

Advanced Time Series Methods

Course Overview and Teaching Plan

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Why this course?

- Many economic and financial variables are observed sequentially: GDP, inflation, interest rates, returns, exchange rates, volatility, and risk measures.
- The objective is not only description, but also **dynamic interpretation, forecasting, policy analysis, and risk measurement**.
- Time ordering creates dependence, and dependence changes both estimation and inference.

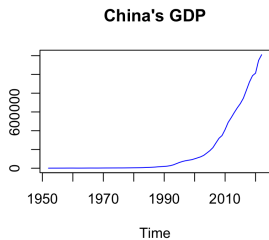
Big picture

This course studies how to model, estimate, interpret, and forecast dependent data in a rigorous econometric framework.

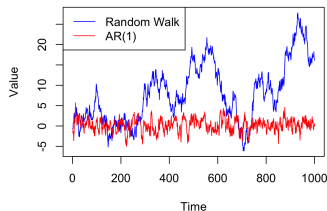
What kinds of questions will we study?

- Is a series stable, drifting, or driven by stochastic trends?
- How persistent are shocks, and how quickly do they die out?
- When do variables move together in the long run?
- How should we model time-varying volatility and risk?
- How do we do reliable inference when data are dependent?
- How do filtering and state-space methods extract latent signals?

Two visual motivations



Real macroeconomic series often contain trend, persistence, and structural change.



Stable dynamics and shock-accumulating dynamics can look very different and require different tools.

Course Roadmap

- 1 Course structure and logistics
- 2 Chapter allocation
- 3 Lecture map
- 4 How we will work in the course
- 5 Closing

Core textbook and scope

- We use the course textbook as the main reference.
- The core material in this lecture series covers **Chapters 2–9**.
- **Chapter 10 (Machine Learning)** is treated as an optional advanced topic rather than a core block.
- R applications are integrated across the course rather than taught as a separate module.

Core emphasis

Univariate foundations → multivariate systems → volatility → nonparametrics → robust inference → filtering → nonstationarity → continuous-time finance.

Course structure in hours

- **Total:** 60 contact hours.
- **Core teaching:** 54 hours.
- **Q&A / tutorial / assignment discussion:** 3 hours.
- **Final exam:** 3 hours.

Class meeting format

Each meeting is 3 contact hours, and each contact hour is 45 minutes.

What you should be able to do by the end

By the end of the course, students should be able to:

- 1 model univariate and multivariate dependence structures;
- 2 distinguish stationary, nonstationary, and cointegrated dynamics;
- 3 analyze volatility and time-varying uncertainty;
- 4 use robust inference tools for dependent data;
- 5 understand filtering, state-space, and continuous-time perspectives;
- 6 connect theory, computation, and empirical interpretation using R.

How the course is organized

Foundations and univariate models → multivariate systems → volatility and nonparametrics → robust inference and filtering → nonstationarity and continuous-time topics

- We begin with the language of stationary dependence and ARMA modeling.
- We then expand to VAR, VECM, SVAR, volatility, filtering, and continuous-time intuition.

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Chapter allocation in the 54 teaching hours

Chapter	Theme	Hours
Chapter 2	Univariate time series	8
Chapter 3	Multivariate linear time series	10
Chapter 4	Volatility models	6
Chapter 5	Nonparametric methods	4
Chapter 6	HAR inference / robust inference	9
Chapter 7	Filtering	5
Chapter 8	Nonstationary processes	7
Chapter 9	Continuous-time finance	5

What this allocation means in practice

- **Most time** goes to the core econometric backbone: univariate models, multivariate dynamics, and robust inference under dependence.
- **Significant time** is also reserved for nonstationary processes, because trends, unit roots, and long-run relations are central in applied work.
- **Specialized but important blocks** include volatility, filtering, and continuous-time finance.

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20-lecture map (Lectures 1–10)

Lec.	Main topic	Core chapter
1	Course map; dependence, stationarity, ergodicity, mixing, and Wold decomposition	Ch. 2
2	$AR(p)$, $MA(q)$, $ARMA(p, q)$; lag polynomials; roots; causality; invertibility	Ch. 2
3	ACF, PACF, model identification, estimation, diagnostics, and forecasting	Ch. 2
4	VAR and VMA representations; covariance-stationary multivariate models	Ch. 3
5	VAR estimation, stability, forecasting, and interpretation	Ch. 3
6	Impulse responses, FEVD, and structural interpretation	Ch. 3
7	Cointegration, VECM, and long-run multivariate dynamics	Ch. 3 / 8
8	Volatility basics: ARCH, GARCH, and conditional variance dynamics	Ch. 4
9	Asymmetric / multivariate volatility and forecasting risk	Ch. 4
10	Nonparametric methods for time series	Ch. 5

20-lecture map (Lectures 11–20)

Lec.	Main topic	Core chapter
11	HAC, long-run variance estimation, and robust inference under dependence	Ch. 6
12	Bootstrap and advanced inference topics for time-series data	Ch. 6
13	Further robust inference applications and empirical implementation	Ch. 6
14	Linear filters and signal extraction	Ch. 7
15	State-space models and the Kalman filter	Ch. 7
16	Trends, random walks, unit roots, and difference stationarity	Ch. 8
17	Nonstationary systems, long-run interpretation, and empirical issues	Ch. 8
18	Continuous-time finance and diffusion-based intuition	Ch. 9
19	Q&A / tutorial / assignment discussion / review	Review
20	Final exam	Exam

How to read the lecture map

- Lectures 1–3 build the univariate foundation and the ARMA toolkit.
- Lectures 4–7 move to multivariate systems and long-run relations.
- Lectures 8–10 introduce specialized modeling blocks: volatility and nonparametrics.
- Lectures 11–15 focus on inference, filtering, and signal extraction.
- Lectures 16–18 address nonstationarity and continuous-time ideas.

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How R is used in this course

- R is integrated throughout the course rather than separated into a standalone programming week.
- In each major block, we will use R for:
 - simulation,
 - visualization,
 - model estimation,
 - diagnostics,
 - interpretation of empirical output.
- The goal is not coding for its own sake, but econometric understanding through implementation.

What I expect from you

- This is a course for **master's and PhD students**, so we will combine intuition with mathematical structure.
- I expect you to understand definitions, derivations, and empirical meaning.
- You should be able to explain major concepts in words, not only reproduce formulas.
- Regular review after each lecture will matter much more than last-minute memorization.

A good weekly routine for this course

- 1 Read the relevant textbook section before class.
- 2 During class, focus on the conceptual logic behind the formulas.
- 3 After class, rewrite the key definitions and derivations in your own words.
- 4 Reproduce the main R examples and change parameters to build intuition.
- 5 Bring questions to the tutorial / review session instead of letting confusion accumulate.

Why the sequence is designed this way

- We start with stationary linear models because they are the basic language of time-series econometrics.
- We then generalize to multivariate systems and structural interpretation.
- After that, we add volatility, nonparametrics, robust inference, filtering, and nonstationarity.
- The final continuous-time block broadens the perspective rather than replacing the discrete-time core.

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The series in one sentence

This lecture series moves from the foundations of dependent-data modeling to a broad toolkit for analyzing persistence, volatility, signal extraction, nonstationarity, and dynamic economic structure.

Welcome to the course

We begin with the foundations of univariate time-series analysis, and then build step by step toward a broader econometric view of dynamic systems.