# VARIATIONAL PRINCIPLES

### Jonathan Gai

### 5th May 2022

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Lecture 1 29 Apr. 2022

### Motivation

#### 0.1 The Brachistochrone Problem

**Problem.** Particle slides on a wire under influence of gravity between two fixed points *A*, *B*. Which shape of the wire gives the shortest travel time, starting from rest?

The travel time is  $T = \int_A^B \frac{d\ell}{v(x,y)}$ , and by energy conservation, and by energy conservation

$$\frac{1}{2}mv^2 + mgy = 0 \implies v = \sqrt{-2gy}.$$

So

$$T[y] = \frac{1}{\sqrt{2g}} \int_0^{x_2} \frac{\sqrt{1 + (y')^2}}{\sqrt{-y}} dx$$

subject to y(0) = 0,  $y(x_2) = y_2$ .

#### 0.2 Geodesics

**Problem.** What is the shortest path  $\gamma$  between two points A, B on a surface.

Take  $\Sigma = \mathbb{R}^2$ . The distance along  $\gamma$  is

$$D[y] = \int_A^B d\ell = \int_{x_1}^{x_2} \sqrt{1 + (y')^2} dx,$$

and we want to minimize D by varying  $\gamma$ .

#### 0.3 Introduction

In general, we want to minimize (maximize)

$$F[y] = \int_{x_1}^{x_2} f(x, y, y') \, \mathrm{d}x$$

among all functions s.t.  $y(x_1) = y_1$ ,  $y(x_2) = y_2$ . The expression is a *functional*. (A function on a space of functions). Functions map numbers to numbers, and functionals map functions to numbers.

Area under the graph is when f = y and the length of a curve is when  $f = \sqrt{1 + (y')^2}$ .

Calculus of variations finds extrema of functionals on space of functions.

**Notation.** •  $C(\mathbb{R})$  is space of continuous functions on  $\mathbb{R}$ .

- $C^k(\mathbb{R})$  is the space of functions with continuous kth derivatives.
- $C^k_{(\alpha,\beta)}$  is the space of functions with continuous kth derivatives and  $f(\alpha) = f(\beta)$ .

We need to specify the function space beforehand. It is a branch of functional analysis—Part III analysis on the space of functions, while Analysis I is analysis on the number line. Variational Principles follows principles in Nature, where the laws follow from extremizing functionals.

**Example** (Fermat's Principle). Light between two pints travel along paths which require least time.

**Example** (Principle of Least Action). Let T be the kinetic energy  $\frac{m|\dot{\mathbf{x}}|^2}{2}$  and potential energy  $V = V(\mathbf{x})$ .

$$S[\gamma] = \int_{t_1}^{t_2} (T - V) \, \mathrm{d}t$$

is minimized along paths of motion.

Leibniz commented on this, saying that "we live in the best of all worlds".

Feynman's take on this: "This is wrong. In quantum theorem the motion takes place along all possible path with different possibilities". [See Part III QFT]

In this course, we discuss

- 1. necessary condition for extrema of the Euler Lagrange Equations;
- 2. lots of examples (geometry, physics, problems with constraints);
- 3. second variation (some sufficient condition of extrema).

The following books will be useful

- 1. Gelfand Fomin "Calculus of Variations";
- 2. DAMTP notes (e.g. P. Townsend);
- 3. Lectures are self-contained.

**Lecture 2** 2 May 2022

## 1 Calculus for Functions of $\mathbb{R}^n$

We consider  $f \in C^2(\mathbb{R}^n)$ , that is, all functions  $f : \mathbb{R}^n \to \mathbb{R}^n$  with continuous second derivatives. The point  $\mathbf{a} \in \mathbb{R}$  is *stationary* if

$$\nabla f(\mathbf{a}) = \partial_1 f, \dots, \partial_n f|_{\mathbf{x} = \mathbf{a}} = 0, \quad \partial_i f = \frac{\partial f}{\partial x^i}.$$

If we expand near  $\mathbf{a}$ , we have

$$f(\mathbf{x}) = f(\mathbf{a}) + (\mathbf{x} - \mathbf{a}) \cdot \nabla |_{\mathbf{a}} + \frac{1}{2} (x_i - a_i)(x_j - a_j) \partial_{ij}^2 f|_{\mathbf{a}} + \mathcal{O}(|\mathbf{x} - \mathbf{a}|^2).$$

The *Hessian matrix* is  $H_{ij} = \partial_i \partial_j f = H_{ij}$ . We shift the origin to set  $\mathbf{a} = \mathbf{0}$ , and diagonalize  $H(\mathbf{0})$  by orthogonal transformation

$$H' = R^T H(\mathbf{0}) R = \begin{pmatrix} \lambda_1 & \lambda_2 & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}, \quad f(\mathbf{x}') - f(\mathbf{0}) = \frac{1}{2} \sum_i \lambda_i (x_i')^2 + \mathcal{O}(|\mathbf{x}|^2).$$

We consider several cases.

- 1. If all  $\lambda_i > 0$ ,  $f(\mathbf{x}') > f(\mathbf{0})$  in all directions, it is a *local minimum*.
- 2. If all  $\lambda_i < 0$ ,  $f(\mathbf{x}') < f(\mathbf{0})$  in all directions, it is a *local maximum*.
- 3. If some  $\lambda_i > 0$  and some  $\lambda_i < 0$ , f increases in some directions, and decreases in other directions. It is a *saddle point*.
- 4. If some  $\lambda_i = 0$  and the rest are of the same sign, we need to consider higher order terms in expansion.

The special case when n=2, we have  $\det(H)=\lambda_1\lambda_2$  and  $\mathrm{Tr}(H)=\lambda_1+\lambda_2$ . If  $\det(H)>0$ ,  $\mathrm{Tr}(H)>0$ , it is a local minimum. If  $\det(H)<0$ ,  $\mathrm{Tr}(H)<0$ , it is a local maximum. If  $\det(H)<0$ , it is a saddle point. If  $\det(H)=0$ , we need to look at third and higher derivatives.

**Remark.** 1. If  $f: D \to \mathbb{R}$  maps from a domain. We need to check boundary values too for global extrema.

2. If f is harmonic on  $\mathbb{R}^2$ , that is  $f_{xx} + f_{yy} = 0 \implies \text{Tr}(H) = 0$ . If  $D \subseteq \mathbb{R}^2$ , all turning points are saddle points, and the extrema must be on the boundary.

**Example.** Take  $f(x,y) = x^3 + Y^3 - 3xy$ , and we have  $\nabla f = (3x^2 - 3y, 3y^2 - 3x)$ . So the critical points are points where  $x^2 - y = 0$  and  $y^2 - x = 0$ . So the stationary points are x = 0, y = 0 and x = 1, y = 1. The Hessian matrix is

$$H = \begin{pmatrix} 6x & -3 \\ -3 & 6y \end{pmatrix}.$$

At (0,0), det(H) = -9 < 0, so it's a saddle point at (0,0). And at (1,1), det(H) = 27 > 0, Tr(H) = 12 > 0, it is a local minimum.

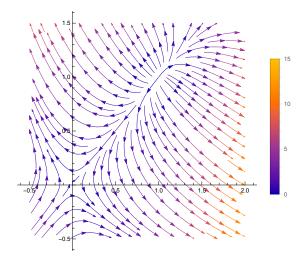


Figure 1: Plot of *f* showing the critical points

### 1.1 Constraints and Lagrange Multipliers

**Example.** Find the circle centered at (0,0) with the smallest radius, which intersects the parabola  $y = x^2 - 1$ .

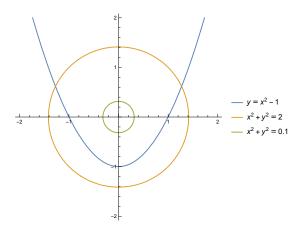


Figure 2: Circles and parabola

The direct method is to solve the constraints

$$f = x^2 + y^2 = x^2 + (x^2 - 1)^2 = x^4 - x^2 + 1.$$

If  $\partial f = 0$ , we have  $4x^3 - 2x = 0$ . So the solutions are (0, -1) with radius 1 and  $(\pm \frac{1}{\sqrt{2}}, -\frac{1}{2})$  with radius  $\frac{\sqrt{3}}{2} < 1$  which is the global minimum.

The other method is *Lagrange Multiplier*. Define  $h(x,y,\lambda)=f(x,y)-\lambda g(x,y)$  where g(x,y)=0 is the constraint and  $\lambda$  is the Lagrange multiplier. Now we extremize the function  $h(x,y)=x^2+y^2-\lambda(y-x^2+1)$  over 3 variables with no constraints; that is,

$$\begin{cases} \frac{\partial h}{\partial x} = 2x + 2\lambda x = 0\\ \frac{\partial h}{\partial y} = 2y - \lambda = 0\\ \frac{\partial h}{\partial \lambda} = y - x^2 + 1 = 0. \end{cases}$$

Note that the last equation just gives the constraint back. The solutions are (0, -1) and  $(\pm \frac{1}{\sqrt{2}}, -\frac{1}{2})$  as before will  $\lambda = -2$  and  $\lambda = -1$ , respectively.

We show that Lagrange Multiplier works by geometry.

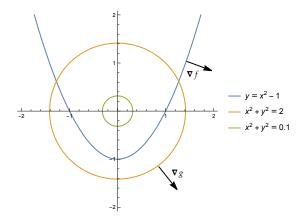


Figure 3: Gradients of the functions

We know that  $\nabla g$  is perpendicular to g=0 and  $\nabla f$  is perpendicular to f being constant. They must be parallel so  $\nabla f = \lambda \nabla g$  and  $\nabla (f - \lambda g) = 0$ .

If we have multiple constraints with  $f : \mathbb{R}^n \to \mathbb{R}$  subject to  $g_a(\mathbf{x}) = 0$  for a = 1, ..., k, then we let

$$h(x_1,\ldots,x_n,\lambda_1,\ldots,\lambda_k)=f-\sum_a\lambda_ag_a.$$

And solve for  $\frac{\partial h}{\partial x^i} = 0$  and  $\frac{\partial h}{\partial \lambda^i} = 0$ .

Lecture 3 4 May 2022

## 2 Euler-Lagrange Equations

If we want to extremize the functional

$$F[y] = \int_{\alpha}^{\beta} f(x, y, y') dx.$$
 (2.1)

for f given. It depends on y fixing only  $y(\alpha) = y_1$  and  $y(\beta) = y_2$ . We assume that

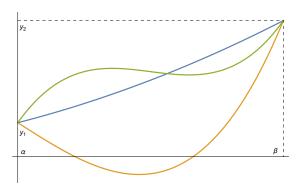


Figure 4: Different possible functions

the extremum exists, y = y(x), and consider small perturbation  $y \to y + \epsilon \eta(x)$  with  $\eta$  satisfying  $\eta(\alpha) = \eta(\beta) = 0$ . We want to compute  $F[y + \epsilon \eta]$ .

#### Lemma 2.1: Fundemental Lemma of Calculus of Variations

If  $g: [\alpha, \beta] \to \mathbb{R}$  is continuous on  $[\alpha, \beta]$  and  $\int_{\alpha}^{\beta} g(x) \cdot \eta(x) dx = 0$  for all  $\eta$  continuous on  $[\alpha, \beta]$  with  $\eta(\alpha) = \eta(\beta) = 0$ , then g = 0 on  $[\alpha, \beta]$ .

*Proof.* Assume otherwise,  $\exists \overline{x} \in (\alpha, \beta)$  such that  $g(\overline{x}) \neq 0$ . Without loss of generality,  $g(\overline{x}) > 0$ , then there exists interval  $[x_1, x_2] \subseteq (\alpha, \beta)$  such that f(x) > c > 0 for  $x \in [x_1, x_2]$  and some c. Take

$$\eta(x) = \begin{cases} (x - x_1)(x_2 - x), & \text{if } x \in [x_1, x_2] \\ 0, & \text{otherwise} \end{cases}$$
(2.2)

So

$$\int_{\alpha}^{\beta} g(x)\eta(x) dx > c \int_{x_1}^{x_2} (x - x_1)(x_2 - x) dx > 0.$$

**Remark.** The function  $\eta$  is an example of a bump function. There are  $C^k$  bump functions

$$\eta(x) = \begin{cases} ((x - x_1)(x_2 - x))^{k+1}, & \text{if } x \in [x_1, x_2] \\ 0, & \text{otherwise} \end{cases}.$$

Now we go back to equation (2.1).

$$F[y + \epsilon \eta] = \int_{\alpha}^{\beta} f(x, y + \epsilon \eta, y' + \epsilon \eta') dx$$

$$= F[y] + \epsilon \int_{\alpha}^{\beta} \frac{\partial f}{\partial y} \eta + \frac{\partial f}{\partial y'} \eta' dx + \mathcal{O}(\epsilon^{2})$$

$$= F[y] + \epsilon \left( \int_{\alpha}^{\beta} \frac{\partial f}{\partial y} \eta - \frac{d}{dx} \left( \frac{\partial f}{\partial y'} \right) \eta dx + \frac{\partial f}{\partial y'} \eta \Big|_{\beta}^{\alpha} \right)$$

At the extremum, we have  $F[y + \epsilon \eta] = F[y] + \mathcal{O}(\epsilon^2)$ ; that is, when  $\frac{\partial F}{\partial \epsilon}\Big|_{\epsilon=0} = 0$ . So

$$\int_{\alpha}^{\beta} \left( \frac{\partial f}{\partial y} - \frac{\mathrm{d}}{\mathrm{d}x} \left( \frac{\partial f}{\partial y'} \right) \right) \eta \, \mathrm{d}x.$$

By Lemma, we know

$$\frac{\mathrm{d}}{\mathrm{d}x} \left( \frac{\partial f}{\partial y'} \right) - \frac{\partial f}{\partial y} = 0. \tag{2.3}$$

This is the Euler-Lagrange equation, and it is a necessary condition for an extremum.

**Remark.** 1. Equation (2.3) is a second order ODE for y(x) with boundary conditions  $y(\alpha) = y_1$  and  $y(\beta) = y_2$ .

- 2. The left-hand side of equation (2.3) is denoted  $\frac{\delta F}{\delta y(x)}$ , and called *functional derivative*. Some books use  $\delta y = \epsilon \eta(x)$  as small variation.
- 3. There are other boundary conditions possible. For example,  $\frac{\partial f}{\partial y'}\Big|_{\alpha,\beta} = 0$ .
- 4. One needs to be careful with derivatives.  $\frac{\partial f}{\partial y} = \left(\frac{\partial f}{\partial y}\right)_{x,y'}$  treating x,y,y' as independent, and the *total derivative* is

$$\frac{\mathrm{d}h}{\mathrm{d}x} = \frac{\partial h}{\partial x} + \frac{\partial h}{\partial y}y' + \frac{\partial h}{\partial y'}y''$$
$$\frac{\mathrm{d}}{\mathrm{d}x} = \frac{\partial}{\partial x} + y'\frac{\partial}{\partial y} + y''\frac{\partial}{\partial y'}.$$

**Example.** If  $f = x \cdot ((y')^2 - y^2)$ , then

$$\partial_x f = (y')^2 - y^2, \quad \partial_y f = -2xy, \quad \partial_{y'} f = 2xy'.$$

So 
$$\frac{df}{dx} = (y')^2 - y^2 - 2xyy' + 2xy'y''$$
.

### 2.1 First Integrals of the Euler-Lagrange Equations

In some cases equation (2.3) (second order ODE) can be integrated once to a first order ODE called the *first integral*.

If f does not explicitly depend on y, that is,  $\frac{\partial f}{\partial y} = 0$ . Then equation (2.3) gives  $\frac{\mathrm{d}}{\mathrm{d}x} \left( \frac{\partial f}{\partial y'} \right) = 0$ , and

$$\frac{\partial f}{\partial y'} = \text{const.} \tag{2.4}$$

**Example.** Geodesics on  $\mathbb{R}^2$  have the functional

$$F[y] = \int_{\alpha}^{\beta} \sqrt{dx^2 + dy^2} = \int_{\alpha}^{\beta} \sqrt{1 + (y')^2} dx.$$

By equation (2.4), we must have  $\frac{\partial f}{\partial y'} = \frac{y'}{\sqrt{1+(y')^2}} = \text{const. So } y' \text{ is constant, calling it } m$ . We have y = mx + c, a straight line.