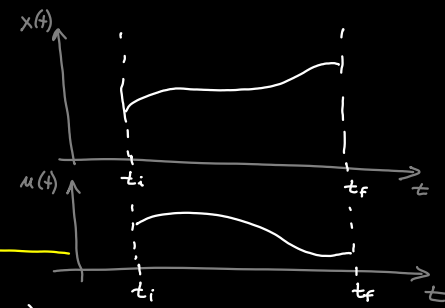


$$\min_{u, x \in C[t_i, t_f]} \left[\phi(x(t_f), t_f) + \int_{t_i}^{t_f} L(x(t), u(t), t) dt \right]$$

s.t. $\dot{x}(t) = f(x(t), u(t), t), \quad x(t_i) = r_i$

$h(x(t), u(t)) \leq 0 \quad \rightarrow \langle x(t_f), t_f \rangle = 0 \quad (x(t_f) = r_f)$



LQ - optimal control:

$$\min_{u, x} \left[\frac{1}{2} x^T(t_f) S_f x(t_f) + \frac{1}{2} \int_0^{t_f} (x^T(t) Q x(t) + u^T(t) R u(t)) dt \right]$$

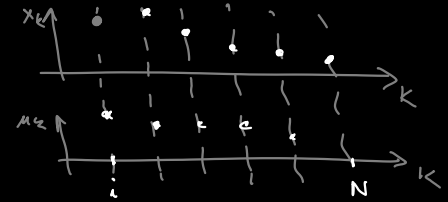
s.t. $\dot{x}(t) = A x(t) + B u(t), \quad x(0) = r_0$

compare with the discrete-time:

or infinite-horizon LQ:

$$\min \int_0^{\infty} (x^T(t) Q x(t) + u^T(t) R u(t)) dt$$

s.t. $\dot{x}(t) = A x(t) + B u(t), \quad x(0) = r_0$



OPTIMIZATION OVER FUNCTIONS $\rightarrow$ CALCULUS OF VARIATIONS

compare with

DIFFERENTIAL CALCULUS

SCALAR CASE



$$f(x^* + \Delta x) \geq f(x^*)$$

$$\Delta f \geq 0 \quad \text{differential}$$

$$\Delta f = df + O(\Delta x^2)$$

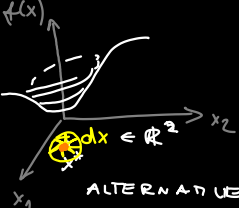
derivative... $f' = \frac{df}{dx}$

$$df = 0 \Rightarrow f' = 0$$

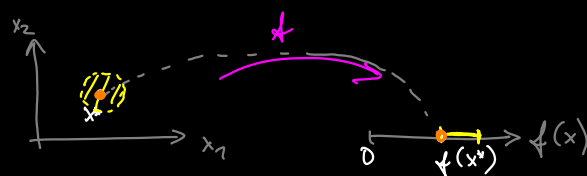
$$df = \left(\frac{\partial f}{\partial x} \right)^T dx$$

gradient inner product

VECTOR CASE $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$



ALTERNATIVE VIEWPOINT:



$$\frac{\delta J}{\delta g^k} = ?$$

arbitrary but fixed

$$\frac{dJ(g^* + \gamma \cdot \alpha)}{d\alpha}$$

CALCULUS OF VARIATIONS

$$\min_{y \in C[a, b]} J(y(x))$$

indep. var. (posit.)
function

functional
(= function of a function)

$$y(x, \alpha) = y^*(x) + \left. \frac{\partial y(x, \alpha)}{\partial \alpha} \right|_{\alpha=0} \cdot \alpha + O(\alpha^2)$$

$$\delta y \dots \text{variation of function}$$

$$\delta y = \frac{\partial y(x, \alpha)}{\partial \alpha} \Big|_{\alpha=0} \cdot \alpha = \gamma(x) \cdot \alpha$$

$\gamma(x) \in C^1$

$$J = J^* + \delta J$$

$$\delta J = \frac{J(y^* + \delta y) - J(y^*)}{\|y^* + \delta y - y^*\|} \approx \frac{J(y^* + \delta y) - J(y^*)}{\|y^*\|}$$

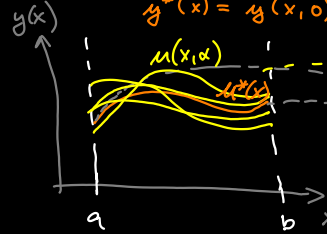
$$\delta J = \int_{t_i}^{t_f} \left(\frac{\delta J}{\delta y(t)} \right) \cdot \delta y(t) dt$$

inner product

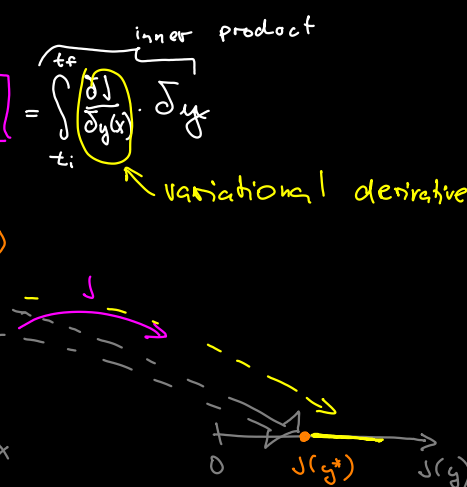
variation of functional

variational derivative

SCALAR γ

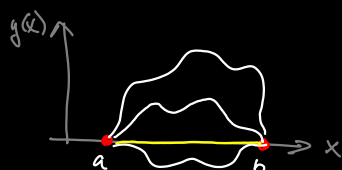


NEIGHBORHOOD?



Examples

1.) Minimum distance between points



$$J = \int_a^b ds = \int_a^b \sqrt{dx^2 + dy^2} = \int_a^b \sqrt{1 + \left(\frac{dy}{dx}\right)^2} dx = \int_a^b \sqrt{1 + y'^2} dx$$

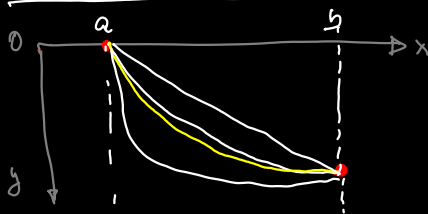
2.) Dido's problem — maximization of area (circumference fixed)



$$\max J, \quad J = \int_a^b y(x) dx$$

s.t. $\int_a^b \sqrt{1 + y'^2} dx = C$

3.) Brachistochrone problem (shortest time)



$$\min J$$

$$J = \int_{t_i}^{t_f} dt = \int_{t_i}^{t_f} \frac{ds}{v} = \int_a^b \frac{\sqrt{1 + y'^2}}{\sqrt{2g y(x)}} dx$$

$$\mathcal{T} + \mathcal{V} = \frac{1}{2} m v^2 - m g y$$

$$(\mathcal{T} + \mathcal{V})_{t_i} = 0 \Rightarrow (\mathcal{T} + \mathcal{V})_t = 0 \Rightarrow \frac{1}{2} m v^2 = m g y$$

$$v = \sqrt{2 g y}$$

$$J(y(x)) = \int_a^b L(x, y(x), y'(x)) dx$$

Basic problem of calculus of variations with fixed ends

$$\min_{y(x)} \int_a^b L(x, y(x), y'(x)) dx, \quad y(a) = y_a, \quad y(b) = y_b$$



$$y^*(x) \rightarrow y^*(x) + \underbrace{\frac{\delta J}{\delta y}}_{\alpha \cdot \eta(x)}$$

$$\delta J = \left. \frac{dJ}{d\alpha} \right|_{\alpha=0}$$

$$\frac{dJ(y^*(x) + \alpha \cdot \eta(x))}{d\alpha} = \frac{d}{d\alpha} \int_a^b L(x, y^*(x) + \alpha \cdot \eta(x), y'^*(x) + \alpha \cdot \eta'(x)) dx = \int_a^b \frac{d}{d\alpha} L(x, y^* + \alpha \cdot \eta, y'^* + \alpha \cdot \eta') dx = \int_a^b \left[\frac{\partial L}{\partial y} \eta + \frac{\partial L}{\partial y'} \eta' \right] dx$$

$\frac{\partial L}{\partial y} \dots$ variational derivative

$$\left[\frac{\partial L}{\partial y'} \eta \right]_a^b = \left[\frac{\partial L}{\partial y'} \eta \right]_a^b + \int_a^b \left[\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial y'} \right] \eta(x) dx$$

$\eta(a) = \eta(b) = 0$

Fact: $\int_a^b f(x) \eta(x) dx = 0$ for arbitrary $\eta(x) \in \mathcal{C}^1 \Rightarrow f = 0$

$$\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} = 0$$

... EULER-LAGRANGE EQUATION
... second-order ODE

$$\text{or} \quad \frac{\partial L}{\partial y} = \frac{d}{dx} \frac{\partial L}{\partial \dot{y}}$$

$$L_r - (L_{ix} + L_{iy} \cdot \dot{y} + L_{ij} \cdot \ddot{y}) = 0$$

+ boundary conditions

$$\delta J = \left. \frac{dJ}{dx} \right|_{x=0} \cdot \alpha = \int_a^b \left[\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} \right] \underbrace{\eta \alpha}_{\delta y} dx = \int_a^b \left[\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} \right] \delta y dx$$

But also

$$\delta J = \delta \int_a^b L(x, y(x), \dot{y}(x)) dx$$

$$= \int_a^b \delta L(x, y(x), \dot{y}(x)) dx = \int_a^b \left[\frac{\partial L}{\partial x} \delta x + \frac{\partial L}{\partial y} \delta y + \frac{\partial L}{\partial \dot{y}} \delta \dot{y} \right] dx$$

$$= \int_a^b \left[\frac{\partial L}{\partial y} \delta y + \frac{\partial L}{\partial \dot{y}} \delta \dot{y} \right] dx = \int_a^b \left[\frac{\partial L}{\partial y} \delta y + \frac{\partial L}{\partial \dot{y}} \left(\delta \dot{y} \right) \right] dx$$

PER PARTES

$$= \left[\underbrace{\frac{\partial L}{\partial \dot{y}} \delta y}_{\delta \left(\frac{dy}{dx} \right) - \frac{d}{dx} (\delta y)} \right]_a^b + \int_a^b \left[\frac{\partial L}{\partial y} \delta y - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} \delta y \right] dx$$

$$\delta J = \int_a^b \left[\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} \right] \delta y dx$$

SPECIAL CASES

1) NO x: $0 = \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} = \left(\frac{\partial L}{\partial \dot{y}} \right)^p = c_1 \text{ constant}$

$$p := \frac{\partial L}{\partial \dot{y}}$$

2) NO y:
$$L_r - (L_{ix} + L_{iy} \cdot \dot{y} + L_{ij} \cdot \ddot{y}) = 0$$
$$L_y \cdot \dot{y} - L_{iy} \cdot \dot{y}^2 - L_{ij} \cdot \ddot{y} \cdot \dot{y} = 0$$
$$\frac{d}{dx} (L_y \cdot \dot{y} - L) = 0$$

$$\Downarrow$$
$$L_y \cdot \dot{y} - L = c_2 \text{ ... constant}$$

$$H := L_y \cdot \dot{y} - L = p \cdot \dot{y} - L$$

$$H(x, y, \dot{y}, p)$$

/ $\dot{y}$

$$\left\{ \begin{aligned} & (L_{ix} \cdot \dot{y} + L_{iy} \cdot \ddot{y}) \cdot \dot{y} + L_y \cdot \dot{y} \\ & - L_y \cdot \dot{y} - L_{iy} \cdot \dot{y}^2 \\ & = L_{iy} \cdot \dot{y}^2 + L_{iy} \cdot \ddot{y} \cdot \dot{y} + L_{ij} \cdot \ddot{y} \cdot \dot{y} - L_y \cdot \dot{y} - L_{iy} \cdot \dot{y}^2 \end{aligned} \right.$$

Ex: NO y: switch from $x \rightarrow t$:
 $L = \frac{1}{2} m \dot{y}^2 + 0$

$$L_y = m \dot{y} =: p$$

NO x: $L = \frac{1}{2} m \dot{y}^2 - m g \cdot y$
 $H = \frac{1}{2} m \dot{y}^2 - \frac{1}{2} m \dot{y}^2 + m g y \text{ ... total en.}$

Also:

$$\dot{y} = H_p$$

$$\dot{p} = \frac{d}{dx} L_{\dot{y}} = L_y = -H_y$$

↑
E.L.

}

$$\begin{bmatrix} \dot{y} \\ \dot{p} \end{bmatrix} = \begin{bmatrix} H_p \\ -H_y \end{bmatrix}$$

Hamilton's canon. eq.

And yet another property:

$$\frac{\partial H_y}{\partial \dot{y}} = p - \frac{\partial L}{\partial \dot{y}} = 0 \Rightarrow H \text{ has a stationary point wrt. the third variable}$$

$$H(x, y, \underline{z}, p) \longrightarrow$$

$$\left. \frac{\partial H(x, y^*, z, p^*)}{\partial z} \right|_{z=\underline{y}} = 0$$

SECOND-ORDER CONDITIONS

- necessary

$$L_{\dot{y}\dot{y}} \geq 0$$

- sufficient

$$L_{\dot{y}\dot{y}} > 0$$

... not enough (should be on a short interval)

⇓

$$H_{\dot{y}\dot{y}} = -L_{\dot{y}\dot{y}}$$

⇓

$$H_{\dot{y}\dot{y}} < 0$$

$$H = \underbrace{L_{\dot{y}}}_{p} \dot{y} - L$$

⇒ H maximized at extremal $H_{\dot{y}} = 0$

$$H(x, y^*, \underline{y}, p^*) \geq H(x, \dot{y}, z, p^*)$$

EQUALITY CONSTRAINTS

$$F(x, y, \dot{y}) = 0 \quad \dots \text{implicit ODE}$$

$$J^{Aug} = \int_a^b L(x, y, \dot{y}) dx + \int_a^b \lambda(t) \cdot F(x, y, \dot{y}) dx = \int_a^b \underbrace{[L + \lambda F]}_{L^{Aug}(x, y, \dot{y}, \lambda)} dx$$

Explicit ODE: $\dot{y} = f(x, y, u(x))$

$$J = \int_a^b L(x, y, \dot{y}) dx = \int_a^b \tilde{L}(x, y, u) dx$$

Ex: $\dot{y} = u$

$$J = \int_a^b L(x, y, \underline{u}) dx$$

OPTIMAL CONTROL - FIXED FINAL STATE, FINITE HOR.

$$\min \int_{t_i}^{t_f} L(t, x(t), u(t)) dt$$

! $x \rightarrow t$
 $y(x) \rightarrow x(t)$

s.t. $\dot{x}(t) = f(x(t), u(t), t), \quad x(t_i) = x_i, \quad x(t_f) = x_f$

either

$$J^{AVG}(t, x, u, \lambda) = \int_{t_i}^{t_f} \underbrace{\left[L(x, u, t) + \lambda^T(t) \cdot (\dot{x}(t) - f(x, u, t)) \right]}_{L^{AVG}(t, x(t), \dot{x}(t), u(t), \lambda(t))} dt$$

E.L. wrt $\begin{bmatrix} x \\ u \\ \lambda \end{bmatrix}$

$$\nabla_x L^{AVG} = \frac{d}{dt} \nabla_{\dot{x}} L^{AVG}$$

$$\nabla_u L^{AVG} = \frac{d}{dt} \nabla_{\dot{u}} L^{AVG} = 0$$

$$\nabla_{\lambda} L^{AVG} = 0$$

vs

$$L(t, x, \dot{x})$$

$$\begin{matrix} \uparrow \\ \begin{bmatrix} x \\ u \\ \lambda \end{bmatrix} \end{matrix} \quad \begin{bmatrix} \dot{x} \\ \dot{u} \\ \dot{\lambda} \end{bmatrix}$$

$$\begin{aligned} \nabla_x L - \nabla_{\dot{x}} f \cdot \lambda &= \dot{x} \\ \nabla_u L - \nabla_{\dot{u}} f \cdot \lambda &= 0 \\ \dot{x} - f(x, u, t) &= 0 \end{aligned}$$

Ex.: LQ-optimal, fixed final state: $L = \frac{1}{2} [x^T Q x + u^T R u]$

$$Qx - A^T \lambda = \dot{x}$$

$$u = \bar{R}^{-1} B^T \lambda$$

$$\begin{aligned} R u - B^T \lambda &= 0 \\ A x + B u &= \dot{x} \end{aligned}$$

$$\begin{aligned} \dot{x} &= A x + B \bar{R}^{-1} B^T \lambda \\ \dot{\lambda} &= Q x - A^T \lambda \end{aligned}$$

Two-point BVP
2n eq's
2n BC's

NEEDED: $x(t_i) = \bar{x}_i$
 $x(t_f) = \bar{x}_f$

MATLAB: `BVP4C`

Using Hamiltonian

$$H(t, x, u, \lambda) = \lambda^T \dot{x} - L$$

$$\text{LQ: } H = \lambda^T \dot{x} - \frac{1}{2} [x^T Q x + u^T R u]$$

Ham. can. eq

$$\begin{aligned} \dot{x} &= \nabla_{\lambda} H \\ \dot{\lambda} &= -\nabla_x H \\ 0 &= \nabla_u H \end{aligned}$$

gram. based $u(t)$
 $\uparrow$
 $\lambda(t)$
 $\uparrow$
 $\lambda(t_f)$
 $\uparrow$

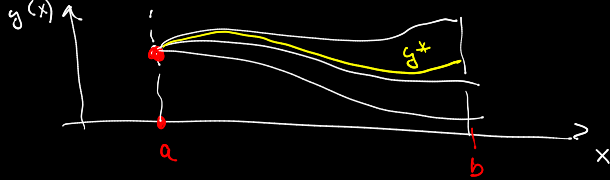
Solving LQ - fixed final state

$$\text{assume } Q=0 \Rightarrow J = \int_0^{t_f} u^T R u dt$$

$$\begin{aligned} \dot{x} &= A x + B \bar{R}^{-1} B^T \lambda \rightarrow x(t) = \dots x(t_f) \\ \dot{\lambda} &= -A^T \lambda \rightarrow \lambda(t) = \dots \lambda(t_f) \end{aligned}$$

FREE FINAL STATE ...

back to CALCULUS OF VARIATIONS



$$\left[\frac{\partial L}{\partial \dot{y}} \delta y \right]_a^b + \int_a^b \left[\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} \right] \delta y \, dx$$

$$\frac{\partial L}{\partial \dot{y}} \delta y \Big|_b + \int_a^b \left[\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} \right] \delta y \, dx = 0$$

$$\frac{\partial L}{\partial y} - \frac{d}{dx} \frac{\partial L}{\partial \dot{y}} = 0$$

$$\frac{\partial L}{\partial \dot{y}} \Big|_b = 0$$

↓ in correct :

$$J = \underbrace{\phi(x_{t_f})} + \int_{t_i}^{t_f} L \, dt$$

$$\int_{t_i}^{t_f} \frac{d}{dt} \phi(x, t) \, dt + \phi(x(t_i))$$

$$J = \int_{t_i}^{t_f} \left[L + \frac{\partial \phi}{\partial x} \dot{x} + \lambda^T (\dot{x} - f(x, u, t)) \right] dt$$

TIME INVARIANT

$$\frac{\partial L^{aug}}{\partial \dot{x}} \Big|_{t_f} = \lambda(t_f) = 0$$

$$\lambda(t_f) + \frac{\partial \phi}{\partial x}(t_f) = 0$$

$$LQ: \phi = \frac{1}{2} x(t_f)^T S x(t_f)$$

$$\lambda(t_f) + S \cdot x(t_f)$$

LQ :

$$\dot{x} = Ax + Bu$$

$$\dot{\lambda} = Qx - A^T \lambda$$

$$u = -\hat{B}^T \lambda$$

BC: $x(t_i) = r_i$

$$S x(t_f) = -\lambda(t_f)$$

$$S x(t) = -\lambda(t)$$

$$\dot{S}x + S\dot{x} = -\dot{\lambda}$$

$$\dot{S}x + S(Ax + B\dot{u}) = -(Qx - \overset{\uparrow}{A^T}\overset{\uparrow}{\lambda})$$

$$\quad \quad \quad \uparrow \quad \quad \quad \uparrow$$

$$\quad \quad \quad -\tilde{R}^{-1}B^T Sx \quad \quad \quad -Sx$$

$$-\dot{S}x = SAx - S B \tilde{R}^{-1} B^T Sx + Qx + A^T Sx$$

$$-\dot{S} = SA + A^T S + Q - S B \tilde{R}^{-1} B^T S$$

diff. RE

$$\dot{S} = 0 \Rightarrow ARE$$