

Digital Communications

— Lecture 03 — On-Off Keying

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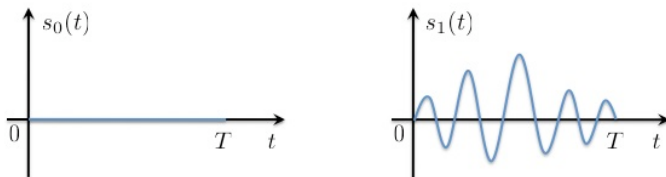
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System Model

Binary memoryless modulation: $M = 2$, $\mathcal{A} = \{0, 1\}$, $\Pi = \{p, 1 - p\}$

Consider **real-valued** signals and noise

$$s_0(t) = 0 \quad s_1(t) = p(t) \quad t \in [0, T)$$



AWGN channel: $\mu_w(t) = 0$, $P_w(f) = \eta_0/2$

Outline

- 1 System Model
- 2 Sufficient Statistic
- 3 Optimum Receiver
- 4 Performance

Signal Constellation

The signal space has dimension $N = 1$

$$\psi_1(t) = \frac{1}{\sqrt{\mathcal{E}_p}} p(t)$$
$$\mathcal{E}_p = \int_0^T p^2(t) dt$$

The signal constellation is

$$s_0 = \langle s_0, \psi_1 \rangle = 0$$
$$s_1 = \langle s_1, \psi_1 \rangle = \sqrt{\mathcal{E}_p}$$



Received Signal

Two possible hypotheses in $t \in [0, T]$

$$H_0 : r(t) = w(t)$$

$$H_1 : r(t) = p(t) + w(t)$$

To represent the noise, thus the received signal, we need an infinite orthonormal set $\{\psi_n(t)\}_{n=1}^{\infty}$. Assume that $\psi_1(t) = p(t)/\sqrt{\mathcal{E}_p}$

$$\begin{aligned} \mathbf{r}|H_0 &= \begin{pmatrix} \langle w, \psi_1 \rangle \\ \langle w, \psi_2 \rangle \\ \vdots \end{pmatrix} = \begin{pmatrix} w_1 \\ w_2 \\ \vdots \end{pmatrix} \\ \mathbf{r}|H_1 &= \begin{pmatrix} \langle s + w, \psi_1 \rangle \\ \langle s + w, \psi_2 \rangle \\ \vdots \end{pmatrix} = \begin{pmatrix} \sqrt{\mathcal{E}_p} \\ 0 \\ \vdots \end{pmatrix} + \begin{pmatrix} w_1 \\ w_2 \\ \vdots \end{pmatrix} \end{aligned}$$

Sufficient Statistic (1/2)

Two considerations are crucial:

- Only the **first component** of the received vector is dependent on the two hypotheses
- The components of the received vector are **statistically independent**

The first component r_1 , or simply r , represents a **sufficient statistic** for the considered detection problem, thus is the only to be computed

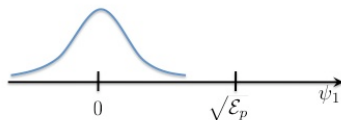
$$r|H_0 = w_1 \sim \mathcal{N}\left(0, \frac{\eta_0}{2}\right)$$

$$r|H_1 = \sqrt{\mathcal{E}_p} + w_1 \sim \mathcal{N}\left(\sqrt{\mathcal{E}_p}, \frac{\eta_0}{2}\right)$$

Sufficient Statistic (2/2)

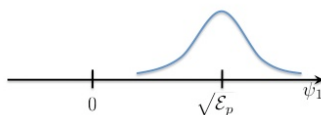
When transmitting the symbol **0** we observe a random variable with pdf

$$f_{r|H_0}(r) = \frac{1}{\sqrt{\pi\eta_0}} \exp\left(-\frac{r^2}{\eta_0}\right)$$



When transmitting the symbol **1** we observe a random variable with pdf

$$f_{r|H_1}(r) = \frac{1}{\sqrt{\pi\eta_0}} \exp\left(-\frac{(r - \sqrt{\mathcal{E}_p})^2}{\eta_0}\right)$$



Optimum Decision (1/2)

Decision is done via $\{\Omega_0, \Omega_1\}$ representing a **partition** of \mathbb{R} , i.e. $\Omega_0 \cup \Omega_1 = \mathbb{R}$ and $\Omega_0 \cap \Omega_1 = \emptyset$, with

$$r \in \Omega_0 \rightarrow \text{rx } 0, \quad r \in \Omega_1 \rightarrow \text{rx } 1$$

Each partition leads to a different **error probability**

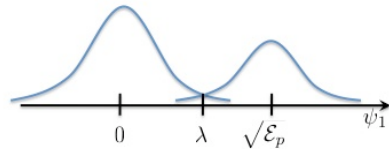
$$\begin{aligned} P_e &\triangleq \Pr\left\{(\text{tx } 0, \text{rx } 1) \cup (\text{tx } 1, \text{rx } 0)\right\} \\ &= \Pr\left\{(\text{rx } 1|\text{tx } 0)\right\} p + \Pr\left\{(\text{rx } 0|\text{tx } 1)\right\} (1-p) \\ &= \Pr\left\{r|H_0 \in \Omega_1\right\} p + \Pr\left\{r|H_1 \in \Omega_0\right\} (1-p) \\ &= \int_{\Omega_1} p f_{r|H_0}(r) dr_1 + \int_{\Omega_0} (1-p) f_{r|H_1}(r) dr_1 \\ &= \int_{\mathbb{R}-\Omega_0} p f_{r|H_0}(r) dr_1 + \int_{\Omega_0} (1-p) f_{r|H_1}(r) dr_1 \\ &= p - \int_{\Omega_0} (p f_{r|H_0}(r) - (1-p) f_{r|H_1}(r)) dr_1 \end{aligned}$$

Optimum Decision (2/2)

The **optimum decision** leads to the **minimum** $\Pr(e)$, thus

$$\Omega_0 = \{r \in \mathbb{R} : pf_{r|H_0}(r) > (1-p)f_{r|H_1}(r)\}$$

$$\Omega_1 = \{r \in \mathbb{R} : pf_{r|H_0}(r) < (1-p)f_{r|H_1}(r)\}$$



$$\Omega_0 = \{r \in \mathbb{R} : r < \lambda\}$$

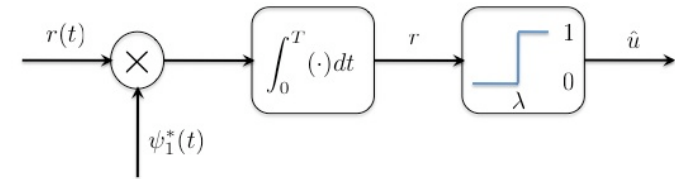
$$\Omega_1 = \{r \in \mathbb{R} : r > \lambda\}$$

with

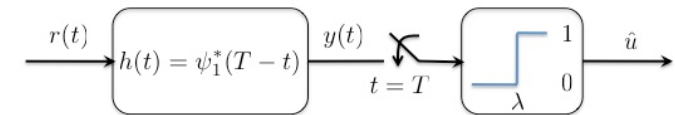
$$\lambda : pf_{r|H_0}(\lambda) = (1-p)f_{r|H_1}(\lambda)$$

Receiver Architecture (1/2)

First option



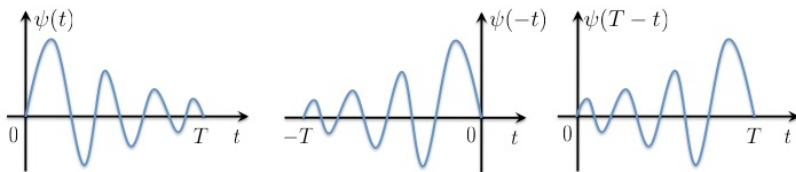
Second option



Receiver Architecture (2/2)

$$y(t) = r(t) \star h(t) = \int_{-\infty}^{+\infty} r(\tau)h(t-\tau)d\tau$$

$$= \int_0^t r(\tau)\psi_1(T-t+\tau)d\tau$$



$$y(t)|_{t=T} = \int_0^T r(\tau)\psi_1(\tau)d\tau = r$$

Optimum Threshold

$$\lambda : pf_{r|H_0}(\lambda) = (1-p)f_{r|H_1}(\lambda)$$

$$\frac{p}{\sqrt{\pi\eta_0}} \exp\left(-\frac{\lambda^2}{\eta_0}\right) = \frac{1-p}{\sqrt{\pi\eta_0}} \exp\left(-\frac{(\lambda - \sqrt{\mathcal{E}_p})^2}{\eta_0}\right)$$

The optimum threshold depends on the **energy**, on the **noise** PSD and on the **a-priori probabilities**:

$$\lambda = \frac{\sqrt{\mathcal{E}_p}}{2} + \frac{\eta_0}{2\sqrt{\mathcal{E}_p}} \log\left(\frac{p}{1-p}\right)$$

In case of transmission with **equiprobable symbols**, i.e. $p = 1/2$, the optimum decision is a **minimum distance decision**:

$$\lambda = \frac{\sqrt{\mathcal{E}_p}}{2}$$

Error Probability (1/2)

$$\begin{aligned}P_e &= 1 - P_c = 1 - p \Pr(c|H_0) - (1-p) \Pr(c|H_1) \\&= 1 - p \Pr(r|H_0 < \lambda) - (1-p) \Pr(r|H_1 > \lambda) \\&= 1 - p \int_{-\infty}^{\lambda} f_{r|H_0}(r) dr - (1-p) \int_{\lambda}^{+\infty} f_{r|H_1}(r) dr \\&= 1 - p \Pr\left(\mathcal{N}\left(0, \frac{\eta_0}{2}\right) < \lambda\right) - (1-p) \Pr\left(\mathcal{N}\left(\sqrt{\mathcal{E}_p}, \frac{\eta_0}{2}\right) > \lambda\right) \\&= 1 - p \left(1 - Q\left(\frac{\lambda}{\sqrt{\eta_0/2}}\right)\right) - (1-p) Q\left(\frac{\lambda - \sqrt{\mathcal{E}_p}}{\sqrt{\eta_0/2}}\right)\end{aligned}$$

In case of equiprobable symbols

$$P_e = Q\left(\sqrt{\frac{\mathcal{E}_p}{2\eta_0}}\right)$$

Error Probability (2/2)

