

# Advanced Applied Econometrics

## Week 1 - Intro and Static Discrete Choice

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# Introduction

- **EMAILS: mail list for the course**
- Organization, see outline

If anything **unclear**, please ask.

**Presentation:** name, study status, main econ interest

# Course schedule

- 1 Methods
- 2 Paper discussion  $\Rightarrow$  you **need to read at home**
- 3 Practical session (numerical methods)
  - $\Rightarrow$  you **need** to do problem sets
  - $\Rightarrow$  you **need** STATA and Python
  - $\Rightarrow$  you **need** to practice coding

# Evaluation

For students who need to be evaluated

## (1) **Special problem sets**

- individually or in pairs
- provide commented code, results, documentation, tables

## (2) **Exam**

- 21.7.2023;

# Plan for today

- Structural econometrics
- Examples of a structural model
- Practical session,
- Python and example with Max Blesch

# What is structural econometrics?

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Structural econometric models...

- “...combine explicit **economic theories with statistical models**” (Reiss and Wolak, 2007)
- “...[infer] underlying parameters that represent **tastes, technology, and other primitives** from observations of individual households and firms...” (Adda and Cooper, 2003)
- ... estimate **features of a data generating process** (i.e., a model) that are (assumed to be) invariant to the policy changes or other counterfactuals of interest (Haile, 2022)

# Descriptive vs structural analyses

- Descriptive analysis: establish facts about observable quantities
- RCT
- Policy evaluation
- Economic model



# Program Evaluation

Haile (2022):

- Program evaluation (indeed, any type of so-called "causal inference" is always a form of structural estimation. It requires a set of maintained hypotheses about the world (i.e., a model) allowing one to define and identify a counterfactual quantity of interest.
- TT, ATE, LATE, QTE, etc. are all precisely defined only under a well specified model of how the data are being generated. Any suggestion that these objects are **"model free" is nonsense**.
- Causality is always defined by a counterfactual.

# Reduced form vs structural

Haile (2022):

- "Reduced form" is sometimes used to mean "equation I won't derive, justify, or take questions on, but which I will nonetheless treat as causal when I talk about conclusions. **This is just bad science.**"

Other views and definitions by Rust (2010), thoughtful (and spiteful) comments on Keane (2013) in Fritjers (2013) vs Rust (2013).

# An Introductory Example

How much revenue raised by **introducing income taxes**?

**Simple answers?**

# Effects of tax change I

How much **additional revenue (AR)**

1  $AR = \text{new tax rate} * \text{wage} * \text{number of hours worked}$

2  $AR = \text{new tax rate} * \text{wage} * \text{hours worked post tax}$

**How many hours worked post tax?**

# Effects of tax change II

**How many hours worked post tax?**

- estimate **observed corr between wages and hours**
- interpret as  $L^S$ -elasticity?

# Effects of tax change III

using theory... why might **number of hours worked change**?

- 1 **hours margin**: change in hours for (pre- & post-) workers
- 2 **participation margin**: change in participation pre vs. post-tax:

## Effects of tax change IV

### 3 **life-cycle** margin:

changing taxes may influence optimal allocation of work over life  
(eg. if progressivity is increased...)

### 4 **wage** margin: **equilibrium (gross) wage may change**

Conclude: Predicting reform effect (AR) requires structural model. But: Model choice depends on **margin** and **identification**.

# Before the practical session

10 minute break



# Practical Session - Economic model simple version of van Soest 1995

Question: The Flat Party wins elections and wants to **introduce a Flat Income Tax (FIT)**.

We want to know **how much revenue is raised under FIT**.

Why do we need a structural model? Alternatives?

How to model this ?

# Practical Session

What minimal necessary ingredients for a model?

- 1 decision to participate
- 2 decision over hours worked (discretize - why?)
- 3 earnings of participants (pre- and post reform)
- 4 NB. recognize heterogeneity (e.g. due to different non-labour incomes)

# (1) Participation and hours

How do we model this?

# (1) Participation and hours II

Individuals **maximize utility** from consumption & leisure

$$U(c, h) = \gamma \left[ \frac{c^\theta}{\theta} - \alpha h \right] + \varepsilon_h$$

by **choosing hours per week**  $h \in [0, 10, 20, 30, 40]$   
faced with constraint posed by wage.

## (2) Consumption

Simplest model of consumption in this context?

## (2) Consumption II

$$c = w * h$$

What does this imply?  
Simplest models of wages?

### (3) Wages

Assume workers draw wage rates from normal wage offer distribution *before* choosing hours

$$\log w = \mu_w + \epsilon_w$$

where  $\epsilon_w$  from  $N[0, \sigma_w]$

# Estimation and Identification

Use maximum likelihood to estimate...

- 1 parameters of the wage distribution
- 2 taste parameters of leisure-consumption choice

How is the model identified? How can we estimate a causal effect based on the model?



# Coding

## Principles

- 1 Use a **script** and **document your code**. Always.
- 2 There is **always** a solution. Not one.
- 3 One solution is **fast**,
- 4 another is **easy to read** & understand,
- 5 another is **flexible**.
- 6 You won't find any immediately. No matter. **Try again**.

# Simulate to estimate

Good practice: Simulate before estimating. Why?

# In practice: Simulation and Estimation

- 1 Simulate economic behaviour in model with your favourite parameters.
- 2 Generate simulated data.
- 3 Save the data. Forget the parameters.
- 4 Write the likelihood function as a function of data and parameters.
- 5 Maximize likelihood of observing your simulated data w.r.t. parameters.
- 6 Rediscover your favourite parameters.