

LECTURE 9

EIVIND ERIKSEN

~~SEP 06 2012~~

DRE 70d7

MATHEMATICS

PLAN:

- ① Optimal control theory - discrete time
finite horizon
- ② Infinite horizon

Reading:

[FMEA] 12
[S] 11-12

① Problem (finite horizon)

$$\max_{(\min)} \sum_{t=0}^T f(t, x_t, u_t) \quad \text{subject to} \quad \begin{cases} x_0 \text{ given} \\ x_{t+1} = g(t, x_t, u_t) \text{ for } t=0, \dots, T-1 \\ u_t \in U \subseteq \mathbb{R} \end{cases}$$

Comments:

- the set $U \subseteq \mathbb{R}$ of admissible controls may depend on the state variable x_t
- Sometimes $f(T, x_T, u_T) = S(x_T)$ and is called a scrap value function

A) Dynamic programming and the Bellman equation

Let $s \leq T$ and define the optimal value function

$$J_s(x) = \max_{(u_s, \dots, u_T)} \sum_{t=s}^T f(t, x_t, u_t) \quad \text{subject to} \quad \begin{cases} x_s = x \\ x_{t+1} = g(t, x_t, u_t) \text{ for } t \geq s \\ u_t \in U \end{cases}$$

Bellman equation:

$$J_s(x) = \max_{u \in U} \left\{ f(s, x, u) + J_{s+1}(g(s, x, u)) \right\} \quad \text{for } s=0, 1, 2, \dots, T-1$$

$$J_T(x) = \max_{u \in U} f(T, x, u)$$

Ex 1: $\max \sum_{t=0}^3 (1+x_t+u_t^2) \quad x_{t+1} = x_t + u_t, \quad x_0 = 0, \quad U \subseteq \mathbb{R}$

$$J_3(x) = \max_{u \in \mathbb{R}} (1+x-u^2) = 1+x \quad \text{at } u_3 = 0$$

$$J_2(x) = \max_{u_2 \in \mathbb{R}} (1+x-u_2^2) + J_3(x+u_2) = \max_u \{1+x-u^2 + 1+x+u\}$$

$$\begin{aligned} ' &= -2u+1 & u &= 1/2 \\ '' &= -2 & & \text{ok.} \end{aligned}$$

$$= 2+2x+1/4 = \underline{9/4+2x}, \quad u_2 = 1/2$$

$$J_1(x) = \max_u (1+x-u^2) + J_2(x+u) = \max_u (1+x-u^2 + 9/4 + 2x+2u)$$

$$\begin{aligned} ' &= -2u+2 & u &= 1 \\ '' &= -2 & & \end{aligned}$$

$$= 1+x-1+9/4+2x+2$$

$$= \underline{17/4+3x}, \quad u_1 = 1$$

$$J_0(x) = \max_u (1+x-u^2) + 17/4 + 3(x+u)$$

$$\begin{aligned} ' &= -2u+3=0 & u &= 3/2 \\ '' &= -2 & & \end{aligned}$$

$$= 1+x-9/4+17/4+3x+9/2 = \underline{4x+15/2} \quad \text{ok } u_0 = 3/2$$

With $x_0=0$:

$$u_0 = 3/2 \quad x_0 = 0 \quad J_0(x) = \underline{4x+15/2} = \underline{\underline{15/2}}$$

$$u_1 = 1 \quad x_1 = 3/2$$

$$u_2 = 1/2 \quad x_2 = 5/2$$

$$u_3 = 0 \quad x_3 = 3$$

Alternative solution methods

B) Euler equation

$$\max \sum_{t=0}^T F(t, x_t, x_{t+1}) \quad \text{subject to} \quad \begin{cases} x_0 \text{ given} \\ x_1, \dots, x_{T+1} \in U = \mathbb{R} \text{ free} \end{cases}$$

- special case: $x_{t+1} = u_t$, $g(t, x_t, u_t) = u_t$, $U = \mathbb{R}$

- many problems are covered by this via transformation

Euler equation:

If $x_0^*, x_1^*, \dots, x_{T+1}^*$ is optimal solution, then it satisfies the difference equations

$$\begin{aligned} F_2'(t, x_t, x_{t+1}) + F_3'(t+1, x_{t+1}, x_t) &= 0 & \text{for } t=1, 2, \dots, T \\ F_3'(t+1, x_{t+1}, x_t) &= 0 & t=T+1 \end{aligned}$$

where F_2', F_3' are the partial derivatives of F with respect to variable 2 and 3.

Ex: $\max \sum_{t=0}^{T-1} \ln(x_t - \beta x_{t+1}) + \ln x_T$

$$\left\{ \begin{aligned} F_2'(t, x_t, x_{t+1}) + F_3'(t+1, x_{t+1}, x_t) &= 0 & t=1, \dots, T \\ \frac{1}{x_t - \beta x_{t+1}} - \frac{\beta}{x_{t-1} - \beta x_t} &= 0 \\ x_{t-1} - \beta x_t &= \beta (x_t - \beta x_{t+1}) \\ \beta^2 x_{t+1} - 2\beta x_t + x_{t-1} &= 0 & \text{(second order difference eqn.)} \end{aligned} \right.$$

$$\left\{ \begin{aligned} F_3'(T, x_T, x_{T+1}) &= 0 \\ 0 &= 0 & \text{(since } F(T, x_T, x_{T+1}) = \ln(x_T)) \\ \text{(no condition)} & \end{aligned} \right.$$

C) Maximum principle and Hamiltonian

$$\max \sum_{t=0}^T f(t, x_t, u_t) \quad \text{s.t.} \quad \begin{cases} x_0 \text{ given} \\ x_{t+1} = g(t, x_t, u_t) \\ x_T \text{ free} \\ u_t \in U \subseteq \mathbb{R} \end{cases}$$

Hamiltonian: $H = \begin{cases} f(t, x, u) + p g(t, x, u) & , t < T \\ f(t, x, u) & , t = T \end{cases}$

Maximum principle (necessary conditions)

If (x_t^*, u_t^*) is optimal, then there is a sequence $p_t, t=0, 1, \dots, T$ with

1) $H'_u(t, x_t^*, u_t^*, p_t) \cdot (u - u_t^*) \leq 0$ for all $u \in U$

2) $p_{t-1} = H'_x(t, x_t^*, u_t^*, p_t)$, $t=1, 2, \dots, T$

3) $p_T = 0$

Sufficient conditions

Suppose that (x_t^*, u_t^*) satisfy the conditions above, and that in addition

$H(t, x, u, p)$ is concave in (x, u) for all t

Then (x_t^*, u_t^*) is optimal.

② Infinite Horizon dynamic programming

This will be explained in lecture 10.

$$\max \sum_{t=0}^{\infty} \beta^t f(x_t, u_t) \quad \text{subject to} \quad \begin{cases} x_0 \text{ given} \\ x_{t+1} = g(x_t, u_t) \\ u_t \in U \subseteq \mathbb{R} \end{cases}$$

Note: * instead of $t \in T$ we have infinite horizon $t \rightarrow \infty$

* $\beta^t f(x_t, u_t) = f(t, x_t, u_t)$; $g(x_t, u_t) = g(t, x_t, u_t)$

($0 < \beta < 1$ is the one-period discount factor)

* we assume that $f(x_t, u_t)$ is bounded, i.e. $|f(x_t, u_t)| < M$ for all t for some number $M > 0$. This implies that the sum

$$\sum_{t=0}^{\infty} \beta^t f(x_t, u_t) \leq \sum_{t=0}^{\infty} \beta^t M = \frac{M}{1-\beta} \quad \text{is finite}$$

Bellman equation: Let $J(x) = J_0(x) = \max_{(u)} \sum_{t=0}^{\infty} \beta^t f(x_t, u_t)$ sub to ... $\begin{cases} x_0 = x \\ \vdots \end{cases}$

$$J(x) = \max_{u \in U} \left\{ f(x, u) + \beta J(g(x, u)) \right\}$$

- Functional equations: we want to solve for the function $J(x)$.

Difficult to find max when $J(x)$ is not known, hence difficult to find $J(x)$ by solving the Bellman equation.

- When $0 < \beta < 1$ and $|f(x_t, u_t)| < M$, the Bellman equation has a unique bounded solution $J^*(x)$. If we "guess" $J(x)$ and it fits in the equation, it is therefore the unique solution.

Ex: Problem 12.3.1 in [FMEAT]. See also (SM) Student Manual (online)