

**Seminar 31-32****Multivariate Time Series models****Problem 1**

What is “spurious regression”?

Using STATA simulate two independent RW-processes  $x$  and  $y$ . Then try to estimate a regression of  $y$  on  $x$ . Comment your results after several attempts.

**Problem 2** (*Exercise 9.3 from Verbeek M. A guide to modern Econometrics, 2004*)

In the file `mts_income.dta` you find quarterly data on UK nominal consumption and income, for 1971:1 to 1985:2 ( $T = 58$ ).

- Test for a unit root in the income and consumption series using several augmented Dickey–Fuller tests.
- Investigate the correlation of income and consumption.
- Perform a regression by OLS explaining consumption from income. Test for cointegration using two different tests.
- Perform a regression by OLS explaining income from consumption. Test for cointegration.
- Compare the estimation results and  $R^2$  from the last two regressions.
- Discuss an error-correction (ECM) model. Interpret the components of these model (what are the long-term and short-term effects).
- Estimate an error-correction model for the change in consumption. Pay attention to the adjustment parameter. Test whether the adjustment coefficient is zero.
- Repeat the last question for the change in income. What do you conclude?
- Discuss a vector autoregressive (VAR) model using these two time series and a VECM representation for this model.

**Table 9.2** Asymptotic critical values residual unit root tests for cointegration (with constant term) (Davidson and MacKinnon, 1993)

Number of variables (incl. $Y_t$ )	Significance level		
	1%	5%	10%
2	−3.90	−3.34	−3.04
3	−4.29	−3.74	−3.45
4	−4.64	−4.10	−3.81
5	−4.96	−4.42	−4.13

**Table 9.3** 5% Critical values CRDW tests for co-integration (Banerjee *et al.*, 1993)

Number of variables (incl. $Y_t$ )	Number of observations		
	50	100	200
2	0.72	0.38	0.20
3	0.89	0.48	0.25
4	1.05	0.58	0.30
5	1.19	0.68	0.35

**Problem 3**

- Download the data file balance2.dta.
- Investigate  $y$ ,  $i$  and  $c$  for stationarity:
  - $y$  is  $\ln(\text{GDP})$
  - $i$  is  $\ln(\text{income})$
  - $c$  is  $\ln(\text{consumption})$
- Investigate these time series for cointegration using Durbin-Watson test.
- Discuss the Johansen test. Use it to test for cointegration and number of cointegrating relations.
- Estimate a vector error-correction (VECM) model, interpret the results and provide tests for specification of the model.

**Problem 4** (9.6 Illustration “Money Demand and Inflation” from Verbeek M. *A guide to modern Econometrics*, 2004)

Read attentively this chapter and try to repeat calculations and interpret the results.

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**VECM is STATA**

$$\Delta y_t = \alpha(\beta y_{t-1} + \mu + \rho t) + \sum_{i=1}^{p-1} \Gamma_i \Delta y_{t-i} + \gamma + \tau t + \epsilon_t$$

Five different trend specifications are available:

Option in <code>trend()</code>	Parameter restrictions	Johansen (1995) notation
<code>trend</code>	none	$H(r)$
<code>rtrend</code>	$\tau = 0$	$H^*(r)$
<code>constant</code>	$\rho = 0$ , and $\tau = 0$	$H_1(r)$
<code>rconstant</code>	$\rho = 0$ , $\gamma = 0$ and $\tau = 0$	$H_1^*(r)$
<code>none</code>	$\mu = 0$ , $\rho = 0$ , $\gamma = 0$ , and $\tau = 0$	$H_2(r)$

**Johansen procedure**

$$\Delta \vec{Y}_t = \delta + \Gamma_1 \Delta \vec{Y}_{t-1} + \dots + \Gamma_{p-1} \Delta \vec{Y}_{t-p+1} + \Pi \vec{Y}_{t-1} + \vec{\varepsilon}_t, \quad \Pi = \gamma \beta'$$

Where  $\beta$  is a matrix of cointegration vectors and  $\gamma$  represents the matrix of weights with which each cointegrating vector enters each of the  $\Delta Y_t$  equations.

$r_k(\Pi) = 0$  if  $\Pi=0$ , i.e. all elements of  $Y_t$  are  $I(1)$  processes without cointegration

$r_k(\Pi) = (\max \text{rank if } \Pi) \text{ is invertible, i.e. all elements of } Y_t \text{ are } I(0) \text{ processes}$

the hypothesis  $H_0: r \leq r_0$  versus the alternative  $H_1: r_0 < r \leq k$ , can be tested using the statistic

$$\lambda_{trace}(r_0) = -T \sum_{j=r_0+1}^k \log(1 - \hat{\lambda}_j). \quad (9.52)$$

This test is the so-called **trace test**. It checks whether the smallest  $k - r_0$  eigenvalues are significantly different from zero. Furthermore, we can test  $H_0: r \leq r_0$  versus the more restrictive alternative  $H_1: r = r_0 + 1$  using

$$\lambda_{max}(r_0) = -T \log(1 - \hat{\lambda}_{r_0+1}). \quad (9.53)$$

This alternative test is called the **maximum eigenvalue test**, as it is based on the estimated  $(r_0 + 1)$ th largest eigenvalue.

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From Enders "Applied Econometric Time Series" (pp. 132-133)

**Dolado, Jenkinson and Sosvilla-Rivero** (1990) suggested the following procedure to test for a unit root when the form of the data-generating process is unknown. The following is a straightforward modification of their method:

**STEP 1:** As shown in Figure 4.7, start with the least restrictive of the plausible models (which will generally include a trend and drift) and use the  $\tau_\tau$  statistic to test the null hypothesis  $\gamma = 0$ . Unit root tests have low power to reject the null hypothesis; hence, if the null hypothesis of a unit root is *rejected*, there is no need to proceed. Conclude that the  $\{y_t\}$  sequence does not contain a unit root.

**STEP 2:** If the null hypothesis is *not rejected*, it is necessary to determine whether too many deterministic regressors were included in Step 1 above.<sup>11</sup> Test for the significance of the trend term under the null of a unit root (e.g., use the  $\tau_{\beta\tau}$  statistic to test the significance of  $a_2$ ). You should try to gain additional confirmation for this result by testing the hypothesis  $a_2 = \gamma = 0$  using the  $\phi_3$  statistic. If the trend is not significant, proceed to Step 3. Otherwise,

if the trend is significant, retest for the presence of a unit root (i.e.,  $\gamma = 0$ ) using the standardized normal distribution. After all, if a trend is inappropriately included in the estimating equation, the limiting distribution of  $a_2$  is the standardized normal. If the null of a unit root is rejected, proceed no further; conclude that the  $\{y_t\}$  sequence does not contain a unit root. Otherwise, conclude that the  $\{y_t\}$  sequence contains a unit root.

**STEP 3:** Estimate (4.35) without the trend [i.e., estimate a model in the form of (4.13)]. Test for the presence of a unit root using the  $\tau_\mu$  statistic. If the null is rejected, conclude that the model does not contain a unit root. If the null hypothesis of a unit root is not rejected, test for the significance of the constant (e.g., use the  $\tau_{a_0}$  statistic to test the significance of  $a_0$  given  $\gamma = 0$ ). Additional confirmation of this result can be obtained by testing the hypothesis  $a_0 = \gamma = 0$  using the  $\phi_1$  statistic. If the drift is not significant, estimate an equation in the form of (4.12) and proceed to Step 4. If the drift is significant, test for the presence of a unit root using the standardized

Figure 4.7 A procedure to test for unit roots.

Estimate  $\Delta y_t = a_0 + \gamma y_{t-1} + a_2 t + \sum \beta_i \Delta y_{t-i} + \varepsilon_t$

