MFE230E Problem Set 4

Due April 21 6:00pm via bCourses

You may NOT use the statsmodels VAR and VECM modules, or any other VAR/VECM package, for this problem set but you may use the statsmodels OLS module.

1. Show that

dels VAR and VECM modules, or any other VAR/VECM package, se the statsmodels OLS module.
$$P_t = \mathsf{E}_t \sum_{i=1}^\infty \frac{D_{t+i}}{(1+R)^i}$$

$$\Longrightarrow P_t - \frac{D_t}{R} = \frac{1}{R} \, \mathsf{E}_t \sum_{i=0}^\infty \frac{\Delta D_{t+1+i}}{(1+R)^i}.$$

What does this equation imply for cointegration of prices and dividends?

2. Start from the definition of returns:

$$R_{t+1} = \frac{P_{t+1} + D_{t+1}}{P_t} - 1 \tag{1}$$

$$r_{t+1} \approx k + \rho p_{t+1} + (1 - \rho)d_{t+1} - p_t.$$
 (2)

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(b) Show that (2) implies (ignoring constants)

$$p_t \approx \mathsf{E}_t \sum_{j=0}^{\infty} \rho^j \left[(1-\rho) d_{t+1+j} - r_{t+1+j} \right]$$

$$p_t - d_t \approx E_t \left[\sum_{j=0}^{\infty} \rho^j \left(\Delta d_{t+1+j} - r_{t+1+j} \right) \right]$$

The approximate identity $r_{t+1} \approx k + \rho p_{t+1} + (1-\rho)d_{t+1} - p_t. \tag{2}$ Show that (2) implies (ignoring constants) $p_t \approx \mathsf{E}_t \sum_{j=0}^\infty \rho^j \left[(1-\rho)d_{t+1+j} - r_{t+1+j} \right]$ $p_t - d_t \approx E_t \left[\sum_{j=0}^\infty \rho^j \left(\Delta d_{t+1+j} - r_{t+1+j} \right) \right].$ $r_{2,t})' \text{ and } x_{1,t}, x_{2,t} \sim I(1). \ x_{1,t} \text{ and } x_{2,t} \sim t_{1,t} \text{ and } x_{2,t} \sim t_{2,t}$ that $z_t = \alpha' \mathbf{X}_t \sim I(0)$. $\mathsf{M}(1)$ 3. Let $\mathbf{X}_t = (x_{1,t}, x_{2,t})'$ and $x_{1,t}, x_{2,t} \sim I(1)$. $x_{1,t}$ and $x_{2,t}$ are cointegrated with cointegration vector $\boldsymbol{\alpha} = (\alpha_1, \alpha_2)'$, so that $z_t = \boldsymbol{\alpha}' \mathbf{X}_t \sim I(0)$. Consider the VECM(1) $\Delta x_{1,t} = \gamma_1 z_{t-1} + \phi_{11} \Delta x_{1,t-1} + \phi_{12} \Delta x_{2,t-1} + \epsilon_{1,t}$ $\Delta x_{2,t} = \gamma_2 z_{t-1} + \phi_{21} \Delta x_{2,t-1} + \epsilon_{1,t}$

$$\Delta x_{1,t} = \gamma_1 z_{t-1} + \phi_{11} \Delta x_{1,t-1} + \phi_{12} \Delta x_{2,t-1} + \epsilon_{1,t}$$

$$\Delta x_{2,t} = \gamma_2 z_{t-1} + \phi_{21} \Delta x_{1,t-1} + \phi_{22} \Delta x_{2,t-1} + \epsilon_{2,t}$$

$$\Leftrightarrow \Delta \mathbf{X}_t = \boldsymbol{\gamma} \boldsymbol{\alpha}' \mathbf{X}_{t-1} + \boldsymbol{\Phi}_1 \Delta \mathbf{X}_{t-1} + \boldsymbol{\epsilon}_t$$

$$\boldsymbol{\gamma} = \begin{pmatrix} \gamma_1 \\ \gamma_2 \end{pmatrix}, \ \boldsymbol{\Phi}_1 = \begin{pmatrix} \phi_{11} & \phi_{12} \\ \phi_{21} & \phi_{22} \end{pmatrix}.$$

$$1$$

$$\Psi(L)\mathbf{X}_t = \boldsymbol{\epsilon}_t$$

(a) Write the VECM(1) as a VAR(p): Solve for $\Psi(L)$ in ' AR? ankaj_kumar@ Solve for $\Psi(L)$ in terms of the VECM coefficients γ and Φ_1 . What is the lag order p of the VAR?

- (b) Now consider the special case $\alpha = (1, -1)'$. What restrictions of the VAR coefficients are $m{lpha}=egin{pmatrix} 1 \ -1 \end{pmatrix}, \ m{\Phi}_1=egin{pmatrix} 0.2 & -0.1 \ 0 & -0.25 \end{pmatrix}$ cases for $m{\gamma}$: implied by the VECM(1)?
- 4. Next, let's run some simulations. Fix

$$\boldsymbol{\alpha} = \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \quad \boldsymbol{\Phi}_1 = \begin{pmatrix} 0.2 & -0.1 \\ 0 & -0.25 \end{pmatrix}$$

for all simulations. Consider four cases for γ : pankaj_kumi

$$\gamma_1 = (0, 0.3)'$$
 $\gamma_2 = (0, 0.03)'$
 $\gamma_3 = (-0.25, 0.1)'$
 $\gamma_4 = (-0, 0)'$

1,2022,10:02:27 PN and $T_1 = 250$ and $T_2 = 2500$. For simplicity, let's analyze a single simulation for each combination of γ_i and T_j .

Note: Decide for yourself how you might run the simulations in order to answer the questions below (e.g. how many simulations to run in total, ...).

- (a) For each combination of γ_i and T_j , estimate a
 - i. VECM(1)

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- ii. VAR in \mathbf{X}_t with the optimal lag length according to the BIC

Which estimation specification(s) are most appropriate for the different cases? What are the tradeoffs of the different specifications?

- 2:27 PM P1 (b) Comment on the estimation results. Are the VAR restrictions that you derived in 3(b) satisfied? Discuss the implications for the cointegration mechanism implied by the different estimation methods. How do γ and T affect the results? (You do not have to report complete estimation outputs, just the "important" information.)
 - (c) Run long-horizon regressions to explore the forecastability of $x_{1,t}$ and $x_{2,t}$ that is a result of cointegration. Select the appropriate horizons for the long-horizon regressions.
 - (d) Study the serial correlation patterns of the errors of the long-horizon regressions.

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- du-May II (e) Compute different specifications of the variance-covariance matrix. Which variance-covariance matrix/matrices are most appropriate?
- 5. Download the file MFE230E_PS4_data.csv from bCourses. The file contains annual data for S&P500 ankai_kL prices and earnings.
 - (a) Test whether log prices and log earnings are stationary or not.

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- (b) Construct the log price-earnings ratio $p_t e_t$ and test whether it is stationary or not.
- Estimate a VECM with one lags and report the results. Describe and interpret the results. Which specification(s) of the variance-covariance of the estimated parameters are appropriate or inappropriate?
- (d) Compute long-horizon regressions for k = 1, 2, 3, 4, 5 for the regressions

$$\Delta p_{t+1} + \dots + \Delta p_{t+k} = \alpha_k + \beta_k (p_t - e_t) + u_{t+k,k}$$

$$\Delta e_{t+1} + \dots + \Delta e_{t+k} = \alpha_k + \beta_k (p_t - e_t) + u_{t+k,k}$$

- (e) Study the heteroskedasticity and serial correlation patterns of the errors of the long-horizon regressions.
- Compute the standard OLS standard errors as well as the White, Newey-West and Hansen-Hodrick standard error corrections. Which specification is most appropriate?
- pankaj_kumar@berkeley.edh Summarize your results. What conclusion can you draw about the behavior of returns and earnings?

pankaj_kumar@berkeley.edu - May 1, 2022.

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