

CHUKA



UNIVERSITY

## UNIVERSITY EXAMINATIONS

**FIRST YEAR EXAMINATION FOR THE DEGREE OF DOCTORATE OF  
PHILOSOPHY IN APPLIED STATISTICS**

**MATH 943: GENERALIZED LINEAR MODELS AND APPLICATIONS****STREAMS: PhD****TIME: 3 HOURS****DAY/DATE: TUESDAY 13/8/2019****2.30 P.M. – 5.30 P.M****INSTRUCTIONS**

- Answer any **THREE** questions.
- Do not write anything on the question paper.

**QUESTION ONE (20 Marks)**

Suppose  $X$  is a binary random variable that takes value 0 with probability  $p$  and value 1 with probability  $1-p$ . let  $X_1, \dots, X_n$  be iid samples of  $X$ .

- Compute a maximum likelihood estimation (MLE) estimate of  $p$ . (5 marks)
- Is  $\hat{p}$  an unbiased estimate of  $p$ ? Prove the answer. (5 marks)
- Compute the expected square error of  $\hat{p}$  in terms of  $p$ . (5 marks)
- Prove that if you know that  $p$  lies in the interval  $\left[\frac{1}{4}; \frac{3}{4}\right]$  and you are given only  $n = 3$  samples of  $X$ , then  $\hat{p}$  is an inadmissible estimator of  $p$  when minimising the expected square error of estimation. (5 marks)

**QUESTION TWO (20 Marks)**

(a) Suppose that the p.d.f of a random variable  $X$  has a 2-component mixture form:

$$p_\alpha(x) = \alpha * p_1(x) + (1 - \alpha) * p_2(x)$$

One component is the density model  $p_1(x)$  and the other component is the density model  $p_2(x)$ . We know both  $p_1(x)$  and  $p_2(x)$ . We do not know  $\alpha$ . Given that  $\{x_1, x_2, \dots, x_n\}$  are iid samples from the distribution of  $X$ , give EM algorithm for estimating  $\alpha$ . Describe the E-Step and M-step clearly in your answer (give clear step by step derivation).

(12 Marks)

**MATH 943**

(b) Consider independent binary features

$$\mathbf{X} = (x_1, \dots, x_d)^t \quad p_i = Pr[x_i = 1 \mid \omega_1] \quad q_i = Pr[x_i = 1 \mid \omega_2]$$

and assuming conditional independence

$$P(\mathbf{X} \mid \omega_1) = \prod_{i=1}^d p_i^{x_i} (1 - p_i)^{1-x_i}$$

$$P(\mathbf{X} \mid \omega_2) = \prod_{i=1}^d q_i^{x_i} (1 - q_i)^{1-x_i}$$

- (i) Derive the likelihood ratio. (4 marks)
- (ii) Derive the discriminant function. (4 marks)

**QUESTION THREE (20 Marks)**

The data in Table 1 relate to grain yield (Y), plant height (X<sub>1</sub>), and tiller number (X<sub>2</sub>) of sorghum.

Table 1: Performance of Sorghum with respect to grain yield (Kg/ha), plant height (cm) and tiller numbers/hill

Grain yield Kg/ha (Y)	Plant height cm (X <sub>1</sub> )	Tiller No./hill (X <sub>2</sub> )
5755	110.5	14.5
5939	105.4	16.0
6010	118.1	14.6
6545	104.5	18.2
6730	93.6	15.4
6750	84.1	17.6
6899	77.8	17.9
7862	75.6	19.4

- (i) Fit a multiple linear regression model of Y on X<sub>1</sub> and X<sub>2</sub>. (8 Marks)
- (ii) Determine variance of β<sub>0</sub>, β<sub>1</sub> and β<sub>2</sub>. (6 Marks)
- (iii) Test if the fitted model is adequate. Take α = 0.05. (6 Marks)

**QUESTION FOUR (20 Marks)**

Linear regression models a real-valued output Y given an input vector X as

$$Y \mid X \sim N(\mu(X), \sigma^2)$$

where the mean is a linear function of the input:  $\mu(X) = \beta^T X = \beta_0 + \beta_1 X_1 + \dots + \beta_p X_p$

Logistic regression models a binary output Y by

$$Y \mid X \sim \text{Bernoulli}(\theta(X))$$

## MATH 943

where the Bernoulli parameter is related to  $\beta^T X$  by the logit transformation

$$\text{logit}(\theta(X)) \equiv \log\left(\frac{\theta(X)}{1 - \theta(x)}\right) = \beta^T X$$

Given data  $\{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\}$ , for each of the two regression models above, show that at the MLE  $\hat{\beta}$

$$\sum_{i=1}^n x_i * y_i = \sum_{i=1}^n x_i * E[Y | X = x_i, \beta = \hat{\beta}]$$

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