

NUMERICAL METHODS-LECTURE III: BELLMAN EQUATIONS: THEORY

Trevor Gallen

Fall, 2015

PRELIMINARIES

- ▶ I assume you have seen Bellman equations before
- ▶ Seeing Bellman equations before is unnecessary
- ▶ I will introduce as if you have forgotten everything
- ▶ I leave the technical requirements to chapters 4 and 9 of Stokey & Lucas with Prescott, (RMED).
- ▶ Most of the problems we write down will be well-conditioned
- ▶ This class : DP math :: rocket engineering : physics
 - ▶ No need to know theory, but can blow up in your face
- ▶ With that shallow warning, we'll be learning Monkey Bellmans

SEQUENCE PROBLEM

- ▶ We can formulate a number of very interesting problems as:

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

$$\text{s.t. } x_{t+1} \in \Gamma(x_t), \quad t = 0, 1, 2, \dots$$

$$x_0 \in X \text{ is given.}$$

SEQUENCE PROBLEM

- ▶ We can formulate a number of very interesting problems as:

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

$$\text{s.t. } x_{t+1} \in \Gamma(x_t), \quad t = 0, 1, 2, \dots$$

$$x_0 \in X \text{ is given.}$$

- ▶ Too abstract for an engineer. What is this saying? Translate it.

SEQUENCE PROBLEM: PIECES

- ▶ $F(\cdot, \cdot)$ is something we're trying to maximize
- ▶ x_{t+1} is the thing we can control
- ▶ F is influenced by:
 - ▶ Something we control right now: x_{t+1}
 - ▶ Something that we don't control right now: x_t
 - ▶ But yesterday, we controlled it (except in first period)
- ▶ Our ability to control x_{t+1} is governed by x_t (via Γ)

GENERALITY? EXAMPLE 1: NCG

We get utility from consumption. Consumption is constrained by capital.

$$\max \text{NPV of } U(c_t) \text{ s.t. } (1-\delta)k_t + Ak_t^\alpha = c_t + i_t, \quad k_{t+1} = (1-\delta)k_t + i_t$$

- ▶ $U(c_t)$ is something we're trying to maximize
- ▶ k_{t+1} is the thing we can control
- ▶ F is influenced by:
 - ▶ Something we control right now: k_{t+1} (via LOM, c_t)
 - ▶ Something that we don't control right now: k_t
- ▶ Our ability to control k_{t+1} is governed by k_t , which gives our budget constraint

GENERALITY? EXAMPLE 2: MCCALL'S MODEL-I

- ▶ Search for job
- ▶ Each period, one offer w from bounded wage distribution with CDF $F(w)$
- ▶ Can reject, get c
- ▶ Can accept, get w forever
- ▶ Utility is equal to whatever income you get (c or w).

GENERALITY? EXAMPLE 2: MCCALL'S MODEL

$$\max \text{NPV of } U(c_t) \quad s.t. \quad c_t = \begin{cases} w & \text{if accept } w \\ c & \text{if not accepted} \end{cases}$$

$$\begin{aligned} w_{t+1} &= w && \text{if accept } w \\ w_{t+1} &\sim F(w) && \text{if not accepted} \end{aligned}$$

- ▶ $U(c_t)$ is something we're trying to maximize
- ▶ We control if w or c
- ▶ F is influenced by:
 - ▶ Our choice of c vs. w if not yet chosen (memoryless)
 - ▶ Our past choice of w if accepted (no choice)
- ▶ Next w is influenced by past w (sometimes)

GENERALITY? EXAMPLE 3: FIRM INVESTMENT

Firm wants to maximize profit given A and K , with profits:

$$a_i \in \{0, 1, \dots, 100\}$$

$$Pr(A_{t+1} = a_i | A_t) \text{ given}$$

$$\pi_t = \begin{cases} -(A_t - K_t)^2 & \text{if no change} \\ -(A_t - K_t^*)^2 - C & \text{if change} \end{cases}$$

$$\begin{cases} K_{t+1} = K_t & \text{if no change} \\ K_{t+1} = K_t^* & \text{if change} \end{cases}$$

Where if change, K_t^* is chosen from the range of a_i .

THE BELLMAN EQUATION - I

- ▶ Starting with a sequence problem:

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

s.t. $x_{t+1} \in \Gamma(x_t)$, $t = 0, 1, 2, \dots$

$x_0 \in X$ is given.

- ▶ We can define an equation:

$$V(x) = \max_{y \in \Gamma(x)} \{F(x, y) + \beta V(y)\}$$

- ▶ Call x the state, y the control, V the value function

THE BELLMAN EQUATION - II

$$V(x) = \max_{y \in \Gamma(x)} \{F(x, y) + \beta V(y)\}$$

- ▶ What are doing?
 - ▶ Want to maximize current and future F 's through choice of all future x 's
 - ▶ Simplify the problem: today, and the problem you wake up with tomorrow
 - ▶ Tomorrow, you maximize today, and the next day's problem
 - ▶ Define everything recursively: the NPV of happiness if you wake up with x and maximize is utility today plus the NPV of happiness when you wake up with whatever you have left tomorrow
- ▶ Call x our state, the thing we wake up with
- ▶ Call y our control, the thing we control
- ▶ Call Γ our law of motion

THE BELLMAN EQUATION - III

$$V(x_t) = \max_{\{y_s\}_{s=t}^{\infty}} \sum_{s=t}^{\infty} \beta^{s-t} F(x_s, y_s)$$

Break the first part out:

$$V(x_t) = \max_{\{y_s\}_{s=t}^{\infty}} \left[\beta^{t-t} F(x_t, y_t) + \sum_{s=t+1}^{\infty} \beta^{s-t} F(x_s, y_s) \right]$$

Break up the optimization and take out a β :

$$V(x_t) = \max_{y_t} \left[F(x_t, y_t) + \beta \max_{\{y_s\}_{s=t+1}^{\infty}} \sum_{s=t+1}^{\infty} \beta^{s-t-1} F(x_s, y_s) \right]$$

THE BELLMAN EQUATION - VI

$$V(x_t) = \max_{y_t} \left[F(x_t, y_t) + \beta \max_{\{y_s\}_{s=t+1}^{\infty}} \sum_{s=t+1}^{\infty} \beta^{s-t-1} F(x_s, y_s) \right]$$

Notice that if we replaced t with $t + 1$ in the definition, we would have:

$$V(x_{t+1}) = \max_{\{y_s\}_{s=t+1}^{\infty}} \sum_{s=t+1}^{\infty} \beta^{s-t-1} F(x_s, y_s)$$

Plugging this in:

$$V(x_t) = \max_{y_t} [F(x_t, y_t) + \beta V(x_{t+1})]$$

Given an x_t , just need a y_t , and whatever the value of all the future decisions we make will be.

SUMMING UP

- ▶ Bellmans are a flexible way of describing a large number of dynamic problems
- ▶ We can write down problems easily
- ▶ Now we want to numerically solve them
- ▶ Next up: value function iteration (extremely easy and intuitive ways to solve)