STAT 714 hw 5

Gauss-Markov model, estimation, projections, miscellaneous results

Do problem 4.26 from Monahan. In addition:

- 1. For some matrices **X**, **A**, and **B**:
 - (a) Show that $AX^TX = BX^TX$ if and only if $AX^T = BX^T$.
 - (b) Let $(\mathbf{X}^T\mathbf{X})^-$ be a generalized inverse of $\mathbf{X}^T\mathbf{X}$. Show that $\mathbf{X}(\mathbf{X}^T\mathbf{X})^-$ is a gen. inverse of \mathbf{X}^T .
 - (c) Give a matrix **P** which is not the identity matrix such that $\mathbf{Pc} = \mathbf{c}$ for any $\mathbf{c} \in \operatorname{Col} \mathbf{X}^T$.
- 2. Let $Y_i = \mu + \sum_{j=1}^d \xi_{ij} \beta_j + \varepsilon_i$ for i = 1, ..., n and assume the matrix $(\xi_{ij})_{1 \le i \le n, 1 \le j \le d}$ has full-column rank. Moreover, suppose $\sum_{j=1}^d \xi_{ij} = 1$ for each i = 1, ..., n.
 - (a) Give a representation of the model as y = Xb + e.
 - (b) Check whether μ is estimable.
 - (c) Give the matrix **C** such that $\mathbf{Cb} = \mathbf{0}$ imposes the constraint $\sum_{j=1}^{d} \sum_{i=1}^{n} \xi_{ij} \beta_j = 0$.
 - (d) Give the least-squares estimator $\hat{\mu}$ under the constraint in (c) in terms of Y_1, \ldots, Y_n .
- 3. Let **X** be an $n \times p$ design matrix with the first column a column of ones and **y** be an $n \times 1$ vector of responses (which do not all take the same value). Let $\mathbf{P}_{\mathbf{X}}$ be the orthogonal projection onto Col **X** and $\mathbf{P}_{\mathbf{1}}$ be the orthogonal projection onto Span $\{\mathbf{1}_n\}$.
 - (a) Give an interpretation of the quantity $\frac{\|(\mathbf{P_X} \mathbf{P_1})\mathbf{y}\|^2}{\|(\mathbf{I} \mathbf{P_1})\mathbf{y}\|^2}$.
 - (b) Give the range of values which the quantity in part (a) can take.
 - (c) Write down the quantity in part (a) in terms of Y_i , \hat{Y}_i , and \bar{Y}_n , where the Y_i are the entries of \mathbf{y} , \hat{Y}_i are the entries of $\mathbf{P}_{\mathbf{X}}\mathbf{y}$, and $\bar{Y}_n = n^{-1}\sum_{i=1}^n Y_i$.
- 4. Let $\mathbf{y} = \mathbf{X}\mathbf{b} + \mathbf{e}$, where $\mathbb{E}\mathbf{e} = \mathbf{0}$ and $\operatorname{Cov}\mathbf{e} = \sigma^2\mathbf{I}_n$, and partition \mathbf{X} as $\mathbf{X} = [\mathbf{x}_1 \ \mathbf{X}_2]$, where \mathbf{x}_1 is a vector and \mathbf{X}_2 is a matrix with one column a column of ones. Suppose \mathbf{X} has full column rank and let $\hat{\mathbf{b}} = (\mathbf{X}^T\mathbf{X})^{-1}\mathbf{X}^T\mathbf{y}$. Partition $\hat{\mathbf{b}}$ as $\hat{\mathbf{b}} = [\hat{b}_1 \ \hat{\mathbf{b}}_2^T]^T$.
 - (a) Show that

$$\operatorname{Var} \hat{b}_1 = \frac{\sigma^2}{1 - R_1^2} \frac{1}{\|(\mathbf{I} - \mathbf{P_1})\mathbf{x}_1\|^2},$$

where $R_1^2 = \|(\mathbf{P}_{\mathbf{X}_2} - \mathbf{P}_1)\mathbf{x}_1\|^2 / \|(\mathbf{I} - \mathbf{P}_1)\mathbf{x}_1\|^2$.

- (b) Give an interpretation of the result in part (a).
- 5. Let $\mathbf{y} = \mathbf{X}\mathbf{b} + \mathbf{e}$, where $\mathbb{E}\mathbf{e} = \mathbf{0}$ and $\operatorname{Cov}\mathbf{e} = \sigma^2\mathbf{I}$. Assume the eigenvalues of $\mathbf{X}^T\mathbf{X}$ are all positive.
 - (a) Show that there is a unique solution to $\mathbf{X}^T \mathbf{X} \hat{\mathbf{b}} = \mathbf{X}^T \mathbf{y}$.
 - (b) Show that the mean squared error $\mathbb{E}\|\hat{\mathbf{b}} \mathbf{b}\|^2$ of $\hat{\mathbf{b}}$ is given by $\sigma^2 \sum_{i=1}^p \lambda_j^{-1}$, where $\lambda_1, \dots, \lambda_n$ are the eigenvalues of $\mathbf{X}^T \mathbf{X}$.
- 6. Let $\mathbf{X} = [\mathbf{x}_1 \cdots \mathbf{x}_p]$ be matrix with full column rank and with unit columns. Find the unit vector \mathbf{a} and the scalar A such that $\mathbf{a}^T \mathbf{x}_j = A$ for all $j = 1, \dots, p$. The vector \mathbf{a} is "equiangular" with every column of \mathbf{X} with angle equal to $\cos^{-1}(A)$.