

Recap:

We talked about finding the UMP test for testing  $H_0: \theta \leq \theta_0$  vs.  $H_1: \theta > \theta_0$ , when the family of distributions  $\{f(\cdot|\theta) : \theta \in \mathcal{R}\}$  assumes an MLR.

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } T(\underline{x}) > b \\ 0 & \text{o.w.} \end{cases}$$

$P_{\theta=\theta_0}(T(\underline{x}) > b) = \alpha$  in the UMP test of level  $\alpha$ .

Example:  $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} N(\theta, \sigma^2)$ ,  $\sigma$  is known.

$H_0: \theta \leq \theta_0$  vs.  $H_1: \theta > \theta_0$ .

We have seen that the  $\{N(\theta, \sigma^2) : \theta \in \mathcal{R}\}$  has MLR in  $\sum_{i=1}^n X_i$ . Thus UMP test

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } \sum_{i=1}^n X_i > b \\ 0 & \text{o.w.} \end{cases}$$

$$P_{\theta_0} \left( \sum_{i=1}^n X_i > b \right) = \alpha$$

$$\Leftrightarrow \phi(\underline{x}) = \begin{cases} 1 & \text{if } \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} > c \\ 0 & \text{o.w.} \end{cases}$$

$$P_{\theta_0} \left( \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} > c \right) = \alpha$$

We know  $\sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} \sim N(0, 1)$  under  $\theta = \theta_0$

$$P_{\theta_0} \left( \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} > c \right) = \alpha \Leftrightarrow P(N(0, 1) > c) = \alpha$$

thus  $c = z_\alpha$  which can be found from the normal table

the UMP test is

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} > z_\alpha \\ 0 & \text{o.w.} \end{cases}$$

So, the classical one-sided Normal test is the UMP test.

p-value: ~~In many~~ In NP lemma  $K = K(\alpha)$  is determined by using the fact that the prob. of  $f_{\theta_1}(\underline{x}) > K$  is  $\alpha$ . Let the rejection region for level  $\alpha$  be denoted by  $R_\alpha$ . In many situations the region  $R_\alpha$  is nested in a way that  $R_\alpha \subset R_{\alpha'}$  if  $\alpha < \alpha'$ . When this is the case, it is a common practice to see not only whether the hypothesis is accepted or rejected, but also determine the smallest level at which the hypothesis is rejected. This is called the p-value. Mathematically

$$p(\underline{x}) = \inf \{ \alpha : \underline{x} \in R_\alpha \}.$$

Example:  $x_1, \dots, x_n \stackrel{i.i.d}{\sim} N(\mu, 1)$   
 $H_0: \mu = 0$  vs.  $H_1: \mu > 0$ , what is the rejection region of ~~a~~ the UMP test of level  $\alpha$ .

$$R_\alpha = \left\{ \bar{x} : \bar{x} > \frac{1}{\sqrt{n}} z_{1-\alpha} \right\} = \left\{ \bar{x} : 1 - \Phi(\sqrt{n} \bar{x}) < \alpha \right\}$$

$$\text{Given data } \inf \{ \alpha : \underline{x} \in R_\alpha \} = 1 - \Phi(\sqrt{n} \bar{x})$$

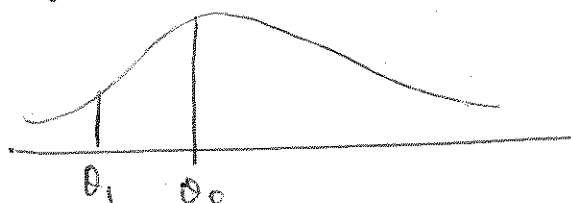
Remark: ~~100~~ ~~100~~ If we are interested in the one parameter exponential family with parameter  $\theta$ , does a UMP exist for testing  $H_0: \theta = \theta_0$  vs.  $H_1: \theta \neq \theta_0$ .

Answer: NO

Let us look at the result in the case of Normal distribution.

$X_1, \dots, X_n \stackrel{iid}{\sim} N(\theta, \sigma^2)$ ,  $\sigma^2$  known. For any  $\theta_1 < \theta_0$  the UMP level  $\alpha$ -test for testing

$H_0: \theta = \theta_0$  vs.  $H_1: \theta = \theta_1$



$H_0: \theta \geq \theta_0$  vs.  $H_1: \theta < \theta_0$

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } T(\underline{x}) < b \\ 0 & \text{o.w.} \end{cases}$$

$$\Leftrightarrow \phi(\underline{x}) = \begin{cases} 1 & \text{if } \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} < b \\ 0 & \text{o.w.} \end{cases}$$

$$P_{\theta_0} \left( \underbrace{\sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma}}_{N(0,1)} < b \right) = \alpha \quad \bullet \quad b = -z_\alpha$$

the test is going to reject the null hypothesis if  $\sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} < -z_\alpha \Leftrightarrow \bar{x} < -\frac{z_\alpha \sigma}{\sqrt{n}} + \theta_0$

the UMP level  $\alpha$ -test for testing

$H_0: \theta = \theta_0$  vs.  $H_1: \theta = \theta_1$ ,  $\theta_1 < \theta_0$  is given

by 
$$\phi(x) = \begin{cases} 1 & \text{if } \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} < -z_\alpha \\ 0 & \text{o.w.} \end{cases}$$

power calculate power for of this test at  $\theta_2 > \theta_0$

$$P_{\theta_2} \left( \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} < -z_\alpha \right)$$

$$= P_{\theta_2} \left( \bar{x} < \theta_0 - \frac{z_\alpha \sigma}{\sqrt{n}} \right)$$

$$= P_{\theta_2} \left( \sqrt{n} \frac{(\bar{x} - \theta_2)}{\sigma} < \frac{\theta_0 - \theta_2}{\sigma/\sqrt{n}} - z_\alpha \right)$$

$$\leq P(Z < -z_\alpha) \quad \text{where } Z \sim N(0,1)$$

this is because under  $\theta_2$ ,  $\sqrt{n} \frac{(\bar{x} - \theta_2)}{\sigma} \sim N(0,1)$

$$= P(Z > z_\alpha)$$

$$= P_{\theta_2} \left( \underbrace{\sqrt{n} \frac{(\bar{x} - \theta_2)}{\sigma}}_{N(0,1)} > z_\alpha \right)$$

$$= P_{\theta_2} \left( \bar{x} > \frac{\sigma z_\alpha}{\sqrt{n}} + \theta_2 \right)$$

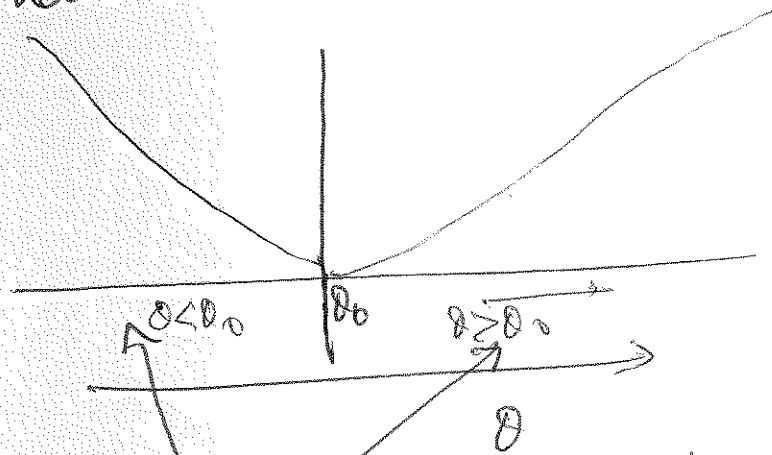
$$= P_{\theta_2} \left( \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} > z_\alpha + \frac{(\theta_2 - \theta_0)\sqrt{n}}{\sigma} \right)$$

$$< P_{\theta_2} \left( \sqrt{n} \frac{(\bar{x} - \theta_0)}{\sigma} > z_\alpha \right) \quad \text{as } \theta_2 > \theta_0$$

when  $\theta_2 > \theta_0$

$$P_{\theta_2} \left( \underbrace{\frac{\sqrt{n}(\bar{X} - \theta_0)}{\sigma}} < -z_\alpha \right) < P_{\theta_2} \left( \underbrace{\frac{\sqrt{n}(\bar{X} - \theta_0)}{\sigma}} > z_\alpha \right)$$

$\Rightarrow$  UMP test for testing  $H_0: \theta = \theta_0$  vs.  $H_1: \theta = \theta_1$ ,  
 when  $\theta_1 < \theta_0$  is  $\frac{\sqrt{n}(\bar{X} - \theta_0)}{\sigma} < -z_\alpha$  which has  
 less power than another test that rejects  
~~to~~ to test  $H_0: \theta = \theta_0$  vs.  $H_1: \theta = \theta_2, \theta_2 > \theta_0$   
 than another test.

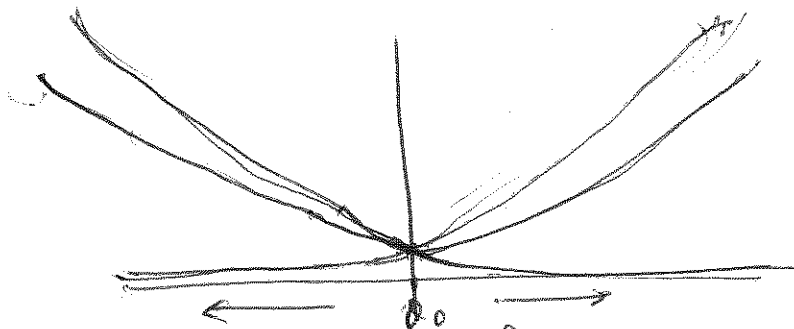


$\frac{\sqrt{n}(\bar{X} - \theta_0)}{\sigma} > z_\alpha$  is most powerful

$\frac{\sqrt{n}(\bar{X} - \theta_0)}{\sigma} < -z_\alpha$  is most powerful

$\Rightarrow H_0: \theta = \theta_0$  vs.  $H_1: \theta \neq \theta_0$  does not admit  
 the UMP test.

$H_0: \theta = \theta_0$  vs.  $H_1: \theta \neq \theta_0$  for  $x_1, \dots, x_n \stackrel{iid}{\sim} U(0, \theta)$ ,  
 then  $\exists$  ~~no~~ UMP test.



the tests for which power function doesn't go below the level in the alternative are known as unbiased tests.

A test  $\phi$  is unbiased at level  $\alpha$  for testing  $H_0: \theta \in \Theta_0$  vs.  $H_1: \theta \in \Theta_1$ , if

(i)  $\beta(\theta) \leq \alpha$  for  $\theta \in \Theta_0$

(ii)  $\beta(\theta) \geq \alpha$  for  $\theta \in \Theta_1$ .

We want to find a test  $\phi$  of level  $\alpha$ , which is Most Powerful among all unbiased tests. If such a test exists then it is called UMPU test.

clearly, if the power function  $\beta(\theta)$  of an unbiased test is cont. then  ~~$\beta(\theta)$~~  in the boundary of  $\Theta_0$  and  $\Theta_1$ ,  $\beta(\theta) = \alpha$

e.g.,  $H_0: \theta = \theta_0$  vs.  $H_1: \theta \neq \theta_0$   
 $\beta(\theta_0) = \alpha$ .

Result: If the densities are s.t. all tests have cont. power functions, then if  $\exists$  a UMP test  $T$  among the tests satisfying  $E[T] = \alpha$  whenever  $\theta$  belongs to the boundary of  $\Theta_0$  and  $\Theta_1$ , then  $T$  is UMPU level  $\alpha$ .

~~Application~~ UMPU tests in one parameter exponential family

$$p_{\theta}(x) = c(\theta) h(x) \exp(\omega(\theta) T(x))$$

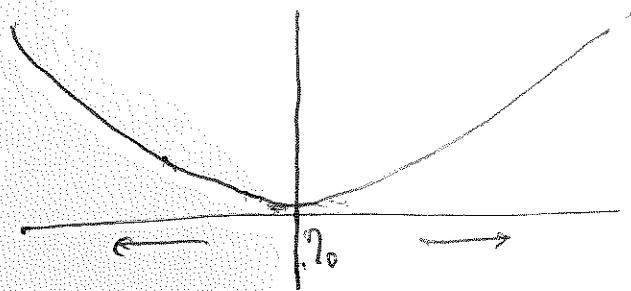
$$\text{let } \eta = \omega(\theta) \Rightarrow p_{\eta}(x) = c_1(\eta) h(x) \exp(\eta T(x))$$

This is called natural exponential family parameterization.

UMPV test: To test  $H_0: \eta = \eta_0$  vs.  $H_1: \eta \neq \eta_0$   
the UMPV test is given by

$$\phi(x) = \begin{cases} 1 & \text{if } T(x) < c_1 \text{ or } T(x) > c_2 \\ 0 & \text{o.w.} \end{cases}$$

$$E_{\eta_0}(\phi(x)) = \alpha, \quad E_{\eta_0}(T(x)\phi(x)) = \alpha E_{\eta_0}(T(x))$$



$$\beta(\eta) = E_{\eta}(\phi(x))$$

this should satisfy

$$\underline{\beta'(\eta_0) = 0}$$

If  $T(\underline{x})$  follows a distribution symmetric around ~~under~~  $R$  under  $H_0$ , then the above two conditions boil down to

$$E_{\eta_0}[\phi(\underline{x})] = \alpha, \quad c_1 + c_2 = 2R.$$

Example:  $X_1, \dots, X_n \stackrel{iid}{\sim} N(0, \sigma^2)$ ,  $\sigma^2$  known.

To test:  $H_0: \theta = \theta_0$  vs.  $H_1: \theta \neq \theta_0$ .

Here normal ~~belongs~~ belongs to the exponential family  $\eta = \theta$  and  $T(\underline{x}) = \bar{x}$ .

UMPV test should look like

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } \bar{x} < c_1 \text{ or } \bar{x} > c_2 \\ 0 & \text{o.w.} \end{cases}$$

$$\Leftrightarrow \phi(\underline{x}) = \begin{cases} 1 & \text{if } \frac{\sqrt{n}(\bar{x} - \theta_0)}{\sigma} < b_1 \text{ or } \frac{\sqrt{n}(\bar{x} - \theta_0)}{\sigma} > b_2 \\ 0 & \text{o.w.} \end{cases}$$

$$\text{subject. } E_{\theta_0}[\phi(\underline{x})] = \alpha \text{ and } E_{\theta_0}\left[\phi(\underline{x}) \frac{\sqrt{n}(\bar{x} - \theta_0)}{\sigma}\right] = \alpha E_{\theta_0}\left[\frac{\sqrt{n}(\bar{x} - \theta_0)}{\sigma}\right]$$

$$T(\underline{x}) = \frac{\sqrt{n}(\bar{x} - \theta_0)}{\sigma} \stackrel{H_0}{\sim} N(0, 1) \text{ under } H_0.$$

$\Rightarrow T(\underline{x})$  has symmetric dist. around 0 under  $H_0$ .

$$E_{\theta_0}\left(\frac{\sqrt{n}(\bar{x} - \theta_0)}{\sigma} \phi(\underline{x})\right) = \alpha \quad \text{--- (1)}$$

$$\textcircled{a} \quad b_1 + b_2 = 2 \cdot 0 = 0 \quad \text{--- (2)}$$



$$\int_{-\infty}^{b_1} f(z) dz + \int_{b_2}^{\infty} f(z) dz = \alpha \quad \text{when } f \text{ is the density of } N(0,1).$$

$$E_{\theta_0}[\phi(X)] = \alpha$$

$$\phi(X) = \begin{cases} 1 & \text{if } \sqrt{n} \frac{(\bar{X} - \theta_0)}{\sigma} < b_1 \quad \text{or} \quad \sqrt{n} \frac{(\bar{X} - \theta_0)}{\sigma} > b_2 \\ 0 & \text{o.w.} \end{cases}$$

$$\begin{aligned} E_{\theta_0}[\phi(X)] &= P_{\theta_0} \left( \sqrt{n} \frac{(\bar{X} - \theta_0)}{\sigma} < b_1 \right) + P_{\theta_0} \left( \sqrt{n} \frac{(\bar{X} - \theta_0)}{\sigma} > b_2 \right) \\ &= \Phi(b_1) + 1 - \Phi(b_2) \end{aligned}$$

thus ~~the~~ ① & ② become

$$b_1 + b_2 = 0 \quad \text{--- (3)}$$

$$\Phi(b_1) + 1 - \Phi(b_2) = \alpha \quad \text{--- (4)}$$

$$\Rightarrow b_1 = -b_2$$

$$\Rightarrow \Phi(-b_2) + 1 - \Phi(b_2) = \alpha$$

$$\Rightarrow 2\Phi(-b_2) = \alpha \quad \Rightarrow \Phi(-b_2) = \frac{\alpha}{2}$$

$$\Rightarrow b_2 = z_{1-\alpha/2} \quad b_1 = -b_2 = z_{\alpha/2} = -z_{1-\alpha/2}$$

$$\phi(X) = \begin{cases} 1 & \text{if } \left| \sqrt{n} \frac{(\bar{X} - \theta_0)}{\sigma} \right| > z_{1-\alpha/2} \\ 0 & \text{o.w.} \end{cases}$$

the classical two sided test is the UMPU test of level  $\alpha$ .

# Likelihood ratio test

The aim is to test  $H_0: \theta \in \Theta_0$  vs.  $H_1: \theta \in \Theta_1$ .  
 $= \Theta \setminus \Theta_0$

Based on samples  $X_1, \dots, X_n$ .

Let  $f_\theta(\underline{x})$  be the likelihood of  $\theta$ . The likelihood ratio test (LRT) ~~is given~~ statistic is given

by

$$\lambda = \frac{\sup_{\theta \in \Theta_0} f_\theta(\underline{x})}{\sup_{\theta \in \Theta} f_\theta(\underline{x})}$$

If  $\Theta_0$  contains not very ~~likely~~ ~~probable~~  $\theta$  values then  $\sup_{\theta \in \Theta_0} f_\theta(\underline{x})$  should be much

smaller than  $\sup_{\theta \in \Theta} f_\theta(\underline{x})$

Thus  $H_0$  is rejected when  $\lambda < c$ , where  $c$  is determined by the level of the test.

i.e.  $\sup_{\theta \in \Theta_0} P_\theta(\lambda < c) \leq \alpha$ . Since  $0 < \lambda < 1$

So in  $c$ .