

# Lecture 2

## Continuous Time Dynamic Optimization Problem

$$\max_{\{x, u\}} \int_0^{\infty} e^{-r \cdot t} \cdot f(x, u) dt$$

$$\text{s.t. } \lambda \text{ or } m: x' = g(x, u) \quad \left[ x' \equiv \frac{dx}{dt} \right]$$

$x(0)$  is known  
Initial condition

$x \rightarrow$  state

$u \rightarrow$  control

$\lambda \rightarrow$  multiplier (present value)  
(costate variable)  
or  
 $m \rightarrow$  multiplier (current value)

③  $\frac{\partial \bar{H}}{\partial m} = x' \Rightarrow x' = g(x, u)$  (the constraint)

## I. Optimal Control [Hamiltonian] ①

Present Value Hamiltonian:  $H = e^{-r \cdot t} \cdot f(x, u) + \lambda \cdot g(x, u)$

or  
Current " " ;

General Definition of the Hamiltonian

$$\bar{H} = f(x, u) + m \cdot g(x, u)$$

$$\begin{aligned} \bar{H} &= e^{r \cdot t} \cdot H = \\ &= e^{r \cdot t} \cdot [e^{-r \cdot t} \cdot f(x, u) + \lambda \cdot g(x, u)] = \\ &= f(x, u) + e^{r \cdot t} \cdot \lambda \cdot g(x, u) \end{aligned}$$

$$\begin{aligned} \max \int_0^{\infty} F(x, u) dt \\ \text{s.t. } x' = g(x, u) \\ \downarrow \\ H = F(x, u) + \lambda \cdot g(x, u) \end{aligned}$$

### FOC

(CV Hamiltonian)

$$m \equiv e^{r \cdot t} \cdot \lambda$$

$$\bar{H} = f(x, u) + m \cdot g(x, u)$$

$$\frac{\partial \bar{H}}{\partial u} (=), \frac{\partial f(x, u)}{\partial u} + m \cdot \frac{\partial g(x, u)}{\partial u} = 0$$

②  $m' = r \cdot m - \frac{\partial \bar{H}}{\partial x}$   
 $(=), m' = r \cdot m - \left[ \frac{\partial f(x, u)}{\partial x} + m \cdot \frac{\partial g(x, u)}{\partial x} \right]$   
 solve a dynamic system wrt  $x, u, m$



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Continuous Time  
Dynamic Optimization  
Problem

$$\max_{\{x, u\}} \int_0^{\infty} e^{-r \cdot t} \cdot f(x, u) dt$$

$$\text{s.t. } x' = g(x, u) \quad \left[ x' \equiv \frac{dx}{dt} \right]$$

$x(0)$  is known  
(initial condition)

$x \rightarrow$  state  
 $u \rightarrow$  control

$$\text{Define } V(x_0) \equiv \max_{\{x, u\}} \int_0^{\infty} e^{-r \cdot (t-t_0)} \cdot f(x, u) dt$$

$$J(t_0, x_0) \equiv e^{-r \cdot t_0} V(x_0)$$

$$\frac{\partial J(t_0, x_0)}{\partial t} = -r \cdot e^{-r \cdot t_0} V(x_0)$$

$$\frac{\partial J(t_0, x_0)}{\partial x} = e^{-r \cdot t_0} \cdot V'(x_0)$$

## II. Dynamic Programming (Bellman Equation) (2)

Define  $J(t_0, x_0)$  ★ the optimal value function starting at time  $t_0$ , in state  $x_0$  : Hamilton - Jacobi - Bellman Equation

$$J(t_0, x_0) \equiv \max_{\{x, u\}} \int_{t_0}^{\infty} e^{-r \cdot t} \cdot f(x, u) dt$$

$$\text{s.t. } x' = g(x, u) \\ x(t_0) = x_0, \text{ known}$$

$$J(t_0, x_0) \equiv \max_{\{x, u\}} \int_{t_0}^{\infty} e^{-r \cdot t + r \cdot t_0 - r \cdot t_0} \cdot f(x, u) dt =$$

$$= J(t_0, x_0) \equiv \max_{\{x, u\}} e^{-r \cdot t_0} \int_{t_0}^{\infty} e^{-r \cdot (t-t_0)} \cdot f(x, u) dt = V(x_0)$$

It can be shown that  $-\frac{\partial J}{\partial t} \approx \max_u [e^{-r \cdot t} \cdot f(x, u) + e^{-r \cdot t} \cdot V'(x) \cdot g(x, u)]$

$$\Rightarrow -e^{r \cdot t} \frac{\partial J}{\partial t} = r \cdot V(x) = \max_u [f(x, u) + V'(x) \cdot g(x, u)]$$



# Lecture 2

## II. Dynamic Programming (Bellman Equation) (3)

Hamilton-Jacobi-Bellman equation

$$r \cdot V(x) = \max_u [f(x,u) + V'(x) \cdot g(x,u)]$$

F.O.C.

$$u: 0 = \frac{\partial f(x,u)}{\partial u} + V'(x) \cdot \frac{\partial g(x,u)}{\partial u}$$

Envelope or Benveniste-Scheinkman Condition

$$r \cdot V'(x) = \frac{\partial f(x,u)}{\partial x} + v''(x) \cdot g(x,u) + V'(x) \cdot \frac{\partial g(x,u)}{\partial x}$$

(+) constraint  $x' = g(x,u)$

Define  $V(x_0) \equiv \max_{to} \int_{to}^{\infty} e^{-r \cdot (t-to)} \cdot f(x,u) dt$

$$J(t_0, x_0) \equiv e^{-r \cdot t_0} V(x_0)$$

$$\frac{\partial J(t_0, x_0)}{\partial t} = -r \cdot e^{-r \cdot t_0} V(x_0)$$

$$\frac{\partial J(t_0, x_0)}{\partial x} = e^{-r \cdot t_0} \cdot V'(x_0)$$

Define  $J(t_0, x_0)$  the optimal value function starting at time  $t_0$ , in state  $x_0$ :

$$J(t_0, x_0) \equiv \max_{to} \int_{to}^{\infty} e^{-r \cdot t} \cdot f(x,u) dt$$

st  $x' = g(x,u)$   
 $x(t_0) = x_0$ , known

$$J(t_0, x_0) \equiv \max_{to} \int_{to}^{\infty} e^{-r \cdot t + r \cdot t_0 - r \cdot t_0} \cdot f(x,u) dt =$$

$$J(t_0, x_0) \equiv \max_{to} e^{-r \cdot t_0} \int_{to}^{\infty} e^{-r \cdot (t-to)} \cdot f(x,u) dt = V(x_0)$$

It can be shown that  $-\frac{\partial J}{\partial t} \approx \max_u [e^{-r \cdot t} \cdot f(x,u) + e^{-r \cdot t} \cdot V'(x) \cdot g(x,u)]$

$$\Rightarrow -e^{r \cdot t} \cdot \frac{\partial J}{\partial t} = r \cdot V(x) = \max_u [f(x,u) + V'(x) \cdot g(x,u)]$$

★ Hamilton-Jacobi-Bellman Equation



# Lecture 2

## Optimal Control

[Pontryagin]

General Optimal Control Problem

$T \rightarrow$  Time  
Horizon

$$\max_u \int_0^T f(t, x, u) dt$$

$$\text{s.t. } x' = g(t, x, u)$$

$x(0) = x_0$  initial condition  
and/or

$x(T) \geq x_T$  terminal condition

$x \rightarrow$  state variables

$u \rightarrow$  control "

initial  
endowment

## Consumer Model

(4)

$\hookrightarrow$  Lives  $\rightarrow [0, T]$

Stock of savings  $S(t)$  at time  $t$

Some of  $S(t)$  can be consumed  $c(t)$

The remaining is saved, gaining a rate  $r$   
Thus  $\rightarrow$  Budget Constraint

$$S'(t) = r \cdot S(t) - c(t)$$

$$S' = r \cdot S - c$$

temporal  
utility

Initial Savings  $S_0$

Life time  
utility

$$\max_{\{c\}} \int_0^T e^{-\rho \cdot t} \cdot U(c) dt$$

$$S' = r \cdot S - c$$

$$S(0) = S_0$$

$$S(T) \geq 0$$

$\rho$  the rate of  
time  
preference

sustainability condition  
[cannot owe in  $T$ ]

# Lecture 2

# Sketch of the Solution Method (5)

## Optimal Control

[Pontryagin]

## General Optimal Control Problem

$T \rightarrow$  Time Horizon

$$\max_u \int_0^T f(t, x, u) dt$$

$$\text{s.t. } x' = g(t, x, u)$$

$x(0) = x_0$  initial condition  
and/or

$x(T) \Rightarrow \bar{x}$  terminal condition

$x \rightarrow$  state variables

$u \rightarrow$  control "

Constrained Optimization problem

1) Terminal condition (1)

2) Dynamic constraint (differential equation) ( $\infty$ )

$$\int_0^T \lambda \cdot g(t, x, u) dt - \int_0^T \lambda \cdot x' dt$$

$$\mathcal{L} = \int_0^T f(t, x, u) dt + \int_0^T \lambda \cdot [g(t, x, u) - x'] dt + \mu \cdot [x(T) - \bar{x}] \quad (1)$$

multiplier

co-state variables

Integrate the term  $-\int_0^T \lambda \cdot x' dt$  in  $\mathcal{L}$  by parts

$$-\int_0^T \lambda \cdot x' dt = -[\lambda(T) \cdot x(T) - \lambda(0) \cdot x(0)] + \int_0^T \lambda' \cdot x dt \quad (2)$$

Substitute (2) in (1)



# Lecture 2

# Sketch of the Solution Method ⑥

## Optimal Control

Observe that

If I define

$$\mathcal{H} = f(t, x, u) + \lambda \cdot g(t, x, u)$$

the FOC (3) + (4)

can be written in terms of derivatives of  $\mathcal{H}$ :

$$\textcircled{3} \rightarrow \frac{\partial \mathcal{H}}{\partial u} = 0$$

$$\textcircled{4} \rightarrow \frac{\partial \mathcal{H}}{\partial x} = -\lambda'$$

$$x' = \frac{\partial \mathcal{H}}{\partial x}$$

$$\lambda(T) \cdot [x(T) - \bar{x}] = 0$$

Hamiltonian

② in ①

$$\mathcal{L} = \int_0^T f(t, x, u) dt + \int_0^T \lambda \cdot g(t, x, u) - [\lambda(T) \cdot x(T) - \lambda(0) \cdot x(0)] + \int_0^T \lambda' \cdot x dt + \mu \cdot [x(T) - \bar{x}]$$

$$\mathcal{L} = \int_0^T \{ f(t, x, u) + \lambda \cdot g(t, x, u) + \lambda' \cdot x \} dt + \mu \cdot [x(T) - \bar{x}] - [\lambda(T) \cdot x(T) - \lambda(0) \cdot x(0)]$$

To maximize  $\mathcal{L}$  by choosing  $x, u$  all it is necessary to do is to maximize

$$\{ f(t, x, u) + \lambda \cdot g(t, x, u) + \lambda' \cdot x \}$$

FOC  $\rightarrow$  { }

$$\frac{\partial \{ \}}{\partial u} = 0 \Rightarrow \frac{\partial f}{\partial u} + \lambda \cdot \frac{\partial g}{\partial u} = 0, \forall t \quad \textcircled{3}$$

$$\frac{\partial \{ \}}{\partial x} = 0 \Rightarrow \frac{\partial f}{\partial x} + \lambda \cdot \frac{\partial g}{\partial x} + \lambda' = 0, \forall t \quad \textcircled{4}$$

$\rightarrow$  Kuhn-Tucker complementarity condition

⊕ Constraints ⊕  $\frac{\partial \{ \}}{\partial \mu} = 0$  + Complementary slackness condition