

Lecture #14

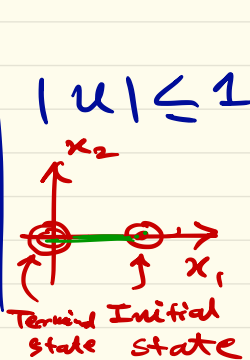
Regarding singular controls:

Main message: Singular controls happen often in practice.

Example ①

Singular controls in Nonlinear Systems

$$\begin{aligned} \min_{u(\cdot)} & \int_0^T 1 \cdot dt \\ \text{s.t.} & \dot{x}_1 = x_2^2 - 1 \\ & \dot{x}_2 = u \end{aligned}$$



$$\begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \text{ given}$$

$$\begin{pmatrix} x_1(T) \\ x_2(T) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \text{ fixed.}$$

$$H = 1 + \lambda_1(x_2^2 - 1) + \lambda_2 u$$

$$\dot{\lambda}_1 = -\frac{\partial H}{\partial x_1} = 0 \Rightarrow \lambda_1 = \text{const.}$$

$$\dot{\lambda}_2 = -\frac{\partial H}{\partial x_2} = -2\lambda_1 x_2$$

PMP $u^* = \begin{cases} -1 & \text{if } \lambda_2(t) > 0 \\ +1 & \text{if } \lambda_2(t) < 0 \\ ?? & \text{if } \lambda_2(t) = 0 \end{cases}$

Can show:

$$u^*(t) = 0 \quad \forall t \in [0, T]$$

\therefore Optimal control is singular for all $t \in [0, T]$

$\Leftrightarrow \overline{x^*}(t)$ is a singular arc for all $t \in [0, T]$

Slightly general set up for singular optimal control in nonlinear systems:

$$\boxed{\underline{x} \in \mathbb{R}^n, u \in \mathbb{R}} \quad \dot{\underline{x}} = \underline{f}(\underline{x}) + \underline{g}(\underline{x})u$$

$$\min_{u(\cdot)} \int_0^T 1 \cdot dt$$

$$\text{s.t. } \dot{\underline{x}} = \underline{f}(\underline{x}) + \underline{g}(\underline{x})u$$

$$|u| \leq 1.$$

$$H = 1 + \underline{\lambda}^T (\underline{f}(\underline{x}) + \underline{g}(\underline{x})u)$$

$$\dot{\underline{\lambda}} = -\frac{\partial H}{\partial \underline{x}} = -\left(\frac{\partial \underline{f}}{\partial \underline{x}}\right)^T \underline{\lambda} - \left(\frac{\partial \underline{g}}{\partial \underline{x}}\right)^T \underline{\lambda} u$$

PMP

$$u^* = \begin{cases} -1 & \text{if } \langle \underline{\lambda}(t), \underline{g}(\underline{x}) \rangle > 0 \\ +1 & \text{if } \langle \underline{\lambda}(t), \underline{g}(\underline{x}) \rangle < 0 \\ ?? & \text{if } \langle \underline{\lambda}(t), \underline{g}(\underline{x}) \rangle = 0 \end{cases}$$

singular

∴ whether $u^*(t)$ is singular or not, depends on "switching f^* "

$$0(t) := \langle \underline{\lambda}(t), \underline{g}(\underline{x}) \rangle$$

$n \times 1 \quad n \times 1$

If the control $u^*(t)$ is singular over some subinterval $[t_1, t_2]$ then

$$\sigma(t) := \langle \underline{\lambda}(t), \underline{g}(x) \rangle \equiv 0 \quad \forall t \in [t_1, t_2]$$

$$\Rightarrow \underline{\lambda}(t) \perp^r \underline{g}(x)$$

Also,

$$\dot{\sigma}(t) \equiv 0 \quad \forall t \in [t_1, t_2]$$

$$\ddot{\sigma}(t) \equiv 0 \quad \forall t \in [t_1, t_2]$$

⋮

etc.

Now, $\dot{\sigma}(t) := \frac{d}{dt} \sigma(t) = \langle \dot{\underline{\lambda}}, \underline{g} \rangle + \langle \underline{\lambda}, \dot{\underline{g}} \rangle$

$$= \left\langle - \left(\frac{\partial f}{\partial x} + \frac{\partial g}{\partial x} u \right)^T \underline{\lambda}, \underline{g} \right\rangle$$

costate ODE RHS

$$+ \left\langle \underline{\lambda}, \frac{\partial g}{\partial x} \dot{\underline{x}} \right\rangle \rightarrow \left(\frac{f(x)}{g(x)} + u \right)$$

chain rule

$$\Rightarrow \dot{\sigma} = \frac{d}{dt} \sigma(t) = \left\langle \underline{\lambda}, -\left(\frac{\partial f}{\partial \underline{x}} + \frac{\partial g}{\partial \underline{x}} u\right) \underline{g} \right\rangle$$

$$+ \left\langle \underline{\lambda}, \frac{\partial g}{\partial \underline{x}} (f(\underline{x}) + g(\underline{x})u) \right\rangle$$

$$= \left\langle \underline{\lambda}, -\frac{\partial f}{\partial \underline{x}} \underline{g} + \frac{\partial g}{\partial \underline{x}} \underline{f} \right\rangle \equiv 0$$

$\forall t \in [t_1, t_2]$

$$= \left\langle \underline{\lambda}, \underbrace{[\underline{f}, \underline{g}]}_{\text{new vector field}}(\underline{x}) \right\rangle$$

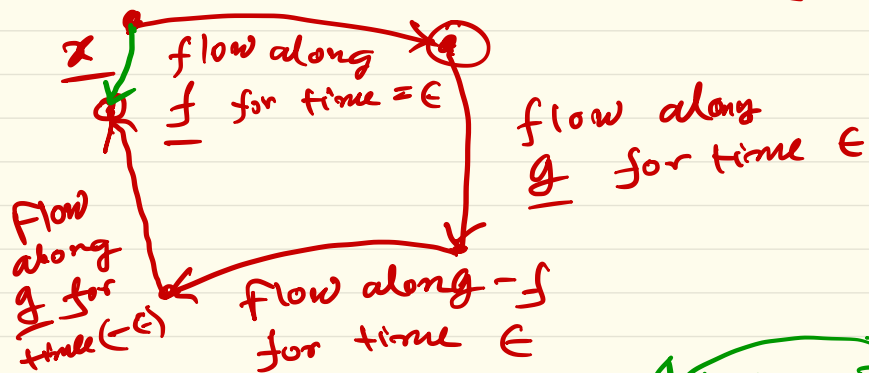
new vector field
called "Lie Bracket"
or "commutator" vector field

Obviously,

of \underline{f} & \underline{g}

$$[\underline{f}, \underline{g}](\underline{x}) = -[\underline{g}, \underline{f}](\underline{x}), \quad [\underline{f}, \underline{f}] = 0.$$

What does it mean: $[\underline{f}, \underline{g}](\underline{x})$



$h(\underline{x})$



Can show that:
this composition

$$\in^4 [\underline{f}, \underline{g}](\underline{x})$$

Example: $\underline{f}(\underline{x}) = A\underline{x}$, $\underline{g}(\underline{x}) = B\underline{x}$

$$\text{Then } [\underline{f}, \underline{g}](\underline{x}) = \frac{\partial \underline{g}}{\partial \underline{x}} \underline{f}(\underline{x}) - \frac{\partial \underline{f}}{\partial \underline{x}} \underline{g}(\underline{x})$$

$$= B(A\underline{x}) - A(B\underline{x})$$

$$= \underline{(BA - AB)} \underline{x}$$

$$= [\underline{f}, \underline{g}](\underline{x})$$

Again, having singular control u^* is equivalent to

$$\sigma(t) \equiv 0 \quad \forall t \in [t_1, t_2] \Leftrightarrow \langle \underline{\lambda}(t), \underline{g}(\underline{x}) \rangle \equiv 0,$$

$$\dot{\sigma}(t) \equiv 0 \quad \forall t \in [t_1, t_2] \Leftrightarrow \langle \underline{\lambda}(t), [\underline{f}, \underline{g}](\underline{x}) \rangle \equiv 0,$$

$$\ddot{\sigma}(t) \equiv 0 \quad \forall t \in [t_1, t_2] \Leftrightarrow \langle \underline{\lambda}(t), [\underline{f}, [\underline{f}, \underline{g}]](\underline{x}) \rangle$$

$$+ \langle \underline{\lambda}(t), [\underline{g}, [\underline{f}, \underline{g}]](\underline{x}) \rangle u^* \equiv 0,$$

Theorem for $n=2$: Let min-time problem, $|u| \leq 1, n=2$.

Suppose $\underline{g}(\underline{x})$ & $[\underline{f}, \underline{g}](\underline{x})$ are linearly independent (for all $\underline{x} \in \mathbb{R}^n$), (i.e.) $\text{rank}[\underline{g} \mid [\underline{f}, \underline{g}]] = 2$.

then u^* is bang-bang..
(no singular control)

Back to example 1:

$$\begin{cases} \dot{x}_1 = x_2^2 - 1 \\ \dot{x}_2 = u \end{cases} \Leftrightarrow \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} x_2^2 - 1 \\ 0 \end{pmatrix}}_{\underline{f(x)}} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_{\underline{g(x)}} u$$
$$\underline{x} := \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2, \quad u \in \mathbb{R}.$$

$$\underline{g(x)} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$

$$\begin{aligned} [\underline{f}, \underline{g}](\underline{x}) &:= \frac{\partial \underline{g}}{\partial \underline{x}} \underline{f}(\underline{x}) - \frac{\partial \underline{f}}{\partial \underline{x}} \underline{g}(\underline{x}) \\ &= \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \underline{f}(\underline{x}) - \begin{pmatrix} 0 & 2x_2 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 2x_2 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} -2x_2 \\ 0 \end{pmatrix} \end{aligned}$$

$$\text{rank} \left[\underline{f} \mid [\underline{f}, \underline{g}](\underline{x}) \right] = \text{rank} \begin{bmatrix} 0 & -2x_2 \\ 1 & 0 \end{bmatrix}$$

If $x_2 = 0$, then NOT linearly indep. $\left[\underline{f}, \underline{g} \right] = 0$

Example (Fuller's Problem)

$$\min_{u(\cdot)} \int_0^T x_1^2 dt, \quad T \text{ free.}$$

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = u \end{cases}$$

$$\left. \begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = u \end{cases} \right\} \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix} = \begin{pmatrix} x_0 \\ v_0 \end{pmatrix} \text{ given, } \begin{pmatrix} x_1(T) \\ y_1(T) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \text{ fixed.}$$

Exercise: ① No singular arc.

Show:

② $u^* \in \{-1, 1\}$, i.e., bang-bang

with # of switching = $+\infty$.

③ Switching takes place on the curve

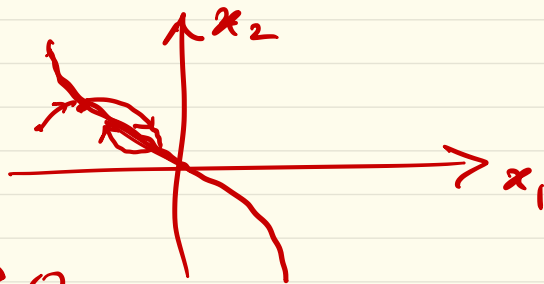
$$\left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2 \mid x_1 + \delta |x_2| x_2 = 0 \right\} \text{ where } \delta \approx 0.445$$

④ Time intervals betⁿ consecutive switches decreases in Geometric Progression.

→ Fuller's phenomenon, also known as Zero behavior

In the min. time problem, switching curve had the same form with $\gamma = \gamma_2$.

We can encode this type of problems as:



$$\min_{u(\cdot)} J(u) = \int_0^T |x_1(t)|^v dt, \quad v \geq 0$$

If $\begin{cases} v = 0, & \text{then min. time OCP, \# of switching} \leq 1 \\ v = 2, & \text{Fuller's problem, \# of switching} = +\infty \end{cases}$

→ what happens in between:

Bifurcation: $\exists \bar{v} \approx 0.35$ s.t. for $v \in [0, \bar{v}]$
 u^* is bang-bang with ≤ 1 switch

If $v > \bar{v}^* \approx 0.35$, then zero behavior (∞ switching)

Maximum Principle

Theory: PMP, costates, inequalities, 2PBVP

Setting: Deterministic
BOTH continuous & discrete time

Computation:

2 ways
After 1990s

Indirect Method

→ Exact OCP

→ Approx. solⁿ of Exact 2PBVP problem
→ shooting type algorithms

Direct Method

→ Approximate the problem
→ Solve Approx. Problem exactly

Dynamic Programming

Theory: Recursion on Value Function

Setting: Deterministic and Stochastic case

(BOTH continuous & discrete time)

continuous time: PDE

discrete time: Recursion on function space

Computation:

Exponential complexity