

Variational Methods & Optimal Control

lecture 22

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More Optimal Control Examples

First we'll cover a bit more terminology, and then some examples primarily focussed on planned growth strategies in economics.

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Formulation of control problems

We break a control problem into two parts

- ▶ **The system state:** $\mathbf{x}(t) = (x_1(t), \dots, x_n(t))^t$

The system state describes the system (e.g. position and velocity of the car in car parking example)

- ▶ **The control:** $\mathbf{u}(t) = (u_1(t), \dots, u_m(t))^t$

We apply the control to the system (e.g. force applied to the car).

The evolution of the system is governed by the set of DEs

$$\dot{\mathbf{x}}(t) = \mathbf{g}(t, \mathbf{x}, \mathbf{u})$$

In a control problem we want to get the system to a particular state $\mathbf{x}(t_1)$ at time t_1 , given initial state $\mathbf{x}(t_0)$.

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Optimal control problems

In an **optimal** control problem we still have the system equations

$\dot{\mathbf{x}}(t) = \mathbf{g}(t, \mathbf{x}, \mathbf{u})$ and we might wish to get to state $\mathbf{x}(t_1)$ given initial state $\mathbf{x}(t_0)$, but now we wish to do so while minimizing a functional

$$F\{\mathbf{x}, \mathbf{u}\} = \int_{t_0}^{t_1} f(t, \mathbf{x}, \mathbf{u}) dt$$

That is, we wish to choose a function $\mathbf{u}(t)$ which minimizes the functional $F\{\mathbf{x}, \mathbf{u}\}$, while satisfying the end-point conditions $\mathbf{x}(t_0) = \mathbf{x}_0$ and $\mathbf{x}(t_1) = \mathbf{x}_1$, and the non-holonomic constraints $\dot{\mathbf{x}}(t) = \mathbf{g}(t, \mathbf{x}, \mathbf{u})$.

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Optimal control problems

Optimization functional

$$F\{\mathbf{x}, \mathbf{u}\} = \int_{t_0}^{t_1} f(t, \mathbf{x}, \mathbf{u}) dt$$

Note that

- ▶ $f(t, \mathbf{x}, \mathbf{u})$ has no dependence on $\dot{\mathbf{u}}$: this is typically because costs depend on the control, not how we change the control, but there might be counter-examples
- ▶ $f(t, \mathbf{x}, \mathbf{u})$ has no dependence on $\dot{\mathbf{x}}$: this is common in control problems, but not universal (we have seen at least one counter example).

System Terminology

- ▶ **linear**: the state equations are a set of linear DEs.
- ▶ **autonomous**: time doesn't appear explicitly in the state equations (e.g. in $g(\mathbf{x}, \mathbf{u})$, or $f(\mathbf{x}, \mathbf{u})$).
 - ▷ also called time-invariant
- ▶ **terminal cost**: the term $\phi(t_1, \mathbf{x}(t_1))$ is called the terminal cost.
- ▶ **controllable**: a solution to the control problem exists.
- ▶ **stable**: a stable equilibrium solution to the system DEs exists.
 - ▷ often we are interested in problems that are unstable, or we wouldn't really need a control

Terminal costs

Sometimes in optimal control we don't fix the end-point $\mathbf{x}(t_1)$, but rather we assign a cost $\phi(t_1, \mathbf{x}(t_1))$ to particular end-points.

So now we wish to choose a control $\mathbf{u}(t)$ which minimizes the functional

$$F\{\mathbf{x}, \mathbf{u}\} = \phi(t_1, \mathbf{x}(t_1)) + \int_{t_0}^{t_1} f(t, \mathbf{x}, \mathbf{u}) dt$$

while satisfying the single end-point condition $\mathbf{x}(t_0) = \mathbf{x}_0$, and the non-holonomic constraint $\dot{\mathbf{x}}(t) = \mathbf{g}(t, \mathbf{x}, \mathbf{u})$.

- ▶ $\phi(t_1, \mathbf{x}(t_1))$ is called the **terminal cost**.

Control Terminology

- ▶ control (driver or automatic)
 - ▷ **planned** (open loop)
 - ▷ **feedback** (closed loop) control depends on current state
- ▶ type of control
 - ▷ movement from A to B
 - ▷ continuous operations (maintain equilibrium)
- ▶ type of cost functional F
 - ▷ minimum time
 - ▷ minimum fuel
 - ▷ quadratic costs
- ▶ admissible controls
 - ▷ unbounded/bounded/bang-bang

Cost functional examples

- ▶ **minimum time:** choose the fastest possible control

$$F\{x, u\} = \int_{t_0}^{t_1} dt$$

- ▶ **minimum fuel:** fuel is expended by the controller, and we wish to minimize this

$$F\{x, u\} = \int_{t_0}^{t_1} |u(t)| dt$$

- ▶ **quadratic costs:**

$$F\{x, u\} = \int_{t_0}^{t_1} x^2(t) + \alpha u^2(t) dt$$

Boundary conditions

- ▶ End time t_1 : can be fixed or free
- ▶ End position $\mathbf{x}(t_1)$: can be fixed or free

In the cases with free boundary conditions, we introduce natural, or transversal boundary conditions.

Example: dynamic production

- ▶ A producer in purely competitive market
 - ▷ A large numbers of independent producers
 - ▷ Standardized product, e.g. potatoes
 - ▷ Firms are "price takers", i.e. they have no significant control over product price
 - ▷ Free entry and exit
 - ▷ Free flow of information
- ▶ wants to find optimal production path $x(t)$, $0 \leq t \leq T$.
- ▶ production target $x(T) = x_T$
- ▶ profit at time t is $\pi(x, \dot{x}, t)$
- ▶ maximize profit functional $F\{x\} = \int_0^T \pi(x, \dot{x}, t) dt$

Example: dynamic production

Profit calculation

- ▶ quadratic production costs $C_1 = a_1 x^2 + b_1 x + c_1$
 - ▷ labor
 - ▷ raw materials
- ▶ production increase costs $C_2 = a_2 \dot{x}^2 + b_2 \dot{x} + c_2$
 - ▷ new buildings
 - ▷ recruiting and training costs
- ▶ revenue $r = px$ where p is the constant price per unit
 - ▷ $p = \text{const}$ due to purely competitive market
- ▶ profit at time t is

$$\pi(x, \dot{x}, t) = px - C_1(x) - C_2(\dot{x})$$

Example: dynamic production

Problem formulation: maximize total profit

$$F\{x\} = \int_0^T px - C_1(x) - C_2(\dot{x}) dt$$

subject to $x(0) = 0$ and $x(T) = x_T$.

- ▶ notice that the control, and rate of change of state are the same (i.e., $u = \dot{x}$) but we write it as above for simplicity
- ▶ autonomous problem
- ▶ the control is planned, and has quadratic costs
- ▶ admissible controls are unbounded

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Example: dynamic production

Euler-Lagrange equations

$$\begin{aligned}\frac{d}{dt} \frac{\partial \pi}{\partial \dot{x}} - \frac{\partial \pi}{\partial x} &= 0 \\ -\frac{d}{dt} \frac{\partial C_2}{\partial \dot{x}} - p + \frac{\partial C_1}{\partial x} &= 0 \\ -\frac{d}{dt} [2a_2\dot{x} + b_2] - p + 2a_1x + b_1 &= 0 \\ -2a_2\ddot{x} - p + 2a_1x + b_1 &= 0 \\ \ddot{x} - \frac{a_1}{a_2}x &= \frac{-p + b_1}{2a_2}\end{aligned}$$

for $a_2 \neq 0$

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Example: dynamic production

Solution (for $a_1, a_2 \neq 0$)

$$x(t) = Ae^{\sqrt{\frac{a_1}{a_2}}t} + Be^{-\sqrt{\frac{a_1}{a_2}}t} + \frac{b_1 - p}{2a_2}$$

where A and B are determined by the fixed end points $x(0) = x_0$ and $x(T) = X_T$.

This gives the optimal production schedule.

- ▶ no dependence on c_1 or c_2 (these are constant costs and so shouldn't effect production strategy)
- ▶ no dependence on b_2 because this is a linear cost in increasing production, and so occurs regardless of how we increase over time (to get to the final production target $x(T) = X_T$).

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Example: dynamic production

What happens if we make the end point $x(T)$ free, i.e. we don't have a production target at time T ?

Then we get a natural boundary condition

$$\left. \frac{\partial \pi}{\partial \dot{x}} \right|_{t=T} = \left. \frac{\partial C_2}{\partial \dot{x}} \right|_{t=T} = 2a_2\dot{x} + b_2 \Big|_{t=T} = 0$$

So, rearranging, we get

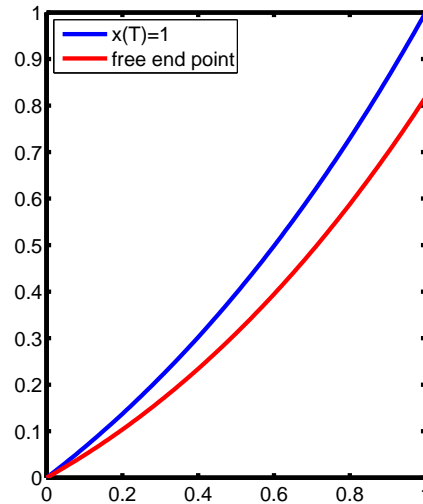
$$\dot{x}(T) = -\frac{b_2}{2a_2}$$

- ▶ constants A and B are determined by end-point conditions $x(0) = x_0$ and $\dot{x}(T) = -\frac{b_2}{2a_2}$

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Example: dynamic production

- ▶ production costs $C_1 = x^2 + 5x$
- ▶ production increase costs
 $C_2 = 2\dot{x}^2 + 5\dot{x}$
- ▶ $p = 10$
- ▶ $T = 1$
- ▶ $x_0 = 0, x_T = 1$



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Example: optimal economic growth

- ▶ GDP $Y(t)$ is a function of labor $L(t)$, and capital $K(t)$
- ▶ The production function $Y(t) = f_2(K, L)$ is homogeneous of degree one, e.g.
$$Y(t) = L(t)f_2(K/L, 1) = L(t)f(K/L)$$

- ▶ Hence we normalize all quantities by population L

$$y = Y/L \quad \text{GDP per capita}$$

$$k = K/L \quad \text{Capital investment per capita}$$

$$c = C/L \quad \text{Consumption per capita}$$

and write $y(t) = f(k)$ where f is assumed to be a strictly concave, monotonically increasing function, with slope decreasing from ∞ at 0, to 0 at ∞ .

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Example: optimal economic growth

How much should be consumed, and how much invested for future consumption?

- ▶ optimal theory of saving (Ramsey, 1928)
- ▶ Total capital at time t is $K(t)$
- ▶ Total population (labor force) $L(t)$, which grows at exogenous rate n , e.g. $\dot{L} = nL$
- ▶ Homogeneous quantity called GDP denoted $Y(t)$
- ▶ GDP can either be consumed $C(t)$ or invested to get $\dot{K}(t)$, or used to replace depreciated capital $\mu K(t)$.

$$Y(t) = C(t) + \dot{K}(t) + \mu K(t)$$

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Example: optimal economic growth

Consider the rate of per capita investment

$$\dot{k} = \frac{d}{dt} \left(\frac{K}{L} \right) = \frac{\dot{K}}{L} - \left(\frac{K\dot{L}}{L^2} \right) = \frac{\dot{K}}{L} - n \frac{K}{L} = \frac{\dot{K}}{L} - nk$$

using the fact that $\dot{L}/L = n$. Now we assumed that GDP could be expended in one of three ways, leading to

$$Y = C + \dot{K} + \mu K$$

which we also divide by L to obtain

$$y = c + \dot{k} + (\mu + n)k$$

which, when we substitute $y = f(k)$ gives

$$c(t) = f(k) - \dot{k} - (\mu + n)k(t)$$

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Example: optimal economic growth

- ▶ We want to maximize the total **utility**
- ▶ Utility of per capita consumption is $U(c)$. This would also be a strictly concave, monotonically increasing function (according to the law of diminishing marginal utility, i.e. $U''(c) < 0 < U'(c)$).
- ▶ Utility in the future is discounted by rate r , e.g. is given by $U(c)e^{-rt}$
- ▶ Our control is how much we consume (and hence what is left to invest \dot{k}), and the state is the per capita investment $k(t)$.

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Example: optimal economic growth

The E-L equations are

$$\frac{d}{dt} \frac{\partial \Psi}{\partial \dot{k}} - \frac{\partial \Psi}{\partial k} = 0$$

where $\Psi(k, \dot{k}) = U(f(k) - \dot{k} - (\mu + n)k(t))e^{-rt}$, so

$$\begin{aligned} -\frac{d}{dt} e^{-rt} \frac{dU}{dc} - e^{-rt} \frac{dU}{dc} \left[\frac{df}{dk} - (\mu + n) \right] &= 0 \\ -e^{-rt} \frac{d}{dt} \frac{dU}{dc} + e^{-rt} \frac{dU}{dc} \left[r - \frac{df}{dk} + (\mu + n) \right] &= 0 \\ -e^{-rt} \frac{d^2 U}{dc^2} \frac{dc}{dt} + e^{-rt} \frac{dU}{dc} \left[r - \frac{df}{dk} + (\mu + n) \right] &= 0 \end{aligned}$$

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Example: optimal economic growth

We want to maximize the total **utility** over time, e.g.

$$F\{c\} = \int_0^T U(c)e^{-rt} dt$$

subject to

$$c(t) = f(k) - \dot{k} - (\mu + n)k(t)$$

with $k(0) = k_0$, and $k(T) = k_T$.

Substitute c into the functional and we get

$$F\{k\} = \int_0^T U(f(k) - \dot{k} - (\mu + n)k(t)) e^{-rt} dt$$

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Example: optimal economic growth

We know $e^{-rt} \neq 0$, so we divide it out, and rearrange to get

$$\frac{dc}{dt} = \left[r + \mu + n - \frac{df}{dk} \right] \frac{U'}{U''}$$

which together with

$$\dot{k} = f(k) - c(t) - (\mu + n)k(t)$$

determines the optimal solution of the system. Remember we are given

- ▶ U the utility
- ▶ f the per capita production as a function of capital

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Example: optimal economic growth

Example, $U(c) = \log(c)$, then $U' = 1/c$ and $U'' = -1/c^2$, so

$$\frac{dc}{dt} = \alpha c \text{ where } \alpha = - \left[r + \mu + n - \frac{df}{dk} \right]$$

so

$$c(t) = Ae^{\alpha t}$$

To solve for k , take linear production model, e.g. $y = \beta k$, and then

$$\dot{k} = \gamma k(t) - c(t) \text{ where } \gamma = (\beta - \mu - n)$$

So

$$k(t) = Be^{\gamma t} + \frac{c(t)}{\gamma - \alpha} = Be^{\gamma t} + \frac{c(t)}{r}$$

with A and B determined by $k(0) = k_0$, and $k(T) = k_T$.

Example: optimal economic growth

To maintain constant consumption $c(t)$ we require $\dot{c} = 0$, and so we must have

$$\frac{df}{dk} = r + \mu + n$$

To maintain constant investment, we require

$$\dot{k} = f(k) - c(t) - (\mu + n)k(t) = 0$$

which together determine a solution (c^*, k^*) , where the system is in equilibrium.

For the example $y = \beta k$

$$k = \frac{r + \mu + n}{\beta} \text{ and } c = (\beta - \mu - n)k$$