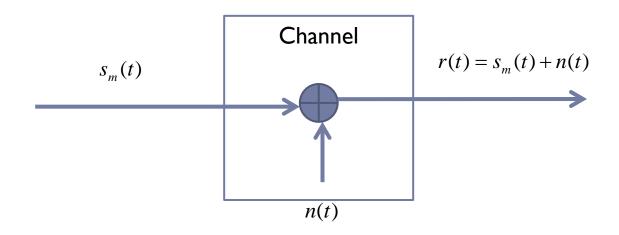
Digital Communications Optimal Receivers

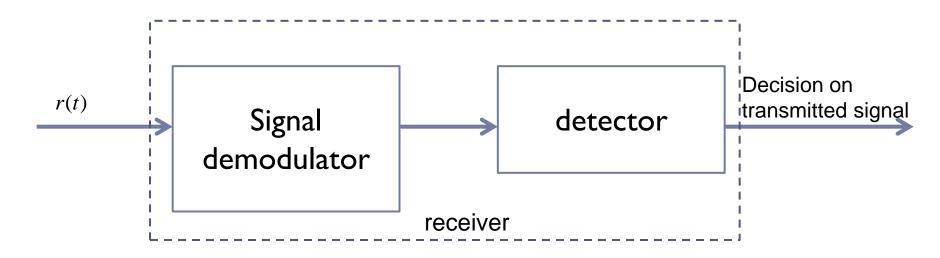
Dott.ssa Ernestina Cianca a.a. 2017-2018



Let us assume that the channel only adds white gaussian noise with power spectral density $N_{\rm 0}/2$

How to process the received signal r(t) in the interval $0 \le t \le T_b$





It is convenient to subdivide the receiver into two parts

<u>Demodulator</u>: to convert the received signal waveform into a vector y whose dimension is equal to the dimension of the transmitted signal waveforms

<u>Detector:</u> to decide which of the two (in case of binary transmission) possible signal waveforms was transmitted, based on observation of the vector *y*



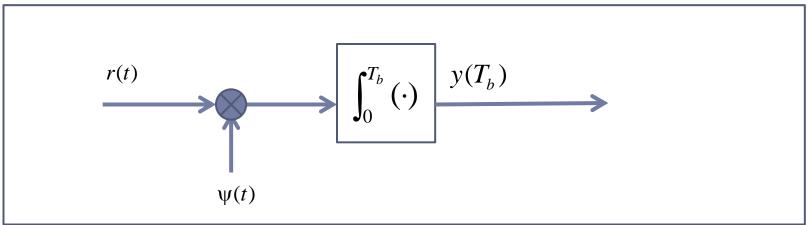
Correlator-type receiver Case #1: binary antipodal signals

$$r(t) = s_m \psi(t) + n(t) \qquad 0 \le t \le T_b$$

where:

 $\Psi^{(t)}$ is the unit energy rectangular pulse

$$s_1 = \sqrt{E_b} \quad s_2 = -\sqrt{E_b}$$



Cross-correlator for binary antipodal

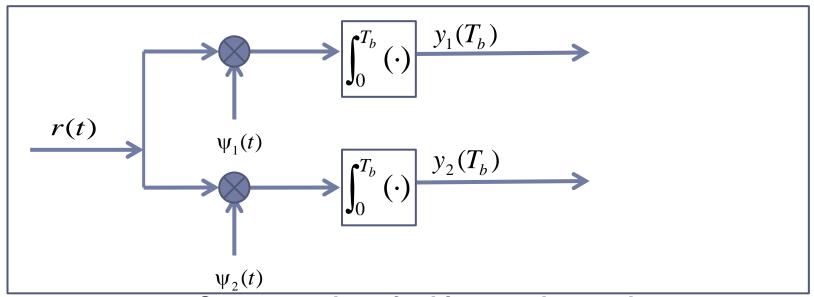
Correlator-type receiver Case #2: binary orthogonal signals

$$r(t) = s_m(t) + n(t) \qquad 0 \le t \le T_h$$

Where $s_m(t)$ is one of the two following orthogonal waveforms:

$$s_1(t) = s_{11} \psi_1(t) + s_{12} \psi_2(t)$$

$$s_2(t) = s_{21} \psi_1(t) + s_{22} \psi_2(t)$$



Fundamental result in communication theory on the:

Detection of a pulse signal of know waveforms that is immersed in additive white noise



MATCHED FILTER



The matched filter (MF) is the optimal linear filter for maximizing the output SNR.



$$S_m(t) + n(t)$$

$$Q_R(\cdot)$$

$$V_o(t) + n_o(t)$$

$$V_o(t_0) + n_o(t_0)$$

$$S_m(t)$$

$$S_m(t)$$

$$S_m(t)$$

$$S_m(t)$$

$$S_m(t)$$

$$S_m(t)$$

$$S_m(t)$$

$$S_m(t)$$

2) n(t) is a zero-mean noise with spectral density $W_{\scriptscriptstyle n}(f)$

Problem: Find
$$q_R(\cdot)$$
 which maximize the ratio $\frac{\left|y_o(t_0)\right|^2}{E\left|n_o(t_0)\right|^2}$



Preliminar considerations:

$$y_o(t) = s_m * q_R(t)$$

$$n_o(t) = n * q_R(t)$$

parseval



$$y_{o}(t_{0}) = \int_{-\infty}^{\infty} q_{R}(\tau) s_{m}(t_{0} - \tau) d\tau = \int_{-\infty}^{\infty} Q_{R}(f) S_{m}(f) e^{j2\pi f t_{0}} df$$

$$E\left[\left|n_{o}(t_{0})\right|^{2}\right] = \int_{0}^{\infty} W_{n}(f)\left|Q_{R}(f)\right|^{2} df =$$

Schwartz's inequality

$$\left| \int_{-\infty}^{\infty} X^*(f) Y(f) df \right|^2 \le \int_{-\infty}^{\infty} |X(f)|^2 df \int_{-\infty}^{\infty} |Y(f)|^2 df$$

Equality holds only if X(f) = kY(f)

Solution:

$$\frac{\left|\frac{y_{o}(t_{0})\right|^{2}}{E\left[\left|n_{o}(t_{0})\right|^{2}\right]} = \frac{\left|\int_{-\infty}^{\infty} Q_{R}(f)S_{m}(f)e^{j2\pi\beta t_{0}}\right|^{2}}{\int_{-\infty}^{\infty} W_{n}(f)\left|Q_{R}(f)\right|^{2}df}$$

$$= \frac{\left|\int_{-\infty}^{\infty} Q_R(f) \sqrt{W_n(f)} \frac{S_m(f)}{\sqrt{W_n(f)}} e^{j2\pi f t_0} df\right|^2}{\int_{-\infty}^{\infty} Q_R(f) \sqrt{W_n(f)} e^{j2\pi f t_0} df$$

$$\int\limits_{0}^{\infty}W_{n}(f)ig|Q_{R}(f)ig|^{2}df$$
 This is the maximum value

$$\int_{-\infty}^{\infty} W_n(f) |Q_R(f)|^2 df \qquad \text{This is the maximum value}$$

$$\leq \int_{-\infty}^{\infty} \left| \frac{S_m(f)}{\sqrt{W_n(f)}} e^{j2\pi f t_0} \right|^2 df \leq \int_{-\infty}^{\infty} \frac{\left|S_m(f)\right|^2}{W_n(f)} df$$



Solution:

The maximum value is achieved when the equality holds, e.g. when:

$$Q_R(f)\sqrt{W_n(f)} = k \left(\frac{S_m(f)}{\sqrt{W_n(f)}}e^{j2\pi f t_0}\right)^{\frac{1}{2}}$$



$$Q_{R}(f) = k \frac{S_{m}^{*}(f)}{W_{n}(f)} e^{-j2\pi f t_{0}}$$



$$Y_o(f) = k \frac{\left|S_m(f)\right|^2}{W_n(f)} e^{-j2\pi f t_0}$$



If the noise is **WHITE noise**

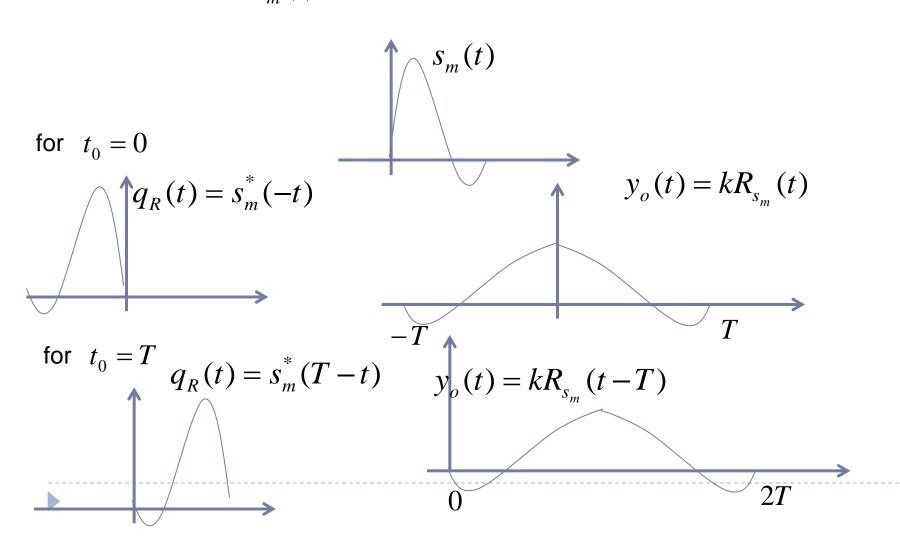
$$W_n(f) = N_0 / 2$$

$$Q_{R}(f) = kS_{m}^{*}(f)e^{-j2\pi f t_{0}} \qquad \qquad \qquad q_{R}(t) = kS_{m}^{*}(t_{0} - t)$$

$$Y_{o}(f) = k\left|S_{m}(f)\right|^{2}e^{-j2\pi f t_{0}} \qquad \qquad \qquad y_{o}(t) = kR_{s_{m}}(t - t_{0})$$



Moreover, if $S_m(t)$ is of finite duration between (0,T)



Moreover, if $S_m(t)$ is of finite duration between (0,T)

$$V_0 = y_o(t_0) = kR_{s_m}(0)$$

$$\sigma^{2} = E \left[n_{o}(t_{0}) \right]^{2} = k^{2} \frac{N_{0}}{2} R_{s_{m}}(0)$$



$$\frac{V_0^2}{\sigma^2} = \frac{2E_{s_m}}{N_0}$$

The peak pulse signal-to-noise ratio of a matched filter depend only on the ratio of the signal energy to the power spectral density of the white noise at the filter input and **NOT** on the particular shape of the waveform that is used



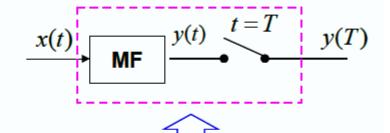
The impulse response of the optimum filter qR, expect for a scaling factor k, is a time-reversed and delayed version of the input signal gR, that is "matched" to the input signal.

No assumption has been made on the statistics of the channel noise, only that is stationary and white.



- Equivalent form Correlator
 - Let $s_i(t)$ be within [0,T]

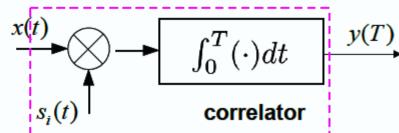
$$y(t) = x(t) * h_m(t) = x(t) * s_i(T - t)$$
$$= \int_0^T x(\tau) s_i(T - t + \tau) d\tau$$



• Observe at sampling time t = T

$$y(T) = \int_0^T x(\tau) s_i(\tau) d\tau = \int_0^T x(t) s_i(t) dt$$

Correlation integration





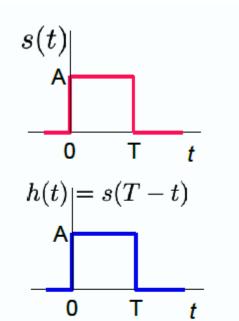
Examples: MATCHED FILTER of a rectangular pulse

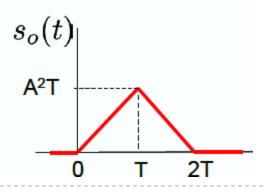
Consider a rectangular pulse s(t)

$$E_s = A^2 T$$

- The impulse response of a filter matched to s(t) is also a rectangular pulse
- The output of the matched filter $s_0(t)$ is h(t) * s(t)
- The output SNR is

$$(SNR)_o = \frac{2}{N_0} \int_0^T s^2(t) dt = \frac{2A^2T}{N_0}$$





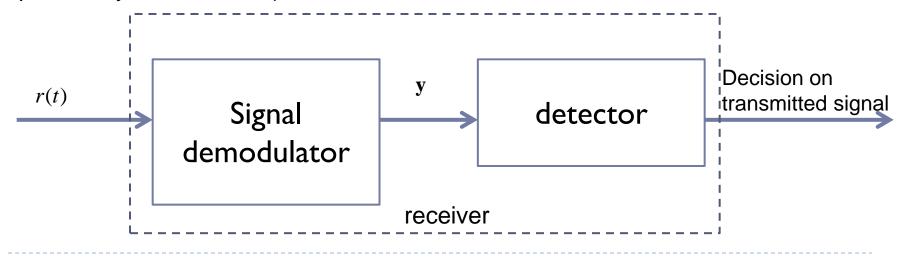


Optimal receiver in AWGN channel

Summary

We have demonstrated that for a signal transmitted over an AWGN channel, either a <u>correlation type demodulator</u> or a <u>matched filter type demodulator</u> produces the vector y which contains all the necessary information in r(t)

Now, we will discuss the design of a **signal detector** that makes a decision of the transmitted signal in each signal interval based on the observation of y, such that the **probability of making an error is minimized** (or correct probability is maximized)

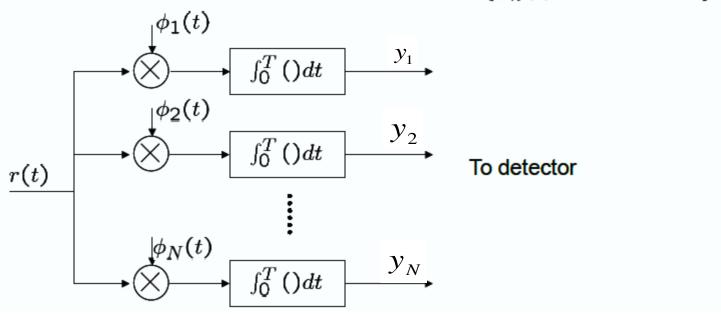




Optimal receiver in AWGN channel

 The received signal r(t) is passed through a parallel bank of N cross correlators which basically compute the projection of r(t) onto the N basis functions

$$\{\phi_k(t), k=1,\dots N\}$$

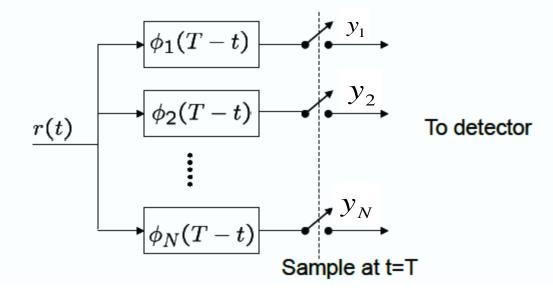




Optimal receiver in AWGN channel

 Alternatively, we may apply the received signal r(t) to a bank of N matched filters and sample the output of filters at t = T. The impulse responses of the filters are

$$h_k(t) = \phi_k(T - t), \quad 0 \le t \le T$$





DETECTION THEORY

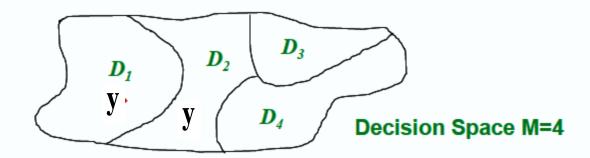
- Given M possible hypotheses H_i (signal m_i) with probability $P_i = P(m_i)$, i = 1, 2, ..., M
 - P_i represents the prior knowledge concerning the probability of the signal m_i – Prior Probability
- The observation is some collection of N real values, denoted by y = (y₁, y₂,...., y_N) with conditional pdf
 f(y|m_i) conditional pdf of observation y given the signal m_i
- Goal: Find the best decision-making algorithm in the sense of minimizing the probability of decision error.





DETECTION THEORY Observation Space

- In general, y can be regarded as a point in some observation space
- Each hypothesis H_i is associated with a decision region D_i:
- The decision will be in favor of H_i if y is in D_i
- Error occurs when a decision is made in favor of another when the signals y falls outside the decision region D_i





 Consider a decision rule based on the computation of the posterior probabilities defined as

$$P(m_i | \mathbf{y}) = P(\text{ signal } m_i \text{ was transmitted given } \mathbf{y} \text{ observed })$$

for $i = 1, ..., M$

- Known as a posterior since the decision is made after (or given) the observation
- Different from the a prior where some information about the decision is known in advance of the observation



- By Bayes' Rule: $P(m_i \mid \mathbf{y}) = \frac{P_i f(\mathbf{y} \mid m_i)}{f(\mathbf{y})}$
- Since our criterion is to minimize the probability of detection error given \mathbf{y} , we deduce that the optimum decision rule is to choose $\hat{m} = m_k$ if and only if $P(m_i | \mathbf{y})$ is maximum for i = k.
- Equivalently,

Choose $\hat{m} = m_k$ if and only if $P_k f(\mathbf{y} \mid m_k) \ge P_i f(\mathbf{y} \mid m_i)$ for all $i \ne k$

 This decision rule is known as maximum a posterior or MAP decision criterion



- If $p_1=p_2=...=p_M$, i.e. the signals $\{m_k\}$ are equiprobable, finding the signal that maximizes $P(m_k|\mathbf{y})$ is equivalent to finding the signal that maximizes $f(\mathbf{y}|m_k)$
- The conditional pdf f(y|m_k) is usually called the likelihood function. The decision criterion based on the maximum of f(y|m_k) is called the Maximum-Likelihood (ML) criterion.
- ML decision rule:

Choose
$$\widehat{m} = m_k$$
 if and only if $f(\mathbf{y} \mid m_k) \ge f(\mathbf{y} \mid m_i)$ for all $i \ne k$

 In any digital communication systems, the decision task ultimately reverts to one of these rules



- If $p_1=p_2=...=p_M$, i.e. the signals $\{m_k\}$ are equiprobable, finding the signal that maximizes $P(m_k|\mathbf{y})$ is equivalent to finding the signal that maximizes $f(\mathbf{y}|m_k)$
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 In any digital communication systems, the decision task ultimately reverts to one of these rules



During a given signaling interval T, a binary baseband system will transmit one of two waveforms

$$s_i(t) = \begin{cases} s_1(t) & 0 \le t < T & \text{for a binary 1} \\ s_2(t) & 0 \le t < T & \text{for a binary 0} \end{cases}$$



$$r(t) = s_i(t) * h_c(t) + n(t)$$
 for $i = 1,2$

At the end of each symbol duration T, the output of the sampler yields a sample y(T) called **test statistic.**

y(T) has a voltage value directly proportional to the energy of the received symbol and that of the noise.



Let us assume that the input noise w(t) is a random Gaussian process and the receiveing filter is linear



the output noise

$$n(T) = n_0$$
 is a zero mean Gaussian random variable

y(T) is a random variable with a mean of either $a_1 \text{ or } a_2$ where

$$a_1 = y_o(T)$$
 when $s_1(t)$ is transmitt ed

$$a_2 = y_o(T)$$
 when $s_2(t)$ is transmitt ed





Conditional pdfs:

$$p(y \mid s_1) = \frac{1}{\sigma_0 \sqrt{2\pi}} \exp \left[-\frac{1}{2} \left(\frac{y - a_1}{\sigma_0} \right)^2 \right]$$

$$p(y \mid s_2) = \frac{1}{\sigma_0 \sqrt{2\pi}} \exp \left[-\frac{1}{2} \left(\frac{y - a_2}{\sigma_0} \right)^2 \right]$$



Notes: after a received waveform has been transformed to a sample, the actual shape of the waveform is no longer important; all waveform types that are transformed to the same value of y(T) are identical for detection purposes. Tha matched filter maps all signals of equal energy into the same point y(T)



The received signal energy (not its shape!) is the important parameter in the detection process.

This is why the detection analysis for baseband signals is the same as that for bandpass

Since y(T) is a voltage signal that is proportional to the energy of the received symbol the larger the magnitue of y(T) the more error free will be the decision process





Detection is done according to:

$$y(T) > \gamma$$
 H_1
 $y(T) < \gamma$ H_2

 H_1, H_2 are the two binary hypothesis

How to choose γ ?

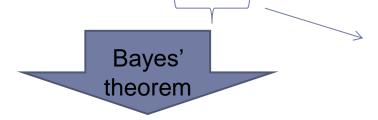


The likelihood ratio test and the Maximum A Posteriori Criterion

The MAP (Maximum A posteriori criterion) select the threshold that maximize the a posteriori probability, i.e.:

choose hypothesis H_1 when $p(s_1 | y) > p(s_2 | y)$

choose hypothesis H_2 when $p(s_1 | y) < p(s_2 | y)$



A posteriori probability

choose hypothesis H_1 when $p(y | s_1)P(s_1) > p(y | s_2)P(s_2)$

choose hypothesis H_2 when $p(y | s_1)P(s_1) < p(y | s_2)P(s_2)$



The likelihood ratio test and the Maximum A Posteriori Criterion

choose hypothesis
$$H_1$$
 when $\frac{p(y \mid s_1)}{p(y \mid s_2)} > \frac{P(s_2)}{P(s_1)}$ choose hypothesis H_2 when $\frac{p(y \mid s_1)}{p(y \mid s_2)} < \frac{P(s_2)}{P(s_1)}$ Likelihood ratio test

If errors are uniformly distributed, this MAP criterion corresponds to a Minimum error criterion (on the average, it yields the minimum number of incorrect decision)



Maximum likelihood Criterion

When the classes are equally likely (in case of two classes $\Rightarrow P(s_1) = P(s_2)$)
The MAP criterion is known as Maximum likelihood criterion



choose hypothesis
$$H_1$$
 when $\frac{p(y | s_1)}{p(y | s_2)} > 1$

choose hypothesis
$$H_2$$
 when $\frac{p(y | s_1)}{p(y | s_2)} < 1$



Maximum likelihood Criterion

In case of binary decision, the optimum threshold is:

$$\gamma_0 = \frac{a_1 + a_2}{2}$$

For antipodal signals $a_1 = -a_2$



$$\gamma_0 = 0$$



Error Probability

$$a_{1} = -a_{2} = \sqrt{E_{b}}$$

$$\gamma = 0$$

$$P_{e} = P(s_{1})P(y < 0 \mid s_{1}) + P(s_{2})P(y > 0 \mid s_{2}) =$$

$$= P(s_{1})P(\sqrt{E_{b}} + n < 0) + P(s_{2})P(-\sqrt{E_{b}} + n > 0)$$

$$P(s_{1}) = P(s_{2}) = 1/2$$

$$P(\sqrt{E_b} + n < 0) = \frac{1}{\sqrt{2\pi\sigma}} \int_{-\infty}^{0} e^{-\frac{1}{2} \left(\frac{v - \sqrt{E_b}}{\sigma}\right)^2} dv$$



variable change

$$P(\sqrt{E_b} + n < 0) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{-\sqrt{\frac{2E_b}{N_0}}} e^{-\frac{1}{2}x^2} dv = Q\left(\sqrt{\frac{2E_b}{N_0}}\right)$$

$$P(-\sqrt{E_b} + n > 0) = Q\left(\sqrt{\frac{2E_b}{N_0}}\right)$$

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

DETECTION OF BINARY SIGNALS IN GAUSSIAN NOISE

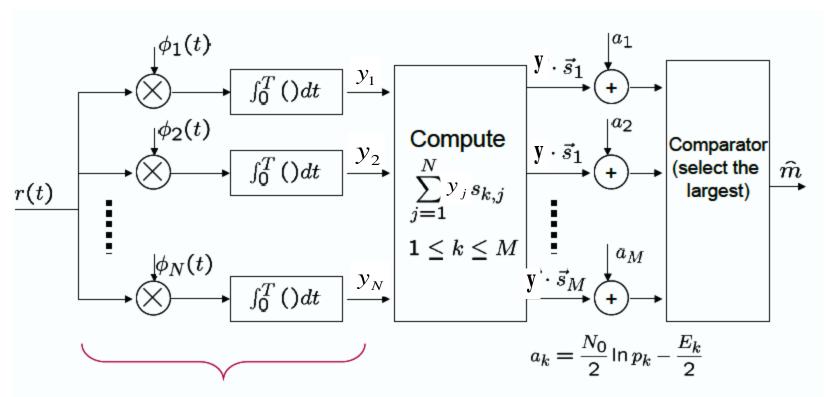
Error Probability

Prove the in case of binary ORTHOGONAL signals the probability of error is

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



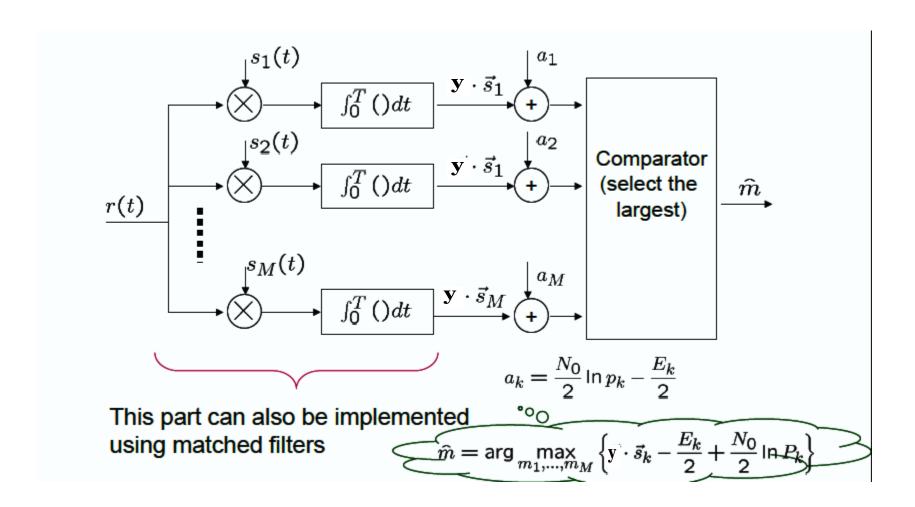
MAP receiver structure Method #1 (signal demodulator + detector)



This part can also be implemented using matched filters



MAP receiver structure Method #2 (integrated demodulator + detector)





MAP receiver structure Method #1 vs Method #2

- Both receivers perform identically
- Choice depends on circumstances
- For instance, if N < M and $\{\phi_j(t)\}$ are easier to generate than $\{s_k(t)\}$, then the choice is obvious



 Signal space can be divided into M disjoint decision regions R₁ R₂, ..., R_M.

If
$$\vec{y} \in R_k$$
 \Longrightarrow decide m_k was transmitted

Select decision regions so that P_e is minimized

- Recall that the optimal receiver sets $\hat{m} = m_k$ iff $\|\mathbf{y} \vec{s}_k\|^2 N_0 \ln P_k$ is minimized
- For simplicity, if one assumes $p_k = 1/M$, for all k, then the optimal receiver sets $\hat{m} = m_k$ iff

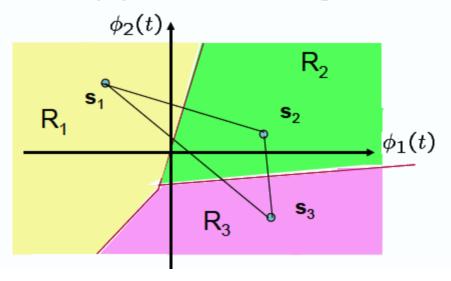
$$\|\mathbf{y} - \vec{s_k}\|^2$$
 is minimized



- Geometrically, this means
 - Take projection of r(t) in the signal space (i.e. ȳ). Then, decision is made in favor of signal that is the closest to y in the sense of minimum Euclidean distance
 - And those observation vectors \mathbf{y} with $\|\mathbf{y} \mathbf{s_k}\|^2 < \|\mathbf{y} \mathbf{s_i}\|^2$ for all $i \neq k$ should be assigned to decision region R_k



- In general, boundaries of decision regions are perpendicular bisectors of the lines joining the original transmitted signals
- Example: three equiprobable 2-dim signals





Decision Regions **Example: QPSK modulation**

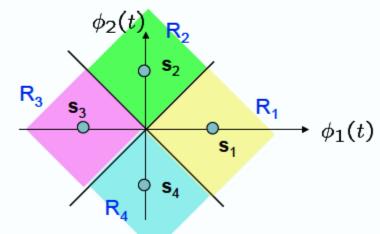
- Assume all signals are equally likely
- All 4 signals could be written as the linear combination of two basis functions
- Constellations of 4 signals

$$s_1 = (1,0)$$

$$s_2 = (0,1)$$

$$s_3 = (-1,0)$$

$$s_4 = (0,-1)$$





- Boundaries are perpendicular to a line drawn between two signal points
- If signals are equiprobable, decision boundaries lie exactly halfway in between signal points
- If signal probabilities are unequal, the region of the less probable signal will shrink



Three equally probable messages m1, m2, and m3 are to be transmitted over an AWGN channel with noise power-spectral density $N_{\rm 0}$ / 2. The messages are

$$s_{1}(t) = \begin{cases} 1 & 0 \le t \le T \\ 0 & otherwise \end{cases}$$

$$s_{2}(t) = -s_{3}(t) = \begin{cases} 1 & 0 \le t \le \frac{T}{2} \\ -1 & \frac{T}{2} \le t \le T \\ 0 & otherwise \end{cases}$$

- 1. What is the dimensionality of the signal space?
- Find an appropriate basis for the signal space (Hint: You can find the basis without using the Gram-Schmidt procedure).
- 3. Draw the signal constellation for this problem.
- 4. Sketch the optimal decision regions R1, R2, and R3.
- 5. Which of the three messages is more vulnerable to errors and why? In other words, which of p(Error | m_i transmitted), i = 1, 2, 3 is larger?



For binary equiprobable signaling over an AWGN channel, regardless of the signaling type, the error probability can be expressed as:

$$P_e = Q \left(\frac{d}{\sqrt{2N_0}} \right)$$

Where d is the Euclidean distance between the two signal points in the constellation

Does it exist a simple expression for the error probability of the general equiprobable M-ary signaling?

NO

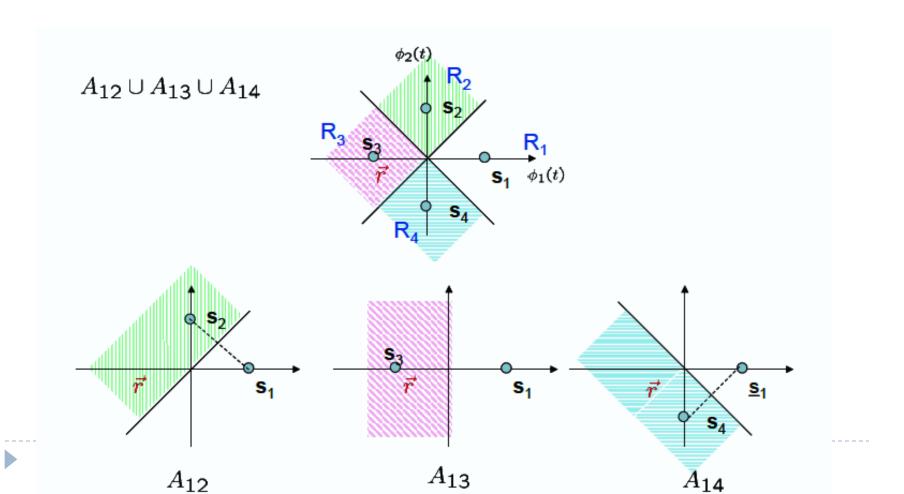
$$P_{M} = \frac{1}{M} \sum_{i=1}^{M} P(e \mid s_{m}) = \frac{1}{M} \sum_{i=1}^{M} \int_{R_{m}^{c}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left(1 - \int_{R_{m}} f(\mathbf{y} \mid s_{m}) d\mathbf{y} \right) = \frac{1}{M} \sum_{i=1}^{M} \left($$

However, there is a simple upper bound known as UNION BOUND



Let us assume that $s_m(t)$ is transmitted The proability of error is the probability that the receiver detects a signal other than $s_m(t)$

Let A_i the event that message i is detected at the receiver



$$P_m = P(error \mid s_m(t) \text{ sent}) = P\left(\bigcup_{\substack{i=1\\i\neq m}}^{M} A_i \mid s_m(t) \text{ sent}\right) \leq \sum_{\substack{i=1\\i\neq m}}^{M} P(A_i \mid s_m(t) \text{ sent})$$

A necessary BUT not sufficient condition for detecting $s_i(t)$ when $s_m(t)$ is sent (if the ML criterion is used):

$$D(\mathbf{y}, s_i) < D(\mathbf{y}, s_m)$$



$$P(A_i \mid s_m(t) \text{ sent}) \le P(D(\mathbf{y}, s_i) < D(\mathbf{y}, s_m))$$



$$P(D(\mathbf{y}, s_i) < D(\mathbf{y}, s_m)) = Q\left(\frac{d_{mi}}{\sqrt{2N_0}}\right)$$



$$P_m \le \sum_{\substack{i=1\\i\neq m}}^M P(A_i \mid s_m(t) \text{ sent}) \le \sum_{\substack{i=1\\i\neq m}}^M Q\left(\frac{d_{mi}}{\sqrt{2N_0}}\right)$$

If we define

$$d_{\min} = \min_{\substack{1 \le m, m \le M \\ m \ne m}} d_{mm}$$



$$Q\left(\frac{d_{\min}}{\sqrt{2N_0}}\right) \leq Q\left(\frac{d_{\min}}{\sqrt{2N_0}}\right)$$



$$P_{m} \leq \sum_{\substack{i=1\\i\neq m}}^{M} Q \left(\frac{d_{\min}}{\sqrt{2N_{0}}} \right) = (M-1)Q \left(\frac{d_{\min}}{\sqrt{2N_{0}}} \right)$$



$$P_{M} = \frac{1}{M} \sum_{\substack{i=1\\i \neq m}}^{M} P_{m} \le (M-1)Q \left(\frac{d_{\min}}{\sqrt{2N_{0}}}\right) \le \frac{M-1}{2} e^{-\frac{d_{\min}^{2}}{4N_{0}}}$$

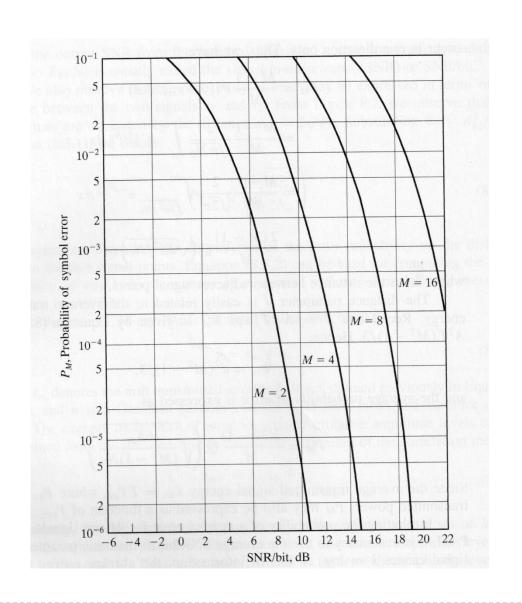
$$Q(x) \le 1/e^{-\frac{x^{2}}{2}}$$

Thi approximation is very useful for high SNR

A good signal set should provide the maximum possible d_{\min}



Probability of Error for M-ary PAM





Probability of Error for M-ary orthogonal signaling

