

In particular we can identify the parameter s with the time t . Equation (11)(a) for the case at hand reads

$$\begin{cases} \dot{p}^i(s) = -H_{x_i}(\mathbf{p}(s), \mathbf{x}(s)) & (i = 1, \dots, n) \\ \dot{p}^{n+1}(s) = 0; \end{cases}$$

the equation (11)(b) is

$$\begin{aligned} \dot{z}(s) &= D_p H(\mathbf{p}(s), \mathbf{x}(s)) \cdot \mathbf{p}(s) + p^{n+1}(s) \\ &= D_p H(\mathbf{p}(s), \mathbf{x}(s)) \cdot \mathbf{p}(s) - H(\mathbf{p}(s), \mathbf{x}(s)). \end{aligned}$$

In summary, the characteristic equations for the Hamilton–Jacobi equation are

$$(64) \quad \begin{cases} \text{(a)} & \dot{\mathbf{p}}(s) = -D_x H(\mathbf{p}(s), \mathbf{x}(s)) \\ \text{(b)} & \dot{z}(s) = D_p H(\mathbf{p}(s), \mathbf{x}(s)) \cdot \mathbf{p}(s) - H(\mathbf{p}(s), \mathbf{x}(s)) \\ \text{(c)} & \dot{\mathbf{x}}(s) = D_p H(\mathbf{p}(s), \mathbf{x}(s)) \end{cases}$$

for $\mathbf{p}(\cdot) = (p^1(\cdot), \dots, p^n(\cdot))$, $z(\cdot)$, and $\mathbf{x}(\cdot) = (x^1(\cdot), \dots, x^n(\cdot))$.

The first and third of these equalities,

$$(65) \quad \begin{cases} \dot{\mathbf{x}} = D_p H(\mathbf{p}, \mathbf{x}) \\ \dot{\mathbf{p}} = -D_x H(\mathbf{p}, \mathbf{x}), \end{cases}$$

are called *Hamilton's equations*. We will discuss these ODE and their relationship to the Hamilton–Jacobi equation in much more detail, just below in §3.3. Observe that the equation for $z(\cdot)$ is trivial, once $\mathbf{x}(\cdot)$ and $\mathbf{p}(\cdot)$ have been found by solving Hamilton's equations. \square

As for conservation laws (Example 5), the initial-value problem for the Hamilton–Jacobi equation does not in general have a smooth solution u lasting for all times $t > 0$.

3.3. INTRODUCTION TO HAMILTON–JACOBI EQUATIONS

In this section we study in some detail the initial-value problem for the Hamilton–Jacobi equation:

$$(1) \quad \begin{cases} u_t + H(Du) = 0 & \text{in } \mathbb{R}^n \times (0, \infty) \\ u = g & \text{on } \mathbb{R}^n \times \{t = 0\}. \end{cases}$$

Here $u : \mathbb{R}^n \times [0, \infty) \rightarrow \mathbb{R}$ is the unknown, $u = u(x, t)$, and $Du = D_x u = (u_{x_1}, \dots, u_{x_n})$. We are given the *Hamiltonian* $H : \mathbb{R}^n \rightarrow \mathbb{R}$ and the initial function $g : \mathbb{R}^n \rightarrow \mathbb{R}$.

Our goal is to find a formula for an appropriate weak or generalized solution, existing for all times $t > 0$, even after the method of characteristics has failed.

3.3.1. Calculus of variations, Hamilton’s ODE.

Remember from §3.2.5 that two of the characteristic equations associated with the Hamilton–Jacobi PDE

$$u_t + H(Du, x) = 0$$

are Hamilton’s ODE

$$\begin{cases} \dot{\mathbf{x}} = D_p H(\mathbf{p}, \mathbf{x}) \\ \dot{\mathbf{p}} = -D_x H(\mathbf{p}, \mathbf{x}), \end{cases}$$

which arise in the classical calculus of variations and in mechanics. (Note the x -dependence in H here.) In this section we recall the derivation of these ODE from a variational principle. We will then discover in §3.3.2 that this discussion contains a clue as to how to build a weak solution of the initial-value problem (1).

a. The calculus of variations. Assume that $L : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ is a given smooth function, hereafter called the *Lagrangian*.

NOTATION. We write

$$L = L(v, x) = L(v_1, \dots, v_n, x_1, \dots, x_n) \quad (v, x \in \mathbb{R}^n)$$

and

$$\begin{cases} D_v L = (L_{v_1} \cdots L_{v_n}) \\ D_x L = (L_{x_1} \cdots L_{x_n}). \end{cases}$$

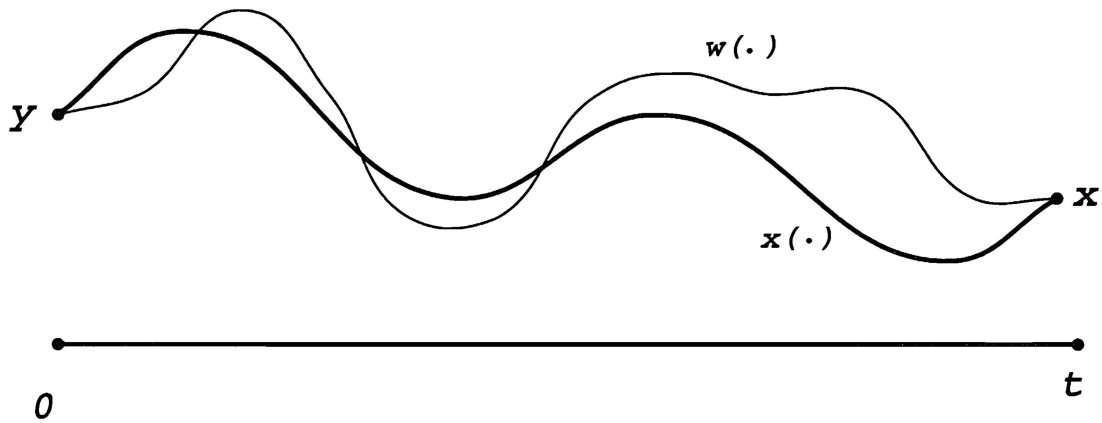
Thus in the formula (2) below “ v ” is the name of the variable for which we substitute $\dot{\mathbf{w}}(s)$, and “ x ” is the variable for which we substitute $\mathbf{w}(s)$. □

Now fix two points $x, y \in \mathbb{R}^n$ and a time $t > 0$. We introduce then the *action* functional

$$(2) \quad I[\mathbf{w}(\cdot)] := \int_0^t L(\dot{\mathbf{w}}(s), \mathbf{w}(s)) ds \quad \left(\dot{\cdot} = \frac{d}{ds} \right),$$

defined for functions $\mathbf{w}(\cdot) = (w^1(\cdot), w^2(\cdot), \dots, w^n(\cdot))$ belonging to the *admissible class*

$$\mathcal{A} := \{ \mathbf{w}(\cdot) \in C^2([0, t]; \mathbb{R}^n) \mid \mathbf{w}(0) = y, \mathbf{w}(t) = x \}.$$



A problem in the calculus of variations

Thus a C^2 curve $\mathbf{w}(\cdot)$ lies in \mathcal{A} if it starts at the point y at time 0 and reaches the point x at time t .

A basic problem in the *calculus of variations* is to find a curve $\mathbf{x}(\cdot) \in \mathcal{A}$ satisfying

$$(3) \quad I[\mathbf{x}(\cdot)] = \min_{\mathbf{w}(\cdot) \in \mathcal{A}} I[\mathbf{w}(\cdot)].$$

That is, we are asking for a function $\mathbf{x}(\cdot)$ which minimizes the functional $I[\cdot]$ among all admissible candidates $\mathbf{w}(\cdot) \in \mathcal{A}$.

We assume next that there in fact exists a function $\mathbf{x}(\cdot) \in \mathcal{A}$ satisfying our calculus of variations problem and will deduce some of its properties.

THEOREM 1 (Euler–Lagrange equations). *The function $\mathbf{x}(\cdot)$ solves the system of Euler–Lagrange equations*

$$(4) \quad -\frac{d}{ds} (D_v L(\dot{\mathbf{x}}(s), \mathbf{x}(s))) + D_x L(\dot{\mathbf{x}}(s), \mathbf{x}(s)) = 0 \quad (0 \leq s \leq t).$$

This is a vector equation, consisting of n coupled second-order equations.

Proof. 1. Choose a smooth function $\mathbf{y} : [0, t] \rightarrow \mathbb{R}^n$, $\mathbf{y}(\cdot) = (y^1(\cdot), \dots, y^n(\cdot))$, satisfying

$$(5) \quad \mathbf{y}(0) = \mathbf{y}(t) = 0,$$

and define for $\tau \in \mathbb{R}$

$$(6) \quad \mathbf{w}(\cdot) := \mathbf{x}(\cdot) + \tau \mathbf{y}(\cdot).$$

Then $\mathbf{w}(\cdot) \in \mathcal{A}$ and so

$$I[\mathbf{x}(\cdot)] \leq I[\mathbf{w}(\cdot)].$$

Thus the real-valued function

$$i(\tau) := I[\mathbf{x}(\cdot) + \tau\mathbf{y}(\cdot)]$$

has a minimum at $\tau = 0$, and consequently

$$(7) \quad i'(0) = 0 \quad \left(' = \frac{d}{d\tau} \right),$$

provided $i'(0)$ exists.

2. We explicitly compute this derivative. Observe

$$i(\tau) = \int_0^t L(\dot{\mathbf{x}}(s) + \tau\dot{\mathbf{y}}(s), \mathbf{x}(s) + \tau\mathbf{y}(s)) ds,$$

and so

$$i'(\tau) = \int_0^t \sum_{i=1}^n L_{v_i}(\dot{\mathbf{x}} + \tau\dot{\mathbf{y}}, x + \tau\mathbf{y}) \dot{y}^i + L_{x_i}(\dot{\mathbf{x}} + \tau\dot{\mathbf{y}}, x + \tau\mathbf{y}) y^i ds.$$

Set $\tau = 0$ and remember (7):

$$0 = i'(0) = \int_0^t \sum_{i=1}^n L_{v_i}(\dot{\mathbf{x}}, \mathbf{x}) \dot{y}^i + L_{x_i}(\dot{\mathbf{x}}, \mathbf{x}) y^i ds.$$

We recall (5) and then integrate by parts in the first term inside the integral, to discover

$$0 = \sum_{i=1}^n \int_0^t \left[-\frac{d}{ds} (L_{v_i}(\dot{\mathbf{x}}, \mathbf{x})) + L_{x_i}(\dot{\mathbf{x}}, \mathbf{x}) \right] y^i ds.$$

This identity is valid for all smooth functions $\mathbf{y}(\cdot) = (y^1(\cdot), \dots, y^n(\cdot))$ satisfying the boundary conditions (5), and so for $0 \leq s \leq t$

$$-\frac{d}{ds} (L_{v_i}(\dot{\mathbf{x}}, \mathbf{x})) + L_{x_i}(\dot{\mathbf{x}}, \mathbf{x}) = 0 \quad (i = 1, \dots, n). \quad \square$$

Critical points. We have just demonstrated that any minimizer $\mathbf{x}(\cdot) \in \mathcal{A}$ of $I[\cdot]$ solves the Euler–Lagrange system of ODE. It is of course possible that a curve $\mathbf{x}(\cdot) \in \mathcal{A}$ may solve the Euler–Lagrange equations without necessarily being a minimizer: in this case we say $\mathbf{x}(\cdot)$ is a *critical point* of $I[\cdot]$. So every minimizer is a critical point, but a critical point need not be a minimizer.

Example. If $L(v, x) = \frac{1}{2}m|v|^2 - \phi(x)$, where $m > 0$, the corresponding Euler–Lagrange equation is

$$m\ddot{\mathbf{x}}(s) = \mathbf{f}(\mathbf{x}(s))$$

for $\mathbf{f} := -D\phi$. This is Newton’s law for the motion of a particle of mass m moving in the force field \mathbf{f} generated by the potential ϕ . (See Feynman–Leighton–Sands [F-L-S, Chapter 19].) \square

b. Hamilton’s equations. We now transform the Euler–Lagrange equations, a system of n second-order ODE, into Hamilton’s equations, a system of $2n$ first-order ODE. We hereafter assume the C^2 function $\mathbf{x}(\cdot)$ is a critical point of the action functional and thus solves the Euler–Lagrange equations (4).

First we set

$$(8) \quad \mathbf{p}(s) := D_v L(\dot{\mathbf{x}}(s), \mathbf{x}(s)) \quad (0 \leq s \leq t);$$

$\mathbf{p}(\cdot)$ is called the *generalized momentum* corresponding to the *position* $\mathbf{x}(\cdot)$ and *velocity* $\dot{\mathbf{x}}(\cdot)$. We next make this important hypothesis:

$$(9) \quad \left\{ \begin{array}{l} \text{Suppose for all } x, p \in \mathbb{R}^n \text{ that the equation} \\ \quad \quad \quad p = D_v L(v, x) \\ \text{can be uniquely solved for } v \text{ as a smooth} \\ \text{function of } p \text{ and } x, v = \mathbf{v}(p, x). \end{array} \right.$$

We will examine this assumption in more detail later: see §3.3.2.

DEFINITION. The Hamiltonian H associated with the Lagrangian L is

$$H(p, x) := p \cdot \mathbf{v}(p, x) - L(\mathbf{v}(p, x), x) \quad (p, x \in \mathbb{R}^n),$$

where the function $\mathbf{v}(\cdot)$ is defined implicitly by (9).

Example (continued). The Hamiltonian corresponding to the Lagrangian $L(v, x) = \frac{1}{2}m|v|^2 - \phi(x)$ is

$$H(p, x) = \frac{1}{2m}|p|^2 + \phi(x).$$

The Hamiltonian is thus the total energy, the sum of the kinetic and potential energies (whereas the Lagrangian is the difference between the kinetic and potential energies). \square

Next we rewrite the Euler–Lagrange equations in terms of $\mathbf{p}(\cdot), \mathbf{x}(\cdot)$:

THEOREM 2 (Derivation of Hamilton’s ODE). *The functions $\mathbf{x}(\cdot)$ and $\mathbf{p}(\cdot)$ satisfy Hamilton’s equations:*

$$(10) \quad \begin{cases} \dot{\mathbf{x}}(s) = D_p H(\mathbf{p}(s), \mathbf{x}(s)) \\ \dot{\mathbf{p}}(s) = -D_x H(\mathbf{p}(s), \mathbf{x}(s)) \end{cases}$$

for $0 \leq s \leq t$. Furthermore,

the mapping $s \mapsto H(\mathbf{p}(s), \mathbf{x}(s))$ is constant.

The equations (10) comprise a coupled system of $2n$ first-order ODE for $\mathbf{x}(\cdot) = (x^1(\cdot), \dots, x^n(\cdot))$ and $\mathbf{p}(\cdot) = (p^1(\cdot), \dots, p^n(\cdot))$ (defined by (8)).

Proof. First note from (8) and (9) that $\dot{\mathbf{x}}(s) = \mathbf{v}(\mathbf{p}(s), \mathbf{x}(s))$.

Let us hereafter write $\mathbf{v}(\cdot) = (v^1(\cdot), \dots, v^n(\cdot))$. We compute for $i = 1, \dots, n$ that

$$\begin{aligned} H_{x_i}(p, x) &= \sum_{k=1}^n p_k v_{x_i}^k(p, x) - L_{v_k}(\mathbf{v}(p, x), x) v_{x_i}^k(p, x) - L_{x_i}(\mathbf{v}(p, x), x) \\ &= -L_{x_i}(q, x) \quad \text{according to (9)} \end{aligned}$$

and

$$\begin{aligned} H_{p_i}(p, x) &= v^i(p, x) + \sum_{k=1}^n p_k v_{p_i}^k(p, x) - L_{v_k}(\mathbf{v}(p, x), x) v_{p_i}^k(p, x) \\ &= v^i(p, x), \quad \text{again by (9)}. \end{aligned}$$

Thus

$$H_{p_i}(\mathbf{p}(s), \mathbf{x}(s)) = v^i(\mathbf{p}(s), \mathbf{x}(s)) = \dot{x}^i(s),$$

and likewise

$$\begin{aligned} H_{x_i}(\mathbf{p}(s), \mathbf{x}(s)) &= -L_{x_i}(\mathbf{v}(\mathbf{p}(s), \mathbf{x}(s)), \mathbf{x}(s)) = -L_{x_i}(\dot{\mathbf{x}}(s), \mathbf{x}(s)) \\ &= -\frac{d}{ds} (L_{v_i}(\dot{\mathbf{x}}(s), \mathbf{x}(s))) \quad \text{according to (4)} \\ &= -\dot{p}^i(s). \end{aligned}$$

Finally, observe

$$\begin{aligned} \frac{d}{ds} H(\mathbf{p}, \mathbf{x}) &= \sum_{i=1}^n H_{p_i}(\mathbf{p}, \mathbf{x}) \dot{p}^i + H_{x_i}(\mathbf{p}, \mathbf{x}) \dot{x}^i \\ &= \sum_{i=1}^n H_{p_i}(\mathbf{p}, \mathbf{x}) (-H_{x_i}(\mathbf{p}, \mathbf{x})) + H_{x_i}(\mathbf{p}, \mathbf{x}) H_{p_i}(\mathbf{p}, \mathbf{x}) = 0. \end{aligned}$$

□

See Arnold [Ar1, Chapter 9] for more on Hamilton’s ODE and Hamilton–Jacobi PDE in classical mechanics. We are employing here different notation than is customary in mechanics: our notation is better overall for PDE theory.

3.3.2. Legendre transform, Hopf–Lax formula.

Now let us try to find a connection between the Hamilton–Jacobi PDE and the calculus of variations problem (2)–(4). To simplify further, we also drop the x -dependence in the Hamiltonian, so that afterwards $H = H(p)$. We start by reexamining the definition of the Hamiltonian in §3.3.1.

a. Legendre transform. We hereafter suppose the Lagrangian $L : \mathbb{R}^n \rightarrow \mathbb{R}$ satisfies these conditions:

$$(11) \quad \text{the mapping } v \mapsto L(v) \text{ is convex}$$

and

$$(12) \quad \lim_{|v| \rightarrow \infty} \frac{L(v)}{|v|} = +\infty.$$

The convexity implies L is continuous.

DEFINITION. *The Legendre transform of L is*

$$(13) \quad L^*(p) := \sup_{q \in \mathbb{R}^n} \{p \cdot v - L(v)\} \quad (p \in \mathbb{R}^n).$$

This is also referred to as the *Fenchel transform*.

Motivation for Legendre transform. Why do we make this definition? For some insight let us note in view of (12) that the “sup” in (13) is really a “max”; that is, there exists some $v^* \in \mathbb{R}^n$ for which

$$L^*(p) = p \cdot v^* - L(v^*)$$

and the mapping $v \mapsto p \cdot v - L(v)$ has a maximum at $v = v^*$. But then $p = DL(v^*)$, provided L is differentiable at v^* . Hence the equation $p = DL(v)$ is solvable (although perhaps not uniquely) for v in terms of p , $v^* = \mathbf{v}(p)$. Therefore

$$L^*(p) = p \cdot \mathbf{v}(p) - L(\mathbf{v}(p)).$$

However, this is almost exactly the definition of the Hamiltonian H associated with L in §3.3.1 (where, recall, we are now assuming the variable x does not appear). We consequently henceforth write

$$(14) \quad H = L^*.$$

Thus (13) tells us how to obtain the Hamiltonian H from the Lagrangian L .

Now we ask the converse question: given H , how do we compute L ?

THEOREM 3 (Convex duality of Hamiltonian and Lagrangian). *Assume L satisfies (11), (12) and define H by (13), (14).*

(i) *Then*

the mapping $p \mapsto H(p)$ is convex

and

$$\lim_{|p| \rightarrow \infty} \frac{H(p)}{|p|} = +\infty.$$

(ii) *Furthermore*

$$(15) \quad L = H^*.$$

Thus H is the Legendre transform of L , and vice versa:

$$L = H^*, \quad H = L^*.$$

We say H and L are *dual* convex functions. The identity (15) implies that the three statements

$$(16) \quad \begin{cases} p \cdot v = L(v) + H(p) \\ p = DL(v) \\ v = DH(p) \end{cases}$$

are equivalent provided H is differentiable at p and L is differentiable at v : see Problem 11.

Proof. 1. For each fixed v , the function $p \mapsto p \cdot v - L(v)$ is linear; and consequently the mapping

$$p \mapsto H(p) = L^*(p) = \sup_{v \in \mathbb{R}^n} \{p \cdot v - L(v)\}$$

is convex. Indeed, if $0 \leq \tau \leq 1$, $p, \hat{p} \in \mathbb{R}^n$, we have

$$\begin{aligned} H(\tau p + (1 - \tau)\hat{p}) &= \sup_v \{(\tau p + (1 - \tau)\hat{p}) \cdot v - L(v)\} \\ &\leq \tau \sup_v \{p \cdot v - L(v)\} \\ &\quad + (1 - \tau) \sup_v \{\hat{p} \cdot v - L(v)\} \\ &= \tau H(p) + (1 - \tau)H(\hat{p}). \end{aligned}$$

2. Fix any $\lambda > 0$, $p \neq 0$. Then

$$\begin{aligned} H(p) &= \sup_{v \in \mathbb{R}^n} \{p \cdot v - L(v)\} \\ &\geq \lambda |p| - L\left(\lambda \frac{p}{|p|}\right) \quad (v = \lambda \frac{p}{|p|}) \\ &\geq \lambda |p| - \max_{B(0, \lambda)} L. \end{aligned}$$

Thus $\liminf_{|p| \rightarrow \infty} \frac{H(p)}{|p|} \geq \lambda$ for all $\lambda > 0$.

3. In view of (14)

$$H(p) + L(v) \geq p \cdot v$$

for all $p, v \in \mathbb{R}^n$, and consequently

$$L(v) \geq \sup_{p \in \mathbb{R}^n} \{p \cdot v - H(p)\} = H^*(v).$$

On the other hand

$$\begin{aligned} H^*(v) &= \sup_{p \in \mathbb{R}^n} \{p \cdot v - \sup_{r \in \mathbb{R}^n} \{p \cdot r - L(r)\}\} \\ &= \sup_{p \in \mathbb{R}^n} \inf_{r \in \mathbb{R}^n} \{p \cdot (v - r) + L(r)\}. \end{aligned}$$

Now since $v \mapsto L(v)$ is convex, according to §B.1 there exists $s \in \mathbb{R}^n$ such that

$$L(r) \geq L(v) + s \cdot (r - v) \quad (r \in \mathbb{R}^n).$$

(If L is differentiable at q , take $s = DL(v)$.) Putting $p = s$ above, we compute

$$H^*(v) \geq \inf_{r \in \mathbb{R}^n} \{s \cdot (v - r) + L(r)\} = L(v). \quad \square$$

b. Hopf–Lax formula. Let us now return to the initial-value problem (1) for the Hamilton–Jacobi equation and conclude from (64) in §3.2.5 that the corresponding characteristic equations are

$$\begin{cases} \dot{\mathbf{p}} = 0 \\ \dot{z} = DH(\mathbf{p}) \cdot \mathbf{p} - H(\mathbf{p}) \\ \dot{\mathbf{x}} = DH(\mathbf{p}). \end{cases}$$

The first and third of these are Hamilton’s ODE, which we in §3.3.1 derived from a minimization problem for associated Lagrangian $L = H^*$. Remembering (16), we can therefore understand the second of the characteristic equations as asserting

$$\dot{z} = DH(\mathbf{p}) \cdot \mathbf{p} - H(\mathbf{p}) = L(\dot{\mathbf{x}}).$$

But at least for such short times that (1) has a smooth solution u , we have $z(t) = u(\mathbf{x}(t), t)$ and consequently

$$u(x, t) = \int_0^t L(\dot{\mathbf{x}}(s)) ds + g(\mathbf{x}(0)).$$

Our intention is to modify this expression, to make sense even for large times $t > 0$ when (1) does not have a smooth solution. The variational principle for the action discussed in §3.3.1 provides the clue. Given $x \in \mathbb{R}^n$ and $t > 0$, we therefore propose to *minimize* among curves $\mathbf{w}(\cdot)$ satisfying $\mathbf{w}(t) = x$ the expression

$$\int_0^t L(\dot{\mathbf{w}}(s)) ds + g(\mathbf{w}(0)),$$

which is the action augmented with the value of the initial data. We accordingly now *define*

$$(17) \quad u(x, t) := \inf \left\{ \int_0^t L(\dot{\mathbf{w}}(s)) ds + g(\mathbf{w}(0)) \mid \mathbf{w}(t) = x \right\},$$

the infimum taken over all C^1 functions $\mathbf{w}(\cdot)$. (Better justification for this guess will be provided much later, in Chapter 10.)

We must investigate the sense in which the function u given by (17) actually solves the initial-value problem for the Hamilton–Jacobi PDE:

$$(18) \quad \begin{cases} u_t + H(Du) = 0 & \text{in } \mathbb{R}^n \times (0, \infty) \\ u = g & \text{on } \mathbb{R}^n \times \{t = 0\}. \end{cases}$$

Recall we are assuming H is smooth,

$$(19) \quad \begin{cases} H \text{ is convex and} \\ \lim_{|p| \rightarrow \infty} \frac{H(p)}{|p|} = +\infty. \end{cases}$$

We henceforth suppose also

$$(20) \quad g : \mathbb{R}^n \rightarrow \mathbb{R} \quad \text{is Lipschitz continuous;}$$

this means $\text{Lip}(g) := \sup_{\substack{x, y \in \mathbb{R}^n \\ x \neq y}} \left\{ \frac{|g(x) - g(y)|}{|x - y|} \right\} < \infty$.

First we note that formula (17) can be simplified:

THEOREM 4 (Hopf–Lax formula). *If $x \in \mathbb{R}^n$ and $t > 0$, then the solution $u = u(x, t)$ of the minimization problem (17) is*

$$(21) \quad u(x, t) = \min_{y \in \mathbb{R}^n} \left\{ tL \left(\frac{x - y}{t} \right) + g(y) \right\}.$$

DEFINITION. *We call the expression on the right-hand side of (21) the Hopf–Lax formula.*

Proof. 1. Fix any $y \in \mathbb{R}^n$ and define $\mathbf{w}(s) := y + \frac{s}{t}(x - y)$ ($0 \leq s \leq t$). Then the definition (17) of u implies

$$u(x, t) \leq \int_0^t L(\dot{\mathbf{w}}(s)) ds + g(y) = tL\left(\frac{x - y}{t}\right) + g(y),$$

and so

$$u(x, t) \leq \inf_{y \in \mathbb{R}^n} \left\{ tL\left(\frac{x - y}{t}\right) + g(y) \right\}.$$

2. On the other hand, if $\mathbf{w}(\cdot)$ is any C^1 function satisfying $\mathbf{w}(t) = x$, we have

$$L\left(\frac{1}{t} \int_0^t \dot{\mathbf{w}}(s) ds\right) \leq \frac{1}{t} \int_0^t L(\dot{\mathbf{w}}(s)) ds$$

by Jensen's inequality (§B.1). Thus if we write $y = \mathbf{w}(0)$, we find

$$tL\left(\frac{x - y}{t}\right) + g(y) \leq \int_0^t L(\dot{\mathbf{w}}(s)) ds + g(y);$$

and consequently

$$\inf_{y \in \mathbb{R}^n} \left\{ tL\left(\frac{x - y}{t}\right) + g(y) \right\} \leq u(x, t).$$

3. We have so far shown

$$u(x, t) = \inf_{y \in \mathbb{R}^n} \left\{ tL\left(\frac{x - y}{t}\right) + g(y) \right\},$$

and leave it as an exercise to prove that the infimum above is really a minimum. \square

We now commence a study of various properties of the function u defined by the Hopf–Lax formula (21). Our ultimate goal is showing this formula provides a reasonable weak solution of the initial-value problem (18) for the Hamilton–Jacobi equation.

First, we record some preliminary observations.

LEMMA 1 (A functional identity). *For each $x \in \mathbb{R}^n$ and $0 \leq s < t$, we have*

$$(22) \quad u(x, t) = \min_{y \in \mathbb{R}^n} \left\{ (t - s)L\left(\frac{x - y}{t - s}\right) + u(y, s) \right\}.$$

In other words, to compute $u(\cdot, t)$, we can calculate u at time s and then use $u(\cdot, s)$ as the initial condition on the remaining time interval $[s, t]$.