

(b) Let
$$Y = \frac{1}{n} \sum_{i=1}^n X_i$$

Then by Eqs. (4.108) and (4.112), and the result of Prob. 5.60, we see that Y is a normal r.v. with zero mean and variance $1/n$ under H_0 , and is a normal r.v. with mean 1 and variance $1/n$ under H_1 . Thus

$$\begin{aligned} P_1 &= P(D_1 | H_0) = \int_{R_1} f_Y(y | H_0) dy = \frac{\sqrt{n}}{\sqrt{2\pi}} \int_{1/2}^{\infty} e^{-(n/2)y^2} dy \\ &= \frac{1}{\sqrt{2\pi}} \int_{\sqrt{n}/2}^{\infty} e^{-z^2/2} dz = 1 - \Phi(\sqrt{n}/2) \\ P_0 &= P(D_0 | H_1) = \int_{R_0} f_Y(y | H_1) dy = \frac{\sqrt{n}}{\sqrt{2\pi}} \int_{1/2}^{\infty} e^{-(n/2)(y-1)^2} dy \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{-\sqrt{n}/2} e^{-z^2/2} dz = \Phi(-\sqrt{n}/2) = 1 - \Phi(\sqrt{n}/2) \end{aligned}$$

Note that $P_1 = P_0$. Using Table A (Appendix A), we have

$$\begin{aligned} P_1 = P_0 &= 1 - \Phi(1.118) = 0.1318 & \text{for } n = 5 \\ P_1 = P_0 &= 1 - \Phi(1.581) = 0.057 & \text{for } n = 10 \end{aligned}$$

- 8.16. In the binary communication system of Prob. 8.6, suppose that $s_0(t)$ and $s_1(t)$ are arbitrary signals and that n observations of the received signal $x(t)$ are made. Let n samples of $s_0(t)$ and $s_1(t)$ be represented, respectively, by

$$\mathbf{s}_0 = [s_{01}, s_{02}, \dots, s_{0n}]^T \quad \text{and} \quad \mathbf{s}_1 = [s_{11}, s_{12}, \dots, s_{1n}]^T$$

where T denotes "transpose of." Determine the MAP test.

For each X_i , we can write

$$\begin{aligned} f(x_i | H_0) &= \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{1}{2}(x_i - s_{0i})^2\right] \\ f(x_i | H_1) &= \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{1}{2}(x_i - s_{1i})^2\right] \end{aligned}$$

Since the noise components are independent, we have

$$f(\mathbf{x} | H_j) = \prod_{i=1}^n f(x_i | H_j) \quad j = 0, 1$$

and the likelihood ratio is given by

$$\begin{aligned} \Lambda(\mathbf{x}) &= \frac{f(\mathbf{x} | H_1)}{f(\mathbf{x} | H_0)} = \frac{\prod_{i=1}^n \exp\left[-\frac{1}{2}(x_i - s_{1i})^2\right]}{\prod_{i=1}^n \exp\left[-\frac{1}{2}(x_i - s_{0i})^2\right]} \\ &= \exp\left[\sum_{i=1}^n (s_{1i} - s_{0i})x_i - \frac{1}{2}(s_{1i}^2 - s_{0i}^2)\right] \end{aligned}$$

Thus, by Eq. (8.15), the MAP test is given by

$$\exp\left[\sum_{i=1}^n (s_{1i} - s_{0i})x_i - \frac{1}{2}(s_{1i}^2 - s_{0i}^2)\right] \underset{H_0}{\overset{H_1}{\gtrless}} \eta = \frac{P(H_0)}{P(H_1)}$$

Taking the natural logarithm of both sides of the above expression yields

$$\sum_{i=1}^n (s_{1i} - s_{0i})x_i \underset{H_0}{\overset{H_1}{\gtrless}} \ln \left[\frac{P(H_0)}{P(H_1)} \right] + \frac{1}{2}(s_{1i}^2 - s_{0i}^2) \quad (8.38)$$