

**STAT 2131:**  
**Applied Statistical Methods I**  
**HW #5**  
**Due Thursday, November 11th**

1. Refer to the data set CH09PR10.txt (from the KNNL book). A personnel officer in a governmental agency administered four newly developed aptitude tests to each of 25 applicants for entry-level clerical positions in the agency. For purpose of the study, all 25 applicants were accepted for positions irrespective of their test scores. After a probationary period, each applicant was rated for proficiency on the job. The scores on the four tests ( $X_1, X_2, x_3, X_4$ ) and the job proficiency score ( $Y$ ) for the 25 employees were recorded.
  - (a) Obtain the scatter plot matrix. Also obtain the correlation matrix of the  $X$  variables. What do the scatter plots suggest about the nature of the functional relationship between the response variable  $Y$  and each of the predictor variables? Are any serious multicollinearity problems evident? Explain.
  - (b) Fit the multiple regression function containing all four predictor variables as first-order terms. Does it appear that all predictor variables should be retained?
  - (c) Consider only the four first order terms of  $X_1, X_2, X_3, X_4$ , find the best subset regression models according to the adjusted  $R^2$  criterion and the AIC criterion.
  - (d) Using forward stepwise regression, find the best subset of predictor variables to predict job proficiency. Use  $\alpha$  limits of 0.05 and 0.1 for adding or deleting a variable.
  - (e) How does the best subset model in part (d) compare to that in part (c)?
2. We have looked at using ordinary least squares/maximum likelihood estimation for the simple linear regression model. This problem considers an alternative estimation procedure. For simplicity, we will assume that the variance  $\sigma^2$  is known.

You observe outcomes  $y_i$  from  $i = 1, \dots, n$  subjects and assume that the data follow the linear model

$$y_i = \beta_0 + x_i\beta_1 + \epsilon_i$$

where  $x_i$  are known deterministic covariates,  $\beta_0$  and  $\beta_1$  are unknown deterministic parameters, and  $\epsilon_i$  are independent and identically distributed mean zero Gaussian random variables with *known* variance 1. Given some  $\lambda > 0$ , you decide to estimate  $\beta_1$  with  $\tilde{\beta}_1$  that minimizes the penalized sum-of-squares

$$\text{PSS}_\lambda(\beta_0, \beta_1) = \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2 + \lambda \beta_1^2$$

so that

$$(\tilde{\beta}_0, \tilde{\beta}_1) = \operatorname{argmin}_{\beta_0, \beta_1 \in \mathbb{R}} \text{PSS}_\lambda(\beta_0, \beta_1).$$

- (a) What is  $\tilde{\beta}_1$  for a given  $\lambda > 0$ ?
- (b) Show that  $\tilde{\beta}_1$  is biased for a given  $\lambda > 0$ .
- (c) Compare the variance of  $\tilde{\beta}_1$  with the variance of the OLS  $\hat{\beta}_1$ . Is one variance always smaller than the other? If so, prove it. If not, under what conditions is  $var(\tilde{\beta}_1) < var(\hat{\beta}_1)$ ?