Lecture 15: Uniformly Most Powerful Tests

Mathematical Statistics II, MATH 60062/70062

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Reference: Casella & Berger, 8.3.2

Recap: Monotone likelihood ratio

A family of PDFs/PMFs $\{g_T(t \mid \theta) : \theta \in \Theta\}$ for a univariate random variable T has a **monotone likelihood ratio (MLR)** if for all $\theta_2 > \theta_1$, the ratio

$$\frac{g_T(t\mid\theta_2)}{g_T(t\mid\theta_1)}$$

is a nondecreasing function of t on $\{t: g_T(t \mid \theta_1) > 0 \text{ or } g_T(t \mid \theta_2) > 0\}.$

Note: If $T \sim g_T(t \mid \theta) = h(t)c(\theta)e^{w(\theta)t}$ and $w(\theta)$ is a nondecreasing function of θ , then $\{g_T(t \mid \theta) : \theta \in \Theta\}$ has an MLR.

Recap: Karlin-Rubin Theorem

Consider testing

$$H_0: \theta \leq \theta_0$$
 versus $H_1: \theta > \theta_0$.

Suppose that T is a sufficient statistic for θ and the family $\{g_T(t\mid\theta):\theta\in\Theta\}$ has an MLR. Then the test that rejects H_0 if and only if $T>t_0$ is a UMP level α test, where

$$\alpha = P_{\theta_0}(T > t_0).$$

Similarly, when testing

$$H_0: \theta \geq \theta_0$$
 versus $H_1: \theta < \theta_0$,

the test that rejects H_0 if and only if $T < t_0$ is a UMP level α test, where $\alpha = P_{\theta_0}(T < t_0)$.

Bernoulli/Binomial UMP test

Suppose X_1, \ldots, X_n are iid $\mathrm{Bern}(\theta)$, where $0 < \theta < 1$. Consider testing

$$H_0: \theta \leq \theta_0 \quad \text{versus} \quad H_1: \theta > \theta_0.$$

We know that

$$T = \sum_{i=1}^{n} X_i$$

is sufficient for θ and $T \sim \text{Bin}(n, \theta)$, and $\{g_T(t \mid \theta) : 0 < \theta < 1\}$ has an MLR.

Therefore, by the Karlin-Rubin Theorem the UMP level α test is

$$\phi(t) = I(t > t_0),$$

where t_0 satisfies

$$\alpha = P_{\theta_0}(T > t_0) = \sum_{t=|t_0|+1}^n \binom{n}{t} \theta_0^t (1 - \theta_0)^{n-t}.$$

Normal UMP test

Suppose X_1, \ldots, X_n are iid $\mathcal{N}(\mu, \sigma_0^2)$, where $-\infty < \mu < \infty$ and σ_0^2 is known. Consider testing

$$H_0: \mu \leq \mu_0$$
 versus $H_1: \mu > \mu_0$.

We know that $T(\boldsymbol{X}) = \bar{X}$ is sufficient for μ and $T \sim \mathcal{N}(\mu, \sigma_0^2/n)$, and $\{g_T(t \mid \mu) : -\infty < \mu < \infty\}$ has an MLR (exercise).

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By the Karlin-Rubin Theorem, the UMP level α test is

$$\phi(t) = I(t > t_0),$$

where t_0 satisfies

$$\alpha = P_{\mu_0}(T > t_0) = P\left(\frac{\bar{X} - \mu_0}{\sigma_0/\sqrt{n}} > \frac{t_0 - \mu_0}{\sigma_0/\sqrt{n}}\right) = 1 - F_Z\left(\frac{t_0 - \mu_0}{\sigma_0/\sqrt{n}}\right)$$

$$\implies t_0 = \frac{z_\alpha \sigma_0}{\sqrt{n}} + \mu_0.$$

Thus, the UMP level α test function for H_0 versus H_1 is

$$\phi(\boldsymbol{x}) = I\left(\bar{x} > \frac{z_{\alpha}\sigma_0}{\sqrt{n}} + \mu_0\right)$$

and the corresponding power function is

$$\beta(\mu) = E_{\mu}[\phi(\boldsymbol{X})]$$

$$= P_{\mu} \left(\bar{X} > \frac{z_{\alpha}\sigma_{0}}{\sqrt{n}} + \mu_{0} \right)$$

$$= P_{\mu} \left(\frac{\bar{X} - \mu}{\sigma_{0}/\sqrt{n}} > z_{\alpha} + \frac{\mu_{0} - \mu}{\sigma_{0}/\sqrt{n}} \right)$$

$$= 1 - F_{Z} \left(z_{\alpha} + \frac{\mu_{0} - \mu}{\sigma_{0}/\sqrt{n}} \right).$$

Nonexistence of UMP test

Using the Karlin-Rubin Theorem, we can find UMP level α tests for

$$H_0: \theta \leq \theta_0$$
 versus $H_1: \theta > \theta_0$.

or

$$H_0: \theta \geq \theta_0$$
 versus $H_1: \theta < \theta_0$.

Unfortunately, with a *two-sided* H_1 ($H_1: \theta \neq \theta_0$), UMP tests do not exist.

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Suppose X_1,\ldots,X_n are iid $\mathcal{N}(\mu,\sigma_0^2)$, where $-\infty<\mu<\infty$ and σ_0^2 is known. Consider testing

$$H_0: \mu = \mu_0$$
 versus $H_1: \mu \neq \mu_0$.

If there exists a UMP test, then for all $\mu \neq \mu_0$ the power function of the test should be greater than the power function of any other level α test.

Nonexistence of UMP test

It is possible to find UMP tests when H_1 is *one-sided*.

• For $H_0': \mu \leq \mu_0$ versus $H_1': \mu > \mu_0$, the UMP level α test function is

$$\phi'(\mathbf{x}) = I\left(\bar{x} > \frac{z_{\alpha}\sigma_0}{\sqrt{n}} + \mu_0\right)$$

and the corresponding power function is

$$\beta'(\mu) = 1 - F_Z \left(z_\alpha + \frac{\mu_0 - \mu}{\sigma_0 / \sqrt{n}} \right).$$

• For $H_0'': \mu \ge \mu_0$ versus $H_1'': \mu < \mu_0$, the UMP level α test function is

$$\phi''(x) = I\left(\bar{x} < -\frac{z_{\alpha}\sigma_0}{\sqrt{n}} + \mu_0\right)$$

and the corresponding power function is

$$\beta''(\mu) = F_Z \left(-z_\alpha + \frac{\mu_0 - \mu}{\sigma_0 / \sqrt{n}} \right).$$

Note that both are also size (and level) α tests for H_0 versus H_1 because

$$\sup_{\mu \in \Theta_0} \beta'(\mu) = \beta'(\mu_0) = 1 - F_Z(z_\alpha) = \alpha$$

and

$$\sup_{\mu \in \Theta_0} \beta''(\mu) = \beta''(\mu_0) = F_Z(-z_\alpha) = \alpha.$$

Therefore,

- $\phi'(x)$ is UMP level α when $\mu > \mu_0$
- $\phi''(x)$ is UMP level α when $\mu < \mu_0$.

Since $\phi'(x) \neq \phi''(x)$ for all $x \in \mathcal{X}$, no UMP test exists for H_0 versus H_1 .

Unbiased tests

When no UMP level α test within the class of all tests, we could further restrict our attention to a smaller class, the class of unbiased tests.

Consider testing

$$H_0: \theta \in \Theta_0$$
 versus $H_1: \theta \in \Theta_0^c$.

A test with power function $\beta(\theta)$ is **unbiased** if $\beta(\theta_1) \geq \beta(\theta_0)$ for all $\theta_1 \in \Theta_0^c$ and for all $\theta_0 \in \Theta_0$. In other words, the power is always larger in the alternative parameter space than it is in the null parameter space.

Uniformly most powerful unbiased tests

The uniformly most powerful unbiased (UMPU) level α test has power function that satisfies

$$\beta(\theta) \ge \beta^*(\theta)$$

for all $\theta \in \Theta_0^c$, where $\beta^*(\theta)$ is the power function of any other unbiased level α test.

Suppose X_1,\dots,X_n are iid $\mathcal{N}(\mu,\sigma_0^2)$, where $-\infty<\mu<\infty$ and σ_0^2 is known. Consider testing

$$H_0: \mu = \mu_0$$
 versus $H_1: \mu \neq \mu_0$.

The UMPU level α test rejects H_0 if and only if

$$\frac{\bar{x} - \mu_0}{\sigma_0/\sqrt{n}} < -z_{\alpha/2} \quad \text{or} \quad \frac{\bar{x} - \mu_0}{\sigma_0/\sqrt{n}} > z_{\alpha/2}$$

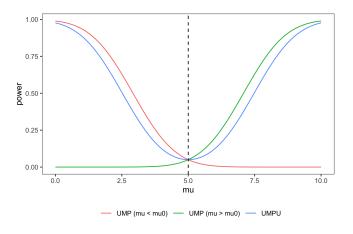
The rejection region of the UMPU level α test is

$$R = \left\{ \boldsymbol{x} \in \mathcal{X} : \left| \frac{\bar{x} - \mu_0}{\sigma_0 / \sqrt{n}} \right| > z_{\alpha/2} \right\}.$$

The power function of the test is

$$\begin{split} \beta(\mu) &= P_{\mu}(\boldsymbol{X} \in R) \\ &= P_{\mu}\left(\frac{\bar{X} - \mu_0}{\sigma_0/\sqrt{n}} > z_{\alpha/2} \text{ or } \frac{\bar{X} - \mu_0}{\sigma_0/\sqrt{n}} < -z_{\alpha/2}\right) \\ &= P\left(Z > z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma_0/\sqrt{n}} \text{ or } Z < -z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma_0/\sqrt{n}}\right) \\ &= 1 - F_Z\left(z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma_0/\sqrt{n}}\right) + F_Z\left(-z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma_0/\sqrt{n}}\right). \end{split}$$

Power function of $\alpha=0.05$ test with parameters n=10, $\mu_0=5$, $\sigma_0^2=4$:



Appendix: Proof of Karlin-Rubin Theorem

Consider testing

$$H_0: \theta \leq \theta_0$$
 versus $H_1: \theta > \theta_0$.

Suppose that T is a sufficient statistic for θ and the family $\{g_T(t\mid\theta):\theta\in\Theta\}$ has an MLR. Then the test that rejects H_0 if and only if $T>t_0$ is a UMP level α test, where

$$\alpha = P_{\theta_0}(T > t_0).$$

Lemma 1: If both g(x) and h(x) are nondecreasing functions of x, then

$$Cov[g(X), h(X)] \ge 0.$$

Let X_1 and X_2 be iid with the same distribution as X. Then

$$E[(h(X_1) - h(X_2))(g(X_1) - g(X_2))]$$

$$= E[h(X_1)g(X_1)] - E[h(X_1)g(X_2)] - E[h(X_2)g(X_1)] + E[h(X_2)g(X_2)]$$

$$= \underbrace{E[h(X_1)g(X_1)] - E[h(X_1)]E[g(X_2)]}_{\text{Cov}[g(X),h(X)]} \underbrace{-E[h(X_2)]E[g(X_1)] + E[h(X_2)g(X_2)]}_{\text{Cov}[g(X),h(X)]}$$

Therefore

$$Cov[g(X), h(X)] = \frac{1}{2}E[(h(X_1) - h(X_2))(g(X_1) - g(X_2))] \ge 0.$$

Lemma 2: Suppose the family $\{g_T(t\mid\theta):\theta\in\Theta\}$ has an MLR. If $\phi(t)$ is a nondecreasing function of t, then $E_{\theta}[\phi(T)]$ is a nondecreasing function of θ .

Suppose $\theta_2 > \theta_1$. Because $\{g_T(t \mid \theta) : \theta \in \Theta\}$ has an MLR,

$$\frac{g_T(t \mid \theta_2)}{g_T(t \mid \theta_1)}$$

is a nondecreasing function of t for $\theta_2 > \theta_1$. By Lemma 1, we know

$$\operatorname{Cov}_{\theta_1}\left[\phi(T), \frac{g_T(T \mid \theta_2)}{g_T(T \mid \theta_1)}\right] \ge 0$$

$$\Longrightarrow \underbrace{E_{\theta_1}\left[\phi(T)\frac{g_T(T\mid\theta_2)}{g_T(T\mid\theta_1)}\right]}_{E_{\theta_2}[\phi(T)]} \ge E_{\theta_1}[\phi(T)]\underbrace{E_{\theta_1}\left[\frac{g_T(T\mid\theta_2)}{g_T(T\mid\theta_1)}\right]}_{1}$$

$$\implies E_{\theta_2}[\phi(T)] \ge E_{\theta_1}[\phi(T)].$$

Now, consider $\phi(t)=I(t>t_0)$, where t_0 is fixed. Clearly, $\phi(t)$ is a nondecreasing function of t. From Lemma 2, we know that

$$E_{\theta}[\phi(T)] = E_{\theta}[I(T > t_0)] = P_{\theta}(T > t_0)$$

is a nondecreasing function of θ .

Consider testing $H_0: \theta \leq \theta_0$ versus $H_1: \theta > \theta_0$, we have shown that the power function

$$\beta(\theta) = P_{\theta}(T > t_0)$$

is a nondecreasing function of θ .

In the Karlin-Rubin Theorem, the condition $\alpha=P_{\theta_0}(T>t_0)$ must be satisfied, where

$$\alpha = \sup_{\theta \le \theta_0} \beta(\theta) = \beta(\theta_0) = P_{\theta_0}(T > t_0).$$

This means that $\phi(t)=I(t>t_0)$ is a size α (and therefore level α) test function. All that remains is to show that this test is uniformly most powerful among level α tests.

Let $\phi^*(x)$ be any other level α test for

$$H_0: \theta \leq \theta_0$$
 versus $H_1: \theta > \theta_0$.

Fix $\theta_1 > \theta_0$ and consider testing

$$H_0^*:\theta=\theta_0\quad \text{versus}\quad H_1^*:\theta=\theta_1$$

instead. Because $\phi^*(\boldsymbol{x})$ is a level α test for H_0 versus H_1 ,

$$E_{\theta_0}[\phi^*(\boldsymbol{X})] \leq \sup_{\theta \leq \theta_0} E_{\theta}[\phi^*(\boldsymbol{X})] \leq \alpha.$$

This means that $\phi^*(x)$ is also a level α test for H_0^* versus H_1^* .

Consider the ratio

$$\frac{g_T(t \mid \theta_1)}{g_T(t \mid \theta_0)}$$

and define

$$k = \inf_{t \in \mathcal{T}} \frac{g_T(t \mid \theta_1)}{g_T(t \mid \theta_0)},$$

where $\mathcal{T} = \{t : t > t_0 \text{ and either } g_T(t \mid \theta_1) > 0 \text{ or } g_T(t \mid \theta_0) > 0\}$. It follows that

$$T > t_0 \iff \frac{g_T(t \mid \theta_1)}{g_T(t \mid \theta_0)} > k.$$

By the Neyman-Pearson Lemma with a sufficient statistic T, we know that $\phi(t)$ is the most powerful level α test for H_0^* versus H_1^* . That is

$$E_{\theta_1}[\phi(T)] \geq E_{\theta_1}[\phi^*(\boldsymbol{X})].$$

Because $\theta_1 > \theta_0$ was chosen arbitrarily,

$$E_{\theta}[\phi(T)] \geq E_{\theta}[\phi^*(\boldsymbol{X})]$$

holds for all $\theta > \theta_0$.