MA 3140: Statistical Inference

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UMP Unbiased tests

Unbiasedness for Hypothesis Testing

Note that in the class Φ_{α} of all size α tests, i.e.,

$$\sup_{\theta \in \Theta_0} \beta_{\phi}^*(\theta) = \sup_{\theta \in \Theta_0} E_{\theta} \phi(\mathbf{X}) = \alpha,$$

there does not exist UMP test for many hypotheses.

A size α test ϕ is said to be unbiased if the power function β_{ϕ}^* satisfies the condition

$$eta_{\phi}^* \le \alpha, \quad \text{for } \theta \in \Theta_0$$
and $eta_{\phi}^* \ge \alpha, \quad \text{for } \theta \in \Theta_1.$ (1)

▶ Thus, an unbiased test rejects a false H_0 more often than a true H_0 .

Unbiasedness for Hypothesis Testing

Let U_{α} be the class of all unbiased size- α tests of H_0 . If there exists a test $\phi \in U_{\alpha}$ that has maximum power at each $\theta \in \Theta_1$, it is referred to as UMP-unbiased size α test.

ightharpoonup Clearly, $U_{\alpha} \subset \Phi_{\alpha}$.

If a UMP test exists in Φ_{α} , it is UMP in U_{α} .

This follows on comparing the power of the UMP test with that of the trivial test $\phi(x) \equiv \alpha$.

Theorem 1

Let $\mathbf{X} = (X_1, \dots, X_n)$ be a random vector with pdf $f(\mathbf{x}, \theta) = c(\theta)e^{\theta T(\mathbf{x})}h(\mathbf{x}), \ \theta \in \Theta \subset \mathbb{R}.$

Then for testing

$$H_3: \theta_1 \leq \theta \leq \theta_2$$
 vs $K_3: \theta < \theta_1$ or $\theta > \theta_2$

there exists a UMP unbiased test given by

$$\phi(\mathbf{x}) = \begin{cases} 1, & \text{if } T(\mathbf{x}) < c_1 \text{ or } T(\mathbf{x}) > c_2 \\ \gamma_i, & \text{if } T(\mathbf{x}) = c_i, \ i = 1, 2 \\ 0, & \text{if } c_1 < T(\mathbf{x}) < c_2 \quad (c_1 < c_2). \end{cases}$$

where c_1 , c_2 , γ_1 γ_2 are determined by

$$E_{\theta_1}\phi(\mathbf{X}) = E_{\theta_2}\phi(\mathbf{X}) = \alpha.$$

Theorem 2

Let $\mathbf{X} = (X_1, \dots, X_n)$ be a random vector with pdf $f(\mathbf{x}, \theta) = c(\theta)e^{\theta T(\mathbf{x})}h(\mathbf{x}), \ \theta \in \Theta \subset \mathbb{R}.$

Then for testing

$$H_4: \theta = \theta_0$$
 vs $K_4: \theta \neq \theta_0$

there exists a UMP unbiased test given by

$$\phi(\mathbf{x}) = \begin{cases} 1, & \text{if } T(\mathbf{x}) < c_1 \text{ or } T(\mathbf{x}) > c_2 \\ \gamma_i, & \text{if } T(\mathbf{x}) = c_i, \ i = 1, 2 \\ 0, & \text{if } c_1 < T(\mathbf{x}) < c_2 \quad (c_1 < c_2). \end{cases}$$

where c_i 's and γ_i 's are determined by

$$E_{\theta_0}\phi(\mathbf{X}) = \alpha; \quad E_{\theta_0}(T(\mathbf{X})\phi(\mathbf{X})) = \alpha E_{\theta_0}T(\mathbf{X}).$$

Example 1

Let $X \sim Bin(n, p)$. We want to test

$$H_4: p = p_0$$
 vs $K_4: p \neq p_0$.

Solution: The pmf

$$f(x,p) = \binom{n}{x} p^{x} (1-p)^{n-x} = \binom{n}{x} (1-p)^{n} e^{x \log \frac{p}{1-p}}$$

belongs to the exponential family with T(x) = x.

The test is

$$\phi(x) = \begin{cases} 1, & \text{if } x < c_1 \text{ or } x > c_2 \\ \gamma_i, & \text{if } x = c_i, \ i = 1, 2 \\ 0, & \text{if } c_1 < x < c_2. \end{cases}$$

The constants c_i 's and γ_i 's are determined by

$$E_{p_0}\phi(X) = \alpha; \tag{2}$$

$$E_{p_0}\{X\phi(X)\} = \alpha E_{p_0}(X). \tag{3}$$

Condition (2) can be written as

$$P_{p_0}(X < c_1 \text{ or } X > c_2) + \gamma_1 P_{p_0}(X = c_1) + \gamma_2 P_{p_0}(X = c_2) = \alpha$$

$$\iff P(c_1 < X < c_2) + (1 - \gamma_1) P(X = c_1) + (1 - \gamma_2) P(X = c_2) = 1 - \alpha$$

$$\iff \sum_{x=c_1+1}^{c_2-1} \binom{n}{x} p_0^x (1-p_0)^{n-x}$$

$$+\sum_{i=1}^{2}(1-\gamma_{i})\binom{n}{c_{i}}p_{0}^{c_{i}}(1-p_{0})^{n-c_{i}}=1-\alpha$$

The LHS can be determined from the tables of Binomial distribution

Condition (3) can be written as

$$E_{p_0}X(1-\phi(X)) = (1-\alpha)E_{p_0}X = (1-\alpha)np_0$$

$$\iff \sum_{x=c_1+1}^{c_2-1} x \binom{n}{x} p_0^x (1-p_0)^{n-x}$$

$$+ \sum_{i=1}^2 (1-\gamma_i)c_i \binom{n}{c_i} p_0^{c_i} (1-p_0)^{n-c_i} = (1-\alpha)np_0$$

$$\iff \sum_{x=c_1+1}^{c_2-1} \binom{n-1}{x-1} p_0^{x-1} (1-p_0)^{n-1-(x-1)}$$

$$+ \sum_{i=1}^2 (1-\gamma_i) \binom{n-1}{c_i-1} p_0^{c_i-1} (1-p_0)^{n-1-(c_i-1)} = 1-\alpha$$

on using

$$x \binom{n}{x} p_0^x (1-p_0)^{n-x} = n p_0 \binom{n-1}{x-1} p_0^{x-1} (1-p_0)^{n-1-(x-1)}.$$

The LHS can be determined from the tables of Binomial distribution.

▶ Consider $n = 10, p_0 = 1/2$. Then the first condition is:

$$\sum_{x=c_1+1}^{c_2-1} {10 \choose x} \left(\frac{1}{2}\right)^{10} + \sum_{i=1}^{2} (1-\gamma_i) {10 \choose c_i} \left(\frac{1}{2}\right)^{10} = 0.9$$

$$\sum_{x=c_1+1}^{c_2-1} {10 \choose x} + \sum_{i=1}^{2} (1-\gamma_i) {10 \choose c_i} = 2^{10} \times 0.9,$$

and the second condition is:

$$\sum_{x=c_1+1}^{c_2-1} {9 \choose x-1} \left(\frac{1}{2}\right)^9 + \sum_{i=1}^2 (1-\gamma_i) {9 \choose c_i-1} \left(\frac{1}{2}\right)^9 = 0.9$$

$$\sum_{x=c_1}^{c_2-2} {9 \choose y} + \sum_{i=1}^2 (1-\gamma_i) {9 \choose c_i-1} = 2^9 \times 0.9$$

These conditions can now be solved using binomial coefficients to get values of c_1 , c_2 , γ_1 and γ_2 .

Example 2

Let $X_1, \ldots, X_n \sim N(0, \sigma^2)$. Find UMP unbiased test for

$$H_4: \sigma^2 = \sigma_0^2$$
 vs $K_4: \sigma^2 \neq \sigma_0^2$.

Solution: The pmf

$$f(\mathbf{x},\sigma) = \frac{1}{(\sigma\sqrt{2\pi})^n} \exp^{-\sum x_i^2/2\sigma^2}$$

belongs to the exponential family with $T(x) = \sum x_i^2$. The UMP unbiased test is

$$\phi(\mathbf{x}) = \begin{cases} 1, & \text{if } \sum x_i^2 < c_1 \text{ or } \sum x_i^2 > c_2 \\ 0, & \text{if } c_1 \le \sum x_i^2 \le c_2. \end{cases}$$

Here, γ_i 's are 0 as X_i 's are constant. Also, recall that $\sum X_i^2/\sigma^2 \sim \chi_n^2$.

Thus, the test becomes

$$\phi(\mathbf{x}) = egin{cases} 1, & ext{if } rac{\sum \chi_i^2}{\sigma_0^2} < c_1 ext{ or } rac{\sum \chi_i^2}{\sigma_0^2} > c_2 \ 0, & ext{if } c_1 \leq rac{\sum \chi_i^2}{\sigma_0^2} \leq c_2. \end{cases}$$

where c_i 's are determined by

$$E_{\sigma_0}\phi(\mathbf{X}) = \alpha; \tag{4}$$

$$E_{\sigma_0} \frac{T(\mathbf{X})}{\sigma_0^2} \phi(\mathbf{X}) = \alpha \ E_{\sigma_0} \frac{T(\mathbf{X})}{\sigma_0^2}. \tag{5}$$

Here,

$$E_{\sigma_0}(1-\phi(\mathbf{X})) = 1-\alpha \Longrightarrow P(c_1 \le W \le c_2) = 1-\alpha, \ W \sim \chi_n^2$$

Also,

$$E_{\sigma_0} \frac{T(\mathbf{X})}{\sigma_0^2} (1 - \phi(\mathbf{X})) = (1 - \alpha) E_{\sigma_0} \frac{T(\mathbf{X})}{\sigma_0^2}$$

$$\iff E_{\sigma_0} W (1 - \phi(\mathbf{X})) = (1 - \alpha) E_{\sigma_0} W$$

$$\iff \int_{c_1}^{c_2} w g_n(w) dw = (1 - \alpha) n$$

$$\iff \int_{c_1}^{c_2} g_{n+2}(w) dw = 1 - \alpha,$$

since $w g_n = w \frac{1}{2^{n/2} \Gamma n/2} \exp^{-w/2} w^{n/2-1} = n g_{n+2}$.

So, the 2 conditions are

$$\int_{c_1}^{c_2} g_n(w) dw = 1 - \alpha \text{ and } \int_{c_1}^{c_2} g_{n+2}(w) dw = 1 - \alpha.$$

Integrating by parts in the second condition and use the first condition, we get

$$c_1^{n/2}e^{-c_1/2}=c_2^{n/2}e^{-c_2/2}.$$

These values of c_1 and c_2 are tabulated by Pacheres (1961, AMS).

If n is large and σ_0 is not close to 0 or ∞ , we can take an approximate test as

$$c_1 = \chi^2_{n,1-\alpha/2}$$
 $c_2 = \chi^2_{n,\alpha/2}$.

Thanks for your patience!