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Solution to Final Exam Practice Problems

1.
$$\begin{pmatrix} y_t \\ x_t \end{pmatrix} f_{t-1} \sim N \begin{pmatrix} \beta \gamma y_{t-1} + \beta \gamma x_{t-1} + \delta y_{t-1} \\ \gamma y_{t-1} \end{pmatrix}$$

(a) and $D(\mathbf{x}_t \mid \mathbf{x}_{t-1}) \neq D(\mathbf{x}_t \mid \mathbf{x}_{t-1}, y_{t-1})$
 $\Longrightarrow y_t$ G.C. \mathbf{x}_t w.r.t. $\mathbf{x}_{t-1}, y_{t-1}$

1. b. The same holds the opposite way

 $\underset{\text{(i.e., } y_t \mid \mathbf{x}_{t-1}, y_{t-1} \sim \text{IN}((\beta\gamma + \delta)y_{t-1} + \beta\gamma x_{t-1}, w) \text{ for some w and this is} }$ not the same conditional distribution as that of $y_t \mid y_{t-1}$).

1. c. Notice using the formula from class that

$$y_t \mid \mathbf{x}_t \sim \mathrm{IN}\left((\beta\gamma + \delta)y_{t-1} + \beta\gamma x_{t-1} + \frac{\beta\sigma_{22} + \sigma_{12}}{\sigma_{22}}(x_t - \gamma y_{t-1}), w\right)$$
$$w = \sigma_{11} - \frac{\sigma_{12}^2}{\sigma_{22}}$$

So β cannot unambiguously be obtained from the conditional distribution, unless $\sigma_{12} = 0$. If $\sigma_{12} = 0$, then we can estimate the conditional model (given above), i.e., regress y_t on x_t , x_{t-1} and y_{t-1} given the original equation

$$\mathbf{y}_t = \beta x_t + \beta \gamma x_{t-1} + \delta y_{t-1} + \epsilon_{yt}$$

2. (a)
$$\Delta y_t = \epsilon_{2t} \Longrightarrow y_t$$
 is I(1)
 $\Longrightarrow x_t$ is I(1).

1. b. $\mathbf{x}_t - \alpha y_t = \epsilon_{1t} \implies$ they are cointegrated with cointegrating vector $(1, -\alpha).$

c.
$$\mathbf{x}_t = \alpha(y_{t-1} + \epsilon_{2t}) + \epsilon_{1t}$$

$$= (x_{t-1} - \epsilon_{1t-1}) + \alpha \epsilon_{2t} + \epsilon_{1t}$$

$$\operatorname{VAR} \begin{cases} x_t = x_{t-1} + \alpha \epsilon_{2t} + \epsilon_{1t} - (x_{t-1} - \alpha y_{t-1}) \\ y_t = y_{t-1} + \epsilon_{2t} \end{cases}$$

$$\Longrightarrow \operatorname{VEC} \begin{cases} \Delta x_t = -z_{t-1} + (\alpha \epsilon_{2t} + \epsilon_{1t}) \\ \Delta y_t = \epsilon_{2t} \end{cases}$$

$$z_{t-1} = (x_{t-1} - \alpha y_{t-1})$$

A more useful solution strategy is

In general, say have an equation with contemporaneous variables in it. Say, for example, that $x_t = \alpha x_{t-1} + \beta y_t + \gamma y_{t-1} + \epsilon_{xt}$ then $(1 - \alpha B) x_t = (\beta + \gamma B) y_t + \epsilon_{xt}$ $\implies x_t - \frac{(\beta + \gamma B)}{(1 - \alpha \beta)} y_t = \frac{\epsilon_{xt}}{(1 - \alpha B)}$

and assuming $|\alpha| < 1 \Longrightarrow \frac{\epsilon_{xt}}{(1-\alpha B)}$ has finite variance $\Longrightarrow x_t - \frac{(\beta+\gamma B)}{(1-\alpha B)} y_t$ is stationary and at B = 1 we get the CI vector

 \implies CI vector is $\left(1, -\frac{(\beta+\gamma)}{(1-\alpha)}\right)$

3. $y_t \mid X_t, f_{t-1} \sim N(X_t\beta + \rho_1 u_{t-1} + \rho_2 u_{t-2}, \sigma_{\epsilon}^2)$

 $\begin{aligned} u_{t-1} &= y_{t-1} - X_{t-1}\beta \\ u_{t-2} &= y_{t-2} - X_{t-2}\beta \\ &\Longrightarrow y_t \mid X_t, \, f_{t-1} \sim N((X_t - \rho_1 X_{t-1} - \rho_2 X_{t-2})\beta + \rho_1 y_{t-2} + \rho_2 y_{t-2}, \sigma_\epsilon^2) \\ &\operatorname{H}_0: \rho_1 &= \rho_2 = 0 \end{aligned}$

Under $H_0, y_t \mid X_t, f_{t-1} \sim N(X_t\beta, \sigma_{\epsilon}^2)$ so Step 1 is: Regress y_t on X_t , construct u_t .

Now $g(\cdot) = (X_t - \rho_1 X_{t-1} - \rho_2 X_{t-2})\beta + \rho_1 y_{t-1} + \rho_2 y_{t-2}$ $g\beta \mid_{\rho_1 = \rho_2 = 0} = X_t$ $g\rho_1 \mid_{\rho_1 = \rho_2 = 0} = (X_{t-1}\beta - Y_{t-1}) = u_{t-1}$ $g\rho_2 \mid_{\rho_1 = \rho_2 = 0} = (X_{t-2}\beta - Y_{t-2}) = u_{t-2}$

So Step 2. Regress \hat{u}_t on X_t , \hat{u}_{t-1} , \hat{u}_{t-2} and $TR^2 \sim \chi_2^2$ from this regression, where T = the number of observations used.

Also, as in class, note that

 $\xi_{LM} = S(y; \tilde{\theta}) I^{-1}(\tilde{\theta}) S(y; \tilde{\theta})$ and you should try to derive these scores and informations of the test statistic directly in addition to the work above.