

UNIVERSITY OF MASSACHUSETTS  
Department of Mathematics and Statistics  
ADVANCED EXAM - LINEAR MODELS  
August 26, 2013 (3 hours)

- Do all problems.
- Start each problem on a new page.
- A total of 70 points is needed to pass.

1. (15 points) Consider a general linear model

$$\underline{Y} = X\underline{\beta} + \underline{\varepsilon},$$

where  $\underline{Y}$  is  $n \times 1$ ;  $X$  is  $n \times p$  with rank  $r(\leq p)$ ;  $\underline{\beta}$  is  $p \times 1$ , and  $\underline{\varepsilon}$  is an  $n$ -dimensional random vector with  $E(\underline{\varepsilon}) = \underline{0}$ .

- (a) Define what it means for  $\underline{c}'\underline{\beta}$  to be estimable and show that  $\underline{c}'\underline{\beta}$  is estimable if and only if  $\underline{c}'$  is a vector in the linear space generated by the row vectors of  $X$ .
  - (b) State and PROOF the Gauss-Markov theorem which relates an estimable linear function of  $\underline{\beta}$  with its BLUE, for  $\text{Cov}(\underline{\varepsilon}) = \sigma^2 I$ .
2. (20 points) Consider the linear model  $Y_i = \beta X_i + \varepsilon_i$  where  $E(\varepsilon_i) = 0$  and  $\text{Var}(\varepsilon_i) = \sigma^2$  with  $\beta$  and  $X_i$  scalars. The  $X_i$  are not random.

- (a) Write out explicitly in terms of the  $Y$ 's and  $X$ 's; what the least squares estimator of  $\beta$  is (call it  $\hat{\beta}$ ) and give an explicit expression for  $\text{Var}(\hat{\beta})$ . Also give an expression for  $\hat{\sigma}^2$ , the optimal unbiased estimate.
  - (b) An alternate estimator arises by considering  $Z_i = Y_i/X_i$ .
    - i. Show that  $\bar{Z} = \sum_{i=1}^n Z_i/n$  is an unbiased estimator for  $\beta$ .
    - ii. Find the variance of  $\bar{Z}$  and show directly that this variance is greater than or equal to the variance of  $\hat{\beta}$ .
    - iii. Why else (other than your result in (ii) above) do you know that the variance of  $\bar{Z}$  is greater than or equal to the variance of  $\hat{\beta}$ ?
3. (10 points) Let  $y \sim N(\mu, \Sigma)$ ,  $r = y' Ay$ , and  $s = B y$ , where  $A$  is a symmetric matrix.
- (a) Prove that if  $B \Sigma A = 0$  then  $r$  and  $s$  are independent.
  - (b) Use the result from part a to show that the the sample mean and sample variance of  $y$  are independent.

4. (10 points) An experiment was run to compare three different primitive altimeters (an altimeter is a device which measures altitude). Each of three pilots used each of three altimeters and the response is the error in reading.

	Altimeter		
	1	2	3
Pilot 1	3	4	7
Pilot 2	6	5	8
Pilot 3	3	4	7

We assume first that the three pilots are FIXED factor levels, and consider it as a two-way fixed effects model with interactions.

- (a) Plot these 9 observations to see any indication of Pilot-Altimeter interactions (no computation)? What should the graph look like if there is no interaction?
- (b) If the design is considered as a two fixed effects design. How will you interpret the the main effects of the altimeters if they are used by the three pilots?
5. (45 points) Let  $Y_{ij}$ ,  $j = 1, 2, \dots, n_i$ ,  $i = 1, 2, \dots, m$  be observable r.v.'s such that

$$Y_{ij} = \alpha_0 + \alpha_i + \varepsilon_{ij},$$

where the  $\alpha$ 's are unknown parameters, the  $\varepsilon_{ij}$  are i.i.d.  $N(0, \sigma^2)$  unobservable r.v.s, and  $\sigma^2 > 0$  is unknown.

- (a) Let  $\psi = \sum_{i=0}^m \ell_i \alpha_i$  with  $\ell_i$  being specified constants. Show that  $\psi$  is estimable if and only if  $\ell_0 = \sum_{i=1}^m \ell_i$ .
- (b) Define  $\mu_i = \alpha_0 + \alpha_i$ ,  $i = 1, 2, \dots, m$ , which will be used for all the following questions as well. Write down the least squares estimates for the  $\mu$ 's. Are they UMVUE's? Why? (Explain briefly). Write down also the MLE,  $\tilde{\sigma}^2$  of  $\sigma^2$  (no need to prove the results), and the unbiased estimator  $\hat{\sigma}^2$  obtained from the MLE.
- (c) Use the full-reduced model approach to construct the F test for testing  $H_0$ : all  $\mu_i$  are equal versus  $H_1$ : at least two  $\mu_i$  are different, using a test of size  $\alpha$ . Write out the F-statistic explicitly and state the distribution of the test statistic, under both null and alternative hypotheses.
- (d) Here we are interested in simultaneous confidence intervals, with confidence  $1 - \alpha$ , for all pairwise differences in the means; e.g., differences  $\mu_j - \mu_{j'}$ ,  $j \neq j'$ .
- i. Give Bonferroni's inequality and explain how it is used to construct simultaneous confidence intervals. (No need to prove anything you can just write down the result.) Be complete.

- ii. For  $n_i = n$  ( $i = 1, \dots, m$ ), define the studentized range distribution and use it to derive Tukey's method for the problem at hand.
  - iii. How would you decide which of the two methods is better here? Why is the Bonferroni method so popular in simultaneous confidence intervals?
- (e) For this question, suppose there is a variable associated with each of the groups (e.g., a dose of some sort) with  $x_i$  denoting the value for group  $i$ , each  $x_i > 0$  and all the  $x_i$  being different.
- i. If it is known that  $V(\epsilon_{ij}) = \sigma^2 x_i$ , explain how you could transform the model to make optimal inferences on the  $\mu$ 's. Could you use a standard one-way analysis of variance routine found in the statistical packages or would you need a general regression package? Explain.
  - ii. Now suppose the  $x_i$  values are used to model the mean with the assumption that  $\mu_i = \beta_0 + \beta_1 x_i$ , but return to assuming that  $V(\epsilon_{ij}) = \sigma^2$ .
    - A. Write the model as  $\underline{Y} = X\underline{\beta} + \underline{\epsilon}$ . Write down the least squares estimate of  $\underline{\beta}$  in terms of  $\underline{Y}$  and  $X$  and the UMVUE for  $\sigma^2$ . No need to simplify these.
    - B. Use the full-reduced model approach to develop the F-test for the null hypothesis of  $H_0 : \mu_i = \beta_0 + \beta_1 x_i$  versus  $H_A : \text{the } \mu_i \text{ are not a linear function of the } x_i, \text{ but change over } i \text{ in some unspecified way}$  - (This is just the model in part (b)). This is a test for "lack of fit". Lay out the general calculation and the degrees of freedom associated with the test, but there is no need to simply the expression.

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