







Forecasting Methods

[ECON 5753]

Sprint 2025 Syllabus

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	Lecture	WCOB 207 TR 11–12:15pm
	Office Hours	WCOB 408 TW 1–2:30pm
	Course Materials	https://github.com/kylebutts/UARK_5753/

Course Summary

This course will provide an introduction to forecasting methods. The class will teach you how to take a set of input variables and produce predictions of some outcome variable. We will survey a set of forecasting methods for your toolbox including: bivariate and multivariate regression; non-parametric and partially linear models; time-series regression; smoothing methods in time-series, and, if time-permitting spatial forecasting methods. The class will teach these methods theoretically and also teach you to estimate these models in the R programming language.

Though the class will also teach you fundamental principles of forecasting: goals of forecasting, fitting of models, evaluating model fit, and limitations of the models. By doing this, the class will equip you with the foundations to expand your toolbox over time and implement these tools as a *careful econometrician*.

Last, the course will try to highlight limitations of forecasting methods; trade-offs between forecasting methods (e.g. interpretability versus predictive accuracy); and help you understand what forecasting methods can not do (e.g. establish causality).

Course Materials

There are two primary textbooks that will be referenced in the class. These textbooks are both available for free online. You may buy a print version, but it is *not necessary* for the course.

1. “**ISLR**”: Gareth, J., Daniela, W., Trevor, H., & Robert, T. (2013). “[An introduction to statistical learning: with applications in R \(2nd edition\)](#)”. Springer.
2. “**FPP3**”: Hyndman, R. J., & Athanasopoulos, G. (2018). “[Forecasting: principles and practice \(3rd edition\)](#)”. OTexts.

The [github repository](#) will assign readings from each textbook as well as additional readings that complement the textbooks.

Coding Software

You will need to download *two* programs:

1. Install R from <https://cloud.r-project.org/>.
2. Install Positron (or RStudio) from <https://github.com/posit-dev/positron/releases>.

Mastering R will take time and dedication, but it is a powerful and adaptable tool that is highly valued by many employers. Invest the necessary effort and time, and you will see the benefits.

Assignments

Problem sets: Throughout the course, problem sets will be assigned. These will feature theoretical questions that will require you to write out responses. Additionally, portions of the assignments will ask you to code up different estimators and interpret your findings in words. Students are encouraged to work in groups of 2 to discuss how to approach the problem sets, but each student must hand in their own set of answers. Missing or late problem sets will receive no credit.

Exams: There will be an in-class midterm and a final exam for the class.

Projects: The class will feature two “capstone” projects that will ask you to select a dataset and use that data to create a forecasting model. Each project will accompany distinct parts of the course: forecasting with cross-sectional data and with time-series data.

The breakdown of your final grade will be as follows:

Assignment	Percent of grade
Problem Sets	20%
Project 1	15%
Project 2	15%
Midterm	25%
Final	25%

Policies

The student who missed exam must provide an official proven emergency which prevents you from attending class on the scheduled exam date within 24 hours after the missed exam to be allowed to take a makeup. Otherwise the student is not eligible to take a makeup exam and the missed exam equals zero points.

There will be due dates on the assignments. Like you, I am a busy person. I may grade the next day or a few days later. You have until I start grading assignments to turn it in without penalty, so take your chances.

If you have any questions during the lecture, feel free to ask right away. Your questions can benefit both you and other students who might have the same questions. If you arrive late or need to leave early, please sit near the door to minimize disruption to the class.

Access and Accommodations

Your experience in this class is important to me. University of Arkansas Academic [Policy Series 1520.10](#) requires that students with disabilities are provided reasonable accommodations to ensure their equal access to course content. If you have already established accommodations with the [Center for Educational Access \(CEA\)](#), please request your accommodations letter early in the semester and contact me privately, so that we have adequate time to arrange your approved academic accommodations.

If you have **not** yet established services through CEA, but have a documented disability and require accommodations (conditions include but not limited to: mental health, attention-related, learning, vision, hearing, physical, health or temporary impacts), contact CEA directly to set up an Access Plan. CEA facilitates the interactive process that establishes reasonable accommodations. For more information on CEA registration procedures contact 479—575—3104, ada@uark.edu or visit <https://cea.uark.edu>.

Course Outline

1. Linear Algebra

Slides

Topics:

- Matrices and vectors
- Transpose
- Identity matrix and matrix inverse
- Dot product as $v'v$
- Statistics as matrix operations
- Derivatives as linear approximation
- Useful matrix derivative rules

Readings:

- Review Notes: [Probability and Statistics](#)
- [Introduction to Computational Finance and Financial Econometrics with R: Chapter 3](#)
- Video: [So You Think You Know How to Take Derivatives? | Steven Johnson](#)

2. Introduction to Forecasting

Topics:

- Goals of forecasting
 - Prediction and Inference
- How to evaluate a model
 - loss-function and mean-squared prediction error
- Overfitting and some solutions
 - Cross-validation
- Types of Datasets
 - cross-sectional, time-series, and panel

Readings:

- ISLR: All of 2.1, 2.2 thru 2.2.2

3. Regression Theory

Topics:

- Conditional expectation function $\mathbb{E}(y \mid X)$
- Bivariate Regression
 - Derivation of bivariate regression
 - Derivation of standard errors
- Forecasting
 - Derivation of standard errors
- Marginal (Predictive) Effects

Readings:

- ISLR: All of 3.1, 3.2, 3.3, and 3.4
- [Introduction to Econometrics with R](#): Section 4.5 on Sample Distribution of $\hat{\beta}$
- [Marginal Effects Book](#): Chapter 5 on Prediction

4. Regression in Practice

- Common Regression Terms
 - Polynomials
 - Indicator variables
 - Discrete variables
 - Interactions
 - Bins
 - Partially linear-model and Splines
- log outcome variables
- 0/1 Outcome variables
 - Linear probability model
 - Logistic regression

Readings:

- ISLR: All of 7.1, 7.2, 7.3, and 7.4

- log readings: [Lecture Notes on Linear Regression Models with Logarithmic Transformations](#) and [FAQ How Do I Interpret A Regression Model When Some Variables Are Log Transformed?](#)
- ISLR: All of 4.1, 4.2, and 4.3
- [Marginal Effects Book](#): Chapter 6 on Slopes

5. Time-series Forecasting

Topics:

- Time-series regression
 - Estimating holiday and seasonal patterns (day of week, monthly effects, etc.)
 - Linear time-trends
 - Piecewise linear time trends
 - * Pre-selected breaks
 - * Data selected breaks
- Smoothing methods
 - One-sided and two-sided rolling averages
 - Exponential smoothing methods
- Prophet model

Readings:

6. Spatial Forecasting (time permitting)

Topics:

- Kriging and other smoothing methods

Readings:

Tentative Schedule

Table 1: Tentative Schedule

Week	Dates	Tuesday	Thursday
01	01/14 - 01/16	Syllabus and Linear Algebra	Linear Algebra
02	01/21 - 01/23	Introduction to Forecasting	Introduction to Forecasting
03	01/28 - 01/30	Conditional Expectation	Bivariate Regression
04	02/04 - 02/06	Bivariate Regression	Bivariate Regression
05	02/11 - 02/13	R Day	Multivariate Regression
06	02/18 - 02/20	Multivariate Regression	Nonparametrics and Partially Linear
07	02/25 - 02/27	R Day	Log Transformations
08	03/04 - 03/06	Logistic Regression	R Day
09	03/11 - 03/13	Clustering(?)	Cross-sectional Project
10	03/18 - 03/20	Midterm Exam	Cross-sectional Project
11	04/01 - 04/03	Time-series Regression	Time-series Regression
12	04/08 - 04/10	R Day	Smoothing Methods
13	04/15 - 04/17	Smoothing Methods	Smoothing Methods
14	04/22 - 04/24	R Day	Prophet
15	04/29 - 05/01	Time-series Project	Time-series Project