

Lectures 20: 18th Feb.

Module 3: Optimal control of continuous-time systems.

Given a CT dynamical system $\dot{x}(t) = f(x(t), u(t), t)$,
the goal is to find an input signal $u(t)$ for $t \in [t_0, T]$
such that the resulting state trajectory $x(t)$, $t \in [t_0, T]$
together with the input signal $u(t)$ minimize the
following functional:

$$\underline{J}(x(t), u(t); t \in [t_0, T]) = \phi(x(T), T) + \int_{t_0}^T L(x(t), u(t), t) dt.$$

Note: i) for a given time t , $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^m$

ii) J is a functional, i.e., it is a function of functions.

input to J are functions $x(t)$, $u(t)$, $t \in [t_0, T]$

output of J is a real-number.

The above problem can be written as

$$\begin{aligned} \min_{\substack{x(t), u(t), \\ t \in [t_0, T]}} & \phi(x(T), T) + \int_{t_0}^T L(x(t), u(t), t) dt \\ \text{s.t.} & \underline{\dot{x}(t) = f(x(t), u(t), t)} && \text{(CT-OPT)} \\ & \psi(x(T), T) = 0 \\ & x(t_0) = x_0 \text{ is known.} \end{aligned}$$

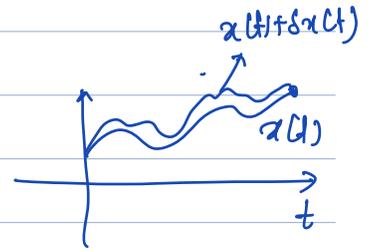
decision variables: $x(t)$, $t \in [t_0, T]$
 $u(t)$, $t \in [t_0, T]$

In order to analyze optimal solutions of the above problem, we need to define quantities for functionals that are analogous to notions of derivative/gradient, Hessian, etc for functions.

$$f(x(t)) = 2x^2(t) + 1$$

$$\frac{\partial f}{\partial x} = 4x(t)$$

ex: $J_1(x(t)) = \int_{t_0}^T [2x^2(t) + 1] dt$



consider a perturbed signal $x(t) + \delta x(t)$

Increment $\Delta J(x(t), \delta x(t)) := J(x(t) + \delta x(t)) - J(x(t))$

$\approx \delta J(x(t), \delta x(t))$ for $\delta x(t)$ sufficiently small.

$\Delta J_1(x(t), \delta x(t))$

$$= \int_{t_0}^T [2(x(t) + \delta x(t))^2 + 1] dt - \int_{t_0}^T [2x(t)^2 + 1] dt$$

$(\delta J)(\delta x(t))$

Recall: (Taylor series)

$$f(x + \Delta x) \approx f(x) + \nabla f(x)^T \Delta x + \frac{1}{2} \Delta x^T H(x) \Delta x + h.o.t.$$

$$= \int_{t_0}^T [4x(t)\delta x(t) + 2\delta x(t)^2] dt$$

$\underbrace{4x(t)\delta x(t)}_{\text{linear in } \delta x(t)} + \underbrace{2\delta x(t)^2}_{\text{quadratic}}$

if $x(t) \in \mathbb{R}^n$
 $\delta x(t) \in \mathbb{R}^n$

$(\cdot)(\delta x(t))_1 + (\cdot)(\delta x(t))_2$

(First) variation of J at $x(t)$ is $\frac{\partial f}{\partial x}$

$$\delta J(x(t), \delta x(t)) = \int_{t_0}^T 4x(t) \delta x(t) dt$$

$$\delta J = \int \frac{D_x J}{Dx(t)} \delta x(t) dt$$

second variation $\delta^2 J(x(t), \delta x(t)) = \int_{t_0}^T 2\delta x(t)^2 dt$

Norm: Given a vector space \mathcal{V} , norm $\|\cdot\|: \mathcal{V} \rightarrow \mathbb{R}$ s.t.

i) $\|x\| \geq 0 \quad \forall x \in \mathcal{V}$, and $\|x\| = 0 \Leftrightarrow x = 0$

ii) $\|\alpha x\| = |\alpha| \|x\|$ for any $\alpha \in \mathbb{R}$, $x \in \mathcal{V}$

iii) $\|x_1 + x_2\| \leq \|x_1\| + \|x_2\|$, $\forall x_1, x_2 \in \mathcal{V}$.

If \mathcal{V} is the space of functions defined over $[0, T]$,

$$\|x(t)\|_2 = \left[\int_0^T x(t)^T x(t) dt \right]^{\frac{1}{2}}$$

$$\|x(t)\|_1 = \int_0^T \|x(t)\|_1 dt, \text{ and so on.}$$

zero element $x(t) = 0 \quad \forall t \in [0, T]$

Local minimum: $x^*(t)$ is a local minimizer of the functional $J(x(t))$

if $J(x(t)) \geq J(x^*(t)) \quad \forall x(t)$ s.t.

$$\|x(t) - x^*(t)\|_2 \leq \epsilon$$

for some $\epsilon > 0$.

or there does not exist any signal $x(t)$ within distance ϵ from $x^*(t)$ which has a strictly smaller functional value.

Fundamental Theorem of Calculus of Variations (analogous FONC)

Let $J(x(t))$ be a differentiable functional of $x(t)$. If $x^*(t)$ is an extremal function, then

$$\delta J(x^*(t), \delta x(t)) = 0 \quad \text{for all admissible } \delta x(t).$$

Sufficient condition for x^* to be a minimizer is $\delta^2 J(x^*) > 0$
 sufficient condition for x^* to be a maximizer is $\delta^2 J(x^*) < 0$.

Leibniz Rule: Consider a functional $J(x(t)) = \int_{t_0}^T f(x(t)) dt$.
 The first variation of J is given by:

$$\delta J(\bar{x}(t), \delta \bar{x}(t)) = \int_{t_0}^T \left[\frac{\partial f}{\partial x}(\bar{x}(t)) \cdot \delta \bar{x}(t) \right] dt + f(\bar{x}(T)) \underline{dT} - f(\bar{x}(t_0)) \underline{dt_0}$$

→ Recall differentiation under integral sign.

Basic Variational Problem

(analogous to unconstrained optimization, precursors to solving CT-OPT)

continuously differentiable

Find $x(t)$, $t \in [t_0, T]$ to minimize

$$J(x(t)) = \int_{t_0}^T V(x(t), \dot{x}(t), t) dt,$$

with $t_0, T, x(t_0) = x_0$ and $x(T) = x_f$ are given/fixed.

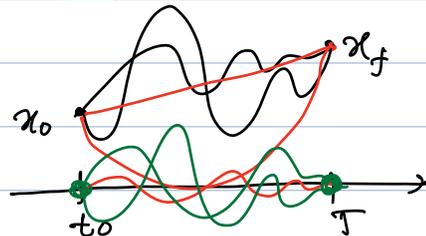
Let $x^*(t)$ be a candidate minimizer.

Then

$$\delta J(x^*(t), \delta x(t)) = 0$$

$\forall \delta x(t)$ s.t. $x^*(t) + \delta x(t)$ must satisfy the boundary conditions.

$$\Rightarrow \underline{\delta x(t) = 0 \text{ at } t=t_0 \text{ and } t=T.}$$



Lecture 21 & 22: 2nd Marsch

Let us compute first variation using Leibniz rule.

$$\delta J(x^*(t), \delta x(t)) = \int_{t_0}^T \left[\frac{\partial V}{\partial x}(x^*(t), \dot{x}^*(t), t) \delta x(t) + \frac{\partial V}{\partial \dot{x}}(x^*(t), \dot{x}^*(t), t) \delta \dot{x}(t) \right] dt \\ + \underbrace{V(x^*(T), \dot{x}^*(T), T)}_{dT} - \underbrace{V(x^*(t_0), \dot{x}^*(t_0), t_0)}_{dt_0}$$

Since T and t_0 are fixed, $dT = dt_0 = 0$ for this problem.

$$\Rightarrow \delta J(x^*(t), \delta x(t)) = \int_{t_0}^T \left[\frac{\partial V}{\partial x}(x^*(t), \dot{x}^*(t), t) \delta x(t) + \frac{\partial V}{\partial \dot{x}}(x^*(t), \dot{x}^*(t), t) \delta \dot{x}(t) \right] dt$$

apply integration by parts.

Note that $\delta \dot{x}(t)$ is not independent of $\delta x(t)$.

$$\int_{t_0}^T \frac{\partial V}{\partial \dot{x}}(x^*(t), \dot{x}^*(t), t) \delta \dot{x}(t) dt = \frac{\partial V}{\partial \dot{x}}(x^*(t), \dot{x}^*(t), t) \delta x(t) \Big|_{t_0}^T \\ - \int_{t_0}^T \frac{d}{dt} \left(\frac{\partial V}{\partial \dot{x}}(x^*(t), \dot{x}^*(t), t) \right) \delta x(t) dt$$

Consequently, we obtain

$$\delta J(x^*(t), \delta x(t)) = \int_{t_0}^T \left[\frac{\partial V}{\partial x}(x^*(t), \dot{x}^*(t), t) - \frac{d}{dt} \left(\frac{\partial V}{\partial \dot{x}}(x^*(t), \dot{x}^*(t), t) \right) \right] \delta x(t) dt \\ + \frac{\partial V}{\partial \dot{x}}(x^*(T), \dot{x}^*(T), T) \delta x(T) \\ - \frac{\partial V}{\partial \dot{x}}(x^*(t_0), \dot{x}^*(t_0), t_0) \delta x(t_0)$$

when $x(T)$ and $x(t_0)$ are fixed, $\delta x(T) = \delta x(t_0) = 0$, and the last two terms are equal to 0.

Thus, if $x^*(t)$ is an extremal signal,

$$\delta J(x^*(t), \delta x(t)) = \int_{t_0}^T \left[\frac{\partial V}{\partial x} \Big|_* - \frac{d}{dt} \left(\frac{\partial V}{\partial \dot{x}} \Big|_* \right) \right] \delta x(t) dt = 0$$

+ admissible $\delta x(t)$.

In practice, we do not need to check for infinitely many $\delta x(t)$.

Fundamental Lemma: If $g(t)$ is a continuous function and

$$\int_{t_0}^T g(t) \delta x(t) dt = 0 \quad \forall \delta x(t) \text{ that are continuous, then}$$

$$g(t) = 0 \quad \forall t \in [t_0, T].$$

proof can be done by contradiction.



Applying the fundamental lemma, we have the following necessary condition for $x^*(t)$ to be an extremal signal.

$$\frac{\partial V}{\partial x} (x^*(t), \dot{x}^*(t), t) - \frac{d}{dt} \left(\frac{\partial V}{\partial \dot{x}} (x^*(t), \dot{x}^*(t), t) \right) = 0$$

+ $t \in [t_0, T]$.

(Euler-Lagrange equation) $\quad \frac{\partial^2 V}{\partial x \partial \dot{x}} \frac{dx}{dt} + \frac{\partial^2 V}{\partial \dot{x}^2} \frac{d\dot{x}}{dt} + \frac{\partial V}{\partial t \partial x}$

Example: $J(x(t)) = \int_0^2 [\dot{x}(t)^2 - 2tx(t)] dt$

$$x(0) = 1, \quad x(2) = 5$$

$\ddot{x}(t)$
↓
2nd order
diff. eqn

here $V(x(t), \dot{x}(t), t) = \dot{x}(t)^2 - 2tx(t)$

$$\frac{\partial V}{\partial x} = -2t, \quad \frac{\partial V}{\partial \dot{x}} = 2\dot{x}(t), \quad \frac{d}{dt} \frac{\partial V}{\partial \dot{x}} = 2\ddot{x}(t)$$

EL eqn:

$$-2t - 2\ddot{x}(t) = 0$$

$$\Rightarrow \ddot{x}(t) = -t$$

$$\Rightarrow \dot{x}(t) = -\frac{t^2}{2} + C_1$$

$$\Rightarrow x(t) = -\frac{t^3}{6} + C_1 t + C_2$$

$$x(0) = C_2 = 1$$

$$x(2) = -\frac{8}{6} + 2C_1 + 1 = 5$$

$$\Rightarrow 2C_1 = 4 + \frac{4}{3} = \frac{16}{3} \Rightarrow C_1 = \frac{8}{3}$$

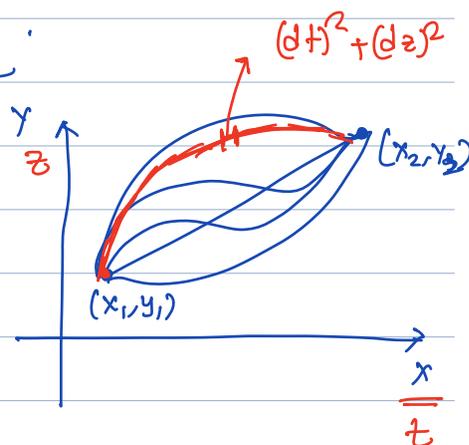
$$x(t) = -\frac{t^3}{6} + \frac{8}{3}t + 1$$

Example

We can express the length of a curve

as

$$J = \int_{x_1}^{x_2} \sqrt{(dt)^2 + (dz)^2} = \int_{x_1}^{x_2} \sqrt{1 + \left(\frac{dz}{dt}\right)^2} dt$$



$$J(z(t)) = \int_{x_1}^{x_2} \underbrace{\left(1 + \dot{z}(t)^2\right)^{\frac{1}{2}}}_{v(z(t), \dot{z}(t), t)} dt$$

$$v(z(t), \dot{z}(t), t) = \left(1 + \dot{z}(t)^2\right)^{\frac{1}{2}}$$

Let $\dot{z}^*(t)$ satisfy the EL eqn.

$$\frac{\partial v}{\partial z} = 0$$

$$\frac{\partial v}{\partial \dot{z}} = \frac{1}{2} \left(1 + \dot{z}(t)^2\right)^{-\frac{1}{2}} \times 2 \dot{z}(t)$$

EL eqn:

$$0 - \frac{d}{dt} \left(\frac{\dot{z}(t)}{\left(1 + \dot{z}(t)^2\right)^{\frac{1}{2}}} \right) = 0 \quad \forall t$$

$$\Rightarrow \frac{\dot{z}(t)}{\sqrt{1 + \dot{z}(t)^2}} = C \quad \forall t \in [x_1, x_2]$$

$$\Rightarrow \dot{z}(t) = C (1 + \dot{z}(t)^2)^{\frac{1}{2}}$$

$$\Rightarrow \dot{z}(t)^2 [1 - C^2] = C$$

$$\Rightarrow \dot{z}(t)^2 = \bar{C}$$

$$\Rightarrow \underline{\dot{z}(t) = \sqrt{\bar{C}}}$$

$$\Rightarrow \boxed{z(t) = C_1 t + C_2}$$

\Rightarrow extremal signal is a straight line,
with C_1, C_2 determined by boundary conditions.

Let us now generalize the above setting.

Let $\underline{x(t)} \in \mathbb{R}^n$, $t_0, t_f, \underline{x(t_0)}$: fixed

$\underline{x(t_f)}$: free variable.

$$J = \int_{t_0}^{t_f} g(\underline{x}(t), \dot{\underline{x}}(t), t) dt \in \mathbb{R}$$

$$\delta J = \int_{t_0}^{t_f} \left[\underbrace{g_x(\underline{x}(t), \dot{\underline{x}}(t), t)}_{\text{row vector}} \delta \underline{x}(t) + \underbrace{g_{\dot{x}}(\underline{x}(t), \dot{\underline{x}}(t), t)}_{\text{column vector}} \delta \dot{\underline{x}}(t) \right] dt$$

$$\underbrace{\begin{bmatrix} \frac{\partial g}{\partial x_1}(\cdot) & \frac{\partial g}{\partial x_2}(\cdot) & \dots & \frac{\partial g}{\partial x_n}(\cdot) \end{bmatrix}}_{\nabla_x g(\cdot)^T} \cdot \begin{bmatrix} \delta x_1(t) \\ \delta x_2(t) \\ \vdots \\ \delta x_n(t) \end{bmatrix}$$

$$g_{\dot{x}} \delta \underline{x}(t) \Big|_{t_0}^{t_f} - \int_{t_0}^{t_f} \frac{d}{dt} g_{\dot{x}} \delta \underline{x}(t) dt$$

once again, we apply integration by parts to the second term to obtain

$$\delta J = \int_{t_0}^{t_f} \left[g_x - \frac{d}{dt} (g_{\dot{x}}) \right] \delta \underline{x}(t) dt + g_{\dot{x}}(\underline{x}(t_f), \dot{\underline{x}}(t_f), t_f) \delta \underline{x}(t_f)$$

Boundary condition: If $\underline{x}(t_f) = \underline{x}_f$ fixed, then $\delta \underline{x}(t_f) = 0$

If $\underline{x}(t_f)$ is free, then $\delta \underline{x}(t_f) \neq 0$

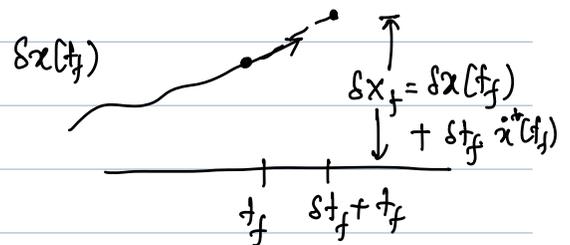
If $x^*(t)$ is an optimal solution, $\delta J(x^*(t), \delta x(t)) = 0$
 \forall admissible $\delta x(t)$.

$$\Rightarrow g_x(x^*(t), \dot{x}^*(t), t) - \frac{d}{dt} g_{\dot{x}}(x^*(t), \dot{x}^*(t), t) = 0 \quad \forall t \in [t_0, t_f]$$

$$g_{\dot{x}}(x^*(t_f), \dot{x}^*(t_f), t_f) = 0$$

Let us now consider the case where both t_f and $x(t_f)$ are free.

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



applying Leibniz rule, we obtain:

$$\delta J = \int_{t_0}^{t_f} (g_x \delta x(t) + g_{\dot{x}} \delta \dot{x}(t)) dt + g(x(t_f), \dot{x}(t_f), t_f) \delta t_f$$

$$= \int_{t_0}^{t_f} \left(g_x - \frac{d}{dt} g_{\dot{x}} \right) \delta x(t) dt + g(x(t_f), \dot{x}(t_f), t_f) \delta t_f$$

$$+ g_{\dot{x}}(x(t_f), \dot{x}(t_f), t_f) \delta x(t_f),$$

where $\delta x_f = \delta x(t_f) + \dot{x}(t_f) \delta t_f$: assuming δx_f and δt_f are free and independent.

$$= \int_{t_0}^{t_f} \left(g_x - \frac{d}{dt} g_{\dot{x}} \right) \delta x(t) dt + g_{\dot{x}}(x(t_f), \dot{x}(t_f), t_f) \delta x_f$$

$$+ \left[g(x(t_f), \dot{x}(t_f), t_f) - g_{\dot{x}}(x(t_f), \dot{x}(t_f), t_f) \dot{x}(t_f) \right] \delta t_f$$

If $x^*(t)$ is an extremal signal, $g_x - \frac{d}{dt} g_{\dot{x}} \Big|_{x^*} = 0$ (EL eqn)

$$g_{\dot{x}} \Big|_{x^*(t_f)} = 0, \quad (g - g_{\dot{x}} \dot{x}) \Big|_{x^*(t_f)} = 0$$

Let us now add constraints.

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

$$\text{s.t. } \underline{f(x(t), \dot{x}(t), t) = 0},$$

$x(t_0), x(t_f), t_0, t_f$ are fixed.

sometimes, there is a terminal constraint

$$m(x(t_f), t_f) = 0:$$

$v \in \mathbb{R}^{k_2}$
 v is not a fⁿ of time $\leftarrow v \in \mathbb{R}^{k_2}$

Recall: for conventional optimization problem: $\min_x g(x)$
 s.t. $f(x) = 0,$

$$L(x, \lambda) = g(x) + \lambda^T f(x).$$

optimality condition: $\begin{cases} \nabla_x L(x, \lambda) = 0 \Rightarrow g_x(x) + \lambda^T f_x(x) = 0 \\ \nabla_\lambda L(x, \lambda) = 0 \Rightarrow f(x) = 0 \end{cases}$

earlier when $f(x) \in \mathbb{R}^m, \lambda \in \mathbb{R}^m$.

Now, $f(x(t), \dot{x}(t), t) \in \mathbb{R}^k$, we need a multiplier $p(t) \in \mathbb{R}^k$, and $p(t)$ should be a function of time t .

Let us now define augmented cost/Lagrangian for the optimization problem stated above.

$$J'(x(t), \dot{x}(t), p(t)) = \int_{t_0}^{t_f} (g(x(t), \dot{x}(t), t) + p(t)^T \underline{f(x(t), \dot{x}(t), t)}) dt$$

$$\delta J' = \int_{t_0}^{t_f} [\quad] \delta x(t) + (f^T) \delta p(t) dt$$

applying Leibniz rule:

$$\int_{t_0}^{t_f} (g_x \delta x + g_{\dot{x}} \delta \dot{x} + p^T f_x \delta x + p^T f_{\dot{x}} \delta \dot{x} + f^T \delta p) dt$$

$$= \int_{t_0}^{t_f} (g_x - \frac{d}{dt} g_{\dot{x}} + p^T f_x - \frac{d}{dt} (p^T f_{\dot{x}}) + f^T \delta p) dt$$

First order necessary condition differential eqⁿ in both x & p .

$$(g_x + p^T f_x) - \frac{d}{dt}(g_x + p^T f_x) = 0 \quad \forall t$$

$$f(x, \dot{x}, t) = 0 \quad \forall t$$

Constraint present in the problem.

Other types of boundary conditions can be handled as discussed above.

Example: $\min_{x(t), u(t)} \int_0^1 (x^2(t) + u^2(t)) dt$

s.t. $\dot{x}(t) = -x(t) + u(t)$

$x(0) = 1, x(1) = 0$

Express the FONC in terms of differential eqⁿ in $x(t)$ & multiplier $p(t)$, and solve for $x(t)$.

Augmented cost:

$$J' = \int_0^1 (x^2(t) + u^2(t) + p(t) [\dot{x}(t) + x(t) - u(t)]) dt$$

$$\delta J' = \int_0^1 \left(\left(\begin{matrix} 2x(t) \\ 2u(t) - p(t) \end{matrix} \right) \delta x(t) + \left(\begin{matrix} 2u(t) - p(t) \\ \dot{x}(t) + x(t) - u(t) \end{matrix} \right) \delta u(t) \right) dt$$

$$\left(2x(t) + p(t) - \frac{d}{dt} p(t) \right) \delta x(t) + \left(\begin{matrix} 2u(t) - p(t) \\ \dot{x}(t) + x(t) - u(t) \end{matrix} \right) \delta p(t) dt$$

FONC:

$$2x(t) + p(t) - \dot{p}(t) = 0 \Rightarrow \boxed{\dot{p}(t) = 2x(t) + p(t)} \text{ costate equation.}$$

$$2u(t) - p(t) = 0 \Rightarrow u(t) = p(t)/2$$

$$\dot{x}(t) + x(t) - u(t) = 0 \Rightarrow \boxed{\dot{x}(t) = -x(t) + \frac{1}{2} p(t)}$$

In this case, since $x(t_f)$ is given, we have no boundary condition on costate $p(t)$.
So, to solve for $x(t)$, we need to eliminate $p(t)$.

$$\begin{aligned}\ddot{x}(t) &= -\dot{x}(t) + \frac{1}{2} \dot{p}(t) \\ &= -\dot{x}(t) + \frac{1}{2} (2x(t) + p(t)) \\ &= -\dot{x}(t) + x(t) + \frac{p(t)}{2} \\ &= -\dot{x}(t) + x(t) + \dot{x}(t) + x(t) = \underline{\underline{2x(t)}}$$

so, extremal $x(t)$ will satisfy $\ddot{x}(t) = 2x(t)$
with boundary conditions $x(0) = 1, x(1) = 0$.

$$u(t) = \dot{x}(t) + x(t)$$

Lecture 23 & 24: 3rd March

- optimality conditions for continuous-time optimal control
(taught on board)

- Notes by Dantu Phani Susya appended below.

$$\rightarrow 2x + p - \frac{dp}{dt} = 0 \Rightarrow \dot{p} = p + 2x \rightarrow \text{co state}$$

$$2u - p = 0 \Rightarrow u = p/2$$

$$2 + \dot{x} - u = 0 \Rightarrow \dot{x} = -x + \frac{p}{2} \rightarrow \text{state}$$

In this case, since $x(t_f)$ is given, we have no boundary condition on p .

$$\rightarrow \ddot{x} = -\dot{x} + \frac{\dot{p}}{2} = -\dot{x} + \frac{p+2x}{2} = -\dot{x} + x + \frac{p}{2} = -\dot{x} + x + x + \dot{x} = 2x$$

$$\Rightarrow \ddot{x} = 2x \Rightarrow x = Ae^{-\sqrt{2}t} + Be^{\sqrt{2}t} = 2x$$

Any extremal $x(t)$ will satisfy $\ddot{x} = 2x$, with boundary conditions.

$$\Rightarrow u(t) = \dot{x}(t) + x(t)$$

Continuous time optimal control

$$\min_{x(t), u(t)} J(x(t), u(t)) = s(x(t_f), t_f) + \int_{t_0}^{t_f} v(x(t), u(t), t) dt$$

$$\text{s.t. } \dot{x}(t) = f(x(t), u(t), t)$$

$$t_0, x(t_0) : \text{given}$$

$$t_f, x(t_f) : \text{free}$$

$p(t)$: multipliers for Lagrangian

minimising J is equivalent to minimising J_2 .

$$J_2(x(t), u(t)) = \int_{t_0}^{t_f} \left[v(x(t), u(t), t) + \underbrace{\frac{ds}{dt}(x(t), t)} \right] dt$$

$$s(x(t_f), t_f) - \underbrace{s(x(t_0), t_0)}_{\text{constant}}$$

$$= \int_{t_0}^{t_f} \left[v(x(t), u(t), t) + \frac{\partial s}{\partial x}(x(t), t) \dot{x}(t) + \frac{\partial s}{\partial t}(x(t), t) \right] dt$$

Augmented cost: $J_2' = \int_{t_0}^{t_f} \left[\overbrace{V(x(t), u(t), t) + p(t)^T (f(x(t), u(t), t) - \dot{x}(t))}^{\text{Hamiltonian}} + \frac{\partial S}{\partial x}(x(t), t) \dot{x}(t) + \frac{\partial S}{\partial t}(x(t), t) \right] dt$

$$= \int_{t_0}^{t_f} \left[H - p^T \dot{x} + \frac{\partial S}{\partial x} \dot{x} + \frac{\partial S}{\partial t} \right] dt = \int_{t_0}^{t_f} L(x, \dot{x}, u, p, t) dt$$

Necessary condition for $x^*(t), u^*(t)$ to be the optimal signal is $\delta J_2' = 0$

$$\Rightarrow \int_{t_0}^{t_f} \left[\frac{\partial L}{\partial x} \delta x + \frac{\partial L}{\partial \dot{x}} \delta \dot{x} + \frac{\partial L}{\partial u} \delta u \right] dt + L \Big|_{t_f} \delta t_f$$

Boundary condition

$$= \int_{t_0}^{t_f} \left[\left(\frac{\partial L}{\partial x} - \frac{d}{dt} \frac{\partial L}{\partial \dot{x}} \right) \delta x + \frac{\partial L}{\partial u} \delta u + \frac{\partial L}{\partial p} \delta p \right] dt + L \Big|_{t_f} \delta t_f + \frac{\partial L}{\partial \dot{x}} \delta x(t_f) = 0$$

$$L = V(x(t), u(t), t) + p(t)^T f(x(t), u(t), t) - p(t)^T \dot{x}(t) + \frac{\partial S}{\partial x}(x(t), t) \dot{x}(t) + \frac{\partial S}{\partial t}(x(t), t)$$

At optimal solution, $\frac{\partial L}{\partial u} = 0 \Rightarrow \frac{\partial V}{\partial u} + \frac{\partial f}{\partial u}(x(t), u(t), t) \cdot p(t) = \frac{\partial H}{\partial u}(x(t), u(t), p(t), t) = 0$

$$\frac{\partial L}{\partial p} = 0 \Rightarrow f(x(t), u(t), t) - \dot{x}(t) = 0 \quad \text{--- (2)} \quad \text{--- (1)}$$

Now, $\frac{\partial L}{\partial x} = \frac{\partial V}{\partial x} + \frac{\partial f}{\partial x} \cdot p + \frac{\partial^2 S}{\partial x^2} \dot{x} + \frac{\partial^2 S}{\partial x \partial t}$

$$\frac{\partial L}{\partial \dot{x}} = -p^T + \frac{\partial S}{\partial x} \Rightarrow \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{x}} \right) = -\dot{p}^T + \frac{\partial^2 S}{\partial x^2} \dot{x} + \frac{\partial^2 S}{\partial x \partial t}$$

$$\Rightarrow \frac{\partial L}{\partial x} - \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{x}} \right) = \frac{\partial V}{\partial x} + \frac{\partial f}{\partial x} \cdot p + \dot{p}^T = 0 \quad \forall t$$

$$\Rightarrow \dot{p}(t) = - \left(\frac{\partial H}{\partial x} \right)^T \quad \text{--- (3)}$$

Boundary conditions at t_f :

$$\left[H(t_f) - p(t_f)^T \dot{x}(t_f) + \frac{\partial S}{\partial x} \Big|_{t_f} \dot{x}(t_f) + \frac{\partial S}{\partial t} \Big|_{t_f} \right] \delta t_f + \frac{\partial S}{\partial x} \Big|_{t_f} \delta x(t_f) - p(t_f)^T \delta x(t_f)$$

$$\Rightarrow \underbrace{\left(H + \frac{\partial S}{\partial t} \right) \Big|_{t_f}}_{=0 \text{ (4)}} \delta t_f + \underbrace{\left(\frac{\partial S}{\partial x} - p^T \right) \Big|_{t_f}}_{=0 \text{ (5)}} \underbrace{(\delta x(t_f) + \dot{x}(t_f) \delta t_f)}_{\delta x_f = \text{perturbation in final value at } t_f + \delta t_f}$$

Optimality conditions

$$H(x, u, p, t) = V(x, u, t) + p^T f(x, u, t)$$

$$\left. \begin{array}{l} \frac{\partial H}{\partial u} = 0 \quad \text{--- (1)} \\ \dot{x} = f(x, u, t) \quad \text{--- (2)} \\ \dot{p} = - \left(\frac{\partial H}{\partial x} \right)^T \quad \text{--- (3)} \end{array} \right\} \forall t \in [t_0, t_f]$$

$$\left. \begin{array}{l} \left(H + \frac{\partial S}{\partial t} \right) \Big|_{t_f} \delta t_f = 0 \quad \text{--- (4)} \\ \left(\frac{\partial S}{\partial x} - p^T \right) \Big|_{t_f} \cdot \delta x_f = 0 \quad \text{--- (5)} \end{array} \right\} \text{Boundary conditions}$$

Special cases:

(a) free final state, free final time

$$H + \frac{\partial S}{\partial t} = 0, \quad \frac{\partial S}{\partial x} = p^T$$

(b) free final state, fixed final time

$$t_f: \text{known}, \quad \delta t_f = 0, \quad \frac{\partial S}{\partial x} = p^T$$

(c) fixed final state, free final time

$$H + \frac{\partial S}{\partial t} = 0, \quad x_f: \text{known}, \quad \delta x_f = 0$$

(d) fixed final state, fixed final time

$$t_f, x_f: \text{known}, \quad \delta x_f = \delta t_f = 0$$

(e) free final time, final state

depends on final time $x(t_f) = \theta(t_f) \Rightarrow \delta x_f = \dot{\theta}(t_f) \delta t_f$

$$\Rightarrow \left[H + \frac{\partial S}{\partial t} + \left(\frac{\partial S}{\partial x} - p^T \right) \dot{\theta} \right] \Big|_{t_f} = 0$$

Sufficient condition for optimal signal

$$\delta^2 J_2' = \int_{t_0}^{t_f} \begin{bmatrix} \delta x^T & \delta u^T \end{bmatrix} \underbrace{\begin{bmatrix} \frac{\partial^2 H}{\partial x^2} & \frac{\partial^2 H}{\partial x \partial u} \\ \frac{\partial^2 H}{\partial x \partial u} & \frac{\partial^2 H}{\partial u^2} \end{bmatrix}} \begin{bmatrix} \delta x(t) \\ \delta u(t) \end{bmatrix} dt > 0 \quad \forall \text{ permissible perturbations}$$

should be a positive definite matrix $\forall t \in [t_0, t_f]$ at the optimal solution.

Example

1. $\min_{u(t)} \frac{1}{2} \int_0^2 u^2(t) dt$

s.t. $\dot{x}_1(t) = x_2(t) \quad x(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad x(2) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$
 $\dot{x}_2(t) = u(t)$

2. $\min_{u(t)} \frac{1}{2} \int_{t_0}^{t_f} u^2(t) dt$

s.t. $\dot{x}_1(t) = x_2(t) \quad x(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad x_2(2) : \text{free}$
 $\dot{x}_2(t) = u_2(t) \quad x_1(2) = 0$

3. $\min \frac{1}{2} (x_1(2) - 4)^2 + \frac{1}{2} (x_2(2) - 2)^2 + \frac{1}{2} \int_0^2 u^2(t) dt$

$x(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad x(2) = \text{free}$
 $t_f = 2 \text{ fixed}$

1. $H = V + p^T f = \frac{u^2}{2} + p_1 x_2 + p_2 u$

$\frac{\partial H}{\partial u} = u + p_2 = 0$

$\frac{\partial H}{\partial x} = \begin{bmatrix} 0 & p_1 \end{bmatrix} \Rightarrow \dot{p} = \begin{bmatrix} 0 \\ -p_1 \end{bmatrix} \Rightarrow \begin{cases} \dot{p}_1 = 0 \Rightarrow p_1 = c_1 \\ \dot{p}_2 = -p_1 \Rightarrow p_2 = -c_1 t + c_2 \end{cases}$

$\Rightarrow u = c_1 t - c_2$

$\dot{x}_2 = u \Rightarrow x_2 = c_1 \frac{t^2}{2} - c_2 t + c_3$

$\dot{x}_1 = x_2 \Rightarrow x_1 = \frac{c_1}{2} \frac{t^3}{3} - \frac{c_2}{2} t^2 + c_3 t + c_4$

$x_1(0) = 1 \Rightarrow c_4 = 1, \quad x_2(0) = 2 \Rightarrow c_3 = 2, \quad x_1(2) = 1, \quad x_2(2) = 0 \rightarrow \text{give } c_1, c_2$

$$2. \quad \delta x_f \neq 0 \Rightarrow \left. \frac{\partial S}{\partial x_2} - p_2 \right|_{t=2} = 0 \Rightarrow p_2(2) = 0$$

$$\Rightarrow 2c_1 = c_2 \quad \text{and} \quad x_1(2) = 0 \rightarrow \text{give } c_1, c_2$$

$$3. \quad H = \frac{u^2}{2} + p_1 x_2 + p_2 u$$

$$\Rightarrow \frac{\partial H}{\partial u} = 0 \Rightarrow u + p_2 = 0$$

$$\dot{p}_1 = -\frac{\partial H}{\partial x_1} = 0 \Rightarrow p_1 = c_1$$

$$\dot{p}_2 = -\frac{\partial H}{\partial x_2} = -p_1 \Rightarrow p_2 = -c_1 t + c_2$$

$$\delta x_f \neq 0 \Rightarrow \left. \frac{\partial S}{\partial x_1} - p_1 \right|_2 = 0, \quad \left. \frac{\partial S}{\partial x_2} - p_2 \right|_2 = 0$$

$$x_1(2) - 4 - p_1 = 0 \Rightarrow x_1(2) = 4 + c_1$$

$$x_2(2) - 2 - p_2 = 0 \Rightarrow x_2(2) = 2 - 2c_1 + c_2$$

$$x_1(0) = 1$$

$$x_2(0) = 2$$

} 4 eqⁿs 4 vars
can be solved