

To find the most powerful test:-

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likelihood ratio test ( Sometimes in the literature, it is called Neyman-Pearson Theorem).

(To address the question:- What should be the form of the rejection region).

A test will reject  $H_0$  if

$$\lambda = \frac{\sup_{\theta \in \mathcal{H}_0} L(\theta \mid x_1, \dots, x_n)}{\sup_{\theta \in \mathcal{H}_1} L(\theta \mid x_1, \dots, x_n)}$$

$\mathcal{H}_0$ : Parameter space under  $H_0$

$\mathcal{H}_1$ : Parameter space under  $H_1$

The MP test at level of significance  $= \alpha$  is

$$\lambda < k, \text{ where } k \text{ is such that } P(\lambda^* < k \mid \theta \in \mathcal{H}_0) \leq \alpha.$$

A few remarks! -

i) Smaller the ration, more likely to reject  $H_0$ .

ii) LRT is the most Powerful test at a given  $\alpha$ .

Recall the first example (Urn containing 10 balls, (Wh & Re)  
5 or 7 of them are White.

$H_0$ :  $p_0 = 0.5$  ag.  $H_1$ :  $p_1 = 0.7$ . ( $p_1 > p_0$ ).

$$\begin{aligned} \text{Consider } \Lambda &= \frac{L(0.5 | x_1, \dots, x_n)}{L(0.7 | x_1, \dots, x_n)} \\ &= \frac{L(p_0 | x_1, \dots, x_n)}{L(p_1 | x_1, \dots, x_n)} = \frac{p_0^{\sum_{i=1}^n x_i} (1-p_0)^{n-\sum_{i=1}^n x_i}}{p_1^{\sum_{i=1}^n x_i} (1-p_1)^{n-\sum_{i=1}^n x_i}} \end{aligned}$$

$$= \left\{ \frac{p_0(1-p_1)}{p_1(1-p_0)} \right\}^{\sum_{i=1}^n x_i} \left( \frac{1-p_0}{1-p_1} \right)^n$$

Observation :-

$$p_0 < p_1 \Rightarrow 1 - p_0 > 1 - p_1 \Rightarrow \left( \frac{1-p_0}{1-p_1} \right) > 1$$

L (\*)

$$\Rightarrow$$

$$p_0 < p_1 \quad | - p_1 < | - p_0$$

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$$\textcircled{1} \wedge \textcircled{2} \Rightarrow p_0(1-p_1) < p_1(1-p_0)$$

$$\Leftrightarrow \frac{p_0(1-p_1)}{p_1(1-p_0)} < 1 \quad \text{---} \quad \text{**}$$

Hence, The LRT test will be

$$R = \left\{ \frac{p_0(1-p_1)}{p_1(1-p_0)} \right\}^{\sum_{i=1}^{x_i}} \left( \frac{1-p_0}{1-p_1} \right)^n < k \}$$

$$\Leftrightarrow R = \left\{ \left\{ \frac{p_0(1-p_1)}{p_1(1-p_0)} \right\} \sum_{i=1}^n x_i < K^* \right\}$$

Rejection region

$$\Leftrightarrow R = \left\{ \sum_{i=1}^n x_i \text{ key } \frac{p_0(1-p_1)}{p_1(1-p_0)} < \log K^* \right\}$$

$$\Leftrightarrow R = \left\{ \sum_{i=1}^n x_i > K^{***} \right\}, \text{ when } K^{***} = \frac{K^*}{\text{key } \frac{p_0(1-p_1)}{p_1(1-p_0)}}$$

And  $K^{***}$  is such that

$$P \left[ \sum_{i=1}^n x_i > K^{***} \mid p = 0.5 \right] \leq \alpha.$$

Another example :-

Suppose  $x_1, \dots, x_n \sim N(\mu, \sigma^2)$   $\hookrightarrow \sigma^2$  is known.

find the MP

Want to test  $H_0: \mu = \mu_0$  ag.  $H_1: \mu = \mu_1$  (suppose  $\mu_1 > \mu_0$ ).

Consider the LRT

$$\Lambda = \frac{L(\mu_0 | x_1, \dots, x_n)}{L(\mu_1 | x_1, \dots, x_n)} = \frac{\left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu_0)^2}}{\left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu_1)^2}}$$
$$= \exp\left(-\frac{1}{2\sigma^2} \left\{ \sum_{i=1}^n (x_i - \mu_0)^2 - \sum_{i=1}^n (x_i - \mu_1)^2 \right\}\right)$$

Now, the rejection region of the MP test will be

$$R = \{ \lambda < k \}$$

$$\Rightarrow R = \left\{ \exp \left( -\frac{1}{2\sigma^2} \left( \sum_{i=1}^n (x_i - \mu_0)^2 - \sum_{i=1}^n (x_i - \mu_1)^2 \right) \right) < k \right\}.$$

$$\Rightarrow R = \left\{ \sum_{i=1}^n (x_i - \mu_0)^2 - \sum_{i=1}^n (x_i - \mu_1)^2 > k^* \right\}$$

$$\Rightarrow R = \left\{ 2n \bar{x}_n (\mu_1 - \mu_0) + n (\mu_0^2 - \mu_1^2) > k^* \right\} .$$

$$\bar{x}_n = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\Rightarrow R = \left\{ 2n \bar{x}_n (\mu_1 - \mu_0) > k^{**} \right\}$$

$$\Rightarrow R = \left\{ \bar{x}_n > k^{***} \right\}, \text{ where } k^{***} \text{ is such that } P[\bar{x}_n > k^{***} | \mu = \mu_0] = \alpha.$$

Explicit expression of  $K^{**} = L$

$$\alpha = P \left[ \bar{X}_n > L \mid \mu = \mu_0 \right]$$

$$\Leftrightarrow P \left[ \frac{\bar{X}_n - \mu_0}{\sigma / \sqrt{n}} > \frac{L - \mu_0}{\sigma / \sqrt{n}} \right] = \alpha$$

$$\Leftrightarrow P \left[ Z > \frac{L - \mu_0}{\sigma / \sqrt{n}} \right] = \alpha, \text{ when } Z \sim N(0, 1).$$

$$\Leftrightarrow 1 - \underline{\Phi} \left( \frac{L - \mu_0}{\sigma / \sqrt{n}} \right) = \alpha.$$

check CDF of  $N(0, 1)$ .

$$\Leftrightarrow L = \mu_0 + \frac{\sigma}{\sqrt{n}} \Phi^{-1}(1 - \alpha).$$

p-value (Another way to summarize the performance of the test)!-

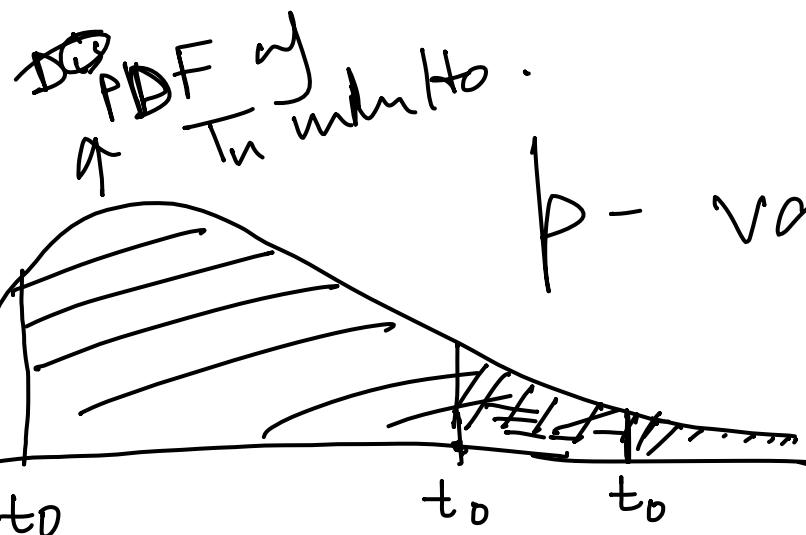
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p-value:-

The smallest significant level at which  $H_0$  would be rejected.

Let  $x_1, \dots, x_n$  be ~~an~~ r.v.s.

$t_0 = T_n(x_1, \dots, x_n)$ , where  $T_n$  is the test statistic.



$$\text{p-value} = P_{H_0}[T_n > t_0]$$

Recall the first example :-

$$p = 0.5 \text{ ag. } p = 0.7$$

Let  $X = 2$  (No. of white ball in 10 draws).

$$\begin{aligned} p\text{-value} &= P_{H_0} [X > 2] \\ &= P [X > 2 | p = 0.5] \\ &= 1 - P [X \leq 1 | p = 0.5] \end{aligned}$$

$$\text{cheK} = 0.9892$$

Under  $H_0$ , it is NOT rare event  
(Cannot disagree with the null).

