

**Instructions:**

1. You have exactly four hours to answer questions in this examination.
2. There are 8 problems of which you must answer 6.
3. Only your first 6 problems will be graded.
4. Write only on one side of the paper, and start each question on a new page.
5. Write your **number** on every page.
6. Do not write your name anywhere on your exam.
7. You must show your work to receive credit.
8. While the eight questions are equally weighted, within a given question, the parts may have different weights.

1. Suppose that  $X$  has a binomial distribution with  $m$  trials and probability  $\mu$ .

- (a) Express the binomial mass function in exponential form in terms of the canonical parameter  $\theta = \text{logit}(\mu)$ .
- (b) Derive the deviance measure of fit  $D(y, \mu)$  for the binomial model, where  $Y = X/m$ .
- (c) Show that the deviance can be approximated by the Pearson chisquared statistic,  $\chi^2$ , if  $m$  is large, where

$$\chi^2 = \frac{m(Y - \mu)^2}{\mu(1 - \mu)}.$$

- (d) Suppose that  $X_i \sim B(m_i, \mu_i)$ ,  $i = 1, \dots, n$ , independently, and that the success probabilities satisfy a linear logistic model. Let  $H_0 \subset H_1$  be nested hypotheses. Show that the corresponding model deviances satisfy the Pythagorean relationship

$$D(\mathbf{y}, \hat{\boldsymbol{\mu}}_0) = D(\mathbf{y}, \hat{\boldsymbol{\mu}}_1) + D(\hat{\boldsymbol{\mu}}_1, \hat{\boldsymbol{\mu}}_0),$$

where  $\mathbf{y} = (y_1, \dots, y_n)$  is the vector of observed proportions, and  $\hat{\boldsymbol{\mu}}_0$  and  $\hat{\boldsymbol{\mu}}_1$  are the vectors of fitted proportions under  $H_0$  and  $H_1$  respectively.

2. Suppose that  $X \sim B(m, \mu)$  where  $\mu = e^\theta / (1 + e^\theta)$ , and let  $Z$  be defined by the equation  $X = m\mu + \sqrt{m\mu(1 - \mu)}Z$ .

- (a) Show that for  $c > 0$

$$E \left\{ \log \frac{X + c}{m\mu} \right\} = \frac{c}{m\mu} - \frac{1 - \mu}{2m\mu} + O(m^{-3/2}).$$

- (b) Find the corresponding expansion for

$$E \left\{ \log \frac{m - X + c}{m(1 - \mu)} \right\}.$$

- (c) Consider the estimate

$$\tilde{\theta} = \log \frac{X + c}{m - X + c}.$$

Show that

$$E(\tilde{\theta}) = \theta + \frac{(1 - 2\mu)(c - \frac{1}{2})}{m\mu(1 - \mu)} + O(m^{-3/2}).$$

- (d) Show that the expected value of the corresponding ML estimate,  $\hat{\theta}$ , does not converge to  $\theta$  as  $m$  increases.

3. Consider the unbalanced random one-way model

$$y_{ij} = \mu + \alpha_i + \epsilon_{ij},$$

where  $\alpha_i$  and  $\epsilon_{ij}$  are independently distributed such that  $\alpha_i \sim N(0, \sigma_\alpha^2)$ , and  $\epsilon_{ij} \sim N(0, \sigma_\epsilon^2)$ ,  $i = 1, 2, \dots, a$ ;  $j = 1, 2, \dots, n_i$ . Let  $SS_\alpha$  be the sum of squares,

$$SS_\alpha = \sum_{i=1}^a \frac{y_{i.}^2}{n_i} - \frac{y_{..}^2}{n},$$

where  $n_{.} = \sum_{i=1}^a n_i$ .

- (a) Write  $SS_\alpha$  as  $\mathbf{y}'\mathbf{A}\mathbf{y}$ , indicating what the matrix  $\mathbf{A}$  is, where  $\mathbf{y}$  is the vector of all observations.  
 (b) Make use of part (a) to show that

$$E(SS_\alpha) = (n_{.} - \frac{1}{n_{.}} \sum_{i=1}^a n_i^2) \sigma_\alpha^2 + (a - 1) \sigma_\epsilon^2$$

- (c) What distribution does  $SS_\alpha$  have if  
 (i)  $\sigma_\alpha^2 = 0$   
 (ii)  $\sigma_\alpha^2 \neq 0$   
 (d) Show that  $SS_\alpha$  and  $SS_E$ , the residual sum of squares, are independent.  
 (e) Show how to obtain the expected value of  $\frac{SS_\alpha}{SS_E}$ .

4. Consider the balanced mixed model

$$y_{ijk} = \mu + \alpha_{(i)} + \beta_{(j)} + \gamma_{i(k)} + (\alpha\beta)_{(ij)} + \epsilon_{i(jk)},$$

$i = 1, 2, \dots, a$ ;  $j = 1, 2, \dots, b$ ;  $k = 1, 2, \dots, c$ , where  $\alpha_{(i)}$  and  $\beta_{(j)}$  are fixed unknown parameters representing the  $i$ th level and  $j$ th level of factors A and B, respectively, and  $\gamma_{i(k)}$  and  $\epsilon_{i(jk)}$  are distributed independently as  $N(0, \sigma_{\gamma(\alpha)}^2)$  and  $N(0, \sigma_\epsilon^2)$ , respectively.

(a) Derive a test statistic for testing each of the following hypotheses:

- (i)  $H_0 : \alpha_{(i)} = 0$  for all  $i$ .  
 (ii)  $H_0 : \beta_{(j)} = 0$  for all  $j$ .

Please specify the distribution and degrees of freedom of the test statistic under  $H_0$  in each case.

- (b) Show how to obtain an exact confidence interval on  $\theta = \sigma_{\gamma(\alpha)}^2 + \sigma_\epsilon^2$ , with a confidence coefficient greater than or equal to 0.95.  
 (c) Suppose that the interaction A\*B is significant and therefore it is necessary to test each factor at fixed levels of the other factor.  
 (i) Give a test statistic that can be used to compare the means of two levels of A (say  $i$  and  $i'$ ) at the fixed level  $j$  of B.  
 (ii) Give a test statistic that can be used to compare the means of two levels of B (say  $j$  and  $j'$ ) at the fixed level  $i$  of A.

In each case, please describe the null distribution of the test statistic

- (d) What is the *best linear unbiased estimator* of the least-squares mean for level  $i$  of A?

5. Let  $X_i$  ( $i = 1, \dots, m$ ) and  $Y_j$  ( $j = 1, \dots, n$ ) be mutually independent with the  $X_i$  iid  $N(\theta_1, \sigma_1^2)$ , and the  $Y_j$  iid  $N(\theta_2, \sigma_2^2)$ , where  $\theta_k \in (-\infty, \infty)$  ( $k = 1, 2$ ) are unknown. Consider estimation of  $\Delta = \theta_2 - \theta_1$  under squared error loss. Define  $\bar{X} = m^{-1} \sum_{i=1}^m X_i$  and  $\bar{Y} = n^{-1} \sum_{j=1}^n Y_j$ .

- (a) Assume  $\sigma_k^2 (> 0)$  ( $k = 1, 2$ ) to be known. Consider independent  $N(0, \tau_1^2)$  and  $N(0, \tau_2^2)$  priors for  $\theta_1$  and  $\theta_2$  respectively. Show that the Bayes estimator of  $\Delta$  is given by  $(1 - B_2)\bar{Y} - (1 - B_1)\bar{X}$ , where  $B_1 = \sigma_1^2 / (\sigma_1^2 + m\tau_1^2)$  and  $B_2 = \sigma_2^2 / (\sigma_2^2 + n\tau_2^2)$ .
- (b) If  $\sigma_k^2$  ( $k = 1, 2$ ) are known, show that  $\bar{Y} - \bar{X}$  is a minimax estimator of  $\Delta$ .
- (c) Suppose now  $\sigma_k^2$  ( $k = 1, 2$ ) are unknown, but  $\sigma_k^2 \leq A_k$  ( $k = 1, 2$ ), where  $A_k (> 0)$  ( $k = 1, 2$ ) are known. Show that  $\bar{Y} - \bar{X}$  continues to be a minimax estimator of  $\Delta$ .

6. (a) Let  $X_1, \dots, X_n$  be iid from the distribution with pdf given by  $f_\mu(x) = (2\pi x^3)^{-1/2} \exp[-(x - \mu)^2 / (2\mu^2 x)]$ , where  $x > 0$  and  $\mu > 0$ . Show that the family of pdf's has monotone likelihood ratio in  $\sum_{i=1}^n X_i$ .

(b) Consider a discrete random variable  $X$  assuming values 1, 2 and 3 with pf's  $p_0$  and  $p_1$  under  $H_0$  and  $H_1$  given by

$x$	1	2	3
$p_0(x)$	.0001	.0500	.9499
$p_1(x)$	.0100	.1000	.8900

- (i) Find the MP size .05 test for testing  $H_0$  against  $H_1$ .
- (ii) The three possible nonrandomized tests at level (NOT size) .05 are given by (1) never reject, (2) reject iff  $x = 1$  and (3) reject iff  $x = 2$ . Find the optimal among these three tests.
- (iii) Comment on the result found in (ii).

7. Suppose that  $\{X_n, n \geq 1\}$  is a sequence of identically distributed random variables with finite mean.

(a) Prove that

$$\frac{1}{n} \max_{1 \leq j \leq n} |X_j| \rightarrow 0 \text{ a.s.}$$

(b) Prove that

$$\lim_{n \rightarrow \infty} E \left[ \frac{1}{n} \max_{1 \leq j \leq n} |X_j| \right] = 0.$$

8. Let  $X_n, n \geq 1$ , be random variables with respective characteristic functions  $\phi_n, n \geq 1$ . Suppose that  $\sup_{n \geq 1} E[g(X_n)] < \infty$  for some nonnegative function  $g$  satisfying  $g(x) \rightarrow \infty$  as  $x \rightarrow \infty$  and  $g(x) \rightarrow \infty$  as  $x \rightarrow -\infty$ . Suppose further that  $\phi_n(t) \rightarrow h(t)$  for some complex-valued function  $h$ . Prove that the sequence of random variables  $\{X_n, n \geq 1\}$  converges in distribution.