

# Advanced Macroeconomics

## Lecture 6 - Introduction to dynamic programming

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## Two different approaches

consider the following problem:

$$\sup_{\{k_{t+1}, c_t\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(c_t)$$

subject to  $k_{t+1} = f(k_t) + (1 - \delta)k_t - c_t$ ,  $k_{t+1} \geq 0$  for all  $t$ ,  $k_0$  given

- ▶ **direct approach**: set up the Lagrangian, find the two optimal infinite sequences  $\{k_{t+1}\}_{t=0}^{\infty}$  and  $\{c_t\}_{t=0}^{\infty}$
- ▶ **dynamic programming approach**: find a **time-invariant policy function**  $h(\cdot)$  mapping wealth into optimal consumption, i.e.  $c_t = h(k_t)$   
iterating together with  $k_{t+1} = f(k_t) + (1 - \delta)k_t - c_t$  starting from  $k_0$  gives the two optimal sequences

## Potential advantages of dynamic programming

it is unclear that finding a policy function is easier than finding an infinite sequence, but it has three advantages:

1. sometimes we can find a closed-form solution for the policy function  $h(\cdot)$
2. sometimes we can characterize theoretical properties of the policy function  $h(\cdot)$
3. various numerical methods are available to solve dynamic programs

We will proceed by

2.1 setting up the Sequence problem

2.2 the Bellman equation

2.3 solving the Bellman equation

## 2.1 Setting up

- ▶ discrete time dynamic optimization
- ▶ infinite-horizon stationary problem

**the sequence problem:** find  $v(x)$  such that

$$v(x_0) = \sup_{\{x_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t F(x_t, x_{t+1})$$

subject to  $x_{t+1} \in \Gamma(x_t)$  (or  $x_{t+1} \in \Gamma(x)$ ), with  $x_0$  given

where

- ▶  $x_t$  is the **state vector** at date  $t$
- ▶  $F(x_t, x_{t+1})$  is the **flow payoff** at date  $t$  ( $F$  is 'stationary')
- ▶  $\beta$  is the **exponential discount factor**,  $\beta^t$  is the exponential discount function

## Example 1

optimal growth with log utility and Cobb-Douglas production function:

$$\sup_{\{k_{t+1}, c_t\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t \ln(c_t)$$

subject to the constraints:  $c_t, k_t > 0$  and  $c_t + k_{t+1} = k_t^\alpha$ ,  $k_0$  given

Question: how do you formulate this as a sequence problem?

[Hint: eliminate redundant variables, and introduce the constraint function  $\Gamma$ ]

## 2.2 The Bellman equation

**the Bellman equation** expresses the value function as a combination of a flow payoff and a discounted continuation payoff

$$v(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta v(x_{+1})\} \quad \forall x$$

- ▶ flow payoff is  $F(x, x_{+1})$
- ▶ current value function is  $v(x)$ , continuation value function is  $v(x_{+1})$
- ▶ equation holds for all feasible values of  $x$

→ compare **the sequence problem** and **the Bellman equation**

- ▶ we call  $v(\cdot)$  the solution to the Bellman equation
- ▶ not trivial to find
- ▶ we haven't even demonstrated yet that there exists a function  $v(\cdot)$  that will satisfy the Bellman equation
- ▶ we will show that the (unique) value function defined by the sequence problem is also the unique solution to the Bellman equation
- ▶ once we determine the value function  $v(x)$ , we can solve for the optimal policy:

$$x_{+1}^* = \arg \max_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta v(x_{+1})\}$$

→  $x_{+1}^*$  is a function of  $x$ :  $h(x) \equiv x_{+1}^*$



a solution to the sequence problem is also a solution to the Bellman equation

$$\begin{aligned} v(x_0) &= \sup_{x_1 \in \Gamma(x)} \sum_{t=0}^{\infty} \beta^t F(x_t, x_{t+1}) \\ &= \sup_{x_1 \in \Gamma(x)} \left\{ F(x_0, x_1) + \sum_{t=1}^{\infty} \beta^t F(x_t, x_{t+1}) \right\} \\ &= \sup_{x_1 \in \Gamma(x)} \left\{ F(x_0, x_1) + \beta \sum_{t=1}^{\infty} \beta^{t-1} F(x_t, x_{t+1}) \right\} \\ &= \sup_{x_1 \in \Gamma(x_0)} \left\{ F(x_0, x_1) + \beta \sup_{x_1 \in \Gamma(x)} \sum_{t=0}^{\infty} \beta^t F(x_{t+1}, x_{t+2}) \right\} \\ &= \sup_{x_1 \in \Gamma(x_0)} \{ F(x_0, x_1) + \beta v(x_1) \} \end{aligned}$$

and vice versa: a solution to the Bellman equation is also a solution to the sequence problem

$$\begin{aligned}v(x_0) &= \sup_{x_1 \in \Gamma(x_0)} \{F(x_0, x_1) + \beta v(x_1)\} \\&= \sup_{x_{+1} \in \Gamma(x)} \{F(x_0, x_1) + \beta [F(x_1, x_2) + \beta v(x_2)]\} \\&= \sup_{x_{+1} \in \Gamma(x)} \{F(x_0, x_1) + \beta F(x_1, x_2) + \beta^2 [F(x_2, x_3) + \beta v(x_3)]\} \\&\vdots \\&= \sup_{x_{+1} \in \Gamma(x)} \{F(x_0, x_1) + \dots + \beta^{n-1} F(x_{n-1}, x_n) + \beta^n v(x_n)\} \\&= \sup_{x_{+1} \in \Gamma(x)} \sum_{t=0}^{\infty} \beta^t F(x_t, x_{t+1})\end{aligned}$$

sufficient condition:  $\lim_{n \rightarrow \infty} \beta^n v(x_n) = 0$  for all feasible  $x$  sequences

$\Rightarrow$  put together: a solution of the Bellman equation will also be a solution to the sequence problem and vice versa

## Example 2

optimal growth (log utility and Cobb-Douglas):

$$\sup_{\{k_{t+1}, c_t\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t \ln(c_t)$$

subject to the constraints:  $c_t, k_t > 0$  and  $c_t + k_{t+1} = k_t^\alpha$ ,  $k_0$  given

sequence problem formulation

with Bellman equation notation

## 2.3 Solving the Bellman equation

► three methods

1. guess a solution and verify
  2. iterate a functional operator analytically
  3. iterate a functional operator numerically
- 
1. in practice: guess a function  $v(x)$  and then check to see that this function satisfies the Bellman equation at all possible values of  $x$

## Example 3

- ▶ for the growth example guess that the solution takes the following form:

$$v(k) = A + B \ln(k)$$

where  $A$  and  $B$  are constants for which we need to find solutions

- ▶ here the value function inherits the functional form of the utility function ( $\ln$ )
- ▶ to solve for the constants rewrite the Bellman equation as

$$v(k) = \sup_{k_{+1} \in \Gamma(k)} \{ \ln(k^\alpha - k_{+1}) + \beta v(k_{+1}) \} \quad \forall k$$

$$A + B \ln(k) = \sup_{k_{+1} \in \Gamma(k)} \{ \ln(k^\alpha - k_{+1}) + \beta (A + B \ln(k_{+1})) \} \quad \forall k$$

- verify that this functional form works and calculate  $A$  and  $B$  (Problem Set)

- ▶ to solve such a problem, we need to use the first order condition (on the right hand side) of the Bellman equation:

$$\frac{\partial F(x, x_{+1})}{\partial x_{+1}} + \beta v'(x_{+1}) = 0$$

- ▶ and the envelope theorem

$$v'(x) = \frac{\partial F(x, x_{+1})}{\partial x}$$

- ▶ heuristic demonstration of the ET

$$\begin{aligned} v'(x) &= \frac{\partial F(x, x_{+1})}{\partial x} + \frac{\partial F(x, x_{+1})}{\partial x_{+1}} \frac{dx_{+1}}{dx} + \beta v'(x_{+1}) \frac{dx_{+1}}{dx} \\ &= \frac{\partial F(x, x_{+1})}{\partial x} + \underbrace{\left[ \frac{\partial F(x, x_{+1})}{\partial x_{+1}} + \beta v'(x_{+1}) \right]}_{=0 \text{ from the FOC}} \frac{dx_{+1}}{dx} \\ &= \frac{\partial F(x, x_{+1})}{\partial x} \end{aligned}$$

## The FOC and the ET provide the Euler equation

- ▶ forwarding the ET by one period we get

$$v'(x_{+1}) = \frac{\partial F(x_{+1}, x_{+2})}{\partial x_{+1}}$$

- ▶ plug this into the FOC

$$\begin{aligned}\frac{\partial F(x, x_{+1})}{\partial x_{+1}} + \beta v'(x_{+1}) &= 0 \\ \frac{\partial F(x, x_{+1})}{\partial x_{+1}} + \beta \frac{\partial F(x_{+1}, x_{+2})}{\partial x_{+1}} &= 0\end{aligned}$$

- ▶ in our growth example  $F(x, x_{+1}) = u(k^\alpha - k_{+1}) = u(c)$  so the above becomes

$$u'(c) = \beta \underbrace{\alpha k_{+1}^{\alpha-1}}_{=R_{+1}} u'(c_{+1})$$

# Iterative solutions to the Bellman equation

the Bellman equation

$$v(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta v(x_{+1})\} \quad \forall x$$

the Bellman operator, operating on function  $w$  is defined as

$$(\mathbf{B}w)(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta w(x_{+1})\} \quad \forall x$$

- ▶ definition is expressed pointwise (for a value of  $x$ ), but holds for all values of  $x$
- ▶ operator  $\mathbf{B}$  maps a function  $w$  to a new function  $\mathbf{B}w$
- ▶ operator  $\mathbf{B}$  maps functions, it is referred to as a functional operator
- ▶ the argument of  $\mathbf{B}$ , the function  $w$ , is not necessarily a solution to the Bellman equation



- ▶ let  $v$  be a solution to the Bellman equation
- ▶ if  $w = v$ , then  $\mathbf{B}v = v$

$$\begin{aligned}(\mathbf{B}v)(x) &= \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta v(x_{+1})\} \quad \forall x \\ &= v(x) \quad \forall x\end{aligned}$$

- ▶ the function  $v$  is a fixed point of  $\mathbf{B}$ , i.e.  $\mathbf{B}$  maps  $v$  to  $v$
- the iterative solution methods for the Bellman equation pick some  $v_0$  and analytically or numerically iterate  $\mathbf{B}^n v_0$  until convergence, until finding a fixed point

how do you iterate  $\mathbf{B}^n w$ ?

$$(\mathbf{B}w)(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta w(x_{+1})\}$$

$$(\mathbf{B}(\mathbf{B}w))(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta(\mathbf{B}w)(x_{+1})\}$$

$$(\mathbf{B}(\mathbf{B}^2 w))(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta(\mathbf{B}^2 w)(x_{+1})\}$$

$\vdots$

$$(\mathbf{B}(\mathbf{B}^n w))(x) = \sup_{x_{+1} \in \Gamma(x)} \{F(x, x_{+1}) + \beta(\mathbf{B}^n w)(x_{+1})\}$$

what does it mean for functions to converge?

$$\lim_{n \rightarrow \infty} \mathbf{B}^n v_0 = v$$

- ▶ informally as  $n$  gets large, the set of remaining elements in the sequence of functions

$$\{\mathbf{B}^n v_0, \mathbf{B}^{n+1} v_0, \mathbf{B}^{n+2} v_0, \dots\}$$

are getting closer and closer

- ▶ two functions are close if the maximum distance between the functions is bounded by  $\varepsilon$

## Contraction mapping and Blackwell's Theorem

- ▶  $\mathbf{B}^n w$  converges as  $n \rightarrow \infty$  because  $\mathbf{B}$  is a contraction mapping
  - ▶  $\mathbf{B}$  is a contraction mapping if operating  $\mathbf{B}$  on any two functions moves them closer together
  - ▶ if  $\mathbf{B}$  is a contraction mapping, then
    - ▶  $\mathbf{B}$  has exactly one fixed point  $v$
    - ▶ for any  $v_0$ ,  $\lim \mathbf{B}^n v_0 = v$
    - ▶  $\mathbf{B}^n v_0$  has an exponential convergence rate
  - ▶ Blackwell's Theorem gives sufficient (but not necessary) conditions for  $\mathbf{B}$  to be a contraction mapping
- ⇒ if Blackwell's conditions are satisfied the Bellman equation can be solved using iterative methods