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Multivariate Time Series Analysis

Exercise Sheet 4

Exercise 1: Implied Models for Components

Consider the VAR(1) model $z_t = \phi_0 + \phi_1 z_{t-1} + a_t$ from the Exercise Sheet 3 again:

$$\phi_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad \phi_1 = \begin{pmatrix} 0.75 & 0 \\ -0.25 & 0.5 \end{pmatrix}, \quad \Sigma_a = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}.$$

- a) Write down the model using lag operator notation. Then rearrange the equation such that all parts based on z_t are on the left-hand side and the remainder is on the right-hand side.
- b) By factoring out z_t on the left, we obtain the lag polynomial $\phi(L)$. Compute its adjoint matrix by hand.

*Hint: Treat the lag operator as some scalar. The adjoint matrix can be computed like the inverse matrix but **without** the scaling by $\frac{1}{\det(\phi(L))}$.*

- c) Pre-multiply the model equation you got in part a) with the adjoint matrix you computed in part b).

Hint: You are supposed to end up with a diagonal matrix.

- d) The result of part c) should be a collection of two univariate ARMA(p,q) models. What is the lag order of both models?
- e) Simulate a trajectory with $T = 1000$ of the original VAR(1) model.

Hint: 'VARMAsim' on Slide 2-6

- f) Fit a VAR(1) model to the data, store the results as a variable and estimate the predictions' mean squared error for each variable in z_t .
- g) Repeat the task by fitting the two ARMA(p,q) models from b) to the data. Again compute the mean squared error for $z_{1,t}$ and $z_{2,t}$ each.

Hint: 'arima'

- h) Compare the MSEs of the VAR(1) estimates and the ARMA(p,q) estimates. Did the VAR(1) and the univariate ARMA(p,q) models perform similarly? If not, provide an intuition why.
- i) How can you manipulate the Σ_a matrix to equalise the MSEs of both the VAR(1) and the ARMA(p,q) models?

Exercise 2: Least Squares Estimation

- a) Again use your simulated time series from exercise 1. Regress $z_{1,t}$ on $z_{1,t-1}$ and $z_{2,t-1}$, then repeat with $z_{2,t}$ as dependent variable (meaning you estimate each row of the VAR(1) specification separately). How similar are the coefficients to those obtained from the VAR(1) regression?
- b) Show that you can generally estimate a VAR(p) by row-wise separate regressions using the derivations starting from Slide 3–3.

Hint: Make sure you understand how the trick in equation (3.3) works.

Exercise 3: Maximum Likelihood Estimation

- a) Let $\epsilon_1, \dots, \epsilon_T$ be an i.i.d. sample from a normal distribution with unknown mean μ and variance σ^2 . Find maximum likelihood estimators for μ and σ^2 .
- b) Prove equation (3.12) in the lecture slides.

This exercise sheet will be discussed in the tutorial on Wednesday, 13 November 2019