

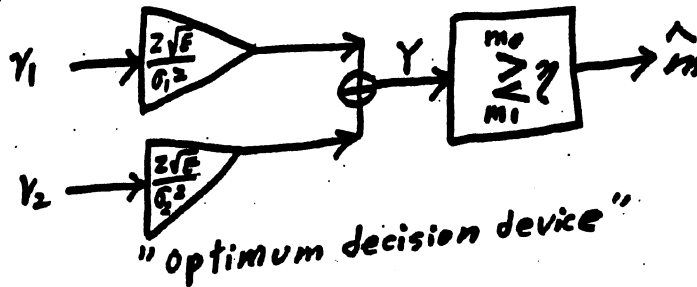
① 2.25: (a)

$$\begin{cases} p(\underline{y} | m_0) = \frac{e^{-\frac{(y_1 - \sqrt{E})^2}{2\sigma_1^2}}}{\sqrt{2\pi\sigma_1^2}} \cdot \frac{e^{-\frac{(y_2 - \sqrt{E})^2}{2\sigma_2^2}}}{\sqrt{2\pi\sigma_2^2}} \\ p(\underline{y} | m_1) = \frac{e^{-\frac{(y_1 + \sqrt{E})^2}{2\sigma_1^2}}}{\sqrt{2\pi\sigma_1^2}} \cdot \frac{e^{-\frac{(y_2 + \sqrt{E})^2}{2\sigma_2^2}}}{\sqrt{2\pi\sigma_2^2}} \end{cases}$$

$$\text{MAP: } p[m_0] p(\underline{y} | m_0) \underset{m_1}{\overset{m_0}{\geq}} p[m_1] p(\underline{y} | m_1)$$

$$\text{Simplify: } \left(\frac{2\sqrt{E}}{\sigma_1^2}\right)y_1 + \left(\frac{2\sqrt{E}}{\sigma_2^2}\right)y_2 \underset{m_1}{\overset{m_0}{\geq}} \ln \frac{p[m_1]}{p[m_0]} = \eta$$

(i)



$$(ii) \ m = m_1, \begin{cases} y_1 = -\sqrt{E} + n_1 \\ y_2 = -\sqrt{E} + n_2 \end{cases}$$

$$Y = \frac{2\sqrt{E}}{\sigma_1^2} (-\sqrt{E} + n_1) + \frac{2\sqrt{E}}{\sigma_2^2} (-\sqrt{E} + n_2)$$

$$= (-2E) \left(\frac{1}{\sigma_1^2} + \frac{1}{\sigma_2^2} \right) + 2\sqrt{E} \left(\frac{n_1}{\sigma_1^2} + \frac{n_2}{\sigma_2^2} \right)$$

$$\rightarrow \bar{Y} = (-2E) \left(\frac{1}{\sigma_1^2} + \frac{1}{\sigma_2^2} \right), \quad \sigma_Y^2 = \frac{4E}{\sigma_1^2} + \frac{4E}{\sigma_2^2}$$

$$P[\text{error} | m_1] = P[Y > \eta] = \int_{\eta}^{\infty} \frac{e^{-\frac{(y - \bar{Y})^2}{2\sigma_Y^2}}}{\sqrt{2\pi\sigma_Y^2}} dy \quad (1)$$

$$m = m_0, \begin{cases} y_1 = \sqrt{E} + n_1 \\ y_2 = \sqrt{E} + n_2 \end{cases}$$

(2)

$$Y = \frac{2\sqrt{E}}{\sigma_1^2} (\sqrt{E} + n_1) + \frac{2\sqrt{E}}{\sigma_2^2} (\sqrt{E} + n_2)$$

$$= (2E) \left(\frac{1}{\sigma_1^2} + \frac{1}{\sigma_2^2} \right) + 2\sqrt{E} \left(\frac{n_1}{\sigma_1^2} + \frac{n_2}{\sigma_2^2} \right)$$

$$\rightarrow \bar{y} = (2E) \left(\frac{1}{\sigma_1^2} + \frac{1}{\sigma_2^2} \right), \sigma_y^2 = \frac{4E}{\sigma_1^2} + \frac{4E}{\sigma_2^2}$$

$$P[\text{error} | m_0] = P[Y < \gamma] = \int_{-\infty}^{\gamma} \frac{e^{-\frac{(y-\bar{y})^2}{2\sigma_y^2}}}{\sqrt{2\pi\sigma_y^2}} dy \quad (2)$$

From (1) & (2) $\rightarrow P[\text{error}] = P[\text{error} | m_0] P[m_0] + P[\text{error} | m_1] P[m_1]$.

(b)

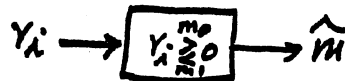
$$\begin{cases} P[\text{error} | m_1] = Q\left(\frac{4E/\sigma^2}{\sqrt{8E/\sigma^2}}\right) = Q\left(\sqrt{\frac{2E}{\sigma^2}}\right), \\ P[\text{error} | m_0] = Q\left(\frac{4E/\sigma^2}{\sqrt{8E/\sigma^2}}\right) = Q\left(\sqrt{\frac{2E}{\sigma^2}}\right), \\ P[\text{error}] = Q\left(\sqrt{\frac{2E}{\sigma^2}}\right). \end{cases}$$

Based on y_1 only, $P[\text{error}] = Q\left(\sqrt{\frac{E}{\sigma^2}}\right)$
 $> Q\left(\sqrt{\frac{2E}{\sigma^2}}\right).$

2.39:
(b)

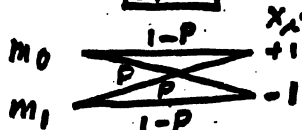
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(i)

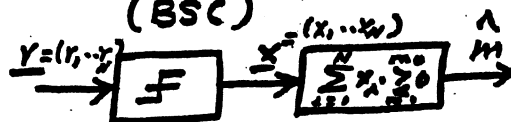


$$M = m_0, Y_i = \sqrt{E} + \eta_i, P[Y_i < 0 | m_0] = P[\sqrt{E} + \eta_i < 0] = P[\eta_i < -\sqrt{E}]$$

$$= \int_{-\infty}^{-\sqrt{E}} \frac{e^{-\eta_i^2/2\sigma^2}}{\sqrt{2\pi\sigma^2}} d\eta_i = Q(\sqrt{E}/\sigma) = P$$



(BSC)



$$(ii) P[\text{error} | m_1] = P[x > 0 | m_1] \begin{cases} x = \sum_{i=1}^N x_i, \\ P[x_i = 1 | m_1] = p, \\ P[x_i = -1 | m_1] = 1-p. \end{cases}$$

$$\leq E[e^{vx}]$$

$$= E[e^{v(x_1 + \dots + x_N)}]$$

$$= E[e^{vx_1}] E[e^{vx_2}] \dots E[e^{vx_N}]$$

$$= [pe^v + (1-p)e^{-v}]^N$$

$$\text{Chernoff Bound: } \frac{\partial}{\partial v} [pe^v + (1-p)e^{-v}]$$

$$= pe^v - (1-p)e^{-v} = 0 \rightarrow e^v = \sqrt{\frac{1-p}{p}} > 1.$$

$$\rightarrow P[\text{error}] \leq \left[\sqrt{4p(1-p)} \right]^N$$

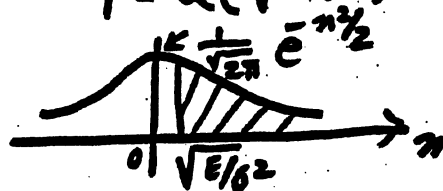
④

→ By symmetry,

$$P[\text{error}] \leq e^{-N[-\ln \sqrt{4p(1-p)}]}$$

$$= e^{-N\left\{-\frac{1}{2} \ln[4p(1-p)]\right\}}$$

(iii) $p = Q(\sqrt{E/\sigma^2})$



$$\rightarrow \frac{1}{2} - \frac{1}{\sqrt{2\pi}} \times \sqrt{\frac{E}{\sigma^2}} \triangleq p$$

$$\rightarrow 4p(1-p) \approx 1 - \frac{2E}{\pi\sigma^2}, \quad \left(\frac{E}{\sigma^2} \ll 1\right)$$

$$\rightarrow -\frac{1}{2} \ln[4p(1-p)] = \frac{1}{2} \times \frac{2E}{\pi\sigma^2} = \frac{E}{\pi\sigma^2}$$

$$\rightarrow P[\text{error}] \leq e^{-\frac{NE}{\pi\sigma^2}} = e^{-\frac{(NE)}{2\sigma^2} \left(\frac{2}{\pi}\right)}$$

Hard

$$\rightarrow P_{\text{soft}}^{\text{Hard}}[\text{error}] \leq \frac{1}{2} e^{-\frac{(NE)}{2\sigma^2}}$$

$$10 \log_{10}\left(\frac{2}{\pi}\right) = \underline{\underline{2 \text{ dB (degradation)}}}$$

3.11: $X, Y \sim \text{i.i.d } N(0,1)$

(5)

(a) $\overline{z(t)} = \underline{\underline{\phi}}$

$$R_z(t_1, t_2) = E[z(t_1)z(t_2)] \\ = \underline{\underline{\cos[2\pi(t_1 - t_2)]}}$$

→ $z(t)$ is a stationary Gaussian random process.

(A)-II $\underline{\underline{\Lambda}}_z = \begin{bmatrix} 1 & \cos 2\pi(t_1 - t_2) \\ \cos 2\pi(t_1 - t_2) & 1 \end{bmatrix}$

$$\underline{\underline{\Lambda}}_z^{-1} = \frac{1}{\begin{bmatrix} 1 - \cos^2 2\pi(t_1 - t_2) \\ 1 - \cos 2\pi(t_1 - t_2) \\ -\cos 2\pi(t_1 - t_2) & 1 \end{bmatrix}} \begin{bmatrix} 1 & -\cos 2\pi(t_1 - t_2) \\ -\cos 2\pi(t_1 - t_2) & 1 \end{bmatrix}$$

→

$$= \frac{1}{\sin^2 2\pi(t_1 - t_2)}$$

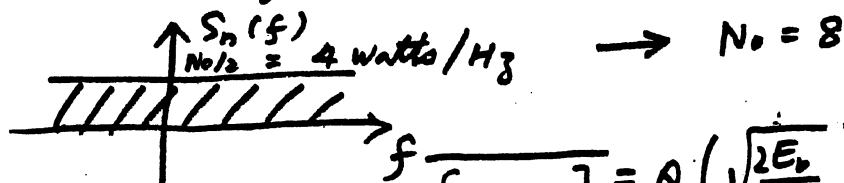
$$\& |\underline{\underline{\Lambda}}_z| = \sin^2 2\pi(t_1 - t_2)$$

→ $p_z(\underline{z}) = e^{-\frac{1}{2} \underline{z} \underline{\underline{\Lambda}}_z^{-1} \underline{z}^*}$

$$= \frac{(2\pi)^{1/2} |\underline{\underline{\Lambda}}_z|^{1/2}}{2\pi |\sin 2\pi(t_1 - t_2)|} e^{-\frac{1}{2} \frac{z_1^2 - [2\cos 2\pi(t_1 - t_2)]z_1 z_2 + z_2^2}{\sin^2 2\pi(t_1 - t_2)}}$$

4 4.7 $\int_0^T x^2(t) dt = \int_0^T y^2(t) dt = 16 \text{ joules} = E_b$ **5**

$\begin{cases} \int_0^T x(t) y(t) dt = 0 \rightarrow x(t) \text{ \& } y(t) \text{ are equal-energy} \\ \text{orthogonal signals.} \end{cases}$



(a) $f = -1 \rightarrow P[\text{error}] = Q\left(\sqrt{\frac{2E_b}{N_0}}\right) = Q\left(\sqrt{\frac{32}{8}}\right) = \underline{\underline{Q(2)}}$

(b) $f = 0 \rightarrow P[\text{error}] = Q\left(\sqrt{\frac{E_b}{N_0}}\right) = \underline{\underline{Q(\sqrt{2})}}$

5 4.8 (a) $\frac{N_0}{2} = 0.15 \rightarrow N_0 = \underline{\underline{0.3}}$, $P[m_0] = P[m_1] = 1/2$,

(i) $\int_0^1 s_0^2(t) dt = 1$, $\int_0^1 s_1^2(t) dt = \frac{1}{2}$, $\int_0^1 s_0(t) s_1(t) dt = 0$.

$E_b = \frac{1}{2} + \frac{1}{2} = \underline{\underline{\frac{3}{2}}}$ $\rightarrow P[\text{error}] = Q\left(\sqrt{\frac{(3/4)}{0.3}}\right) = Q(\sqrt{2.5}) = \underline{\underline{Q(1.5)}}$

(ii) $\int_0^1 s_0^2(t) dt = 4$, $\int_0^2 s_1^2(t) dt = 8$, $\int_0^2 s_0(t) s_1(t) dt = 4$,

$E_b = \frac{4+8}{2} = \underline{\underline{6}}$, $f = \frac{4}{6} = \underline{\underline{\frac{2}{3}}}$.

$P[\text{error}] = Q\left(\sqrt{\frac{(1-f)E_b}{N_0}}\right) = Q\left(\sqrt{\frac{\frac{1}{3} \times 6}{0.3}}\right) = \underline{\underline{Q\left(\sqrt{\frac{20}{3}}\right) = Q(2.58)}}$

(iii) $S_0(f)$ $\xrightarrow{\text{Parseval Theorem}}$ $\int_{-\infty}^{\infty} S_0^2(t) dt = \int_{-1}^1 S_0^2(f) df = 2$

$\int_{-\infty}^{\infty} S_1^2(t) dt = \int_{-1/2}^{-1/2} S_1^2(f) df + \int_{1/2}^{3/2} S_1^2(f) df$

(page 238)

$= 2$

$\int_{-\infty}^{\infty} S_0(t) S_1(t) dt = 2 \cdot \int_{1/2}^1 1 df = \underline{\underline{1}}$

$\rightarrow E_b = \frac{2+2}{2} = 2$, $f = \frac{1}{2} = \underline{\underline{1/2}}$

$P[\text{error}] = Q\left(\sqrt{\frac{(1-f)E_b}{N_0}}\right) = Q\left(\sqrt{\frac{\frac{1}{2} \times 2}{0.3}}\right) = Q\left(\sqrt{\frac{10}{3}}\right) = \underline{\underline{Q(1.82)}}$

$$\rightarrow p(\underbrace{z_1, z_2, z_3}_{\underline{z}}) = p(\underbrace{z_1, z_3}_{z_1, z_3}) p(\underbrace{z_2}_{z_2}) \quad (7)$$

$$\left(\begin{array}{l} z_2 \text{ is independent of } z_1 \text{ and } z_3 \\ \overline{z_2 z_3} = \overline{z_2 z_1} = 0 \end{array} \right)$$

$$\overline{z_1 z_3} = -1 \rightarrow \underline{\underline{z_1 = -z_3}} \quad (\text{"degenerate" case})$$

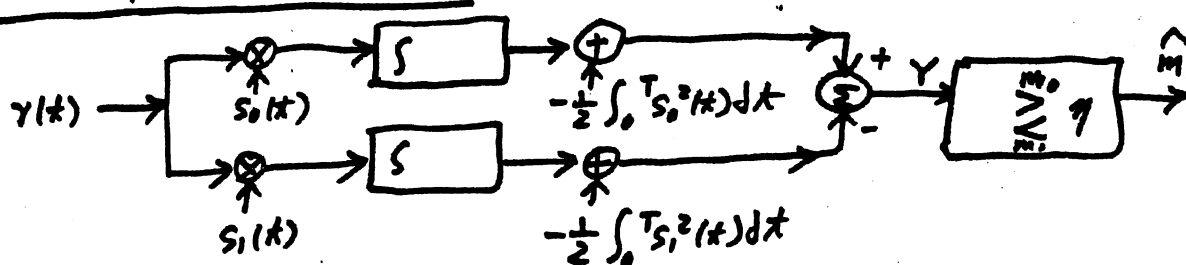
$$\rightarrow p(\underbrace{z_1, z_3}_{z_1, z_3}) = p(\underbrace{z_1}_{z_1}) \delta(z_1 + z_3)$$

$$\begin{aligned} \underline{\underline{p(z_1, z_2, z_3)}} &= p(\underbrace{z_1}_{z_1}) p(\underbrace{z_2}_{z_2}) \delta(z_1 + z_3) \\ &= \frac{e^{-\frac{(z_1^2 + z_2^2)}{2}}}{2\pi} \delta(z_3 + z_1) \end{aligned}$$

4.8 (b) : $P[m_0] = 1/4$, $P[m_1] = 3/4$

(9)

new optimal receiver



where $\eta = \frac{N_0}{2} \log \frac{P[m_1]}{P[m_0]}$, $Y \triangleq \int_0^T y(t) [s_0(t) - s_1(t)] dt - \frac{1}{2} \int_0^T (s_0^2(t) - s_1^2(t)) dt$

$P[Y > \eta | m = m_1] \triangleq P[\text{error} | m_1]$

$$= \int_{\eta}^{\infty} \frac{e^{-(y + E_b(1-\beta))^2 / 2 N_0 E_b(1-\beta)}}{\sqrt{2\pi N_0 E_b(1-\beta)}} dy$$

$$= Q\left(\frac{\eta + E_b(1-\beta)}{\sqrt{N_0 E_b(1-\beta)}}\right)$$

$$P[Y < \eta | m = m_0] = \int_{-\infty}^{\eta} \frac{e^{-(y - E_b(1-\beta))^2 / 2 N_0 E_b(1-\beta)}}{\sqrt{2\pi N_0 E_b(1-\beta)}} dy$$

$$= Q\left(\frac{E_b(1-\beta) - \eta}{\sqrt{N_0 E_b(1-\beta)}}\right)$$

$$P[\text{error}] = P[m_0] Q\left(\frac{E_b(1-\beta) - \eta}{\sqrt{N_0 E_b(1-\beta)}}\right) + P[m_1] Q\left(\frac{E_b(1-\beta) + \eta}{\sqrt{N_0 E_b(1-\beta)}}\right)$$

(i) $\overline{p(\text{error})} = \frac{1}{4} Q(1.23) + \frac{3}{4} Q(1.92)$

(ii) $\overline{p(\text{error})} = \frac{1}{4} Q(2.36) + \frac{3}{4} Q(2.79)$

(iii) $\overline{p(\text{error})} = \frac{1}{4} Q(1.52) + \frac{3}{4} Q(2.12)$