

Frequently Asked Questions

1. What is a Best Critical Region?

Answer:

Let a simple null hypothesis $H_0: \theta = \theta_0$ be tested against the simple alternative hypothesis $H_1: \theta = \theta_1$. Let C be a critical region for testing H_0 of size α against H_1 . C is called the best critical region of size α for testing H_0 against H_1 if C has at least the same power as any other critical region of size α that is if C is a critical region of size α and C' is any other critical region of size α then Power of C greater than or equal to Power of C' .

In testing of hypothesis we keep α the level of significance at a fixed level (say 0.05 or 0.01) and try to minimize the Type II error. The sample space may be partitioned into several ways so that each critical region, w has the same size α . Of all these critical regions choose that which has least type II error. This is called the best critical region of size α .

2. What do you mean by MP test?

Answer:

Among the critical regions of the same size α that which renders the minimum Type II error is called the most powerful critical region. The test based on the most powerful critical region is called the most powerful test. Therefore among all tests possessing the same size of Type I error, choose one for which the size of the Type II error is small as possible. This test is called the Most Powerful test.

3. State the Neyman- Pearson Lemma for finding out the best test.

Answer:

Let the probability density function of the population be $f(x, \theta)$ where θ represents the parameters. Draw a sample (x_1, x_2, \dots, x_n) from this population. Let L be the likelihood function then $L = f(x_1, x_2, \dots, x_n)$.

Let L_1 stands for the likelihood function when H_1 is true and L_0 stands for the likelihood function when H_0 is true (H_0 and H_1 are simple hypothesis)

Let α be the level of significance. Let k be any constant such that the size of the critical region defined as below is α

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} > k \right\}$$

4. Briefly explain the Procedure to obtain a Best Critical Region.

Answer:

- Obtain likelihood functions under null hypothesis and alternative hypothesis and call them as L_0 and L_1
- Obtain the ratio L_1/L_0
- Obtain the critical region as a region satisfying the condition $L_1 / L_0 > k$, where k is determined such that Probability of x belongs to the above mentioned Critical region under null hypothesis is equal to the size of the test

5. Find the BCR of size α for testing $H_0: \mu = \mu_0$ against $H_1: \mu = \mu_1 (> \mu_0)$ when a sample of size n is drawn from the Normal population with an known variance σ^2

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from a Normal population with parameters μ and σ^2

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(x_i, \mu)}{L_0(x_i, \mu)} > k \right\}$$

$$P[x \in C / H_0] = \alpha \text{ --- (*)}$$

Where k is such that

$$\frac{L_1(x_i, \mu)}{L_0(x_i, \mu)} = \frac{\left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n e^{\frac{-1}{2\sigma^2} \sum_i (x_i - \mu_1)^2}}{\left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n e^{\frac{-1}{2\sigma^2} \sum_i (x_i - \mu_0)^2}} = e^{\frac{-1}{2\sigma^2} [\sum_i (x_i - \mu_1)^2 - \sum_i (x_i - \mu_0)^2]}$$

$$\Rightarrow e^{\frac{-1}{2\sigma^2} [\sum_i x_i^2 + n\mu_1^2 - 2\mu_1 \sum_i x_i - (\sum_i x_i^2 + n\mu_0^2 - 2\mu_0 \sum_i x_i)]}$$

$$\Rightarrow e^{\frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2 \sum_i x_i(\mu_1 - \mu_0)]}$$

$$\frac{L_1(x_i, \mu)}{L_0(x_i, \mu)} > k \Rightarrow e^{\frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2 \sum_i x_i(\mu_1 - \mu_0)]} > k$$

Now

$$\Rightarrow \frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2 \sum_i x_i(\mu_1 - \mu_0)] > \ln k$$

$$\Rightarrow \frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2n\bar{x}(\mu_1 - \mu_0)] > \ln k$$

$$\Rightarrow \bar{x}(\mu_1 - \mu_0) > [\ln k + \frac{n}{2\sigma^2} (\mu_1^2 - \mu_0^2)] \frac{\sigma^2}{n} = \lambda_1 \text{ (say)}$$

When $\mu_1 > \mu_0$

$$\frac{L_1(x_i, \mu)}{L_0(x_i, \mu)} > k \Rightarrow \bar{x} > \frac{\lambda_1}{(\mu_1 - \mu_0)} = \lambda_2$$

$$\frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} > \frac{\lambda_1 - \mu_0}{(\mu_1 - \mu_0)} = \lambda_2 \text{ --- (1)}$$

When x_i follows Normal with parameters

μ and σ^2 , a sample mean $\bar{x} \sim N(\mu, \sigma^2 / n)$. Then $\frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} \sim N(0,1)$ under Null hypothesis

By substituting equation (1) in equation (*) we get

$$P[x \in C / H_0] = \alpha \Rightarrow P\left[\frac{L_1(xi, \mu)}{L_0(xi, \mu)} > k / H_0\right] = \alpha$$

$$\Rightarrow P\left[\frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} > \lambda_2\right] = \alpha \Rightarrow P[Z > \lambda_2] = \alpha$$

Now from the table of Normal Probabilities λ_2 can be read . Using the relation among λ_1, λ_2 and k , k can be determined. Hence the BCR can be found

The BCR when $\mu_1 > \mu_0$ is given by $C = \{(x_1, x_2, \dots, x_n) : Z > \lambda_2 = Z_{\alpha/2}\}$

6. Find the BCR of size α for testing $H_0: \mu = \mu_0$ against $H_1: \mu = \mu_1 (< \mu_0)$ when a sample of size n is drawn from the Normal population with an known variance σ^2

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from a Normal population with parameters μ and σ^2

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(xi, \mu)}{L_0(xi, \mu)} > k \right\}$$

$$P[x \in C / H_0] = \alpha \text{-----} (*)$$

Where k is such that

$$\frac{L_1(xi, \mu)}{L_0(xi, \mu)} = \frac{\left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n e^{\frac{-1}{2\sigma^2} \sum_i (xi - \mu_1)^2}}{\left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n e^{\frac{-1}{2\sigma^2} \sum_i (xi - \mu_0)^2}} = e^{\frac{-1}{2\sigma^2} [\sum_i (xi - \mu_1)^2 - \sum_i (xi - \mu_0)^2]}$$

$$\Rightarrow e^{\frac{-1}{2\sigma^2} [\sum_i xi^2 + n\mu_1^2 - 2\mu_1 \sum_i xi - (\sum_i xi^2 + n\mu_0^2 - 2\mu_0 \sum_i xi)]}$$

$$\Rightarrow e^{\frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2 \sum_i xi(\mu_1 - \mu_0)]}$$

$$\frac{L_1(xi, \mu)}{L_0(xi, \mu)} > k \Rightarrow e^{\frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2 \sum_i xi(\mu_1 - \mu_0)]} > k$$

Now

$$\Rightarrow \frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2 \sum_i xi(\mu_1 - \mu_0)] > \ln k$$

$$\Rightarrow \frac{-1}{2\sigma^2} [n(\mu_1^2 - \mu_0^2) - 2n\bar{x}(\mu_1 - \mu_0)] > \ln k$$

$$\Rightarrow \bar{x}(\mu_1 - \mu_0) > [\ln k + \frac{n}{2\sigma^2} (\mu_1^2 - \mu_0^2)] \frac{\sigma^2}{n} = \lambda_1 (\text{say})$$

When $\mu_1 < \mu_0$

$$\frac{L_1(xi, \mu)}{L_0(xi, \mu)} < k \Rightarrow \bar{x} < \frac{\lambda_1}{(\mu_1 - \mu_0)} = \lambda_2$$

$$\frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} < \frac{\lambda_1 - \mu_0}{(\mu_1 - \mu_0) / \sigma / \sqrt{n}} = \lambda_2 \text{ --- (2)}$$

By substituting equation (2) in equation (*) we get

$$P[x \in C / H_0] = \alpha \Rightarrow P\left[\frac{L_1(xi, \mu)}{L_0(xi, \mu)} < k / H_0\right] = \alpha$$

$$\Rightarrow P\left[\frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} < \lambda_2\right] = \alpha \Rightarrow P[Z < \lambda_2] = \alpha$$

When $\mu_1 < \mu_0$ the BCR of size α is given by $C = \{(x_1, x_2, \dots, x_n) : Z < \lambda_2 = -Z_{\alpha/2}\}$

7. Find the BCR of size α for testing $H_0: \sigma = \sigma_0$ against $H_1: \sigma = \sigma_1 (> \sigma_0)$ when a sample of size n is drawn from the Normal population with a known mean μ

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from a Normal population with parameters μ and σ^2

By NP Lemma a size α BCR is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k \right\}$$

$$P[x \in C / H_0] = \alpha \text{ --- (*)}$$

Where k is such that

$$\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} = \frac{\left(\frac{1}{\sigma_1^2}\right)^{n/2} \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{1}{2\sigma_1^2} \sum_i (xi - \mu)^2}}{\left(\frac{1}{\sigma_0^2}\right)^{n/2} \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{-\frac{1}{2\sigma_0^2} \sum_i (xi - \mu)^2}}$$

$$\Rightarrow \left(\frac{\sigma_0^2}{\sigma_1^2}\right)^{n/2} e^{\frac{-1}{2} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_0^2}\right) \sum_i (xi - \mu)^2}$$

$$\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k \Rightarrow \left(\frac{\sigma_0^2}{\sigma_1^2}\right)^{n/2} e^{\frac{-1}{2} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_0^2}\right) \sum_i (xi - \mu)^2} > k$$

Now

$$\Rightarrow \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right) - \frac{1}{2} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k$$

$$\Rightarrow \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right) - \frac{1}{2} \left(\frac{\sigma_0^2 - \sigma_1^2}{\sigma_1^2 \sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k$$

$$\Rightarrow \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right) + \frac{1}{2} \left(\frac{\sigma_1^2 - \sigma_0^2}{\sigma_1^2 \sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k$$

$$\Rightarrow \frac{1}{2} \left(\frac{\sigma_1^2 - \sigma_0^2}{\sigma_1^2 \sigma_0^2} \right) \sum_i (xi - \mu)^2 > \ln k - \frac{n}{2} \ln \left(\frac{\sigma_0^2}{\sigma_1^2} \right)$$

$$\Rightarrow (\sigma_1^2 - \sigma_0^2) \sum_i (xi - \mu)^2 > [\ln k - \frac{n}{2} \ln \left(\frac{\sigma_0^2}{\sigma_1^2} \right)] 2\sigma_1^2 \sigma_0^2 = \lambda_1 (\text{say})$$

When $\sigma_1 > \sigma_0$

$$\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k \Rightarrow \sum_i (xi - \mu)^2 > \lambda_1 / (\sigma_1^2 - \sigma_0^2)$$

$$\Rightarrow \frac{\sum_i (xi - \mu)^2}{\sigma_0^2} > \frac{\lambda_1 / (\sigma_1^2 - \sigma_0^2)}{\sigma_0^2} = \lambda_2 (\text{say})$$

$$\text{Now size of the test} = \alpha \text{ implies } P\left[\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k / \sigma = \sigma_0\right] = \alpha \text{-----} (*)$$

When xi follows Normal with parameters

μ and σ^2 , and when $\sigma = \sigma_0$, $\frac{\sum_i (xi - \mu)^2}{\sigma_0^2} \sim \chi^2(n)$. Then from the table of

probabilities of Chi square variable with n degrees of freedom we have

$$\lambda_2 = \chi^2_{\alpha}(n)$$

By substituting equation (1) in equation (*) we get

$$P[x \in C / H_0] = \alpha \Rightarrow P\left[\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k / \sigma = \sigma_0\right] = \alpha$$

$$\Rightarrow P\left[\frac{\sum_i (xi - \mu)^2}{\sigma_0^2} > \lambda_2\right] \Rightarrow P\left[\frac{\sum_i (xi - \mu)^2}{\sigma_0^2} > \lambda_2 = \chi^2_{\alpha}(n)\right] = \alpha$$

Now from the table of Chi square Probabilities λ_2 can be read. Using the relation among λ_1 , λ_2 and k , k can be determined. Hence the BCR can be found

$$\text{The BCR when } \sigma_1 > \sigma_0 \text{ is given by } C = \left\{ (x_1, x_2, \dots, x_n) : \frac{\sum_i (xi - \mu)^2}{\sigma_0^2} > \chi^2_{\alpha}(n) \right\}$$

8. Find the BCR of size α for testing $H_0: \sigma = \sigma_0$ against $H_1: \sigma = \sigma_1 (< \sigma_0)$ when a sample of size n is drawn from the Normal population with a known mean μ

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from a Normal population with parameters μ and σ^2

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k \right\}$$

$$P[x \in C / H_0] = \alpha \text{-----} (*)$$

Where k is such that

$$\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} = \frac{\left(\frac{1}{\sigma_1^2}\right)^{n/2} \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{\frac{-1}{2\sigma_1^2} \sum_i (xi - \mu)^2}}{\left(\frac{1}{\sigma_0^2}\right)^{n/2} \left(\frac{1}{\sqrt{2\pi}}\right)^n e^{\frac{-1}{2\sigma_0^2} \sum_i (xi - \mu)^2}}$$

$$\Rightarrow \left(\frac{\sigma_0^2}{\sigma_1^2}\right)^{n/2} e^{\frac{-1}{2} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_0^2}\right) \sum_i (xi - \mu)^2}$$

$$\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} > k \Rightarrow \left(\frac{\sigma_0^2}{\sigma_1^2}\right)^{n/2} e^{\frac{-1}{2} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_0^2}\right) \sum_i (xi - \mu)^2} > k$$

Now

$$\Rightarrow \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right) - \frac{1}{2} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k$$

$$\Rightarrow \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right) - \frac{1}{2} \left(\frac{\sigma_0^2 - \sigma_1^2}{\sigma_1^2 \sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k$$

$$\Rightarrow \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right) + \frac{1}{2} \left(\frac{\sigma_1^2 - \sigma_0^2}{\sigma_1^2 \sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k$$

$$\Rightarrow \frac{1}{2} \left(\frac{\sigma_1^2 - \sigma_0^2}{\sigma_1^2 \sigma_0^2}\right) \sum_i (xi - \mu)^2 > \ln k - \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right)$$

$$\Rightarrow (\sigma_1^2 - \sigma_0^2) \sum_i (xi - \mu)^2 > \left[\ln k - \frac{n}{2} \ln\left(\frac{\sigma_0^2}{\sigma_1^2}\right)\right] 2\sigma_1^2 \sigma_0^2 = \lambda_1 (\text{say})$$

When $\sigma_1 < \sigma_0$

$$\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} < k \Rightarrow \sum_i (xi - \mu)^2 < \lambda_1 / (\sigma_1^2 - \sigma_0^2)$$

$$\Rightarrow \frac{\sum_i (xi - \mu)^2}{\sigma_0^2} < \frac{\lambda_1 / (\sigma_1^2 - \sigma_0^2)}{\sigma_0^2} = \lambda_2 (\text{say})$$

Now size of the test $= \alpha$ implies $P\left[\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} < k / \sigma = \sigma_0\right] = \alpha \text{-----} (*)$

When x_i follows Normal with parameters

μ and σ^2 , and when $\sigma = \sigma_0$, $\frac{\sum (xi - \mu)^2}{\sigma_0^2} \sim \chi^2(n)$. Then from the table of probabilities of Chi square variable with n degrees of freedom we have $\lambda_2 = \chi^2_{1-\alpha}(n)$

By substituting equation (1) in equation (*) we get

$$P[x \in C / H_0] = \alpha \Rightarrow P\left[\frac{L_1(xi, \sigma^2)}{L_0(xi, \sigma^2)} < k / \sigma = \sigma_0\right] = \alpha$$

$$\Rightarrow P\left[\frac{\sum (xi - \mu)^2}{\sigma_0^2} < \lambda_2\right] \Rightarrow P\left[\frac{\sum (xi - \mu)^2}{\sigma_0^2} < \lambda_2 = \chi^2_{1-\alpha}(n)\right] = \alpha$$

Now from the table of Chi square Probabilities λ_2 can be read . Using the relation among λ_1, λ_2 and k , k can be determined. Hence the BCR can be found

The BCR when $\sigma_1 < \sigma_0$ is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{\sum (xi - \mu)^2}{\sigma_0^2} < \chi^2_{1-\alpha}(n) \right\}$$

9. Obtain a BCR of size α for testing $H_0: \lambda = \lambda_0$ against $H_1: \lambda = \lambda_1 (< \lambda_0)$ when a sample of size n is drawn from the Poisson population with unknown parameter λ ?

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from Poisson population with parameter λ

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} > k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$\frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} = \frac{\frac{e^{-n\lambda_1} \lambda_1^{\sum x_i}}{\prod_{i=1}^n x_i!}}{\frac{e^{-n\lambda_0} \lambda_0^{\sum x_i}}{\prod_{i=1}^n x_i!}} = e^{-n(\lambda_1 - \lambda_0)} \left(\frac{\lambda_1}{\lambda_0} \right)^{\sum x_i}$$

Now $\frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} > k \Rightarrow e^{-n(\lambda_1 - \lambda_0)} \left(\frac{\lambda_1}{\lambda_0} \right)^{\sum x_i} > k$

$$\Rightarrow -n(\lambda_1 - \lambda_0) + \sum_i x_i \ln \left(\frac{\lambda_1}{\lambda_0} \right) > \ln k$$

$$\Rightarrow \sum_i x_i \ln \left(\frac{\lambda_1}{\lambda_0} \right) > \ln k + n(\lambda_1 - \lambda_0) = k_1 \text{ (say)}$$

When $\lambda_1 < \lambda_0$

$$\frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} < k \Rightarrow \sum_i x_i < \frac{k_1}{\ln \left(\frac{\lambda_1}{\lambda_0} \right)} = k_2$$

The BCR when $\lambda_1 < \lambda_0$ is given by $C = \left\{ (x_1, x_2, \dots, x_n) : \sum_i x_i < k_2 \right\}$

10. If x_1, x_2, \dots, x_n is a random sample of size n from a distribution having probability density function of the form $f(x, \theta) = \theta x^{\theta-1}$. Obtain a BCR for testing $H_0: \theta=1$ against $H_1: \theta=2$

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from population with probability density function of the form $f(x, \theta) = \theta x^{\theta-1}$

Required to test $H_0: \theta=1$ against $H_1: \theta=2$

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(xi, \theta)}{L_0(xi, \theta)} \geq k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$L(xj, \theta) = \prod_j f(xj, \theta) = \theta^n \prod_j x_j^{\theta-1}$$

$$\frac{L_1(xi, \theta)}{L_0(xi, \theta)} = \frac{\theta_1^n \prod_j x_j^{\theta_1-1}}{\theta_0^n \prod_j x_j^{\theta_0-1}} = \frac{2^n \prod_j x_j^{2-1}}{1^n \prod_j x_j^{1-1}} = 2^n \prod_j x_j$$

$$\frac{L_1(xi, \theta)}{L_0(xi, \theta)} \geq k \Rightarrow 2^n \prod_j x_j \geq k \Rightarrow \prod_j x_j \geq \frac{k}{2^n} = c$$

$$\text{A size } \alpha \text{ BCR is given by } C = \left\{ (x_1, x_2, \dots, x_n) : \prod_j x_j \geq c \right\}$$

11. Find BCR of size α for testing $H_0: p=p_0$ against $H_1: p=p_1 (>p_0)$ when a sample of size m is drawn from a Binomial population with parameters n and p (n is known)

Answer:

Let x_1, x_2, \dots, x_m be a random sample of size n from Binomial population with parameters n and p

By NP Lemma a size α BCR is given by

$$C = \left\{ (x_1, x_2, \dots, x_m) : \frac{L_1(xi, p)}{L_0(xi, p)} > k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$\frac{L_1(xi, p)}{L_0(xi, p)} = \frac{(1-p_1)^{mn-\sum xi} p_1^{\sum xi}}{(1-p_0)^{mn-\sum xi} p_0^{\sum xi}} = \left(\frac{p_1}{p_0}\right)^{\sum xi} \left(\frac{1-p_1}{1-p_0}\right)^{mn-\sum xi}$$

$$= \left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right)^{\sum xi} \left(\frac{1-p_1}{1-p_0}\right)^{mn}$$

$$\text{Now } \frac{L_1(xi, p)}{L_0(xi, p)} > k \Rightarrow \left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right)^{\sum xi} \left(\frac{1-p_1}{1-p_0}\right)^{mn} > k$$

$$\sum xi \log\left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right) + mn \log\left(\frac{1-p_1}{1-p_0}\right) > \log k$$

$$\sum xi \log\left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right) > \log k - mn \log\left(\frac{1-p_1}{1-p_0}\right) = k_1(\text{say})$$

When $p_1 > p_0$

$$\frac{L_1(xi, p)}{L_0(xi, p)} > k \Rightarrow \sum xi > \frac{k_1}{\log\left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right)} = k_2$$

The BCR when $p_1 > p_0$ is given by $C = \left\{ (x_1, x_2, \dots, x_m) : \sum_i x_i > k_2 \right\}$

12. Find the BCR of size α for testing $H_0: \lambda = \lambda_0$ against $H_1: \lambda = \lambda_1 (> \lambda_0)$ when a sample of size n is drawn from the Poisson population with unknown parameter λ ?

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from Poisson population with parameter λ

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} > k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$\frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} = \frac{\frac{e^{-n\lambda_1} \lambda_1^{\sum x_i}}{\prod_{i=1}^n x_i!}}{\frac{e^{-n\lambda_0} \lambda_0^{\sum x_i}}{\prod_{i=1}^n x_i!}} = e^{-n(\lambda_1 - \lambda_0)} \left(\frac{\lambda_1}{\lambda_0} \right)^{\sum x_i}$$

$$\text{Now } \frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} > k \Rightarrow e^{-n(\lambda_1 - \lambda_0)} \left(\frac{\lambda_1}{\lambda_0} \right)^{\sum x_i} > k$$

$$\Rightarrow -n(\lambda_1 - \lambda_0) + \sum_i x_i \ln \left(\frac{\lambda_1}{\lambda_0} \right) > \ln k$$

$$\Rightarrow \sum_i x_i \ln \left(\frac{\lambda_1}{\lambda_0} \right) > \ln k + n(\lambda_1 - \lambda_0) = k_1 \text{ (say)}$$

When $\lambda_1 > \lambda_0$

$$\frac{L_1(xi, \lambda)}{L_0(xi, \lambda)} > k \Rightarrow \sum_i x_i > \frac{k_1}{\ln \left(\frac{\lambda_1}{\lambda_0} \right)} = k_2$$

The BCR when $\lambda_1 > \lambda_0$ is given by $C = \left\{ (x_1, x_2, \dots, x_n) : \sum_i x_i > k_2 \right\}$

13. Find BCR of size α for testing $H_0: p=p_0$ against $H_1: p=p_1 (<p_0)$ when a sample of size m is drawn from a Binomial population with parameters n and p (n is known)

Answer:

Let x_1, x_2, \dots, x_m be a random sample of size n from Binomial population with parameters n and p

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_m) : \frac{L_1(x_i, p)}{L_0(x_i, p)} > k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$\frac{L_1(x_i, p)}{L_0(x_i, p)} = \frac{(1-p_1)^{mn-\sum x_i} p_1^{\sum x_i}}{(1-p_0)^{mn-\sum x_i} p_0^{\sum x_i}} = \left(\frac{p_1}{p_0}\right)^{\sum x_i} \left(\frac{1-p_1}{1-p_0}\right)^{mn-\sum x_i}$$

$$= \left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right)^{\sum x_i} \left(\frac{1-p_1}{1-p_0}\right)^{mn}$$

Now $\frac{L_1(x_i, p)}{L_0(x_i, p)} > k \Rightarrow \left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right)^{\sum x_i} \left(\frac{1-p_1}{1-p_0}\right)^{mn} > k$

$$\sum x_i \log\left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right) + mn \log\left(\frac{1-p_1}{1-p_0}\right) > \log k$$

$$\sum x_i \log\left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right) > \log k - mn \log\left(\frac{1-p_1}{1-p_0}\right) = k_1(\text{say})$$

When $p_1 < p_0$

$$\frac{L_1(x_i, p)}{L_0(x_i, p)} < k \Rightarrow \sum x_i < \frac{k_1}{\log\left(\frac{p_1(1-p_0)}{p_0(1-p_1)}\right)} = k_2$$

The BCR when $p_1 > p_0$ is given by $C = \left\{ (x_1, x_2, \dots, x_m) : \sum_i x_i < k_2 \right\}$

14. If x_1, x_2, \dots, x_n is a random sample of size n from an exponential distribution having probability density function of the form $f(x, \theta) = \theta e^{-\theta x}$, $x > 0$. Derive BCR for testing $H_0: \theta = \theta_0$ against $H_1: \theta = \theta_1 (> \theta_0)$. Obtain the null distribution of the test statistic

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from population with probability density function of the form $f(x, \theta) = \theta e^{-\theta x}$, $x > 0$.

Required to test $H_0: \theta = \theta_0$ against $H_1: \theta = \theta_1$.

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} \geq k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$L(x_i, \theta) = \prod_i f(x_i, \theta) = \theta^n e^{-\theta \sum_i x_i}$$

$$\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} = \frac{\theta_1^n e^{-\theta_1 \sum_i x_i}}{\theta_0^n e^{-\theta_0 \sum_i x_i}} = \left(\frac{\theta_1}{\theta_0} \right)^n e^{-[\theta_1 - \theta_0] \sum_i x_i}$$

$$\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} > k \Rightarrow \left(\frac{\theta_1}{\theta_0} \right)^n e^{-[\theta_1 - \theta_0] \sum_i x_i} > k$$

$$\Rightarrow n \ln \left(\frac{\theta_1}{\theta_0} \right) - (\theta_1 - \theta_0) \sum_i x_i > \ln k$$

$$\Rightarrow (\theta_1 - \theta_0) \sum_i x_i \leq -\ln k + n \ln \left(\frac{\theta_1}{\theta_0} \right) = \lambda_1 (\text{say})$$

When $\theta_1 > \theta_0$

$$\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} > k \Rightarrow \sum_i x_i \leq \lambda_1 / (\theta_1 - \theta_0) = \lambda_2 (\text{say})$$

$$P\left[\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} > k / \theta = \theta_0\right] = \alpha \text{-----} (*)$$

Now size of the test $= \alpha$ implies

When x_i follows exponential with parameter θ
and when $\theta = \theta_0$, $\sum_i x_i \sim \gamma(n, \theta)$.

$$P[x \in C / H_0] = \alpha \Rightarrow P\left[\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} \geq k / \theta = \theta_0\right] = \alpha$$

$$\Rightarrow P\left[\sum_i x_i \leq \lambda_2\right] = \alpha$$

Now from the table of Chi square Probabilities λ_2 can be read. Using the relation among λ_1 , λ_2 and k , k can be determined. Hence the BCR can be found

The BCR when $\theta_1 > \theta_0$ is given by $C = \left\{ (x_1, x_2, \dots, x_n) : \sum_i x_i \leq \lambda_2 \right\}$

The null distribution of the test statistic is the distribution of $\sum_i x_i$ when $\theta = \theta_0$ is

nothing but the probability density function of Gamma distribution with parameters n

and θ_0 and is given by $f(y) = \frac{\theta_0^n}{\Gamma(n)} e^{-\theta_0 y} y^{n-1}, y \geq 0,$

15. If x_1, x_2, \dots, x_n is a random sample of size n from an exponential distribution having probability density function of the form $f(x, \theta) = \theta e^{-\theta x}, x > 0$. Derive BCR for testing $H_0: \theta = \theta_0$ against $H_1: \theta = \theta_1 (< \theta_0)$. Obtain the null distribution of the test statistic

Answer:

Let x_1, x_2, \dots, x_n be a random sample of size n from population with probability density function of the form $f(x, \theta) = \theta e^{-\theta x}, x > 0$.

Required to test $H_0: \theta = \theta_0$ against $H_1: \theta = \theta_1$.

By NP Lemma a size α BCR is given is given by

$$C = \left\{ (x_1, x_2, \dots, x_n) : \frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} \geq k \right\}$$

$$P[x \in C / H_0] = \alpha$$

Where k is such that

$$L(x_i, \theta) = \prod_i f(x_i, \theta) = \theta^n e^{-\theta \sum_i x_i}$$

$$\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} = \frac{\theta_1^n e^{-\theta_1 \sum_i x_i}}{\theta_0^n e^{-\theta_0 \sum_i x_i}} = \left(\frac{\theta_1}{\theta_0} \right)^n e^{-[\theta_1 - \theta_0] \sum_i x_i}$$

$$\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} > k \Rightarrow \left(\frac{\theta_1}{\theta_0} \right)^n e^{-[\theta_1 - \theta_0] \sum_i x_i} > k$$

$$\Rightarrow n \ln \left(\frac{\theta_1}{\theta_0} \right) - (\theta_1 - \theta_0) \sum_i x_i > \ln k$$

$$\Rightarrow (\theta_1 - \theta_0) \sum_i x_i \leq -\ln k + n \ln \left(\frac{\theta_1}{\theta_0} \right) = \lambda_1 (\text{say})$$

When $\theta_1 < \theta_0$

$$\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} < k \Rightarrow \sum_i x_i > \lambda_2 (\text{say})$$

Now size of the test $= \alpha$ implies $P\left[\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} < k / \theta = \theta_0\right] = \alpha \text{-----} (*)$

$$P[x \in C / H_0] = \alpha \Rightarrow P\left[\frac{L_1(x_i, \theta)}{L_0(x_i, \theta)} < k / \theta = \theta_0\right] = \alpha$$

$$\Rightarrow P\left[\sum_i x_i > \lambda_2\right] = \alpha$$

Hence the BCR can be found

$$\text{The BCR when } \theta_1 < \theta_0 \text{ is given by } C = \left\{ (x_1, x_2, \dots, x_n) : \sum_i x_i > \lambda_2 \right\}$$