

6 Hypothesis Testing

- $\Theta = H_0 \cup H_a$, $H_0 \cap H_a = \emptyset$
- $H_0 \quad H_a$
- $\alpha(\theta) = \text{prob. } H_0 \text{ rejected given } z \in \mathbb{X}$
- $\alpha = \infty \rightarrow [0, 1]$
- $R(\theta, \alpha) = c_0 E_{\theta}(1) I_{H_0}(\theta) + c_a (1 - E_{\theta}(1)) I_{H_a}(\theta)$
- no optimal α so restrict to $\mathcal{D}_\alpha = \{\alpha : E_{\theta}(1) \leq \alpha \forall \theta \in H_0\}$
size α test fns
- set $c_0 = c_a = 1$ find $\alpha \in \mathcal{D}_\alpha$ that maximizes
the power function $E_{\theta}(1)$ for each $\theta \in H_a$
- $P_\theta(A) = \int_A f_\theta(x) \mu(dx)$ where $\mu = \text{"volume measure"}$

Theorem ① (Neyman-Pearson's The Fundamental Lemma)

Suppose $\Theta = \{\theta_0, \theta_1, \dots\}$ and $H_0 = \{\theta_0\}$. Then

(i) $\exists \alpha \in \mathcal{D}_\alpha$ satisfying $E_{\theta_0}(1) = \alpha$ and st. α
is of the form

$$\alpha = \begin{cases} 1 & f_{\theta_0}(x)/f_{\theta_1}(x) > k_0 \\ 0 & \text{otherwise} \end{cases}$$

for some $k_0 \in \mathbb{R} \cup \{-\infty\}$, $\lambda \in [0, 1]$.

(ii) α in (i) is MP size α .

(iii) If α is MP size α then

$$M\{z : \alpha_{\theta_0}(z) \neq \alpha(z), f_{\theta_0}(z)/f_{\theta_1}(z) \neq k_0\} = 0$$

and $E_{\theta_0}(\alpha_{\theta_0}) = \alpha$ unless $E_{\theta_1}(\alpha_{\theta_0}) = 1$.

Corollary ② If α is MP size α then $\alpha \leq E_{\theta_1}(\alpha)$

with the inequality strict when $0 < \alpha < 1$.

Proof of the Fundamental Lemma (FL)

First note that $\{\varepsilon : f_{\Theta_0}(\varepsilon) = f_{\Theta_1}(\varepsilon) = 0\}$ has P_{Θ_0} and P_{Θ_1} measure 0 and so can be deleted from \mathbb{X} .

Proof of (i) and (ii)

Suppose $\alpha=1$. Then putting $b_0=0, \delta=1$ we see

that $\alpha(\varepsilon) \equiv 1$ satisfies (i) and $E_{\Theta_0}(\alpha) = 1 = E_{\Theta_1}(\alpha)$

and clearly α is MP-size α .

Suppose $\alpha=0$ Then put $b_0=\infty, \delta=1$ and define

$$\alpha(\varepsilon) = \sum_0^1 f_{\Theta_1}(\varepsilon)/f_{\Theta_0}(\varepsilon) = \infty < \infty. \text{ Then } E_{\Theta_0}(\alpha) = 0$$

and $E_{\Theta_1}(\alpha) = P_{\Theta_1}(\{\varepsilon : f_{\Theta_0}(\varepsilon) = 0\})$. If α_y is of size 0 then $E_{\Theta_0}(\alpha_y) = 0$ implying $\alpha_y = 0$ a.e. P_{Θ_0}

$$\text{So } \mu(\{x : \alpha_y(x) > 0, f_{G_0}(x) > 0\}) = 0.$$

$$\text{Therefore, } E_{G_1}(\alpha_y) = \sum_{\{x : \alpha_y(x) > 0\}} \alpha_y(x) f_{G_1}(x) \mu(dx)$$

$$= \sum_{\{x : \alpha_y(x) > 0, f_{G_0}(x) = 0\}} + \sum_{\{x : \alpha_y(x) > 0, f_{G_0}(x) > 0\}}$$

$$= \sum_{\{x : \alpha_y(x) > 0, f_{G_0}(x) = 0\}} f_{G_1}(x) \mu(dx) \leq P_{G_1}(\{x : f_{G_0}(x) = 0\}) > 0$$

$$= E_{G_1}(\alpha_y)$$

Suppose $0 < \alpha < 1$

$$(i) \text{ Consider } 1 - \alpha(k) = P_{G_0}(f_{G_1}(x) / f_{G_0}(x) \leq k),$$

Then $1 - \alpha(k)$ is the cdf of $f_{G_1}(x)/f_{G_0}(x)$ when

$x \sim f_{G_0}$. Therefore $1 - \alpha(k)$ is an increasing function of k , $1 - \alpha(-\infty) = 0$, $1 - \alpha(\infty) = 1$ and it is right continuous. Let $k_0 = \inf\{k : 1 - \alpha \leq 1 - \alpha(k)\}$.

Then by right contg we have $1 - \alpha(k_0 - 0) \leq 1 - \alpha \leq 1 - \alpha(k_0)$. Further $P_{G_0}(f_{G_1}(x) / f_{G_0}(x) = k_0) = (1 - \alpha(k_0)) - (1 - \alpha(k_0 - 0)) = \alpha(k_0 - 0) - \alpha(k_0)$.

Then define $\alpha(x) = \begin{cases} 1 & f_{G_1}(x) / f_{G_0}(x) > k_0 \\ \frac{\alpha - \alpha(k_0)}{\alpha(k_0 - 0) - \alpha(k_0)} & (0 \text{ when } f_{G_0}(x) = 0) \\ 0 & \text{otherwise} \end{cases}$

(Hg)

$$\text{Then } \mathbb{E}_{\theta_0}(\alpha) = P_{\theta_0}(f_{\theta_1}(x)/f_{\theta_0}(x) > k_0)$$

$$+ \frac{\alpha - \alpha(k_0)}{\alpha(k_0 - 0) - \alpha(k_0)} P(f_{\theta_1}(x)/f_{\theta_0}(x) = k_0)$$

$$= \alpha(k_0) + \frac{\alpha - \alpha(k_0)}{\alpha(k_0 - 0) - \alpha(k_0)} (\alpha(k_0 - 0) - \alpha(k_0)) = \alpha.$$

(ii) Consider α_2 with α as in (i). Then

$$\mathbb{E}_{\theta_1}(\alpha) - \mathbb{E}_{\theta_1}(\alpha_2) = \sum_x (\alpha(x) - \alpha_2(x)) f_{\theta_1}(x) \mu(dx)$$

$$= \sum_{\{x : \alpha(x) - \alpha_2(x) \neq 0\}} (\alpha(x) - \alpha_2(x)) f_{\theta_1}(x) \mu(dx)$$

$$\text{now } \{x : \alpha(x) - \alpha_2(x) \neq 0\} :$$

$$= \{x : \alpha(x) - \alpha_2(x) < 0\} \cup \{x : \alpha(x) - \alpha_2(x) > 0\}$$

$$\text{and } \{x : \alpha(x) - \alpha_2(x) < 0\}$$

$$= \{x : f_{\theta_1}(x)/f_{\theta_0}(x) < k_0, \alpha(x) - \alpha_2(x) < 0\}$$

$$\cup \{x : f_{\theta_1}(x)/f_{\theta_0}(x) = k_0, \alpha(x) - \alpha_2(x) < 0\}$$

since $f_{\theta_1}(x)/f_{\theta_0}(x) > k_0$ implies $\alpha(x) - \alpha_2(x) = 1 - \alpha_2(x) > 0$

$$\Rightarrow \sum_{\{x : f_{\theta_1}(x)/f_{\theta_0}(x) < k_0, \alpha(x) - \alpha_2(x) < 0\}} (\alpha(x) - \alpha_2(x)) k_0 f_{\theta_0}(x) \mu(dx) +$$

$$\sum_{\{x : f_{\theta_1}(x)/f_{\theta_0}(x) = k_0, \alpha(x) - \alpha_2(x) < 0\}} (\alpha(x) - \alpha_2(x)) k_0 f_{\theta_0}(x) \mu(dx)$$

$$= k_0 \int_{\mathbb{R}} (\alpha(x) - \alpha_y(x)) f_{\Theta_0}(x) \mu(dx)$$

$$= k_0 (\mathbb{E}_{\Theta_0}(\alpha) - \mathbb{E}_{\Theta_0}(\alpha_y)) = k_0(\alpha - \mathbb{E}_{\Theta_0}(\alpha_y)) > 0$$

since α_y is of size α . Therefore, we must

have $\mathbb{E}_{\Theta_1}(\alpha) > \mathbb{E}_{\Theta_0}(\alpha)$ and α is MP sized.

(iii) Let α_y be a MP size α set and α be as in (i) and $\mathbb{X}_y = \{x : \alpha(x) \neq \alpha_y(x)\} \cap \{x : f_{\Theta_1}(x) \neq k_0 f_{\Theta_0}(x)\}$.

$$\text{Then } 0 > (\mathbb{E}_{\Theta_1}(\alpha) - \mathbb{E}_{\Theta_0}(\alpha)) - k_0(\alpha - \mathbb{E}_{\Theta_0}(\alpha_y))$$

$$= \int_{\mathbb{X}_y} (\alpha(x) - \alpha_y(x)) (f_{\Theta_1}(x) - k_0 f_{\Theta_0}(x)) \mu(dx)$$

$$= \int_{\mathbb{X}_y} (\alpha(x) - \alpha_y(x)) (f_{\Theta_1}(x) - k_0 f_{\Theta_0}(x)) \mu(dx)$$

$$> 0 \text{ since } (\alpha(x) - \alpha_y(x))(f_{\Theta_1}(x) - k_0 f_{\Theta_0}(x)) > 0 \text{ on } \mathbb{X}_y$$

and so we must have that $\nu(\mathbb{X}_y) = 0$. (This

implies that $\alpha_y(x) = \alpha(x)$ when $f_{\Theta_1}(x) \neq k_0 f_{\Theta_0}(x)$

except perhaps on a set having ν -measure 0.)

If $\mu(\{\{x : f_{\Theta_1}(x)/f_{\Theta_0}(x) = k_0\}\}) = 0$ then

$\alpha_y = \alpha$ a.e. μ and $\mathbb{E}_{\Theta_1}(\alpha_y) = \alpha$.

Consider the situation when $\mu(\{x | f_{\theta_0}(x)/f_{\theta_0}(x) \geq b_0\}) > 0$

and $E_{\theta_0}(\alpha) = E_{\theta_0}(\alpha_0) + 1$. Suppose

$$\alpha > E_{\theta_0}(\alpha_0) = \sum_{\{f_{\theta_0}(x)/f_{\theta_0}(x) \geq b_0\}} \alpha_x(x) f_{\theta_0}(x) \mu(dx)$$

$$+ \sum_{\{f_{\theta_0}(x)/f_{\theta_0}(x) < b_0\}} \alpha_x(x) f_{\theta_0}(x) \mu(dx)$$

$$\text{which implies } \sum_{\{x | f_{\theta_0}(x) < b_0\}} \alpha_x(x) f_{\theta_0}(x) \mu(dx) \\ \leq b_0$$

$$< \sum_{\{x | f_{\theta_0}(x) < b_0\}} \alpha_x(x) f_{\theta_0}(x) \mu(dx). \quad \text{Assuming } b_0 \neq 0,$$

~~$$\text{this implies } \frac{1}{b_0} \sum_{\{x | f_{\theta_0}(x) < b_0\}} \alpha_x(x) f_{\theta_0}(x) \mu(dx) \\ < \frac{1}{b_0} \sum_{\{x | f_{\theta_0}(x) < b_0\}} \alpha(x) f_{\theta_0}(x) \mu(dx)$$~~

~~$$\text{which implies}$$~~

$$E_{\theta_0}(\alpha_0) < E_{\theta_0}(\alpha) \quad (\times). \quad \text{When } b_0 = 0$$

$$\text{then } E_{\theta_0}(\alpha) = 1. \quad \text{When } b_0 = \infty \text{ then } E_{\theta_0}(\alpha)$$

$$= 0 \text{ then } \alpha = 0 \text{ since } \alpha \text{ is exact size of } \alpha \text{ and } \alpha$$

$$E_{\theta_0}(\alpha_0) = 0.$$

Proof of Corollary ④

Consider $\alpha_0(\omega) = \alpha$. Then $F_{\alpha_0}(\omega) \geq$

$F_{\alpha_0}(\omega_0) = \alpha$. Suppose $F_{\alpha_0}(\omega) < \alpha$.

Then ω_0 is MP size α which by (iif) implies

$$\mu(\{ \omega : F_{\alpha_0}(\omega) = k_0 f_{\alpha_0}(\omega)^3 \}) = 1$$

which implying $k_0 = 1$ since f_{α_0} is a density

for $\omega \in \mathbb{W}$. But this implies $F_{\alpha_0} = f_{\alpha_0}$. \times

- note - if $\mu(\{ \omega : F_{\alpha_0}(\omega) = k f_{\alpha_0}(\omega)^3 \}) = 0$
 then randomized tests are not required and MP size α test is unique

- to find k_0 and δ calculate

$$\alpha(k) = P_{\alpha_0}(f_{\alpha_0}(\omega) / f_{\alpha_0}(\omega) > k) \text{ and}$$

find k_0 st. $\alpha(k_0-1) > \alpha$ & $\alpha(k_0+1) < \alpha$ and then put

$$\delta = \begin{cases} \frac{\alpha - \alpha(k_0)}{\alpha(k_0+1) - \alpha(k_0)} & \text{when } \alpha(k_0) < \alpha \\ \alpha(k_0-1) - \alpha(k_0) & \text{otherwise.} \end{cases}$$

eg ① $x = (x_1, \dots, x_n)$ iid $N(\mu, 1)$

- $\mu \in \{\mu_0, \mu_1\}$ with $\mu_2 < \mu_1$, and test $H_0 = \{\mu_0\}$ versus $H_a = \{\mu_1\}$

- we can restrict to tests fors that depend on the data only thru a sufficient statistic and \bar{x} is sufficient

- $\bar{x} \stackrel{H_0}{\sim} N(\mu_0, 1/n)$, $\bar{x} \stackrel{H_a}{\sim} N(\mu_1, 1/n)$

- to find k_0 and δ

$$\log(f_{\mu_1}(\bar{x})/f_{\mu_0}(\bar{x})) > k_0$$

$$\text{iff } \log(f_{\mu_1}(\bar{x})/f_{\mu_0}(\bar{x})) = -\frac{n}{2}(\bar{x}-\mu_1)^2 + \frac{n}{2}(\bar{x}-\mu_0)^2 > \log k_0$$

$$\text{iff } -(\bar{x}-\mu_1)^2 + (\bar{x}-\mu_0)^2 > \frac{2}{n} \log k_0$$

$$\text{iff } 2(\mu_1 - \mu_0)\bar{x} > -\frac{2}{n} \log k_0 + \mu_1^2 - \mu_0^2$$

$$\text{iff } \bar{x} > \frac{1}{2(\mu_1 - \mu_0)} \left(-\frac{2}{n} \log k_0 + \mu_1^2 - \mu_0^2 \right) \quad (\text{note } \mu_1 > \mu_0)$$

$$\text{iff } \sqrt{n}(\bar{x} - \mu_0) > \frac{\sqrt{n}}{2(\mu_1 - \mu_0)} \left(-\frac{2}{n} \log k_0 + \mu_1^2 - \mu_0^2 \right) - \sqrt{n}\mu_0 = k'_0$$

$$\therefore P_{\mu_0}(f_{\mu_1}(\bar{x})/f_{\mu_0}(\bar{x}) > k_0) = P_{\mu_0}(\sqrt{n}(\bar{x} - \mu_0) > k'_0)$$

$$= 1 - \Phi(k'_0) \text{ and we choose } k'_0 = z_{1-\alpha} \text{ where}$$

$z_{1-\alpha}$ is the α -th quantile of the $N(0, 1)$ distn.

and for this choice of k_0 (which determines k_0)

$$\text{we have } P_{\mu_0} (f, \bar{x}) / f_0(\bar{x}) > k_0 = \alpha$$

and so $\gamma = 0$. Therefore the MP size α

test for testing $\sum \mu_0^3$ versus $\sum \mu_1^3$ when

$$\mu_1 > \mu_0 \Leftrightarrow Q(\alpha) = 1 \text{ when } \ln(\bar{x} - \mu_0) > z_{1-\alpha}$$

and $Q(\alpha) = 0$ otherwise (so we reject H_0 whenever \bar{x} is large).

Note $\alpha = .05, \mu_0 = 0, \mu_1 = 1, z_{.95} = 1.644854$
 $\log k_0 = \ln(1.644854)$ only if $n \geq 11$
so can reject H_0 even when evidence favors H_0

Notice that this does not depend on μ_1

and so α is MP size α for $H_0 = \sum \mu_0^3$ versus

$H_a = (\mu_0, \infty)$ (i.e. for any alternative $> \mu_0$).

Also if $\mu < \mu_0$ then by the same argument

Q is the MP size $E_{\mu_0}(\alpha)$ test of $H_0 = \sum \mu_3^3$

vs $H_a = \sum \mu_0^3$ and by the Corollary $E_{\mu}(\alpha) \leq E_{\mu_0}(\alpha) = \alpha$.

Therefore, α is size α for $H_0 = (-\infty, \mu_0]$

versus $H_a = (\mu_0, \infty)$. Now suppose that

α_y is size α for $H_0 = (-\infty, \mu_0]$ vs $H_a = (\mu_0, \infty)$.

Then α_y is size α for $H_0 = \{\mu_0\}$ vs $H_a = \{\mu_0\}$
for $\mu_1 > \mu_0$ and so $E_{\mu_1}(\alpha_y) \leq E_{\mu_0}(\alpha)$.

This proves that α is UMP size α .

For $H_0 = (-\infty, \mu_0]$ vs $H_a = (\mu_0, \infty)$.

- note - we can repeat this argument to

find the UMP size α test for

$H_0 = \{\mu_0, \infty\}$ vs $H_a = (-\infty, \mu_0)$ and obtain

$$\alpha_y(\alpha) = 1 \text{ when } \tau_n(\bar{x} - \mu_0) \leq z_\alpha.$$

- now consider the two-sided problem

$H_0 = \{\mu_0\}$ vs $H_a = (-\infty, \mu_0) \cup (\mu_0, \infty)$

and suppose α_{yy} is UMP size α for this.

- but then α_{yy} is size α for $H_0 = \{\mu_0\}$

vs $H_a = \{\mu_0\}$ and is MP, so by FL(iii)

we have $\alpha_{yy}(\alpha) = \alpha(\alpha)$ (when $\mu_1 > \mu_0$) and

$\alpha_{\text{size}}(x) = \alpha_2(x)$ (when $\mu_1 < \mu_0$) which is impossible

- Therefore there is no UMP size α test for the two-sided problem (have to restrict to unbiased tests and then $\alpha(x) = 1$ when $\sqrt{n}(\bar{x} - \mu_0) > z_{\alpha/2}$ and 0 otherwise, can be shown to be UMP unbiased of size α).

$$\text{eq ②} - x = (x_1, \dots, x_n) \stackrel{\text{iid}}{\sim} U(0, \theta)$$

$\leftarrow H_0 = \{\Theta_0\}$ vs $H_a = \{\Theta_1\}$ where $\Theta_0 < \Theta_1$

- we can restrict to tests which are fun
of the suff. stat. \bar{x}_{mn}

$$\begin{aligned} P_G(X_m \leq x) &= P_G(X_1 \leq x, \dots, X_n \leq x) \\ &= \prod_{i=1}^n P_G(X_i \leq x) = \left(\frac{1}{n} \sum_{s=1}^n ds \right) = \frac{x^n}{n!} \end{aligned}$$

$$\text{and so } f_0(u) = n u^{n-1} / \epsilon n$$

WOW

$$f_{\theta_1}(x_m)/f_{\theta_0}(x_m) = \left(\frac{\theta_0}{\theta_1}\right)^n \quad 0 < x_m < \theta_0$$

{ 8 $\theta_0 < x_m < \theta_1$

$$P_{\Theta_0} \left(\frac{f_{\Theta_1}(x_m)}{f_{\Theta_0}(x_m)} > k_0 \right) = \begin{cases} 1 & k_0 < \left(\frac{\Theta_0}{\Theta_1}\right)^n \\ 0 & k_0 > \left(\frac{\Theta_0}{\Theta_1}\right)^n \end{cases}$$

For $0 < \alpha < 1$, $v_0 = \left(\frac{G_0}{\theta}\right)^{\alpha}$ and $\chi = \alpha$.

$$\text{and } \phi(x) = \begin{cases} x & 0 < x_{\text{end}} \leq \theta_0 \\ 1 & x_{\text{end}} > \theta_0 \end{cases}$$

- so the MP size α test here is to randomly reject with probability α whenever $0 \leq x_m \leq G_0$ and reject when $x_m > G_0$

- The "boundary" here is the set $\{0, \theta_0\}$
 (this is continuous model when we have to randomize)
- also this test does not depend on θ_1 , so it is UMP size α for $H_0 = \{\theta_0\}$ vs $H_a = (\theta_0, \infty)$
- also for $\theta < \theta_0$ we have $E_{\theta}(\phi) \geq \alpha$ and so (as in the previous example) the test is UMP size α for $H_0 = \{0, \theta_0\}$ vs $H_a = (\theta_0, \infty)$.

e.g. $\mathbf{x} = (x_1, \dots, x_n) \sim \text{Poisson}(\lambda)$

- $H_0 = \{\lambda_0\}$ vs $H_a = \{\lambda\} \quad \lambda_0 < \lambda$

- Then $n\bar{\lambda}$ is sufficient so we can base the test on $n\bar{\lambda} \sim \text{Poisson}(\lambda)$

$$- f_{\lambda_0}(n\bar{\lambda}) / f_{\lambda_0}(n\bar{\lambda}) = \left(\frac{\lambda}{\lambda_0}\right)^{n\bar{\lambda}} e^{-n(\lambda - \lambda_0)} \rightarrow$$

$$\text{iff } n\bar{\lambda} \log(\lambda/\lambda_0) - n(\lambda - \lambda_0) > \log k$$

$$\text{iff } n\bar{\lambda} > (\log(\lambda/\lambda_0))^{-1} (k + n(\lambda - \lambda_0)) \\ (\text{note } \log(\lambda/\lambda_0) > 0)$$

- so let $k_0 \in \mathbb{N}$ be st. $P_{\lambda_0}(n\bar{\lambda} > k_0) \leq \alpha$
and $P_{\lambda_0}(n\bar{\lambda} \leq k_0-1) > \alpha$

- Then MP size α test of $H_0 = \{\lambda_0\}$ vs $H_a = \{\lambda\}$ is

$$\alpha(x) = \begin{cases} 1 & n\bar{\lambda} > k_0 \\ \frac{\alpha - P_{\lambda_0}(n\bar{\lambda} > k_0)}{P_{\lambda_0}(n\bar{\lambda} > k_0) - P_{\lambda_0}(n\bar{\lambda} \leq k_0)} = \delta & n\bar{\lambda} \leq k_0 \end{cases} =$$

- since this test does not involve λ , it is
MP size α for $H_0 = \{\lambda_0\}$ vs $H_a = (\lambda_0, \infty)$

- now suppose $\lambda < \lambda_0$ and consider the problem $H_0 = \{\lambda\}$ vs $H_a = \{\lambda_0\}$

- Then the same argument shows that α is MP of size $E_{\lambda}(\alpha)$ for the problem, and so by the Corollary of the FL $E_{\lambda}(\alpha) \leq E_{\lambda_0}(\alpha) = \infty$
- Therefore α is of size α for $H_0 = (0, \lambda_0]$ vs $H_a = (\lambda_0, \infty)$ and so must be UMP-size α for this problem.