

Solution to Exercise 12.16

***12.16** The $k \times k$ matrix $\mathbf{X}_\bullet^\top (\boldsymbol{\Sigma}^{-1} \otimes \mathbf{P}_W) \mathbf{X}_\bullet$ given in expression (12.66) is positive semidefinite by construction. Show this property explicitly by expressing the matrix in the form $\mathbf{A}^\top \mathbf{A}$, where \mathbf{A} is a matrix with k columns and at least k rows that should depend on a $g \times g$ nonsingular matrix $\boldsymbol{\Psi}$ which satisfies the relation $\boldsymbol{\Psi} \boldsymbol{\Psi}^\top = \boldsymbol{\Sigma}^{-1}$.

Show that a positive semidefinite matrix expressed in the form $\mathbf{A}^\top \mathbf{A}$ is positive definite if and only if \mathbf{A} has full column rank. In the present case, the matrix \mathbf{A} fails to have full column rank if and only if there exists a k -vector $\boldsymbol{\beta}$, different from zero, such that $\mathbf{A} \boldsymbol{\beta} = \mathbf{0}$. Since $k = \sum_{i=1}^g k_i$, we may write the vector $\boldsymbol{\beta}$ as $[\boldsymbol{\beta}_1 \ \cdots \ \boldsymbol{\beta}_g]$, where $\boldsymbol{\beta}_i$ is a k_i -vector for $i = 1, \dots, g$. Show that there exists a nonzero $\boldsymbol{\beta}$ such that $\mathbf{A} \boldsymbol{\beta} = \mathbf{0}$ if and only if, for at least one i , there is a nonzero $\boldsymbol{\beta}_i$ such that $\mathbf{P}_W \mathbf{X}_i \boldsymbol{\beta}_i = \mathbf{0}$, that is, if $\mathbf{P}_W \mathbf{X}_i$ does not have full column rank.

Show that, if $\mathbf{P}_W \mathbf{X}_i$ has full column rank, then there exists a unique solution of the estimating equations (12.60) for the parameters $\boldsymbol{\beta}_i$ of equation i .

If we set $\mathbf{A} \equiv (\boldsymbol{\Psi}^\top \otimes \mathbf{P}_W) \mathbf{X}_\bullet$, then we see that

$$\begin{aligned} \mathbf{A}^\top \mathbf{A} &= \mathbf{X}_\bullet^\top (\boldsymbol{\Psi} \otimes \mathbf{P}_W) (\boldsymbol{\Psi}^\top \otimes \mathbf{P}_W) \mathbf{X}_\bullet \\ &= \mathbf{X}_\bullet^\top (\boldsymbol{\Psi} \boldsymbol{\Psi}^\top \otimes \mathbf{P}_W) \mathbf{X}_\bullet = \mathbf{X}_\bullet^\top (\boldsymbol{\Sigma}^{-1} \otimes \mathbf{P}_W) \mathbf{X}_\bullet. \end{aligned}$$

In these manipulations, we make use of the fact that the orthogonal projection matrix \mathbf{P}_W is symmetric and idempotent. Since $\boldsymbol{\Psi}^\top \otimes \mathbf{P}_W$ is $gn \times gn$, and \mathbf{X}_\bullet is $gn \times k$, it follows that \mathbf{A} is a $gn \times k$ matrix. The requirement on the number of rows is satisfied because $gn > k$.

For the second part of the question, the positive semidefinite matrix $\mathbf{A}^\top \mathbf{A}$ is positive definite if and only if $\boldsymbol{\beta}^\top \mathbf{A}^\top \mathbf{A} \boldsymbol{\beta} = 0$ implies that $\boldsymbol{\beta} = \mathbf{0}$. But $\boldsymbol{\beta}^\top \mathbf{A}^\top \mathbf{A} \boldsymbol{\beta} = \|\mathbf{A} \boldsymbol{\beta}\|^2$, and so the quadratic form is zero if and only if $\|\mathbf{A} \boldsymbol{\beta}\| = 0$, that is, if and only if $\mathbf{A} \boldsymbol{\beta} = \mathbf{0}$. If this last relation implies that $\boldsymbol{\beta} = \mathbf{0}$, then by definition \mathbf{A} has full column rank.

The matrix \mathbf{A} can be expressed explicitly as follows:

$$\begin{aligned} \mathbf{A} &= (\boldsymbol{\Psi}^\top \otimes \mathbf{P}_W) \mathbf{X}_\bullet \\ &= \begin{bmatrix} \psi_{11} \mathbf{P}_W & \cdots & \psi_{g1} \mathbf{P}_W \\ \vdots & \ddots & \vdots \\ \psi_{1g} \mathbf{P}_W & \cdots & \psi_{gg} \mathbf{P}_W \end{bmatrix} \begin{bmatrix} \mathbf{X}_1 & \cdots & \mathbf{0} \\ \vdots & \ddots & \vdots \\ \mathbf{0} & \cdots & \mathbf{X}_g \end{bmatrix} \\ &= \begin{bmatrix} \psi_{11} \mathbf{P}_W \mathbf{X}_1 & \cdots & \psi_{g1} \mathbf{P}_W \mathbf{X}_g \\ \vdots & \ddots & \vdots \\ \psi_{1g} \mathbf{P}_W \mathbf{X}_1 & \cdots & \psi_{gg} \mathbf{P}_W \mathbf{X}_g \end{bmatrix}, \end{aligned}$$

where ψ_{ij} is the ij^{th} element of Ψ . We can postmultiply this by a k -vector β that is partitioned as $[\beta_1 \vdots \cdots \vdots \beta_g]$. The result is

$$\mathbf{A}\beta = \begin{bmatrix} \psi_{11}\mathbf{P}_W\mathbf{X}_1 & \cdots & \psi_{g1}\mathbf{P}_W\mathbf{X}_g \\ \vdots & \ddots & \vdots \\ \psi_{1g}\mathbf{P}_W\mathbf{X}_1 & \cdots & \psi_{gg}\mathbf{P}_W\mathbf{X}_g \end{bmatrix} \begin{bmatrix} \beta_1 \\ \vdots \\ \beta_g \end{bmatrix} = \begin{bmatrix} \sum_{j=1}^g \psi_{j1}\mathbf{P}_W\mathbf{X}_j\beta_j \\ \vdots \\ \sum_{j=1}^g \psi_{jg}\mathbf{P}_W\mathbf{X}_j\beta_j \end{bmatrix}.$$

If we stack the $n \times 1$ blocks of the $gn \times 1$ vector $\mathbf{A}\beta$ horizontally in an $n \times g$ matrix rather than vertically, we obtain

$$\begin{aligned} & [\sum_{j=1}^g \psi_{j1}\mathbf{P}_W\mathbf{X}_j\beta_j \quad \cdots \quad \sum_{j=1}^g \psi_{jg}\mathbf{P}_W\mathbf{X}_j\beta_j] \\ &= [\mathbf{P}_W\mathbf{X}_1\beta_1 \quad \cdots \quad \mathbf{P}_W\mathbf{X}_g\beta_g] \begin{bmatrix} \psi_{11} & \cdots & \psi_{1g} \\ \vdots & \ddots & \vdots \\ \psi_{g1} & \cdots & \psi_{gg} \end{bmatrix} \\ &= [\mathbf{P}_W\mathbf{X}_1\beta_1 \quad \cdots \quad \mathbf{P}_W\mathbf{X}_g\beta_g] \Psi. \end{aligned}$$

Clearly, the vector $\mathbf{A}\beta$ is zero if and only if the matrix in the last line above is zero. But Ψ is a nonsingular $g \times g$ matrix, and so $\mathbf{A}\beta$ is zero for arbitrary nonzero β if and only if the entire matrix $[\mathbf{P}_W\mathbf{X}_1\beta_1 \quad \cdots \quad \mathbf{P}_W\mathbf{X}_g\beta_g]$ is zero. But that can only be the case if $\mathbf{P}_W\mathbf{X}_i\beta_i$ is zero for all $i = 1, \dots, g$. Consequently, $\mathbf{A}\beta$ is zero with a nonzero β if and only if there is at least one i such that $\mathbf{P}_W\mathbf{X}_i\beta_i = \mathbf{0}$ with nonzero β_i .

We now turn to the third part of the question. If $\mathbf{X}_\bullet^\top(\Sigma^{-1} \otimes \mathbf{P}_W)\mathbf{X}_\bullet$ is nonsingular, then the equations (12.60) have a unique solution for β_\bullet , and the result is trivial. The only case that needs further study is therefore the one in which the matrix \mathbf{A} does not have full column rank. Suppose then that \mathbf{A} is of rank $r < k$. Then \mathbf{A} can be partitioned, possibly after a reordering of its columns, as $\mathbf{A} = [\mathbf{A}_1 \quad \mathbf{A}_1\mathbf{B}]$, where \mathbf{A}_1 is $gn \times r$ with full column rank, and \mathbf{B} is $r \times (k - r)$. This result simply makes explicit the fact that $k - r$ columns of \mathbf{A} are linear combinations of the other r columns.

Two points need to be established, namely, existence and uniqueness. For existence, observe that equations (12.60) can be written as

$$\mathbf{A}^\top(\mathbf{A}\beta_\bullet - (\Psi^\top \otimes \mathbf{P}_W)\mathbf{y}_\bullet) = \mathbf{0}. \quad (\text{S12.22})$$

Partition β_\bullet as $[\beta_\bullet^1 \vdots \beta_\bullet^2]$, conformably with the partition of \mathbf{A} , so that β_\bullet^1 has r elements, and β_\bullet^2 has $k - r$ elements. Thus $\mathbf{A}\beta_\bullet = \mathbf{A}_1(\beta_\bullet^1 + \mathbf{B}\beta_\bullet^2)$. Now set $\beta_\bullet^2 = \mathbf{0}$. Then we can show that equations (S12.22) have a unique solution for β_\bullet^1 . Indeed, equation (S12.22) becomes

$$\begin{bmatrix} \mathbf{A}_1^\top \\ \mathbf{B}^\top \mathbf{A}_1^\top \end{bmatrix} (\mathbf{A}_1\beta_\bullet^1 - (\Psi^\top \otimes \mathbf{P}_W)\mathbf{y}_\bullet) = \mathbf{0}. \quad (\text{S12.23})$$

Note that if

$$\mathbf{A}_1^\top (\mathbf{A}_1 \boldsymbol{\beta}_\bullet^1 - (\boldsymbol{\Psi}^\top \otimes \mathbf{P}_W) \mathbf{y}_\bullet) = \mathbf{0}, \quad (\text{S12.24})$$

then (S12.23) is true, since the last $k - r$ rows are just linear combinations of the first r rows. But \mathbf{A}_1 has full column rank of r , and so the $r \times r$ matrix $\mathbf{A}_1^\top \mathbf{A}_1$ is nonsingular. Thus the equations (S12.24) have a unique solution for the r -vector $\boldsymbol{\beta}_\bullet^1$, as claimed. Denote this solution by $\tilde{\boldsymbol{\beta}}_\bullet^1$, and by $\tilde{\boldsymbol{\beta}}_\bullet$ the k -vector $[\tilde{\boldsymbol{\beta}}_\bullet^1 \ ; \ \mathbf{0}]$.

In order to show uniqueness, observe that any other solution to equation (S12.22), say $\hat{\boldsymbol{\beta}}_\bullet$, is such that

$$\mathbf{A}^\top \mathbf{A} (\hat{\boldsymbol{\beta}}_\bullet - \tilde{\boldsymbol{\beta}}_\bullet) = \mathbf{0}. \quad (\text{S12.25})$$

This follows by subtracting (S12.22) evaluated at $\hat{\boldsymbol{\beta}}_\bullet$ from the same equation evaluated at $\tilde{\boldsymbol{\beta}}_\bullet$. If we write $\boldsymbol{\beta} \equiv \hat{\boldsymbol{\beta}}_\bullet - \tilde{\boldsymbol{\beta}}_\bullet$, then, by an argument used earlier, (S12.25) implies that $\mathbf{A}\boldsymbol{\beta} = \mathbf{0}$, and, by the second part of this exercise, this implies that $\mathbf{P}_W \mathbf{X}_i \boldsymbol{\beta}_i = \mathbf{0}$ for all $i = 1, \dots, g$, where $\boldsymbol{\beta}_i$ is defined as above as the i^{th} block of $\boldsymbol{\beta}$. If, for some i , $\mathbf{P}_W \mathbf{X}_i$ has full column rank, then it follows that $\boldsymbol{\beta}_i = \mathbf{0}$. Thus the two solutions $\tilde{\boldsymbol{\beta}}_\bullet$ and $\hat{\boldsymbol{\beta}}_\bullet$ have the same i^{th} block, which is therefore defined uniquely, as we wished to show.