

### Homework #3 EN.550.413

Problem 1: Let  $y = X\beta + \varepsilon$  be a linear model where  $X$  is of size  $n \times p$  and the error terms  $\varepsilon$  are independent, normally dist. with mean 0 and variance  $\sigma^2$ . Suppose furthermore that the columns of  $X$  can be partitioned as  $X = [W \ Z]$  where  $W$  is of size  $n \times q$  and is of full-column rank and  $Z$  is of size  $n \times (p-q)$  and is of full-column rank, for some  $q$  satisfying  $1 \leq q \leq p$ , and  $W^T Z = 0$ . We now partition  $\beta$  as  $\beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$  where  $\beta_1$  is of size  $q \times 1$  and  $\beta_2$  is of size  $(p-q) \times 1$ . Let  $\hat{\beta} = \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix}$  be the least square estimate of  $\beta$ .

(a) Show that  $\hat{\beta}_1 = (W^T W)^{-1} W^T y$  and  $\hat{\beta}_2 = (Z^T Z)^{-1} Z^T y$ .

Pf: Using the methods of least squares to determine the least square estimates, with the model  $y = X\beta + \varepsilon$ , we have

$$Q = (y - X\beta)^T (y - X\beta) = y^T y - y^T X\beta - \beta^T X^T y + \beta^T X^T X \beta$$

(and  $y^T X\beta = (y^T X\beta)^T = \beta^T X^T y$  since it is a  $1 \times 1$  matrix.)

$$= y^T y - 2\beta^T X^T y + \beta^T X^T X \beta$$

Then,  $\frac{\partial Q}{\partial \beta} = -2X^T y + 2X^T X \beta = 0 \Rightarrow X^T X \beta = X^T y$

Now,  $X^T X$  is not necessarily invertible, so we consider the partitioned/block matrix of  $X = [W \ Z]$  and  $\beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$ . We have,

$$X^T X \beta = \begin{bmatrix} W^T \\ Z^T \end{bmatrix} [W \ Z] \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} W^T \\ Z^T \end{bmatrix} [y] = X^T y$$

$$\Rightarrow \begin{bmatrix} W^T W & W^T Z \\ Z^T W & Z^T Z \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} W^T y \\ Z^T y \end{bmatrix}$$

Now, since  $W^T Z = 0_{p \times (p-q)} \Rightarrow (W^T Z)^T = Z^T W = 0_{(p-q) \times p}$

