

**Theorem.** Let  $\theta$  be a real parameter and let  $X$  have density  $f(x|\theta)$  with MLR in  $T(x)$ . Then

(i) for testing  $H_0 : \theta \leq \theta_0$  versus  $H_1 : \theta > \theta_0$ , there exists a UMP level  $\alpha$  test given by

$$\phi(x) = \begin{cases} 1 & \text{when } T(x) > C; \\ \gamma & \text{when } T(x) = C; \\ 0 & \text{when } T(x) < C, \end{cases} \quad (*)$$

where  $C$  and  $\gamma$  are determined by

$$E_{\theta_0} \phi(X) = \alpha. \quad (**)$$

(ii) The power function  $E_\theta \phi(X)$  of this test is strictly increasing for all points  $\theta$  for which  $E_\theta \phi(X) < 1$ .

(iii) For all  $\theta'$ , the test given by  $(*)$  and  $(**)$  is UMP for testing  $H_0 : \theta \leq \theta'$  versus  $H_1 : \theta > \theta'$  at the level  $\alpha' = E_{\theta'} \phi(X)$ .

(iv) For any  $\theta < \theta_0$ , the test  $(*)$  and  $(**)$  minimizes  $E_\theta \phi(X)$  among all tests satisfying  $(**)$ .

**Proof.** We first prove the existence of  $\phi$  satisfying  $(*)$  and  $(**)$ . Then we show that (ii) holds for this  $\phi$ . Then we prove the UMP part of (i).

Consider testing  $H_0 : \theta = \theta_0$  versus  $H_1 : \theta = \theta_1, \theta_1 > \theta_0$ . From N-P Lemma, MP test for this is of the form:

$$\phi(x) = \begin{cases} 1 & \text{if } \frac{p_{\theta_1}(x)}{p_{\theta_0}(x)} > k; \\ 0 & \text{if } \frac{p_{\theta_1}(x)}{p_{\theta_0}(x)} < k. \end{cases}$$

But  $p_\theta(x) = f(x|\theta)$  has MLR in  $T(x)$ . Therefore,

$$\frac{f(x|\theta_1)}{f(x|\theta_0)} > k \text{ iff } T(x) > C \text{ for some } C.$$

Hence,

$$\phi(x) = \begin{cases} 1 & \text{if } T(x) > C; \\ 0 & \text{if } T(x) < C. \end{cases}$$

Now, from N-P Lemma (a), there exist  $C$  and  $\gamma$  satisfying  $(*)$  and  $(**)$ . These do not depend on  $\theta_1$ . (Note, due to MLR, dependence of MP on  $\theta_1$  has been eliminated.) From N-P Lemma (b) this test is MP for testing  $H_0 : \theta = \theta'$  versus  $H_1 : \theta = \theta''$  at level  $\alpha' = E_{\theta'} \phi(X)$  if  $\theta' < \theta''$ . This is because, from MLR,

$$\frac{f(x|\theta'')}{f(x|\theta')} > k_1 \text{ iff } T(x) > C_1 \text{ for some } C_1$$

and the MP test is exactly of the same form as  $\phi$ . Now from the corollary to N-P Lemma, we have  $E_{\theta''}\phi(X) > E_{\theta'}\phi(X)$ , whenever  $\theta'' > \theta'$ . This proves (ii). The fact that  $E_\theta\phi(X)$  is strictly increasing implies that

$$E_\theta\phi(X) \leq \alpha \quad \forall \theta \leq \theta_0. \quad (***)$$

Now let us check what test is UMP for testing  $H_0 : \theta \leq \theta_0$  versus  $H_1 : \theta > \theta_0$ . For this, consider the class  $\mathcal{C}_1$  of all tests that satisfy (\*\*\*)�. (We want to know what is best in  $\mathcal{C}_1$ .) This class  $\mathcal{C}_1$  is contained in the class  $\mathcal{C}_2$  of tests that satisfy  $E_{\theta_0}\phi(X) \leq \alpha$ . In other words, we have,

$$\mathcal{C}_1 = \{\phi : E_\theta\phi(X) \leq \alpha \quad \forall \theta \leq \theta_0\} \subset \mathcal{C}_2 = \{\phi : E_{\theta_0}\phi(X) \leq \alpha\},$$

since  $E_\theta\phi(X) \leq \alpha \quad \forall \theta \leq \theta_0$  implies  $E_{\theta_0}\phi(X) \leq \alpha$ . From N-P Lemma (b), MP level  $\alpha$  test for testing  $H_0 : \theta = \theta_0$  versus  $H_1 : \theta = \theta_1$  (for any  $\theta_1 > \theta_0$ ) is given by (\*) and (\*\*). That means that  $\phi$  satisfying (\*) and (\*\*) is best (MP) in  $\mathcal{C}_2$  and this  $\phi$  belongs to  $\mathcal{C}_1$  also. Therefore it is best in  $\mathcal{C}_1$ . However, this test does not depend on  $\theta_1$ . Therefore it is UMP for  $H_1 : \theta > \theta_0$ . i.e., for any  $\theta > \theta_0$ ,  $E_\theta\phi(X) \geq E_{\theta^*}\phi(X)$  for all  $\theta^* \in \mathcal{C}_1$ .

**Example.**  $X \sim \text{Binomial}(n, p)$  and it is of interest to test  $H_0 : p \leq p_0$  versus  $H_1 : p > p_0$ . (Consider a clinical trial where the efficacy of a drug is being checked.) MLR exists in  $T(x) = x$ . Therefore, UMP exists and is given by

$$\phi(x) = \begin{cases} 1 & \text{if } x > x_0; \\ 0 & \text{if } x < x_0; \\ \gamma & \text{if } x = x_0. \end{cases}$$

Choose  $\gamma$  and  $x_0$  to satisfy

$$E_{p_0}\phi(X) = P_{p_0}(X > x_0) + \gamma P_{p_0}(X = x_0) = \alpha.$$

Let  $\alpha = 0.05$ ,  $n = 10$  and  $p_0 = 1/2$ . We have

$x$	$f(x p = 1/2)$	$P_{p=1/2}(X \geq x)$
10	0.000977	0.00098
9	0.009766	0.01074
8	0.043945	0.05469

Thus,  $x_0 = 8$  and

$$\gamma = \frac{\alpha - P(X > x_0|p = p_0)}{P(X = x_0|p = p_0)} = \frac{0.05 - 0.01074}{0.04395} = 0.8933.$$

**Example.**  $X_1, \dots, X_n$  i.i.d  $N(\mu, \sigma^2)$ ,  $\sigma^2$  known. Test  $H_0 : \mu \leq \mu_0$  versus  $H_1 : \mu > \mu_0$ .  $\bar{X}$  is sufficient for  $\mu$  and  $\bar{X} \sim N(\mu, \sigma^2/n)$ . It was shown previously,

directly, that the likelihood ratio is an increasing function of  $T(\bar{x}) = \bar{x}$ . This means MLR which can also be shown using the fact that  $N(\mu, \sigma^2)$ ,  $\sigma^2$  known is a one-parameter exponential family. Since

$$f(\bar{x}|\mu) = \sqrt{\frac{n}{2\pi}} \frac{1}{\sigma} \exp\left(-\frac{n}{2\sigma^2}(\bar{x} - \mu)^2\right) = \exp\left(\frac{n\mu}{\sigma^2}\bar{x} - \frac{n}{2\sigma^2}\mu^2 - \frac{n}{2\sigma^2}\bar{x}^2 + \dots\right),$$

which establishes MLR in  $T(\bar{x}) = \bar{x}$ . Therefore UMP level  $\alpha$  test is

$$\phi(\mathbf{x}) = \begin{cases} 1 & \text{if } \bar{x} > C; \\ \gamma & \text{if } \bar{x} = C; \\ 0 & \text{if } \bar{x} < C. \end{cases}$$

Choice of  $\gamma$  is not needed since  $P_\mu(\bar{X} = C) = 0$ . We need

$$\alpha = P_{\mu_0}(\bar{X} > C) = P_{\mu_0}\left(\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} > \frac{C - \mu_0}{\sigma/\sqrt{n}}\right),$$

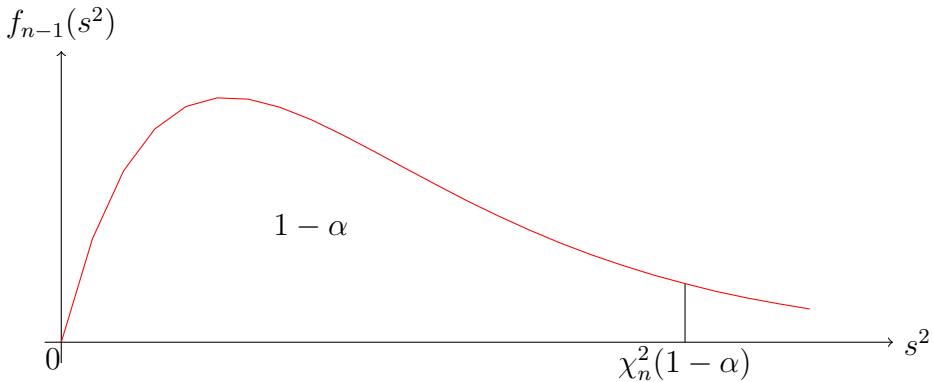
so that  $\frac{C - \mu_0}{\sigma/\sqrt{n}} = z_{1-\alpha}$  or  $C = \mu_0 + z_{1-\alpha} \frac{\sigma}{\sqrt{n}}$ . Thus, the UMP test simply rejects  $H_0$  if  $\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} > z_{1-\alpha}$ . This is simply the standard  $Z$ -test.

**Example.**  $X_1, \dots, X_n$  i.i.d  $N(\mu_0, \sigma^2)$ ,  $\mu_0$  known. Test  $H_0 : \sigma^2 \leq \sigma_0^2$  versus  $H_1 : \sigma^2 > \sigma_0^2$ .  $S^2 = \sum_{i=1}^n (X_i - \mu_0)^2$  is sufficient for  $\sigma^2$  and  $S^2 = \sum_{i=1}^n (X_i - \mu_0)^2 / \sigma^2 \sim \chi_n^2$ .

$$\begin{aligned} f_{S^2}(s^2 | \sigma^2) &= \exp\left(-\frac{s^2}{2\sigma^2}\right) \left(\frac{s^2}{2\sigma^2}\right)^{n/2-1} \times \text{constant} \\ &= \exp\left(-\frac{1}{2\sigma^2}s^2 - \frac{n}{2}\log(\sigma^2) + (\frac{n}{2} - 1)\log(s^2) + \dots\right). \end{aligned}$$

Thus we have a one-parameter exponential family with MLR in  $T(s^2) = s^2$ . Therefore the UMP level  $\alpha$  test is

$$\phi(\mathbf{x}) = \begin{cases} 1 & \text{if } s^2 > C; \\ 0 & \text{if } s^2 < C. \end{cases}$$



$C$  is determined by

$$\alpha = P_{\sigma_0^2} (S^2 > C) = P_{\sigma_0^2} \left( \frac{S^2}{\sigma_0^2} > \frac{C}{\sigma_0^2} \right).$$

Since  $S^2/\sigma_0^2 \sim \chi_n^2$ , when  $\sigma^2 = \sigma_0^2$ , we have that  $C/\sigma_0^2 = \chi_n^2(1 - \alpha)$