## **Introduction to Numerical Methods for Macroeconomics**

Macroeconomics 3: TA class #1

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# **About Myself**

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- ▶ JMP: markups on lending rates and markdowns on deposit rates
- Side project: Unemployment, SDFs and Dual Labor Markets in Europe

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Material https://github.com/AndreaPasqualini/numerical\_methods\_macroeconomics

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- Now it's time for fun: empirical tools!

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Objective: get familiar with projection methods and related applications

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## Advantages of Python

- Free and open-source, reliable tool
- Unbeatable flexibility

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### Advantages of Python

- Free and open-source, reliable tool
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- ▶ Many options to work with Python: VSCode, Spyder, PyCharm, Jupyter Notebooks, etc.
- ► These classes: Jupyter Notebooks (VSCode behind the scenes)

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### Perturbation methods

- Rely on Taylor expansion
- Require differentiability of the model
- Low computational costs

## Projection methods

- Rely on Bellman equations
- Allow for heterogeneity, discontinuities
- High computational costs

## Intro to Numerical Methods for Economists: Hands-on Example

### **Example: Neoclassical Stochastic Growth Model**

$$\begin{aligned} \max_{C_t, K_{t+1}} \mathbf{E}_0 & \sum_{t=0}^{\infty} \beta^t u(C_t) \\ \text{s.t.} & \begin{cases} C_t + K_{t+1} = Z_t K_t^{\alpha} + (1+\delta)K_{t-1} & \forall \ t \\ \log(Z_{t+1}) = (1-\rho)\mu + \rho \log(Z_t) + \log(\varepsilon_{t+1}) & \forall \ t \\ \varepsilon_{t+1} & \stackrel{iid}{\sim} \mathcal{N}(0, \sigma^2) & \forall \ t \\ C_t, K_{t+1} > 0 & \forall \ t \\ K_0, Z_0 \text{ given} \end{cases} \end{aligned}$$

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### Example: Neoclassical Stochastic Growth Model

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#### Variables

- Endogenously predetermined: Z<sub>t</sub>, K<sub>t</sub>
- **Exogenous shocks:**  $\varepsilon_{t+1}$
- ▶ Controls:  $C_t$ ,  $K_{t+1}$
- Forward looking:  $C_{t+1}$

### Equations for the equilibrium

$$\begin{cases} u'(C_t) = \beta \cdot \mathbf{E}_t \left( u'(C_{t+1}) \left[ \alpha K_{t+1}^{\alpha-1} + 1 - \delta \right] \right) \\ C_t + K_{t+1} = Z_t K_t^{\alpha} + (1 + \delta) K_t \\ \log(Z_{t+1}) = (1 - \rho)\mu + \rho \log(Z_t) + \log(\varepsilon_{t+1}) \end{cases}$$

Focus on the equations that characterize the equilibrium (w/ CRRA utility)

$$\begin{cases} C_t^{-\gamma} = \beta \cdot \mathbf{E}_t \left( C_{t+1}^{-\gamma} \left[ \alpha K_{t+1}^{\alpha-1} + 1 - \delta \right] \right) \\ C_t + K_{t+1} = Z_t K_t^{\alpha} + (1 + \delta) K_t \\ \log(Z_{t+1}) = (1 - \rho)\mu + \rho \log(Z_t) + \log(\varepsilon_{t+1}) \end{cases}$$

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- A (deterministic) steady state
- Derivatives of each equation

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Log-linear representation of the model (1st order Taylor expansion around the steady state)

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Can solve this system of linear equations with linear algebra (e.g., Schur decomposition)

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Focus on the optimization problem (write the associated Bellman equation)

$$\begin{split} V(K,Z) &= \max_{C,K'} u(C) + \beta \mathbf{E} \left( V(K',Z')|Z \right) \\ \text{s.t. } \begin{cases} C + K' &= ZK^{\alpha} + (1+\delta)K \\ \log(Z') &= (1-\rho)\mu + \rho \log(Z) + \log(\varepsilon'), \quad \varepsilon' \stackrel{\textit{iid}}{\sim} \mathcal{N}(0,\sigma^2) \end{cases} \end{split}$$

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Can crack this by letting the computer loop the contraction mapping **T** 

For more complicated supply-demand models (let the "real" equilibrium price be  $P^*$ )

- 1. Guess an equilibrium price  $P^{(h)}$
- 2. Obtain the policy functions associated to the Bellman equation, for the given price  $P^{(h)}$
- 3. Define the excess demand function D(P)
- 4. Observe that D(P) is decreasing in P
  - ▶ If  $D(P^{(h)}) > 0$ , then  $P^{(h)} < P^*$
  - ► If  $D(P^{(h)}) < 0$ , then  $P^{(h)} > P^*$
- 5. Propose a new guess  $P^{(h+1)}$  accordingly
- 6. Repeat steps 1–5 until  $|P^{(h+1)} P^{(h)}| < \epsilon$

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- Analyses of policy functions (if non-trivial)
- ► Impulse-Response Functions (IRFs)
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- Do these mechanisms matter quantitatively? Write model, compare simulations with data

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### Why do projection methods matter?

- ▶ Models that suffer from derivatives (i.e., where higher order moments matter)
- Models with heterogeneous agents
- Models with binding constraits

# Intro to Python

Moving to a Jupyter Notebook