

**Problems to be submitted for the Assignment:**

- 1) Text book, Chapter 3 Exercises #4.1, #4.2
- 2) Text book, Chapter 3 Exercises #4.4
- 3) Text book, Chapter 3 Exercise #4.17
- 4) Text book, Chapter 3 Exercise #4.19
- 5) Text book, Chapter 3 Exercise #4.26
- 6) Text book, Chapter 3 Exercise #4.28
- 7) Consider the general linear model  $\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$ ,  $E[\boldsymbol{\varepsilon}] = \mathbf{0}$ ,  $Var[\boldsymbol{\varepsilon}] = \sigma^2\mathbf{V}$ , where  $\mathbf{V}$  is a known positive definite matrix. Let  $\mathbf{M}$  be an arbitrary, symmetric positive definite matrix, and let  $\tilde{\boldsymbol{\beta}}_{\mathbf{M}}$  denote a solution to the following optimization problem:

$$\min_{\boldsymbol{\beta}} (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})' \mathbf{M} (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}).$$

Note that this solution is not unique when  $\mathbf{X}$  is not of full column rank. Note that the generalized least square estimator  $\tilde{\boldsymbol{\beta}}_{GLS}$  corresponds to  $\mathbf{M} = \mathbf{V}^{-1}$ .

Let  $l'\boldsymbol{\beta}$  be any estimable function.

- Show that  $E[l'\tilde{\boldsymbol{\beta}}_{\mathbf{M}}] = E[l'\tilde{\boldsymbol{\beta}}_{GLS}]$
- Show that  $Var(l'\tilde{\boldsymbol{\beta}}_{\mathbf{M}}) \geq Var[l'\tilde{\boldsymbol{\beta}}_{GLS}]$ .

[Remarks: Please do not assume that  $\mathbf{X}'\mathbf{M}\mathbf{X}$  is invertible. Note that  $l'\tilde{\boldsymbol{\beta}}_l$  is the ordinary least squares (OLS) estimator and  $l'\tilde{\boldsymbol{\beta}}_{GLS}$  is the Gauss-Markov estimator of an estimable function  $l'\boldsymbol{\beta}$ .]

**Try to work on as many of the following Unassigned Problems as possible:**

- 1) Text book, Chapter 4 Exercises #4.6, #4.7, #4.9, #4.11, #4.12, #4.13, #4.17, #4.20, #4.21