## Ph.D. Qualifying Exam of Statistical Methods, February 2008

1. A chemical engineer is investigating the yield of a process. There are k process variables,  $x_1, \ldots, x_k$ , and each variable can be run at a low and a high level. Thus this engineer runs a  $2^k$  full factorial design with n center points, and  $x_i, i = 1, \ldots, k$  are defined on a coded scale from -1 to 1. To represent the results, this engineer decides to fit a main effects only model, say

$$y = \beta_0 + \beta_1 x_1 + \ldots + \beta_k x_k + \varepsilon_1$$

where  $\varepsilon$  is the random error. Find the relationship between the least-squares estimates of  $\beta_i, i = 1, ..., k$  and main effect estimates. (10 points)

- 2. Find the least-squares estimator of  $\beta$  in the model  $\mathbf{y} = \mathbf{X}\beta + \varepsilon$  subject to a set of equality constraints on  $\beta$ , say  $T\beta = c$ . (10 points)
- 3. Consider the single-factor fixed-effects analysis of variance with 3 treatments:

$$y_{ij} = \mu + \tau_i + \varepsilon_{ij}, i = 1, 2, 3, j = 1, \dots, n,$$

where  $\varepsilon_{ij}$  is an normal error with zero mean and variance  $\sigma^2$ . This model can be expressed in terms of the regression model with two indicate variables.

- (1) Find the least-squares estimates of these regression coefficients and show the relationship with estimates of  $\mu$  and  $\tau_i$ . (5 points)
- (2) Show how this regression model could be used to test the hypothesis, H<sub>0</sub>: τ<sub>1</sub> = τ<sub>2</sub> = τ<sub>3</sub> = 0, and write down the corresponding test statistic and its distribution under H<sub>0</sub>. (10 points)
- 4. Consider a *p*-dimensional random vector, **U**, with zero mean vector and covariance matrix,  $\Sigma$ . Let  $\lambda_1 \geq \lambda_2 \geq \ldots \geq \lambda_p$  be the eigenvalues of  $\Sigma$ , and  $P_1, \ldots, P_p$ be the corresponding eigenvectors.
  - (1) Find a  $p \times p$  matrix W such that the components of  $\mathbf{Z} = W\mathbf{U}$  are uncorrelated and their variances are all equal to 1. (5 points)
  - (2) Let B be any  $p \times 1$  vector such that ||B|| = 1. Show that

$$\max_{B \perp P_1, \dots, P_{i-1}} Var(B'\mathbf{U}) = \lambda_i,$$

and "max" is attained when  $B = P_i$ . (5 points)

(3) Let  $E_p$  denote the *p*-dimensional Euclidean space. Show

$$\min_{S_{i-1}} \max_{\|B\|=1, B \perp S_{i-1}} Var(B'\mathbf{U}) = \lambda_i,$$

where  $S_{i-1}$  is a space of (i-1) dimensions in  $E_p$ . (10 points)

- 5. Let  $X = (X_1, \ldots, X_p)'$  come from a *p*-dimensional multivariate normal distribution with mean vector  $\mu$  and covariance matrix  $\Sigma$ , and A be a  $p \times p$  symmetric matrix.
  - (1) Show that the covariance of X with X'AX is

$$Cov(X, X'AX) = 2\Sigma A\mu.$$
 (5 points)

- (2) Prove that X'AX and BX are distributed independently if and only if  $B\Sigma A = 0$ . (10 points)
- 6. Consider the simple regression model is

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots \beta_k x_k + \varepsilon,$$

where  $E(\varepsilon) = 0$  and  $Var(\varepsilon) = \sigma^2$ . Suppose that n > k uncorrelated observations are available, and let  $y_i$  denote the *i*th observed response and  $x_{ij}$  denote the *i*th level of regressor  $x_j$ . Here inverse matrix of  $(\mathbf{X}'_n \mathbf{X}_n)$  is assumed to exist, where  $\mathbf{X}_n$  is the model matrix of these *n* observations. Let  $\beta = (\beta_0, \beta_1, \dots, \beta_k)'$ be the coefficient vector. Suppose the (n + 1)th observation is obtained. Compute the covariance matrix of current least-squares estimator,  $\hat{\beta}_{n+1}$ , based on  $(\mathbf{X}'_n \mathbf{X}_n)^{-1}$ . (15 points)

- 7. Consider the  $2_{IV}^{6-2}$  design with the generators I = ABCE and I = BCDF.
  - (1) Construct this design. (5 points)
  - (2) Show the complete alias structure of this design. (10 points)