

University of Ruhuna
Bachelor of Science General Degree
(Level III) Semester I Examination - June/July 2015

Subject : Mathematics

Course unit: MAT313β /MMS3113 (Mathematical Statistics II)

Time :Two (02) Hours

Answer four (04) Questions only

1. Let $\hat{\theta}$ be a point estimator for parameter θ .

Explain what is meant by saying that, $\hat{\theta}$ is an unbiased estimator for θ .

a) Let Y_1, Y_2, \dots, Y_n be a random sample of size n from a population with mean μ and the variance σ^2 .

(i) Show that $\bar{Y} = \frac{\sum_{i=1}^n Y_i}{n}$ is an unbiased estimator for μ and $S^2 = \frac{\sum_{i=1}^n (Y_i - \bar{Y})^2}{(n-1)}$ is an unbiased estimator for σ^2 .

(ii) Let $T = \sum_{i=1}^n a_i Y_i$.

Show that T is an unbiased estimator of μ , if $\sum_{i=1}^n a_i = 1$; where a_1, a_2, \dots, a_n are constants.

b) Let X_1, X_2, \dots, X_n be a random sample of size n from a population with the mean μ and the variance σ_1^2 and Y_1, Y_2, \dots, Y_n be a random sample of size n from a population with the mean μ and the variance σ_2^2 .

If the two samples are independent,

(i) show that $\omega \bar{X} + (1 - \omega) \bar{Y}$ is an unbiased estimator of μ ; where $0 \leq \omega \leq 1$,

$$\bar{X} = \frac{\sum_{i=1}^n X_i}{n} \quad \text{and} \quad \bar{Y} = \frac{\sum_{i=1}^n Y_i}{n}.$$

(ii) show that the variance of this estimator is a minimum when $\omega = \frac{\sigma_2^2}{\sigma_1^2 + \sigma_2^2}$.
Find the value of the minimum variance.

2. a) Let Y_1, Y_2, \dots, Y_n be a random sample of size n from a population with the probability density function

$$f_Y(y, \theta) = \begin{cases} \frac{2}{\theta} - \frac{2y}{\theta^2} & , \text{ if } 0 < y < \theta, \\ 0 & , \text{ otherwise} \end{cases}$$

Find

- (i) $E(Y)$
(ii) the moment estimator of θ .

- b) Explain the method of maximum likelihood estimation.

- (i) Let Y_1, Y_2, \dots, Y_n be a random sample of size n from a population with probability density function

$$f_Y(y, \alpha) = \begin{cases} 2\alpha y e^{-\alpha y^2} & , \text{ if } y > 0, \alpha > 0, \\ 0 & , \text{ otherwise} \end{cases}$$

Find the maximum likelihood estimator of α .

- (ii) Let Y_1, Y_2, \dots, Y_n be a random sample of size n obtained from a Poisson distribution with probability mass function

$$Pr(Y = y) = \frac{e^{-\lambda} \lambda^y}{y!}, \quad y=0,1,2,\dots$$

Find the maximum likelihood estimator of λ .

3. a) Define the property consistent, of an estimator $\hat{\theta}_n$ of the parameter θ , used in parametric point estimation.

Let Y_1, Y_2, \dots, Y_n be a random sample of size n from a distribution with probability density function,

$$f_Y(y, \theta) = \begin{cases} \theta y^{\theta-1} & , \text{ if } 0 < y < 1, \quad \theta > 0, \\ 0 & , \text{ otherwise} \end{cases}$$

Show that $\bar{Y} = \frac{\sum_{i=1}^n Y_i}{n}$, is a consistent estimator of $\frac{\theta}{\theta+1}$.

- b) State Cramer-Rao inequality for the variance of an unbiased estimator of a function $\tau(\theta)$ of a parameter θ , in the usual notation.

When does the equality hold?

Let Y_1, Y_2, \dots, Y_n be a random sample of size n obtained from a normal distribution with known mean μ and unknown variance σ^2 .

$$\text{Let } T = \frac{\sum_{i=1}^n (Y_i - \mu)^2}{n}.$$

Show that T is a minimum variance unbiased estimator of σ^2 .
Obtain Cramer-Rao lower bound for the variance of T .

4. (a) State Neyman factorization theorem for a sufficient statistic.

Let Y_1, Y_2, \dots, Y_n be a random sample of size n from a population with probability density function $f_Y(y, \theta) = a(\theta)b(y)\exp\{c(\theta)d(y)\}$.

Show that $\sum_{i=1}^n d(Y_i)$ is a sufficient statistic for parameter θ .

- (b) Let Y_1, Y_2, \dots, Y_n be a random sample of size n obtained from a Poisson distribution with probability mass function

$Pr(Y = y) = \frac{e^{-\lambda}\lambda^y}{y!}$, $y=0,1,2,\dots$; where λ is an unknown parameter.

Show that

(i) $S = \sum_{i=1}^n Y_i$ is a sufficient statistic for λ ,

(ii) $\bar{Y} = \frac{\sum_{i=1}^n Y_i}{n}$ is an unbiased estimator for λ and

(iii) \bar{Y} is the minimum variance unbiased estimator for λ .

In the usual notation, state the theorem you used.

5. Let X_1, X_2, \dots, X_{n_1} be a random sample of size n_1 from a normal distribution with unknown mean μ_1 and the unknown variance σ^2 . Let Y_1, Y_2, \dots, Y_{n_2} be a random sample of size n_2 from a normal distribution with unknown mean μ_2 and the same unknown variance σ^2 . Assuming that the two samples are independent and the sample sizes are small, explain how you would construct $100(1 - \alpha)\%$ confidence interval for $(\mu_1 - \mu_2)$.

It is needed to compare two kind of fertilizers used in 'Rambutan' plantation. In the plantation some plants were treated with the fertilizer A and the rest were with the fertilizer B. From the plantation, a random sample of five plants treated with fertilizer A and a random sample of five plants treated with fertilizer B, were selected and the harvest was recorded as follows.

No. of fruits of plants treated with fertilizer A	850	833	848	796	803
No. of fruits of plants treated with fertilizer B	771	789	792	827	786

Construct 95% confidence interval for the difference between the mean number of fruits of plants treated with fertilizer A and fertilizer B.
State your assumptions clearly.

6. a) Explain the following terms used in Statistical Hypothesis testing.

- (i) Type I error
- (ii) Type II error
- (iii) Power function

b) Let Y be a random variable of a distribution whose probability density function is,

$$f_Y(y, \theta) = \begin{cases} \frac{1}{\theta} e^{-\frac{y}{\theta}} & , \text{ if } y > 0, \theta > 0, \\ 0 & , \text{ otherwise} \end{cases}$$

A single observation from this distribution is used to test the null hypothesis

$H_0 : \theta = 10$, against the alternative $H_1 : \theta \neq 10$.

If the null hypothesis is rejected when the observed value is less than 8 and greater than 12, find the probability of Type I error.

Also find the probability of Type II error if $\theta = 12$.

Obtain the power function of the test.
