

Digital Communication Systems Engineering with Software-Defined Radio

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Lecture 10

Mathematical Framework

- ▶ Suppose that a signal $s_i(t)$, $i = 1, 2$, was transmitted across an AWGN channel with noise signal $n(t)$
- ▶ A receiver intercepts $r(t)$ and needs to determine whether $s_1(t)$ or $s_2(t)$ was sent
- ▶ This situation can be evaluated using *Hypothesis Theory*:

$$\mathcal{H}_1 : r(t) = s_1(t) + n(t), \quad 0 \leq t \leq T$$

$$\mathcal{H}_0 : r(t) = s_2(t) + n(t), \quad 0 \leq t \leq T$$

where \mathcal{H}_0 and \mathcal{H}_1 are “Hypothesis 0” and “Hypothesis 1”

Decision Rule

- ▶ Suppose that $s_1(t)$ was transmitted
- ▶ In general, we can determine the level of correlation between two signals $x(t)$ and $y(t)$ over the time interval $0 \leq t \leq T$ using the expression:

$$\int_0^T x(t)y(t)dt$$

- ▶ Consequently, our decision rule on whether $s_1(t)$ or $s_2(t)$ was transmitted given that we observe $r(t)$ is defined as:

$$\int_0^T r(t)s_1(t)dt \geq \int_0^T r(t)s_2(t)dt \quad (1)$$

where we assume that $s_1(t)$ was transmitted

- ▶ The received signal $r(t)$ is more correlated to $s_1(t)$ than $s_2(t)$

Error Event

- Assuming $s_1(t)$ was transmitted, an *error event* occurs when:

$$\int_0^T r(t)s_1(t)dt \leq \int_0^T r(t)s_2(t)dt \quad (2)$$

- Since $r(t) = s_1(t) + n(t)$, we can substitute this into the error event in order to obtain:

$$\int_0^T s_1^2(t)dt + \int_0^T n(t)s_1(t)dt \leq \int_0^T s_1(t)s_2(t)dt + \int_0^T n(t)s_2(t)dt$$

$$E_{s_1} - \rho_{12} \leq \int_0^T n(t)(s_2(t) - s_1(t))dt$$

$$E_{s_1} - \rho_{12} \leq z$$

Characterizing z

- ▶ Since $n(t)$ is Gaussian $\rightarrow z$ is Gaussian
 - ▶ Integration is equivalent to a summation across an infinite number of samples
 - ▶ Summing up Gaussian random variables results in a Gaussian random variable
- ▶ Since $z \sim \mathcal{N}(0, \sigma^2)$, we need to calculate σ^2 :

$$\begin{aligned}\sigma^2 = E\{z^2\} &= \frac{N_0}{2} \int_0^T (s_1(t) - s_2(t))^2 dt \\ &= \frac{N_0}{2} (E_{s_1} + E_{s_2} - 2\rho_{12}) \rightarrow \text{Assume } E_{s_1} = E_{s_2} = E \\ &= N_0(E - \rho_{12})\end{aligned}$$

$$\text{where } E = E_i = \int_0^T s_i^2(t) dt \text{ and } \rho_{12} = \int_0^T s_1(t)s_2(t) dt$$

Solving for the Probability of Error

- ▶ The probability of an error occurring given that a “1” was transmitted, i.e., $P(e|1)$, is equal to:

$$\begin{aligned} P(z \geq E - \rho_{12}) &= Q\left(\frac{E - \rho_{12}}{\sigma}\right) \rightarrow \begin{array}{l} \text{Since } z \sim \mathcal{N}(0, \sigma^2) \\ \text{and } E - \rho_{12} \text{ is constant} \end{array} \\ &= Q\left(\sqrt{\frac{(E - \rho_{12})^2}{\sigma^2}}\right) \rightarrow \text{Use } \sigma^2 = N_0(E - \rho_{12}) \\ &= Q\left(\sqrt{\frac{E - \rho_{12}}{N_0}}\right) \end{aligned}$$

where the Q-function is defined as:

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-t^2/2} dt \quad (3)$$

Optimizing the Probability of Error

- ▶ To minimize $P(e|1)$, we need to optimize ρ_{12}
 - ▶ Intuitively, the best choice we can make is $s_2(t) = -s_1(t)$, which gives us $\rho_{12} = -E$
 - ▶ This yields $P(e|1) = Q\left(\sqrt{\frac{2\bar{E}_b}{N_0}}\right)$
- ▶ As an exercise for the student, show that the total probability of error is equal to:

$$P_e = P(e|1)P(1) + P(e|0)P(0) = Q\left(\sqrt{\frac{E - \rho_{12}}{N_0}}\right) \quad (4)$$

- ▶ When $E_{s_1} \neq E_{s_2}$, then $d_{\min}^2 = E_{s_1} + E_{s_2} - 2\rho_{12}$, which yields:

$$P_e = Q\left(\sqrt{\frac{d_{\min}^2}{2N_0}}\right) \quad (5)$$

Signal Vectors

- ▶ Let $\phi_j(t)$ be an orthonormal set of functions on the time interval $[0, T]$ such that:

$$\int_0^T \phi_i(t) \phi_j(t) dt = \begin{cases} 1 & i = j \\ 0 & \text{otherwise} \end{cases}$$

- ▶ Let $s_i(t)$ be the modulation signal that we can represent in terms of the orthonormal functions:

$$s_i(t) = \sum_{k=1}^N s_{ik} \phi_k(t) \quad (6)$$

which can be represented by the vector:

$$s_i(t) \rightarrow \mathbf{s}_i = (s_{i1}, s_{i2}, s_{i3}, \dots, s_{iN}) \quad (7)$$

Vector Representation of $s_i(t)$

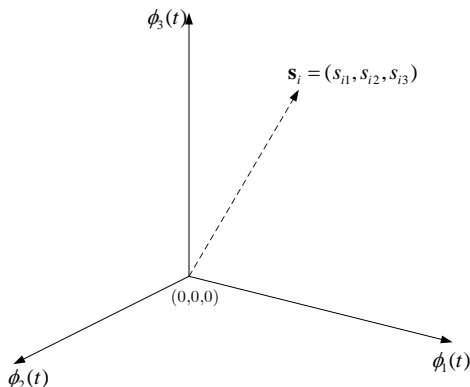


Figure : Sample Vector Representation of $s_i(t)$ in Three Dimensional Space using Basis Functions $\phi_1(t)$, $\phi_2(t)$, and $\phi_3(t)$.

Vector Manipulations

- ▶ To find s_{ik} , solve:

$$\int_0^T s_i(t) \phi_l(t) dt = \sum_{k=1}^N s_{ik} \int_0^T \phi_k(t) \phi_l(t) dt = s_{il} \quad (8)$$

- ▶ The vector dot product between $s_i(t)$ and $s_j(t)$ is equal to:

$$\int_0^T s_i(t) s_j(t) dt = \mathbf{s}_i \cdot \mathbf{s}_j = \rho_{ij} \quad (9)$$

while the energy of a signal $s_i(t)$ is equal to:

$$E_{s_i} = \int_0^T s_i^2(t) dt = \mathbf{s}_i \cdot \mathbf{s}_i = \|\mathbf{s}_i\|^2 \quad (10)$$

Euclidean Distance

- To compute the *Euclidean Distance* using signal space vectors, we need to solve:

$$\begin{aligned}d_{\min}^2 &= \int_0^T \Delta s_{ij}^2(t) dt = \int_0^T (s_i(t) - s_j(t))^2 dt \\&= \|\mathbf{s}_i - \mathbf{s}_j\|^2 = (\mathbf{s}_i - \mathbf{s}_j) \cdot (\mathbf{s}_i - \mathbf{s}_j) \\&= E_{s_i} + E_{s_j} - 2\rho_{ij}\end{aligned}$$

where:

$$\rho_{ij} = \int_0^T s_i(t)s_j(t)dt = \mathbf{s}_i \cdot \mathbf{s}_j \quad (11)$$

Solving for the Power Efficiency

- ▶ Choose orthonormal basis functions $\phi_i(t)$, $i = 1, 2, \dots, k$, where k is the dimension of the signal space
- ▶ Find \mathbf{s}_i , $i = 1, 2, \dots, M$ where $\mathbf{s}_i = (s_{i1}, s_{i2}, \dots, s_{ik})$ and $s_{ij} = \int_0^T s_i(t)\phi_j(t)dt$
- ▶ Consequently, solve for:

$$d_{\min}^2 = \min_{i \neq j} \|\mathbf{s}_i - \mathbf{s}_j\|^2$$

$$\bar{E}_s = \frac{1}{M} \sum_{i=1}^M \|\mathbf{s}_i\|^2$$

$$\bar{E}_b = \bar{E}_s / \log_2(M)$$

$$\epsilon_p = d_{\min}^2 / \bar{E}_b$$