

## Problem Set 5

1.  $Y_t = \Phi_1 Y_{t-1} + \Phi_2 Y_{t-2} + \varepsilon_t, \quad \varepsilon_t \sim WN(0, \Sigma), \quad Y_t \in \mathbb{R}^{k \times 1}$

(a) The companion form of the VAR(2) model:

$$\begin{pmatrix} Y_t \\ Y_{t-1} \end{pmatrix} = \begin{pmatrix} \Phi_1 & \Phi_2 \\ I_k & 0 \end{pmatrix} \begin{pmatrix} Y_{t-1} \\ Y_{t-2} \end{pmatrix} + \begin{pmatrix} \varepsilon_t \\ 0 \end{pmatrix}$$

where  $I_k$  is the  $(k \times k)$  identity matrix.

This is a VAR(1) model for  $Y_t$ :

$$Y_t = \Phi Y_{t-1} + U_t$$

where  $Y_t = (y_t', y_{t-1}')'$ ,  $U_t = (\varepsilon_t', 0)'$  are  $(2k \times 1)$  vectors and  $\Phi$  is a  $(2k \times 2k)$  matrix.

(b)  $Y_t$  is weakly stationary if the eigenvalues of  $\Phi$  are less than 1 in modulus.

Solve

$$\begin{aligned} 0 &= |\Phi - \lambda I_{2k}| = \det \left[ \begin{pmatrix} \Phi_1 - \lambda I_k & \Phi_2 \\ I_k & -\lambda I_k \end{pmatrix} \right] \\ &= \det(-\lambda I_k) \det(\Phi_1 - \lambda I_k - \Phi_2(-\lambda I_k)^{-1} I_k) \end{aligned}$$

Simplify the following:

$$\textcircled{1} \quad -\lambda I_k = \text{diag}(-\lambda, -\lambda, \dots, -\lambda)$$

$$\Rightarrow \det(-\lambda I_k) = \prod_{i=1}^k (-\lambda) = (-1)^k \lambda^k$$

$$\textcircled{2} \quad \Phi_2(-\lambda I_k)^{-1} I_k = \Phi_2(-\lambda)^{-1} I_k^{-1} I_k = \frac{1}{\lambda} \Phi_2$$

$$\Rightarrow \det(\Phi_1 - \lambda I_k - \Phi_2(-\lambda I_k)^{-1} I_k)$$

$$= \det(\Phi_1 - \lambda I_k + \frac{1}{\lambda} \Phi_2)$$

$$= \det \left[ \frac{1}{\lambda} (\lambda \Phi_1 - \lambda^2 I_k + \Phi_2) \right]$$

$$= \frac{1}{\lambda^k} \det(\Phi_2 + \lambda \Phi_1 - \lambda^2 I_k),$$

$$= \frac{1}{\lambda^k} \det[-(\lambda^2 I_k - \lambda \Phi_1 - \Phi_2)]$$

$$= \frac{(-1)^k}{\lambda^k} \det(\lambda^2 I_k - \lambda \Phi_1 - \Phi_2)$$

by the scalar property of determinants.

The above expressions allow us to simplify

$$\begin{aligned} 0 &= |\underline{\Phi} - \lambda I_k| = (-1)^k \lambda^k \frac{(-1)^k}{\lambda^k} \det(\lambda^2 I_k - \lambda \underline{\Phi}_1 - \underline{\Phi}_2) \\ &= (-1)^{2k} \det(\lambda^2 I_k - \lambda \underline{\Phi}_1 - \underline{\Phi}_2) \\ &= \det(\lambda^2 I_k - \lambda \underline{\Phi}_1 - \underline{\Phi}_2) \\ &= |I_k \lambda^2 - \underline{\Phi}_1 \lambda - \underline{\Phi}_2|, \end{aligned}$$

which brings us to the characteristic polynomial equation for finding the roots  $\lambda$  (i.e. eigenvalues) of  $y_t$ . When  $|\lambda| < 1$ , for all  $\lambda$ , the given VAR(2) model is weakly stationary.

(c) Take  $y_t = \underline{\Phi}_1 y_{t-1} + \underline{\Phi}_2 y_{t-2} + \varepsilon_t$

$$\begin{aligned} &= \underline{\Phi}_1 L y_t + \underline{\Phi}_2 L^2 y_t + \varepsilon_t \\ \Rightarrow (I_k - \underline{\Phi}_1 L - \underline{\Phi}_2 L^2) y_t &= \varepsilon_t \end{aligned}$$

The MA coefficients  $\psi_i$  are found by solving

$$(I_k - \underline{\Phi}_1 L - \underline{\Phi}_2 L^2)(I_k + \psi_1 L + \psi_2 L^2 + \psi_3 L^3 + \dots) = I_k$$

$$\Rightarrow \underline{\Phi}(L) \psi(L) = I_k$$

$$\psi(L) = \underline{\Phi}^{-1}(L)$$

Equating the coefficients of  $L, L^2, L^3$ , we get

$$\psi_1 = \underline{\Phi}_1,$$

$$\psi_2 = \underline{\Phi}_1 \psi_1 + \underline{\Phi}_2,$$

$$\psi_3 = \underline{\Phi}_1 \psi_2 + \underline{\Phi}_2 \psi_1.$$

2. (a)

$$\begin{pmatrix} y_{1t} \\ y_{2t} \end{pmatrix} = \begin{pmatrix} 0.4 & 0.6 \\ 0.25 & 0.5 \end{pmatrix} \begin{pmatrix} y_{1,t-1} \\ y_{2,t-1} \end{pmatrix} + \begin{pmatrix} 0.1 & 0.4 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} y_{1,t-2} \\ y_{2,t-2} \end{pmatrix} + \begin{pmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \end{pmatrix}$$

$$\varepsilon_t \sim WN(0, \Sigma)$$

Using our results from part 1 (b),

$$0 = |\Phi - \lambda I_2| = \det(I_2 \lambda^2 - \Phi, \lambda - \Phi_2) \\ = \det \begin{pmatrix} \lambda^2 - 0.4\lambda - 0.1 & -0.6\lambda - 0.4 \\ -0.25\lambda & \lambda^2 - 0.5\lambda \end{pmatrix}$$

Check  $\lambda = 1$ :

$$\det \begin{pmatrix} 1 - 0.4 - 0.1 & -0.6 - 0.4 \\ -0.25 & 1 - 0.5 \end{pmatrix} \\ = \det \begin{pmatrix} 0.5 & -1 \\ -0.25 & 0.5 \end{pmatrix} \quad \text{Note } \text{col}(1) = -\frac{1}{2} \text{col}(2) \\ \Rightarrow \text{linearly dependent columns} \Rightarrow \det = 0 \\ = (0.5)(0.5) - (-1)(-0.25) \\ = 0.25 - 0.25 \\ = 0$$

$\therefore \lambda = 1$  is a root of the determinant  
 $\Rightarrow y_t$  is unit root non-stationary

(b) VECM:

$$\Delta y_t + y_{t-1} = \Phi_1 y_{t-1} + \Phi_2 (y_{t-1} - \Delta y_{t-1}) + \varepsilon_t \\ \Rightarrow \Delta y_t = (-I_k + \Phi_1 + \Phi_2) y_{t-1} - \Phi_2 \Delta y_{t-1} + \varepsilon_t$$

$$\text{Let } \Pi = I_k - \Phi_1 - \Phi_2, \Gamma_1 = -\Phi_2.$$

$$\Rightarrow \Delta y_t = -\Pi y_{t-1} + \Gamma_1 \Delta y_{t-1} + \varepsilon_t.$$

$$\therefore \begin{pmatrix} y_{1t} - y_{1,t-1} \\ y_{2t} - y_{2,t-1} \end{pmatrix} = \begin{pmatrix} 0.5 & -1 \\ -0.25 & 0.5 \end{pmatrix} \begin{pmatrix} y_{1,t-1} \\ y_{2,t-1} \end{pmatrix} \\ + \begin{pmatrix} -0.1 & -0.4 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} y_{1,t-2} - y_{1,t-1} \\ y_{2,t-2} - y_{2,t-1} \end{pmatrix} + \begin{pmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \end{pmatrix}$$

(c) Long-Run Relation:

In part (a), we showed the model is a bivariate cointegrated VAR process.

$\Rightarrow \exists (2 \times 1)$  vector  $\beta = (\beta_1, \beta_2)'$  such that

$$\beta' y_t = \beta_1 y_{1t} + \beta_2 y_{2t} \sim I(0).$$

Normalize  $\beta$  to  $\tilde{\beta} = (I_2, -\tilde{\beta})$ . Then

$$\tilde{\beta}' y_t = y_{1t} - \tilde{\beta} y_{2t},$$

which implies

$$y_{1t} = \tilde{\beta} y_{2t} \text{ (long-run equilibrium relation).}$$

Let  $\alpha = (\alpha_1, \alpha_2)'$  be a vector of coefficients.

Using matrix  $\Pi$  from part (b), we solve

$$\Pi = \alpha \tilde{\beta}'$$

$$\begin{pmatrix} 0.5 & -1 \\ -0.25 & 0.5 \end{pmatrix} = \begin{pmatrix} \alpha_1 & -\alpha_1 \tilde{\beta} \\ \alpha_2 & -\alpha_2 \tilde{\beta} \end{pmatrix}$$

$$\alpha_1 = 0.5 \Rightarrow -(0.5)\tilde{\beta} = -1 \Rightarrow \tilde{\beta} = 2 \quad \checkmark$$

$$\alpha_2 = -0.25 \Rightarrow -(-0.25)\tilde{\beta} = 0.5 \Rightarrow \tilde{\beta} = 2 \quad \checkmark$$

$$\therefore \beta = (I_2, -2)$$

$$\Rightarrow y_{1t} = -2 y_{2t}$$

### 3. Trivariate Process

$$Y_t = \begin{pmatrix} X_t \\ Y_t \\ Z_t \end{pmatrix} = \begin{pmatrix} 0.8 & 0.1 & 0.1 \\ -0.16 & 1.08 & 0.08 \\ 0.24 & -0.12 & 0.88 \end{pmatrix} \begin{pmatrix} X_{t-1} \\ Y_{t-1} \\ Z_{t-1} \end{pmatrix} + \begin{pmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \end{pmatrix}$$

$$\varepsilon_t \sim (\varepsilon_{1t} \ \varepsilon_{2t} \ \varepsilon_{3t})' \sim WN(0, \Sigma), \quad \Sigma \text{ SPD}$$

(a) VECM:

$$\begin{aligned} & \begin{pmatrix} X_t - X_{t-1} \\ Y_t - Y_{t-1} \\ Z_t - Z_{t-1} \end{pmatrix} + \begin{pmatrix} X_{t-1} \\ Y_{t-1} \\ Z_{t-1} \end{pmatrix} \\ &= \begin{pmatrix} 0.8 & 0.1 & 0.1 \\ -0.16 & 1.08 & 0.08 \\ 0.24 & -0.12 & 0.88 \end{pmatrix} \begin{pmatrix} X_{t-1} \\ Y_{t-1} \\ Z_{t-1} \end{pmatrix} + \begin{pmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \end{pmatrix} \\ \begin{pmatrix} \Delta X_t \\ \Delta Y_t \\ \Delta Z_t \end{pmatrix} &= \left[ \begin{pmatrix} 0.8 & 0.1 & 0.1 \\ -0.16 & 1.08 & 0.08 \\ 0.24 & -0.12 & 0.88 \end{pmatrix} - \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right] \begin{pmatrix} X_{t-1} \\ Y_{t-1} \\ Z_{t-1} \end{pmatrix} + \begin{pmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \end{pmatrix} \\ &= \begin{pmatrix} -0.2 & 0.1 & 0.1 \\ -0.16 & 0.08 & 0.08 \\ 0.24 & -0.12 & -0.92 \end{pmatrix} \begin{pmatrix} X_{t-1} \\ Y_{t-1} \\ Z_{t-1} \end{pmatrix} + \begin{pmatrix} \varepsilon_{1t} \\ \varepsilon_{2t} \\ \varepsilon_{3t} \end{pmatrix} \end{aligned}$$

$$\Rightarrow \Delta Y_t = -\Pi Y_{t-1} + \varepsilon_t$$

$$\text{where } \Pi = \begin{pmatrix} 0.2 & -0.1 & -0.1 \\ 0.16 & -0.08 & -0.08 \\ -0.24 & 0.12 & 0.92 \end{pmatrix}$$

(b) Reduced Row-Echelon form:

$$\begin{array}{l} R2 - 0.8R1 \\ R3 + 1.2R1 \end{array} \rightarrow \begin{pmatrix} 0.2 & -0.1 & -0.1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\Rightarrow \text{Rank}(\Pi) = 1 < 3 \Rightarrow m = 1,$$

where  $m$  is the number of cointegrating relations

$$\text{write } \Pi = \underset{(3 \times 3)}{\alpha} \cdot \underset{(3 \times 1)}{\beta'},$$

$$\text{where } \text{Rank}(\alpha) = \text{Rank}(\beta) = 1$$

The VECM model in part 3(a) becomes

$$\Delta Y_t = -\alpha \beta' Y_{t-1} + \varepsilon_t$$

$$\text{Let } \beta = (I_1, \beta'_1)', \quad \underset{(2 \times 1)}{\beta_1} = \begin{pmatrix} \tilde{\beta}_1 \\ \tilde{\beta}_2 \end{pmatrix}, \text{ such that}$$

$$\beta' Y_t = x_t + \tilde{\beta}_1 y_t + \tilde{\beta}_2 z_t \sim I(0)$$

$$\Rightarrow \Pi = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} (I \ \tilde{\beta}_1 \ \tilde{\beta}_2) = \begin{pmatrix} 0.2 & -0.1 & -0.1 \\ 0.16 & -0.08 & -0.08 \\ -0.24 & 0.12 & 0.92 \end{pmatrix}$$

$$\Rightarrow \begin{cases} \alpha_1 = 0.2 \\ \alpha_1 \tilde{\beta}_1 = 0.2 \tilde{\beta}_1 = -0.1 \Rightarrow \tilde{\beta}_1 = -0.5 = \tilde{\beta}_2 \end{cases}$$

\* Normalized cointegrating vector:

$$\beta = (1 \ -0.5 \ -0.5)'$$

\* Cointegrating relation:

$$\beta' Y_t = x_t - 0.5 y_t - 0.5 z_t$$

\* Long-run relation:

$$x_t = 0.5 y_t + 0.5 z_t$$

\* Speed of adjustment coefficients:

$$\alpha = (0.2 \ 0.16 \ -0.24)'$$

(c) Since we assume the standard regularity conditions hold, then OLS estimation of the model will yield consistent and efficient estimates of the VAR coefficients.

→ Consistent Estimates:

The series  $\{x_t\}$ ,  $\{y_t\}$ , and  $\{z_t\}$  are cointegrated such that a stationary ( $I(0)$ ) linear combination exists, so we have consistent estimates of the cointegrating vector and adjustment parameters.

→ Efficient Estimates:

Every equation includes the same set of lagged variables, so we do not need to estimate each equation separately by OLS.