

# MAASAI MARA UNIVERSITY REGULAR UNIVERSITY EXAMINATIONS 2023/2024 ACADEMIC YEAR

## FOURTH YEAR SECOND SEMESTER SCHOOL OF PURE, APPLIED AND HEALTH SCIENCES. DEGREE IN APPLIED STATISTICS WITH COMPUTING.

### COURSE CODE: STA 4244-1

### COURSE TITLE: SAMPLING THEORY AND METHODS II

DATE:

TIME:

**INSTRUCTIONS TO CANDIDATES** 

Answer Question ONE and any other TWO questions This paper consists of FOUR printed pages. Please turn over.

#### **QUESTION ONE (20 MARKS)**

Assuming that the random sample  $(x_i, y_i)$ , i = 1, 2, ..., n is drawn by SRSWOR and a known population mean  $\overline{X}$  of X and  $\overline{Y}$  of Y. Given also that

$$\varepsilon_0 = \frac{\bar{y} - \bar{Y}}{\bar{Y}} \quad and \ \varepsilon_1 = \frac{\bar{x} - \bar{X}}{\bar{X}}$$

a.

(i) Show that 
$$E(\varepsilon_0^2) = \frac{f}{n}C_Y^2$$

Where 
$$f = \frac{N-n}{N}$$
,  $S_Y^2 = \frac{1}{N-1} \sum_{i=1}^N (Y_i - \overline{Y})^2$  and  $C_Y = \frac{S_Y}{\overline{Y}}$  (4 mks)

- (ii) Show that  $E(\varepsilon_0 \varepsilon_1) = \frac{f}{n} \rho C_X C_Y$  where  $C_X = \frac{S_X}{\bar{X}}$  and  $\rho$  is the correlation coefficient between X and Y. (5 mks)
- b. Show that the ratio estimate  $\overline{\widetilde{Y_R}}$  is the best linear unbiased estimator of  $\overline{Y}$  when:
- (i) The relationship between  $y_i$  and  $x_i$  is linear passing through the origin, that is  $y_i = \beta x_i + e_i$  where  $e'_i s$  are independent with  $E(e_i | x_i) = 0$  and  $\beta$  is the slope parameter.

(ii) And the line is proportional to 
$$x_i$$
, that is,  
 $Var(y_i|x_i) = E(e_i^2) = Cx_i$  where C is constant. (8 mks)

c. Give the procedure of sampling by Lahiri's method. (3 mks)

#### **QUESTION TWO (15 MARKS)**

- (a) Give the advantages and disadvantages of Lahiri's method of sampling procedure.
- (4 mks)
- (b) Let
  - $Y_i$ : Value of study variable for the  $i^{th}$  unit of the population, i = 1, 2, ..., N.
  - $X_i$ : Known value of an auxiliary variable (size) for the  $i^{th}$  unit of the population.

 $P_i$ : Probability of selection of  $i^{th}$  unit in the population at any given draw and is proportional to size  $X_i$ .  $Z_i = \frac{Y_i}{NP_i}$ , i = -1, 2, ..., N. Consider the varying probability scheme and with replacement for a sample of size n. Letting  $y_r$  = the value of  $r^{th}$  observation in the sample and Pr =initial probability of selecting yr. Using  $z_r = \frac{y_r}{Np_r}$ , r = 1, 2, ..., n, then,

(i) 
$$\overline{Z} = \frac{1}{n} \sum_{i=1}^{n} z_i$$
 is an unbiased estimator of the population mean  $\overline{Y}$ . Prove.  
(4 mks)  
(ii) Show that  $Var(\overline{Z}) = \frac{\sigma_z^2}{n}$ 
(7 mks)

#### **QUESTION THREE (15 MARKS)**

(a) Given that  $\varepsilon_0 = \frac{\bar{y} - \bar{Y}}{\bar{Y}}$  and  $\varepsilon_1 = \frac{\bar{x} - \bar{X}}{\bar{X}}$ and that  $\widetilde{\overline{Y_R}} = \frac{\bar{y}}{\bar{x}}\bar{X}$ where  $\bar{X}$  is the population mean of X  $\bar{Y}$  is the population mean of Y  $\bar{x}$  is the sample mean of x  $\bar{y}$  is the sample mean of y

Writing  $\overline{\widetilde{Y_R}}$  in terms of  $\varepsilon's$ 

(i) Show that  $\overline{\widetilde{Y}_R} = (1 + \varepsilon_0)(1 + \varepsilon_1)^{-1}\overline{Y}$  (2 mks) (ii)  $\overline{Y}_R = \overline{Y}(1 + \varepsilon_0)(1 + \varepsilon_1)^{-1}\overline{Y}$  (2 mks)

(ii) Show that 
$$Y_R = Y(1 + \varepsilon_0 - \varepsilon_1 + \varepsilon_1^2 - \varepsilon_1\varepsilon_0 + \cdots)$$
 (4 mks)  
(iii) Show that if we assume that the higher powers  $\varepsilon_0$  and  $\varepsilon_1$  more than 2 are poslicily

- (iii) Show that if we assume that the higher powers  $\varepsilon_0$  and  $\varepsilon_1$  more than 2 are negligibly small, then the Bias of  $\overline{Y_R}$  is given by  $Bias\left(\overline{Y_R}\right) = E\left(\overline{Y_R} - \overline{Y}\right) = \frac{f}{n}\overline{Y}C_X(C_X - \rho C_Y)$  (4 mks)
- (b) Given the Mean Squared Error of  $\widehat{Y_R}$  is given by

$$MSE\left(\widehat{Y_R}\right) = \sum_{i=1}^{N} (Y_i - RX_i)^2$$

 $= \sum_{i=1}^{N} (Y_i - \overline{Y}) + R^2 \sum_{i=1}^{N} (X_i - \overline{X})^2 - 2R \sum_{i=1}^{N} (X_i - \overline{X}) (Y_i - \overline{Y}).$ 

Prove by assuming that  $\overline{Y} = R\overline{X}$  (5 mks)

### **QUESTION FOUR (15 MARKS)**

Defining the product estimator of the population mean  $\overline{Y}$  as

$$\left(\frac{\widetilde{Y}_p}{\widetilde{Y}_p}\right) = \frac{\overline{yx}}{\overline{x}}$$
 assuming that the population mean  $\overline{X}$  is known and letting  
 $\varepsilon_0 = \frac{\overline{y} - \overline{Y}}{\overline{Y}}$  and  $\varepsilon_1 = \frac{\overline{x} - \overline{X}}{\overline{X}}$   
a.

(i) Show that the 
$$Bias\left(\overline{Y_p}\right)$$
 is given by  
 $Bias\left(\overline{\widetilde{Y_p}}\right) = \frac{f}{n\overline{X}}S_{XY}$  (6 mks)  
(ii) Writing  $\overline{\widetilde{Y_p}}$  in terms of  $\varepsilon_0$  and  $\varepsilon_1$ ,  
Show that  
 $MSE\left(\overline{\widetilde{Y_p}}\right) = \overline{Y}^2 E(\varepsilon_1^2 + \varepsilon_0^2 + 2\varepsilon_1\varepsilon_2$  (4 mks)

b. Compare the variances of sample mean under SRSWOR with that of the product estimator. (5 mks)