

$$t_f \equiv T$$

Lecture #6

Cost function/objective

$$t_f = T$$

$\min_{U(\cdot)}$

$$J(u) = \underbrace{\phi(x(T), T)}_{\text{terminal cost}} + \int_{t_0=0}^T \underbrace{L(x, u, t)}_{\text{"cost-to-go"}} dt$$

Bolza form (general form)
for OCP

Lagrange form

(when $\phi(\cdot) \equiv 0$)
(only cost-to-go)

Mayer form

($L \equiv 0$)
(only terminal cost)

equivalent

Equivalence betw. forms :

Mayer form \rightarrow Lagrange form

$$\phi(x(t), t) \stackrel{t_f}{=} \phi(t_0, x_0) + \int \frac{d}{dt} \phi(t, x(t)) dt$$

$$= \underbrace{\phi(t_0, x_0)}_{\text{indep. of } u(\cdot)} + \int_{t_0}^{t_f} \left\{ \frac{\partial \phi}{\partial t} + \langle \nabla_x \phi, \dot{x} \rangle \right\} dt$$

$\underset{u(\cdot)}{\text{argmin}}$ $\phi(x(t), t)$
Mayer form.

$= \underset{u(\cdot)}{\text{argmin}} \int_{t_0}^{t_f} \left\{ \frac{\partial \phi}{\partial t} + \langle \nabla_x \phi, \dot{x} \rangle \right\} dt$

L
Lagrange form



Lagrange form \rightarrow Mayer form

Introduce extra state variable \tilde{x} as

$$\dot{\tilde{x}} = L(t, \underline{x}, \underline{u})$$

$\tilde{x}(t_0)$ = arbitrary constant

(only changes the cost by

an additive constant,

does NOT change $\arg\min$)

$$\therefore \int_{t_0}^{t_f} L(t, \underline{x}, \underline{u}) dt = \tilde{x}(t_f)$$

(plays the role of ϕ)
by construction



First order Necessary Conditions for Optimality (1958)

<u>OPT</u>	<u>CoV</u>	<u>OCP</u> (Pontryagin & Boltyanski)
KKT condition	EL equation	<u>Non</u>

Hamiltonian $H(\underline{x}(t), \underline{u}(t), \underline{\lambda}(t), t)$ (scalar)

$$:= L(\underline{x}(t), \underline{u}(t), t) + \underline{\lambda}^T(t) f(\underline{x}(t), \underline{u}(t), t)$$

State eqⁿ:

$$\dot{\underline{x}}(t) = \frac{\partial H}{\partial \underline{\lambda}} = f(\underline{x}, \underline{u}(t), t), \quad \underline{x}(t) \in \mathbb{R}^n$$

$$\underline{x}(0) = \underline{x}_0 \text{ (given)}$$

Co-state eqⁿ:

$$\dot{\underline{\lambda}}(t) = - \frac{\partial H}{\partial \underline{x}}, \quad \underline{\lambda}(t) \in \mathbb{R}^n$$

(time-varying) Lagrange \uparrow multiplier/ trajectories

We call $\underline{\lambda}(t)$ as co-state

③ Pontryagin's Maximum Principle (PMP) :

$$\frac{\partial \mathcal{L}}{\partial x_1} = \frac{\partial H}{\partial u}, \quad \underline{u} \in \mathbb{R}^m$$

④ Transversality Condition:

$$\left(\nabla_{\underline{x}} \phi + (\nabla_{\underline{x}} \Psi)^T \underline{v} - \lambda \right)^T \Big|_{t=t_f=T} d\underline{x}(T) + \underbrace{d\underline{x}(T)}_{\text{final state}}$$

$$\left(\frac{\partial \phi}{\partial t} + \left(\frac{\partial \Psi}{\partial t} \right)^T \underline{v} + H \right) \Big|_{t=T} d\underbrace{T}_{\text{final time}} = 0$$

\underline{v} : Lagrange multiplier vector
but constant vector

Hamiltonian H :

$$\frac{d}{dt} H(\underline{x}, \underline{u}, \underline{\lambda}, t)$$

$$= \frac{\partial H}{\partial t} + (\nabla_{\underline{x}} H)^T \cancel{\underline{x}} + (\nabla_{\underline{u}} H)^T \underline{u} + (\underline{\lambda})^T f$$

(chain rule)

$$= \frac{\partial H}{\partial t} + (\nabla_{\underline{u}} H)^T \underline{u} + (\underbrace{\nabla_{\underline{x}} H + \underline{\lambda}}_0)^T f$$

(from PMP)
(Condition 3)

$\equiv 0$
(Condition 2)

$$= \frac{\partial H}{\partial t}$$

$\Rightarrow H^*$ is constant if the OCP is time-invariant
(i.e., neither f nor L depends explicitly on t)

Proof of (1)-(4) : summary / outline : Book by Liberzon.

Actual proof : Agrachev - Sachkov \checkmark
we will not prove it

Example 1 : (Shortest planar path is straight line, revisited)

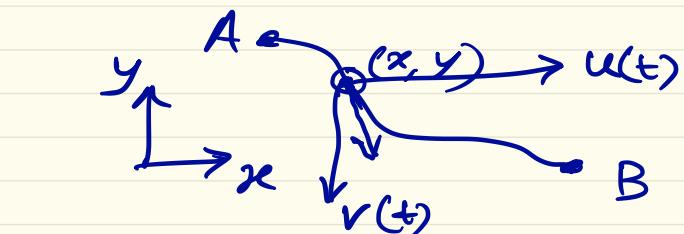
① State eqns

$$\dot{x} = u$$

$$\dot{y} = v$$

$$B \int \sqrt{1 + \left(\frac{dy}{dx}\right)^2} dx$$

$$A = \sqrt{1 + \frac{v^2}{u^2}} dt$$



2 states, 2 controls

$$\phi = 0$$

Hamiltonian :

$$H = \sqrt{1 + \frac{v^2}{u^2}} + \underline{\lambda}^T \underline{f}$$

$$= \sqrt{1 + \frac{v^2}{u^2}} + \lambda_1^{(t)} u + \lambda_2^{(t)} v$$

Also $\Psi = 0$

$$\begin{cases} \dot{x} = u \\ \dot{y} = v \end{cases} \quad \text{--- (1)}$$

Apply co-state eq^{ns} : (condition ②)

$$\dot{\lambda}_1 = -\frac{\partial H}{\partial x} = 0 \Rightarrow \lambda_1(t) = c_1$$

$$\dot{\lambda}_2 = -\frac{\partial H}{\partial y} = 0 \Rightarrow \lambda_2(t) = c_2$$

Apply PMP (condition ③)

$$0 = \frac{\partial H}{\partial u} = \frac{v^2/u^2}{\sqrt{u^2+v^2}} + \cancel{\lambda_1}^T c_1$$

$$0 = \frac{\partial H}{\partial v} = \frac{v/u}{\sqrt{u^2+v^2}} + \cancel{\lambda_2}^T c_2$$

2 eq^{ns}
in 2
variables

$\therefore u = K_1$ (some constant that is nonlinear
fns. of other constants C_1, C_2)

$v = K_2$ (another constant depends on C_1 & C_2)

But state eq \approx :

$$\begin{aligned} \dot{x} = u &= K_1 \Rightarrow x^* = K_1 t + \tilde{K}_1 \\ \dot{y} = v &= K_2 \Rightarrow y^* = K_2 t + \tilde{K}_2 \end{aligned} \quad \left. \begin{array}{l} \tilde{K}_1 \\ \tilde{K}_2 \end{array} \right\}$$

eliminate t

\therefore if t is optimal.

$$y^* = l_1 x^* + l_2$$

$$l_1 = K_2 / K_1,$$

$$l_2 = \tilde{K}_2 - \frac{K_2}{K_1} \tilde{K}_1$$

Applies $(x_1, y_1) \equiv A$
 $(x_2, y_2) \equiv B$ } to determine the constants

$$H^* = H(x^*, u^*, \lambda^*, t)$$

$$= \sqrt{1 + \frac{k_2^2}{k_1^2}} + c_1 k_1 + c_2 k_2$$

$$= \text{constant} \quad (\text{verified})$$



Example 2 : (Temperature(θ) control in a room)

Newton's Law of heating/cooling:

$$\dot{\theta} = -\alpha(\theta - \theta_a) + \beta u, \quad \begin{matrix} \alpha, \beta = \text{constants} \\ \theta_a = \text{ambient temperature} \end{matrix}$$

$$\text{Let } x(t) := \theta(t) - \theta_a$$

$$\Rightarrow \boxed{\dot{x} = -\alpha x + \beta u}, \quad x(0) = x_0 \text{ known. } u = \text{control}$$

minimize $u(\cdot)$ $\int_0^T \underbrace{\frac{1}{2} (u(t))^2}_{\text{Energy}} dt$, $T = \text{final time}$
fixed

$x(0) = x_0$ (known)

$$H = L + \lambda f$$

$$= \frac{1}{2} u^2 + \lambda(t) (-\alpha x + \beta u)$$

$$\dot{x} = \frac{\partial H}{\partial \lambda} = -\alpha x + \beta u \quad \leftarrow \text{state eq}^{**}$$

$$\dot{\lambda} = -\frac{\partial H}{\partial x} = +\alpha \lambda(t) \quad \leftarrow \text{costate eq}^{**}$$

$$0 = \frac{\partial H}{\partial u} = u + \lambda(t) \beta$$

$\Rightarrow \boxed{u^*(t) = -\beta \lambda(t)}$

For now, let's pretend that $\underline{\lambda(T)}$ is known.

Then $\dot{x} = +\alpha x$

$$\Rightarrow \boxed{\lambda(t) = e^{-\alpha(T-t)} \lambda(T)}$$

(integrate back in time)

$$\text{Then } \dot{x} = -\alpha x(t) + \beta \cdot u(t)$$

$$= -\alpha x(t) + \beta \cdot (-\beta \lambda(t))$$

$$= -\alpha x - \beta^2 \lambda(t)$$

$$= -\alpha x - \beta^2 e^{-\alpha(T-t)} \lambda(T)$$

$$\Rightarrow x(t) = x_0 e^{-\alpha t} - \frac{\beta^2}{\alpha} \lambda(T) e^{-\alpha T} \sinh(\alpha t)$$

Case I Fixed $x(T)$.
 Give T is also fixed.

$$\left. \begin{array}{l} dx(T) = 0 \\ dT = 0 \end{array} \right\} x(0) = 0$$

\therefore Condition (4) gives us nothing new.

Then the only strategy possible to
 Compute $x(T)$ is:

$$x(t) \Big|_{t=T} = x(0) e^{-\alpha T} - \frac{\beta^2}{2\alpha} x(T) \left\{ 1 - e^{-2\alpha T} \right\}$$

$x(T)$
 (given)

$$\Rightarrow x(T) = f_n(x(T), T)$$

\therefore solved.

Case II free terminal state

$x(\tau)$

but we need to put $\phi(x(\tau), \tau)$

$$J = \underbrace{\frac{1}{2} s \underbrace{(x(\tau) - 10)^2}_{\text{cost}}}_{\text{some positive constant.}} + \frac{1}{2} \int_0^\tau u^2 dt$$

Then we need condition (4).

$$d\tau = 0$$

$$dx(\tau) \neq 0$$

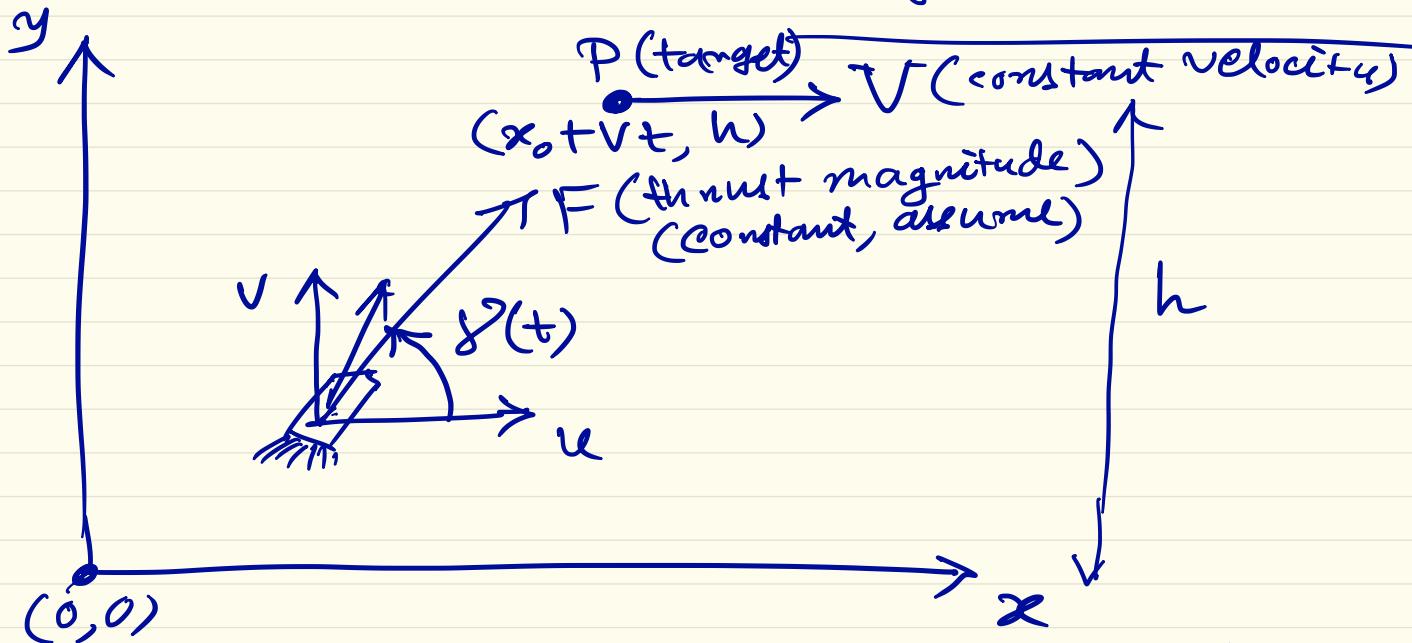
\therefore coeff. of $dx(\tau)$, must be $= 0$.

that gives $\lambda(\tau) = \underline{s(x(\tau) - 10)}$



Example 3:

Thrust angle Programming: Intercepting moving target by a missile



Intercept problem vs. Rendezvous Problem
(Missile problem) (Moon landing)

State vector
of the missible $(\underline{x}) = \begin{pmatrix} x \\ y \\ u \\ v \end{pmatrix} \in \mathbb{R}^4$

$$a := \frac{F}{m} \text{ (Known thrust accel.)}$$

Target P has initial posⁿ x_0
(it moves @ const. altitude w)

Control: thrust angle $\delta(t)$.

Controlled dynamics (state ODE)

state (\underline{x})	$\dot{x} = u$ $\dot{y} = v$ $\dot{u} = a \cos \delta$ $\dot{v} = a \sin \delta$	}	I.C.
	$x(0) = 0$ $y(0) = 0$ $u(0) = 0$ $v(0) = 0$		