

01:640:481 - Homework 7

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1. 12.14

A single observation of a random variable having a geometric distribution is to be used to test the null hypothesis that its parameter equals θ_0 against the alternative that it equals $\theta_1 > \theta_0$. Use the Neyman-Pearson lemma to find the best critical region of size α .

Solution: Let X be the random variable with the geometric distribution. The likelihood function is given by

$$L(\theta) = \theta(1 - \theta)^{x-1}$$

By NPL the ratio of the likelihoods is bounded by a constant k . That is,

$$\begin{aligned} \frac{L(\theta_0)}{L(\theta_1)} &= \left(\frac{\theta_0 \cdot (1 - \theta_0)^{x-1}}{\theta_1 \cdot (1 - \theta_1)^{x-1}} \right) \leq k \\ \ln \left(\frac{\theta_0 \cdot (1 - \theta_0)^{x-1}}{\theta_1 \cdot (1 - \theta_1)^{x-1}} \right) &\leq \ln(k) \\ x \cdot \ln \left(\frac{1 - \theta_0}{1 - \theta_1} \right) &\leq \ln(k) - \ln \left(\frac{\theta_0 \cdot (1 - \theta_0)^{-1}}{\theta_1 \cdot (1 - \theta_1)^{-1}} \right) \\ x &\leq \frac{\ln(k) - \ln \left(\frac{\theta_0 \cdot (1 - \theta_0)^{-1}}{\theta_1 \cdot (1 - \theta_1)^{-1}} \right)}{\ln \left(\frac{1 - \theta_0}{1 - \theta_1} \right)} \end{aligned}$$

Thus if we take the critical region to be $x \leq c$ where c is the above expression, we have the best critical region of size α .

2. 12.24

Given a random sample of size n from a normal population with unknown mean and variance, find an expression for the likelihood ratio statistic for testing the null hypothesis $\sigma = \sigma_0$ against the alternative hypothesis $\sigma \neq \sigma_0$.

Solution: Let X be the random variable with the normal distribution with a pdf of

$$f(x|\mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

The likelihood function is given by

$$L(\mu, \sigma) = \frac{1}{(2\pi)^{n/2} \sigma^n} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2}$$

The likelihood ratio statistic is given by $\Lambda = \frac{L(\mu, \sigma_0)}{L(\mu, \hat{\sigma}_{MLE})}$ where $\hat{\sigma}_{MLE}$ is the maximum

likelihood estimator of σ . The MLE of σ is $\hat{\sigma}_{MLE} = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2}$ Thus

$$\begin{aligned}\Lambda &= \frac{\sigma_{MLE}}{\sigma_0} e^{-\frac{1}{2\sigma_{MLE}^2} \sum_{i=1}^n (x_i - \mu)^2 + \frac{1}{2\sigma_0^2} \sum_{i=1}^n (x_i - \mu)^2} \\ &= \left(\frac{\sum_{i=1}^n (x_i - \bar{x})}{n\sigma_0^2} \right)^{n/2} e^{\frac{1}{2} \left[\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{\sigma_0^2} - n \right]}\end{aligned}$$

3. 13.1

Given a random sample of size n from a normal population with the known variance σ^2 , show that the null hypothesis $\mu = \mu_0$ can be tested against the alternative hypothesis $\mu \neq \mu_0$ with the use of a one-tailed criterion based on the chi-square distribution.

Solution: Let X be a random variable. The sample mean is given by \bar{X} and $Z = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$ is a standard normal random variable under the null hypothesis. $Z^2 \sim \chi_1^2$ under the null hypothesis. Thus by the 1 tail test

$$\frac{n(\bar{X} - \mu_0)^2}{\sigma^2} \geq \chi_{1,\alpha}^2$$

Where $\chi_{1,\alpha}^2$ is the critical value of the chi-square distribution with 1 degree of freedom at the α level of significance.