

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

$$\begin{aligned} \frac{\mu_1 - \mu_0}{\sigma^2} \sum_{i=1}^n x_i &\underset{H_0}{\overset{H_1}{\geq}} \frac{n(\mu_1^2 - \mu_0^2)}{2\sigma^2} \\ \text{or} \quad \frac{1}{n} \sum_{i=1}^n x_i &\underset{H_0}{\overset{H_1}{\geq}} \frac{\mu_1 + \mu_0}{2} \end{aligned} \quad (8.36)$$

Equation (8.36) indicates that the statistic

$$s(X_1, \dots, X_n) = \frac{1}{n} \sum_{i=1}^n X_i = \bar{X}$$

provides enough information about the observations to enable us to make a decision. Thus, it is called the *sufficient statistic* for the maximum likelihood test.

- 8.14.** Consider the same observations  $X_i, i = 1, \dots, n$ , of radar signals as in Prob. 8.13, but now, under  $H_0$ ,  $X_i$  have zero mean and variance  $\sigma_0^2$ , while under  $H_1$ ,  $X_i$  have zero mean and variance  $\sigma_1^2$ , and  $\sigma_1^2 > \sigma_0^2$ . Determine the maximum likelihood test.

In a similar manner as in Prob. 8.13, we obtain

$$\begin{aligned} f(\mathbf{x} | H_0) &= \frac{1}{(2\pi\sigma_0^2)^{n/2}} \exp\left(-\frac{1}{2\sigma_0^2} \sum_{i=1}^n x_i^2\right) \\ f(\mathbf{x} | H_1) &= \frac{1}{(2\pi\sigma_1^2)^{n/2}} \exp\left(-\frac{1}{2\sigma_1^2} \sum_{i=1}^n x_i^2\right) \end{aligned}$$

With  $\sigma_1^2 - \sigma_0^2 > 0$ , the likelihood ratio is

$$\Lambda(\mathbf{x}) = \frac{f(\mathbf{x} | H_1)}{f(\mathbf{x} | H_0)} = \left(\frac{\sigma_0}{\sigma_1}\right)^n \exp\left[\left(\frac{\sigma_1^2 - \sigma_0^2}{2\sigma_0^2\sigma_1^2}\right) \sum_{i=1}^n x_i^2\right]$$

and the maximum likelihood test is

$$\left(\frac{\sigma_0}{\sigma_1}\right)^n \exp\left[\left(\frac{\sigma_1^2 - \sigma_0^2}{2\sigma_0^2\sigma_1^2}\right) \sum_{i=1}^n x_i^2\right] \underset{H_0}{\overset{H_1}{\geq}} 1$$

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

$$\sum_{i=1}^n x_i^2 \underset{H_0}{\overset{H_1}{\geq}} n \left[ \ln\left(\frac{\sigma_1}{\sigma_0}\right) \right] \left( \frac{2\sigma_0^2\sigma_1^2}{\sigma_1^2 - \sigma_0^2} \right) \quad (8.37)$$

Note that in this case,

$$s(X_1, \dots, X_n) = \sum_{i=1}^n X_i^2$$

is the sufficient statistic for the maximum likelihood test.

- 8.15.** In the binary communication system of Prob. 8.6, suppose that we have  $n$  independent observations  $X_i = X(t_i), i = 1, \dots, n$ , where  $0 < t_1 < \dots < t_n \leq T$ .

(a) Determine the maximum likelihood test.

(b) Find  $P_1$  and  $P_{II}$  for  $n = 5$  and  $n = 10$ .

(a) Setting  $\mu_0 = 0$  and  $\mu_1 = 1$  in Eq. (8.36), the maximum likelihood test is

$$\frac{1}{n} \sum_{i=1}^n x_i \underset{H_0}{\overset{H_1}{\geq}} \frac{1}{2}$$