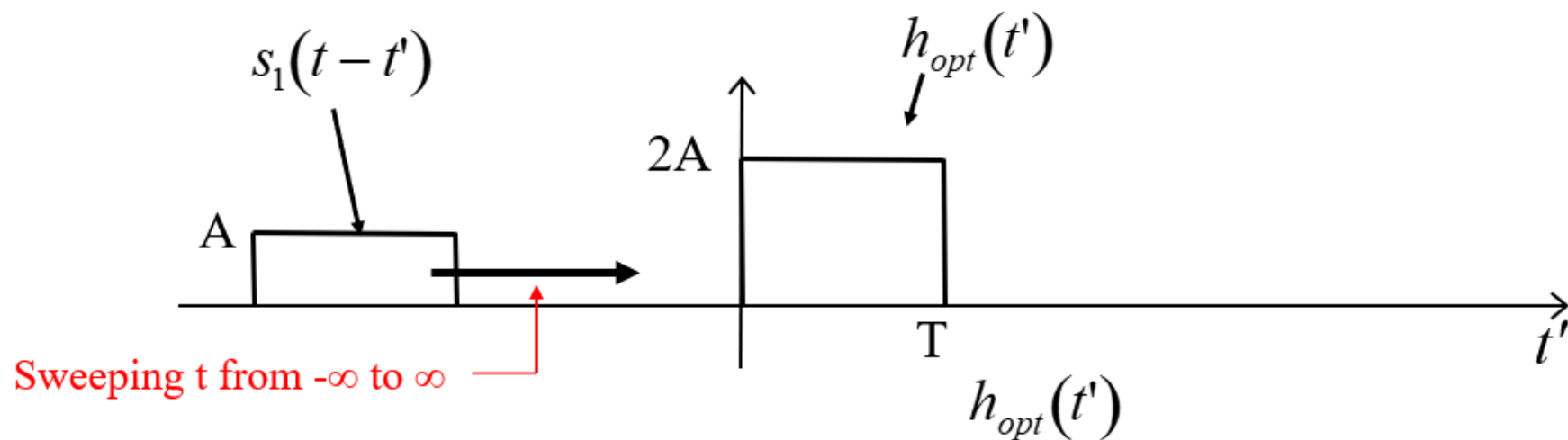
$$V(t) = \int [s_1(T-t') - s_0(T-t')] y(t-t') dt'$$

Case 1: $y(t') = s_1(t')$ “1”



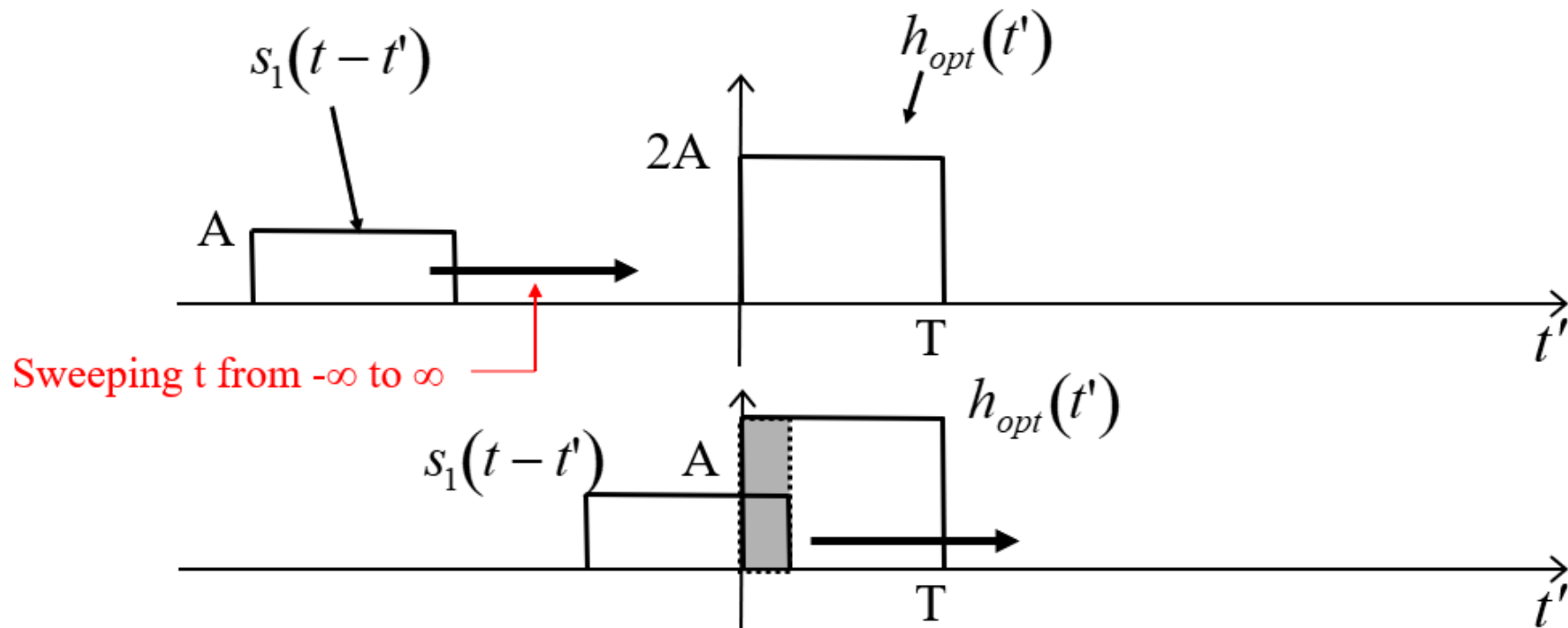
$$V(t) = \int [s_1(T - t') - s_0(T - t')] \boxed{y(t - t')} dt'$$

Case 1: $\boxed{y(t') = s_1(t')}$ “1”



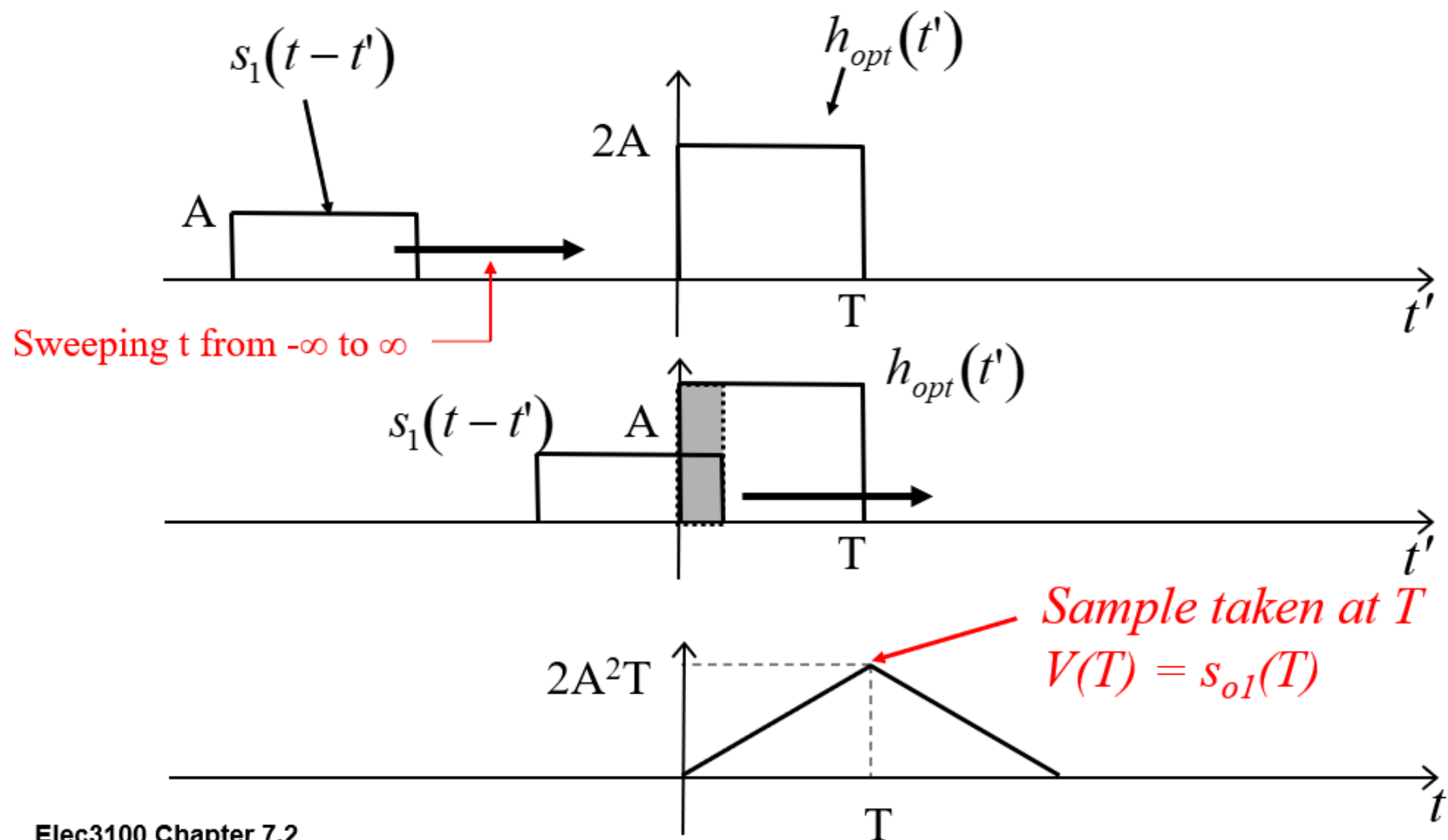
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Case 1: $y(t') = s_1(t')$ “1”



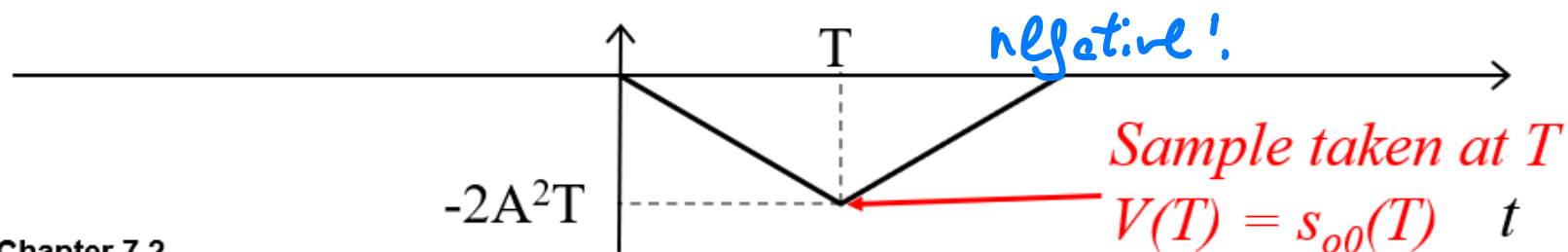
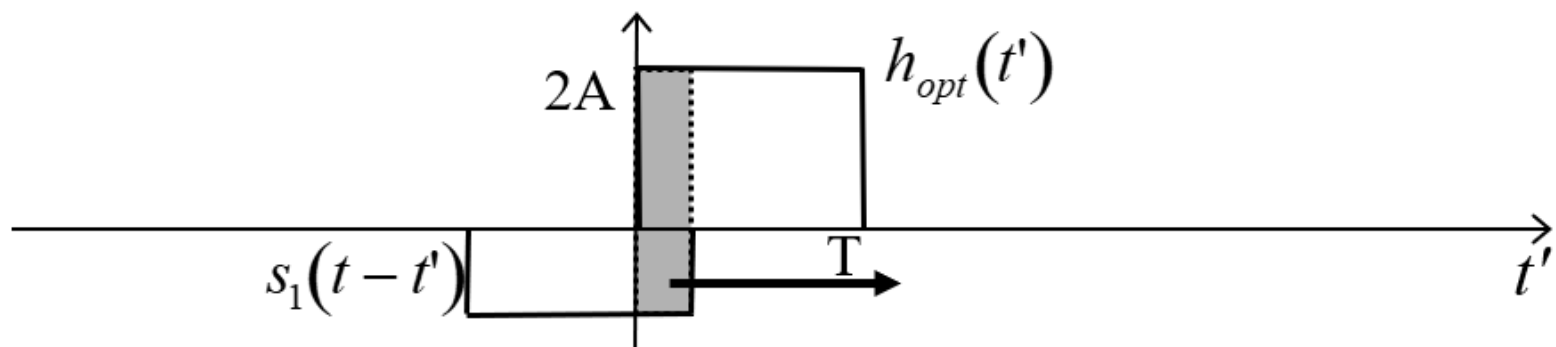
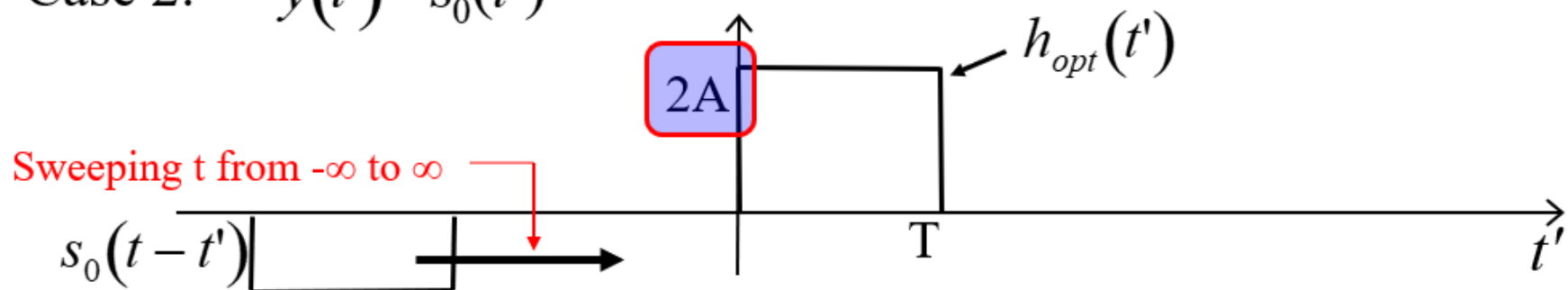
$$V(t) = \int [s_1(T - t') - s_0(T - t')] y(t - t') dt'$$

Case 1: $y(t') = s_1(t')$ "1"



$$V(t) = \int [s_1(T - t') - s_0(T - t')] y(t - t') dt'$$

Case 2: $y(t') = s_0(t')$ "0"



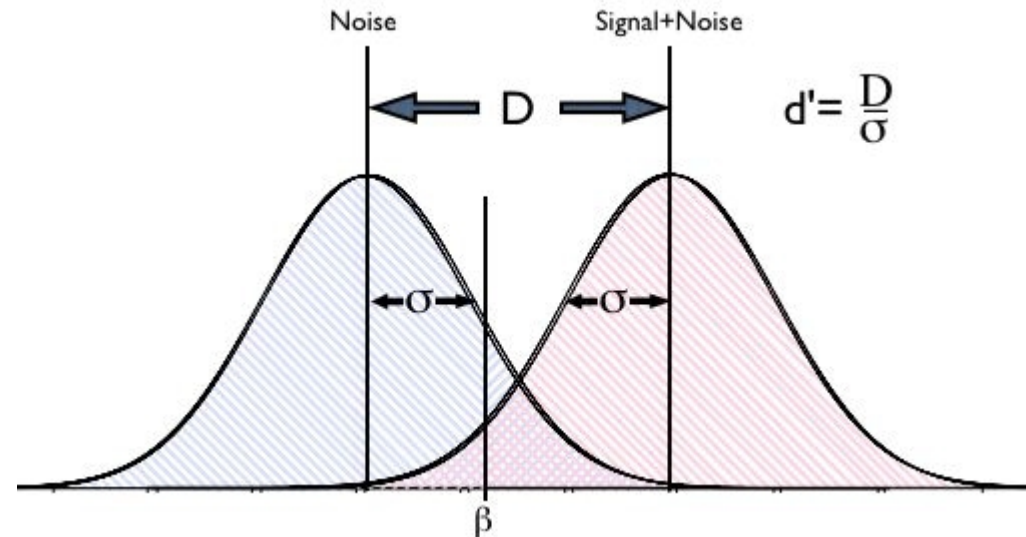
Ch7.2: Optimal Receiver

- ❑ Error Probability for General Signals & Receivers
 - ❑ Optimum Threshold

- ❑ Input-Output Relation

- ❑ **The Matched Filter**

- ❑ Optimal Receiver
- ❑ Examples
- ❑ **Optimal Signals**



Optimal Signal

$$BER = Q(\xi)$$

From Schwarz inequality,

Parseval's Theorem

$$\xi_{\max}^2 = \frac{2}{N_0} \int_{-\infty}^{\infty} |G(f)|^2 df = \frac{2}{N_0} \int_{-\infty}^{\infty} g^2(t) dt$$

$$\xi_{\max}^2 = \frac{2}{N_0} \int_0^T (s_1(t) - s_0(t))^2 dt = \frac{2E_g}{N_0}$$

Want this
be large!

Eg. Not same as E_b

$$E_g = \int_0^T (s_1(t) - s_0(t))^2 dt$$

Third optimum: Optimal Signal

Physical meaning?

$$= \int_0^T s_1^2(t) dt + \int_0^T s_0^2(t) dt - 2 \int_0^T s_1(t) s_0(t) dt$$

Want negative
correlation!

$$= E_1 + E_0 - 2\rho_{10} \sqrt{E_1 E_0}$$

where ρ_{10} is the correlation coefficient

$$-1 \leq \rho_{10} = \frac{1}{\sqrt{E_1 E_0}} \int_0^T s_1(t) s_0(t) dt \leq 1$$

not only Energy,
but also
correlation!

Cross correlation!
-1 \Rightarrow optimal!

Optimal Receiver: Matched Filter

- The **Optimal Filter** is the Matched Filter

$$h_{opt}(t) = s_1(T-t) - s_0(T-t)$$

$$P_e = Q \left[\sqrt{\frac{(s_{o1} - s_{o0})^2}{4\sigma^2}} \right] = Q \left[\sqrt{\frac{\zeta_{\max}^2}{4}} \right]$$

$$= Q \left[\sqrt{\frac{E_g}{2N_0}} \right]$$

maximize this:

$$E_g = E_1 + E_0 - 2\rho_{10}\sqrt{E_1E_0}$$

NOT E_b

Summary



We have three things to optimize:

- **O1: Optimal Signal** *Eg*
Make the two signals as **dissimilar** as possible (Why?)
- **O2: Optimal Processing Unit**
Maximize the **difference** between two signals (How?)
- **O3: Optimum Threshold** (We like **symmetry**, right?)
Make good use of the signal difference

Rx {

In your opinion, which one is the most important?

Next time → optimal ratio