

Econ 2142, Fall 2013

# Answers to practice midterm questions

1a.)

$$x_t = \mu + (1 + \theta L)\varepsilon_t$$

$$(1 - \rho L)x_t = (1 - \rho L)\mu + (1 - \rho L)(1 + \theta L)\varepsilon_t$$

$$y_t = (1 - \rho)\mu + [1 + (\theta - \rho)L - \theta\rho L^2]\varepsilon_t$$

$$\hat{P}(y_t|1, \varepsilon_{t-1}, \varepsilon_{t-2}) = (1 - \rho)\mu + (\theta - \rho)\varepsilon_{t-1} - \theta\rho\varepsilon_{t-2}$$

1b.)  $y_t \sim \text{MA}(2)$  which is stationary for all  $\rho$

$$E(y_t) = (1 - \rho)\mu$$

$$s_Y(\omega) = \frac{\sigma^2(1 + \theta e^{-i\omega})(1 - \rho e^{-i\omega})(1 + \theta e^{i\omega})(1 - \rho e^{i\omega})}{2\pi}$$

1c.) This is proportional to the spectrum at frequency 0, which from part (b) is zero.

1d.) Can equivalently think of starting with  $(1 - \rho L)\varepsilon_t$  and then filtering by  $(1 + \theta L)$ . The representation  $(1 - \rho L)\varepsilon_t$  is noninvertible, but its autocovariances  $\gamma_0 = (1 + \rho^2)\sigma^2$  and  $\gamma_1 = \rho\sigma^2$  are identical to those of  $(1 - \rho^{-1}L)v_t$  with  $E(v_t^2) = \tau^2$ . Solving the equation for  $\gamma_1$  we have  $2\sigma^2 = 0.5\tau^2$  or  $\tau^2 = 4\sigma^2$ . The invertible representation is thus

$$\begin{aligned} y_t &= (1 - \rho)\mu + (1 - \rho^{-1}L)(1 + \theta L)v_t \\ &= c + v_t + (\theta - 0.5)v_{t-1} - 0.5\theta v_{t-2}. \end{aligned}$$

2a.) Need eigenvalues of  $\mathbf{F}$  inside unit circle for

$$\mathbf{F} = \begin{bmatrix} \Phi_1 & \Phi_2 \\ \mathbf{I}_n & \mathbf{0} \end{bmatrix}$$

or equivalently that solutions  $z$  to  $|\mathbf{I}_n - \Phi_1 z - \Phi_2 z^2| = 0$  are outside unit circle.

2b.) Let

$$\mathbf{x}_t = \begin{bmatrix} 1 \\ \mathbf{y}_{t-1} \\ \mathbf{y}_{t-2} \end{bmatrix}$$

$$\mathbf{\Pi} = [\mathbf{c} \quad \Phi_1 \quad \Phi_2]$$

$$\hat{\mathbf{\Pi}} = \left( \sum_{t=1}^T \mathbf{y}_t \mathbf{x}_t' \right) \left( \sum_{t=1}^T \mathbf{x}_t \mathbf{x}_t' \right)^{-1}$$

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or equivalently row  $i$  of  $\hat{\Pi}$  is obtained by OLS regression of  $y_{it}$  on  $\mathbf{x}_t$ . Also

$$\hat{\Omega} = T^{-1} \sum_{t=1}^T \left( \mathbf{y}_t - \hat{\Pi} \mathbf{x}_t \right) \left( \mathbf{y}_t - \hat{\Pi} \mathbf{x}_t \right)'$$

2c.) Stated assumptions along with stationarity condition are sufficient.

2d.)

$$\frac{\partial \mathbf{y}_{t+2}}{\partial \boldsymbol{\varepsilon}'_t} = \boldsymbol{\Phi}_1^2 + \boldsymbol{\Phi}_2$$

2e.)

$$(\mathbf{I}_n - \boldsymbol{\Lambda}L)\mathbf{B}_0\mathbf{y}_t = (\mathbf{I}_n - \boldsymbol{\Lambda}L)\boldsymbol{\lambda} + (\mathbf{I}_n - \boldsymbol{\Lambda}L)\mathbf{B}_1\mathbf{y}_{t-1} + \mathbf{v}_t$$

$$\mathbf{B}_0\mathbf{y}_t = \mathbf{k} + \mathbf{C}_1\mathbf{y}_{t-1} + \mathbf{C}_2\mathbf{y}_{t-1} + \mathbf{v}_t$$

for  $\mathbf{k} = (\mathbf{I}_n - \boldsymbol{\Lambda})\boldsymbol{\lambda}$ ,  $\mathbf{C}_1 = \boldsymbol{\Lambda}\mathbf{B}_0 + \mathbf{B}_1$ ,  $\mathbf{C}_2 = -\boldsymbol{\Lambda}\mathbf{B}_1$ . Thus

$$\hat{\mathbf{B}}_0^{-1}(\hat{\mathbf{B}}_0^{-1})' = \hat{\Omega}$$

$$\hat{\mathbf{B}}_0^{-1}(\hat{\boldsymbol{\Lambda}}\hat{\mathbf{B}}_0 + \hat{\mathbf{B}}_1) = \hat{\boldsymbol{\Phi}}_1$$

$$-\hat{\mathbf{B}}_0^{-1}\hat{\boldsymbol{\Lambda}}\hat{\mathbf{B}}_1 = \hat{\boldsymbol{\Phi}}_2$$

$$\hat{\mathbf{B}}_0^{-1}(\mathbf{I}_n - \hat{\boldsymbol{\Lambda}})\hat{\boldsymbol{\lambda}} = \hat{\mathbf{c}}$$

2e.)

$$\frac{\partial \mathbf{y}_{t+s}}{\partial \mathbf{v}'_t} = \hat{\boldsymbol{\Psi}}_s \hat{\mathbf{B}}_0^{-1}$$

where  $\hat{\boldsymbol{\Psi}}_s$  denotes the first  $n$  rows and  $n$  columns of  $\hat{\mathbf{F}}^s$  and

$$\hat{\mathbf{F}} = \begin{bmatrix} \hat{\boldsymbol{\Phi}}_1 & \hat{\boldsymbol{\Phi}}_2 \\ \mathbf{I}_n & \mathbf{0} \end{bmatrix}$$