

by A.2.1]. Since $\mathbf{A}\boldsymbol{\beta} = \mathbf{M}\mathbf{X}\boldsymbol{\beta} = \mathbf{M}\boldsymbol{\theta}$, we therefore find the restricted least squares estimate of $\boldsymbol{\theta}$ by minimizing $\|\mathbf{Y} - \boldsymbol{\theta}\|^2$ subject to $\boldsymbol{\theta} \in \mathcal{C}(\mathbf{X}) = \Omega$ and $\mathbf{M}\boldsymbol{\theta} = \mathbf{0}$, that is, subject to

$$\boldsymbol{\theta} \in \mathcal{N}(\mathbf{M}) \cap \Omega \quad (= \omega, \text{ say}).$$

If \mathbf{P}_Ω and \mathbf{P}_ω are the projection matrices projecting onto Ω and ω , respectively, then we want to find $\hat{\boldsymbol{\theta}}_\omega = \mathbf{P}_\omega \mathbf{Y}$. Now, from B.3.2 and B.3.3,

$$\mathbf{P}_\Omega - \mathbf{P}_\omega = \mathbf{P}_{\omega^\perp \cap \Omega},$$

where $\omega^\perp \cap \Omega = \mathcal{C}(\mathbf{B})$ and $\mathbf{B} = \mathbf{P}_\Omega \mathbf{M}'$. Thus

$$\begin{aligned}\hat{\boldsymbol{\theta}}_\omega &= \mathbf{P}_\omega \mathbf{Y} \\ &= \mathbf{P}_\Omega \mathbf{Y} - \mathbf{P}_{\omega^\perp \cap \Omega} \mathbf{Y} \\ &= \hat{\boldsymbol{\theta}}_\Omega - \mathbf{B}(\mathbf{B}'\mathbf{B})^{-1}\mathbf{B}'\mathbf{Y}.\end{aligned}$$

EXERCISES 3j

$(\mathbf{z}) \perp \mathcal{C}(\mathbf{X}) \text{ so } \mathbf{P}\mathbf{z} = \mathbf{0}$

1. If \mathbf{P} projects onto $\mathcal{C}(\mathbf{X})$, show that $\mathbf{Z}'(\mathbf{I}_n - \mathbf{P})\mathbf{Z}$ is nonsingular.

2. Prove that if \mathbf{X}_1 is $n \times r$ of rank r and consists of a set of r linearly independent columns of \mathbf{X} , then $\mathbf{X} = \mathbf{X}_1 \mathbf{L}$, where \mathbf{L} is $r \times p$ of rank r .

3. Prove that \mathbf{B} has full column rank [i.e., $(\mathbf{B}'\mathbf{B})^{-1} = (\mathbf{B}'\mathbf{B})^{-1}$].

4. If \mathbf{X} has full rank and $\hat{\boldsymbol{\theta}}_\omega = \mathbf{X}\hat{\boldsymbol{\theta}}_H$, show that

$$\hat{\boldsymbol{\theta}}_H = \hat{\boldsymbol{\beta}} - (\mathbf{X}'\mathbf{X})^{-1}\mathbf{A}'(\mathbf{A}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{A}')^{-1}\mathbf{A}\hat{\boldsymbol{\beta}}.$$

[This is a special case of (3.38).] ↪

5. Show how to modify the theory above to take care of the case when the restrictions are $\mathbf{A}\boldsymbol{\beta} = \mathbf{c}$ ($\mathbf{c} \neq \mathbf{0}$).

3.10 GENERALIZED LEAST SQUARES

Having developed a least squares theory for the full-rank model $\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$, where $E[\boldsymbol{\varepsilon}] = \mathbf{0}$ and $\text{Var}[\boldsymbol{\varepsilon}] = \sigma^2 \mathbf{I}_n$, we now consider what modifications are necessary if we allow the ε_i to be correlated. In particular, we assume that $\text{Var}[\boldsymbol{\varepsilon}] = \sigma^2 \mathbf{V}$, where \mathbf{V} is a known $n \times n$ positive-definite matrix.

Since \mathbf{V} is positive-definite, there exists an $n \times n$ nonsingular matrix \mathbf{K} such that $\mathbf{V} = \mathbf{K}\mathbf{K}'$ (A.4.2). Therefore, setting $\mathbf{Z} = \mathbf{K}^{-1}\mathbf{Y}$, $\mathbf{B} = \mathbf{K}^{-1}\mathbf{X}$, and