

Ex Brachistochrone

$$f = \sqrt{\frac{1+(y')^2}{y}} \rightarrow f(y, z) = \sqrt{\frac{1+z^2}{y}} \quad \int_2 = \frac{z}{y \sqrt{1+z^2}}$$

$$\left( \begin{array}{l} \frac{y' \cdot z}{y \sqrt{1+z^2}} \Big|_{z=y'} = -\frac{1}{2} y^{-3/2} \sqrt{1+z^2} \Big|_{z=y'} = \int_1 \\ \frac{(y')^2}{y \sqrt{1+(y')^2}} - \frac{1}{2y^{3/2}} \sqrt{1+(y')^2} = c_1 \end{array} \right)$$

redo correctly

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$$y' f_z(y, y') - f(y, y') = c$$

$$\rightarrow \frac{y'}{\sqrt{y}} \frac{y'}{\sqrt{1+(y')^2}} - \frac{\sqrt{1+(y')^2}}{\sqrt{y}} = c$$

$$\rightarrow \frac{(y')^2 - (1+(y')^2)}{\sqrt{y} \sqrt{1+(y')^2}} = -c$$

$$\rightarrow \frac{-1}{\sqrt{y} \sqrt{1+(y')^2}} = -c$$

$$\rightarrow \sqrt{y} \sqrt{1+(y')^2} = \frac{1}{c} = c_1$$

$$\rightarrow y' = \frac{\sqrt{c_1^2 - y}}{\sqrt{y}}$$

(equivalent to what we had before)

Free End-Point Problem

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

$$D_1 = \{y \in C^1[a, b]; y(a) = y_a\}$$

$$D_2 = \{y \in C^1[a, b]; y(b) = y_b\}$$

$$D_3 = C^1[a, b]$$

Minimize  $J(y)$  on  $D_1$ ,  $D_2$ , or  $D_3$ .Minimize  $J(y)$  on  $D_1$ :Suppose  $y_0 \in D_1$  gives the minLet  $v \in C^1[a, b]$  such that  $v(a) = 0$ .

Then  $y_0 + \alpha v \in D_1 \quad \forall \alpha \in \mathbb{R}$

Let  $g(\alpha) = J(y_0 + \alpha v) \quad g: \mathbb{R} \rightarrow \mathbb{R}$

$\alpha=0$  gives the min value of  $g \Rightarrow g'(0) = 0$ .

$$0 = g'(0) = \delta J(y_0; v) = \int_a^b (f_y[y_0]v + f_z[y_0]v') dx$$

$$= \int_a^b (f_y[y_0] - (f_z[y_0])') v dx + f_z[y_0]v \Big|_a^b$$

$$= \int_a^b (f_y[y_0] - (f_z[y_0])') v dx + f_z[y_0(b)]v(b)$$

$\forall v \in C^1[a, b]$  st  $v(a) = 0$ .

(1) Consider  $v \in C^1[a, b]$  st  $v(a) = v(b) = 0$

$$\int_a^b (f_y[y_0] - (f_z[y_0])') v dx = 0$$

Fundamental lemma  $\Rightarrow f_y[y_0] - (f_z[y_0])' = 0$ , E-L eqn.

(2)  $f_z[y_0(b)]v(b) = 0 \quad \forall v \in C^1[a, b]$  st  $v(a) = 0$

$$\text{Let } v = x - a \Rightarrow f_z[y_0(b)](b-a) = 0 \Rightarrow \boxed{f_z[y_0(b)] = 0}$$

Natural Boundary Condition

Theorem: (1) If  $y_0$  minimizes  $J$  on  $D_1$ , then  $y_0$  satisfies the E-L equation and  $f_z[y_0(b)] = 0$ .

(2) If  $y_0$  minimizes  $J$  on  $D_2$  then  $y_0$  satisfies the E-L equation and  $f_z[y_0(a)] = 0$ .

(3) If  $y_0$  minimizes  $J$  on  $D_3$  then  $y_0$  satisfies the E-L equation and  $f_z[y_0(a)] = 0, f_z[y_0(b)] = 0$ .

Theorem: Suppose  $f$  is [strongly] convex with respect to  $(y, z)$  &  $(f_z[y_0])' = f_y[y_0]$

(1) If  $y_0(a) = y_a, f_z[y_0(b)] = 0$  then  $y_0$  minimizes  $J$  [uniquely] on  $D_1$ .

(2) If  $y_0(b) = y_b, f_z[y_0(a)] = 0$  then  $y_0$  minimizes  $J$  [uniquely] on  $D_2$ .

(3) If  $f_z[y_0(a)] = 0, f_z[y_0(b)] = 0$  then  $y_0$  minimizes  $J$  [uniquely up to an additive constant or uniquely] on  $D_3$ .

Proof:  $J(y_0+v) \geq J(y_0) + \int J(y_0; v) = J(y_0)$  — (1)

$\forall v \in C^1[a,b]$  So in (1),  $v(a)=0$ ; in (2)  $v(b)=0$

$$f(x, y+v, y'+v') \geq f(x, y, y') + f_y(x, y, y')v + f_z(x, y, y')v' \quad (2)$$

If  $f$  is strongly convex wrt  $(y, z)$  equality holds in (2)

$$\Rightarrow v=0 \text{ or } v'=0 \Rightarrow vv'=0 \Rightarrow (v/2)'=0$$

In case (1),  $v(a)=0 \Rightarrow v=0$

In case (2),  $v(b)=0 \Rightarrow v=0$

In case (3)  $v = \text{const.} = c$

$$J(y_0+c) = J(y_0)$$

$c$  may be forced to be 0

Equality in (1) is equivalent to equality in (2)  $\square$

Example  $J(y) = \int_1^2 \frac{(y')^2}{x} dx$

Minimize  $J$  on  $D_2 = \{y \in C^1[1,2]; y(2)=3\}$

$$f(x, z) = \frac{z^2}{x}, \quad f_{zz} = \frac{2}{x} > 0 \quad f \text{ is strictly convex wrt } z$$

$\Rightarrow f$  is strongly convex wrt  $(y, z)$

E-L eqn:  $(f_z)' = f_y$

$$\left(\frac{2y'}{x}\right)' = 0 \rightarrow \frac{2y'}{x} = c_1 \Rightarrow y = \frac{c_1}{4}x^2 + c_2$$

nat. bd. cond:  $f_z[y(1)] = 0$

$$\rightarrow \frac{2y'(1)}{1} = 0 \rightarrow y'(1) = 0 \rightarrow c_1 = 0 \rightarrow y = c_2$$

$y=3$   
is the unique minimizer on  $D_2$

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Minimize the above on  $C^1[1,2]$

Again,  $y = \frac{c_1}{4}x^2 + c_2$

nat. bd. cond:  $f_z[y(1)] = 0, \quad f_y[y(1)] = 0$

$$\rightarrow y'(1) = 0$$

$$\rightarrow y'(2) = 0$$

$$y' = \frac{c_1}{2}x \rightarrow y' = 0 \rightarrow y \text{ constant}$$

$$J(c) = 0 \quad \forall c$$

Minimizer is unique up to an additive constant

Example  $J(y) = \int_1^2 \frac{y^2}{x} dx$

Minimize  $J$  on  $C^1[1,2]$

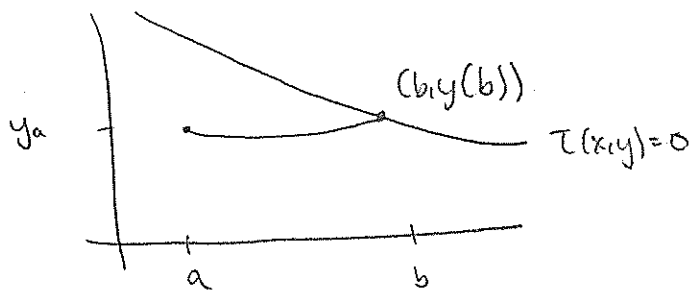
$f = f(x,y) = y^2/x$   $f$  strongly convex w.r.t  $(y,z)$   
 E-L eq'n  $(f_z[y])' = f_y \rightarrow 2y/x = 0 \rightarrow y=0$   
 $y=0$  is the unique minimizer.

Variable End-Point Problems

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

subject to  $y(a) = y_a$  &  $\tau(b, y(b)) = 0$ ,  $b > a$

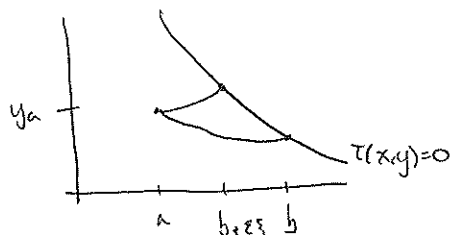
$\tau$  is a  $C^1$  function such that  $\nabla \tau = (\tau_x, \tau_y) \neq (0,0)$



Suppose  $(y_0, b_0)$  is a minimizer

Consider  $h(\epsilon) = J(y_0 + \epsilon v, b_0 + \epsilon \xi)$

$v = v(x)$ ,  $\xi$  is a const. where  $\tau(b_0 + \epsilon \xi, y_0(b_0 + \epsilon \xi) + \epsilon v(b_0 + \epsilon \xi)) = 0$  - (1)



$h$  has a min at  $\epsilon=0 \rightarrow h'(0) = 0$

$h(\epsilon) = \int_a^{b_0 + \epsilon \xi} f(x, y_0 + \epsilon v, y_0' + \epsilon v') dx$

$0 = h'(0) = \int_a^{b_0} \{ f_y[y_0] v + f_z[y_0] v' \} dx$   
Leibniz' rule

For  $\xi = 0$ :  $\int_a^{b_0} \{f_y[y_0]v' + f[y_0]v'\} dx = 0$

$v(b) = 0, v(a) = 0 \Rightarrow \int_a^{b_0} \{f_y[y_0] - (f_z[y_0])'\} v dx = 0$

$\Rightarrow f_y[y_0] = (f_z[y_0])' = 0$  E-L eqn

For  $\xi \neq 0$ :  $v(a) = 0$   
 $0 = \int_a^{b_0} \underbrace{\{f_y[y_0] - (f_z[y_0])'\}}_{=0} v dx + f_z[y_0(b_0)] v(b) + f[y_0(b_0)] \xi$

$f[y_0(b_0)] + f_z[y_0(b_0)] v(b) = 0$  — (2)

$v(b_0)$  is determined by differentiating (1) wrt  $t$  & setting  $\xi = 0$ :

$\tau_x(b_0, y_0(b_0)) \xi + \tau_y(b_0, y_0(b_0)) (y_0'(b_0) \xi + v(b_0)) = 0$

$\Rightarrow v(b_0) = \xi \left( -y_0'(b_0) - \frac{\tau_x(b_0, y_0(b_0))}{\tau_y(b_0, y_0(b_0))} \right)$

Put this into (2):

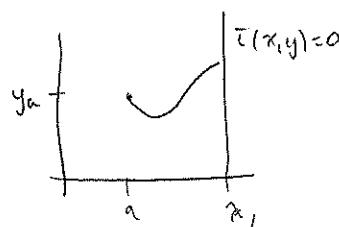
$\tau_y(b_0, y_0(b_0)) f[y_0(b_0)] = f_z[y_0(b_0)] \left( \tau_y(b_0, y_0(b_0)) y_0'(b_0) + \tau_x(b_0, y_0(b_0)) \right)$

or  $\tau_y f = f_z (\tau_y y' + \tau_x)$

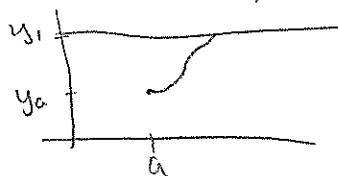
[Transversality Condition]

Special cases:

(a)  $\tau(x, y) = x - x_1$   
 $f_z[y_0(x_1)] = 0$

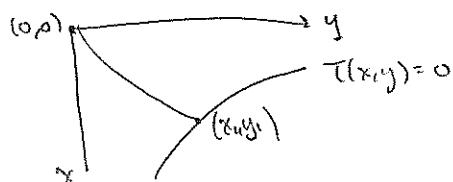


(b)  $\tau(x, y) = y - y_1$   
 $\tau_y = 1, \tau_x = 0$   
 $y_0'(b_0) f_z[y_0(b_0)] - f[y_0(b_0)] = 0$   
 $y' f_z - f = 0$  Hamiltonian



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Seek minimum time curve from  $(0,0)$  to  $(x_1, y_1)$ ,  $\tau(x_1, y_1) = 0$



$$f = \frac{\sqrt{1+z^2}}{\sqrt{\lambda}} \quad f_z = \frac{z}{\sqrt{\lambda} \sqrt{1+z^2}}, \quad z = y'$$

$$T_y(x_i, y_i) \frac{\sqrt{1+y'(x_i)^2}}{\sqrt{\lambda}} = \frac{y'(x_i)}{\sqrt{\lambda} \sqrt{1+y'(x_i)^2}}$$

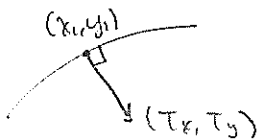
$$T_y(1+(y')^2) = T_y(y')^2 + y' T_x \rightarrow T_y = y' T_x \text{ at } (x_i, y_i) = (x_i, y_i)$$

$$(T_x, T_y) \cdot (y', -1) = 0$$

$(T_x, T_y)$  is orthogonal to  $T(x, y) = 0$

$(x, y(x))$  has tangent  $(1, y')$

$(y', -1)$  is orthogonal to  $(x, y(x)) \Rightarrow$  the curve  $y = y(x)$  is orthogonal to  $T(x, y) = 0$  at  $(x_i, y_i)$



### Dido's Problem

$$y = y(x), \quad -a \leq x \leq a, \quad a > 0$$

$$y(x) \geq 0, \quad y \in C^1(-a, a) \cap C^0[-a, a]$$

$$\text{Maximize } \int_{-a}^a y(x) dx \quad \text{subject to } \int_{-a}^a \sqrt{1+y'(x)^2} dx = L$$

$$T_y f = f_z (T_y y' + T_x) \quad \text{at } x=b \leftarrow ?$$

$$\text{Minimize } J(y, a) = - \int_{-a}^a y dx$$

$$\text{subject to } G(y, a) = \int_{-a}^a \sqrt{1+(y')^2} dx = L$$

$$f = -y, \quad g = \sqrt{1+z^2}$$

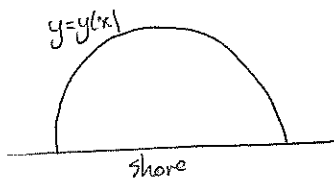
$$f = f_x + \lambda g = -y + \lambda \sqrt{1+z^2}$$

$$\text{E-L eqn } E_v f: (f_z [y])' = \tilde{f}_y [y]$$

$$\left( \frac{\lambda y'}{\sqrt{1+(y')^2}} \right)' = -1$$

$$\rightarrow \text{aside} \rightarrow \begin{cases} \frac{y''}{(1+(y')^2)^{3/2}} = -\frac{1}{\lambda} \\ \text{curvature} = -1/\lambda \end{cases}$$

$$\frac{\lambda y'}{\sqrt{1+(y')^2}} = -x + C$$



For  $\lambda > 0$ ,  $\tilde{f}$  is strictly convex wrt  $(y, z)$

Expect unique minimizer. If  $y(x)$  is a minimizer then  $y(-x)$  is also a minimizer  $\Rightarrow y(x) = y(-x) \Rightarrow y'(0) = 0 \Rightarrow c = 0$

$$\frac{\lambda y'}{\sqrt{1+(y')^2}} = -x$$

For  $y(x) > 0$ ,  $x > 0$ ,  $y(a) = 0 \rightarrow y'(x) < 0 \rightarrow \lambda > 0$

$$\frac{(y')^2}{1+(y')^2} = \left(\frac{x}{\lambda}\right)^2 \rightarrow y' = \pm \frac{x}{\sqrt{\lambda^2 - x^2}}$$

$y' < 0$  for  $x > 0$  implies  $y' = -x/\sqrt{\lambda^2 - x^2}$

$$\rightarrow y = \sqrt{\lambda^2 - x^2} + d$$

Transversality condition

$T(x, y) = y$  endpoints are on  $T(x, y) = 0$ ,  $T_y = 1$ ,  $T_x = 0$

$\tilde{f} = \tilde{f}_2 y'$  or  $y' \tilde{f}_2 = \tilde{f} = 0$  at  $x = \pm a$

$$\frac{y' \lambda y'}{\sqrt{1+(y')^2}} - (-y + \sqrt{1+(y')^2}) = 0$$

$$\frac{\lambda (y')^2}{\sqrt{1+(y')^2}} - \sqrt{1+(y')^2} + y = 0$$

$$\left(\frac{-\lambda}{\sqrt{1+(y')^2}} + y\right) \Big|_{x=\pm a} = 0 \rightarrow \lim_{x \rightarrow a^-} |y'| = \infty$$

$$y' = \frac{-x}{\sqrt{\lambda^2 - x^2}} \quad |y'| \rightarrow \infty \text{ as } x \rightarrow \pm a \Rightarrow \lambda = a$$

$$y = \sqrt{a^2 - x^2} + d$$

$$y(\pm a) = 0 \Rightarrow d = 0. \text{ Then } y = \sqrt{a^2 - x^2}$$

Constraint:  $\pi a = L \Rightarrow a = L/\pi$

$$\therefore y = \sqrt{\left(\frac{L}{\pi}\right)^2 - x^2}$$