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# Multivariate Time Series Analysis Exercise Sheet 7

#### Exercise 1: The optimal forecast

a) Show that the stationary VAR(1) process  $z_t = \phi z_{t-1} + a_t$  with  $a_t$  a standard white noise has the following causal representation:

$$z_t = \sum_{i=0}^{\infty} \theta_i \, a_{t-i}.$$

b) Assume the linear forecasting model  $y_T(h) = \Psi y_T$  and show that  $\Psi = \phi^h$  minimises the MSE of  $y_T(h)$  given that  $y_t$  is a VAR(1) process.

## Exercise 2: Properties of forecast errors

a) Show that for a general VAR(p) process

$$z_{T+h} - z_T(h) = e_T(h) = \sum_{i=0}^{h-1} \theta_i a_{T+h-i},$$

where  $z_T(h)$  is assumed to be the optimal forecast.

Hint: (5.9)

- b) Assume that  $a_t \sim N(0, \Sigma_a)$ . Derive the distribution of  $e_T(h)$ .
- c) Prove that  $Cov(e_T(h)) \to \Gamma_0$  as  $h \to \infty$ .

#### Exercise 3: Forecast intervals

Derive the confidence ellipsoid for  $e_T(h)$  (see slide 5-14) from (5.9) based on your results in Exercise 2.

### Exercise 4: Delta Method

For this task, assume both  $y_t$  and  $x_t$  to be  $K \times 1$  vectors and  $x_t \stackrel{iid}{\sim} [\mu_x, \Sigma_x]$ .

- a) Let  $y_t = f(x_t) = \phi_1 x_t$ . Compute the mean and variance of  $y_t$ .
- b) Derive the distribution of  $\sqrt{T}(\overline{y_T} E(y))$  from your results in a).
- c) Now let  $f(\cdot)$  be some function  $f(x): \mathbb{R}^k \to \mathbb{R}^k$ . Derive the first order Taylor expansion for f(x) at  $\mu_x$  and write it down in detail.

  Hint: You need the Jacobian matrix.
- d) Based on the expression obtained in c), show that a CLT applies for  $\sqrt{T}(f(\overline{x_T}) f(\mu_x))$ , and derive the distribution. Hint: Factor out deterministic parts. Since  $f(\cdot)$  is deterministic, f(c) is deterministic if c is.
- e) Lastly, assume the variable  $x_t$  to be known (meaning it is not stochastic). We want to predict  $y_t$  using  $y_t = \phi_1 x_t$ . Unfortunately, we only have  $\hat{\phi}_1$  which is stochastic with  $\sqrt{T} \left( \hat{\phi}_1 \phi_1 \right) \stackrel{d}{\to} N \left( 0, \Sigma_{\phi_1} \right)$ . Can we say something about the distribution of the prediction error  $\hat{y}_t y_t$ ?

This exercise sheet will be discussed in the tutorial on Wednesday, 4 December 2019