

STAT 5430

Lec 22, F, Mar 14

Bay's theorem
Intro to testing → Homework 5 posted, due M, Mar 29
(aft'r break)

Hypothesis Testing I

finding best
test for
Simple H_0 vs
Simple
 H_1 !

Let $f(\underline{x}|\theta)$, $\underline{x} = (x_1, x_2, \dots, x_n)$, $\theta \in \Theta$, be the joint pdf/pmf of $\underline{X} = (X_1, \dots, X_n)$.

We want to test the hypothesis

- . $H_0 : \theta = \theta_0$ vs $H_1 : \theta = \theta_1$ where $\theta_0, \theta_1 \in \Theta$, $\theta_0 \neq \theta_1$.
- given parameters* *"best test"*

Definition: A test function $\varphi(\underline{x})$ is called a **most powerful** (MP) test of size α if

1. $E_{\theta_0} \varphi(\underline{X}) = \alpha$. *prob of Type I error at θ_0 is set to $\alpha \in [0, 1]$*
 $1 - E_{\theta_1} \bar{\varphi}(\underline{X}) \leq 1 - E_{\theta_0} \bar{\varphi}(\underline{X})$
2. $E_{\theta_1} \varphi(\underline{X}) \geq E_{\theta_1} \bar{\varphi}(\underline{X})$ holds for any other test rule $\bar{\varphi}(\underline{x})$ with $E_{\theta_0} \bar{\varphi}(\underline{X}) \leq \alpha$.
 \Rightarrow prob of Type II error at θ_1 or $1 - E_{\theta_1} \bar{\varphi}(\underline{X}) \leq$ prob of Type II error at θ_1 or $1 - E_{\theta_1} \bar{\varphi}(\underline{X})$

A MP test does exist at least for simple H_0 vs simple H_1 , as described below.

Theorem: (Neyman-Pearson Lemma) Let $f(\underline{x}|\theta)$, $\theta \in \Theta$, be the joint pdf/pmf of X_1, \dots, X_n . Then for testing $H_0 : \theta = \theta_0$ vs $H_1 : \theta = \theta_1$, a MP test of size α exists for all $\alpha \in [0, 1]$ and is given by

$$\varphi(\underline{x}) = \begin{cases} 1 & \text{if } f(\underline{x}|\theta_1) > k f(\underline{x}|\theta_0) \\ \gamma & \text{if } f(\underline{x}|\theta_1) = k f(\underline{x}|\theta_0) \\ 0 & \text{if } f(\underline{x}|\theta_1) < k f(\underline{x}|\theta_0) \end{cases}$$

$L(\theta) = f(\underline{x}|\theta)$
likelihood function

where $\gamma \in [0, 1]$ and $0 \leq k \leq \infty$ are constants satisfying

$$E_{\theta_0} \varphi(\underline{X}) = \alpha. \quad \leftarrow \text{pick } k \text{ & } \gamma \text{ to get size } \alpha \text{ at } \theta_0 \quad (5)$$

$\frac{L(\theta_1)}{L(\theta_0)}$	$>$	κ	$\text{reject } H_0 \text{ w prob } \frac{1}{\kappa}$]
$=$	κ	$\text{reject } H_0 \text{ w. prob } \gamma$		
$<$	κ	$\text{don't reject } H_0$		

*form
makes
MP
test!*

Hypothesis Testing I

Most Powerful Tests, cont'd

Remarks on Neyman-Pearson Lemma:

- Let $L(\theta) \equiv f(x|\theta)$, the likelihood function at θ . Then the MP test in (4) rejects H_0 for all x such that the likelihood $L(\theta_1)$ of θ_1 is “large” compared to the likelihood of $L(\theta_0)$ of θ_0 ; the “largeness” factor k is determined by (5).
- In general, the choice of (γ, k) satisfying (5) is not unique.
- If the distribution of $f(x|\theta_1)/f(x|\theta_0)$ under θ_0 is continuous, then we may set $\gamma = 0$.
*i.e. We usually only care concerned with r for discrete data!
(for continuous data, take r=0)*

Hypothesis Testing I

$$EX_1 = 3\theta$$

Illustration of Most Powerful Test (Continuous Case)

Example: Let X_1, \dots, X_n be a random sample from $\text{Gamma}(\alpha = 3, \theta)$, $\theta > 0$. Find a MP test of size α for $H_0 : \theta = \theta_0$ vs $H_1 : \theta = \theta_1$ (where $0 < \theta_0 < \theta_1$).

Solution: $f(\underline{x} | \theta) = \text{joint density of } \underline{x}$

$$= \prod_{i=1}^n \left(\frac{x_i^2 e^{-x_i/\theta}}{2\theta^3} \right) = \left(\prod_{i=1}^n x_i^2 \right) (2\theta^3)^{-n} e^{-\sum_{i=1}^n x_i/\theta}$$

By the theorem, the MP test of size is given by

$$\phi(\underline{x}) = \begin{cases} 1 & f(\underline{x} | \theta_1) > k f(\underline{x} | \theta_0) \\ \gamma & = \\ 0 & < \end{cases}$$

where $E_{\theta_0} \phi(\underline{x}) = \alpha$.

Note: $f(\underline{x} | \theta_1) \geq k f(\underline{x} | \theta_0)$

$$\Leftrightarrow (2\theta_1^3)^{-n} \left(\prod_{i=1}^n x_i^2 \right) e^{-\sum_{i=1}^n x_i/\theta_1} \stackrel{<}{\underset{>}{\gtrless}} k (2\theta_0^3)^{-n} \left(\prod_{i=1}^n x_i^2 \right) e^{-\sum_{i=1}^n x_i/\theta_0}$$

$$\Leftrightarrow -n \log(2\theta_1^3) - \sum_{i=1}^n x_i/\theta_1 \stackrel{<}{\underset{>}{\gtrless}} \log k + -n \log(2\theta_0^3) - \sum_{i=1}^n x_i/\theta_0$$

$$\Leftrightarrow \underbrace{\left(\sum_{i=1}^n x_i \right)}_{> 0} \left(\frac{1}{\theta_0} - \frac{1}{\theta_1} \right) \stackrel{<}{\underset{\equiv K_1}{\gtrless}} \underbrace{\log k - n \log(2\theta_0^3) + n \log(2\theta_1^3)}$$

since $\theta_0 < \theta_1$

$$\Leftrightarrow \left(\sum_{i=1}^n x_i \right) \stackrel{<}{\gtrless} \left(\frac{1}{\theta_0} - \frac{1}{\theta_1} \right) K_1 \equiv K_2$$

The MP test (of size α) is

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } \sum_{i=1}^n x_i > k_2 \\ r & \text{if } \sum_{i=1}^n x_i = k_2 \\ 0 & \text{if } \sum_{i=1}^n x_i < k_2 \end{cases}$$

where $0 \leq r \leq 1$ & k_2 are chosen so that $E_{\theta_0} \phi(\underline{x}) = \alpha$.

Note:

$$\alpha = E_{\theta_0} \phi(\underline{x}) = 1 \cdot P_{\theta_0} \left(\sum_{i=1}^n x_i > k_2 \right) + r P_{\theta_0} \left(\sum_{i=1}^n x_i = k_2 \right)$$

$$\sum_{i=1}^n x_i \stackrel{\text{since } \underline{x} \text{ are continuous}}{\sim} \text{gamma}(3n, \theta_0) = P_{\theta_0} \left(\sum_{i=1}^n x_i > k_2 \right)$$

$$= P_{\theta_0} \left(\frac{2 \sum_{i=1}^n x_i}{\theta_0} > \frac{2k_2}{\theta_0} \right)$$

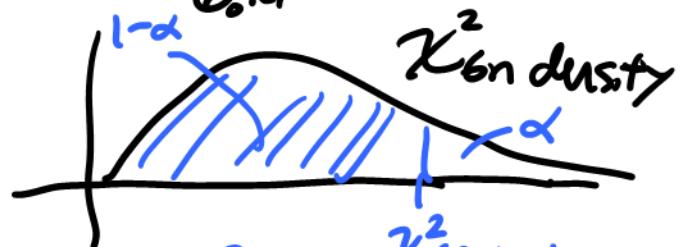
$$= P \left(\chi^2_{6n} > \frac{2k_2}{\theta_0} \right)$$

$\chi^2_{6n, 1-\alpha}$

$$\text{So, } k_2 = \frac{\theta_0}{2} \chi^2_{6n, 1-\alpha}$$

Note: $\frac{2x_i}{\theta_0} \sim \text{gamma}(3, 2)$
 $\frac{2 \sum_{i=1}^n x_i}{\theta_0} \sim \chi^2_{6n}$

$$\text{So } \frac{2 \sum_{i=1}^n x_i}{\theta_0} \sim \chi^2_{6n}$$



$$P(\chi^2_{6n} \leq k^2_{6n, 1-\alpha}) = 1 - \alpha$$

So the MP test (size α) of $H_0: \theta = \theta_0$ vs $H_1: \theta = \theta_1$

is $\phi(\underline{x}) = \begin{cases} 1 & \text{if } \sum_{i=1}^n x_i > \frac{\theta_0}{2} \chi^2_{6n, 1-\alpha} (\theta_1, \theta_0) \\ 0 & \text{otherwise} \end{cases}$