"It is a mistake to confound strangeness with mystery." Sherlock Homes (A Study in Scarlet)

### Definition

- 1. Any statement about the unknown parameter  $oldsymbol{ heta}$  is called a *hypothesis*
- 2. One of the complementary hypothesis is called *Null Hypothesis* (denoted by  $H_0$ ) and other is called *Alternative Hypothesis* (denoted by  $H_1$  or  $H_A$ ).

ightharpoonup Example:  $X_1, \cdots, X_n \stackrel{iid}{\sim} N(\theta_1, \sigma^2)$  - regular diet program,  $Y_1, \cdots, Y_n \stackrel{iid}{\sim} N(\theta_2, \sigma^2)$  - Caloric restricted diet program

$$H_0: \theta_1 = \theta_2$$
 vs  $H_1: \theta_1 \geq \theta_2$ 

 $\triangleright$  Note:  $\Theta_0$  and  $\Theta_1$  are often called *Null* and *Alternative* space of parameter and the hypotheses are expressed as

$$H_0: \theta \in \Theta_0$$
 vs  $H_1: \theta \in \Theta_1$ 

### Definition

A hypothesis that completely specifies the distribution of  $X_1, \dots, X_n$  is called a *simple hypothesis* otherwise it is called *composite hypothesis*.

- $\triangleright$  Example:  $\theta_1 = \theta_2$ ,  $\theta_1 = \theta_2 = 2$ ,  $\theta_1 > \theta_2$ .
  - After observing  $X_1 = x_1, \dots, X_n = x_n$ , we need to decide which hypothesis,  $H_0$  or  $H_1$ , we will accept. Let  $\mathfrak{X}$  denote the set of all possible realization of  $X_1, \dots, X_n$ . Testing function (rule) plays the same role as estimator in point estimation.

### Definition

- 1. A function  $\phi: \mathfrak{X} \to [0,1]$  is called a *testing function*.
- 2. If a testing function takes a values in  $\{0,1\}$ , i.e.  $\phi:\mathfrak{X}\to\{0,1\}$ , it is called a *simple testing function*.

 $\triangleright$  Note: The interpretation of definition 1 is that after observing  $X_1=x_1,\cdots,X_n=x_n$ , reject  $H_0$  with probability  $\phi(x_1,\cdots,x_n)$  and accept  $H_0$  with probability  $1-\phi(x_1,\cdots,x_n)$ . This is called a randomized procedure.

### Definition

- $R_{\phi} = \{\mathbf{x} : \phi(\mathbf{x}) = 1\}$  is called the *rejection region* or *critical region*
- $A_{\phi} = \{ \mathbf{x} : \phi(\mathbf{x}) = 0 \}$  is called the *acceptance region*

Finding test - LRT

### Definition

Let  $X_1, \dots, X_n$  have joint pdf/pmf  $f(\mathbf{x}|\theta)$ ,  $\theta \in \Theta$ . Let  $\Theta_0$  be a proper subset of  $\Theta$ . Define the likelihood ratio

$$\lambda(\mathbf{x}) = \frac{\sup_{\theta \in \Theta_0} f(\mathbf{x}|\theta)}{\sup_{\theta \in \Theta} f(\mathbf{x}|\theta)}.$$

Then the Likelihood Ratio Test (LRT) of size  $\alpha$  for testing  $H_0: \theta \in \Theta_0$  vs  $H_1: \theta \in \Theta_0^c$  is

$$\phi(\mathbf{x}) = \begin{cases} 1, & \lambda(\mathbf{x}) < k, \\ \gamma, & \lambda(\mathbf{x}) = k, \\ 0, & \lambda(\mathbf{x}) > k, \end{cases}$$

where k and  $\gamma$  satisfy  $\sup_{\theta \in \Theta_0} E_{\theta}[\phi(\mathbf{x})] = \alpha$ .

Finding test - LRT

### Note:

1. Let  $\hat{\theta}_0$  be the MLE of  $\theta$  under  $H_0$  and  $\hat{\theta}$  be the MLE of  $\theta$  without any restriction. Then,

$$\lambda(\mathbf{x}) = \frac{f(\mathbf{x}|\hat{\theta}_0)}{f(\mathbf{x}|\hat{\theta})}.$$

2.  $0 \le \lambda(x) \le 1$ .

ightharpoonup Example 8.2.2:  $X_1, \dots, X_n \stackrel{iid}{\sim} N(\theta, \sigma^2)$ ,  $\sigma^2$  is known.

$$H_0: \theta = \theta_0$$
 vs  $H_1: \theta \neq \theta_0$ 

Finding test - LRT

ightharpoonup Example 8.2.6:  $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$ ,  $\mu$  and  $\sigma^2$  are unknown.

$$H_0: \mu = \mu_0$$
 vs  $H_1: \mu 
eq \mu_0$ 

Finding test - LRT

### Theorem

 $X_1, \dots, X_n \stackrel{iid}{\sim} f(x|\theta)$ .  $\lambda(\mathbf{x})$  is a likelihood ratio for testing

$$H_0: \theta \in \Theta_0 \quad \textit{vs} \quad H_1: \theta \in \Theta_0^c$$

Then under the regularity conditions (CRLB) on  $f(x|\theta)$  and  $H_0$ 

$$-2\ln[\lambda(\mathbf{x})] \stackrel{D}{\to} \chi_k^2,$$

where k = # of free parameters for  $\theta \in \Theta$  - # of free parameters for  $\theta \in \Theta_0$ . This yields the approximate size  $\alpha$  test

$$\phi(\mathbf{x}) = \begin{cases} 1, & -2\ln[\lambda(\mathbf{x})] > \chi^2_{1-\alpha,k}, \\ \gamma, & -2\ln[\lambda(\mathbf{x})] = \chi^2_{1-\alpha,k}, \\ 0, & -2\ln[\lambda(\mathbf{x})] < \chi^2_{1-\alpha,k}. \end{cases}$$

Evaluating the test

*Q:* How to compare several testing function ? or How to construct a good testing functions?

Errors in Testing

		True status of Nature	
		H₀ is true	$H_1$ is true
Action	Accept H <sub>0</sub>	O.K.	Type II error
	Reject $H_0$	Type I error	O.K.

- Type I error: Reject  $H_0$  when  $H_0$  is true
- Type II error: Accept  $H_0$  when  $H_0$  is false

Evaluating the test

### Definition

The power function  $\beta_{\phi}(\theta)$  of a test  $\phi(\mathbf{x})$  is the function defined as

$$\beta_{\phi}(\theta) = P_{\theta}[\phi(\mathbf{X}) = 1] = E_{\theta}[\phi(\mathbf{X})] = P_{\theta}(\mathbf{X} \in R_{\phi})$$

### Note:

- ▶  $\sup_{\theta \in \Theta_0} \beta_{\phi}(\theta)$  is called the *size of the test*  $\phi$ . Thus, any test such that  $\sup_{\theta \in \Theta_0} \beta_{\phi}(\theta) = \alpha$  is called as a *size*  $\alpha$  *test*.
- ▶ Test  $\phi$  such that  $\sup_{\theta \in \Theta_0} \beta_{\phi}(\theta) \leq \alpha$  is called a *level*  $\alpha$  *test*.
- ▶  $\theta \in \Theta_1$ ,  $\beta_{\phi}(\theta) = 1 Pr[Type \ II \ error]$ .

### Evaluating the test

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

$$H_0: \theta = \theta_0$$
 vs  $H_1: \theta > \theta_0$ 

Consider a test function of size  $\alpha = 0.10$ .

$$\phi(\mathbf{x}) = \begin{cases} 1 & \bar{x} > \theta_0 + c\sigma/\sqrt{n} \\ 0 & elsewhere. \end{cases}$$

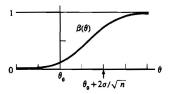
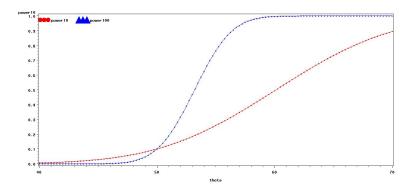


Figure 8.3.2. Power function for Example 8.3.3

Evaluating the test



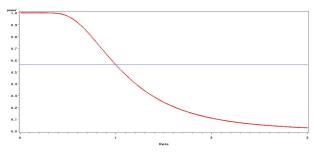
### Evaluating the test

 $\triangleright$  Example:  $X_1, \dots, X_n \stackrel{iid}{\sim}$  exponential( $\theta$ ).

$$H_0: \theta \geq 1$$
 vs  $H_1: \theta < 1$ 

Consider a test function

$$\phi(\mathbf{x}) = egin{cases} 1 & ar{x} < 1 \ 0 & \textit{elsewhere}. \end{cases}$$



Evaluating the test - MP test

### Definition

A test function  $\phi[\mathbf{X}=(X_1,\cdots,X_n)]$  is said to be the *most* powerful test of size  $\alpha$  for testing

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

if

- 1.  $E_{\theta_0}[\phi(\mathbf{X})] = \alpha$ ,  $[\beta_{\phi}(\theta_0) = \alpha]$
- 2. for any other test function  $\tilde{\phi}(\mathbf{X})$  with  $E_{\theta_0}[\tilde{\phi}(\mathbf{X})] \leq \alpha$ ,

$$E_{\theta_1}[\phi(\mathbf{X})] \ge E_{\theta_1}[\tilde{\phi}(\mathbf{X})], \ [\beta_{\phi}(\theta_1) \ge \beta_{\tilde{\phi}}(\theta_1)]$$

MP test has the smallest probability of type II error among all test rules with probability of type I error no bigger than  $\alpha$ .

Evaluating the test - MP test

ho Example:  $H_0: \theta = \theta_0$  vs  $H_1: \theta = \theta_1$ 

X = x	0	1	2
${p(x \theta_0)}$	0.05	0.05	0.90
$p(x \theta_1)$	0.90	0.08	0.02
$p(x \theta_1)/p(x \theta_0)$	18	1.6	0.022

Size  $\alpha = 0.05$  tests?

Find the MP test of size 0.05? Choose the test that has the largest/smallest ratio?

Evaluating the test - MP test

### Theorem (Neyman-Pearson Lemma)

 $X_1, \dots, X_n$  has a joint pdf/pmf  $f(\mathbf{x}|\theta)$ ,  $\theta \in \Theta$ . Consider the testing the hypotheses,

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

Then, for any  $0 \le \alpha \le 1$ , there exist a MP test of size  $\alpha$  given below;

$$\phi(\mathbf{x}) = \begin{cases} 1 & \text{if} \quad f(\mathbf{x}|\theta_1) > kf(\mathbf{x}|\theta_0), \\ \gamma & \text{if} \quad f(\mathbf{x}|\theta_1) = kf(\mathbf{x}|\theta_0), \\ 0 & \text{if} \quad f(\mathbf{x}|\theta_1) < kf(\mathbf{x}|\theta_0), \end{cases}$$

where the constants k and  $\gamma$  are chose to satisfy

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

Evaluating the test - MP test

#### Note:

1. The MP test  $\phi$  reject  $H_0$  if the likelihood ratio

$$L = \frac{f(\mathbf{x}|\theta_1)}{f(\mathbf{x}|\theta_0)}$$

is large.

- 2. In general, there may be more than one choice of k and  $\gamma$  that  $\beta_{\phi}(\theta_0) = \alpha$ . Then each is MP test of size  $\alpha$ .
- 3. When  $f(\mathbf{x}|\theta_1)/f(\mathbf{x}|\theta_0)$  has a continuous distribution under the null,  $H_0$ ,  $\gamma=0$  is usually taken and considered as the MP test of size  $\alpha$ .

Evaluating the test - MP test

ightharpoonup Example:  $X_1, \cdots, X_n \stackrel{iid}{\sim} \mathsf{Gamma}(3, \theta)$ .

$$H_0: \theta = \theta_0$$
 vs  $H_1: \theta = \theta_1(>\theta_0)$ 

Find the MP-test of size  $\alpha$ .

Evaluating the test - MP test

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

$$H_0: \mu = \mu_0$$
 vs  $H_1: \mu = \mu_1(>\mu_0)$ 

Find the MP-test of size  $\alpha$ .

Evaluating the test - UMP test

### Definition

Let  $f(\mathbf{x}|\theta)$ ,  $\theta \in \Theta$  be the joint pdf/pmf of  $X_1, \dots, X_n$ . Let  $\Theta_0$  and  $\Theta_1$  be the nonempty disjoint subsets of  $\Theta$ . A test rule  $\phi(\mathbf{x})$  is said to be an *uniformly most powerful (UMP)* test of size  $\alpha$  for testing

$$H_0: \theta \in \Theta_0 \quad \textit{vs} \quad H_1: \theta \in \Theta_1$$

if

- 1.  $\max_{\theta \in \Theta_0} E_{\theta}[\phi(\mathbf{X})] = \alpha$
- 2. for any other test  $\tilde{\phi}(\mathbf{x})$  with  $\max_{\theta \in \Theta_0} E_{\theta}[\tilde{\phi}(\mathbf{X})] \leq \alpha$ , we have

$$E_{\theta}[\phi(\mathbf{X})] \geq E_{\theta}[\tilde{\phi}(\mathbf{X})]$$

for each  $\theta \in \Theta_1$ .

Evaluating the test - UMP test

#### Note:

- 1. A UMP test has the smallest probability of type II error for every  $\theta \in \Theta_1$ among all the test with size  $\leq \alpha$ .
- 2. Condition 2 is a really strong requirement. Unlike the simple versus simple case, UMP test may not exist for composite  $H_0$  and for composite  $H_1$ .
- 3. NP lemma can be used to show that UMP test does not exist or identify the UMP test if it exists. HOW? (See next slide)

Evaluating the test - UMP test

- a. Fix  $\theta_0 \in \Theta_0$  appropriately (usually boundary of  $\Theta_0$ ).
- b. Choose any  $\theta_1 \in \Theta_1$
- c. Then find a MP test of size  $\alpha$ ,  $\phi(\mathbf{x})$ , for

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

lf

i  $\phi(\mathbf{x})$  does not depend on  $\theta_1$ 

ii  $\max_{\theta \in \Theta_0} E_{\theta}[\phi(\mathbf{X})] = \alpha$ 

then  $\phi(\mathbf{x})$  is the UMP-test of size  $\alpha$ .

Evaluating the test - UMP test

$$ightharpoonup$$
 Example:  $X_1,\cdots,X_n\stackrel{iid}{\sim} N(\mu,\sigma^2).$  
$$H_0:\mu=\mu_0\quad \textit{vs}\quad H_1:\mu>\mu_0$$
 
$$H_0:\mu\leq\mu_0\quad \textit{vs}\quad H_1:\mu>\mu_0$$

Evaluating the test - UMP test

$$ho$$
 Example:  $X_1, \cdots, X_n \stackrel{iid}{\sim} f(x|\lambda)$ . 
$$f(x|\lambda) = \lambda e^{-\lambda x}, \quad x > 0$$
 
$$H_0: \lambda \leq \lambda_0 \quad vs \quad H_1: \lambda > \lambda_0$$

Evaluating the test - UMP test

### Definition

Let  $f(\mathbf{x}|\theta)$ ,  $\theta \in \Theta$  be the joint pdf/pmf of  $X_1, \dots, X_n$ . The family is said to have *Monotone Likelihood Ratio (MLR)* in a statistic  $T(\mathbf{X})$  if, for all  $\theta'' > \theta'$ ,  $\theta'', \theta' \in \Theta$ , there exist a nondecreasing function of T, g, such that

$$L = \frac{f(\mathbf{x}|\theta'')}{f(\mathbf{x}|\theta')} = g_{\theta', \ \theta''}[T(\mathbf{x})]$$

in a support of x.

Note:

- ▶ if  $g_{\theta', \theta''}(x)$  is decreasing then  $g_{\theta', \theta''}(-x)$  is increasing.
- if  $f(\mathbf{x}|\theta'') > 0$  and  $f(\mathbf{x}|\theta') = 0$  then  $L = \infty$ .

Evaluating the test - UMP test

$$ightharpoonup$$
 Example:  $X_1, \cdots X_n \stackrel{iid}{\sim} f(x|\theta)$  
$$f(x|\theta) = c(\theta)h(x)\exp[w(\theta)t(x)]$$

Evaluating the test - UMP test

### **Theorem**

Let  $X_1, \dots, X_n$  have joint pdf/pmf  $f(\mathbf{x}|\theta)$ ,  $\theta \in \Theta$ . Assume the family has MLR in  $T(\mathbf{X})$ . Then

1. A UMP test of size  $\alpha$  for

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

is given by

$$\phi(\mathbf{x}) = \begin{cases} 1, & T(\mathbf{x}) > k, \\ \gamma, & T(\mathbf{x}) = k, \\ 0, & T(\mathbf{x}) < k, \end{cases}$$

where k and  $\gamma$  are determined by

$$P_{\theta_0}[T(\mathbf{X}) > k] + \gamma P_{\theta_0}[T(\mathbf{X}) = k] = \alpha.$$

Evaluating the test - UMP test

### Theorem (-Continued)

2. A UMP test of size  $\alpha$  for

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

is given by

$$\phi(\mathbf{x}) = \begin{cases} 1, & T(\mathbf{x}) < k, \\ \gamma, & T(\mathbf{x}) = k, \\ 0, & T(\mathbf{x}) > k, \end{cases}$$

where k and  $\gamma$  are determined by

$$P_{\theta_0}[T(\mathbf{X}) < k] + \gamma P_{\theta_0}[T(\mathbf{X}) = k] = \alpha.$$

Evaluating the test - UMP test

ightharpoonup Example:  $X_1, \dots, X_n \stackrel{iid}{\sim} \mathsf{Unif}[0, \theta], \ \theta > 0.$ 

$$H_0: \theta \leq \theta_0 \quad \textit{vs} \quad H_1: \theta > \theta_0$$

Find a UMP test of size  $\alpha$ .

Evaluating the test - UMP test

$$ightharpoonup$$
 Example:  $X_1,\cdots,X_n \stackrel{iid}{\sim} f(x|\eta)$  
$$f(x|\eta) = e^{-(x-\eta)}, \quad x > \eta.$$
 
$$H_0: \eta \leq \eta_0 \quad vs \quad H_1: \eta > \eta_0$$

Find a UMP test of size  $\alpha$ .