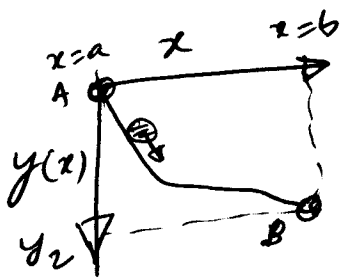


Calculus of Variations Example.



min time for ball to go from A to B
 given $y(x_1) = 0$, $y(x_2) = y_2$

$$dt = \frac{ds}{v} \quad ; \quad ds = \text{differential distance}$$

$$ds = \sqrt{dx^2 + dy^2}$$

$$\therefore T = \int_{x=a}^{x=b} \frac{ds}{v} = \int_a^b \frac{\sqrt{1+y'^2}}{v} dx$$

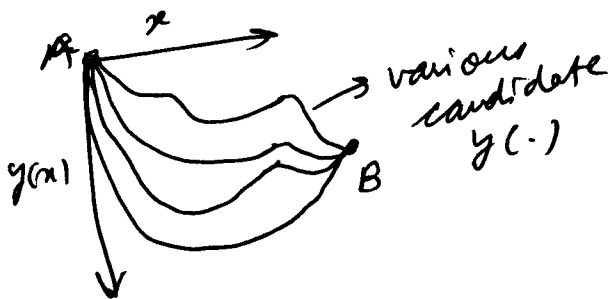
$$\frac{1}{2} m v_2^2 - \frac{1}{2} m v_1^2 = mg(y_1 - y_2)$$

Taking $v_1 = 0$, $y_1 = 0$ then

$$v = \sqrt{2gy}$$

$$\therefore T = \frac{1}{\sqrt{2g}} \int_a^b \frac{\sqrt{1+y'^2}}{\sqrt{y}} dx$$

find $y(x)$, $x \in [a, b]$ to minimize T .



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

$$y(a) = y_1 \quad ; \quad y(b) = y_2$$

Calculus of Variations (Simplified Version)

min $J(y(\cdot)) = \int_a^b f(x, y(x), y'(x)) dx$; $y(a) = y_1$; $y(b) = y_2$

Assume $y^*(\cdot)$ is a local minimizer

$$\hat{y}(x) = y^*(x) + \epsilon \eta(x)$$

$$F(\epsilon) = \int_a^b f(x, \hat{y}(x), \hat{y}'(x)) dx$$

$$F'(0) = \int_a^b (f_y \cdot \eta(x) + f_x \cdot \eta'(x)) dx$$

$$= \int_a^b (f_y \cdot \eta(x) + \frac{d}{dx} f_x \cdot \eta(x)) dx + f_x \cdot \eta(x) \Big|_a^b$$

(∵ $\eta(a) = 0$
 $\eta(b) = 0$)

∴ Euler-Lagrange necessary condition is

$$\frac{d}{dx} f_x(x, y^*(x), y'^*(x)) = f_y(x, y^*(x), y'^*(x)) \quad \forall x \in [a, b]$$

If the problem has one fixed point $y(a) = y_1$ and other end free then at that end we will have to satisfy

$$f_x \Big|_b = 0$$

Two Independent Variable Problem

$$J = \iint_D f(x, y, w, w_x, w_y) dx dy$$

points given on the boundary ∂D .

$$w(x, y) = w(x, y) + \epsilon \eta(x, y)$$

$$\therefore \eta(x, y) = 0 \quad \forall (x, y) \in \partial D$$

$$F(\epsilon) = \iint_D f(x, y, w, w_x, w_y) dx dy$$

solve $F'(\epsilon) = 0$ at $\epsilon = 0$

$$F'(0) = \iint_D (f_{\omega} \eta + f_{\omega_x} \eta_x + f_{\omega_y} \eta_y) dx dy = 0 \quad \text{--- (A)}$$

Green's Theorem: $\iint_D \left(\frac{\partial P}{\partial x} + \frac{\partial Q}{\partial y} \right) dx dy = \int_C (P dy - Q dx)$

Take $P = \eta G$ and $Q = \eta F$, then

$$\iint_D \left(G \frac{\partial \eta}{\partial x} + F \frac{\partial \eta}{\partial y} \right) dx dy = - \iint_D \eta \left(\frac{\partial G}{\partial x} + \frac{\partial F}{\partial y} \right) dx dy + \int_C \eta (G dy - F dx) \quad \text{--- (B)}$$

Applying (B) to (A) where $G = \omega_x$; $F = \omega_y$

we get

$$F'(0) = \iint_D \eta \left[f_{\omega} - \frac{\partial}{\partial x} \omega_x - \frac{\partial}{\partial y} \omega_y \right] dx dy + \int_C \eta (f_{\omega_y} dy - f_{\omega_x} dx) = 0$$

$$\therefore \boxed{f_{\omega} = \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial \omega_x} \right) + \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial \omega_y} \right)}$$

Linear Transformation

A mapping T of a vector space X into a vector space Y is called a linear transformation ($T \in L(X, Y)$) if

- (i) $T(x+y) = T(x) + T(y) \quad \forall x, y \in X$ and
- (ii) $T(\alpha x) = \alpha T(x) \quad \forall x \in X$ and $\alpha \in F$ (field, i.e. real number or complex number)

Adjoint Transformation

For each $G \in L(X, X)$, \exists a unique $G^* \in L(X, X)$ called the adjoint of G , s.t. $(x, G^* y) = (G x, y) \quad \forall x, y \in X$

note: (x, y) is the inner product of two vectors x and y

e.g. For G a real matrix, $G^* = G^T$
 " " " complex " $G^* = \hat{G}^T$
 \hat{G} = complex conjugate of G .

Image Restoration

$$u_0 = Ru + \eta$$

\uparrow noisy image \downarrow linear operator representing blur (convolution) \uparrow noise \rightarrow noiseless image

The image can be written as a one dimensional vector.

$$u, u_0 \in \mathbb{R}^M; R \in M \times M$$

$$\inf_u \int_{\Omega} (u_0 - Ru)^T (u_0 - Ru) dx \quad (3.2)$$

$$f = (u_0 - Ru)^T (u_0 - Ru) = u_0^T u_0 - u_0^T R u - u^T R^T u_0 + u^T R^T R u$$

Applying Euler-Lagrange here gives $\delta_u = 0$

$$\delta_u = -R^T u_0 - R^T u_0 + 2R^T R u = 0$$

$$R^T u_0 - R^T R u = 0 \quad (3.3)$$

ill-posed problem if $R^T R$ singular.

Note:

$$\nabla_x (x^T A y) = A y$$

$$\nabla_x (y^T A x) = A^T y$$

$$\nabla_x (x^T A x) = 2 A x$$

Regularization

$$F(u) = \int_{\Omega} (u_0 - Ru)^T (u_0 - Ru) + \lambda (\nabla u^T \nabla u) dx \quad (3.4)$$

Here compared to $f(x, y, z)$; y is u and ∇u is z

Euler Lagrange $\frac{d}{dx} \delta_x = \delta_u$

$$\Rightarrow R^T R u - R u_0 - \lambda \Delta u = 0 \quad (3.5)$$

Note:

$$\delta_x = 2 \nabla u = 2 \frac{\partial u}{\partial x}$$

$$\frac{d}{dx} \delta_x = 2 \frac{\partial^2 u}{\partial x^2} = 2 \Delta u$$

Boundary condition $\frac{\partial u}{\partial n} = 0$ on $\partial \Omega$ (n is the outward normal to $\partial \Omega$)

Effect of Δu in (3.5) smooths out edges (Figure 3.3)

Problem is the L_2 norm on ∇u i.e. the $\nabla u^T \nabla u$ term

using 2 independent variables version of Euler-Lagrange

$$\nabla u^T \nabla u = \frac{\partial u}{\partial x} + \frac{\partial u}{\partial y} \quad ; u = u(x, y)$$

$$f_u = \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial u_x} \right) + \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial u_y} \right)$$

$$R^T R u - R u_0 - \lambda \Delta u = 0$$

$$\text{where } \Delta u = \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2}$$

Take L_1 norm based problem

$$E(u) = \frac{L}{2} \int_{\Omega} |u_0 - Ru|^2 dx + \lambda \int_{\Omega} \phi(|\nabla u|) dx \quad (3.6)$$

Euler-Lagrange gives

$$R^T R u - \lambda \operatorname{div} \left(\frac{\phi'(|\nabla u|) \nabla u}{|\nabla u|} \right) = R^T u_0 \quad (3.7)$$

note:

$$\text{gives a vector } v = (v_1 \ v_2 \ v_3)$$
$$\operatorname{div} v = \frac{\partial v_1}{\partial x} + \frac{\partial v_2}{\partial y} + \frac{\partial v_3}{\partial z}$$