

LECTURE 5: ERROR PROBABILITY COMPUTATION

INTRODUCTION TO ERROR PROBABILITY COMPUTATION

To measure the goodness of a digital radio link: error probability

$$\textbf{SYMBOL ERROR RATE} = \text{SER} = P_s(e) =$$
$$P_s(e) = P(\underline{s}_R[n] \neq \underline{s}_T[n])$$

$$\textbf{BIT ERROR RATE} = \text{BER} = P_b(e) =$$
$$P(u_{\underline{R}}[i] \neq u_{\underline{T}}[i])$$

INTRODUCTION TO ERROR PROBABILITY COMPUTATION

$$R_b$$

Bit Rate

$$T_b = 1/R_b$$

Bit time

$$T = k T_b$$

Symbol time

$$R = 1/T$$

Symbol Rate

INTRODUCTION TO ERROR PROBABILITY COMPUTATION

$$E_b$$

Energy per information bit

$$E_s$$

Energy per transmitted signal

$$S = E_b R_b = E_s R$$

Signal power

INTRODUCTION TO ERROR PROBABILITY COMPUTATION

N_0

Noise spectral density

B

Signal Bandwidth

$N = N_0 B$

Noise power

INTRODUCTION TO ERROR PROBABILITY COMPUTATION

$$S/N$$

Signal to Noise ratio

$$E_b/N_0$$

Signal to Noise ratio referred to an information bit

Connection:

$$\frac{S}{N} = \frac{E_b}{N_0} \frac{R_b}{B} = \frac{E_b}{N_0} \eta$$

where

$$\eta = \frac{R_b}{B}$$

spectral efficiency

INTRODUCTION TO ERROR PROBABILITY COMPUTATION

The system performance are expressed as a function of E_b/N_0

This ratio is proportional to the received power

$$S = \frac{S}{N} N = \frac{E_b}{N_0} \frac{R_b}{B} N_0 B = \frac{E_b}{N_0} R_b N_0$$

SER COMPUTATION

Definition: $P_S(e) = P(\underline{s}_R \neq \underline{s}_T)$

We have

$$P_S(e) = \sum_{i=1}^m P_S(e \mid \underline{s}_T = \underline{s}_i) P(\underline{s}_T = \underline{s}_i) = \frac{1}{m} \sum_{i=1}^m P_S(e \mid \underline{s}_T = \underline{s}_i)$$

We have to compute

$$P_S(e \mid \underline{s}_T = \underline{s}_i) = P(\underline{s}_R \neq \underline{s}_T \mid \underline{s}_T = \underline{s}_i)$$

SER COMPUTATION

First formulation:

$$\begin{aligned} P_S(e \mid \underline{s_T} = \underline{s_i}) &= P(\underline{s_R} \neq \underline{s_T} \mid \underline{s_T} = \underline{s_i}) = 1 - P(\underline{s_R} = \underline{s_T} \mid \underline{s_T} = \underline{s_i}) = \\ &= 1 - P(\underline{\rho} \in V(\underline{s_i}) \mid \underline{s_T} = \underline{s_i}) \end{aligned}$$

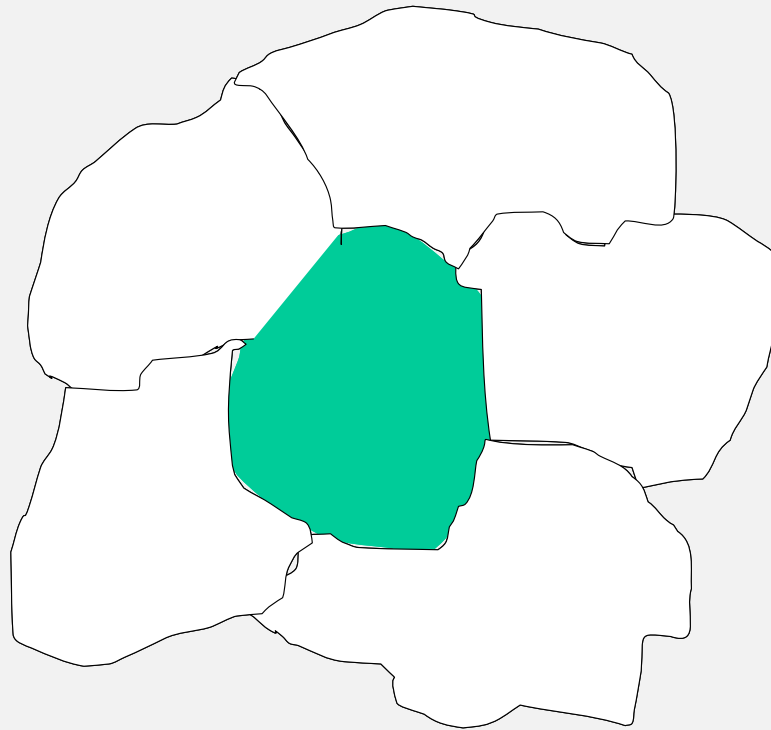
Second formulation:

$$\begin{aligned} P_S(e \mid \underline{s_T} = \underline{s_i}) &= P(\underline{s_R} \neq \underline{s_T} \mid \underline{s_T} = \underline{s_i}) = P(\underline{\rho} \notin V(\underline{s_i}) \mid \underline{s_T} = \underline{s_i}) = \\ &= \sum_{j \neq i} P(\underline{s_R} = \underline{s_i} \mid \underline{s_T} = \underline{s_i}) = \sum_{j \neq i} P(\underline{\rho} \in V(\underline{s_j}) \mid \underline{s_T} = \underline{s_i}) \end{aligned}$$

SER COMPUTATION

First formulation

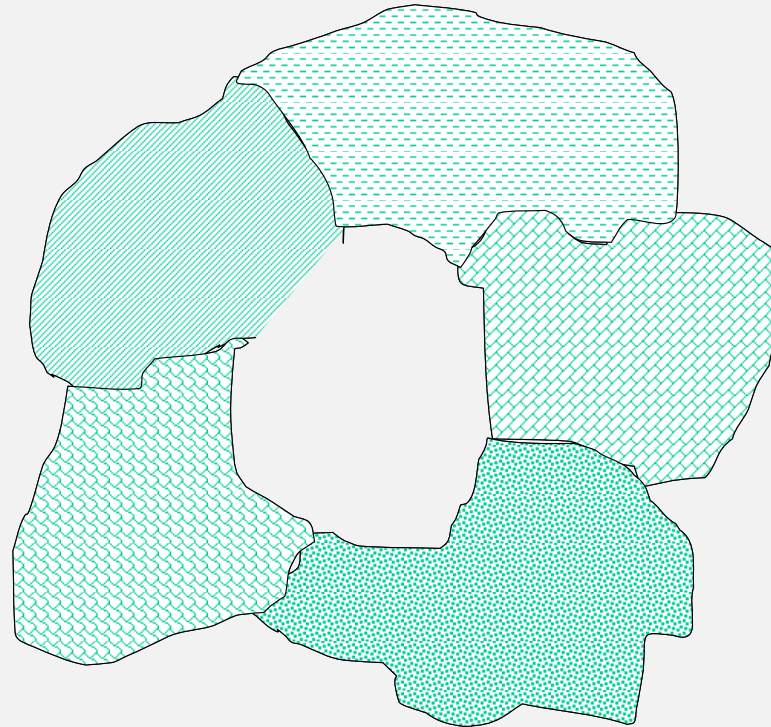
$$P_s(e \mid \underline{s}_T = \underline{s}_i) = 1 - P(\underline{\rho} \in V(\underline{s}_i) \mid \underline{s}_T = \underline{s}_i)$$



SER COMPUTATION

Second formulation

$$P_S(e \mid \underline{s}_T = \underline{s}_i) = P(\underline{\rho} \notin V(\underline{s}_i) \mid \underline{s}_T = \underline{s}_i) = \sum_{j \neq i} P(\underline{\rho} \in V(\underline{s}_j) \mid \underline{s}_T = \underline{s}_i)$$



BER COMPUTATION

When the received signal is correct ($\underline{s}_R = \underline{s}_T$), also the information vector is correct ($\underline{v}_R = \underline{v}_T$).

When the received signal is wrong ($\underline{s}_R \neq \underline{s}_T$), the binary information vector is certainly wrong ($\underline{v}_R \neq \underline{v}_T$), but the number of information bits depends on the labeling and is given by

$$\frac{d_H(\underline{v}_R, \underline{v}_T)}{k}$$

BER COMPUTATION

We have

$$P_b(e) = \frac{1}{m} \sum_{i=1}^m P_b(e | \underline{s}_T = \underline{s}_i)$$

where

$$P_b(e | \underline{s}_T = \underline{s}_i) = \sum_{j \neq i} P_b(e, \underline{s}_R = \underline{s}_j | \underline{s}_T = \underline{s}_i) =$$

$$= \sum_{j \neq i} \frac{d_H(\underline{v}_j, \underline{v}_i)}{k} P(\underline{s}_R = \underline{s}_j | \underline{s}_T = \underline{s}_i) =$$

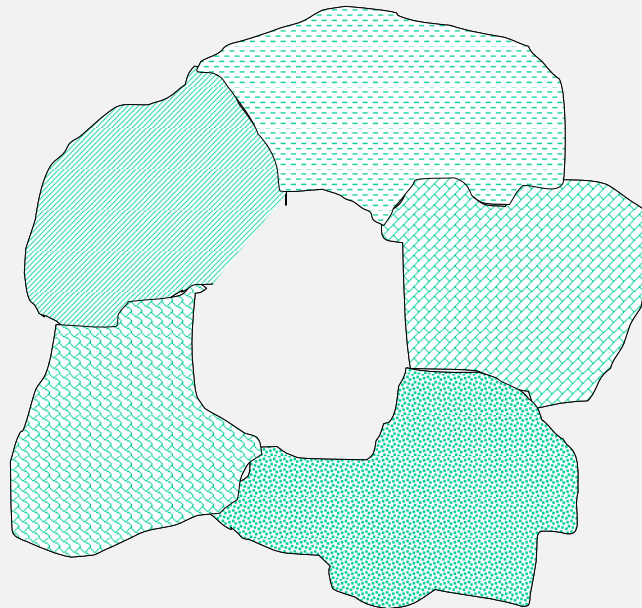
$$= \sum_{j \neq i} \frac{d_H(\underline{v}_j, \underline{v}_i)}{k} P(\underline{\rho} \in V(\underline{s}_j) | \underline{s}_T = \underline{s}_i)$$

$$\left[\text{where } \underline{v}_i = e^{-1}(\underline{s}_i) \text{ and } \underline{v}_j = e^{-1}(\underline{s}_j) \right]$$

BER COMPUTATION

$$P_b(e) = \frac{1}{m} \sum_{i=1}^m P_b(e | \underline{s}_T = \underline{s}_i)$$

$$P_b(e | \underline{s}_T = \underline{s}_i) = \sum_{j \neq i} \frac{d_H(\underline{v}_j, \underline{v}_i)}{k} P(\underline{\rho} \in V(\underline{s}_j) | \underline{s}_T = \underline{s}_i)$$



ERFC

Given a Gaussian random variable n with

- mean μ

- variance σ^2

- density

$$f_n(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

We have

$$P(n > x) = \int_x^{+\infty} f_n(x) dx = \frac{1}{2} \operatorname{erfc}\left(\frac{x-\mu}{\sqrt{2}\sigma}\right)$$

ERFC

where

$$\operatorname{erfc}(x) = \frac{2}{\sqrt{\pi}} \int_x^{+\infty} e^{-t^2} dt$$

In fact

$$\begin{aligned} P(n > x) &= \int_x^{+\infty} f_n(x) dx = \int_x^{+\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) dx = \\ &= \frac{1}{\sqrt{\pi}} \int_{\frac{(x-\mu)}{\sqrt{2}\sigma}}^{+\infty} e^{-t^2} dt = \frac{1}{2} \operatorname{erfc}\left(\frac{x-\mu}{\sqrt{2}\sigma}\right) \end{aligned}$$

In case of zero mean and variance $N_0/2$

$$P(n > x) = \frac{1}{2} \operatorname{erfc}\left(\frac{x-\mu}{\sqrt{2}\sigma}\right) = \frac{1}{2} \operatorname{erfc}\left(\frac{x}{\sqrt{N_0}}\right)$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

Let us consider a one-dimensional constellation ($d=1$) composed by two signals ($m=2$), symmetrical with respect to the origin:

$$M = \{ \underline{s_1} = (+A) \quad \underline{s_2} = (-A) \}$$

The Voronoi regions are:

$$V(\underline{s_1}) = \{ \underline{\rho} = (\rho_1) \text{ , } \rho_1 \geq 0 \}$$

$$V(\underline{s_2}) = \{ \underline{\rho} = (\rho_1) \text{ , } \rho_1 \leq 0 \}$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

We have:

$$P_S(e) = \frac{1}{m} \sum_{i=1}^m P_S(e | \underline{s}_T = \underline{s}_i) = \frac{1}{2} \left[P_S(e | \underline{s}_T = \underline{s}_1) + P_S(e | \underline{s}_T = \underline{s}_2) \right]$$

Let us compute

$$P_S(e | \underline{s}_T = \underline{s}_1)$$

and

$$P_S(e | \underline{s}_T = \underline{s}_2)$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

$$P_s(e | \underline{s}_T = \underline{s}_1) = P(\underline{\rho} \in V(\underline{s}_2) | \underline{s}_T = \underline{s}_1) = P(\rho_1 < 0 | \underline{s}_T = \underline{s}_1)$$

We have:

$$\underline{r} = \underline{s}_T + \underline{n} \quad \underline{r} = \underline{\rho} \quad \underline{s}_T = \underline{s}_1$$

where $\underline{\rho} = (\rho_1) \quad \underline{s}_1 = (s_{11}) = (+A) \quad \underline{n} = (n_1)$

It follows:

$$\rho_1 = A + n_1$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

$$P_S(e \mid \underline{s}_T = \underline{s}_1) = P(\rho_1 < 0 \mid \underline{s}_T = \underline{s}_1) = P(A + n_1 < 0) = P(n_1 < -A)$$

The random variable n_1 is Gaussian, with mean zero and variance $N_0/2$

$$P_S(e \mid \underline{s}_T = \underline{s}_1) = P(n_1 < -A) = P(n_1 > A) = \frac{1}{2} \operatorname{erfc} \left(\frac{A}{\sqrt{N_0/2}} \right)$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

Compute now the error probability for $\underline{s}_T = \underline{s}_2$

$$P_S(e | \underline{s}_T = \underline{s}_2) = P(\underline{\rho} \in V(\underline{s}_1) | \underline{s}_T = \underline{s}_2) = P(\rho_1 > 0 | \underline{s}_T = \underline{s}_2)$$

We have:

$$\underline{r} = \underline{s}_T + \underline{n} \quad \underline{r} = \underline{\rho} \quad \underline{s}_T = \underline{s}_2$$

then

$$\underline{\rho} = (\rho_1) \quad \underline{s}_2 = (s_{21}) = (-A) \quad \underline{n} = (n_1)$$

$$\rho_1 = -A + n_1$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

$$P_s(e \mid \underline{s}_T = \underline{s}_2) = P(-A + n_1 > 0) = P(n_1 > A)$$

$$P_s(e \mid \underline{s}_T = \underline{s}_2) = \frac{1}{2} \operatorname{erfc} \left(\frac{A}{\sqrt{N_0}} \right)$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

We have

$$P_S(e | \underline{s_T} = \underline{s_1}) = P_S(e | \underline{s_T} = \underline{s_2})$$

It follows:

$$P_S(e) = \frac{1}{2} \left[P_S(e | \underline{s_T} = \underline{s_1}) + P_S(e | \underline{s_T} = \underline{s_2}) \right] = P_S(e | \underline{s_T} = \underline{s_1})$$

then

$$P_S(e) = P_S(e | \underline{s_T} = \underline{s_1}) = \frac{1}{2} \operatorname{erfc} \left(\frac{A}{\sqrt{N_0}} \right)$$

[note that

$$P_S(e) = P_S(e | \underline{s_T} = \underline{s_1}) = \frac{1}{2} \operatorname{erfc} \left(\frac{d}{2\sqrt{N_0}} \right)$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

We have:

$$P_s(e) = P_s(e | \underline{s}_T = \underline{s}_1) = \frac{1}{2} \operatorname{erfc} \left(\frac{A}{\sqrt{N_0}} \right)$$

We want to write it as a function of E_b/N_0 .

$$E(\underline{s}_1) = E(\underline{s}_2) = A^2$$

$$E_s = \frac{E(\underline{s}_1) + E(\underline{s}_2)}{2} = A^2$$

$$E_b = \frac{E_s}{k} = E_s = A^2$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

Fundamental result

$$P_s(e) = \frac{1}{2} \operatorname{erfc} \left(\sqrt{\frac{E_b}{N_0}} \right)$$

SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

For this constellation we can establish this binary labeling:

$$e : H_1 \Leftrightarrow M$$

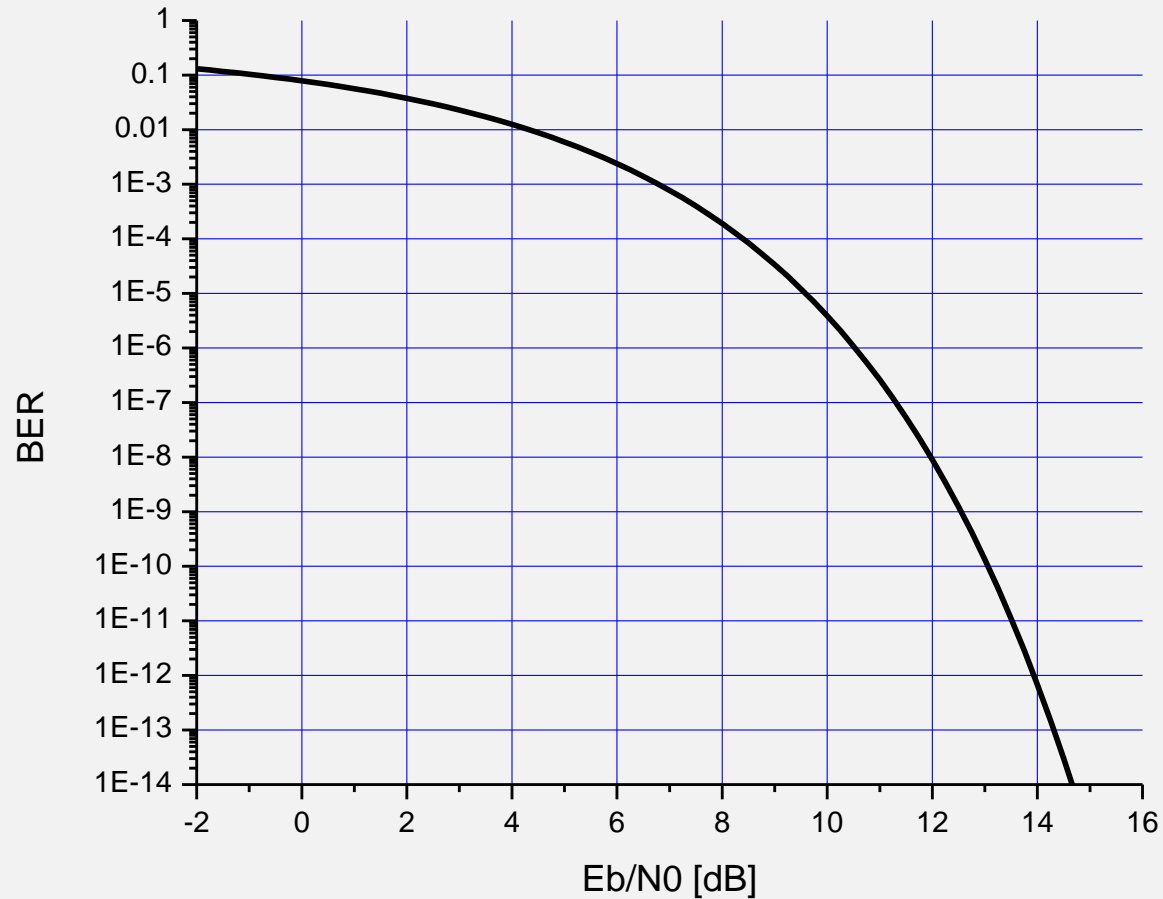
$$\underline{v_1} = (0) \Leftrightarrow \underline{s_1}$$

$$\underline{v_2} = (1) \Leftrightarrow \underline{s_2}$$

In this case, if the received signal is wrong, the information bit is wrong, too.

$$P_b(e) = P_s(e) = \frac{1}{2} \operatorname{erfc} \left(\sqrt{\frac{E_b}{N_0}} \right)$$

SER/BER computation for binary antipodal signals



SER/BER COMPUTATION FOR BINARY ANTIPODAL SIGNALS

Different signal constellations (different transmitted waveforms) with the same vector constellations achieve equal BER performance!

As an example, the BER performance of a binary antipodal constellation does not depend on the versor

$$b_1(t) = \frac{1}{\sqrt{T}} P_T(t)$$

$$b_1(t) = \sqrt{\frac{2}{T}} P_T(t) \cos(2\pi f_0 t)$$