

***UNIVERSITY OF OSLO***  
***DEPARTMENT OF ECONOMICS***

Exam: **ECON4160 – Econometrics – Modeling and systems estimation**

Date of exam: Wednesday, November 25, 2015      **Grades are given: December 15, 2015**

Time for exam: 09.00 a.m. – 12.00 noon

The problem set covers 8 pages (incl. cover sheet)

Resources allowed:

- Open book examination where 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:** 25 November 2015

**Time of day:** 09:00—12:00

This is a 3 hour school exam.

**Guidelines:**

In the grading, question A will count 30%, and question B will count 70%.

## Question A (30 %)

1. Assume that the conditional expectation of  $Y$  grows linearly with  $X$ . Consider the  $n$  variable pairs  $(Y_1, X_1), \dots, (Y_n, X_n)$ , and assume that the variable pairs are mutually independent and have identical normal distributions. In this case, what are the expressions for the maximum likelihood estimators of the two parameters

$$\frac{\partial}{\partial X_i} E(Y_i | X_i) \text{ and } Var(\varepsilon_i | X_i) = \sigma^2 ?$$

2. How can you estimate the parameter  $\frac{\partial}{\partial X_i} E(Y_i | X_i)$  efficiently, if the assumption  $Var(\varepsilon_i | X_i) = \sigma^2$  is changed to  $Var(\varepsilon_i | X_i) = \sigma^2 X_i^2$  ?
3. Assume that the relationship

$$(1) \quad Y_i = \beta_1 + \beta_2 X_i + \varepsilon_{1i},$$

is an equation in a model consisting of two equations. Discuss identification and estimation of the parameter  $\beta_2$  in the following three cases: (We denote parameters in the second equation by  $\gamma_j$  in all three cases, and that the only unobservable variables are disturbances.)

- (a) The second equation is

$$(2) \quad Z_i = \gamma_0 + \gamma_1 X_i + \varepsilon_{2i}$$

and we assume  $Cov(\varepsilon_{1i}, \varepsilon_{2i}) = \omega_{12} \neq 0$ , and that  $X_i$  is uncorrelated with both disturbances.

(b) The second equation is

$$(3) \quad X_i = \gamma_0 + \varepsilon_{2i}$$

and we assume  $\omega_{12} = 0$ .

(c) The second equation is

$$(4) \quad X_i = \gamma_0 + \gamma_1 Y_i + \gamma_2 Z_{1i} + \gamma_3 Z_{2i} + \varepsilon_{2i}$$

and we assume  $\omega_{12} \neq 0$ .

4. In case 3(b) and 3(c), is the second equation of the model identified?

## Question B (70 %)

We have collected annual data for hourly wages in Norwegian manufacturing for the period 1970 to 2013. We also have data for the value of labour productivity in this sector. In the print-out from PcGive in Table 1, we denote the logarithms of these two variables as LW (wages) and LZ (value of labour productivity) respectively.

1. Explain why the evidence in Table 1 gives reason to conclude that both LW and LZ are integrated of order one,  $I(1)$ . (You can consider the degree of augmentation as a given thing).

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Unit-root tests
The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7
The sample is: 1972 - 2013 (44 observations and 2 variables)

LW: ADF tests (T=42, Constant+Trend; 5%=-3.52 1%=-4.19)
D-lag   t-adf      beta Y_1   sigma   t-DY_lag  t-prob
  1     -2.811    0.94888   0.01856   3.623   0.0008
  0     -4.145*    0.92148   0.02125

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LZ: ADF tests (T=42, Constant+Trend; 5%=-3.52 1%=-4.19)
D-lag   t-adf      beta Y_1   sigma   t-DY_lag  t-prob
  1     -1.887    0.93151   0.03735   -0.3623  0.7192
  0     -1.915    0.93131   0.03693

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Unit-root tests
The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7
The sample is: 1973 - 2013 (43 observations and 2 variables)

DLW: ADF tests (T=41, Constant; 5%=-2.93 1%=-3.60)
D-lag   t-adf      beta Y_1   sigma   t-DY_lag  t-prob
  1     -3.079**   0.55138   0.02025   1.143   0.2603
  0     -2.864**   0.62787   0.02033

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DLZ: ADF tests (T=41, Constant; 5%=-2.93 1%=-3.60)
D-lag   t-adf      beta Y_1   sigma   t-DY_lag  t-prob
  1     -4.655**   -0.11150  0.03953   0.2770  0.7833
  0     -6.553**   -0.063503 0.03905

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Table 1: Augmented Dickey-Fuller (ADF) tests to determine the order of integration of  $LW_t$  and  $LZ_t$ .

2. According to theory, the system of collective wage bargaining in Norway creates a long-run dependency between the hourly wage and the manufacturing firms' ability to pay, as measured by the value of labour productivity.

To test this theory, we estimate the following Engle-Granger regression by OLS:

$$(5) \quad LW_t = \beta_0 + \beta_1 LZ_t + u_t, \quad t = 1970, \dots, 2013$$

where  $u_t$  is the disturbance. In the print-out in Table 2, the residual from the Engle-Granger regression has been labeled EGLWresiduals.

Use the results in Table 2 to form a conclusion about whether the theory of a long-run relationship between LW and LZ is supported or not. The critical values of the Engle-Granger test of no long-run (cointegrating) relationship are: 5% =  $-3.33$ , and 1% =  $-3.90$ . As part of your answer, explain why you use these critical values, instead of the critical values for the ADF tests given in Table 1.

EQ( 1) Modelling LW by OLS  
 The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7  
 The estimation sample is: 1970 - 2013

	Coefficient	Std.Error	t-value	t-prob	Part.R^2
Constant	-0.172383	0.04508	-3.82	0.0004	0.2582
LZ	0.963630	0.008637	112.	0.0000	0.9966
sigma	0.0504044	RSS		0.106705506	
R^2	0.996637	F(1,42) =		1.245e+004	[0.000]**

Unit-root tests

The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7  
 The sample is: 1973 - 2013 (43 observations and 1 variables)

EGLWresiduals: ADF tests (T=41; 5%=-1.95 1%=-2.62)

D-lag	t-adf	beta	Y_1	sigma	t-DY_lag	t-prob	AIC	F-prob
1	-2.853**	0.61462	0.03731		0.5051	0.6163	-6.529	
0	-2.935**	0.64303	0.03696				-6.572	0.6163

Table 2: Results for an Engle-Granger regression between  $LW_t$  and  $LZ_t$ , and unit-root tests for the disturbance of that regression.

3. An alternative test for the null hypothesis of no long-run relationship can be based on a conditional equilibrium correction model (ECM). You find estimation results for such a model in Table 3. In that print-out, the variables DLW and DLZ are the differences of LW and LZ, for example  $DLW=LW-LW_{-1}$ , where  $LW_{-1}$  denotes the first lag of LW. The ECM of LW also includes two other conditioning variables: DLKPI, which is the inflation rate, and DLNH, which is the change in the length of the normal working day. It is relevant to condition on these two variables because compensation for increases in the cost of living, and for shorter-hours is part of the bargaining between the unions and the firms. We base our analysis on the assumption that DLKPI and DLNH are  $I(0)$  variables.
  - (a) Based on the results in Table 3, and the information that the relevant 1% critical value for the ECM-test of no relationship is  $-3.29$ , explain why it is reasonable to conclude that LW is cointegrated with LZ.
  - (b) Can you give some intuition on why the evidence in support of

cointegration may be stronger when we use the ECM-test than when the Engle-Granger test is used?

- (c) Use the results to find the estimated long-run elasticity of the wage level (W) with respect to the value of labour productivity (Z). The estimated standard-error of the long-run elasticity can be shown to be 0.032. Is a long-run elasticity of 1 supported empirically if you use a significance level of 5%?
- (d) How could the standard-error be calculated? Explain in words.

EQ(2) Modelling DLW by OLS  
 The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7  
 The estimation sample is: 1972 - 2013

	Coefficient	Std.Error	t-value	t-prob	Part.R^2
DLW_1	0.247629	0.09799	2.53	0.0163	0.1581
Constant	-0.00492564	0.03616	-0.136	0.8925	0.0005
DLZ	0.100648	0.05653	1.78	0.0839	0.0853
DLZ_1	0.00917294	0.05715	0.160	0.8734	0.0008
LW_1	-0.250837	0.04727	-5.31	0.0000	0.4530
LZ_1	0.237388	0.04547	5.22	0.0000	0.4449
DLKPI	0.567936	0.1122	5.06	0.0000	0.4297
DLNH	-0.685046	0.2384	-2.87	0.0070	0.1953
sigma	0.0124393	RSS		0.00526103907	
R^2	0.911595	F(7,34) =	50.08	[0.000]**	
Adj.R^2	0.893394	log-likelihood		129.092	
no. of observations	42	no. of parameters		8	
mean(DLW)	0.0691441	se(DLW)		0.0380983	
AR 1-2 test:	F(2,32) =	0.55526	[0.5792]		
ARCH 1-1 test:	F(1,40) =	1.9135	[0.1743]		
Normality test:	Chi^2(2) =	0.33955	[0.8439]		
Hetero test:	F(14,27) =	1.3089	[0.2652]		
RESET23 test:	F(2,32) =	3.0802	[0.0598]		

Table 3: Results for ECM of  $LW_t$

4. As a way of imposing a long-run elasticity of one, we define the variable

$$ECMwage = LW - LZ,$$

which we assume to be  $I(0)$  from now on, and re-estimate the ECM for wages. Table 4 shows the results. Table 5 shows the result for a marginal model of DLZ.

- (a) Explain how you can test the weak exogeneity of LZ with respect to the cointegration parameters with the aid of the information given. What does the evidence indicate?
- (b) Assume that you are asked by the Norwegian Productivity Commission to estimate the dynamic effects on wages of a shock to average labour productivity. Could you use the results reported in Table 4 and 5 to give an answer? Explain how you would motivate your answer.
- (c) A colleague suggests that to appropriately address question 4 (b), a SVAR model needs to be considered. Do you think using a SVAR is suitable to deal with this question?

EQ( 3) Modelling DLW by OLS

The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7

The estimation sample is: 1972 - 2013

	Coefficient	Std.Error	t-value	t-prob	Part.R^2
DLW_1	0.333612	0.09621	3.47	0.0014	0.2557
Constant	-0.0768366	0.01971	-3.90	0.0004	0.3027
DLZ	0.134225	0.05796	2.32	0.0265	0.1329
DLZ_1	0.0570952	0.05652	1.01	0.3193	0.0283
DLKPI	0.681517	0.1071	6.36	0.0000	0.5365
DLNH	-0.582600	0.2485	-2.34	0.0249	0.1357
ECMwage_1	-0.210894	0.04668	-4.52	0.0001	0.3683
sigma	0.0131938	RSS		0.00609264708	
R^2	0.897621	F(6,35) =	51.14	[0.000]**	
Adj.R^2	0.88007	log-likelihood		126.01	
no. of observations	42	no. of parameters		7	
mean(DLW)	0.0691441	se(DLW)		0.0380983	
AR 1-2 test:	F(2,33) =	2.7736	[0.0770]		
ARCH 1-1 test:	F(1,40) =	0.00067955	[0.9793]		
Normality test:	Chi^2(2) =	1.9124	[0.3844]		
Hetero test:	F(12,29) =	0.84777	[0.6042]		
Hetero-X test:	F(27,14) =	1.4910	[0.2185]		
RESET23 test:	F(2,33) =	3.5640	[0.0397]*		

Table 4: Results for an ECM of  $LW_t$  with cointegration imposed in the form of the variable ECMwage.

EQ( 4) Modelling DLZ by OLS

The dataset is: C:\SW20\ECON4160\H2015\Exam\LoennIndogFastland.in7

The estimation sample is: 1972 - 2013

	Coefficient	Std.Error	t-value	t-prob	Part.R^2
DLZ_1	0.105475	0.1616	0.653	0.5180	0.0117
Constant	0.130999	0.05231	2.50	0.0169	0.1484
DLW_1	0.558570	0.2605	2.14	0.0389	0.1132
DLKPI	-0.144616	0.3070	-0.471	0.6404	0.0061
DLNH	0.0181977	0.7147	0.0255	0.9798	0.0000
ECMwage_1	0.284411	0.1256	2.26	0.0297	0.1247
sigma	0.037941	RSS		0.0518226244	
R^2	0.452038	F(5,36) =	5.94	[0.000]**	
Adj.R^2	0.375932	log-likelihood		81.0541	
no. of observations	42	no. of parameters		6	
mean(DLZ)	0.0693511	se(DLZ)		0.0480278	
AR 1-2 test:	F(2,34) =	0.18983	[0.8280]		
ARCH 1-1 test:	F(1,40) =	0.44294	[0.5095]		
Normality test:	Chi^2(2) =	0.31240	[0.8554]		
Hetero test:	F(10,31) =	0.60697	[0.7958]		
Hetero-X test:	F(20,21) =	1.6381	[0.1349]		
RESET23 test:	F(2,34) =	4.8045	[0.0145]*		

Table 5: Results for a marginal model of  $DLZ_t$ .