

Dynamic Optimisation

MATLAB AND MICRODATA PROGRAMMING GROUP

HILARY 2014
21 FEBRUARY



Outline

- 1 Direct Attack
- 2 The Bellman Equation
- 3 Uncertainty

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Choice over time

Dynamic problems have two aspects: stocks and flows.

- The state variable summarises stocks
- The control variable is the variable being chosen (ie flows)

$$U = \sum_{t=1}^T \beta^{t-1} u(c_t), \quad (1)$$

$$k_{t+1} = f(k_t, c_t). \quad (2)$$

A Dynamic Household

Attach functional forms to (1) and (2):

$$u(c_t) = \ln(c_t)$$

$$f(k_{t+1}) = k_t - c_t$$

Then ...

A Dynamic Household

$$\begin{aligned} \max_{\{c_t\}_1^T} \sum_{t=1}^T \beta^{t-1} \ln(c_t) \quad \text{s.t.} \quad & \sum_{t=1}^T c_t + k_{T+1} = k_1 \quad (3) \\ & c_t \geq 0 \\ & k_t \geq 0. \end{aligned}$$

MATLABbing it

We should be able to solve this problem by “direct attack” in MATLAB

- A function to maximise
- A vector of maximands
- A vector of upper and lower bounds
- A(n) (in)equality constraint
- our old friend `fmincon`

```
1  function V = flowUtility(T,Beta,C)
2      % flowUtility(T,Beta,C) takes T periods of
3      % consumption of size C (a Tx1 vector), and
4      % calculates the total utility of consumption
5      % assuming an additively separable utility
6      % function and discount rate  $\beta$ .
7
8      t = [1:1:T];
9      V = Beta.^(t-1)*log(C);
10     V = -V;
11
12  return
```


Sensitivity

We have assumed a particular functional form, and values for input parameters

- Here we are imposing these, rather than recovering them
- Of course, we can re-solve the model based on alternative assumptions. . .
 - Alternative values of β
 - Alternative utility functions
 - Alternative forms of the flow equation (see chapter)

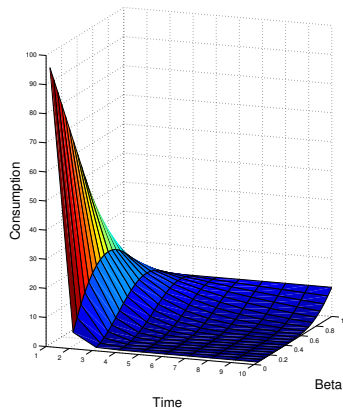
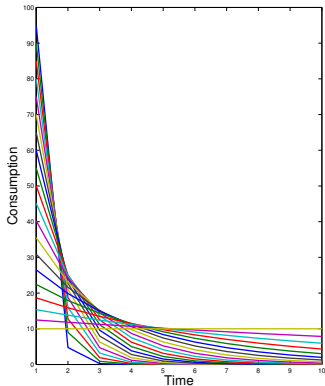


Figure: Sensitivity of Consumption to Discount Rate

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The Bellman Equation

Generally when people speak about ‘dynamic programming’ in economics, they refer to the class of models solved using value function iteration.

- While solvers like `fmincon` are useful as a general outline, often we need more flexible methods of attack
- This is where the Bellman equation comes in handy
- Essentially, breaks down the problem into sequentially much smaller problems

The Bellman Equation

$$V(k_t) = \max_{c_t} \{u(c_t) + \beta V(k_{t+1})\} \quad (4)$$

Iteration...

So, we can break this down into sub-problems:

$$\begin{aligned}
 V(k_T) &= \max_{c_T} \{u(c_T) + \beta V(k_{T+1})\} \\
 V(k_{T-1}) &= \max_{c_{T-1}} \{u(c_{T-1}) + \beta V(k_T)\} \\
 V(k_{T-2}) &= \max_{c_{T-2}} \{u(c_{T-2}) + \beta V(k_{T-1})\} \\
 &\vdots \\
 V(k_2) &= \max_{c_2} \{u(c_2) + \beta V(k_3)\} \\
 V(k_1) &= \max_{c_1} \{u(c_1) + \beta V(k_2)\}
 \end{aligned} \tag{5}$$

Iteration II

Now, all we need is a place to start...

$$V(k_{T+1}) = 0 \quad \forall k \quad (6)$$

MATLABbing it

We'll solve Bellman equations numerically with MATLAB

- Essentially, 'brute force' grid search
- Requires 'gridding' state variables (if not binary)
- Let's check out `backwardsInduc.m`

'Memoization'

A brief final point here: this is computationally intense, but we can avoid a lot of repeated heavy lifting

- 'Memoization' (aka computer programming in 'Nature')
- This is something that comes in very handy when simulating and solving these problems

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Uncertainty

What we've seen so far is actually remarkably flexible.

- Generalises quite simply (in theory) to multiple state and control variables
- Though in practice, curse of dimensionality
- Perhaps the only major thing we're missing is stochastic elements

The Bellman Equation

$$V(k_t) = \max_{c_t} \{u(c_t) + \beta \mathbb{E}[V(k_{t+1})]\} \quad (7)$$

Decisions Under Uncertainty

So, now the decision must be framed in terms of consumption now and *expected* consumption in the future.

- In this case, the backwards iteration step is similar
- However, the iterating forwards to solve the model depends upon progressive realisations of shocks
- If time: `finiteStochastic.m`,
`simulateStochastic.m`

Simulations

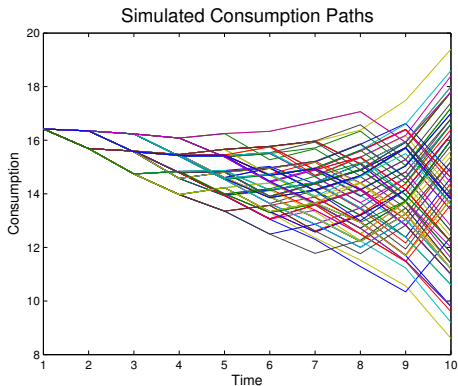


Figure: Simulated Consumption in a Stochastic Model

Summary

This week:

- Finite horizon dynamic optimisation
- Bellman equations
- A little bit of model simulation

Next week:

- Infinte horizons
- Using Bellman again
- Estimation!!

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