

Lecture 1 What's macroeconometrics?

Tomasz Woźniak

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Tomasz Woźniak - short CV

Lived in.

Inowrocław, Poland Kraków, Poland Firenze, Italy Melbourne, Australia

Education.

Ph.D. in Economics at the European University Institute M.Res. in Economics at the European University Institute M.S. in IT and Econometrics at the Cracow University of Economics

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Research Interests.

Econometrics
Multivariate Time Series Analysis
Bayesian Inference

Economic Forecasting Causality Analysis

Topics I worked on.

Granger causality for time-varying volatility how risk associated with one financial asset transmits to another financial asset's risk

Granger causality for state-dependent variables how variables affect one another in an economy following business cycles

Heteroskedastic models for monetary policy

how to use volatility of variables to estimate the effect of monetary policy on the real economy more precisely

What's macroeconometrics?

Organization of the subject

Research project

Models we are working with

What's macroeconometrics?

Macroeconometrics focuses on developing methodology for empirical macroeconomic research.

Its main objectives include

- verification of economic theories
- forecasting future developments of essential variables
- providing analyses for data-driven decision-making at governing institutions and their stakeholders

It uses dedicated econometric models and procedures that determine the feasibility, robustness, and reliability of the applied research.

The characterization of econometric modeling includes

- System modeling of many variables in one model
- ► Identification of economically interpretable objects of interest
- Incorporation of economic theory assumptions into econometric model specifications
- ► Efficient extraction of information from the data implying e.g. modeling unit-root nonstationary variables

Macroeconometrics develops dedicated econometric models and procedures that are data, application, and objective-specific.

The development of these methods makes the empirical research possible and reliable.

Macroeconometrics as a field became highly technical and heavily computational.



This subject prepares the students to develop their own methods which requires:

- knowing the models and their properties
- deriving estimation procedures
- writing computer programs for these procedures and computation of interpretable values
- employing these programs for data analysis

Organisation of the subject

Contact Info

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Lectures

Lectures.

In-person active learning sessions are scheduled on:

Wednesdays 5:45 - 7:15 pm, FBE 221 (Theatre 4)

Thursdays 3:15 – 4:45 pm, FBE 211 (Theatre 2)

Attendance is monitored

Consultations.

Wednesdays 4:30–5:30 pm FBE 350

Assessment

| Week | Task | Grade |
|---------|---|------------|
| 4 | Test 1: Concepts and Tools | 10% |
| 5 | RP1: question, data, model, hypothesis | 10% |
| 6 | Test 2: Bayesian Estimation | 10% |
| 8 10 | RP2: estimation procedure and algorithm RP3: empirical analysis | 10% 10% |
| 4-10 | Learning repository contribution | 10% |
| 11 | RP Presentation | 10% |
| 12+ | RP Final report | 30% |

RP stands for Research Project

Learning outcomes

LO1: Develop original econometric methodology for applied macroeconomic analyses

LO2: Propose econometric techniques and models to verify hypotheses that inform fiscal or monetary policy

LO3: Derive Bayesian estimation procedure for the newly proposed macroeconometric model

LO4: Write computer programs in R that implement the derived estimation procedure

LO5: Apply the computer program in the forecasting or structural analyses of Australian macroeconomic data

LO6: Transparently create econometric data analysis using the newly proposed methodology in a fully reproducible report developed collaboratively

Generic skills

GS1: Obtain and format data from the original sources in an automated workflow

GS2: Document the essential data properties and incorporate them in the econometric modelling

GS3: Handle statistical distributions of parameters and forecasted values to make the econometric analysis feasible

GS4: Apply linear algebra operations and basic statistical theory to facilitate model estimation, hypothesis verification, and reliable forecasting

GS5: Create visualisations of data and estimation results that inform economic interpretations

Generic skills

GS6: Use functional programming to implement econometric procedures

GS7: Propose economic interpretations based on the empirical evidence

GS8: Obtaining, providing, and implementing constructive and actionable feedback

GS9: Managing a programming and data analysis project using git and GitHub

GS10: Communicating research outcomes in plain language and using visualisations

| Lecture | Topic | | |
|--------------------|--|--|--|
| Concepts and Tools | | | |
| 1 | What's macroeconometrics? | | |
| 2 | Maximum likelihood estimation | | |
| 3 | Bayesian estimation | | |
| 4 | Numerical optimization and integration | | |
| 5 | Understanding unit-rooters | | |
| 6 | Macroeconometric research themes | | |

| Lecture | Topic |
|---------|--------------------------------------|
| Macro | economic Forecasting with Fat Data |
| 7 | Vector Autoregressions |
| 8 | Large Bayesian VARs |
| 9 | Forecasting with Bayesian VARs |
| 10 | Forecasting with Large Bayesian VARs |

| Lecture | Topic |
|---------|--|
| Mo | deling Effects of Monetary Policy |
| 11 | Structural VAR models |
| 12 | Structural VAR tools |
| 13 | Bayesian estimation of Structural VARs |
| 14 | Modeling effects of monetary policy |

| Lecture | Topic | | |
|---|--|--|--|
| Modeling Trend Inflation | | | |
| 15 | Unobserved Component models | | |
| 16 | Bayesian estimation using simulation smoother | | |
| 17 | Modeling trend inflation | | |
| Modeling Conditional Heteroskedasticity | | | |
| 18 | Stochastic Volatility models | | |
| 19 | Bayesian estimation using auxiliary mixtures | | |
| Topics in Climate Change | | | |
| 20 | Forecasting CO ₂ Emissions for the 21st Century | | |

| Lecture | Topic | | | |
|--------------------------------|------------------------------------|--|--|--|
| Research Project Presentations | | | | |
| 21 | Student Presentations | | | |
| 22 | Student Presentations | | | |
| Le | Lecturer's Research Presentation | | | |
| 23 | Robust macroeceonometric modelling | | | |
| 24 | bsvars package presentation | | | |

Introduction to R

The objective of the complementary four sessions is to facilitate the beginning of working with R.

Session 1: Introduction to R

Session 2: Basic programing in R

Session 3: Numerical integration

Session 4: Numerical optimization

Session 5: Quarto documents

Session 6: Project development with git and GitHub

Session 7: Working with template repository on GitHub

| Week | Task | Grade |
|------|---|-------|
| 5 | RP1: question, data, model, hypothesis | 10% |
| 8 | RP2: estimation procedure and algorithm | 10% |
| 10 | RP3: empirical analysis | 10% |
| 11 | RP Presentation | 10% |
| 12+ | RP Final report | 30% |

The report includes:

- ▶ a proposal of an original model and a hypothesis to be verified
- derivation and coding of the Bayesian estimation procedure
- empirical investigation answering the proposed hypothesis

Submission format.

- ► a fully reproducible report generated with **Quarto**
- ► the report is a website hosted on **GitHub**
- the project is developed collaboratively using management tools git and GitHub

Each of the PR1-PR3 and the Presentation consist of:

| Task | Grade |
|--|-------|
| GitHub and Canvas submission | 8% |
| Providing feedback to your peer | 1% |
| Implementing the feedback in your report | 1% |

Vector Autoregressions.

$$y_t = A_1 y_{t-1} + \dots + A_p y_{t-p} + \mu_0 + \epsilon_t$$

 $\epsilon_t | Y_{t-1} \sim iid (\mathbf{0}_N, \Sigma)$

System modelling — all variables are endogenous

Dynamics — captures system dynamics of the variables

Forecasting — a go to model for predictive applications

Extensions capturing important data features improve forecasting precision

Structural Vector Autoregressions.

$$y_t = A_1 y_{t-1} + \dots + A_p y_{t-p} + \mu_0 + \epsilon_t$$
 $B\epsilon_t = u_t$ $u_t | Y_{t-1} \sim iid(\mathbf{0}_N, I_N)$

Structural relationships are explicitly modelled

Economic theory informs identification of structural shocks

Dynamic causal effects can be estimated and interpeted

Policy decision-making is based on evidence provided by SVARs

Unobserved Component Models.

$$\begin{aligned} y_t &= \tau_t + \epsilon_t \\ \tau_t &= \mu + \tau_{t-1} + \eta_t \\ \epsilon_t &= \alpha_1 \epsilon_{t-1} + \dots + \alpha_p \epsilon_{t-p} + e_t \\ \eta_t | Y_{t-1} &\sim \textit{iid} \mathcal{N} \left(0, \sigma_\eta^2 \right) \\ e_t | Y_{t-1} &\sim \textit{iid} \mathcal{N} \left(0, \sigma_e^2 \right) \end{aligned}$$

Trend and cycle decomposition of a variable

Long-run trend is highly-persistent

Oscillating cycle captures short-term dynamics

Inflation trend analysis and output gap estimation are the main applications

Macroeconometrics means cooperation!