

## More on Hypothesis Testing, Local Power, and Unbiased Test

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**Example 1.** Suppose  $X$  follows a Poisson distribution,  $p_\theta(x) = \theta e^{-\theta x}$ . We want to test  $H_0 : \theta = 1$  versus  $H_1 : \theta = \theta_1 > 1$ . Then we perform a *likelihood ration test*,

$$\phi(x) = \begin{cases} 1 & p_1(x)/p_0(x) = \theta_1 e^{-\theta_1 x}/e^{-x} > k \\ 0 & \text{otherwise} \end{cases}$$

Equivalently,  $\phi = 1$  when  $x < \frac{\log(\theta_1/k)}{\theta_1 - 1} = k'$ .  $k'$  can be set such that level

$$\alpha = P_0(X < k') = \int_0^{k'} e^{-x} dx = 1 - e^{-k'},$$

so that  $k' = -\log(1 - \alpha)$ . Therefore,

$$\phi(x) = \begin{cases} 1 & x < -\log(1 - \alpha) \\ 0 & \text{otherwise} \end{cases}$$

## 1 Uniformly Most Powerful (UMP) Tests

In short, a test  $\phi^*$  is UMP if  $\mathbb{E}_\theta \phi^* \geq \mathbb{E}_\theta \phi, \forall \theta \in \Omega$  for  $\phi^*$  and  $\phi$  having level  $\alpha$ .

**Definition 2.** A family  $p_\theta(x), \theta \in \Omega \subseteq \mathbb{R}$  has monotone likelihood ratios if there exists a statistic  $T$  such that whenever  $\theta_1 < \theta_2$ , the ratio  $p_{\theta_2}(x)/p_{\theta_1}(x)$  is monotone (non-decreasing) in  $T$ .

**Example 3.** Exponential family random variable  $X$  has  $p_\theta(x) = h(x) \exp(\eta(\theta)T(x) - B(\theta))$  with  $\eta(\cdot)$  strictly increasing. Then for  $\theta_2 > \theta_1$ ,

$$\frac{p_{\theta_2}(x)}{p_{\theta_1}(x)} = \exp\{(\eta(\theta_2) - \eta(\theta_1))T(x) + B(\theta_1) - B(\theta_2)\}$$

is non-decreasing in  $T(x)$  and therefore has monotone likelihood ratios.

**Theorem 4.** (12.9 in Keener) Suppose that  $p_\theta$  has monotone likelihood ratios, then

1 The test  $\phi^*$  given by

$$\phi^*(x) = \begin{cases} 1 & T(x) > c \\ \gamma & T(x) = c \\ 0 & T(x) < c \end{cases}$$

is UMP for  $H_0 : \theta \leq \theta_0$  versus  $H_1 : \theta > \theta_0$ .

2 The power function  $\beta(\theta) = \mathbb{E}_\theta \phi^*$  is non-decreasing and strictly increasing for  $\beta(\theta) \in (0, 1)$ .

3 If  $\theta_1 \leq \theta_0$ , then  $\phi^*$  minimize  $\mathbb{E}_{\theta_1} \phi$  among all tests such that  $\mathbb{E}_{\theta_0} \phi = \mathbb{E}_{\theta_0} \phi^* = \alpha$ .

**Example 5.**  $X_i$ 's are iid with  $U(0, \theta)$ .  $M(x) = \min\{x_1, \dots, x_n\}$  and  $T(x) = \max\{x_1, \dots, x_n\}$ .

$$p_\theta(x) = \begin{cases} 1/\theta^n & M(x) > 0, T(x) < \theta \\ 0 & \text{otherwise} \end{cases}$$

If  $\theta_2 > \theta_1$ , then

$$\frac{p_{\theta_2}(x)}{p_{\theta_1}(x)} = \begin{cases} \theta_1^n/\theta_2^n & T(x) < \theta_1 \\ 0 & \text{otherwise} \end{cases}$$

. Thus, it has monotone likelihood ratios.  $H_0 : \theta \leq 1$  versus  $H_1 : \theta > 1$ . Follow by the Theorem,

$$\phi^*(x) = \begin{cases} 1 & T(x) \geq c \\ 0 & T(x) < c \end{cases}$$

is UMP. The level is  $P_1(T \geq c) = 1 - c^n$  is set to  $\alpha$ . Thus,  $c = (1 - \alpha)^{1/n}$ . The power of the test

$$\beta_\phi(\theta) = P_\theta(T \geq c) = \begin{cases} 0 & \theta < c \\ 1 - (\frac{c}{\theta})^n = 1 - \frac{1-\alpha}{\theta^n} & \text{otherwise} \end{cases}$$

## 2 Local Power

*Local Power* is a popular concept. We first assume enough regularity such that

$$\beta'(\theta) = \int \phi(x) \frac{\partial p_\theta(x)}{\partial \theta} \mu(dx).$$

A test  $\phi^*$  is locally most powerful (LMP) for  $H_0 : \theta = \theta_0$  versus  $H_1 : \theta > \theta_0$  if it maximizes  $\beta'_\phi(\theta_0)$  among all tests with level  $\alpha$ . The major characterization for LMP test is that the test should maximize  $\int \phi(x) \frac{\partial \log p_\theta(x)}{\partial \theta} p_\theta(x) \mu(dx)$  subject to  $\int \phi p_\theta d\mu = \alpha$ . Therefore, the test is in the form

$$\phi^*(x) = \begin{cases} 1 & \frac{\partial \log p_\theta(x)}{\partial \theta} > k \\ 0 & \frac{\partial \log p_\theta(x)}{\partial \theta} < k \end{cases}$$

**Example 6.**  $X_i$ 's iid has density  $f_\theta(x)$ , then  $\log p_\theta(x) = \sum_i \log f_\theta(x_i)$ . From CLT,

$$\frac{1}{\sqrt{n}} \frac{\partial \log p_\theta(x)}{\partial \theta} \Rightarrow N(0, I(\theta)).$$

Level of  $\phi^*$  is approximately

$$p_{\theta_0}(\frac{1}{\sqrt{n}} \frac{\partial \log p_\theta(x)}{\partial \theta} < \frac{k}{\sqrt{n}}) \approx 1 - \Phi(\frac{k}{\sqrt{nI(\theta_0)}}),$$

which is  $\alpha$  if  $k = z_\alpha \sqrt{nI(\theta_0)}$ .

## 3 Unbiased Tests

**Definition 7.** A test is *unbiased* if the test has power  $\beta_\phi(\theta) \leq \alpha, \forall \theta \in \Omega_0$  and  $\beta_\phi(\theta) \geq \alpha, \forall \theta \in \Omega_1$ .

**Theorem 8.** (12.26 in Keener) For  $\alpha \in (0, 1)$ , let  $\theta_0 \in \Omega^0$ ,  $X$  is exponential family random variable with  $\eta$  differentiable and  $0 < \eta'(\theta_0) < \infty$ . Then there exist a two-sided, level  $\alpha$  test  $\phi^*$  with  $\beta'_{\phi^*}(\theta_0) = 0$  and any such test is UMP for  $H_0 : \theta = \theta_0$  versus  $H_1 : \theta \neq \theta_0$  among unbiased tests at level  $\alpha$ .