

Winter Term 2019/2020

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## Multivariate Time Series Analysis

### Solution Exercise Sheet 9

#### 1 Exercise 1: Granger Causality – Theory

Let  $z_t = (x_t, y_t)'$  be a stationary time series with two dimensions. Define the forecast error as the univariate series  $e_T(h) = y_{T+h} - y_T(h)$  with  $y_T(h) = \mathbb{E}(y_{T+h} | \Omega_T)$ . The information set  $\Omega_T$  contains all relevant variables available whereas  $\Omega_T^{\setminus x} = \Omega_T \setminus \{x_t\}_{t=0}^T$  omits the variable  $x$  entirely. (This setting is the univariate equivalent to definition 6.1 on Slide 6-4.)

a) Prove that  $\mathbb{E}(e_T(h) | \Omega_T^{\setminus x}) = 0$ .

*Solution:*

$$z_t = \begin{pmatrix} x_t \\ y_t \end{pmatrix}$$

$$\mathbb{E}(e_T(h) | \Omega_T^{\setminus x}) = \mathbb{E}(y_{T+h} - \mathbb{E}(y_{T+h} | \Omega_T) | \Omega_T^{\setminus x})$$

$$\stackrel{\text{LIE}}{=} \mathbb{E}(\mathbb{E}(y_{T+h} - y_{T+h} | \Omega_T) | \Omega_T^{\setminus x})$$

since  $\Omega_T^{\setminus x} \subseteq \Omega_T$

LIE = Law of Iterated Expectations

$$= 0$$

b) Prove that  $\text{Var}(e_t(h) | \Omega_T) \leq \text{Var}(e_t(h) | \Omega_T^{\setminus x})$

*Solution:*

2 Theorems necessary for the proof:

## 1 Conditional Jensen's Inequality

$g(\cdot) : \mathbb{R}^m \rightarrow \mathbb{R}$  is convex (like  $\chi^2$ ), then for any random vectors  $(y, x)$  for which  $\mathbb{E}(\|y\|) < \infty$  and  $\mathbb{E}(\|g(y)\|) < \infty$ ,  $g(\mathbb{E}(y|x)) \leq \mathbb{E}(g(y)|x)$ . It is the other way around for concave functions.

## 2 Conditioning Theorem

If  $\mathbb{E}(\|y\|) < \infty$ , then  $\mathbb{E}(g(x)y|x) = g(x) \cdot \mathbb{E}(y|x)$ . If in addition  $\mathbb{E}(\|g(x)y\|) < \infty$ , then  $\mathbb{E}(g(x)y) = \mathbb{E}(g(x)\mathbb{E}(y|x))$ .

Back to Granger:

$e_T(h) = y_{T+h} - y_T(h)$  is a scalar. We know that  $\mathbb{E}(e_T(h)|\Omega_T^{\setminus x}) = 0$ ,  $\mathbb{E}(e_T(h)|\Omega_T) = 0$  and  $\text{Var}(e_T(h)) < \infty$  since  $y_t$  is a weakly stationary (w.s.) process. Furthermore, w.s. implies that  $\mathbb{E}(y_t) < \infty$ ,  $\mathbb{E}(y_t^2) < \infty$ .

From Jensen's Inequality it follows:

$$\begin{aligned} \left[ \mathbb{E}(y_{T+h}|\Omega_T^{\setminus x}) \right]^2 &\stackrel{\text{LIE}}{=} \left[ \mathbb{E} \left[ \mathbb{E}(y_{T+h}|\Omega_T) | \Omega_T^{\setminus x} \right] \right]^2 \\ &\leq \mathbb{E} \left[ \left[ \mathbb{E}(y_{T+h}|\Omega_T) \right]^2 | \Omega_T^{\setminus x} \right] \end{aligned}$$

Taking conditional expectations:

$$\mathbb{E} \left[ \left( \mathbb{E} \left[ y_{T+h} | \Omega_T^{\setminus x} \right] \right)^2 \right] \leq \mathbb{E} \left( \left[ \mathbb{E}(y_{T+h}|\Omega_T) \right]^2 \right)$$

This extends to:

$$\begin{aligned} \left[ \mathbb{E}(y_{T+h}) \right]^2 &\leq \mathbb{E} \left( \left[ \mathbb{E}(y_{T+h}|\Omega_T^{\setminus x}) \right]^2 \right) \\ \text{Since } \mathbb{E}(y_{T+h}) &= \mathbb{E} \left[ \mathbb{E}(y_{T+h}|\Omega_T^{\setminus x}) \right] \end{aligned}$$