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FINC 520, Spring 2007
Homework 2, due Monday, April 16.

1. In class, we discussed the $p^{t h}$ order VAR:

$$
y_{t}=c+\phi_{1} y_{t-1}+\phi_{2} y_{t-2}+\ldots+\phi_{p} y_{t-p}+\varepsilon_{t},
$$

where $\varepsilon_{t}$ is a white noise with variance-covariance matrix, $\Omega$, and orthogonal with $y_{t-s}, s>0$. Find a definition of $\xi_{t}$, so that the $\operatorname{VAR}(\mathrm{p})$ can be written in the following $\operatorname{VAR}(1)$ form:

$$
\xi_{t}=F \xi_{t-1}+v_{t}
$$

2. Write the $\operatorname{VAR}(\mathrm{p})$ in transposed form:

$$
y_{t}^{\prime}=x_{t}^{\prime} \Pi+\varepsilon_{t}^{\prime}
$$

where

$$
\underset{(n p+1) \times}{x_{t}}=\left(\begin{array}{c}
1 \\
y_{t-1} \\
\vdots \\
y_{t-p}
\end{array}\right), \underset{n \times(n p+1)}{\Pi^{\prime}}=\left[\begin{array}{cccc}
c & \phi_{1} & \cdots & \phi_{p}
\end{array}\right] .
$$

Suppose we have data over the period $t=1, \ldots, T$ and write:

$$
Y=\left(\begin{array}{c}
y_{1}^{\prime} \\
y_{2}^{\prime} \\
\vdots \\
y_{T}^{\prime}
\end{array}\right), X=\left(\begin{array}{c}
x_{1}^{\prime} \\
x_{2}^{\prime} \\
\vdots \\
x_{T}^{\prime}
\end{array}\right), \varepsilon=\left(\begin{array}{c}
\varepsilon_{1}^{\prime} \\
\varepsilon_{2}^{\prime} \\
\vdots \\
\varepsilon_{T}^{\prime}
\end{array}\right)
$$

so that the VAR can be written as follows:

$$
\underset{T \times n}{Y}=\underset{T \times(n p+1)(n p+1) \times n}{X}+\underset{T \times n}{\varepsilon} .
$$

Premultiply this by $X^{\prime}$ and then take expectations:

$$
E\left[X^{\prime} Y\right]=E\left[X^{\prime} X\right] \Pi+E\left[X^{\prime} \varepsilon\right]
$$

Note that

$$
E X^{\prime} \varepsilon=E \sum_{t=1}^{T} x_{t} \varepsilon_{t}^{\prime}=0
$$

because of the assumed orthogonality properties of $\varepsilon_{t}$. Then,

$$
\Pi=\left\{E\left[X^{\prime} X\right]\right\}^{-1} E\left[X^{\prime} Y\right]
$$

Ergodicity (we established in class that ergodicity is satisfied) suggests using the following estimator for $\Pi$ :

$$
\hat{\Pi}=\left(X^{\prime} X\right)^{-1} X^{\prime} Y
$$

(a) Compute $\hat{\Pi}$ using the data in the MATLAB $m$ file that has been provided, with $p=4$. Let

$$
y_{t}=\left(\begin{array}{c}
R_{t} \\
\log \frac{G D P_{t}}{G D P_{t-1}} \\
\pi_{t}
\end{array}\right)
$$

where $R_{t}$ denotes the 3 month Tbill rate, $\pi_{t}$ denotes the quarterly inflation rate and $G D P_{t}$ denotes Gross Domestic Product in quarter $t$.
(b) Construct $F$, the matrix you constructed for question 1. Note how your way of handling the constant term causes there to be one eigenvalue equal to unity this matrix. Compute $\mu=E y_{t}=$ $\left[I-\phi_{1}-\phi_{2}-\phi_{3}-\phi_{4}\right]^{-1} c$. Compare $\mu$ with:

$$
P\left[y_{t+500} \mid y_{t}, y_{t-1}, \ldots .\right]
$$

Why should $\mu$ and this projection be so similar? Is the $\operatorname{VAR}(\mathrm{p})$ model you constructed covariance stationary and ergodic? Explain.
(c) Construct a 'predicted time series' on the long rate for the $T=$ 201 quarters, 1955Q4-2005Q4. Graph it together with the actual 5 year rate, $R_{t}^{l}$. To do this, you'll have to do the calculations discussed in class. However, the geometric sum formula described there will not work if you do it with the version of $F$ in question

1, because that has a unit root in it. You can simply add $I+F+$ $\ldots .+F^{19}$. How does the expectations hypothesis do? Do the same graph, only this time do it with

$$
y_{t}=\left(\begin{array}{c}
R_{t} \\
R_{t}^{l} \\
\log \frac{G D P_{t}}{G D P_{t-1}} \\
\pi_{t}
\end{array}\right)
$$

Do the results make any difference with this adjustment?
(d) Graph the spectrum, $(0, \pi)$, for the long rate, $R_{t}^{l}$, and the long rate as predicted by the model. (Hint: for this, note that each of these long rates can be expressed as $\gamma^{\prime} y_{t}$, for a suitable choice of the column vector, $\gamma$. Then, the spectrum of $\gamma^{\prime} y_{t}$ is just $\gamma^{\prime} S_{y}\left(e^{-i \omega}\right) \gamma$.)
3. Calculate the variance, $\Gamma_{0}$, of $\operatorname{VAR}(\mathrm{p})$ process you estimated in question 2. Do so using the vectorization and the Riccatti equation methods discussed in class. Also compute $\Gamma_{0}$ using

$$
\Gamma_{0}=\int_{-\pi}^{\pi} S_{y}\left(e^{-i \omega}\right) d \omega
$$

using the Riemann approximation to the integral. How many approximation points do you have to use, in order for the integral approach to yield an accurate answer (say, to within 3 significant digits)? Do the same for

$$
\Gamma_{1}=\int_{-\pi}^{\pi} S_{y}\left(e^{-i \omega}\right) e^{i \omega} d \omega
$$

