

Dr. Yannick Hoga Thilo Reinschlüssel

Multivariate Time Series Analysis

Exercise Sheet 6

Exercise 1: Model Selection – Review

- a) Is the MSE scale-invariant? (Does it matter if you use percentage points instead of decimals?)
- b) What is the fundamental trade-off which information criteria are supposed to balance?
- c) Does a linear transformation affect the value of the information criteria? Does it also influence the locations of the minima (from which we pick the optimal order)?
- d) Finally: Are OLS standard errors scale invariant?

Exercise 2: Simplification and Forecasting – Macroeconomic Data

Reconsider Exercise 2 from Exercise Sheet 5. Again, please import/load the dataset ‘us_macrodata.Rda’ into your workspace and compute the growth rates of the variables appearing non-stationary. There are still 5 variables—CPI, Real GDP, the unemployment rate, general private investment and the debt-to-GDP ratio. All series have been sampled quarterly and were seasonally adjusted before downloaded from *FRED*.

- a) Fit a VAR(p) model according to the Hannan-Quinn information criterion.
- b) Use the estimated coefficients and the associated standard errors to compute the t-statistics ($H_0 : \phi_{p,ij} = 0$ vs $H_1 : \phi_{p,ij} \neq 0$) for each coefficient separately. Then count how many coefficients are not significantly different from zero at the 5% level.
- c) Estimate the refined model using the command `refVAR` by setting a threshold corresponding to the 5% level from b).
 - i) How many variables have been set to 0?
 - ii) Does the number coincide with your count in task b)?
 - iii) What may be the reason for the two numbers differing? (*Hint: Slide 4-21*)
- d) Compare the values of all information criteria offered to you both for the ‘ordinary’ VAR and the refined VAR model. Which model is best? Is the recommendation unanimous?
- e) Proceed to compare the MSEs of the ‘ordinary’ VAR model and the refined model. Is the model picked by the information criteria again superior? Explain your results.
- f) Calculate the numbers of coefficients estimated both for the ‘ordinary’ model and the refined model. Then perform a Ljung-Box test on the residuals of both models.
 - i) Do the models absorb the dynamics in the data completely?
 - ii) Explain the massive differences of the two tests at $m = \{3, 4\}$.
Hint: Mind the option ‘adj’ using the command mq!

- g) Now estimate a VAR(1) model (with intercept). How does it compare to the VAR(4) model in terms of MSE?
- h) Lastly, compute the forecasts' MSEs (referred to as MSFE) for both models using the command `VARpred`. Please use a forecast horizon h of 10.
 - i) Does the model superior in f) still prevail at every h ?
 - ii) Explain what conceptual difference between MSE and MSFE drives the results in i).
 - iii) To which values do the forecasts converge to if $h \rightarrow \infty$?

Hint: VARpred objects offer the RMSE based on the forecasts.

Exercise 3: Simplification and Forecasting – Exchange Rates

Reconsider Exercise 3 from Exercise Sheet 5. Please reuse the file 'quandl_fx_download.R' from last time to download and prepare the data. (Feel free register at www.quandl.com to get an API key for unlimited downloads.)

- a) Fit a VAR(1) model to the data regardless of the information criteria.
- b) Now fit the refined model based on your VAR(1) setting the threshold to 1.96. How many coefficients have been set to zero?
- c) Compare the MSEs of the 'ordinary' model and the refined model.
- d) Thirdly, estimate a VAR(0) with intercept by regression. Compare its MSFE with the forecast errors of the 'ordinary' VAR(1) model regarding a forecast horizon $h = 10$.

Hint: Write down the loss function and find the FOC's trivial solution.

- e) How do your results in d) align with the insights you gained in exercise 2h)?
- f) What can you do to reproduce the findings from 2h) in this setting?

Hint: Degrees of freedom.

Exercise 4: Forecast Errors

Show that Equation (5.1) in the lecture slides implies that

$$\mathbb{E}[\hat{z}_{T,T+h}^{(i)} - z_{T+h}]^2 \geq \mathbb{E}[z_T^{(i)}(h) - z_{T+h}]^2,$$

where $\hat{z}_{T,T+h}^{(i)}$ and $z_T^{(i)}(h)$ denote the i -th components of the respective forecasts ($i = 1, \dots, K$) for the observation z_{T+h} . This means that the optimal *univariate* forecasts are simply the components of the optimal *multivariate* forecast $z_T(h)$.

This exercise sheet will be discussed in the tutorial on Wednesday, 27 November 2019