PROBLEMS AND SOLUTIONS

PROBLEMS

04.2.1. A Range Equality for Block Matrices with Orthogonal Projectors

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When considering the general Gauss–Markov model $Y = X\beta + e$, where E(e) = 0, Cov(e) = V, one often needs to know whether the range inclusion $range(X) \subseteq range(V)$ holds. Show that if V is the following tridiagonal block matrix:

$$V = \begin{bmatrix} P + Q & PQ & & \\ QP & P + Q & \ddots & \\ & \ddots & \ddots & PQ \\ & & QP & P + Q \end{bmatrix}_{n \times n}$$

where P and Q are two orthogonal projectors of the same size, then

range(V) = range $\begin{bmatrix} P+Q & & \\ & \ddots & \\ & & P+Q \end{bmatrix}_{n \times n}$.

04.2.2. Characterizations of Hermitian Projectors

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Let P in $\mathbb{C}^{n \times n}$ be a projector with P^+ its Moore–Penrose inverse and P^H its conjugate transpose. Werner (2002) provides a list of equivalent conditions for P to be Hermitian: (i) $P = PP^HP$, (ii) $P^+P^+ = P^+$, (iii) $P^+ = P$, and (iv) $P^+ = P^H$. Extend this list and show that also condition (a) (resp. (b)) is sufficient and necessary for a projector P to be Hermitian:

- (a) the composition $P^H P$ is a projector,
- (b) the composition PP^H is a projector.

REFERENCE

Werner, H.J. (2002) Partial isometry and idempotent matrices. Solution 28-7.5. *IMAGE, The Bulletin of the International Linear Algebra Society* **29**, 31–32.

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SOLUTION

04.2.1. Fixed Effects Estimation of the Population-Averaged Slopes in a Panel Data Random Coefficient Model—Solution

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a. Just substitute $\mathbf{b}_i = \boldsymbol{\beta} + \mathbf{d}_i$ into (1) and rearrange:

$$y_{it} = a_i + \mathbf{x}_{it}(\boldsymbol{\beta} + \mathbf{d}_i) + u_{it} = a_i + \mathbf{x}_{it}\boldsymbol{\beta} + (\mathbf{x}_{it}\mathbf{d}_i + u_{it})$$
$$\equiv a_i + \mathbf{x}_{it}\boldsymbol{\beta} + v_{it}, \qquad t = 1, \dots, T,$$

where $v_{it} \equiv \mathbf{x}_{it} \mathbf{d}_i + u_{it}$.

b. The time-demeaned equation is simply

$$\ddot{\mathbf{y}}_{it} = \ddot{\mathbf{x}}_{it} \boldsymbol{\beta} + (\ddot{\mathbf{x}}_{it} \mathbf{d}_i + \ddot{u}_{lt}) = \ddot{\mathbf{x}}_{it} \boldsymbol{\beta} + \ddot{v}_{it}, \qquad t = 1, \dots, T,$$
(5)

where $\ddot{v}_{it} \equiv \ddot{\mathbf{x}}_{it} \mathbf{d}_i + \ddot{u}_{it}, t = 1, \dots, T.$

c. The fixed effects estimator is simply the pooled ordinary least squares estimator applied to the time-demeaned equation (5). So we can write the fixed effects estimator as

$$\hat{\boldsymbol{\beta}} = \left(N^{-1}\sum_{i=1}^{N}\sum_{t=1}^{T}\mathbf{\ddot{x}}_{it}'\mathbf{\ddot{x}}_{it}\right)^{-1} \left(N^{-1}\sum_{i=1}^{N}\sum_{t=1}^{T}\mathbf{\ddot{x}}_{it}'\mathbf{\ddot{y}}_{it}\right)$$
$$= \boldsymbol{\beta} + \left(N^{-1}\sum_{i=1}^{N}\sum_{t=1}^{T}\mathbf{\ddot{x}}_{it}'\mathbf{\ddot{x}}_{it}\right)^{-1} \left(N^{-1}\sum_{i=1}^{N}\sum_{t=1}^{T}\mathbf{\ddot{x}}_{it}'\mathbf{\ddot{v}}_{it}\right)$$
$$= \boldsymbol{\beta} + \left(N^{-1}\sum_{i=1}^{N}\mathbf{\ddot{x}}_{i}'\mathbf{\ddot{x}}_{i}\right)^{-1} \left(N^{-1}\sum_{i=1}^{N}\sum_{t=1}^{T}\mathbf{\ddot{x}}_{it}'\mathbf{\ddot{v}}_{it}\right).$$

By the rank condition (3) and the weak law of large numbers (WLLN), $(N^{-1}\sum_{i=1}^{N} \ddot{\mathbf{X}}_{i}' \ddot{\mathbf{X}}_{i})^{-1} \xrightarrow{p} \mathbf{A}^{-1}$, where $\mathbf{A} \equiv \mathbf{E}(\ddot{\mathbf{X}}_{i}' \ddot{\mathbf{X}}_{i})$ Consistency follows, again by the WLLN, if

$$\mathbf{E}(\mathbf{\ddot{x}}_{it}'\mathbf{\ddot{v}}_{it}) = \mathbf{0}, \qquad t = 1, \dots, T.$$

Sufficient is $E(\mathbf{\ddot{x}}_{it}\mathbf{\ddot{x}}_{it}\mathbf{d}_i) = \mathbf{0}$ and $E(\mathbf{\ddot{x}}_{it}\mathbf{\ddot{u}}_{it}) = \mathbf{0}$, t = 1, ..., T. But $E(\mathbf{\ddot{u}}_{it}|\mathbf{\ddot{x}}_{it}) = 0$, t = 1, ..., T. But $E(\mathbf{\ddot{u}}_{it}|\mathbf{\ddot{x}}_{it}) = 0$, t = 1, ..., T. But $E(\mathbf{\ddot{u}}_{it}|\mathbf{\ddot{x}}_{it}) = 0$, t = 1, ..., T. Further, (4) is equivalent to $E(\mathbf{d}_i|\mathbf{\ddot{x}}_{it}) = \mathbf{0}$, t = 1, ..., T; the law of iterated expectations then implies that $E(\mathbf{\ddot{x}}_{it}\mathbf{\ddot{x}}_{it}\mathbf{d}_i) = \mathbf{0}$.

Condition (4) only restricts the mean dependence between \mathbf{b}_i , and the timedemeaned variables, $\ddot{\mathbf{x}}_{ii}$; it places no direct restriction on the dependence between \mathbf{b}_i , and $\bar{\mathbf{x}}_i$. Often, it is reasonable to think that although the time averages of some covariates would be correlated with unobserved heterogeneity—in particular, with \mathbf{b}_i —deviations about the time average might not be. So the fixed effects estimator, although not robust to arbitrary dependence between \mathbf{b}_i and $(\mathbf{x}_{i1}, \mathbf{x}_{i2}, \dots, \mathbf{x}_{iT})$, does have some useful robustness properties in the general random coefficient panel data model.

d. Zero correlation does not suffice because we need \mathbf{b}_i to be uncorrelated with $\mathbf{\ddot{x}}'_{it}\mathbf{\ddot{x}}_{it}, t = 1, 2, ..., T$.