Automated Modeling and Forecasting Vector Autoregressive Processes

Vector Autoregressiv Procossos

VAR Modeling with AuFVAR Data Initial Analysis Model Settings Selection Structural Breaks Estimation and Forecasting Residual Analysis Empirical Automated Modeling and Forecasting Vector Autoregressive Processes

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#### VAR Model: Theoretical Basics

Standard Modeling Steps:

2 Lag order selection

6 Residual analysis

4 Forecasting

• VAR(p) model for k-variate process Y<sub>t</sub>:

 $Y_t = \nu + A_1 Y_{t-1} + \ldots + A_p Y_{t-p} + \varepsilon_t.$ 

8 Parameter estimation for selected model

Identification of relevant variables and data initial analysis

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Matthias Fischer Vector Autoregressive Processes VAR Modeling with AuFVAR Data Initial Analysis Model Settings Selection Structural Breaks Estimation and Forecasting Residual

### Agenda

#### Automated Modeling and Forecasting

Vector Autoregressive

Processes

1 Vector Autoregressive Processes

2 VAR Modeling with AuFVAR Data Initial Analysis Model Settings Selection Structural Breaks Estimation and Forecasting Residual Analysis

3 Empirical Example: Advertisement Spendings

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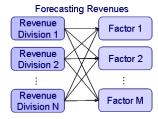
# Special Characteristics of AuFVAR

1 Incorporation of time trend and season as exogenous variables:

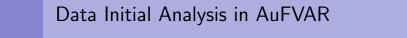
$$Y_t = \nu + \underbrace{n_{1t}\nu_1 + \ldots + n_{1(s-1)}\nu_{s-1}}_{Season \ Dummies} + \underbrace{\gamma t_t}_{Time \ Trend} + A(L)Y_t + \varepsilon_t.$$

2 Definition of a VAR model with *n* time series in each model

- S Automated model structure selection (settings und lag order) by means of the information criteria and backtesting
- Repeated forecasts for different VAR models with similar time series structure



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Initial Analysis

of Time Series

Granger Causality



Automated





1 Stationarity Test: ADF Test • Function: adf.test (modified) ② Granger Causality Test • Function: grangertest (standard)

#### 3 Log Level Transformation

Stationarity Test

Phase 1

Natural Logarithm of Time Series

Log-

Transformation

### Test for Structural Breaks

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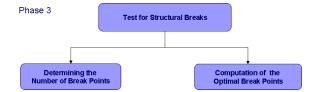
Automated





Structural Breaks





- Testing for the structural breaks: Function breakpoints, library strucchange
  - Determining the number of breakpoints
  - 2 Computation of the optimal breakpoints
  - Adjustment of the model definition

# Model Selection

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Processes

Model Settings Selection

Automated

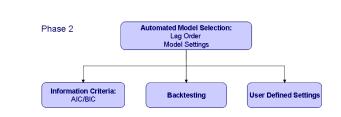
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Forecasting

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Forecasting



Definition the different VAR models (with/without trend/season, lag order variation) and selection the model with the least criterion value:

Information Criteria, e.g.

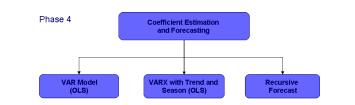
$$AIC(p) = \ln \det(\widetilde{\Sigma}_{\varepsilon}(p)) + rac{2}{T}pK^2,$$

2 Backtesting (out-of-sample):

$$RMSE = \sqrt{\frac{1}{h} \sum_{i=1}^{h} (y_{T+h} - y_t(h))^2}.$$

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# Estimation and Forecasting



- 1 Estimation of the appropriate model after testing for structural breaks
  - Function: estVARXIs for the models with intercept and lagged variables (standard)
  - Function: *estVARXIsM* for the models with additional time trend and/or season (modified)
- 2 Forecasting the time series, computation of confidence bounds and corresponding plots

### Residual Analysis

#### Automated Modeling and Forecasting Vector Autoregressive Processes

Processes VAR Modeling with AuFVAR Data Initial Analysis Model Setting Selection Structural Breaks Estimation ar Forecasting **Residual** Analysis



- 1 Portmanteau Test for Autocorrelation:
  - The standard test statistic, if time series length  $T \ge 100$
  - The modified test statistic otherwise
- Test for Nonnormality based on Skewness and Kurtosis (Lütkepohl, 1993)

#### Example: Time Series and Forecasting

#### Data

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Processes

Empirical Example

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- Variables to be forecasted:
  - Common Advertisement Earnings (with/without media)
- Further possible variables for the VAR model:
  - 1 ZEW Index (Centre for European Economic Research)
  - 2 Incoming Orders (Germany)
  - 3 CDAX (Composite DAX)

