Math 222A Lecture 9 Notes

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1 Solutions to Hamilton-Jacobi Equations via Calculus of Variations

1.1 Recap: Connecting Hamilton-Jacobi equations to calculus of variations using the Legendre transform

Last time, we wanted to compare Hamilton-Jacobi equations to calculus of variations. The Hamilton-Jacobi equations are of the form

$$\begin{cases} u_t + H(x, \partial u) = 0 & \text{in } \mathbb{R} \times \mathbb{R}^n \\ u(0) = u_0 & \text{in } \mathbb{R}. \end{cases}$$

The characteristics given to this equation are

$$\begin{cases} \dot{x} = H_p \\ \dot{p} = -H_x \\ \dot{z} = p \cdot H_p - H, \end{cases}$$

with initial data $x(0) = x_0$ and $p(0) = \partial u_0$. The first two equations are called the **Hamilton flow**.

In calculus of variations, we have a **Lagrangian** $L : \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}$, and we want to minimize an **action functional**

$$\min_{x \in \mathcal{A}} \underbrace{\int_{0}^{T} L(x, \dot{x}) \, dt}_{\mathcal{L}(x)},$$

where $\mathcal{A} = \{x : [0,T] \to \mathbb{R} \text{ Lipschitz } | x(0) = x_0, x(T) = x_T\}$. Minimizers satisfy the **Euler-Lagrange equation**

$$L_x(x,\dot{x}) - \frac{d}{dt}L_q(x,\dot{x}) = 0.$$

Last time, we connected these two setups. We saw that

- L is strictly convex and coercive if and only if H is strictly convex and coercive.
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$$H(x,p) = \max_{q \in \mathbb{R}^n} -L(x,q) + p \cdot q,$$

which is maximized at $p = L_q(x, q)$. This relation gives

$$H(x,p) + L(x,q) \ge p \cdot q$$

with equality when $p = L_q(x, q)$. This expression is symmetric in p and q, so it allows us to cast q in terms of p: $q = H_p(x, p)$. This relationship is known as the **Legendre** transform.

Remark 1.1. The Legendre transform well-defined and is an involution, only assuming convexity.

Example 1.1. If we remove strict convexity and coercivity, we can get functions which are not defined everywhere. For example, take

$$\begin{cases} L(0) = 0\\ L(q) = \infty \quad q \neq 0. \end{cases}$$

What is H in this case?

We have not incorporated the initial data of the Hamilton-Jacobi equations into our calculus of variations. We will do this by adding $u_0(x_0)$ to the minimization problem (so when T = 0, we get $u_0(x_0)$) and removing the condition $x(0) = x_0$ from our set \mathcal{A} . So we are minimizing

$$\min_{x \in \mathcal{A}} \int_0^T L(x, \dot{x}) \, dt + u_0(x_0) = u(T, x_T),$$

with $\mathcal{A} = \{x : [0, T] \to \mathbb{R} \text{ Lipschitz } | x(T) = x_T \}.$

1.2 Existence of minimizers for the Euler-Lagrange equation

We want to prove the following:

Theorem 1.1. The minimal value function $u(T, x_T)$ in the calculus of variations is the solution to the Hamilton-Jacobi equations.

First, we should ask: Does a minimum solution to the Euler-Lagrange equation exist? The answer is yes, as long as L is convex, coercive, and Lipschitz in x and if $u_0 \in$ Lip. However, there is no guarantee of uniqueness. We will not prove this, but here is some intuition:

Here is the trivial case:

Proposition 1.1. Suppose we have a continuous function $F : K \to \mathbb{R}$ with K compact. Then min F is attained.

Proof. Let x_n be a minimizing sequence: $F(x_n) \to \inf F$. Then $x_n \to x_0$ along a subsequence. Then $F(x_n) \to F(x_0)$, so x_0 is the minimizer.

What if we try to apply this to calculus of variations? Suppose we have a minimizing sequence $x_n : [0,T] \to \mathbb{R}^n$. Then $\mathcal{L}(x_n) \to u(T,x_T)$ but in what topology? Is x_n in a bounded set? We know that $\mathcal{L}(x_n)$ is bounded. If $L(x,q) = q^2$, for example, we could conclude that $\int_0^T (\dot{x}_n)^2 \leq c$. Then \dot{x}_n is bounded in $L^2([0,T])$. This would imply that x_n is bounded in $C^{1/2}$ using Hölder's inequality: $(|x_n(t) - x_n(s)| \leq c|t - s|^{1/2})$. This implies that x_n is equicontinuous (and equibounded by the $x(T) = x_T$ assumption). So the Arzelà Ascoli theorem says that $x_n \to x$ uniformly. Then

$$\lim_{n \to \infty} \mathcal{L}(x_n) = \lim_{n \to \infty} \int_0^T L(x_n, \dot{x}_n) \, dt + \underbrace{u(x_{n,0})}_{\to u_0(x_0)}$$

We can pass to the limit without a problem for x, but convergence with respect to \dot{x} is trouble.

The limit of the integral may not exist, but maybe we can hope for

$$\int_0^T L(x, \dot{x}) \, dt \le \liminf_{n \to \infty} \int_0^T L(x_n, \dot{x}_n) \, dt.$$

This is lower semicontinuity for the map $x \mapsto \mathcal{L}(x)$. The key observation is that convexity of \mathcal{L} implies lower semicontinuity of \mathcal{L} :

Proof. The convexity inequality tells us that

$$L(\dot{x}_n) \ge L(\dot{x}) + L_q(\dot{x})(\dot{x}_n - \dot{x}).$$

Integrating gives us

$$\int_0^T L(\dot{x}_n) \, dt \ge \int_0^T L(\dot{x}) \, dt + \int_0^T L_q(\dot{x})(\dot{x} - \dot{x}_n) \, dt$$

We are done if $\lim_{n\to\infty} \int L_q(\dot{x})(\dot{x}_n - \dot{x}) = 0$. We have replaced our nonlinear dependence on $\dot{x}_n - \dot{x}$ by a linear property.

Since $\dot{x} \in L^2$, we can approximate $L_q(\dot{x})$ by smooth functions. Suppose $y_k \in C^{\infty}$ with $y_k \to L(\dot{x})$ in L^2 . It is enough to see that

$$\lim_{n \to \infty} \int_0^T y_k(\dot{x}_n - \dot{x}) \, dt = 0$$

In this context, we can integrate by parts. The integral equals

$$\int_0^T y_k(\dot{x}_n - x) \, dt = \int_0^T \dot{y}_k(x_n - x) \, dt + y_k(x_n - x) \big|_0^T \xrightarrow{n \to \infty} 0$$

by uniform convergence of $x_n \to x$.

Example 1.2. Recall our double well potential.



In this case, if x_n is a wiggle approximating the 0 trajectory, we have $L(\dot{x}_n) = 0$ by $L(\dot{x}) = L(0) > 0$.

Remark 1.2. The Hamilton-Jacobi equation can be solved for a short time using characteristics. In calculus of variations, the analogue turns out to be that minimizers are unique for a short time.

We want to think of two minimizers in calculus of variations as characteristics that intersect.



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1.3 Proving that Euler-Lagrange equation minimizers solve Hamilton-Jacobi equations

Here is the "proof" of our theorem.

Proof. Suppose x is a minimizer for the action functional. We can choose a intermediate point t, and first minimize relative to the time t.



$$\min_{x} \int_{0}^{T} L(x, \dot{x}) dt + u_{0}(x_{0}) = \min_{x} \int_{0}^{t} L(x, \dot{x}) ds + u_{0}(x_{0}) + \int_{t}^{T} L(x, \dot{x}) ds$$

If $x|_{[0,T]}$ is a minimizer, then $x|_{[0,t]}$ is also a minimizer. So

$$u(x_T, x_0) = \min u(x_t, x_0) + \int_t^T L(x, \dot{x}) \, ds$$

This is called the **dynamic programming principle**.¹ This principle tells us that for minimizers,

$$u(x_T, x_0) = u(x_t, x_0) + \int_t^T L(x, \dot{x}) \, ds,$$

which we can differentiate with respect to t to get

$$\frac{d}{dt}u(x_t, x_0) = L(x, \dot{x})$$
$$= p \cdot q - H(x, p)$$
$$= p \cdot H_p - H.$$

We conclude that $u(t, x_t)$ from the calculus of variations is the same as the $u(t, x_t)$ from the Hamilton-Jacobi equation because they solve the same equation with the same initial data at time 0.

¹This is discussed near the end of Evans' book.

Remark 1.3. This is not an entirely correct proof. How do we know that there is an optimal trajectory starting at x_0 ? If the time is short enough, we can guarantee a minimizer starting at x_0 , but this is exactly the issue of uniqueness of minimizers. This proof can be made rigorous for short times.

Remark 1.4. More generally, this is related to control theory, where we try to find

$$u(x_0, T) = \min \int_0^T L(x, u) \, dt + u_0(x(0)), \qquad \dot{x} = h(x, f)$$

Here, we can choose some weight of influence by changing f, and we are trying to optimize some cost functional. The function $u(x_0, T)$ solves a Hamilton-Jacobi equation.

We can think of our calculus of variations problem as the case where the ODE for x is given by $\dot{x} = f$.

Remark 1.5. Calculus of variations allows us to obtain meaningful solutions for Hamilton-Jacobi equations after characteristics begin to intersect. Instead of picking which characteristic to continue, we can just look for a minimizer for a calculus of variations problem in longer time.