

ECE 5654 Lecture 8

Union Bound and Probability of Error

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Learning Objectives

At the end of this lecture, the student should be able to:

- Describe the functional blocks for implementing the optimal receiver in an AWGN channel
- Union bound for probability of error in AWGN.
- Probability of error for various modulation schemes.
- Power vs Bandwidth tradeoffs.

MAP Detector for AWGN

- MAP for AWGN:

$$\begin{aligned} g_{MAP}(r) &= \underset{1 \leq m \leq M}{\operatorname{argmax}} \quad \frac{N_0}{2} \log(p_m) - \frac{\|r - s_m\|^2}{2} \\ &= \underset{1 \leq m \leq M}{\operatorname{argmax}} \quad \frac{N_0}{2} \log(p_m) - \left(\frac{\|r\|^2}{2} + \frac{\|s_m\|^2}{2} - \frac{2r \cdot s_m}{2} \right) \end{aligned}$$

- $\|r\|^2$ does not depend on m so it can be ignored

$$\begin{aligned} g_{MAP}(r) &= \underset{1 \leq m \leq M}{\operatorname{argmax}} \quad \frac{N_0}{2} \log(p_m) - \frac{\|s_m\|^2}{2} + r \cdot s_m \\ &= \underset{1 \leq m \leq M}{\operatorname{argmax}} \quad \underbrace{\frac{N_0}{2} \log(p_m) - \frac{E_m}{2}}_{\eta_m: \text{ signal dependent bias}} + r \cdot s_m \\ &= \underset{1 \leq m \leq M}{\operatorname{argmax}} \quad \eta_m + r \cdot s_m \end{aligned}$$

Decision regions for MAP

- MAP receiver:

$$g_{MAP}(r) = \underset{1 \leq m \leq M}{\operatorname{argmax}} \quad \eta_m + r \cdot s_m$$

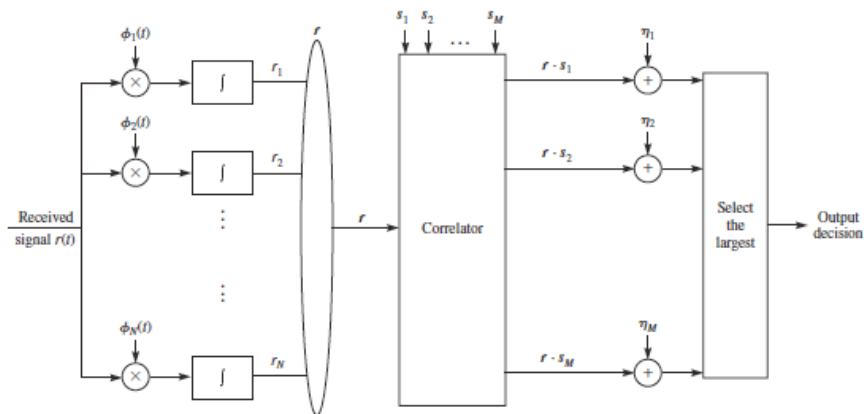
- Signal dependent bias: $\eta_m = \frac{N_0}{2} \log(p_m) - \frac{E_m}{2}$
- Output dependent term: $r \cdot s_m = \int_{-\infty}^{\infty} r(t) s_m(t) dt$
- Decision region D_m :

$$D_m = \{r : \eta_m + r \cdot s_m > \eta_{m'} + r \cdot s_{m'}, \forall m' \neq m\}$$

The Correlation Receiver

- For each m , the receiver needs to obtain $\eta_m + r \cdot s_m$
- Receiver observes $r(t)$
- First step is to obtain the vector r from the received signal $r(t)$
- $r_j = \int_{-\infty}^{\infty} r(t)\phi_j(t) dt, j = 1, 2, \dots, N$
- $r = (r_1, r_2, \dots, r_N)$
- Using r and s_m , computes the inner product: $r \cdot s_m$
- Adds the bias term η_m to get $\eta_m + r \cdot s_m$
- Compare the results and choose m which maximizes this result.

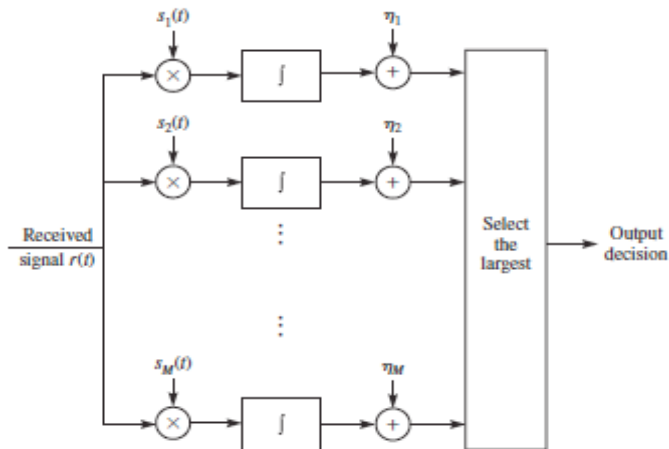
Correlation Receiver Implementation 1



N correlators

N : signal space dimensionality

Correlation Receiver Implementation 2



M correlators M : number of signals

Typically $N \ll M$ Implementation 1 is preferable over 2

The Matched Filter Receiver

- In both correlation receiver implementations, the receiver computes

$$r_x = \int_{-\infty}^{\infty} r(t)x(t) dt$$

where $x(t)$ could either be

- Implementation 1: $\phi_1(t), \phi_2(t), \dots, \phi_N(t)$
- Implementation 2: $s_1(t), s_2(t), \dots, s_M(t)$
- Matched filter: if we define $h(t) = x(T - t)$ and consider a filter with impulse response $h(t)$, this filter is called a **filter matched to $x(t)$** , or a matched filter.

Matched Filter

- If input $r(t)$ is applied to the matched filter $h(t) = x(T - t)$, then its output is

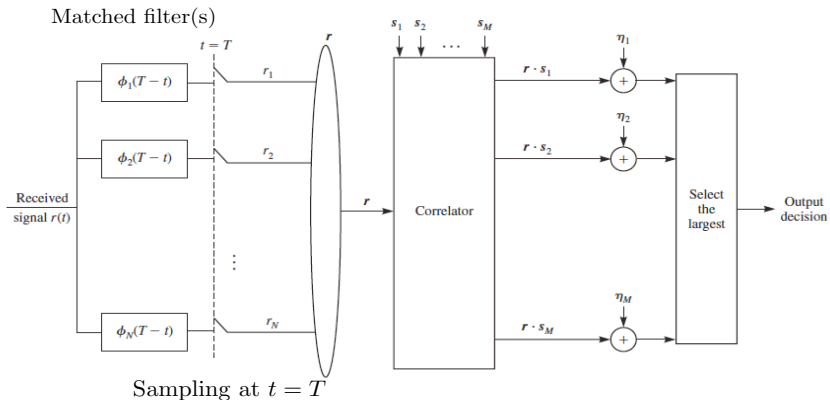
$$\begin{aligned} y(t) &= r(t) * h(t) = \int_{-\infty}^{\infty} r(\tau) h(t - \tau) d\tau \\ &= \int_{-\infty}^{\infty} r(\tau) x(T - t + \tau) d\tau \end{aligned}$$

- If we sample the output of the matched filter at $t = T$, we obtain

$$r_x = y(T) = \int_{-\infty}^{\infty} r(\tau) x(\tau) d\tau$$

- Hence, the correlator r_x can be obtained by sampling the output of the matched filter at time $t = T$.
- T could be arbitrary but the sampling must be done exactly at $t = T$, where T is the value used in the design of the matched filter.

Matched Filter Receiver



- From a practical perspective, T must be chosen such that
- the resulting filters are causal, i.e., $h(t) = 0$ for $t < 0$.
 - this puts a practical limit on possible values of T

Frequency Domain Interpretation of Matched Filter

- MF of a signal $s(t)$ is $h(t) = s(T - t)$
- Fourier transform of $h(t) = H(f) = S^*(f)e^{-j2\pi fT}$
- Magnitude remains the same, $|S(f)| = |H(f)|$
- Phase of $H(f) = -\text{Phase of } S(f) - 2\pi fT$
- Represents the sampling delay of T

Matched Filter maximizes SNR

- We will now show that MF has an interesting property that it maximizes the signal-to-noise ratio (SNR) at the receiver.
- Suppose that $r(t) = s(t) + n(t)$ (the received signal) is passed through a filter with impulse response $h(t)$
- The output is

$$\begin{aligned}y(t) &= r(t) * h(t) = [s(t) + n(t)] * h(t) \\&= s(t) * h(t) + n(t) * h(t) \\&= y_s(t) + \nu(t)\end{aligned}$$

- Output has two parts:
 - Signal part: $y_s(t)$, with Fourier transform $H(f)S(f)$
 - Noise part: $\nu(t)$, with PSD $\frac{N_0|H(f)|^2}{2}$
- We sample the output at time $t = T$

MF maximizes SNR

- Sampling these components at $t = T$ results in: (book has a Typo)

$$y_s(T) = \int_{-\infty}^{\infty} H(f)S(f)e^{j2\pi fT} df$$

and a zero-mean Gaussian noise component $\nu(T)$, which has the variance:

$$\text{Var}[\nu(T)] = \frac{N_0}{2} \int_{-\infty}^{\infty} |H(f)|^2 df = \frac{N_0}{2} \cdot E_h$$

- E_h is the energy in $h(t)$.
- Let us define the SNR at the filter output as:

$$\text{SNR} = \frac{y_s^2(T)}{\text{Var}[\nu(T)]}$$

MF maximizes SNR

- Using Cauchy-Schwartz inequality:

$$\begin{aligned} y_s^2(T) &= \left(\int_{-\infty}^{\infty} H(f) S(f) e^{j2\pi fT} df \right)^2 \\ &\leq \int_{-\infty}^{\infty} |H(f)|^2 df \cdot \int_{-\infty}^{\infty} |S(f) e^{j2\pi fT}|^2 df \\ &= E_h \cdot E_s \end{aligned}$$

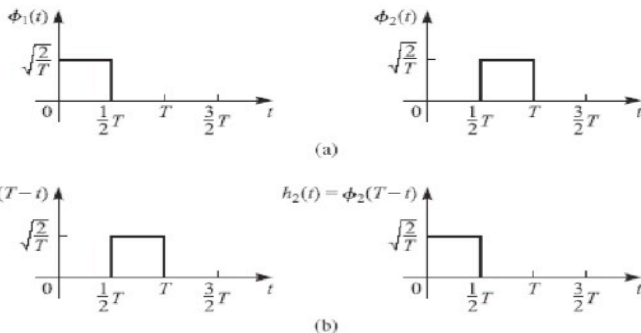
which holds with equality if and only if $H(f) = \alpha S^*(f) e^{-j2\pi fT}$ for some complex constant α .

- So we can upper bound the SNR as

$$SNR = \frac{y_s^2(T)}{\text{Var}[\nu(T)]} \leq \frac{E_h E_s}{\frac{N_0}{2} E_h} = \frac{2E_s}{N_0}$$

- This shows that MF achieves this upper bound and maximizes SNR.

Example 1 Biorthogonal signaling

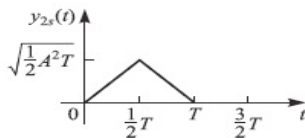
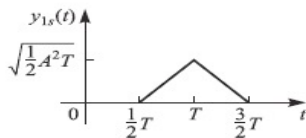


- Biorthogonal signaling
- $M = 4$ signals
- $s_1(t) = A\sqrt{T/2}\phi_1(t) = [A\sqrt{T/2} \ 0]$; $s_3(t) = -s_1(t)$
- $s_2(t) = A\sqrt{T/2}\phi_2(t) = [0 \ A\sqrt{T/2}]$; $s_4(t) = -s_2(t)$

Example 1

- We have two matched filters, one for each basis function.
- Suppose $s_1(t)$ is transmitted, then we can find the (noise-free) responses of the two matched filters.

$$y_{1s}(t) = \int_{-\infty}^{\infty} h_1(\tau) s_1(t - \tau) d\tau, \quad y_{2s}(t) = \int_{-\infty}^{\infty} h_2(\tau) s_1(t - \tau) d\tau$$



- We sample these outputs at $t = T$;
 $y_{1s}(T) = \sqrt{A^2 T/2}$ and $y_{2s}(T) = 0$
- Hence, output SNR $= (A^2 T/2)/(N_0/2) = \frac{A^2 T}{N_0} = 2E_1/N_0$

Probability of Error (for ML Receiver)

- Recall that for equiprobable symbols $p_m = 1/M$,
- MAP reduces to ML and hence ML receiver is the optimal receiver.
- Error Probability for ML detection becomes:

$$\begin{aligned} P_e &= \frac{1}{M} \sum_{m=1}^1 P_{e|m} \\ &= \frac{1}{M} \sum_{m=1}^1 \sum_{1 \leq m' \leq M, m' \neq m} \int_{D_{m'}} p(r|s_m) dr \end{aligned}$$

- For very specific constellations, the decision regions D_m 's are regular enough that the integrals can be computed in closed form.
- Hence, it is useful to have upper bounds for the error probability.

Union Bound

- The union bound is the simplest and widely used bound which gives a good approximation (particularly) at high SNR.
- We will start with the union bound for a general communication channel (described by a probabilistic transformation) and then return to the AWGN channel as a special case.
- Let us look at the term $P_{e|m}$ and specifically the integral

$$\int_{D_{m'}} p(r|s_m) dr$$

- Let's take a closer look at the decision region $D_{m'}$

Union Bound

- The decision region $D_{m'}$ is defined (for ML) as

$$D_{m'} = \{r : p(r|s_{m'}) > p(r|s_k), 1 \leq k \leq M, k \neq m'\}$$

- Let's define a new region $D_{mm'}$ as

$$D_{mm'} = \{r : p(r|s_{m'}) > p(r|s_m)\}$$

- $D_{mm'}$ is the decision region for the symbol m' in a binary equiprobable system with two signals, m and m'
- Comparing $D_{m'}$ and $D_{mm'}$, it is clear that $D_{m'}$ is contained in the region $D_{mm'}$, i.e., $D_{m'} \subseteq D_{mm'}$. This implies that

$$\int_{D_{m'}} p(r|s_m) dr \leq \int_{D_{mm'}} p(r|s_m) dr$$

Union Bound Continued..

- So we have

$$\int_{D_{m'}} p(r|s_m) dr \leq \int_{D_{mm'}} p(r|s_m) dr$$

- The r.h.s. above is the error probability of a binary equiprobable system with signals s_m and $s_{m'}$, when the signal s_m is sent.
- If we denote this error probability as

$$P_{m \rightarrow m'} = \int_{D_{mm'}} p(r|s_m) dr$$

we can then upper bound the probability $P_{e|m}$ as

$$P_{e|m} = \sum_{m' \neq m} \int_{D_{m'}} p(r|s_m) \leq \sum_{m' \neq m} P_{m \rightarrow m'}$$

- Putting everything together, we have the union bound for a general communication channel

$$\begin{aligned} P_e &= \frac{1}{M} \sum_{m=1}^M P_{e|m} \\ &\leq \frac{1}{M} \sum_{m=1}^M \sum_{m' \neq m} P_{m \rightarrow m'} \end{aligned}$$

- In summary, union bound upper bounds the probability by the pairwise error probabilities, which are often easier to obtain.
- In previous lecture, for the AWGN channel, we obtained the error probability for a binary equiprobable signaling scheme was shown to be

$$P_{m \rightarrow m'} = P_b = Q \left(\sqrt{\frac{d_{mm'}^2}{2N_0}} \right)$$

Union Bound for AWGN

- For the AWGN channel, the union bound gives

$$P_e \leq \frac{1}{M} \sum_{m=1}^M \sum_{m' \neq m} Q \left(\sqrt{\frac{d_{mm'}^2}{2N_0}} \right)$$

- $d_{mm'}$ is the distance between constellation points s_m and $s_{m'}$
- If the distance structure of the constellation has symmetries, this bound can be further simplified.
- The $Q(\cdot)$ function can be upper bounded as $Q(x) \leq \frac{1}{2}e^{-x^2/2}$.
- Using this, we can further bound

$$P_e \leq \frac{1}{2M} \sum_{m=1}^M \sum_{m' \neq m} e^{-d_{mm'}^2/4N_0}$$

Impact of Minimum Distance

- Let us define d_{min} , the minimum distance of a constellation as

$$d_{min} = \min_{m, m', m \neq m'} \|s_m - s_{m'}\|$$

- Then, as the $Q(\cdot)$ function is decreasing,

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

- Using this, the upper bound can be further relaxed to

$$\begin{aligned} P_e &\leq \frac{1}{M} \sum_{m=1}^M \sum_{m' \neq m} Q\left(\sqrt{\frac{d_{mm'}^2}{2N_0}}\right) \leq \frac{1}{M} \sum_{m=1}^M \sum_{m' \neq m} Q\left(\sqrt{\frac{d_{min}^2}{2N_0}}\right) \\ &= (M-1)Q\left(\sqrt{\frac{d_{min}^2}{2N_0}}\right) \end{aligned}$$

Lower bound on Error Probability

- Recall that

$$\begin{aligned} P_e &= \frac{1}{M} \sum_{m=1}^M P_{e|m} \\ &= \frac{1}{M} \sum_{m=1}^M \int_{D_m^c} p(r|s_m) dr \end{aligned}$$

where D_m^c the complementary decision region of message m .

- $D_m^c = \cup_{m' \neq m} D_{m'}$
- Hence $D_{m'}$ is contained in D_m^c for all $m' \neq m$
- This implies that for all $m' \neq m$

$$\int_{D_m^c} p(r|s_m) dr \geq \int_{D_{m'}} p(r|s_m) dr$$

Lower bound on Error Probability

- Recall that $P_e = \frac{1}{M} \sum_{m=1}^M \int_{D_m^c} p(r|s_m) dr$
- For all $m' \neq m$

$$\int_{D_m^c} p(r|s_m) dr \geq \int_{D_{m'}} p(r|s_m) dr$$

- The r.h.s can be interpreted as the probability of error of a binary signaling scheme with messages s_m and $s_{m'}$ when s_m is sent.
- hence, we can now obtain a lower bound on the error probability as

$$\begin{aligned} P_e &\geq \frac{1}{M} \sum_{m=1}^M \int_{D_{m'}} p(r|s_m) dr \\ &= \frac{1}{M} \sum_{m=1}^M Q \left(\sqrt{\frac{d_{mm'}^2}{2N_0}} \right) \end{aligned}$$

- Note that this bound holds for all $m' \neq m$ and can be optimized.

Lower bound on Error Probability

- The optimized lower bound on the error probability is then given as

$$P_e \geq \frac{1}{M} \sum_{m=1}^M \max_{m' \neq m} Q \left(\sqrt{\frac{d_{mm'}^2}{2N_0}} \right)$$

- As we saw before $Q(\cdot)$ is a decreasing function, choosing m' that gives the maximum value is equivalent to finding such m' with minimum distance from m
- If we denote d_{min}^m as the distance from m to its nearest neighbor, then

$$P_e \geq \frac{1}{M} \sum_{m=1}^M Q \left(\frac{d_{min}^m}{\sqrt{2N_0}} \right)$$

Lower bound on Error Probability

- Now, note that

$$Q\left(\frac{d_{min}^m}{\sqrt{2N_0}}\right) \geq \begin{cases} Q\left(\frac{d_{min}}{\sqrt{2N_0}}\right) & \text{if there exists at least one} \\ & \text{signal at distance } d_{min} \text{ from } m \\ 0 & \text{otherwise} \end{cases}$$

- Hence, if we denote N_{min} as the number of message points which are at a distance d_{min} from at least one other point in the constellation, then we have

$$P_e \geq \frac{N_{min}}{M} Q\left(\frac{d_{min}}{\sqrt{2N_0}}\right)$$

- In summary, we have

$$\frac{N_{min}}{M} Q\left(\frac{d_{min}}{\sqrt{2N_0}}\right) \leq P_e \leq (M-1) Q\left(\frac{d_{min}}{\sqrt{2N_0}}\right)$$

Conclusions

- Next lecture, we will look at optimal detection and error probability for band-limited signaling schemes (such as ASK, PSK and QAM).
- We will also do the same for Power-limited signaling schemes (such as orthogonal signaling and FSK).
- Bandwidth vs Power tradeoffs.