

## STAT 610: Discussion 5

### 1 Summary

- Hypothesis test: after observing samples  $\mathbf{X} \sim \mathbb{P}_\theta$ , decide either to retain  $H_0 : \theta \in \Theta_0$  (null hypothesis) or to reject  $H_0$  and conclude  $H_1 : \theta \in \Theta_0^c$  (alternative hypothesis) is true.
- Basic definitions:
  - Type I error: reject  $H_0$  if  $H_0$  is true.
  - Type II error: retain  $H_0$  if  $H_0$  is false.
  - Let  $R$  denote the rejection region for a test. Then

$$\beta(\theta) = \mathbb{P}_\theta(\mathbf{X} \in R) = \begin{cases} \mathbb{P}(\text{Type I error}) & \text{if } \theta \in \Theta_0 \\ 1 - \mathbb{P}(\text{Type II error}) & \text{if } \theta \in \Theta_0^c \end{cases}$$

is called the power function.

- A test with power function  $\beta(\theta)$  is a size (level)  $\alpha$  test if  $\sup_{\theta \in \Theta_0} \beta(\theta) = \alpha$  ( $\leq \alpha$ ).
- The likelihood ratio test statistic for  $H_0 : \theta \in \Theta_0$  v.s.  $H_1 : \theta \in \Theta_0^c$  is

$$\lambda(\mathbf{x}) = \frac{\sup_{\theta \in \Theta_0} L(\theta | \mathbf{x})}{\sup_{\theta \in \Theta} L(\theta | \mathbf{x})}.$$

A *likelihood ratio test* (LRT) is any test with rejection region  $\{\mathbf{x} : \lambda(\mathbf{x}) \leq c\}$ , where  $0 \leq c \leq 1$ . Normally, we choose  $c$  such that the test

$$T(\mathbf{x}) = \begin{cases} 1 & \text{if } \lambda(\mathbf{x}) \leq c \\ 0 & \text{if } \lambda(\mathbf{x}) > c \end{cases}$$

has size  $\alpha$ , that is,  $\sup_{\theta \in \Theta_0} \beta(\theta) = \sup_{\theta \in \Theta_0} \mathbb{P}_\theta(\lambda(\mathbf{x}) \leq c) = \alpha$ .

- **Theorem 8.2.4.** If  $W(\mathbf{X})$  is a sufficient statistic for  $\theta$  and  $\lambda^*(w)$  and  $\lambda(\mathbf{x})$  are the LRT statistic based on  $W$  and  $\mathbf{X}$ , then  $\lambda^*(W(\mathbf{x})) = \lambda(\mathbf{x})$ .
- A test  $T^*$  of size  $\alpha$  is a *uniformly most powerful* (UMP) test if and only if  $\beta_{T^*}(\theta) \geq \beta_T(\theta)$  for all  $\theta \in \Theta_0^c$  and  $T$  of level  $\alpha$ .
- **Theorem 8.3.12. (Neyman-Pearson Lemma)** Consider  $H_0 : \theta = \theta_0$  v.s.  $H_1 : \theta = \theta_1$ , then

$$T^*(\mathbf{x}) = \begin{cases} 1 & f(\mathbf{X} | \theta_1) > c f(\mathbf{X} | \theta_0) \\ 0 & f(\mathbf{X} | \theta_1) < c f(\mathbf{X} | \theta_0) \end{cases},$$

is a UMP test of size  $\alpha_c := E(T^* | \theta_0)$ . Note that  $f(\mathbf{X} | \theta_1) = c f(\mathbf{X} | \theta_0)$  can be arbitrary.

## 2 Questions

1. Suppose that we observe  $m$  i.i.d  $\text{Ber}(\theta)$  random variables, denoted by  $Y_1, \dots, Y_m$ . Show that the LRT of  $H_0 : \theta \leq \theta_0$  v.s.  $H_1 : \theta > \theta_0$  will reject  $H_0$  if  $\sum_{i=1}^m Y_i > b$ .

$$\{x \mid \lambda(x) \leq c\} \equiv \{\sum Y_i > b\}$$

2. A random sample  $X_1, \dots, X_n$  is drawn from a Pareto distribution with pdf

$$f(x \mid \theta, \nu) = \frac{\theta \nu^\theta}{x^{\theta+1}} I_{[\nu, \infty)}(x), \quad \theta > 0, \nu > 0.$$

- (a) Find the MLEs of  $\theta$  and  $\nu$ .  
(b) Show that the LRT of

$$H_0 : \theta = 1 \quad \text{v.s.} \quad H_1 : \theta \neq 1, \nu \text{ unknown},$$

has reject region of the form  $\{\mathbf{x} : T(\mathbf{x}) \leq c_1 \text{ or } T(\mathbf{x}) \geq c_2\}$ , where  $0 < c_1 < c_2$  and

$$\begin{aligned} T &= \log \left[ \frac{\prod_{i=1}^n X_i}{X_{(1)}^n} \right]. \\ \{M(\mathbf{x}) < c\} &\equiv \{g(T) < c\} \\ &\equiv \{T < c_1 \text{ or } T > c_2\}. \end{aligned}$$

3. Suppose  $X$  is one observation from a population with distribution  $\text{Beta}(\theta, 1)$ .

- (a) For testing  $H_0 : \theta \leq 1$  v.s.  $H_1 : \theta > 1$ , find the size and the power function of the test that rejects  $H_0$  if  $X > 0.5$ .  
(b) Find the most powerful level  $\alpha$  test of  $H_0 : \theta = 1$  v.s.  $H_1 : \theta = 2$ .

$$Q_1: Y_1, \dots, Y_m \sim \text{Ber}(\theta)$$

$$\begin{cases} H_1: \theta \leq \theta_0 \\ H_0: \theta > \theta_0 \end{cases}$$

$$\lambda(X) = \frac{\sup_{\theta \leq \theta_0} L(\theta | X)}{\sup_{\theta \in \Theta} L(\theta | X)}$$

$\hat{\theta} = \bar{X}$

Find the solution for numerator:

$$L(\theta | X) = \binom{m}{\sum Y_i} \theta^{\sum Y_i} (1-\theta)^{m-\sum Y_i}$$

$$\ell(\theta | X) = \log \binom{m}{\sum Y_i} + \sum Y_i \log \theta + (m - \sum Y_i) \log (1-\theta)$$

$$\ell'(\theta | X) = 0 + \frac{\sum Y_i}{\theta} - \frac{m - \sum Y_i}{1-\theta}$$

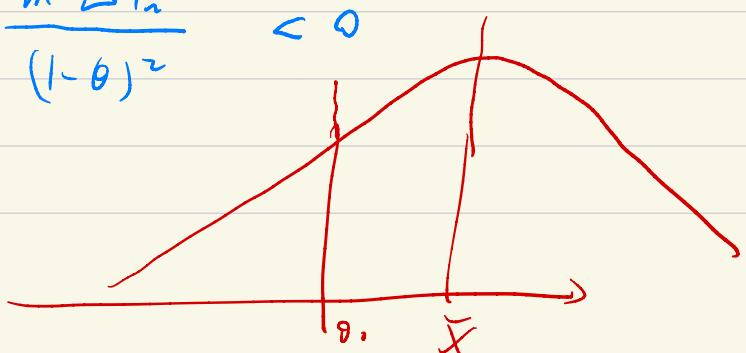
$$\text{Solve } \ell'(\theta | X) = 0 \Rightarrow \hat{\theta} = \bar{X}$$

But,  $\bar{X}$  may not in  $(-\infty, \theta_0)$

$$\ell''(\theta | X) = -\frac{\sum Y_i}{\theta^2} - \frac{m - \sum Y_i}{(1-\theta)^2} < 0$$

$\Rightarrow \ell$  is concave

so if  $\bar{X} > \theta_0$



$\Rightarrow \theta_0$  is the maximizer -

so in the numerator,  $\hat{\theta} = \min\{\bar{X}, \theta_0\}$

$$M(\bar{x}) = \frac{L(\hat{\theta} | \bar{y})}{L(\bar{x} | \bar{y})} = \frac{\binom{m}{\sum y_i} \hat{\theta}^{\sum y_i} (1-\hat{\theta})^{m-\sum y_i}}{\binom{m}{\sum y_i} \bar{x}^{\sum y_i} (1-\bar{x})^{m-\sum y_i}}$$

$$= \begin{cases} 1 & \text{if } \bar{x} < \theta_0 \\ \left(\frac{\theta_0}{\bar{x}}\right)^{\sum y_i} \left(\frac{1-\theta_0}{1-\bar{x}}\right)^{m-\sum y_i} & \bar{x} > \theta_0 \end{cases}$$

Rejection region.

$$\left\{ \left( \frac{\theta_0}{\bar{x}} \right)^{m\bar{x}} \left( \frac{1-\theta_0}{1-\bar{x}} \right)^{m(1-\bar{x})} < c \right\}$$

$\underbrace{\qquad}_{\text{II}}$

$g(\bar{x})$

If  $g(\bar{x}) \downarrow$

We have equivalent RR.

$$\Rightarrow \left\{ \bar{x} > b \right\}$$

$$g(t) = \underbrace{\exp \left\{ m t \left\{ \log \theta_0 - \log t \right\} + \ln(1-t) \left\{ \log(1-\theta_0) - \log(1-t) \right\} \right\}}$$

$$g'(t) = g(t) \times \left[ m \left( \log \theta_0 - \log t \right) + \cancel{m t \cdot \frac{1}{1-t}} \right]$$

$$- m \left( \log(1-\theta_0) - \log(1-t) \right) + \cancel{\ln(1-t)} \frac{1}{1-t}$$

$$= g(t) \left[ m \left\{ \log \frac{\theta_0}{t} - \log \frac{1-\theta_0}{1-t} \right\} \right]$$

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$$\Rightarrow \frac{\theta_0}{t} < 1 \Rightarrow \log \frac{\theta_0}{t} < 0$$

$$t > \theta_0$$

$$\Rightarrow \frac{1-\theta_0}{1-t} > 1 \Rightarrow \log \frac{1-\theta_0}{1-t} > 0$$

$$\text{So } g'(t) < 0$$

$\Rightarrow \{ \bar{X} > b \}$  is the R/Z  
for L/K/T  $\nexists$

$$Q2: f(x|\theta, v) = \frac{\theta v^\theta}{x^{\theta+1}} \mathbf{1}\{x > n\}$$

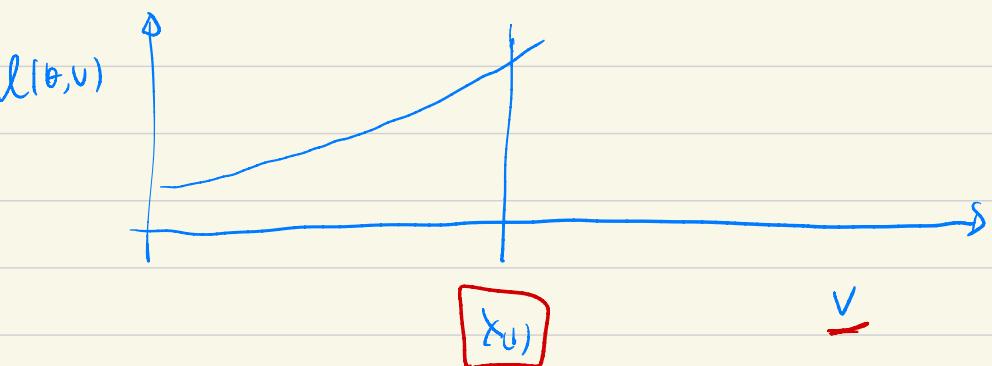
a) MLE

$$L(\theta, v) = \theta^n v^{n\theta} \left(\prod x_i\right)^{-\theta-1} \mathbf{1}\{x_{(1)} \geq n\}$$

Consider  $n < x_{(1)}$

$$\ell(\theta, v) = n \log \theta + n \theta \log v - (\theta+1) \log \prod x_i$$

$$\partial_v \ell(\theta, v) = 0 + \frac{n\theta}{v} + 0 > 0$$



$$\text{So } \hat{v} = x_{(1)}$$

$$\begin{aligned} \partial_\theta \ell(\theta, x_{(1)}) &= \frac{n}{\theta} + n \log x_{(1)} - \log \prod x_i \\ &= \frac{n}{\theta} - \boxed{\log \prod \frac{x_i}{x_{(1)}}} = T \end{aligned}$$

$$\text{Solve } \partial_\theta \ell(\theta, x_{(1)}) = 0$$

$$\hat{\theta} = \frac{n}{T}$$

$$\text{check for } \frac{\partial l(\theta, x_{(1)})}{\partial \theta} < 0$$

So we have the MLE  $(x_{(1)}, \frac{n}{T})$

b)  $L(X) = \frac{L(\theta=1, v=x_{(1)})}{L(\theta=\frac{n}{T}, v=x_{(1)})}$

$$= \frac{x_{(1)}^n (\pi x_{(1)})^{-2}}{\left(\frac{n}{T}\right)^n x_{(1)}^{\frac{n^2}{T}} (\pi x_{(1)})^{-\frac{n}{T}-1}}$$

$$= \frac{T^n}{n^n} x_{(1)}^{n(1-\frac{n}{T})} (\pi x_{(1)})^{-(1-\frac{n}{T})}$$

$$= \frac{T^n}{n^n} \left\{ \frac{\pi x_{(1)}}{x_{(1)}^n} \right\}^{-(1-\frac{n}{T})} e^T$$

$$= \frac{T^n}{n^n} e^{-T+n}$$

The KR for LKT

$$\left\{ \frac{T^n}{n!} e^{-T+n} < c \right\}$$

$$\equiv \left\{ \underbrace{\frac{T^n e^{-T}}{n!}}_{g(T)} < c \right\}$$

$g(T)$

$$g(0) = 0$$

$$\lim_{t \rightarrow \infty} g(t) = 0$$

$$g'(T) = n T^{n-1} e^{-T} - T^n e^{-T}$$

$$g'(T) = 0 \Rightarrow n T^{n-1} e^{-T} = T^n e^{-T}$$

$$\Rightarrow n T^{n-1} = T^n$$

$$\Rightarrow \boxed{n} = T$$

$\therefore$  we only have one solution.



So, the RK for LLT

$$\{ T < c_1, \quad T > c_2 \}$$

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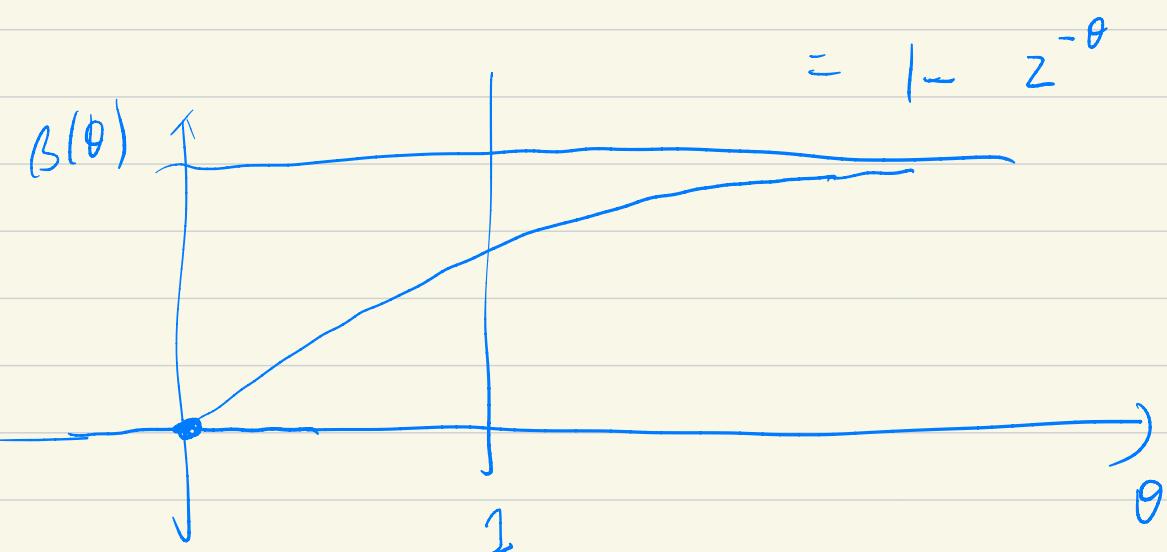
$$Q3: X \sim \text{Beta}(\theta, 1) \quad f_X(x) = \frac{1}{\text{Beta}(\theta, 1)} x^{\theta-1} \quad x \in [0, 1]$$

$$= \theta x^{\theta-1} \quad x \in [0, 1]$$

a)  $\{X > 0.5\}$

$$\beta(\theta) = P(X > 0.5 | \theta)$$

$$= \int_{0.5}^1 \theta x^{\theta-1} dx = x^\theta \Big|_{0.5}^1$$



$\Rightarrow$  Size is  $\beta(1) = 1 - \frac{1}{2} = \frac{1}{2}$

Power is  $1 - z^{-\theta}$  if

b) Apply N-P Lemma.

$\phi$  ::  $X$  is continuous

$\Rightarrow$  don't need to discuss  $f_1(x) = c f_0(x)$

$$f_1(x) = 2x$$

$$f_0(x) = 1$$

$$\text{RR} \quad \{ \quad f_1(x) > c f_0(x) \}$$

$$= \{ \quad 2x > c \quad \}$$

$$\Rightarrow \{ \quad x > c \quad \}$$

By N.P this is the RR for Unif-distr.

$$\alpha = P(X > c \mid \theta = 1)$$

$$= \int_c^1 1 dx = 1 - c$$

$$\text{So } c = 1 - \alpha$$

so the UMP test for  $\alpha$

is  $\{ X > 1 - \alpha \}$

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