

1. Suppose  $\mathbf{c}^T \boldsymbol{\beta}$  is an estimable function, i.e.,  $\mathbf{c}^T \in \mathcal{R}(\mathbf{X})$ , the row space of  $\mathbf{X}$ . Then show that

(a)  $\mathcal{R}(\mathbf{X}) = \mathcal{R}(\mathbf{X}^T \mathbf{X}) = \mathcal{C}(\mathbf{X}^T \mathbf{X})$ .

(b)  $\mathbf{c}^T \in \mathcal{R}(\mathbf{X}^T \mathbf{X})$  if and only if  $\mathbf{c}^T \mathbf{G} \mathbf{X}^T \mathbf{X} = \mathbf{c}^T$ , where  $\mathbf{G}$  is any generalized inverse of  $\mathbf{X}^T \mathbf{X}$ .

**Remark.** Note that, to check whether  $\mathbf{c}^T \in \mathcal{R}(\mathbf{X})$  is often a tedious job in practice but from (a) and (b), we get an equivalent condition for estimability of a linear function, i.e., by checking whether  $\mathbf{c}^T \mathbf{G} \mathbf{X}^T \mathbf{X} = \mathbf{c}^T$ .

2. Consider the linear model

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon},$$

with  $\mathbb{E}(\boldsymbol{\varepsilon}) = 0$  and  $\text{Var}(\boldsymbol{\varepsilon}) = \sigma^2 \mathbf{I}$ . Suppose  $\text{Rank}(\mathbf{X}) \leq p$ . Suppose  $\mathbf{c}_1^T \boldsymbol{\beta}$  and  $\mathbf{c}_2^T \boldsymbol{\beta}$  are both estimable functions, then show that  $\text{Cov}[\mathbf{c}_1^T \hat{\boldsymbol{\beta}}, \mathbf{c}_2^T \hat{\boldsymbol{\beta}}] = \sigma^2 \mathbf{c}_1^T \mathbf{G} \mathbf{c}_2$ , where  $\mathbf{G}$  is any generalized inverse of  $\mathbf{X}^T \mathbf{X}$ .

3. If  $\text{Rank}(\mathbf{X})$  has full rank,  $p$ , then we know  $\hat{\boldsymbol{\beta}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y}$  starting from the normal equations as described in the class and we also know  $\mathbb{E}(\hat{\boldsymbol{\beta}}) = \boldsymbol{\beta}$  is an unbiased estimator. Show:

(a)  $\text{Cov}(\hat{\boldsymbol{\beta}}) = \sigma^2 (\mathbf{X}^T \mathbf{X})^{-1}$ .

(b)  $\mathbf{c}^T \boldsymbol{\beta}$  is estimable for any  $\mathbf{c}^T \in \mathbb{R}^p$ .

(c)  $\text{Var}(\mathbf{c}^T \hat{\boldsymbol{\beta}}) = \sigma^2 \mathbf{c}^T (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{c}$ , for any  $\mathbf{c}^T \in \mathbb{R}^p$ .

(d)  $\text{Cov}(\mathbf{c}_1^T \hat{\boldsymbol{\beta}}, \mathbf{c}_2^T \hat{\boldsymbol{\beta}}) = \sigma^2 \mathbf{c}_1^T (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{c}_2$ , for any  $\mathbf{c}_1^T, \mathbf{c}_2^T \in \mathbb{R}^p$ .