HW7

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$\mathbf{Q}\mathbf{1}$

Problem 8.6 a) - b), Casella and Berger (2nd Edition)

Suppose that we have two independent random samples: X_1, \ldots, X_n are exponential(θ), and Y_1, \ldots, Y_m are exponential(μ).

a)

Find the LRT of

$$H_0: \theta = \mu$$
 versus $H_1: \theta \neq \mu$.

Answer

The likelihood ratio test (LRT) statistic is:

$$\lambda(x, y) = \frac{\max_{\theta} L(\theta \mid x, y)}{\max_{\theta, \mu} L(\theta, \mu \mid x, y)}$$

Under H_0 ($\theta = \mu$):

The MLE for θ is obtained from the combined sample:

$$\hat{\theta}_0 = \frac{\sum_{i=1}^n X_i + \sum_{j=1}^m Y_j}{n+m}$$

Under the full model:

The MLEs are the sample means:

$$\hat{\theta} = \bar{X} = \frac{\sum X_i}{n}, \quad \hat{\mu} = \bar{Y} = \frac{\sum Y_j}{m}$$

Substituting the MLEs, we get:

$$\lambda(x,y) = \frac{(\hat{\theta}_0)^{-(n+m)} e^{-(n+m)}}{(\hat{\theta})^{-n} e^{-n} (\hat{\mu})^{-m} e^{-m}} = \frac{(\bar{X})^n (\bar{Y})^m}{\left(\sum_{n+m} \frac{X_i + \sum_j Y_j}{n+m}\right)^{n+m}}$$

Simplifying, this becomes:

$$\lambda(x,y) = \frac{(n+m)^{n+m} (\sum X_i)^n (\sum Y_j)^m}{n^n m^m (\sum X_i + \sum Y_j)^{n+m}}$$

Rejection Rule: Reject H_0 if $\lambda(x,y) \leq c$, where c is chosen for significance level α .

$$\varphi(x,y) = \begin{cases} 1 & \text{if } \lambda(x,y) \le c, \\ 0 & \text{otherwise,} \end{cases}$$

where

$$\lambda(x,y) = \frac{(n+m)^{n+m} (\sum X_i)^n (\sum Y_j)^m}{n^n m^m (\sum X_i + \sum Y_j)^{n+m}}$$

And c is chosen such that $\mathbb{P}(\varphi(X,Y)=1\mid H_0)=\alpha$.

b)

Show that the test in part a) can be based on the statistic

$$T = \frac{\sum X_i}{\sum X_i + \sum Y_i}.$$

Answer

Let
$$T = \frac{\sum X_i}{\sum X_i + \sum Y_j}$$
.

Rewriting $\lambda(x,y)$ in terms of T:

$$\lambda(x,y) = \frac{(n+m)^{n+m}}{n^n m^m} \left(\frac{\sum X_i}{\sum X_i + \sum Y_j} \right)^n \left(\frac{\sum Y_j}{\sum X_i + \sum Y_j} \right)^m = \frac{(n+m)^{n+m}}{n^n m^m} T^n (1-T)^m$$

Since $\lambda(x,y)$ depends on the data only through T, the LRT can be based entirely on T.

Rejection Region:

The test rejects H_0 when T is too small or too large, i.e.,

$$T \le a$$
 or $T \ge b$

where a and b are critical values satisfying:

$$P(T \le a \mid H_0) + P(T \ge b \mid H_0) = \alpha$$

Distribution of T under H_0 : Under H_0 ($\theta = \mu$):

- $\sum X_i \sim \text{Gamma}(n, \theta)$
- $\sum Y_j \sim \text{Gamma}(m, \theta)$

Thus,

$$T = \frac{\sum X_i}{\sum X_i + \sum Y_j} \sim \text{Beta}(n, m)$$

Alternatively, $\frac{T}{1-T} = \frac{\sum X_i/n}{\sum Y_j/m} \sim F_{2n,2m}$, which can be used to compute critical values.

$\mathbf{Q2}$

Problem 8.28, Casella and Berger (2nd Edition)

Let $f(x|\theta)$ be the logistic location probability density function:

$$f(x|\theta) = \frac{e^{(x-\theta)}}{(1+e^{(x-\theta)})^2}, \quad -\infty < x < \infty, \quad -\infty < \theta < \infty.$$

a)

Show that this family has an MLR.

Answer

To show the family has a monotone likelihood ratio (MLR) in x, consider the likelihood ratio for $\theta_2 > \theta_1$:

$$\frac{f(x|\theta_2)}{f(x|\theta_1)} = e^{\theta_1 - \theta_2} \left[\frac{1 + e^{x - \theta_1}}{1 + e^{x - \theta_2}} \right]^2.$$

Define $g(x) = \frac{1 + e^{x - \theta_1}}{1 + e^{x - \theta_2}}$. Its derivative is:

$$g'(x) = \frac{e^{x-\theta_1}(1+e^{x-\theta_2}) - e^{x-\theta_2}(1+e^{x-\theta_1})}{(1+e^{x-\theta_2})^2} = \frac{e^{x-\theta_1} - e^{x-\theta_2}}{(1+e^{x-\theta_2})^2} > 0,$$

where the inequality holds because $\theta_2 > \theta_1$. Thus, g(x) is strictly increasing in x, and so is the likelihood ratio.

Conclusion: The family $\{f(x|\theta)\}\$ has MLR in x.

b)

Based on one observation X, find the most powerful size α test of

$$H_0: \theta = 0$$
 versus $H_1: \theta = 1$.

For $\alpha = 0.2$, find the size of the Type II error.

Answer

By the Neyman-Pearson Lemma, the most powerful (MP) test rejects H_0 when:

$$\frac{f(x|1)}{f(x|0)} = e^{-1} \left(\frac{1 + e^x}{1 + e^{x-1}} \right)^2 > k.$$

Since the likelihood ratio is increasing in x (from part (a)), the MP test rejects if X > k', where k' is determined by the size α .

The CDF of the logistic distribution is:

$$F(x|\theta) = \frac{e^{x-\theta}}{1 + e^{x-\theta}}.$$

Under H_0 , the size condition is:

$$\mathbb{P}(X > k' \mid \theta = 0) = 1 - F(k'|0) = \frac{1}{1 + e^{k'}} = \alpha.$$

Solving for k':

$$k' = \log\left(\frac{1-\alpha}{\alpha}\right).$$

For $\alpha = 0.2$:

$$k' = \log(4) \approx 1.386.$$

Under H_1 :

$$\beta = \mathbb{P}(X \le k' \mid \theta = 1) = F(k'|1) = \frac{e^{k'-1}}{1 + e^{k'-1}} \approx \frac{e^{0.386}}{1 + e^{0.386}} \approx 0.595.$$

Conclusion: The MP level-0.2 test rejects when X > 1.386, with a Type II error rate of approximately 0.595.

c)

Show that the test in part b) is UMP size α for testing

$$H_0: \theta \leq 0$$
 versus $H_1: \theta > 0$.

What can be said about UMP tests in general for the logistic location family?

Answer

- 1. MLR Property: From part (a), the family has MLR in X.
- 2. Karlin-Rubin Theorem: Since the MP test for $\theta = 0$ vs $\theta = 1$ rejects for large X and does not depend on the specific $\theta_1 = 1$, it is uniformly most powerful (UMP) for $H_0: \theta \leq 0$ vs $H_1: \theta > 0$.

General Conclusion:

For the logistic location family, UMP tests exist for one-sided hypotheses and take the form "Reject H_0 if X > c."

Q3

Problem 8.29 a) - b), Casella and Berger (2nd Edition)

Let X be one observation from a Cauchy(θ) distribution.

The Cauchy(θ) density is given by:

$$f(x|\theta) = \frac{1}{\pi} \cdot \frac{1}{1 + (x - \theta)^2}, \quad x \in \mathbb{R}, -\infty < \theta < \infty.$$

a)

Show that this family does not have an MLR.

Hint:

Show that the Cauchy(θ) family $\{f(x|\theta): \theta \in \mathbb{R} = \Theta\}$, based on one observation X, does not have monotone likelihood ratio (MLR) in t(X) = X or t(X) = -X. That is, the ratio

$$\frac{f(x|\theta_2)}{f(x|\theta_1)}$$

might not be monotone (either increasing or decreasing) in x.

Answer

For $\theta_2 > \theta_1$, the likelihood ratio is:

$$\frac{f(x|\theta_2)}{f(x|\theta_1)} = \frac{1 + (x - \theta_1)^2}{1 + (x - \theta_2)^2}.$$

1. Limiting Behavior:

$$\lim_{x \to \pm \infty} \frac{f(x|\theta_2)}{f(x|\theta_1)} = 1.$$

- 2. Non-Monotonicity:
 - The ratio achieves a maximum at finite x. For example, let $\theta_1 = 0, \theta_2 = 1$:

$$\frac{f(x|1)}{f(x|0)} = \frac{1+x^2}{1+(x-1)^2}.$$

- At x = 0: Ratio = 1.

- At x = 1: Ratio = 2.

 $- As x \to \infty$: Ratio $\to 1$.

• The ratio increases from x = 0 to x = 1 and then decreases, proving non-monotonicity.

Conclusion: The Cauchy(θ) family lacks MLR in X or -X.

b)

Show that the test

$$\phi(x) = \begin{cases} 1 & \text{if } 1 < x < 3 \\ 0 & \text{otherwise} \end{cases}$$

is most powerful of its size for testing

$$H_0: \theta = 0$$
 versus $H_1: \theta = 1$.

Calculate the Type I and Type II error probabilities.

Hint:

Show that the test given is equivalent to rejecting H_0 if

$$f(x|\theta=1) > 2f(x|\theta=0)$$

and not rejecting otherwise. Conclude that this must be the most powerful (MP) test for its size. Justify why.

Answer

Consider the test:

$$\phi(x) = \begin{cases} 1 & \text{if } 1 < x < 3, \\ 0 & \text{otherwise.} \end{cases}$$

By the Neyman-Pearson Lemma, the MP test rejects H_0 when:

$$\frac{f(x|1)}{f(x|0)} = \frac{1+x^2}{1+(x-1)^2} > k.$$

- The ratio $\frac{f(x|1)}{f(x|0)}$ has critical points at $x = \frac{1 \pm \sqrt{5}}{2}$.
- At x = 1 and x = 3:

$$\frac{f(1|1)}{f(1|0)} = \frac{f(3|1)}{f(3|0)} = 2.$$

• The set $\{x: \frac{f(x|1)}{f(x|0)} > 2\} = (1,3)$ exactly matches $\phi(x)$.

Thus, $\phi(x)$ is the most powerful test for its size.

Under H_0 :

$$\alpha = \mathbb{P}(1 < X < 3 \mid \theta = 0) = \frac{1}{\pi} \left(\tan^{-1}(3) - \tan^{-1}(1) \right) \approx 0.1476.$$

Under H_1 :

$$\beta = 1 - \mathbb{P}(1 < X < 3 \mid \theta = 1) = 1 - \frac{1}{\pi} \left(\tan^{-1}(2) - \tan^{-1}(0) \right) \approx 0.6476.$$

Conclusion:

- $\phi(x)$ is MP with $\alpha \approx 0.1476$ and $\beta \approx 0.6476$.
- The rejection region (1,3) is unique for this α , as guaranteed by Neyman-Pearson.

$\mathbf{Q4}$

Consider one observation X from the probability density function

$$f(x \mid \theta) = 1 - \theta^2 \left(x - \frac{1}{2} \right), \quad 0 \le x \le 1, \quad 0 \le \theta \le 1.$$

We wish to test:

$$H_0: \theta = 0$$
 vs. $H_1: \theta > 0$

a)

Find the UMP test of size $\alpha = 0.05$ based on X. Carefully justify your answer.

Answer

1. Likelihood Ratio Analysis:

For $\theta_2 > \theta_1 \geq 0$, the likelihood ratio is:

$$\frac{f(x \mid \theta_2)}{f(x \mid \theta_1)} = \frac{1 - \theta_2^2(x - \frac{1}{2})}{1 - \theta_1^2(x - \frac{1}{2})}$$

2. Monotonicity Properties:

- When x > ½: the ratio is decreasing in x
 When x < ½: the ratio is increasing in x

So, the family does not have global monotone likelihood ratio in X, but the likelihood function tilts rightward under H_1 , which suggests rejecting for large X values is most powerful.

3. UMP Test Construction:

Under $H_0: \theta = 0, X \sim \text{Uniform}(0,1)$. So we define the test:

$$\phi(x) = \begin{cases} 1 & \text{if } x > c \\ 0 & \text{otherwise} \end{cases}$$

4. Critical Value Calculation:

$$P_{\theta=0}(X>c) = 1 - c = 0.05 \implies c = 0.95$$

Final UMP Test:

$$\phi(x) = \begin{cases} 1 & \text{if } x > 0.95 \\ 0 & \text{otherwise} \end{cases}$$

b)

Find the likelihood ratio test statistic $\lambda(X)$ based on X, expressed as a function of X.

Answer

The likelihood ratio test statistic is:

$$\lambda(X) = \frac{f(X \mid 0)}{\max_{\theta \in [0,1]} f(X \mid \theta)} = \frac{1}{\max_{\theta} \left[1 - \theta^2 (X - \frac{1}{2})\right]}$$

Case 1: $X \ge \frac{1}{2}$

Maximum occurs at $\theta = 0$:

$$\max_{\theta} f(X \mid \theta) = 1$$

Case 2: $X < \frac{1}{2}$

Maximum occurs at $\theta = 1$:

$$\max_{\theta} f(X \mid \theta) = 1 + \left(\frac{1}{2} - X\right) = 1.5 - X$$

Final LRT Statistic:

$$\lambda(X) = \begin{cases} \frac{1}{1.5 - X} & \text{if } X < \frac{1}{2} \\ 1 & \text{if } X \ge \frac{1}{2} \end{cases}$$

c)

Find the likelihood ratio test (LRT) of size $\alpha = 0.05$ for the above hypotheses.

Answer

1. Rejection Region:

From part (b), $\lambda(X) = 1$ for $X \ge \frac{1}{2}$, and is increasing for $X < \frac{1}{2}$. So to make the test most powerful while maintaining the correct size, we reject for large values of X.

2. Size Condition:

$$P_{\theta=0}(X > k) = 1 - k = 0.05 \implies k = 0.95$$

Final LRT:

Reject H_0 when X > 0.95

Note: The LRT coincides with the UMP test derived in part (a) because:

- 1. Although the MLR is not strictly monotone in X, the density under H_1 favors larger values of X
- 2. The test based on rejecting for large X values maximizes power subject to size, satisfying both Neyman-Pearson and UMP conditions