

# Mathematics III

## Problem Set 3: Dynamic Optimization I

### Suggested Solutions

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Deadline is *Mon 11 December at 14:00*. Submission via email: jose.elias.gallegos@iies.su.se. By that same time, I will upload solutions to my webpage, [www.joseeliasgallegos.com](http://www.joseeliasgallegos.com). I suggest you to have a look at them before coming to class.

#### Exercise 1: Bang-Bang Control

Consider the following dynamic optimization problem,

$$\begin{aligned} \max_{u \in [-1,1]} \int_0^1 (2x - x^2) dt \\ \text{s.t. } \dot{x} = u \\ x(0) = 0 \\ x(1) = 0 \end{aligned}$$

(a) Write down the conditions of the maximum principle. Are these conditions sufficient?

*Answer:*

Pontryagin's Maximum Principle provides necessary conditions for the optimality of an admissible pair to the standard problem.

**Theorem A.** *Suppose  $(\mathbf{x}^*(t), \mathbf{u}^*(t))$  solves the standard problem. Then, there exists a continuous and piece-wise differentiable function  $\lambda(t)$  and a number  $\lambda_0 \in \{0, 1\}$  such that for all  $t \in [0, T]$ ,  $(\lambda_0, \lambda(t)) \neq (0, \mathbf{0})$ , and*

- $\mathbf{u}^*(t)$  maximizes  $\mathcal{H} \forall \mathbf{u} \in \mathcal{U}$ ,

$$\mathcal{H}(t, \mathbf{x}^*(t), \mathbf{u}^*(t), \lambda(t)) \geq \mathcal{H}(t, \mathbf{x}^*(t), \mathbf{u}(t), \lambda(t)) \quad \forall \mathbf{u} \in \mathcal{U}$$

- Whenever  $\mathbf{u}^*(t)$  is continuous,

$$\dot{\lambda}(t) = -\frac{\partial \mathcal{H}}{\partial \mathbf{x}(t)}$$

• Corresponding to the terminal condition, a transversality condition holds:

- $\lambda(T) = \mathbf{0}$  ( $x(T)$  free)
- $\lambda(T) \geq \mathbf{0}$  ( $\mathbf{x}(T) \geq \mathbf{x}_1$ )  
 $\lambda(T)[\mathbf{x}(T) - \mathbf{x}_1] = \mathbf{0}$
- $\mathbf{x}(T) = \mathbf{x}_1$

(b) Show that  $\lambda(t)$  is decreasing

*Answer:*

Form the Hamiltonian

$$\mathcal{H} = 2x(t) - x(t)^2 + \lambda(t)u(t)$$

hence,

$$u^*(t) = \begin{cases} 1 & \text{if } \lambda(t) > 0 \\ -1 & \text{if } \lambda(t) < 0 \end{cases}$$

We know from the maximum principle that  $\dot{\lambda}(t) = -\frac{\partial \mathcal{H}}{\partial x(t)}$ . Hence,

$$\dot{\lambda}(t) = -[2 - 2x(t)] = 2[x(t) - 1]$$

Notice that  $\dot{x}(t) = u^*(t)$ . Hence,  $\dot{x}(t) \leq 1$ ,

$$\frac{dx}{dt} \leq 1 \implies dx \leq dt \implies \int dx \leq \int dt \implies x(t) + C \leq t + D \implies x(t) \leq t + E$$

We need to pin down  $E$ . Notice that the above expression can hold with equality, and that  $x(0) = 0$ . Hence,  $0 = 0 + E \implies E = 0$ . As a result,  $x(t) \leq t$ . Therefore,  $x(t) \leq 1 \forall t \in [0, 1)$ , and  $x(T) = x(1) = 0$  by the text.

[It is also easily seen if you take  $dt = t - 0 = t$  and  $dx = x(t) - x(0) = x(t)$ ]

Finally,

$$\dot{\lambda}(t) = 2[x(t) - 1] < 0 \quad \forall t \in [0, 1]$$

So we have proved that  $\lambda(t)$  is strictly decreasing in  $[0, 1]$ !

(c) Suppose that  $\lambda(t)$  has a unique crossing with the zero-line at  $\hat{t} = 0.5$ . Use this to derive  $u^*(t)$ ,  $\lambda(t)$  and  $x^*(t)$ .

*Answer:*

Thanks to the hint  $\lambda(\hat{t}) = 0$  and  $\lambda(t)$  decreasing, we know that  $\lambda(t) > 0$  in  $[0, \hat{t}]$  and  $\lambda(t) < 0$  in  $[\hat{t}, T] = [\hat{t}, 1]$ . Lets do the two cases

- $[0, \hat{t}]$ : in this case  $u^*(t) = \dot{x}^*(t) = 1$ . This implies that  $x(t) = t + C$ . We need to pin down  $C$ : since  $x(0) = 0 \implies 0 = 0 + C \implies C = 0$ , and therefore  $x^*(t) = t$ . Finally

$$\begin{aligned}\dot{\lambda}(t) &= 2[x(t) - 1] = 2(t - 1) \\ d\lambda &= 2(t - 1)dt \\ \int d\lambda &= 2 \int (t - 1)dt \\ \lambda(t) &= t(t - 2) + D\end{aligned}$$

We need to pin down  $D$ . We know that  $\lambda(\hat{t}) = \lambda(0.5) = 0 \implies 0 = 0.5(0.5 - 2) + D \implies D = \frac{3}{4}$ . Hence,  $\lambda(t) = t(t - 2) + \frac{3}{4}$ .

- $[\hat{t}, T] = [\hat{t}, 1]$ : in this case  $u^*(t) = \dot{x}^*(t) = -1$ . This implies that  $x(t) = -t + C$ . We need to pin down  $C$ : since  $x(T) = x(1) = 0 \implies 0 = -1 + C \implies C = 1$ , and therefore  $x^*(t) = 1 - t$ . Finally

$$\begin{aligned}\dot{\lambda}(t) &= 2[x(t) - 1] = -2t \\ d\lambda &= -2tdt \\ \int d\lambda &= -2 \int tdt \\ \lambda(t) &= -t^2 + D\end{aligned}$$

We need to pin down  $D$ . We know that  $\lambda(\hat{t}) = \lambda(0.5) = 0 \implies 0 = -(0.5)^2 + D \implies D = \frac{1}{4}$ . Hence,  $\lambda(t) = -t^2 + \frac{1}{4}$ .

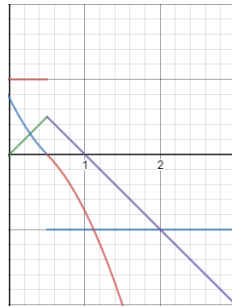


Figure 1:  $u^*$ : strong red and strong blue,  $x^*$ : green and purple,  $\lambda^*$ : light blue and light red

## Exercise 2: A Distance Problem

Suppose  $x_0 < x_1 < x_0 + T$ . Solve the problem

$$\begin{aligned} \max_{u \in [0,1]} \int_0^T x \, dt \\ \text{s.t. } \dot{x} &= u \\ x(0) &= x_0 \\ x(T) &= x_1 \end{aligned}$$

What classes of problems does this problem attempt to solve? Discuss.

*Answer:*

A cake-eating problem. Let us now solve it. Form the Hamiltonian

$$\mathcal{H} = x(t) + \lambda(t)u(t)$$

hence,

$$u^*(t) = \begin{cases} 1 & \text{if } \lambda(t) > 0 \\ 0 & \text{if } \lambda(t) < 0 \end{cases}$$

By the maximum principle we know that  $\dot{\lambda}(t) = -\frac{\partial \mathcal{H}}{\partial x(t)}$ . Hence,

$$\dot{\lambda}(t) = -1 \implies d\lambda = -dt \implies \lambda(t) = -t + C$$

We know that  $\lambda(t)$  is decreasing, and we know  $u(t)$  in terms of  $\lambda(t)$ . Let us now guess how does  $\lambda(t)$  behave

- Suppose  $\lambda(t) > 0 \forall t$ . This implies  $u^*(t) = 1 = \dot{x}(t)$ , which in turn gives  $x(t) = t + D$ . We need to pin down  $D$ ,

$$\begin{aligned} x(0) = x_0 = D &\implies x(t) = t + x_0 \\ x(T) = x_1 &\implies D = x_1 - T \implies x(t) = t + x_1 - T \end{aligned}$$

which is not consistent since  $x_0 \neq x_1 - T$ .

- Suppose  $\lambda(t) < 0 \forall t$ . This implies  $u^*(t) = 0 = \dot{x}(t)$ , which in turn gives  $x(t) = D$ . We need to pin down  $D$ ,

$$x(0) = x_0 = D \implies x(t) = x_0$$

$$x(T) = x_1 = D \implies x(t) = x_1$$

which is not consistent since  $x_0 \neq x_1$ .

- Hence, it must be that  $\lambda(t)$  is sometimes positive and sometimes negative. Since we know that it is decreasing, it must be positive in the beginning and negative later:  $\lambda(t) > 0$  in  $[0, \hat{t}]$  and  $\lambda(t) < 0$  in  $[\hat{t}, T]$

- $[0, \hat{t}]$ : since  $\lambda(t) > 0$ ,  $u^*(t) = 1 = \dot{x}(t)$ , which in turn gives  $x(t) = t + D$ . We need to pin down  $D$ ,

$$x(0) = x_0 = D \implies x^*(t) = t + x_0$$

and we need to find  $\lambda(t)$ . We need to pin down  $C$ . We know that  $\lambda(\hat{t}) = 0$ . Hence,  $0 = -\hat{t} + C \implies \hat{t} = C$ , so that  $\lambda^*(t) = -t + \hat{t}$ .

- $[\hat{t}, T]$ : since  $\lambda(t) < 0$ ,  $u^*(t) = 0 = \dot{x}(t)$ , which in turn gives  $x(t) = D$ . We need to pin down  $D$ ,

$$x(T) = x_1 = D \implies x^*(t) = x_1$$

and we need to find  $\lambda(t)$ . How can we pin down  $\hat{t}$  (that is, the critical point)? Well, it seems reasonable that the state  $x^*(t)$  is the same in both cases at  $x^*(\hat{t})$  (if not, there would be a discontinuity in the state function...that cannot happen!) Hence,

$$\underbrace{x^*(\hat{t})}_{t \in [0, \hat{t}]} = \underbrace{x^*(\hat{t})}_{t \in [\hat{t}, T]} \implies \hat{t} + x_0 = x_1 \implies \hat{t} = x_1 - x_0$$

Therefore,  $\lambda^*(t) = -t + x_1 - x_0$ .

### Exercise 3: Growth in Consumption

Suppose a person seeks to maximize,

$$\max_{x(t)} \int_0^{\infty} u(c) e^{-\delta t} dt \quad s.t. \quad \dot{k} = b(1-x)k, \quad x \in [0, 1]$$

where  $c = f(xk)$ ,  $f(0) = 0$ ,  $f'(\cdot) > 0$  and  $f''(\cdot) < 0$ .

(a) Derive the sufficient first-order conditions associated with the maximization problem.

*Answer:*

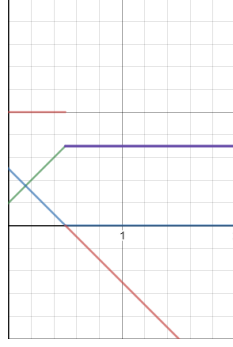


Figure 2:  $u^*$ : strong red and strong blue,  $x^*$ : green and purple,  $\lambda^*$ : light blue and light red (using  $x_0 = 0.2$  and  $x_1 = 0.7$ ).

(a) Forming the Hamiltonian,

$$\mathcal{H} = u\{f[x(t)k(t)]\}e^{-\delta t} + \lambda(t)\{b[1 - x(t)]k(t)\}$$

Taking the FOC's,

$$\frac{\partial \mathcal{H}}{\partial x(t)} \stackrel{!}{=} 0 \implies \lambda(t) = \frac{u'\{f[x(t)k(t)]\}f_x[x(t)k(t)]k(t)e^{-\delta t}}{b}$$

$$\frac{\partial \mathcal{H}}{\partial k(t)} \stackrel{!}{=} -\dot{\lambda}(t) \implies u'\{f[x(t)k(t)]\}f_k[x(t)k(t)]x(t)e^{-\delta t} + \lambda(t)b[1 - x(t)] = -\dot{\lambda}(t)$$

$$\lim_{t \rightarrow \infty} \lambda(t)k(t) = 0$$

(b) Assume that  $f(xk) = A(xk)^\alpha$  where  $\alpha \in (0, 1)$  and that  $u(c) = \log c$ . Show that on the optimal path the growth of  $c$  is given by  $\alpha(b - \delta)$ . When does  $c$  grow forever?

*Answer:*

(b) Let me now be sloppy and don't write everything in terms of  $(t)$ . Expressions will be more transparent. Forming the Hamiltonian,

$$\mathcal{H} = \log[A(xk)^\alpha]e^{-\delta t} + \lambda b(1 - x)k$$

Taking the FOC's,

$$\frac{\partial \mathcal{H}}{\partial x(t)} \stackrel{!}{=} 0 \implies \lambda b k = \frac{\alpha}{x} e^{-\delta t} \tag{1}$$

$$\frac{\partial \mathcal{H}}{\partial k(t)} \stackrel{!}{=} -\dot{\lambda} \implies \dot{\lambda} = -\frac{\alpha}{k} e^{-\delta t} - \lambda b(1 - x) \tag{2}$$

$$\lim_{t \rightarrow \infty} \lambda k = 0$$

we want to obtain  $c = A(xk)^\alpha$ . Hence, let's find  $xk$ . Rearranging (1),

$$xk = \frac{\alpha}{\lambda b} e^{-\delta t} \quad (3)$$

we need  $\lambda$  now. Rearranging again (1),

$$\lambda b x = \frac{\alpha}{k} e^{-\delta t} \quad (4)$$

Introducing (4) into (2),

$$\dot{\lambda} = -\lambda b$$

Hence, multiplying both sides by  $dt$  and integrating,

$$\begin{aligned} \frac{\dot{\lambda}}{\lambda} = -b &\implies \int_0^t \frac{1}{\lambda} \frac{d\lambda}{dt} dt = - \int_0^t b dt \implies \\ &\implies [\log \lambda]_0^t = -[bs]_0^t \implies \\ &\implies \log \lambda(t) = \log \lambda(0) - bt \implies \\ &\implies \lambda(t) = \lambda(0)e^{-bt} \end{aligned} \quad (5)$$

where I have written  $\lambda$  in terms of  $t$  just to clarify the result of the integral. From now on, I will denote  $\lambda(0) = \lambda_0$ . Introducing (5) into (3),

$$xk = \frac{\alpha}{\lambda_0 b} e^{(b-\delta)t}$$

Therefore,

$$c(t) = A(xk)^\alpha = A \left( \frac{\alpha}{\lambda_0 b} \right)^\alpha e^{\alpha(b-\delta)t}$$

Finally, in continuous time we define growth as

$$g_c = \frac{\dot{c}}{c} = \frac{1}{c} \frac{dc}{dt} \underset{\text{elasticities}}{\equiv} \frac{d \log c}{dt}$$

Taking therefore the log (it just simplifies calculations, you could also do it without taking the log),

$$\log c = \alpha \log \frac{A\alpha}{\lambda_0 b} + \alpha(b - \delta)t$$

and we can get the growth rate

$$g_c = \frac{d \log c}{dt} = \alpha(b - \delta)$$

hence,  $g_c > 0$  if  $b > \delta$ .

## Exercise 4: Autonomous Systems

Consider the optimal control problem

$$\max_{x(t)} \int_0^1 \left( -\frac{1}{2}u^2 - x \right) dt \quad \text{s.t.} \quad \dot{x} = 2(1 - u), \quad x(0) = 1$$

where  $x(1)$  is free.

(a) State the necessary and sufficient conditions for a candidate solution

*Answer:*

(a) **Necessary:** if there exists a continuous and piecewise differentiable function  $\lambda(t)$  s.t.  $\forall t \in [0, 1]$ ,

1.  $u^*(t)$  maximizes  $\mathcal{H} \forall u \in \mathcal{U}$ ,

$$\mathcal{H}[t, x^*(t), u^*(t), \lambda(t)] \geq \mathcal{H}[t, x(t), u(t), \lambda(t)]$$

2. Whenever  $u^*(t)$  is differentiable,  $\dot{\lambda} = -\frac{\partial \mathcal{H}}{\partial x}$ .
3. One of the transversality conditions hold,

- $\lambda(1) = 0$  ( $x_1$  free)
- $\lambda(1) \geq 0; \lambda(1)[x(1) - x_1] = 0$  ( $x(1) \geq x_1$ )
- $x(1) = x_1$

**Sufficient:** suppose  $[t, x^*(t), u^*(t), \lambda(t)]$  satisfies the above necessary conditions. Moreover, suppose that  $\mathcal{H}[t, x(t), u(t), \lambda(t)]$  is concave in  $(x, u) \forall t \in [0, 1]$ . Then,  $[t, x^*(t), u^*(t), \lambda(t)]$  is optimal.

(b) Find  $u^*(t)$ ,  $x^*(t)$  and  $\lambda^*(t)$  using these conditions.

*Answer:*

- (b) Forming the hamiltonian,

$$\mathcal{H} = -\frac{1}{2}u^2 - x + \lambda 2(1 - u)$$

Before starting the analysis, lets check if the hamiltonian is concave,



$$\begin{aligned}
\frac{\partial \mathcal{H}}{\partial u} &= -u - \lambda 2 \\
\frac{\partial^2 \mathcal{H}}{\partial u^2} &= -1 \\
\frac{\partial \mathcal{H}}{\partial x} &= -1 \\
\frac{\partial^2 \mathcal{H}}{\partial x^2} &= 0 \\
\frac{\partial^2 \mathcal{H}}{\partial u \partial x} &= \frac{\partial^2 \mathcal{H}}{\partial x \partial u} = 0
\end{aligned}$$

Hence, it is concave! Again, taking the FOC of the hamiltonian,

$$\begin{aligned}
\frac{\partial \mathcal{H}}{\partial u} \stackrel{!}{=} 0 &\implies u = -\lambda 2 \\
\frac{\partial \mathcal{H}}{\partial x} \stackrel{!}{=} -\dot{\lambda} &\implies \dot{\lambda} = 1
\end{aligned} \tag{6}$$

Hence,  $\lambda = t + C$ . Since we know that  $\lambda(T) = \lambda(1) = 0$ ,

$$\lambda(1) = 1 + C = 0 \implies C = -1$$

and  $\lambda^*(t) = t - 1$ . Introducing  $\lambda^*$  into (6) we obtain

$$u^*(t) = -2(t - 1)$$

From the law of motion (equivalent to resource constraint),

$$\dot{x} = 2(1 - u) = 2[1 + 2(t - 1)] = 2t - 1$$

and hence,  $x^*(t) = 2t(t - 1) + D$ . We know that  $x(0) = 1$ , hence

$$x(0) = 0 + D = 1 \implies D = 1$$

and  $x^*(t) = 2t(t - 1) + 1$ .

(c) Show that the Hamiltonian is constant along the optimal trajectory. Explain why.

*Answer:*

(c) Plugging in the values we obtain in (b),  $\mathcal{H}^* = -1$ . Let me explain why,

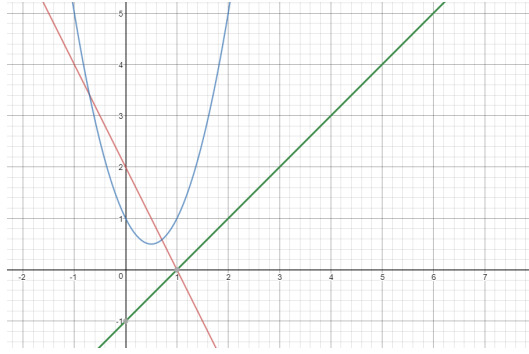


Figure 3:  $u^*$ : red,  $x^*$ : blue,  $\lambda^*$ : green.

$$\begin{aligned}
 \dot{\mathcal{H}} &= \frac{d\mathcal{H}[t, x(t), u(t), \lambda(t)]}{dt} = \\
 &= \frac{d\mathcal{H}}{dt} + \underbrace{\frac{d\mathcal{H}}{du}}_{=0 \text{ if optimum is interior}} \dot{u} + \frac{d\mathcal{H}}{dt} \dot{x} + \frac{d\mathcal{H}}{dt} \underbrace{\dot{\lambda}}_{-\frac{d\mathcal{H}}{dx}} = \\
 &= \frac{d\mathcal{H}}{dt} + \frac{d\mathcal{H}}{dt} \left( \dot{x} - \underbrace{\frac{d\mathcal{H}}{d\lambda}}_{\dot{x}} \right) = \\
 &= \frac{d\mathcal{H}}{dt} = 0
 \end{aligned}$$

Hence, if  $\mathcal{H}$  does not depend directly on  $t$ , then  $\mathcal{H} \equiv cte$  along optimal trajectory.