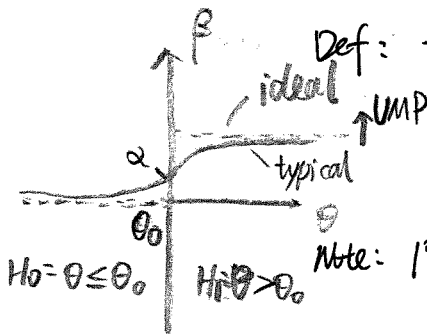


# IV

## Error Probabilities and Most Powerful Tests

Def: Type I Error means incorrectly rejecting  $H_0$  ;  
Type II Error means incorrectly accepting  $H_0$  .

Note: 1° With the above def. , the prob of Type I Error is  $P_\theta(\underline{X} \in R)$  ,  $\theta \in \Theta_0$  ; the prob of Type II Error is  $1 - P_\theta(\underline{X} \in R)$  ,  $\theta \in \Theta_0^c$  .



Def: The power function of a hypothesis test with rejection region  $R$  is

$$\beta(\theta) \equiv P_\theta(\underline{X} \in R)$$

Note: 1° The ideal power fn. is  $\beta(\theta) = \begin{cases} 0, & \theta \in \Theta_0 \\ 1, & \theta \in \Theta_0^c \end{cases}$  , which cannot be attained in most cases. A good test should have a power fn. as "close" to the ideal as possible.

2° Typically, the power fn. depends on the sample size  $n$ , so it can help determine an appropriate sample size.

3° For a fixed sample size, it's usually impossible to make both types of error prob. arbitrarily small. Naturally, we consider successive optimization.

Def: A test with power fn.  $\beta(\theta)$  is a size  $\alpha$  test, if  $\sup_{\theta \in \Theta_0} \beta(\theta) = \alpha$  ,  $0 < \alpha \leq 1$  .

Def: A ~~test~~ test with power fn.  $\beta(\theta)$  is a level  $\alpha$  test, if  $\sup_{\theta \in \Theta_0} \beta(\theta) \leq \alpha$  ,  $0 < \alpha \leq 1$  .

Note: 1° Typical choices of  $\alpha$  are 0.01, 0.05, 0.10.

2° For LRTs, a monotonic relation can be built between critical value  $c$  and size  $\alpha$ . In general

$$\sup_{\theta \in \Theta_0} P_{\theta}(\lambda(\underline{X}) \leq c) = \alpha$$

3° Cutoff points:

- $z_{\alpha}$  —  $P(Z > z_{\alpha}) = \alpha$ ,  $Z \sim N(0,1)$
- $t_{n-1, \frac{\alpha}{2}}$  —  $P(T_{n-1} > t_{n-1, \frac{\alpha}{2}}) = \frac{\alpha}{2}$ ,  $T_{n-1} \sim t_{n-1}$
- $\chi_{p, 1-\alpha}^2$  —  $P(\chi_p^2 > \chi_{p, 1-\alpha}^2) = 1-\alpha$ , ~~scribble~~

Def: A test with power fn.  $\beta(\theta)$  is unbiased if  $\beta(\theta') \geq \beta(\theta'')$ ,  $\forall \theta' \in \Theta_0^c$ ,  $\forall \theta'' \in \Theta_0$ .

Def: Given hypotheses  $H_0: \theta \in \Theta_0$ ;  $H_1: \theta \in \Theta_0^c$ ,  $C$  is a class of tests. A test in  $C$  with power fn.  $\beta(\theta)$  is a uniformly most powerful (UMP) class  $C$  test, if

$$\beta(\theta) \geq \beta'(\theta), \forall \theta \in \Theta_0^c, \forall \beta'(\theta) \text{ of a test in } C.$$

Note: 1° UMP tests do not exist in many realistic problems.

2° A hypothesis is simple, if it corresponds to only one value of  $\theta$ ; alternatively, a composite hypothesis corresponds to more than one value of  $\theta$ .

Thm: (Neyman-Pearson Lemma)

Given hypotheses  $H_0: \theta = \theta_0$ ;  $H_1: \theta = \theta_1$ ,

Conditions: and  $\underline{X} \sim f(x|\theta_i), (i=0,1)$ .

(1) A test has rejection region and acceptance region:

$$R = \{x \mid \text{scribble} f(x|\theta_1) > k f(x|\theta_0)\}$$

$$R^c = \{x \mid \text{scribble} f(x|\theta_1) < k f(x|\theta_0)\}$$

where  $k > 0$ .

(2) ~~condition~~ ~~condition~~ ~~condition~~  $P_{\theta_0}(X \in R) = \alpha$  ~~condition~~

Then:

a) (Sufficiency)

Any test that satisfies (1) and (2) is a UMP level  $\alpha$  test.

b) (Necessity)

If there exists a test satisfying (1) and (2) with  $k > 0$ , then every UMP level  $\alpha$  test is a size  $\alpha$  test, i.e. satisfies (2), and every UMP level  $\alpha$  test satisfies (1) except for a measure-zero set on sample space, for both  $\theta_0$  and  $\theta_1$ .

Note: 1° Condition (2) simply says the test is size  $\alpha$ .

2°  $R \cup R^c$  is not equal to the sample space, which is such stated for convenience of proof, but it's understandable.

Eg: 1° (UMP binomial test)

Given population  $X \sim \text{Binomial}(2, \theta)$

Hypotheses are:  $H_0: \theta = \frac{1}{2}$ ;  $H_1: \theta = \frac{3}{4}$ .

The ratios of pmfs are:

$$\frac{P(0|\theta=\frac{3}{4})}{P(0|\theta=\frac{1}{2})} = \frac{1}{4}, \quad \frac{P(1|\theta=\frac{3}{4})}{P(1|\theta=\frac{1}{2})} = \frac{3}{4}, \quad \frac{P(2|\theta=\frac{3}{4})}{P(2|\theta=\frac{1}{2})} = \frac{9}{4}$$

The test satisfying condition (1) in Neyman-Pearson Lemma with:

1°  $\frac{3}{4} < k < \frac{9}{4}$ , then  $R = \{2\}$ ,  $\alpha = P(2|\theta=\frac{1}{2}) = \frac{1}{4}$ .

So the test is a UMP level  $\frac{1}{4}$  test. (And the only one)

2°  $\frac{1}{4} < k < \frac{3}{4}$ , then  $R = \{1, 2\}$ ,  $\alpha = P(\{1, 2\}|\theta=\frac{1}{2}) = \frac{3}{4}$ .

So the test is a UMP level  $\frac{3}{4}$  test. (And the only one)

3°  $0 < k < \frac{1}{4}$ , then  $R = \{0, 1, 2\}$ ,  $\alpha = 1$ .

4°  $k > \frac{9}{4}$ , then  $R = \emptyset$ ,  $\alpha = 0$ .

2° (UMP normal test)

Given population  $X \sim N(\theta, \sigma^2)$  with  $\sigma^2$  known.

Hypotheses are:  $H_0: \theta = \theta_0$ ;  $H_1: \theta = \theta_1$  ( $\theta_0 > \theta_1$ )

We know  $\bar{X}$  is a sufficient stat. for  $\theta$ , and <sup>the</sup> ratio of pdfs is.

$$\frac{f_{\bar{X}}(\bar{x} | \theta = \theta_1)}{f_{\bar{X}}(\bar{x} | \theta = \theta_0)} = \exp\left\{-\frac{n}{2\sigma^2} \left[ 2(\theta_0 - \theta_1)\bar{x} + (\theta_1^2 - \theta_0^2) \right]\right\}$$

The test based on  $\bar{X}$  that satisfies condition (1) in Neyman-Pearson lemma gives rejection region:

$$S = \left\{ \bar{x} \mid \bar{x} < \frac{-\sigma^2 \ln k}{n(\theta_0 - \theta_1)} + \frac{\theta_0 + \theta_1}{2} \right\} \quad (k > 0)$$

and it has size

$$\alpha = P_{\theta_0}(\bar{X} \in S) = \Phi\left(-\frac{\sigma \ln k}{\sqrt{n}(\theta_0 - \theta_1)} - \frac{\sqrt{n}(\theta_0 - \theta_1)}{2\sigma}\right)$$

By corollary of Neyman-Pearson lemma, this test is a UMP level  $\alpha$  test.

Note: ~~1° Hypotheses that specify only one possible distn. for sample  $X$  are called simple hypotheses;~~

1° Hypotheses like  $H: \theta \geq \theta_0$  or  $H: \theta < \theta_0$  are called one-sided hypotheses; Hypotheses that assert either a parameter is large or small, like  $H: \theta \neq \theta_0$ , are called two-sided hypotheses.

Def: r.v.  $T \sim g(t|\theta)$ ,  $\theta \in \Theta \subseteq \mathbb{R}$  has a monotone likelihood ratio (MLR), if

$$\forall \theta_2 > \theta_1, \frac{g(t|\theta_2)}{g(t|\theta_1)} \text{ is monotone on the union of supports of } g(t|\theta_2) \text{ and } g(t|\theta_1).$$

where  $\frac{c}{0}$  is defined as  $+\infty$  if  $c > 0$ .

Note: 1° ~~Normal~~ Gaussian (known variance), Poisson, and binomial have an MLR.

2° Any exponential family with  $g(t|\theta) = h(t)c(\theta)e^{w(\theta)t}$  has an MLR if  $w(\theta)$  is nondecreasing.

Thm: (Karlin-Rubin)

Suppose  $T$  is a sufficient stat. for  $\theta$ ,  $T \sim g(t|\theta)$  has a nondecreasing MLR.

Hypotheses are  $H_0: \theta \leq \theta_0$ ,  $H_1: \theta > \theta_0$

Then  $\forall t_0$ , the test with rejection region  $R = \{x: T(x) > t_0\}$  is a UMP level  $\alpha = P_{\theta_0}(T > t_0)$  test.

Eg: 1° (UMP normal test)

Population  $X \sim N(\theta, \sigma^2)$ ,  $\sigma^2$  known,  $\Theta = \mathbb{R}$ .

Hypotheses are  $H_0: \theta \geq \theta_0$ ;  $H_1: \theta < \theta_0$ .

Since  $\bar{X}$  is sufficient for  $\theta$ , and has an MLR, the Karlin-Rubin thm give that the test with

$$R = \{x: \bar{x} < -\frac{\sigma^2 z_{\alpha}}{\sqrt{n}} + \theta_0\}$$

is a UMP level  $\alpha$  test.