# Optimal Control Lecture 12

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- Calculus of variation
- Optimal Control

## **Piecewise-smooth extremals**

- So far we focused on admissible x(t) that are continuous with continuous first derivatives
- We want to expand to class if piecewise-smooth admissible functions
  - control input is no smooth (e.f., subject to saturation)



Illustration of a piecewise continuous control  $u \in \hat{C}[t_0, t_t]$  (red line), and the corresponding piecewise continuously differentiable response  $x \in \hat{C}^1[t_0, t_t]$  (blue line).

intermediate state constraints are imposed

Objective: determine vector function  $x^{\star}(t)$  in the class of functions with *piecewise-continuous* first derivative that is a local extremum of

$$J(\mathbf{x}(t)) = \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}, t) d(t)$$

and respects  $x(t_0)=x_0\in \mathbb{R}^n,\, x(t_f)=x_f$  for given and fixed  $t_0,\, x_0,\, t_f$  and  $x_f$ 

#### **Piecewise-smooth extremal**

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• Assume  $\dot{x}$  has a discontinuity at  $t_1 \in (t_0, t_f)$ , where  $t_1$  is not fixed (or known)

$$J(\mathbf{x}(t)) = \int_{t_0}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}, t) d(t) = \underbrace{\int_{t_0}^{t_1} g(\mathbf{x}(t), \dot{\mathbf{x}}, t) dt}_{J_1} + \underbrace{\int_{t_1}^{t_f} g(\mathbf{x}(t), \dot{\mathbf{x}}, t) dt}_{J_2} = \underbrace{\int_{t_0}^{t_1} (\frac{\partial g}{\partial \mathbf{x}} \delta \mathbf{x} + \frac{\partial g}{\partial \dot{\mathbf{x}}} \delta \dot{\mathbf{x}}) dt + g(t_1^-) \delta t_1 + \int_{t_1}^{t_f} (\frac{\partial g}{\partial \mathbf{x}} \delta \mathbf{x} + \frac{\partial g}{\partial \dot{\mathbf{x}}} \delta \dot{\mathbf{x}}) dt - g(t_1^+) \delta t_1}_{I_0}$$

$$\begin{split} \delta J = & \int_{t_0}^{t_1} (g_x - \frac{d}{dt}g_{\dot{x}})\delta x dt + g(t_1^-)\delta t_1 + g_{\dot{x}}(t_1^-)\delta x(t_1^-) \\ & \int_{t_1}^{t_f} (g_x - \frac{d}{dt}g_{\dot{x}})\delta x dt - g(t_1^+)\delta t_1 - g_{\dot{x}}(t_1^+)\delta x(t_1^+) \end{split}$$

$$x_{0}$$
  
 $x_{0}$   
 $x_{0$ 

$$\label{eq:constraint} \begin{split} \text{from lefthand side } & \delta x_1 = \delta x(t_1^-) + \dot{x}(t_1^-) \delta t_1, \\ \text{from righthand side } & \delta x_1 = \delta x(t_1^+) + \dot{x}(t_1^+) \delta t_1, \end{split}$$

- Continuity requires that these two expression for δx<sub>1</sub> be equal
- Already now that it is possible  $\dot{x}(t_1^-) \neq \dot{x}(t_1^+), \text{ so possible that } \\ \delta x(t_1^-) \neq \delta x(t_1^+)$

$$\begin{split} \delta J &= \int_{t_0}^{t_1} (g_x - \frac{d}{dt} g_{\dot{x}}) \delta x dt + [g(t_1^-) - g_{\dot{x}}(t_1^-) \dot{x}(t_1^-)] \delta t_1 + g_{\dot{x}}(t_1^-) \delta x_1 \\ &\int_{t_1}^{t_f} (g_x - \frac{d}{dt} g_{\dot{x}}) \delta x dt - [g(t_1^+) - g_{\dot{x}}(t_1^+) \dot{x}(t_1^+)] \delta t_1 - g_{\dot{x}}(t_1^+) \delta x_1 \end{split}$$

$$\begin{split} &\frac{\partial g}{\partial x}(x^{\star}(t),\dot{x}^{\star}(t),t) - \frac{d}{dt}[\frac{\partial g}{\partial \dot{x}}(x^{\star}(t),\dot{x}^{\star}(t),t)] = 0, \quad t \in (t_0,t_f) \\ &x^{\star}(t_0) = x_0, \end{split}$$

Weierstrasss-Erdmann Condition

$$\begin{split} &g_{\dot{x}}(t_1^-) = g_{\dot{x}}(t_1^+), \\ &g(t_1^-) - g_{\dot{x}}(t_1^-) \dot{x}(t_1^-) = g(t_1^+) - g_{\dot{x}}(t_1^+) \dot{x}(t_1^+) \end{split}$$

#### **Piecewise-smooth extremal**

Typical scenarios that introduce corners is when there exists intermediate constraints

$$\mathbf{x}(\mathbf{t}_1) = \mathbf{\theta}(\mathbf{t}_1)$$

 $\bullet\,$  Constraint couples the allowable variations in  $\delta x_1$  and  $\delta t_1$ 

$$\delta x_1 = \frac{\mathsf{d}\theta}{\mathsf{d}t} \delta t_1 = \dot{\theta}(t_1) \delta t_1$$

$$\begin{aligned} \bullet \ \delta J &= \int_{t_0}^{t_0} (g_x - \frac{d}{dt} g_{\dot{x}}) \delta x dt + [g(t_1^-) - g_{\dot{x}}(t_1^-) \dot{x}(t_1^-)] \delta t_1 + g_{\dot{x}}(t_1^-) \delta x_1 + \int_{t_1}^{t_f} (g_x - \frac{d}{dt} g_{\dot{x}}) \delta x dt - [g(t_1^+) - g_{\dot{x}}(t_1^+) \dot{x}(t_1^+)] \delta t_1 - g_{\dot{x}}(t_1^+) \delta x_1 \\ \delta J &= \int_{t_0}^{t_1} (g_x - \frac{d}{dt} g_{\dot{x}}) \delta x dt + [g(t_1^-) - g_{\dot{x}}(t_1^-) \dot{x}(t_1^-)] \delta t_1 + g_{\dot{x}}(t_1^-) (\dot{\theta}(t_1^-) \delta t_1) + \\ \int_{t_1}^{t_f} (g_x - \frac{d}{dt} g_{\dot{x}}) \delta x dt - [g(t_1^+) - g_{\dot{x}}(t_1^+) \dot{x}(t_1^+)] \delta t_1 - g_{\dot{x}}(t_1^+) (\dot{\theta}(t_1^+) \delta t_1) + \end{aligned}$$

$$\begin{split} &\frac{\partial g}{\partial x}(x^{\star}(t),\dot{x}^{\star}(t),t) - \frac{d}{dt}[\frac{\partial g}{\partial \dot{x}}(x^{\star}(t),\dot{x}^{\star}(t),t)] = 0, \quad t \in (t_0,t_f) \\ &x^{\star}(t_0) = x_0, \\ &g(t_1^-) + g_{\dot{x}}(t_1^-)[\dot{\theta}(t_1^-) - \dot{x}(t_1^-)] = g(t_1^+) + g_{\dot{x}}(t_1^+)[\dot{\theta}(t_1^+) - \dot{x}(t_1^+)] \\ &x(t_1) = \theta(t_1) \end{split}$$

 $\bullet \ \ \, \text{Note here that} \ \, g_{\dot{x}}(t_1^-)=g_{\dot{x}}(t_1^+) \ \, \text{no longer is needed. Instead we have} \ \, x(t_1)=\theta(t_1).$ 

**Example**: Find the shortest length path joining the points x = 0, t = -2 and x = 0 and t = 1 that touches the curve  $x = t^2 + 3$  at some point.



Solution is the dashed blue lines

## **Optimal control**

We are going to focus on solving

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

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

• Use Lagrange multiplier to write

$$J = h(x(t_f), t_f) + \int_{t_0}^{t_f} \big(g(x(t), u(t), t) + p(t)^\top (a(x(t), u(t), t) - \dot{x}(t))\big) dt$$

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

• Use Lagrange multiplier to write

$$J = h(x(t_f), t_f) + \int_{t_0}^{t_f} \big(g(x(t), u(t), t) + p(t)^\top (a(x(t), u(t), t) - \dot{x}(t))\big) dt$$

• Define the Hamiltonian to help with sorting out the equations  $H(x, u, p, t) = g(x(t), u(t), t) + p(t)^{\top} a(x(t), u(t), t)$ 

## **Optimal control**

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$$\begin{split} \delta J_{a} = & (h_{x} - p(t_{f}))^{\top} \delta x_{f} + \left[h_{t_{f}} + g + p^{\top}(a - \dot{x}) + p^{\top} \dot{x}\right]_{t_{f}} \delta t_{f} + \\ & + \int_{t_{0}}^{t_{f}} \left[ (H_{x} + \dot{p})^{\top} \delta x(t) + H_{u}^{\top} \delta u(t) + (a - \dot{x})^{\top} \delta p(t) \right] dt \end{split}$$

first order conditions for extremal solution

$$\begin{split} \dot{p} &= -H_x, & (n \text{ dimensional}) \\ 0 &= H_u, & (m \text{ dimensional}) \\ 0 &= H_p \rightarrow \quad \dot{x} = a(x,u,t), & (n \text{ dimensional}) \end{split}$$

Boundary conditions  $\boldsymbol{x}(t_0) = \boldsymbol{x}_0,$  and

 $\bullet \ \ \, \text{if} \ t_f \ \, \text{free}$ 

$$\mathbf{h}_{t_f} + \mathbf{g} + \mathbf{p}^\top \mathbf{a} = \mathbf{h}_{t_f} + \mathbf{H}(\mathbf{t}_f) = \mathbf{0}$$

• if 
$$x_i(t_f)$$
 is fixed:  $x_i(t_f) = x_{i_f}$ 

• if 
$$x_i(t_f)$$
 is free, then  $p_i(t_f) = \frac{\partial h}{\partial x_i}(t_f)$ 

## Constrained functional optimization: example

Minimum-time path through a region of position dependent vector velocity (Zernelo's problem)



- The forward velocity of the ship V is constant but its steering angle  $\theta$  can be controlled.
- in the depicted example, it is assume that the current's velocity vector is only in x direction



see [Bryson and Ho]

Next couple of slides are for self-study

## Constrained functional optimization

Determine vector function  $w^{\star}(t): \mathbb{R} \to \mathbb{R}^{n+m}$  in the class of functions with continuous first derivative that is a local extremum of

$$J(w(t),t) = \int_{t_0}^{t_f} g(w(t),\dot{w},t)d(t)$$

and respects

$$\begin{split} f(w(t),\dot{w}(t),t) &= 0_n, \text{ set of } n \text{ differential equations,} \\ w(t_0) &= w_0 \in \mathbb{R}^{n+m}, \\ w(t_f) &= w_f \quad \text{various terminal conditions possible.} \end{split}$$

• in control problems 
$$w = \begin{bmatrix} u \in \mathbb{R}^m \\ x \in \mathbb{R}^n \end{bmatrix}$$

To derive the first order necessary conditions we proceed with the following

• Similar to function parameter optimization, augment the cost functional with the constraint using Lagrange multiplier

$$J_{\mathfrak{a}}(w(t),t) = \int_{t_0}^{t_f} \left( g(w(t), \dot{w}, t) + p(t)^{\top} f(w(t), \dot{w}, t) \right) \mathsf{d}(t)$$

- Notice that p(t) is is time-varying (gives more degree of freedom)
- If constraint is satisfied the augmented cost functional is same as J(w(t), t)

Let

 $g_{\mathfrak{a}}(w(t),p(t),\dot{w}(t),t) \equiv g(w(t),\dot{w}(t),t) + p(t)^{\top}f(w(t),\dot{w}(t),t)$ 

Then

$$J_{\mathfrak{a}}(w(t), p(t), \delta x, \delta p, t) = \int_{t_0}^{t_f} \left( g_{\mathfrak{a}}(w(t), p(t), \dot{w}, t) \right) d(t)$$

• Invoke Fundamental Theorem of Calculus of Variation:  $\delta J_{\alpha}(w^{\star}(t), p^{\star}(t), \delta x, \delta p, t) = 0$ 

• The variations  $\delta x$  and  $\delta p$  are independent from one another.

The first order necessary conditions are

$$\text{Euler equation:} \quad \begin{cases} \frac{\partial g_{\alpha}(w(t),p(t),\dot{w},t)}{\partial w} - \frac{d}{dt} [\frac{\partial g_{\alpha}(w(t),p(t),\dot{w},t)}{\partial \dot{w}}] = 0, \\ \\ \frac{\partial g_{\alpha}(w(t),p(t),\dot{w},t)}{\partial p} - \frac{d}{dt} [\frac{\partial g_{\alpha}(w(t),p(t),\dot{w},t)}{\partial \dot{p}}] = 0, \end{cases}$$

$$\begin{cases} \frac{\partial g_{\alpha}(w(t),p(t),\dot{w},t)}{\partial w} - \frac{d}{dt} \left[ \frac{\partial g_{\alpha}(w(t),p(t),\dot{w},t)}{\partial \dot{w}} \right] = 0, \quad (n+m \text{ dimensional}) \\ w(t_0) = w_0, \\ w(t_f) = w_f, \end{cases}$$

$$F(w,\dot{w},t) = 0, \quad (n \text{ dimensional}).$$

• 
$$w^{\star}(t) : \mathbb{R} \to \mathbb{R}^{n+m}, p^{\star}(t) : \mathbb{R} \to \mathbb{R}^{n}$$

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