Exercise class 12

Introduction to Programming and Numerical Analysis

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UNIVERSITY OF COPENHAGEN



Plan for today

Dynamic programming

Model project

Plan for today

- 1. Now-15.50: Work on optimization problems
- 2. 15.50-16.00: We talk about optimization in class
- 3. 16.00-16.15: Break
- 4. 16.15-16.35: Intro to dynamic programming
- 5. 16.35-17.00: Solving dynamic optimization problems in class

reminology...

State variable: A variable determining the *state of the world* as it is when I make my decision. State variables evolve over time according to a *transition process*.

Example: Cash-on-hand in consumption-savings model.

Terminology...

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Choice variable: The variable I *choose* to maximize utility (*also control variable*). ample: Consumption in consumption-savings model.

Policy function: The optimal *choice* as a function of *state*.

Value function: The total value of today's state: Utility today + discounted value of tomorrow's state.

Period 1:

$$egin{aligned} V_1(\emph{m}_1) &= \max_{\emph{c}_1} \emph{u}(\emph{c}_1) + eta \mathbb{E}\left[V_2(\emph{m}_2)
ight] \ s.t. \quad \emph{m}_2 &= R(\emph{m}_1 - \emph{c}_1) + \emph{y}_2 \ \emph{y}_2 &\sim \emph{U}(0,1) \end{aligned}$$

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Period 2:

$$V_2(m_2) = \max_{c_2} u(c_2)$$

s.t. $m_2 \ge c_2$

State variable(s)? Choice variable(s)? Policy function(s)?

State variable(s)? m_1 , m_2 Choice variable(s)? Policy function(s)?

State variable(s)? m_1 , m_2 Choice variable(s)? c_1 , c_2 Policy function(s)?

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State variable(s)? m_1, m_2
Choice variable(s)? c_1, c_2
Policy function(s)? c_1^*(m_1), c_2^*(m_2)
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Period 1:

$$V_1(m_1) = \max_{c_1} u(c_1) + \beta \mathbb{E} \left[\frac{V_2(m_2)}{c_1} \right]$$

s.t. $m_2 = R(m_1 - c_1) + y_2$
 $y_2 \sim U(0, 1)$

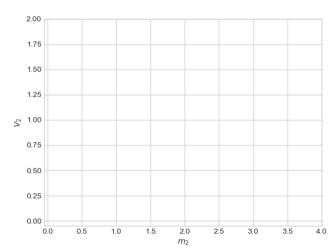
The optimal choice today depends on the value function tomorrow - which in turn depends on the optimal choice tomorrow.

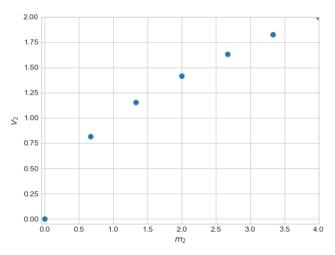
So we need to solve tomorrow's problem first - backwards induction.

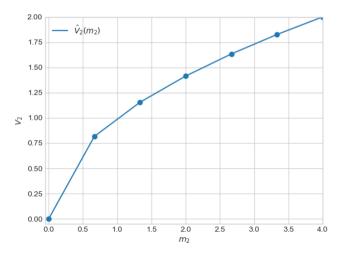
Period 2:

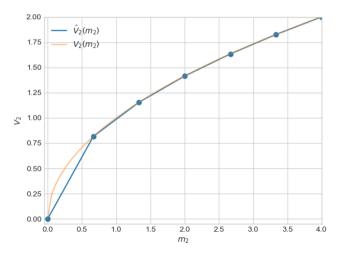
$$egin{aligned} V_2(m_2) &= \max_{c_2} u(c_2) \ s.t. \quad m_2 \geq c_2 \end{aligned}$$

This is just a standard constrained optimization problem. We can solve this for any value of m_2 , we like.









Back to period 1...

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Now we know the (approximate) period 2 value function, so we can solve the problem in period 1.

We still need to take expectations with respect to $\hat{V}_2(m_2)$, where m_2 is stochastic...

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Now we know the (approximate) period 2 value function, so we can solve the problem in period 1.

We still need to take expectations with respect to $\hat{V}_2(m_2)$, where m_2 is stochastic... \rightarrow Monte Carlo Integration for continuous m_2 , weighted average for discrete m_2 !

Build an analyze an economic model of your choosing:

- 1. Extend a model from this course
- 2. A model you know from other courses
- 3. Something you find interesting!

Preferably, you use your numerical skill to do some analysis that would not have been possible using just analytical tools (eg. non-Cobb-Douglass production or utility functions, heterogeneity, uncertainty etc.)

MAY 12th IS A HARD DEADLINE! I have to approve you for the exam by May 15th, so there will be **no time for resubmissions!**

Next time...

Video lectures:

- Structural estimation
- OLG and Ramsey models

Exercises

Work on data project