

MEM-264 Applied Statistics

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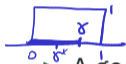
9th lecture - 09-03-2021

Comparing test functions : Power

$$\phi(x) = \begin{cases} 1, & t(x) \in C_\alpha^{(1)} \quad (\text{Reject } H_0 \text{ with probability } 1) \\ \gamma, & t(x) \in C_\alpha^{(\gamma)} \quad (\text{Reject } H_0 \text{ with probability } \gamma \in (0,1)) \\ 0, & \text{otherwise} \quad (\text{Reject } H_0 \text{ with probability } 0) \equiv (\text{Do not Reject } H_0) \end{cases}$$

Power function of a test ϕ

$$\Gamma \sim U[0,1] \quad \text{observation: } \gamma^* \quad w(\theta) = P\{\text{Reject } H_0\} = \mathbb{E}\{\phi(X)\}$$



► A good is one that makes $w(\theta)$ as large as possible on Θ_1 , while satisfying $w(\theta) \leq \alpha$ for all $\theta \in \Theta_0$.

$$P\{\text{Reject } H_0\} = P\{t(X) \in C_\alpha^{(1)}\} + P\{t(X) \in C_\alpha^{(\gamma)}\} \cdot \gamma$$

$$P\{\text{Reject } H_0\} = P\{t(X) \in C_\alpha^{(1)}\} + P\{t(X) \in C_\alpha^{(\gamma)}, \Gamma \leq \gamma\} =$$

$$= P\{t(X) \in C_\alpha^{(1)}\} + P\{t(X) \in C_\alpha^{(\gamma)}\} \underbrace{P\{\Gamma \leq \gamma\}} =$$

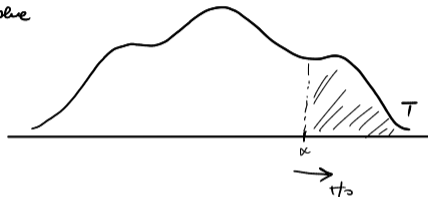
$$= 1 \cdot P\{t(X) \in C_\alpha^{(1)}\} + \gamma P\{t(X) \in C_\alpha^{(\gamma)}\} + 0 \cdot P\{t(X) \in (C_\alpha^{(1)} \cup C_\alpha^{(\gamma)})^c\}$$

$$= \mathbb{E}\{\phi(X)\}$$

- ▶ **Error of type I:** Reject H_0 , but $\theta \in \Theta_0$

For example

if H_0 : simple



- ▶ **Error of type II:** Do not reject H_0 , but $\theta \in \Theta_1$.

$$f(x; \theta_0) \quad \text{vs} \quad f(x; \theta_1)$$

$$H_0 : \theta = \theta_0 \quad \text{vs} \quad H_1 : \theta = \theta_1$$

We define the **likelihood ratio** $\Lambda(x)$ by

$$\Lambda(x) = \frac{f(x; \theta_1)}{f(x; \theta_0)}$$

Definition : Likelihood Ratio Test

$$\phi_0(x) = \begin{cases} 1, & \Lambda(x) > K \\ \gamma(x), & \Lambda(x) = K \\ 0, & \Lambda(x) < K \end{cases}$$

where $K \geq 0$ is a constant and $\gamma(x)$ a function such that $0 \leq \gamma(x) \leq 1$ for all x .

Neyman-Pearson theorem

(a) ► Optimality

For any K and $\gamma(x)$, the test ϕ_0 has maximum power among all test whose sizes are no greater than the size of ϕ_0 .

(b) ► Existence

Given $\alpha \in (0, 1)$, there exist constants K and γ_0 s.t the Likelihood Ratio Test has size exactly α .

(c) ► Uniqueness

If the test ϕ has size α , and is of maximum power amongst all possible tests of size α , then ϕ is a likelihood test, except possibly on a set of zero probability under both H_0 and H_1 .

Neyman-Pearson theorem : Proof

Size :

$$f(x; \theta_0)$$

$$\phi = \begin{cases} > 0 & \text{Reject with probability} \\ 0 & \text{Do not reject} \end{cases}$$

$$(K, \gamma(x))$$



*) $H_0: \theta = \theta_0$

$$\mathbb{E}_{\theta_0} \{ \phi \} = \alpha$$

Let ϕ be an arbitrary test function for which $\mathbb{E}_{\theta_1} \{ \phi(x) \} \leq \mathbb{E}_{\theta_0} \{ \phi_0(x) \}$ (*)

Let's define $V(x) = (\phi_0(x) - \phi(x)) (f(x; \theta_1) - K f(x; \theta_0))$

* if $\Lambda(x) > K \Leftrightarrow f(x; \theta_1) - K f(x; \theta_0) > 0$ then $\phi_0(x) = 1 \Rightarrow \phi_0(x) - \phi(x) \geq 0 \Rightarrow V(x) \geq 0$

* if $\Lambda(x) < K \Leftrightarrow f(x; \theta_1) - K f(x; \theta_0) < 0$ then $\phi_0(x) = 0 \Rightarrow \phi_0(x) - \phi(x) \leq 0 \Rightarrow V(x) \geq 0$

* if $\Lambda(x) = K \Rightarrow f(x; \theta_1) - K f(x; \theta_0) = 0$ then $V(x) = 0$

for any x we have $V(x) \geq 0$.

$$0 \leq \int_{-\infty}^{+\infty} V(x) dx = \int_{\mathbb{R}} \phi_0(x) f(x; \theta_1) dx - \int_{\mathbb{R}} \phi(x) f(x; \theta_1) dx + K \left\{ \int_{\mathbb{R}} \phi(x) f(x; \theta_0) dx - \int_{\mathbb{R}} \phi_0(x) f(x; \theta_0) dx \right\} =$$

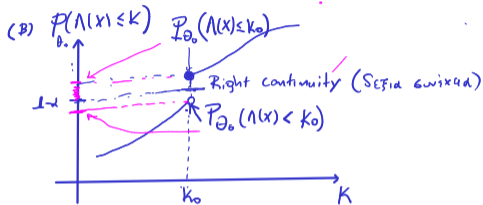
$$= \mathbb{E}_{\theta_1} \{ \phi_0(x) \} - \mathbb{E}_{\theta_1} \{ \phi(x) \} + K \underbrace{\left\{ \mathbb{E}_{\theta_0} \{ \phi(x) \} - \mathbb{E}_{\theta_0} \{ \phi_0(x) \} \right\}}_{\leq 0}$$

therefore $\mathbb{E}_{\theta_1} \{ \phi_0(x) \} \geq \mathbb{E}_{\theta_1} \{ \phi(x) \}$ (**)

Neyman-Pearson theorem : Proof

$$(*) , (K*) \Rightarrow \mathbb{E}_{\theta_0} \{ \phi_0(x) \} \geq \mathbb{E}_{\theta} \{ \phi(x) \}$$

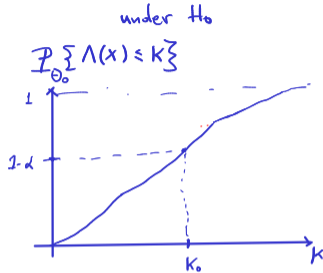
$$W_{\phi_0}(\theta) \geq W_{\phi}(\theta)$$



$$P_{\theta_0}(\Lambda(x) < k_0) \leq 1-\alpha < P_{\theta_0}(\Lambda(x) \leq k_0)$$

$$\gamma = \frac{P_{\theta_0}(\Lambda(x) \leq k_0) - (1-\alpha)}{P_{\theta_0}(\Lambda(x) \leq k_0) - P_{\theta_0}(\Lambda(x) < k_0)}$$

$$\text{Size} = \underline{\underline{\alpha}}$$



$$\exists k_0 \text{ such that } P_{\theta_0} \{ \Lambda(x) \leq k_0 \} = 1-\alpha$$

$$\Downarrow$$

$$P_{\theta_0} \{ \Lambda(x) > k_0 \} = \alpha$$

Let $\gamma = 0$ then

the size of the test is α

Neyman-Pearson theorem : Proof

Let ϕ_0 be a likelihood ratio test with $K, \gamma(x)$ and size α .

Suppose ϕ is a test of the same size α and the same power.

$$U(x) = (\phi_0(x) - \phi(x)) (\mathbb{P}(x; \theta_1) - K \mathbb{P}(x; \theta_0)) \quad U(x) \geq 0$$

$$\int_{\mathbb{R}} U(x) dx \geq 0 \xrightarrow[\text{same power}]{\text{same size}} \int_{\mathbb{R}} U(x) dx = 0 \Rightarrow$$

$$\phi_0(x) = \phi(x)$$

or


$$1(x) = K$$

α almost surely

Simple H_0 vs simple H_1

Example : $X \sim \text{Bin}(5, \theta)$, $H_0 : \theta = 0.5$ vs $H_1 : \theta = 0.75$

\propto coin
fair $\theta_0 = 0.5$
 $\theta_1 = 0.75$



α is given!

ϕ_0 :

we are going to determine the constant K and the function $\gamma(x)$

$$f(x; \theta) = \binom{5}{x} \theta^x (1-\theta)^{5-x}$$

$$\Lambda(x) = \frac{\theta_1^x (1-\theta_1)^{5-x}}{\theta_0^x (1-\theta_0)^{5-x}} = \left(\frac{\theta_1}{\theta_0}\right)^x \left(\frac{1-\theta_1}{1-\theta_0}\right)^{5-x} = \left(\frac{3}{2}\right)^x \left(\frac{1-\frac{3}{4}}{1-\frac{1}{2}}\right)^{5-x} = \left(\frac{3}{2}\right)^x \left(\frac{1}{2}\right)^{5-x} = 3^x 2^{-x} 2^{x-5} = 2^{-5} 3^x$$

$$\mathcal{P}_{\theta_0} \left\{ \Lambda(x) \leq K \right\} = \mathcal{P}_{\theta_0} \left\{ X \leq \frac{5 \log 2 + \log K}{\log 3} \right\} = \mathbb{F}_X \left(\frac{5 \log 2 + \log K}{\log 3} ; \theta_0 \right)$$

$$\Lambda(x) \leq K \Leftrightarrow 2^{-5} 3^x \leq K \Rightarrow 3^x \leq 2^5 K \Rightarrow x \log 3 \leq 5 \log 2 + \log K \Rightarrow$$

$$x \leq \frac{5 \log 2 + \log K}{\log 3}$$

$$x \in \{0, 1, 2, 3, 4, 5\} \quad \theta_0 = 0.5$$

Simple H_0 vs simple H_1

$$\alpha = 0.05 \quad 1 - \alpha = 0.95$$

Example : $X \sim \text{Bin}(5, \theta)$, $H_0 : \theta = 0.5$ vs $H_1 : \theta = 0.75$

x	0	1	2	3	4	5
F	0.031	0.19	0.5	0.81	0.97	1

$$1 - \alpha = 0.95$$

$$\gamma = \frac{0.97 - 0.95}{0.97 - 0.81} = 0.125$$

$$\frac{5 \log 2 + \log k_0}{\log 3} = 4$$

$$\log k_0 = 4 \log 3 - 5 \log 2 = \log \frac{3^4}{2^5}$$

$$k_0 = \frac{3^4}{2^5} = 2.5125$$

$$\phi_0(x) = \begin{cases} 1, & \Lambda(x) > 2.5125 \\ 0.125, & \Lambda(x) = 2.5125 \\ 0, & \text{otherwise} \end{cases}$$