## UNIVERSITY OF OSLO <br> DEPARTMENT OF ECONOMICS

Exam: ECON4160 - Econometrics - Modeling and systems estimation
Date of exam: Monday, December 2, $2013 \quad$ Grades are given: January 2, 2014
Time for exam: 2:00 p.m. - 5:00 p.m.
The problem set covers 6 pages (including cover page)
Resources allowed:

- All printed and written resources, in addition to calculator, are allowed

The grades given: A-F, with A as the best and E as the weakest passing grade. F is fail.

Exam in: ECON 4160: Econometrics: Modelling and Systems Estimation
Day of exam: 2 December 2013
Time of day: 14:00-17:00
This is a 3 hour school exam.

## Guidelines:

In the grading, question $A$ will count $1 / 3$, and question $B$ will count $2 / 3$.

## Question A (1/3)

1. Consider the regression model

$$
\begin{equation*}
Y_{t}=\phi_{0}+\beta X_{t}+\varepsilon_{t} \tag{1}
\end{equation*}
$$

where $Y_{t}$ and $X_{t}$ are random variables, and where $\varepsilon_{t}$ has the classical properties of a regression model disturbance. Explain briefly, what is meant by the following two exogeneity concepts:
(a) Weak exogeneity of $X_{t}$ with respect to the parameters of interest, $\phi_{0}$ and $\beta$.
(b) Super exogeneity of $X_{t}$.
2. Assume that the regression model is not (1), but

$$
\begin{equation*}
Y_{t}=\phi_{0}+\phi_{1} Y_{t-1}+\beta_{1} X_{t}+\varepsilon_{t} \tag{2}
\end{equation*}
$$

and that the parameters of interest are the characteristics root(s) that determine whether $Y_{t}$ is a weakly stationary (covariance stationary) time series variable or not. Is, in general, $X_{t}$ weakly exogenous in this case? Motivate your answer.
3. Consider an econometric equation for $Y_{t}\left\{Y_{t} ; t=1,2, \ldots, T\right\}$ with $k$ explanatory variables. The equation can be written in matrix notation as

$$
\begin{equation*}
\mathbf{y}=\mathbf{X} \boldsymbol{\beta}+\boldsymbol{\varepsilon} \tag{3}
\end{equation*}
$$

The vector $\mathbf{y}$ is $T \times 1$ and the matrix $\mathbf{X}$ is $T \times k$. $\varepsilon$ is the $T \times 1$ disturbance vector. $\boldsymbol{\beta}$ is the $k \times 1$ coefficient vector.
In each of the three cases below, give, without proofs, the expression for the consistent and asymptotically efficient method of moments estimator of $\boldsymbol{\beta}$. Assume $E(\boldsymbol{\varepsilon})=\mathbf{0}$ in all cases:
(a) $E\left(\varepsilon \varepsilon^{\prime}\right)=\sigma^{2} \mathbf{I}$, where $\mathbf{I}$ is the identity matrix of dimension $T \times T$, $\sigma^{2}>0$, and $\operatorname{plim}\left(\frac{1}{T} \mathbf{X}^{\prime} \boldsymbol{\varepsilon}\right)=\mathbf{0}$.
(b) $E\left(\varepsilon \varepsilon^{\prime}\right)=\sigma^{2} \mathbf{I}$, and $\operatorname{plim}\left(\frac{1}{T} \mathbf{X}^{\prime} \boldsymbol{\varepsilon}\right) \neq \mathbf{0}, \operatorname{plim}\left(\frac{1}{T} \mathbf{W}^{\prime} \boldsymbol{\varepsilon}\right)=\mathbf{0}, \operatorname{plim}\left(\frac{1}{T} \mathbf{W}^{\prime} \boldsymbol{X}\right)=$ $\boldsymbol{\Sigma}_{W X}$ (invertible) and $\operatorname{plim}\left(\frac{1}{T} \mathbf{W}^{\prime} \mathbf{W}\right)=\boldsymbol{\Sigma}_{W W}$ (positive definite) where the matrix $\mathbf{W}$ is $T \times k$.
(c) As in (a), but $E\left(\varepsilon \varepsilon^{\prime}\right)=\sigma^{2} \mathbf{I}$ is replaced by $E\left(\varepsilon \varepsilon^{\prime}\right)=\sigma^{2} \Omega$, where $\boldsymbol{\Omega}$ is a symmetric and positive definite $T \times T$ matrix.

## Question B (2/3)

In this exercise, we shall first consider the following open VAR, often called a VAR-X, with two exogenous variables, $Z_{1}$ and $Z_{2}$ :

$$
\begin{align*}
\binom{Y_{t}}{X_{t}} & =\binom{a_{10}}{a_{20}}+\left(\begin{array}{ll}
a_{11} & a_{12} \\
a_{21} & a_{22}
\end{array}\right)\binom{Y_{t-1}}{X_{t-1}} \\
& +\left(\begin{array}{ll}
\gamma_{11} & \gamma_{12} \\
\gamma_{21} & \gamma_{22}
\end{array}\right)\binom{Z_{1, t}}{Z_{2, t}}+\binom{\varepsilon_{1, t}}{\varepsilon_{2, t}} \tag{4}
\end{align*}
$$

$\varepsilon_{t}=\left(\varepsilon_{1 t}, \varepsilon_{2 t}\right)^{\prime}$ is bivariate normal with $E\left(\varepsilon_{t}\right)=\mathbf{0}$ and $\operatorname{Var}\left(\varepsilon_{t}\right)=\boldsymbol{\Omega}$.

1. It can be shown that (you are not supposed to show this) if we multiply on both sides of the equality sign by the matrix

$$
\boldsymbol{B}=\left(\begin{array}{cc}
1 & -b_{12} \\
-b_{21} & 1
\end{array}\right)
$$

(4) can be written as a simultaneous equation system in the two variables $Y_{t}$ and $X_{t}$ :

$$
\begin{align*}
Y_{t}-b_{12} X_{t} & =\pi_{10}+\pi_{11} Y_{t-1}+\pi_{12} X_{t-1}+\eta_{11} Z_{1, t}+\eta_{12} Z_{2, t}+\epsilon_{1 t}  \tag{5}\\
-b_{21} Y_{t}+X_{t} & =\pi_{20}+\pi_{21} Y_{t-1}+\pi_{22} X_{t-1}+\eta_{21} Z_{1, t}+\eta_{22} Z_{2, t}+\epsilon_{2 t} \tag{6}
\end{align*}
$$

where all the parameters of the SEM in (5) and (6) are functions of the parameters of the reduced form VAR system in (4). Discuss the identification of the two equations in the following two cases:
(a) $\eta_{11}=0$ and $\eta_{22}=0$
(b) $\pi_{12}=0, \pi_{21}=0$ and $\eta_{22}=0$
2. Which estimation method would you use to estimate the parameters of (5) and(6) in the cases where you have concluded that one or both of the equations are identified? Does it matter for your choice of estimation method whether the two structural disturbances $\epsilon_{1 t}$ and $\epsilon_{2 t}$ are contemporaneously uncorrelated or not?
3. For the remaining questions, we consider the case of a closed VAR (corresponding to setting $\gamma_{11}=\gamma_{12}=\gamma_{21}=\gamma_{22}=0$ in (4)). At the end of the question set, you find OLS estimation results for such a VAR.
Use the relevant results to test, first, the null hypothesis that $Y$ is not Granger-causing $X$, and, second, that $X$ is not Granger-causing $Y$.
4. Based on your answer in 3., explain why $X_{t}$ can be regarded as strongly exogenous in the ARDL model:

$$
\begin{equation*}
Y_{t}=\beta_{0}+\rho Y_{t-1}+\beta_{1} X_{t}+\beta_{2} X_{t-1}+\varepsilon_{t} \tag{7}
\end{equation*}
$$

5. Under Estimation results for question $B$ you also find results for the estimation of the ECM version of (7), together with a marginal model for $\Delta X_{t}$.
(a) The endogenous variables are different in the two models. The numbers of parameters are also different. Nevertheless, the "loglikelihood" is the same in the VAR as in the second model. Why is that?
(b) The VAR residuals are correlated contemporaneously, as the results show. Will the disturbances of the ECM equation and the marginal model for $\Delta X_{t}$, in the second model, also be contemporaneously correlated?
6. Assume that we get to know that $Y_{t}$ and $X_{t}$ are $I(1)$ variables. Under this assumption, and using the estimation results at the end of the question set, show that the null-hypothesis of no long-run relationship between $Y_{t}$ and $X_{t}$ can be rejected.

For reference: the $5 \%$ critical value of the relevant test statistic is -3.21 , and the $1 \%$ critical value is -3.79 .
7. What is the estimated long-run effect on $Y_{t}$ of a change in $X_{t}$ ? What additional results would you need to be able to construct a confidence interval for that parameter, and how would you proceed?

## Estimation results for question B.

## Results for question B3 and B4.

Estimation method is OLS
The estimation sample is: 3-201

| URF equation for: Y |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Coefficient | Std.Error | t-value | t-prob |
| Y_1 |  | 0.507680 | 0.07350 | 6.91 | 0.0000 |
| X_1 |  | 0.394077 | 0.05876 | 6.71 | 0.0000 |
| Constant | U | 0.962592 | 0.1687 | 5.71 | 0.0000 |
| sigma $=0.994659 \quad$ RSS $=193.9118976$ |  |  |  |  |  |
| URF equation for: X |  |  |  |  |  |
|  |  | Coefficient | Std.Error | t-value | t-prob |
| Y_1 |  | 0.0126066 | 0.07717 | 0.163 | 0.8704 |
| X_1 |  | 0.990516 | 0.06169 | 16.1 | 0.0000 |
| Constant | U | 1.03115 | 0.1771 | 5.82 | 0.0000 |
| sigma $=1.04433 \quad$ RSS $=213.7638469$ |  |  |  |  |  |
| log-likel |  | -526.222426 | -T/2log\|Om | mega\| | 38.5151101 |
| \|Omega| |  | 0.679032334 | $\log \left\|\mathrm{Y}^{\prime} \mathrm{Y} / \mathrm{T}\right\|$ |  | 8.24714056 |
| no. of ob | vations | ns 199 | no. of par | rameters | 6 |
| \| correlation of URF residuals (standard deviations on diagon |  |  |  |  |  |
| Y | 0.99466 | $66 \quad 0.592$ | 269 |  |  |
| X | 0.59269 | 691.0 | 443 |  |  |

Memo: Residual diagnostics (have been omitted to save space) do not give any indication of misspecification of the two equation's disturbances. They can be assumed to be white-noise.

## Results for question B5-B7.

Estimation method is OLS

| Equation for: DY |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Coefficient | Std.Error | t-value | t-prob |
| Y_1 |  | -0.499436 | 0.05935 | -8.41 | 0.0000 |
| DX |  | 0.564496 | 0.05493 | 10.3 | 0.0000 |
| X_1 |  | 0.399431 | 0.04745 | 8.42 | 0.0000 |
| Constant | U | 0.380511 | 0.1475 | 2.58 | 0.0106 |
| sigma $=0.80113$ |  |  |  |  |  |
| Equation for: DX |  |  |  |  |  |
|  |  | Coefficient | Std.Error | t-value | t-prob |
| Y_1 |  | 0.0126066 | 0.07717 | 0.163 | 0.8704 |
| X_1 |  | -0.00948438 | 0.06169 | -0.154 | 0.8780 |
| Constant | U | 1.03115 | 0.1771 | 5.82 | 0.0000 |
| sigma $=1.04433$ |  |  |  |  |  |
| log-likelihood |  | -526.222426 | -T/2log\|Om | mega\| | 38.5151101 |
|  |  | ons 199 | no. of par | rameters | 7 |

