# Optimal Control Lecture 10

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Suggested ready: Section 4.1 and 4.2 of Ref[1] (see class website or the class syllabus for the list of references)

We are going to focus on solving

$$\begin{split} u^{\star}(t)\Big|_{t\in[t_0,t_f]} &= \underset{u(t)\in\mathcal{U}}{\operatorname{argmin}}(J = h(x(t_f),t_f) + \int_{t_0}^{t_f} g(x(t),u(t),t)), \ \text{ s.t.} \\ \dot{x}(t) &= f(x(t),u(t),t), \\ x(t_0), \ t_0 \ \text{is given}, \\ m(x(t_f),t_f) &= 0 \leftarrow \ \text{when final state is constrained}, \end{split}$$

 $x(t):\mathbb{R}\to\mathbb{R}^n,\quad u(t):\mathbb{R}\to\mathbb{R}^m,\quad f:\mathbb{R}^n\times\mathbb{R}^m\times\mathbb{R}\to\mathbb{R}^n.$ 

Observations:

- J is a function of x(t), u(t) both functions over  $t \in [t_0, t_f]$
- J is a functional (function of a function)

#### Static parameter optimization:

 objective: determine a point that minimizes a specific function (the performance measure)

#### Optimization in continuous-time:

 objective: determine <u>a function</u> that minimizes a specific functional (the performance measure)

### Function vs. functional

Def (function): A function f is a rule of correspondence that assigns to each element q in a certain set  $\mathcal{D}$  (domain of the function) a unique element in a set  $\mathcal{R}$  (range or image of the function)

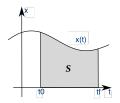
Def (functional): A functional J is a rule of correspondence that assigns to each function x in a certain class  $\Omega$  (domain of the functional) a unique real number. The set of real numbers associated with the functions  $\Omega$  is called the range of the functional.

- functional: function of function
- domain is a class of functions

**Example**: x: continuous function of t defined in the interval  $[t_0, t_f]$  and

$$J(\mathbf{x}) = \int_{t_0}^{t_f} \mathbf{x}(t) dt$$

is a functional. Its range is the area under x(t) curves.



- discrete-time optimal control: can be cast as parameter optimization with a finite dimensional decision variable and constraints
- continuous-time optimal control: infinite dimensional decision variable
  - Continuous-time optimal control: use Calculus of Variation

#### Calculus of Variation

- field of mathematical analysis that deals with maximization/minimization of functionals
- functionals are defined as integrals involving functions and their derivatives
- interest is in extremal functions that make the functional attain
  - maximum
  - minimum
  - or stationary functions (those where the rate of change of the functional is zero)

$$\begin{split} q^{\star} &= \text{argmin } f(q) \\ \text{Point } q^{\star} \text{ is a minimizer of a function } f(q) \text{ iff} \\ f(q^{\star}) \leqslant f(q) & J(x) \\ \text{for all admissible } q \text{ in } \|q - q^{\star}\| \leqslant \varepsilon \end{split}$$

$$x^{\star} = \text{argmin} \ J = \int_{t_0}^{t_f} g(x(t), \dot{x}(t), t)$$

Function  $x^{\star}(t)$  is a minimizer of functional J(x(t)) iff

$$J(x^{\star}(t)) \leqslant J(x(t))$$

for all admissible x(t) in  $\|x(t)-x^\star(t)\|\leqslant\varepsilon.$ 

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Tools we need for our studies:

• How to measure closeness of two functions?

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 How to compute/approximate variation of a functional due to 'small' changes in its arguments, which are functions?

#### Norm:

in n-dimensional Euclidean space: rule of correspondence which assigns to each point q a real number.

$$\textcircled{1} \|q\| \geqslant 0 \text{ and } \|q\| = 0 \text{ iff } q = 0$$

2 
$$\|\alpha q\| = |\alpha| \|q\|$$
 for all  $\alpha \in \mathbb{R}$ 

$$\|q^1 + q^2\| \leqslant \|q^1\| + \|q^2\|$$

$$q^1 \text{ and } q^2 \text{ close together} \Leftrightarrow \|q^1 - q^2\| \text{ is small}$$

of a function: rule of correspondence which assigns to each function  $x \in \Omega$ , defined for  $t \in [t_0, t_f]$ , a real number.

 $\label{eq:constraint} \|x\| \geqslant 0 \text{ and } \|x\| = 0 \text{ iff } x(t) = 0 \text{ for all } t \in [t_0,t_f]$ 

2 
$$\|\alpha x\| = |\alpha| \|x\|$$
 for all  $\alpha \in \mathbb{R}$ 

**3** 
$$||x^1 + x^2|| \leq ||x^1|| + ||x^2||$$

Intuitively speaking norm of the difference of two functions should be

- zero if the functions are identical
- small, if the functions are "close"
- large if the functions are "far apart"

Examples

•  $\|x\|_2 = (\int_{t_0}^{t_f} x^\top(t) x(t) dt)^{1/2}$ 

• 
$$||x|| = \max_{t_0 \leqslant t \leqslant t_f} (|x(t)|)$$
, (scalar x)

#### Increment:

of a function f: If q,  $q + \Delta q \in D$ , the increment of f is

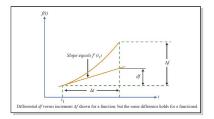
$$\Delta \mathbf{f} = \mathbf{f}(\mathbf{q} + \Delta \mathbf{q}) - \mathbf{f}(\mathbf{q}).$$

of a functional J: If x and  $x+\delta x$  are functions for which the functional J is defined, then increment of J is

$$\Delta \mathbf{J} = \mathbf{J}(\mathbf{x} + \delta \mathbf{x}) - \mathbf{J}(\mathbf{x}).$$

 $\delta x$  is the variation of the function x

### The variation of a functional



variation of a functional  $\sim$  differential of a function

The increment of a functional can be written as

$$\Delta J(\mathbf{x}(t), \delta \mathbf{x}(t)) = \delta J(\mathbf{x}(t), \delta \mathbf{x}(t)) + g(\mathbf{x}(t), \delta \mathbf{x}(t)) \| \delta \mathbf{x}(t) \|,$$

where  $\delta J$  is a linear in  $\delta x(t)$ . If

$$\lim_{\|\delta x(t)\|\to 0} (g(x(t), \delta x(t))) = 0,$$

then J is said to be *differentiable* on x and  $\delta J$  is the variation of J evaluated for a function x.

A variation of the functional is a linear approximation of this increment, i.e.,  $\delta J(x(t), \delta x(t))$  is linear in  $\delta x(t)$ .

$$\Delta J(x(t), \delta x(t)) = \delta J(x(t), \delta x(t)) + H.O.T.,$$

How to compute variation of  $J(x(t)) = \int_{t_0}^{t_f} f(x(t)) dt$  (assuming f has first and second continuous derivative)?

 $\delta J(x(t), \delta x(t)) = \int_{t_0}^{t_f} \frac{\partial f(x(t))}{\partial x(t)} \cdot \delta x dt + f(x(t_f)) \delta t_f - f(x(t_0)) \delta t_0$ 

See next page for the derivation

### The variation of a functional: example 1 (cont'd)

$$\begin{split} \Delta J(x(t), \delta x(t)) &= J(x(t) + \delta x(t)) - J(x(t)) \\ &= \int_{t_0 + \delta t_0}^{t_f + \delta t_f} (f(x(t) + \delta x(t)) \, dt - \int_{t_0}^{t_f} f(x(t)) \, dt \\ &= - \int_{t_0}^{t_0 + \delta t_0} (f(x(t) + \delta x(t)) \, dt + \int_{t_f}^{t_f + \delta t_f} (f(x(t) + \delta x(t)) \, dt \\ &+ \int_{t_0}^{t_f} (f(x(t) + \delta x(t)) \, dt - \int_{t_0}^{t_f} f(x(t)) \, dt \end{split}$$

• 
$$\int_{t_0}^{t_0+\delta t_0} f(x(t) + \delta x(t)) dt \approx (f(x(t_0) + \delta x(t_0))\delta t_0 = -f(x(t_0)) \delta t_0 + H.O.T$$

 $\blacktriangleright \int_{t_f}^{t_f + \delta t_f} f(x(t) + \delta x(t)) dt \approx (f(x(t_f) + \delta x(t_f)) \delta t_f = f(x(t_f)) \delta t_f + H.O.T$ 

$$\int_{t_0}^{t_f} f(x(t) + \delta x(t)) dt - \int_{t_0}^{t_f} f(x(t)) dt = \int_{t_0}^{t_f} (f(x(t) + \delta x(t)) - f(x(t))) dt$$

$$= \int_{t_0}^{t_f} (f(x(t) + \frac{\partial f(x(t))}{\partial x(t)} \cdot \delta x + H.O.T - f(x(t))) dt \approx \int_{t_0}^{t_f} \frac{\partial f(x(t))}{\partial x(t)} \cdot \delta x dt$$

 $\delta J(x(t), \delta x(t)) = \int_{t_0}^{t_f} \frac{\partial f(x(t))}{\partial x(t)} \cdot \delta x dt + f(x(t_f)) \delta t_f - f(x(t_0)) \delta t_0$ 

How to compute variation of  $J(x(t)) = \int_{t_0}^{t_f} g(x(t), \dot{x}(t), t) dt$  for fixed  $t_0$  (assuming f has first and second continuous derivative)?

$$\begin{split} \delta J(x(t),\delta x(t)) = & \int_{t_0}^{t_f} (g_x - \frac{d}{dt}g_{\dot{x}}) \cdot \delta x dt + g_{\dot{x}}(x(t_f),\dot{x}(t_f),t_f) \delta x(t_f) + \\ & g(x(t_f),\dot{x}(t_f),t_f) \delta t_f \end{split}$$

 $(g_x = \frac{\partial g}{\partial x}, g_{\dot{x}} = \frac{\partial g}{\partial \dot{x}})$ 

See next page for the derivation

## The variation of a functional: example 2 (cont'd)

$$\begin{split} \Delta J(\mathbf{x}(t), \delta \mathbf{x}(t)) &= J(\mathbf{x}(t) + \delta \mathbf{x}(t)) - J(\mathbf{x}(t)) \\ &= \int_{t_0}^{t_f + \delta t_f} (g(\mathbf{x}(t) + \delta \mathbf{x}(t), \dot{\mathbf{x}}(t) + \delta \dot{\mathbf{x}}(t), t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_f}^{t_f + \delta t_f} (g(\mathbf{x}(t) + \delta \mathbf{x}(t), \dot{\mathbf{x}}(t) + \delta \dot{\mathbf{x}}(t), t) \, dt \\ &+ \int_{t_0}^{t_f} (g(\mathbf{x}(t) + \delta \mathbf{x}(t), \dot{\mathbf{x}}(t) + \delta \dot{\mathbf{x}}(t), t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &+ \int_{t_f}^{t_f + \delta t_f} (g(\mathbf{x}(t) + \delta \mathbf{x}(t), \dot{\mathbf{x}}(t) + \delta \dot{\mathbf{x}}(t), t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= g(\mathbf{x}(t_f), \dot{\mathbf{x}}(t_f), t_f) \, \delta t_f + H.O.T \end{split}$$

$$* \int_{t_0}^{t_f} (g(\mathbf{x}(t) + \delta \mathbf{x}(t), \dot{\mathbf{x}}(t) + \delta \dot{\mathbf{x}}(t), t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} (g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) + g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) + g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt + g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt - \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt + g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt + g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) + \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, dt \\ &= \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt + g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) + \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}(t) \, dt \\ &= \int_{t_0}^{t_f} (g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \, \delta \mathbf{x}($$

### Extremal of a functional: fundamental theorem of the calculus of variation

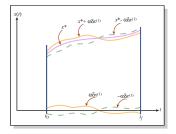
 $\begin{array}{ll} \mbox{Minimizer of a function } f(q) \mbox{ is } q^{\star} \mbox{ if } & \mbox{Minimizer of a functional } J(x(t)) \mbox{ is } x^{\star}(t) \mbox{ if } \\ f(q^{\star}) \leqslant f(q) & \mbox{ } J(x^{\star}(t)) \leqslant J(x(t)) \\ \mbox{for all admissible } q \mbox{ in } \|q - q^{\star}\| \leqslant \varepsilon & \mbox{for all admissible } x(t) \mbox{ in } \|x(t) - x^{\star}(t)\| \leqslant \varepsilon. \\ \end{array}$ 

#### Fundamental theorem of the calculus of variation

- Let x be a vector function of t in the class  $\Omega$ , and J(x) be a differential functional of x.
- Assume that all  $x \in \Omega$  are not constrained by any boundaries. If  $x^*$  is an extremal function, the variation of J must vanish in  $x^*$

$$\delta J(\mathbf{x}^{\star}, \delta \mathbf{x}) = \mathbf{0}$$

for all admissible  $x \in \Omega$ .



### Optimal control problems of interest

We are going to study

$$\begin{split} u^{\star}(t)\Big|_{t\in[t_0,t_f]} &= \underset{u(t)\in\mathcal{U}}{\operatorname{argmin}}(J = h(x(t_f),t_f) + \int_{t_0}^{t_f} g(x(t),u(t),t)dt), \ \text{ s.t.} \\ \dot{x}(t) &= f(x(t),u(t),t), \\ x(t_0), \ t_0 \ \text{is given}, \\ m(x(t_f),t_f) &= 0 \leftarrow \ \text{when final state is constrained}, \end{split}$$

 $x(t):\mathbb{R}\to\mathbb{R}^n,\quad u(t):\mathbb{R}\to\mathbb{R}^m,\quad f:\mathbb{R}^n\times\mathbb{R}^m\times\mathbb{R}\to\mathbb{R}^n.$ 

We will first focus on the special case below.<sup>1</sup>

$$\begin{split} x^{\star}(t) \Big|_{t \in [t_0, t_f]} &= \text{argmin} \left( J(x(t)) = \int_{t_0}^{t_f} g(x(t), \dot{x}(t), t) dt \right) \, s.t. \\ x(t_0) &= x_0, \\ x(t_f) &= x_f \quad (\text{various terminal conditions }) \end{split}$$

<sup>1&</sup>quot;Think of it as a case that we can find u(t) in terms of  $(x(t), \dot{x}(t))$  from  $\dot{x}(t) = f(x(t), u(t), t)$ . Then the optimal control problem above can be cast with u(t) eliminated."

### First order optimality conditions

$$\begin{split} x^{\star}(t) \Big|_{t \in [t_0, t_f]} &= \text{argmin} \left( J(x(t)) = \int_{t_0}^{t_f} g(x(t), \dot{x}(t), t) dt \right) \, s.t. \\ x(t_0) &= x_0, \\ x(t_f) &= x_f \quad (\text{various terminal conditions }) \end{split}$$

Variation

$$\begin{split} \delta J(\mathbf{x}(t), \delta \mathbf{x}(t)) = & g(\mathbf{x}(t_{f}), \dot{\mathbf{x}}(t_{f}), t_{f}) \, \delta t_{f} + g_{\dot{\mathbf{x}}}(\mathbf{x}(t_{f}), \dot{\mathbf{x}}(t_{f}), t_{f}) \, \delta \mathbf{x}(t_{f}) + \\ & \int_{t_{0}}^{t_{f}} \left( g_{\mathbf{x}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) - \frac{d}{dt} g_{\dot{\mathbf{x}}}(\mathbf{x}(t), \dot{\mathbf{x}}(t), t) \right) \delta \mathbf{x}(t) \, dt \end{split}$$

From this variation, for different terminal conditions, we are going to derive first order necessary conditions for optimality using the Fundamental Theorem of Calculus of Variation.

- Both  $t_f$  and  $x(t_f)$  are specified and are given
  - $\bullet\,$  In this case  $\delta t_{\rm f}=0$  and  $\delta x(t_{\rm f})=0$
  - $\delta J(x(t),\delta x(t)) = \int_{t_0}^{t_f} \left( g_x(x(t),\dot{x}(t),t) \frac{\mathrm{d}}{\mathrm{d}t}g_{\dot{x}}(x(t),\dot{x}(t),t) \right) \delta x(t) \, \mathrm{d}t = 0 \Rightarrow$

the (first order) necessary condition for a maximum or minimum is called Euler Equation

$$g_x(x(t),\dot{x}(t),t) - \frac{d}{dt}g_{\dot{x}}(x(t),\dot{x}(t),t) = 0$$

In this case we solve the Euler Equation with the boundary conditions  $\chi(0) = \chi_0$  and