Lecture 0: Class Setup

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FGV EPGE

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Course Information

- Course: Econometrics I (Part I);
- Instructor: Raul Guarini Riva raul.riva@fgv.br;
- **Teaching Assistant**: Taric Latif tariclatif@gmail.com;
- Class Schedule: Tuesdays and Thursdays, 11:00–13:00;
- **Office Hours**: Fridays, 9:00-10:00;
- TA Session Schedule: check with Taric;
- GitHub repository: https://github.com/epge-classes/econometrics-fgv-2025
- After class, you should review the online syllabus in detail;

We will use Github for course materials, assignments, and so on.

Why Should You Take This Class Seriously?

- Most students will engage in empirical research across diverse fields of Economics:
 - o Labor, Development, Urban, Finance, Health, IO, even Macro...!
 - o Obviously important if you want to do Econometrics;
- So far, most of what you learned is about **cross-sectional data** and **OLS**;
- We go beyond cross-sectional data to tackle: **Time Series** and **Panel Data**;
- Time Series data is your bread and butter in Macro, Finance, and many other fields;
- Panel Data methods (fixed effects, diff-in-diff, ...) are everywhere in empirical research;
- We will also cover a bit of non-parametric methods and bootstrap techniques;

Learning Goals (by Week 5)

After the first half of the course, you should be able to:

- 1. Understand and implement non-parametric kernel-based methods;
- 2. Apply bootstrap methods for i.i.d. data;
- 3. Analyze time series data:
 - o Concepts: stationarity, ergodicity, persistence, autocorrelation, ...
 - o Estimation and forecasting with ARMA models;
 - Asymptotic distribution of the OLS estimator under autocorrelation;
- 4. Implement GMM estimation under heteroskedasticity and autocorrelation;
 - Super relevant for Panel Data methods;
 - Also very useful to estimate structural models;

Books? Slides?

- I will provide slides for the lectures;
- These slides will be mostly based on two books + a famous reference here and there:
 - Econometrics, by Bruce Hansen (hello, old friend!);
 - Time Series Analysis, by James Hamilton (very comprehensive, also very long);
 - Stochastic Limit Theory, by James Davidson (seriously dark magic)
- I will highlight the recommended readings in the slides, and reading the books is mandatory.
- Slides are not a substitute for reading the books;
- I will make slides available on Github just before class;

Tentative Weekly Topics

- Week 1: Non-parametric estimation + Bootstrap for i.i.d. data;
- Week 2:
 - The structure of time series data, stationarity, ergodicity, persistence, autocorrelation, ...;
 - ARMA models + the Box-Jenkins methodology + Wold Decomposition;
- Week 3:
 - Estimation of ARMA models, MLE estimators...
 - LLN's and CLT's for dependent data focus on using the results;
- Week 4:
 - LLN and CLT's for dependent data focus on using the results rather than proving every single little thing;
 - Intro to GMM;
- Week 5:
 - GMM asymptotic theory;
 - GMM and HAC estimation + numerical pitfalls;

Evaluation and **Grades**

- Grades for the first and second half of the class are independent;
- Final grade = $0.5 \times (Grade \text{ on Part I}) + 0.5 \times (Grade \text{ on Part II});$
- The first part has 3 problem sets and 1 exam:

Grade on Part I =
$$\frac{1}{9} \times PS1 + \frac{1}{9} \times PS2 + \frac{1}{9} \times PS3 + \frac{2}{3} \times Exam$$
 (1)

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- There will one private repository for each group and each problem set;
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- You should deliver both a report answering the questions and the code used to generate the results;
- Feel free to use any civilized programming language (Python, R, Julia, Matlab, ...);
- No, you cannot use Stata or pre-packaged solutions in any of the languages above
 details will be greefied in the greefier sets.
 - details will be provided in the problem sets;

Attendance

- It is not mandatory at all you are adults;
- However, if you come, you have to be on time;
- 10-minute grace period after that, you will not be allowed to enter the classroom;



One more thing...

This is really important!

- This is the first time I teach this class;
- I need ongoing feedback;
- Is there something great about? Tell me! If there something terrible about it? Also tell me!
- Don't be shy negative feedback will never be punished in any way;
- If anything, on the contrary!