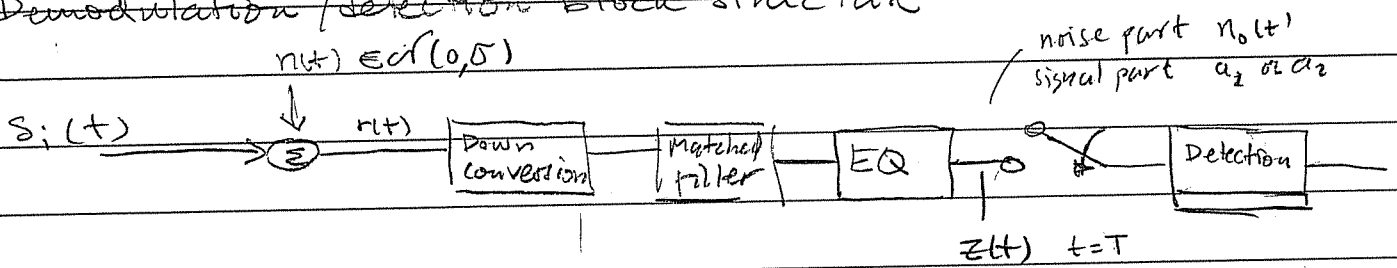


Detection of Binary signals in Noise

Demodulation / detection block structure



at sampling point  $t$  filtered Gaussian = Gaussian,  $\mathcal{N}(0, \sigma_0)$

$$z(T) = a_i(T) + n_0(t) \quad (1) \quad i = 1, 2 \quad (\text{Binary})$$

$a_1 \leftrightarrow$  binary "one"

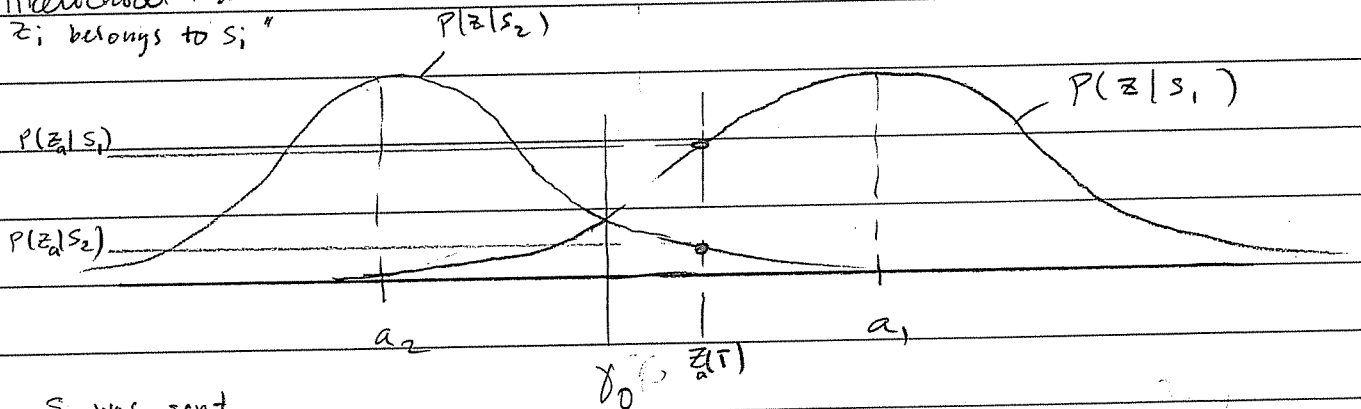
$a_2 \leftrightarrow$  "zero"

Thus,  $z(T) \in \mathcal{N}(a_i, \sigma_0)$

If  $s_i$  was transmitted the pdf is

$$P(z | s_i) = \frac{1}{\sigma_0 \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{z - a_i}{\sigma_0} \right)^2} \quad (2)$$

"Likelihood that  $z_i$  belongs to  $s_i$ "



$s_1$  was sent

$$z(T) \underset{H_2}{\overset{H_1}{\gtrless}} z_0 \quad (3)$$

$s_2$  was sent

When do we choose  $S_1$  and when do we choose  $S_2$ ?

(App B.2.2)

A Reasonable decision rule is:

$$P(S_1|Z) \stackrel{H_1}{\geq} P(S_2|Z) \tag{4}$$

$H_2$

Aposteriori probability

observation

Choose  $H_1$  if the a posteriori probability  $P(S_1|Z)$  is greater than  $P(S_2|Z)$ , other wise choose  $H_2$

By the use of Bayes' Theorem;

$$P(S_i|Z_j) = \frac{p(Z_j|S_i) P(S_i)}{p(Z_j)} \quad \begin{matrix} \text{Aposteriori} & \text{Likelihood that } Z_j \text{ belongs to } S_i & \text{A priori probability} \\ i=1, \dots, M \\ j=1, \dots \end{matrix}$$

$p(Z_j)$  ← normalized factor probability over all events

$Z_j = j$ th sample of  $Z$

$$p(Z_j) = \sum_{i=1}^M p(Z_j|S_i) P(S_i) \quad \text{— probability of the received sample } Z_j \text{ over all signal classes } S_i$$

Note that  $p(Z_j)$  and  $p(Z_j|S_i)$  are continuous since  $Z_j$  is continuous because of the noise which is continuous.

we may rewrite (4) as

$$\frac{p(Z|S_1) P(S_1)}{p(Z)} \stackrel{H_1}{\geq} \frac{p(Z|S_2) P(S_2)}{p(Z)} \tag{5}$$

$H_2$

$$\frac{p(Z|S_1)}{p(Z|S_2)} \stackrel{H_1}{\geq} \frac{P(S_2)}{P(S_1)} \tag{6}$$

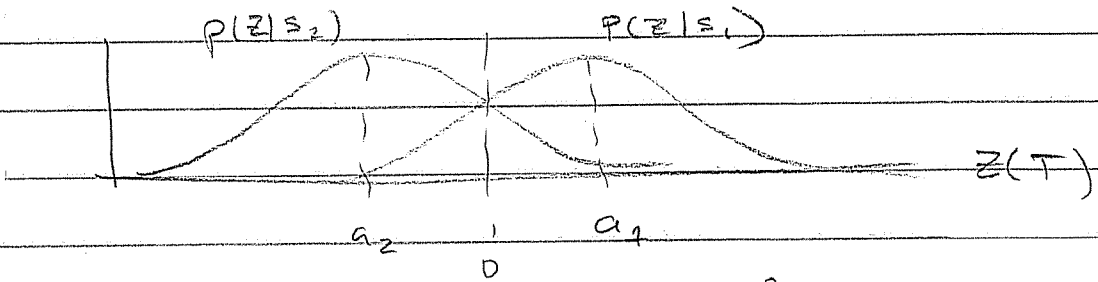
$H_2$

(6) is called the likelihood ratio test (MAP criterion or min err criterion)

If we do not know anything about the a priori probabilities, then set  $P(s_2) = P(s_1)$ . The MAP criteria is also called the maximum likelihood criteria

$$\Lambda(z) \triangleq \frac{p(z|s_1)}{p(z|s_2)} \underset{H_2}{\overset{H_1}{>}} 1 \quad (7)$$

Example:



Let  $p(n_0) = \frac{1}{\sigma_0 \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{n_0}{\sigma_0} \right)^2}$  (8)

then  $\Lambda(z)$  is given by

$$\frac{\frac{1}{\sigma_0 \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{z-a_1}{\sigma_0} \right)^2}}{\frac{1}{\sigma_0 \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{z-a_2}{\sigma_0} \right)^2}} \underset{H_2}{\overset{H_1}{>}} 1 \quad (9)$$

$$e^{\left\{ \frac{z(a_1-a_2)}{\sigma_0^2} - \frac{a_1^2-a_2^2}{2\sigma_0^2} \right\}} \underset{H_2}{\overset{H_1}{>}} 1 \quad (10) \quad [\text{take the loge}]$$

$$\frac{z(a_1-a_2) - \frac{(a_1^2-a_2^2)}{2}}{\sigma_0^2} \underset{H_2}{\overset{H_1}{>}} 0 \quad (11)$$

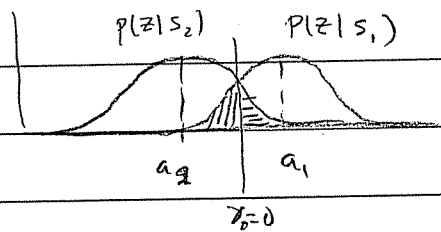
$$z \underset{H_2}{\overset{H_1}{>}} \frac{(a_1^2-a_2^2)}{2(a_1-a_2)} = \frac{a_1+a_2}{2} \triangleq \gamma_0 \quad (12)$$

↙ optimum threshold.

If  $a_1 = 1$  and  $a_2 = -1$ , then  $\gamma_0 = 0$  and  $z \underset{H_2}{\overset{H_1}{>}} 0$  which is a threshold at zero.

# Error probability

Assume binary decision making.



Two possibilities for an error  $e$  to occur:

1.  $S_1$  is sent but the detector selects  $H_2$  i.e.  $z < 0$
2.  $S_2$  is sent " " " "  $H_1$  i.e.  $z > 0$

The probability for these occurrences are

$$P(e|s_1) = P(H_2|s_1) = \int_{-\infty}^0 p(z|s_1) dz \quad (13)$$

$$P(e|s_2) = P(H_1|s_2) = \int_0^{\infty} p(z|s_2) dz$$

The probability of a bit error is then the sum of all error events i.e. (App B.1.1)

$$P_B = P(e) = \sum_{i=1}^2 P(e|s_i) P(s_i) = P(e|s_1) P(s_1) + P(e|s_2) P(s_2) = P(H_2|s_1) P(s_1) + P(H_1|s_2) P(s_2) \quad (14)$$

↑ (Probability of an error if  $s_i$  was sent)

If now  $P(s_1) = P(s_2) = \frac{1}{2}$  and if the pdf's are symmetric we obtain

$$P_B = \frac{1}{2} P(H_2|s_1) + \frac{1}{2} P(H_1|s_2) = P(H_2|s_1) = P(H_1|s_2)$$

$$P_B = P(H_2|s_1) = \int_{-\infty}^0 p(z|s_1) dz = \int_0^{\infty} p(z|s_2) dz \quad (15)$$

Now, by inserting  $P(z|s_2) = N(a_2, \sigma_0)$ , we obtain

$$P_B = \int_{-\infty}^{\infty} \frac{1}{\sigma_0 \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{z-a_2}{\sigma_0} \right)^2} dz \quad (16)$$

Substituting  $\frac{z-a_2}{\sigma_0}$  for  $u$  gives  $\left[ \frac{du}{dz} = \frac{1}{\sigma_0} \right]$

$$P_B = \int_{\frac{a_1-a_2}{2\sigma_0}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du \triangleq Q\left(\frac{a_1-a_2}{2\sigma_0}\right) \quad (17)$$

$Q(x) \triangleq \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-\frac{u^2}{2}} du$  is called the  $Q(\cdot)$  function (18)  
 or the complementary error function  
 or co-error function.

Another form of the co-error function is

$$\text{erfc}(x) \triangleq \frac{2}{\sqrt{\pi}} \int_x^{\infty} e^{-u^2} du \quad (19)$$

$Q(x)$  and  $\text{erfc}(x)$  are related by

$$\begin{aligned} \text{erfc}(x) &= 2Q(x\sqrt{2}) \\ Q(x) &= \frac{1}{2} \text{erfc}\left(\frac{x}{\sqrt{2}}\right) \end{aligned} \quad (20)$$

They can not be evaluated in closed form and are therefore tabulated. See Table B.1.

A good approximation is (for  $x > 3$ )

$$Q(x) \approx \frac{1}{x\sqrt{2\pi}} e^{-\frac{x^2}{2}} \quad (21)$$

## Optimizing error performance

For binary signaling (baseband) we know that

$$P_B = Q\left(\frac{a_1 - a_2}{2\sigma_0}\right) = Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-\frac{u^2}{2}} du \quad (22)$$

where  $a_1 - a_2$  is the distance between the binary signals.

To minimize  $P_B$  we need to maximize

$$x \triangleq \frac{a_1 - a_2}{2\sigma_0} \quad (= \text{minimize integral}) \quad (23)$$

If the matched filter is matched to the input difference signal  $(s_1(t) - s_2(t))$  (= maximize energy of distance)

$$\left(\frac{S}{N}\right)_T = \frac{(a_1 - a_2)^2}{\sigma_0^2} = \frac{E_d}{\frac{N_0}{2}} = \frac{2E_d}{N_0} \quad (24)$$

where

$$E_d = \int_0^T [s_1(t) - s_2(t)]^2 dt \quad (25)$$

Inserted into (22) we obtain

$$P_B = Q\left(\sqrt{\frac{E_d}{2N_0}}\right) \quad (26)$$

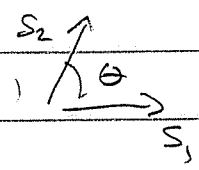
where  $E_d$  is the energy of the matched filter input and  $\frac{N_0}{2}$  is the input noise spectral density.

Calculate  $E_d$ :

$$E_d = \int_0^T (s_1(t) - s_2(t))^2 dt = \underbrace{\int_0^T s_1^2 dt}_{E_b} + \underbrace{\int_0^T s_2^2 dt}_{E_b} - 2 \int_0^T s_1 s_2 dt \quad (27)$$

Now, let the correlation between  $s_1$  and  $s_2$  be defined as

$$\rho = \frac{1}{E_b} \int_0^T s_1(t) s_2(t) dt = \cos \theta \quad (28)$$



Normalization with Energy/bit

Inserting (28) into (27) gives

$$E_d = 2E_b - 2\rho E_b = 2E_b(1-\rho) \quad (29)$$

which inserted into (26) yields the probability of error

$$P_B = Q\left(\sqrt{\frac{E_d}{2N_0}}\right) = Q\left(\sqrt{\frac{2E_b(1-\rho)}{2N_0}}\right) = Q\left(\sqrt{\frac{E_b(1-\rho)}{N_0}}\right) \quad (30)$$

$$= \begin{cases} Q\left(\sqrt{\frac{2E_b}{N_0}}\right) & \text{for } \rho = -1 \text{ (antipodal signaling)} \\ Q\left(\sqrt{\frac{E_b}{N_0}}\right) & \text{for } \rho = 0 \end{cases}$$

