

# < Chapter 8 > slide #5.

①

$X_i \stackrel{iid}{\sim} \mathcal{N}(\theta, \sigma^2)$ ,  $\sigma^2$  known,  $i=1, \dots, n$ ,  $H_0: \theta = \theta_0$  vs  $H_a: \theta \neq \theta_0$

$$\lambda(x) = \frac{f(x | \hat{\theta}_0)}{f(x | \hat{\theta})} = \frac{\prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(X_i - \theta_0)^2}{2\sigma^2}\right\}}{\prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(X_i - \bar{X})^2}{2\sigma^2}\right\}}$$

$$= \exp\left\{\frac{\sum X_i}{\sigma^2}(\theta_0 - \bar{X}) - \frac{n}{2\sigma^2}(\theta_0^2 - \bar{X}^2)\right\}$$

$$\sum X_i = n\bar{X} \Rightarrow \exp\left\{\frac{n}{2\sigma^2}(2\bar{X}\theta_0 - 2\bar{X}^2 - \theta_0^2 + \bar{X}^2)\right\}$$

$$= \exp\left\{-\frac{n}{2\sigma^2}(\bar{X}^2 - 2\bar{X}\theta_0 + \theta_0^2)\right\} = \exp\left\{-\frac{n}{2\sigma^2}(\bar{X} - \theta_0)^2\right\}$$

Then,  $-2 \log \lambda(x) = -2 \left\{-\frac{n}{2\sigma^2}(\bar{X} - \theta_0)^2\right\}$

$$= \left\{\frac{\bar{X} - \theta_0}{\sigma/\sqrt{n}}\right\}^2 \stackrel{\text{under } H_0}{\sim} \chi^2_{(1)}$$

\* When  $X_i$ 's are normal,  $-2 \log \lambda(x)$  follows  $\chi^2_{(1)}$ . \* We reject  $H_0$  if  $\lambda(x) \leq c$ .

\* Under certain conditions (conditions required for CRLB),  $-2 \log \lambda(x)$  approximately follows  $\chi^2_{(1)}$ .

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$$\lambda(x) = \frac{\prod_{i=1}^n e^{-(X_i - \theta_0)}}{\prod_{i=1}^n e^{-(X_i - \hat{\theta})}} = \frac{e^{-\sum X_i + n\theta_0}}{e^{-\sum X_i + n\bar{X}}} = \exp\{n(\theta_0 - \bar{X})\}$$

In LRT reject  $H_0$  if  $\lambda(x) \leq c \Leftrightarrow \log \lambda(x) \leq \log c$

$$\Leftrightarrow n(\theta_0 - \bar{X}) \leq \log c$$

$$\Leftrightarrow \bar{X} \geq \theta_0 - \frac{\log c}{n}$$

The reject region depends on the sample only through the sufficient statistic.

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$X_i$ 's iid exponential( $\theta$ ),  $i=1, \dots, n$ . Consider a test for  $\phi(x) = \begin{cases} 1 & \bar{x} < c \\ 0 & \text{o/w.} \end{cases}$  (2)

$$\begin{aligned} \beta_{\phi}(\theta) &= P_{\theta}[\phi(x) = 1] \\ &= P_{\theta}[\bar{x} < c] = P_{\theta}[\sum X_i < nc], \quad \sum X_i \sim \text{Gamma}(n, \theta) \\ &= \int_0^{nc} \frac{1}{\Gamma(n)\theta^n} t^{n-1} e^{-t/\theta} dt. \quad ; \text{Power function.} \end{aligned}$$

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$X_i$ 's iid Gamma(3,  $\theta$ ),  $i=1, \dots, n$ .  $H_0: \theta = \theta_0$  vs  $H_1: \theta = \theta_1 (>\theta_0)$

Find MP-test of size  $\alpha$ . ( $=0.05$ )

We reject  $H_0$  if  $L = \frac{f(x|\theta_1)}{f(x|\theta_0)} > k$ .

$$\begin{aligned} L &= \frac{\prod_{i=1}^n \frac{1}{\Gamma(3)\theta_1^3} x_i^{3-1} e^{-x_i/\theta_1}}{\prod_{i=1}^n \frac{1}{\Gamma(3)\theta_0^3} x_i^{3-1} e^{-x_i/\theta_0}} \\ &= \left(\frac{\theta_0}{\theta_1}\right)^3 \exp\left\{-\sum x_i \left(\frac{1}{\theta_1} - \frac{1}{\theta_0}\right)\right\} = \left(\frac{\theta_0}{\theta_1}\right)^3 \exp\left\{n\bar{x} \frac{(\theta_1 - \theta_0)}{\theta_0 \theta_1}\right\} \end{aligned}$$

$\Rightarrow$  Note  $L$  is an increasing function of  $\bar{x}$  since  $\theta_1 - \theta_0 > 0$ .

Thus, rejecting  $H_0$  if  $L > k$  is the same as rejecting  $H_0$  if  $\bar{x} > c$ .

Or, if  $\sum X_i > C_1 = nc$ . As  $T = \sum X_i \stackrel{H_0}{\sim} \text{Gamma}(3n, \theta_0)$ ,

$C_1$  can be found numerically s.t.  $\int_{C_1}^{\infty} \frac{1}{\Gamma(3n)\theta_0^{3n}} t^{3n-1} e^{-t/\theta_0} dt = \alpha (=0.05)$ ;

\* When  $n$  is large,  $(E[T] = 3n\theta_0, \text{Var}(T) = 3n\theta_0^2)$  Find  $C$  s.t.

$P\left(\frac{\sum X_i - 3n\theta_0}{\sqrt{3n\theta_0^2}} > c\right) = \alpha$ . So, MP test of (approximate) size  $\alpha (=0.05)$  is

reject  $H_0$  if  $\sum X_i > 3n\theta_0 + c\sqrt{3n\theta_0^2}$  where  $c = 1.645$  for  $\alpha = 0.05$ .

$X_i$ 's  $\stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$ ,  $\sigma^2$  known.  $i=1, \dots, n$ .

$$H_0: \mu = \mu_0 \quad \text{vs} \quad H_1: \mu = \mu_1 (> \mu_0).$$

Find MP-test of size  $\alpha$ .

$$L = \frac{\exp\left\{-\frac{\sum (X_i - \mu_1)^2}{2\sigma^2}\right\}}{\exp\left\{-\frac{\sum (X_i - \mu_0)^2}{2\sigma^2}\right\}}$$

$$= \exp\left\{\frac{n\bar{X}}{2\sigma^2} (\mu_1 - \mu_0) - \frac{n}{2\sigma^2} (\mu_1^2 - \mu_0^2)\right\}.$$

As  $L$  is an increasing function of  $\bar{X}$ , reject  $H_0$  if  $L > k$  is the same as rejecting  $H_0$  if  $\bar{X} > c$ , where  $c$  is determined as

$$P(\bar{X} > c) \stackrel{H_0}{=} \alpha \quad \text{or} \quad P\left(\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} > c_1\right) = \alpha.$$



$\therefore$  MP test: Reject  $H_0$  if  $\bar{X} > \mu_0 + C_1 \frac{\sigma}{\sqrt{n}}$ , where  $C_1 = Z_{1-\alpha}$ .

⊗  $H_a: \mu = \mu_1 (< \mu_0)$ .

Then,  $L$  is a decreasing function of  $\bar{X}$ .



$\Rightarrow$  MP test is "Reject  $H_0$  if  $\bar{X} < \mu_0 + C_1 \frac{\sigma}{\sqrt{n}}$ , where  $C_1 = Z_{\alpha}$ .

⊗  $H_a: \mu = \mu_1 (\neq \mu_0)$

We have UMP test for the two cases of  $(\mu_1 > \mu_0)$ ,  $(\mu_1 < \mu_0)$ , and they are not identical. So, for the hypothesis  $H_0: \mu = \mu_0$  vs  $H_a: \mu \neq \mu_0$ , we do NOT have a UMP test.



$X_i$ 's iid  $f(x|\lambda) = \lambda e^{-\lambda x}$ ,  $x > 0$ ,  $\lambda > 0$ . Find UMP test of size  $\alpha$ .  
 This is a scale parameter.

$$\frac{L(x|\lambda_1)}{L(x|\lambda_0)} = \frac{\prod_{i=1}^n \lambda_1 e^{-\lambda_1 x_i}}{\prod_{i=1}^n \lambda_0 e^{-\lambda_0 x_i}} = \left(\frac{\lambda_1}{\lambda_0}\right)^n \exp\left\{-\sum_{i=1}^n x_i (\lambda_1 - \lambda_0)\right\} \geq k \quad \text{--- (a)}$$

$$\Rightarrow \exp\left\{-\sum_{i=1}^n x_i (\lambda_1 - \lambda_0)\right\} \geq \left(\frac{\lambda_0}{\lambda_1}\right)^n k$$

$$\Rightarrow -\sum_{i=1}^n x_i (\lambda_1 - \lambda_0) \geq \log\left(\frac{\lambda_0}{\lambda_1}\right)^n k \equiv k_1$$

$$\Rightarrow -\sum_{i=1}^n x_i (\lambda_1 - \lambda_0) \geq k_1$$

$$\Rightarrow \sum_{i=1}^n x_i \leq c$$

$$\begin{aligned} H_0: \lambda &\leq \lambda_0 \\ H_1: \lambda &> \lambda_0 \end{aligned}$$

By Neyman - Pearson Theorem, rejecting  $H_0$  if  $\sum x_i \leq c$  is UMP test. [Note: small " $\sum x_i$ " indicates small mean or large scale  $\lambda$ ]

⊗ Alternative solution.

From (a), we see the family has MLR in  $T(x) = \sum x_i$ .

(Note (a) is a decreasing function of  $\sum x_i$ , so an increasing function of  $-\sum x_i$ .)

Then, by Karlin - Rubin Thm, a UMP test is

rejecting  $H_0$  if  $-\sum x_i \geq k$  or, equivalently  $\sum x_i \leq c$ .

When  $n=100$  is given, then choose  $c$  s.t.

$$\int_0^c \frac{1}{P(n)(1/\lambda_0)^n} x^{n-1} e^{-\lambda_0 x} dx = 0.05$$

In R,  $c$  is found by  $\text{qgamma}(0.05, 100, 1/\lambda_0)$   
 (Note:  $n$  is the shape parameter,  $1/\lambda_0$  is the rate or mean.)

Note that it makes  $P(\sum x_i \leq c) = 0.05$  under  $H_0$ .  $\otimes$  scale =  $1/\text{rate}$

Ex)  $X_i$ 's iid  $f(x|\eta) = e^{-(x-\eta)}$ ,  $x > \eta$ ,  $\eta > 0$ ,  $i=1, \dots, n$ .

Find a UMP test of size  $\alpha$  for  $H_0: \eta \leq \eta_0$  vs  $H_1: \eta > \eta_0$ .

$$\frac{L(X|\eta_1)}{L(X|\eta_0)} = \frac{e^{-\sum X_i + n\eta_1} I(X_{(1)} > \eta_1)}{e^{-\sum X_i + n\eta_0} I(X_{(1)} > \eta_0)} = \exp\{\underbrace{n(\eta_1 - \eta_0)}_{>0}\} \underbrace{\frac{I(X_{(1)} > \eta_1)}{I(X_{(1)} > \eta_0)}}_{\substack{0 \text{ or } 1 \\ \Rightarrow \begin{array}{c} \text{---} | \text{---} | \text{---} | \text{---} \\ \eta_0 \quad \eta_1 \quad X_{(1)} \quad X_{(1)} \end{array}}}.$$

$\Rightarrow$  This is MLR in  $T(X) = X_{(1)}$

By Karlin-Rubin Theorem, rejecting  $H_0$  if  $X_{(1)} > c$ , where

$P(X_{(1)} > c) \stackrel{H_0}{=} \alpha$  is a UMP test of size  $\alpha$ .

Now, find the value of  $c$ .

$$P(X_{(1)} > c) = P(X_1 > c)^n = \left[ 1 - \int_{\eta_0}^c e^{-(t-\eta_0)} dt \right]^n$$

$\nearrow$   
 This is the probability of rejecting  $H_0$  when  $H_0$  is true.  
 $\therefore c = \frac{1}{n} \log \alpha + \eta_0.$

$$= [1 + (e^{c-\eta_0} - 1)]^n = e^{n(c-\eta_0)} = \alpha.$$