

Functional Derivatives

Newtonian, Hamiltonian, Lagrangian mechanics

Principle of least action

Legendre Transform

Noether's Theorem

Lecture 6

2/17

How should we
learn physics?

Newtonian Mechanics

$$F = m\ddot{x}$$

- Good for force balance
- Hrd to associate w/ invariants
- Coordinate dependant



Hamiltonian Mechanics

$$\dot{x} = J \nabla H$$

- Energy centric
- Good for designing schemes

Legendre Transform

Lagrangian Mechanics

$$x = \underset{\hat{x}}{\operatorname{argmin}} S(\hat{x})$$

- optimization centric
- easy constraints, treatment of symmetries

Lagrangian mechanics

Lagrangian

$$L = \underset{\substack{\uparrow \\ \text{kinetic}}}{T} - \underset{\substack{\uparrow \\ \text{potential}}}{V}$$

$$L = \int_{\Omega} \mathcal{L}[u] dx$$

\uparrow Lagrangian density

Action

$$S = \int_0^T L dt$$

Principle of least action

$$u = \underset{\hat{u}}{\operatorname{argmin}} \int_0^T \int_{\Omega} \mathcal{L}[u] dx dt$$

- Put informally, the path that a system will evolve through for a fixed initial and final state will be the $u(t)$ which minimizes the action.
- To make sense of this, we need to minimize expression like S . Like standard calculus: take a derivative of a function $f(x)$, set to zero to find extremal pts. Here though, x is a coordinate, f a function. u is a function, S is a functional
 $S: V \rightarrow \mathbb{R}, \quad V$ a space of functions

i.e. the input space is now infinite dimensional

A few definitions/options for how to do that. We'll show both so you understand what people mean by them, but then get less abstract.

Fredet derivative

- Let V be func. space, $F: V \rightarrow \mathbb{R}$. The differential $\delta F[u]$ is defined as, for $g \in V$

$$F[u + g] - F[u] = \delta F[u] + \varepsilon \|g\|$$

such that $\lim_{\varepsilon \rightarrow 0} \varepsilon \|g\| = 0$

- This is a challenging definition to work with

Gâteaux derivative

Idea Introduce a scalar parameter and use the vector calc def of a directional derivative

$$\begin{aligned}\delta F[u, g] &= \lim_{\varepsilon \rightarrow 0} \frac{F[u + \varepsilon g] - F[u]}{\varepsilon} \\ &= \left. \frac{d}{d\varepsilon} F[u + \varepsilon g] \right|_{\varepsilon=0}\end{aligned}$$

To compute

Given $F[u] = \int_{\Omega} \mathcal{L}(x, u, D\tilde{u}) dx$

δF is defined via, g varying a bit,

$$(\delta F, g) = \lim_{\varepsilon \rightarrow 0} \frac{F[u + \varepsilon g] - F[u]}{\varepsilon}$$

Example Kinetic energy

$$K = \frac{1}{2} \int \rho u^2 dx, \text{ take var. w.r.t. } u$$

$$(\delta_u K, \delta u) = \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int_{\Omega} \rho \frac{(u + \varepsilon \delta u)^2}{2} - \frac{\rho u^2}{2} dx$$

$$= \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int \rho \left[\frac{u^2}{2} + \varepsilon u \delta u + \frac{\varepsilon^2 \delta u^2}{2} - \frac{\rho u^2}{2} \right] dx$$

$$= \lim_{\varepsilon \rightarrow 0} \int \rho u \delta u + \frac{\varepsilon}{2} \rho \delta u^2$$

$$= \int \rho u du$$

$$= (\rho u, \delta u)$$

And so we identify

$$\delta_u K = \rho u \quad (\text{the momentum})$$

Be careful physicists will derive by setting δu to δu only ok for const. functionals

Further properties

Denote $\frac{\delta}{\delta u} F[u] = \delta_u F[u]$

• Linearity $\frac{\delta}{\delta u} (\lambda F[u] + \mu G[u]) = \lambda \delta_u F[u] + \mu \delta_u G[u]$

• Product rule $\frac{\delta}{\delta u} (F[u] G[u]) = \delta_u F G + F \delta_u G$

• Chain rule $\frac{\delta}{\delta u} (F[g(u)]) = \frac{\delta F[g(u)]}{\delta g(u)} \frac{dg(u)}{du}$

The Euler-Lagrange equations

Assume a generic functional density

$$F[u] = \int_{\Omega} f(x, u, \nabla u) dx$$

$$(\delta_u F, \delta u) = \left[\frac{d}{dz} \int f(x, u + z \delta u, \nabla u + z \nabla \delta u) dx \right]_{z=0}$$

chain rule

$$= \int_{\Omega} \frac{\partial f}{\partial u} \delta u + \frac{\partial f}{\partial \nabla u} \cdot \nabla \delta u dx$$

$$= \int_{\Omega} \left(\frac{\partial f}{\partial u} - \nabla \cdot \frac{\partial f}{\partial \nabla u} \right) \cdot \delta u + \int_{\partial \Omega} \frac{\partial f}{\partial \nabla u} \cdot \nabla u \cdot dA$$

Assume vanishing

$$\text{So } \delta_u F = \frac{\partial f}{\partial u} - \sum_i \frac{\partial}{\partial x_i} \frac{\partial f}{\partial x_i}$$

$$= \frac{\partial f}{\partial u} - \sum_i \frac{\partial}{\partial x_i} \frac{\partial f}{\partial (\partial_{x_i} u)}$$

- We can just apply this as a formula and skip all the limits
- For different form f , get a different ~~equ~~ eqn

ex

$$F[u] = \int_0^1 f(x, u, \partial_x u, \partial_{xx} u) dx \Rightarrow$$

$$\delta_u F = \frac{\partial f}{\partial u} - \frac{d}{dx} \frac{\partial f}{\partial (\partial_x u)} + \frac{d^2}{dx^2} \frac{\partial f}{\partial (\partial_{xx} u)}$$

Returning to the least action principle

$$S[u] = \int_a^b \int_{\Sigma} \mathcal{L}(x, u, \dot{u}) \, dx \, dt$$

Finally we can state that the path which admits an extremal value of S satisfies

$$\delta_u S[u] = 0$$

Applying the E-L eqns

$$\left[\frac{\partial \mathcal{L}}{\partial u} - \frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{u}} = 0 \right]$$

Again, depending on \mathcal{L} we can get different versions of this

Ex Linear pendulum

$$T = \frac{1}{2} m v^2 = \frac{1}{2} m l^2 \dot{\theta}^2$$

$$U = mgh$$

$$h = l(1 - \cos \theta) \approx \frac{1}{2} l \theta^2$$

$$L = \frac{1}{2} m l^2 \dot{\theta}^2 - \frac{1}{2} m g l \theta^2$$

$$\partial_{\theta} L = -m g l \theta$$

$$\partial_{\dot{\theta}} L = m l^2 \dot{\theta}$$

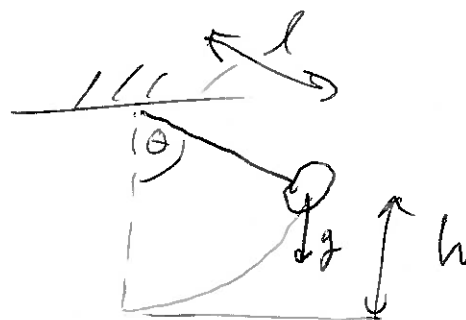
E-L eqn

$$\partial_{\theta} L - \frac{d}{dt} \partial_{\dot{\theta}} L = 0$$

\Rightarrow

$$-m g l \theta - \frac{d}{dt} m l^2 \dot{\theta} = 0$$

$$\ddot{\theta} = -\frac{g}{l} \theta$$



Legendre transform

A technique to switch between the Lagrangian and Hamiltonian description

Given a Lagrangian

$$L(q_1, \dots, q_N, \dot{q}_1, \dots, \dot{q}_N)$$

The Legendre transform induces the conjugate momenta p_1, \dots, p_N

$$\begin{cases} H(q_1, \dots, q_N, p_1, \dots, p_N) = \sum_i p_i \dot{q}_i - L(q_1, \dots, q_N, \dot{q}_1, \dots, \dot{q}_N) \\ p_i = \partial_{\dot{q}_i} L \end{cases}$$

Ex Getting Hamiltonian from Lagrangian for pendulum

$$q = \theta, \quad L = \frac{1}{2} m l^2 \dot{\theta}^2 - \frac{1}{2} m g l \theta^2$$

$$p_\theta = \partial_{\dot{\theta}} L = m l^2 \dot{\theta} \Rightarrow \dot{\theta} = \frac{1}{m l^2} p$$

$$L = \frac{1}{2} (m l^2)^{-1} p^2 - \frac{1}{2} m g l \theta^2$$

$$H = p \dot{\theta} - L$$

$$= \frac{1}{m l^2} p^2 - \frac{1}{2} (m l^2)^{-1} p^2 - \frac{1}{2} m g l \theta^2$$

$$= \frac{p^2}{2 m l^2} - \frac{m g l \theta^2}{2}$$

Constrained Lagrangians

def If constraint is equality and a function of coordinate only (no \dot{q} 's) it is holonomic

$$f_i(q, t) = 0$$

We can augment the Lagrangian w/ a Lagrange multiplier

$$\hat{L}(q, \dot{q}, \lambda) = L(q, \dot{q}) + \sum_i \lambda_i f_i(q, t)$$

Where λ is a field (in contrast to KKT we considered previously)

Applying Euler-Lagrange we obtain equations of motion

$$\begin{cases} \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}_i} \right) = \frac{\partial L}{\partial q_i} + \lambda(t) \frac{\partial f_i}{\partial q_i} \\ f_i(q, t) = 0 \end{cases}$$

Noether's theorem

Informally - symmetries which leave the Lagrangian unchanged each have a corresponding conserved quantity.

Consider the maps

$$t \rightarrow t' = t + \delta t$$
$$q \rightarrow q' = q + \delta q$$

Assume N of such maps

$$\delta t = \sum_r \epsilon_r T_r$$
$$\delta q = \sum_r \epsilon_r Q_r$$

Then $\left(\frac{\partial L}{\partial \dot{q}} \cdot \dot{q} - L \right) T_r - \frac{\partial L}{\partial \dot{q}} \cdot Q_r$

Are conserved quantities.

Examples

(1) Time Invariance

$$t \rightarrow t + \delta t$$

$$N=1, T=1, Q=0$$

$$\frac{\partial L}{\partial \dot{q}} \cdot \dot{q} - L$$

is conserved (this is actually the Legendre transform providing H !)

(2) Translation invariance

$$q \rightarrow q + \delta q$$

$$N=1, T=0, Q=1$$

$$\frac{\partial L}{\partial \dot{q}} = \text{const}$$

(this is the definition of conjugate momentum!)

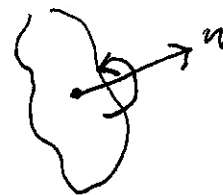
③ Rotational invariance

$L = r \times p$ is angular momentum

Consider

$$r \rightarrow r + \delta\theta \, n \times r$$

Rotation of $\delta\theta$ about a given axis n .



$$N=1, T=0 \quad Q = n \times r$$

$$\begin{aligned} \text{const} &= \frac{\partial L}{\partial \dot{q}} \cdot Q = p \cdot (n \times r) \\ &= n \cdot (r \times p) \\ &= n \cdot L \end{aligned}$$

When machine learning dynamics, we will design architectures which respect this set of symmetries