

in $U = \mathbb{R}^n \times (0, \infty)$, subject to the initial condition

$$(57) \quad u = g \quad \text{on } \Gamma = \mathbb{R}^n \times \{t = 0\}.$$

Here $\mathbf{F} : \mathbb{R} \rightarrow \mathbb{R}^n$, $\mathbf{F} = (F^1, \dots, F^n)$, and, as usual, we have set $t = x_{n+1}$. Also, “div” denotes the divergence with respect to the spatial variables (x_1, \dots, x_n) , and $Du = D_x u = (u_{x_1}, \dots, u_{x_n})$.

Since the direction $t = x_{n+1}$ plays a special role, we appropriately modify our notation. Writing now $q = (p, p_{n+1})$ and $y = (x, t)$, we have

$$G(q, z, y) = p_{n+1} + \mathbf{F}'(z) \cdot p,$$

and consequently

$$D_q G = (\mathbf{F}'(z), 1), \quad D_y G = 0, \quad D_z G = \mathbf{F}''(z) \cdot p.$$

Clearly the noncharacteristic condition (35) is satisfied at each point $y^0 = (x^0, 0) \in \Gamma$. Furthermore equation (21)(a) becomes

$$(58) \quad \begin{cases} \dot{x}^i(s) = F^{i'}(z(s)) & (i = 1, \dots, n) \\ \dot{x}^{n+1}(s) = 1. \end{cases}$$

Hence $x^{n+1}(s) = s$, in agreement with our having written $x_{n+1} = t$ above. In other words, we can identify the parameter s with the time t .

Equation (21)(b) reads $\dot{z}(s) = 0$. Consequently

$$(59) \quad z(s) = z^0 = g(x^0);$$

and (58) implies

$$(60) \quad \mathbf{x}(s) = \mathbf{F}'(g(x^0))s + x^0.$$

Thus the projected characteristic $\mathbf{y}(s) = (\mathbf{x}(s), s) = (\mathbf{F}'(g(x^0))s + x^0, s)$ ($s \geq 0$) is a straight line, along which u is constant.

Crossing characteristics. But suppose now we apply the same reasoning to a different initial point $z^0 \in \Gamma$, where $g(x^0) \neq g(z^0)$. *The projected characteristics may possibly then intersect at some time $t > 0$.* Since Theorem 1 tells us $u \equiv g(x^0)$ on the projected characteristic through x^0 and $u \equiv g(z^0)$ on the projected characteristic through z^0 , an apparent contradiction arises. The resolution is that *the initial-value problem (56), (57) does not in general have a smooth solution, existing for all times $t > 0$.* \square

We will discuss in §3.4 the interesting possibility of extending the local solution (guaranteed to exist for short times by Theorem 2) to all times $t > 0$, as a kind of “weak” or “generalized” solution.

Remark. Let us also note we can eliminate s from equations (59), (60) to obtain an implicit formula for u . Indeed given $x \in \mathbb{R}^n$ and $t > 0$, we see that since $s = t$,

$$\begin{aligned} u(\mathbf{x}(t), t) &= z(t) = g(\mathbf{x}(t) - t\mathbf{F}'(z^0)) \\ &= g(\mathbf{x}(t) - t\mathbf{F}'(u(\mathbf{x}(t), t))). \end{aligned}$$

Hence

$$(61) \quad u = g(x - t\mathbf{F}'(u)).$$

This implicit formula for u as a function of x and t is a nonlinear analogue of equation (3) in §2.1. It is easy to check (61) does indeed give a solution, provided

$$1 + tDg(x - t\mathbf{F}'(u)) \cdot \mathbf{F}''(u) \neq 0.$$

In particular if $n = 1$, we require

$$1 + tg'(x - tF'(u))F''(u) \neq 0.$$

Note that if $F'' > 0$, but $g' < 0$, then this will definitely be false at some time $t > 0$. This failure of the implicit formula (61) reflects also the failure of the characteristic method. \square

c. \mathbf{F} fully nonlinear.

The form of the full characteristic equations can be quite complicated for fully nonlinear first-order PDE, but sometimes a remarkable mathematical structure emerges.

Example 6 (Characteristics for the Hamilton–Jacobi equation). We look now at the general Hamilton–Jacobi PDE

$$(62) \quad G(Du, u_t, u, x, t) = u_t + H(Du, x) = 0,$$

where $Du = D_x u = (u_{x_1}, \dots, u_{x_n})$. Then writing $q = (p, p_{n+1})$, $y = (x, t)$, we have

$$G(q, z, y) = p_{n+1} + H(p, x);$$

and so

$$D_q G = (D_p H(p, x), 1), \quad D_y G = (D_x H(p, x), 0), \quad D_z G = 0.$$

Thus equation (11)(c) becomes

$$(63) \quad \begin{cases} \dot{x}^i(s) = \frac{\partial H}{\partial p_i}(\mathbf{p}(s), \mathbf{x}(s)) & (i = 1, \dots, n) \\ \dot{x}^{n+1}(s) = 1. \end{cases}$$

In particular we can identify the parameter s with the time t .

Equation (11)(a) for the case at hand reads

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

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

$$L = L(q, x) = L(q_1, \dots, q_n, x_1, \dots, x_n) \quad (q, x \in \mathbb{R}^n)$$

and

$$\begin{cases} D_q L = (L_{q_1} \cdots L_{q_n}) \\ D_x L = (L_{x_1} \cdots L_{x_n}). \end{cases}$$

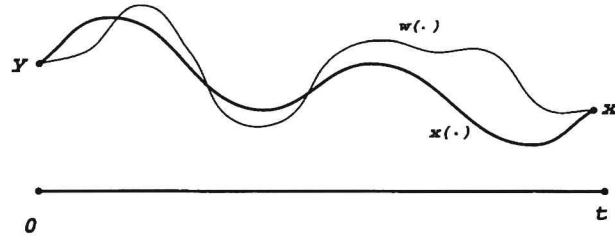
Thus in the formula (2) below “ q ” is the name of the variable for which we substitute $\dot{\mathbf{w}}(s)$, and “ x ” is the variable for which we substitute $\mathbf{w}(s)$. \square

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

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

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 $w(\cdot)$ 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 then to find a curve $x(\cdot) \in \mathcal{A}$ satisfying

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

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

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

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

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

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

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

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

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

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

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

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

Thus the real-valued function

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

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

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

provided $i'(0)$ exists.

2. We explicitly compute this derivative. Observe

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

and so

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

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

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

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

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

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

$$-\frac{d}{ds} (L_{q_i}(\dot{x}, x)) + L_{x_i}(\dot{x}, x) = 0$$

for $0 \leq s \leq t$, $i = 1, \dots, n$. □

Remark. We have just demonstrated that any minimizer $x(\cdot) \in \mathcal{A}$ of $I[\cdot]$ solves the Euler–Lagrange system of ODE. It is of course possible that a curve $x(\cdot) \in \mathcal{A}$ may solve the Euler–Lagrange equations without necessarily being a minimizer: in this case we say $x(\cdot)$ 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 $L(q, x) = \frac{1}{2}m|q|^2 - \phi(x)$, 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 ODE.

We now convert 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

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

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

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

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

DEFINITION. The Hamiltonian H associated with the Lagrangian L is

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

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

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

Remark. 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)). \square

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

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

$$\begin{aligned} \frac{\partial H}{\partial x_i}(p, x) &= \sum_{k=1}^n p_k \frac{\partial q^k}{\partial x_i}(p, x) - \frac{\partial L}{\partial q_k}(q, x) \frac{\partial q^k}{\partial x_i}(p, x) - \frac{\partial L}{\partial x_i}(q, x) \\ &= -\frac{\partial L}{\partial x_i}(q, x) \quad \text{by (9),} \end{aligned}$$

and

$$\begin{aligned} \frac{\partial H}{\partial p_i}(p, x) &= q^i(p, x) + \sum_{k=1}^n p_k \frac{\partial q^k}{\partial p_i}(p, x) - \frac{\partial L}{\partial q_k}(q, x) \frac{\partial q^k}{\partial p_i}(p, x) \\ &= q^i(p, x), \quad \text{again by (9).} \end{aligned}$$

Thus

$$\frac{\partial H}{\partial p_i}(\mathbf{p}(s), \mathbf{x}(s)) = q^i(\mathbf{p}(s), \mathbf{x}(s)) = \dot{x}^i(s);$$

and likewise

$$\begin{aligned} \frac{\partial H}{\partial x_i}(\mathbf{p}(s), \mathbf{x}(s)) &= -\frac{\partial L}{\partial x_i}(\mathbf{q}(\mathbf{p}(s), \mathbf{x}(s)), \mathbf{x}(s)) = -\frac{\partial L}{\partial x_i}(\dot{\mathbf{x}}(s), \mathbf{x}(s)) \\ &= -\frac{d}{ds} \left(\frac{\partial L}{\partial q_i}(\dot{\mathbf{x}}(s), \mathbf{x}(s)) \right) \quad \text{according to (4)} \\ &= -\dot{p}^i(s). \end{aligned}$$

Finally, observe

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

\square

Remark. See Arnold [AR, 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. \square

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) the mapping $q \mapsto L(q)$ is convex

and

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

The convexity implies L is continuous.

DEFINITION. The Legendre transform of L is

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

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 $q^* \in \mathbb{R}^n$ for which

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

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

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

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)
$$H = L^*.$$

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

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)
$$L = H^*.$$

Remark. 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. \square

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

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

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

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

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

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

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

3. In view of (14)

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

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

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

On the other hand

$$(16) \quad \begin{aligned} H^*(q) &= \sup_{p \in \mathbb{R}^n} \{p \cdot q - \sup_{r \in \mathbb{R}^n} \{p \cdot r - L(r)\}\} \\ &= \sup_{p \in \mathbb{R}^n} \inf_{r \in \mathbb{R}^n} \{p \cdot (q - r) + L(r)\}. \end{aligned}$$

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

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

(If L is differentiable at q , take $s = DL(q)$.) Taking $p = s$ in (16), we compute

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

□

b. Hopf–Lax formula.

Let us now return to the initial-value problem (1). Recall that the calculus of variations problem with Lagrangian L , discussed in §3.3.1, led to Hamilton’s ODE for the associated Hamiltonian H . Since these ODE are also the characteristic equations of the Hamilton–Jacobi PDE, we conjecture there is probably a direct connection between this PDE and the calculus of variations.

So if $x \in \mathbb{R}^n$ and $t > 0$ are given, we should presumably try to minimize the action

$$\int_0^t L(\dot{w}(s)) \, ds$$

over functions $w : [0, t] \rightarrow \mathbb{R}^n$ satisfying $w(t) = x$. But what should we take for $w(0)$? As we must somehow take into account the initial condition for our PDE, let us try modifying the action to include the function g evaluated at $w(0)$:

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

Next let us construct a candidate for a solution to the initial-value problem (1), in terms of a variational principle entailing this modified action. We accordingly set

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

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

We propose now to investigate the sense in which u so defined 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

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

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 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 $w(s) := y + \frac{s}{t}(x - y)$ ($0 \leq s \leq t$). Then the definition (17) of u implies

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

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 $w(\cdot)$ is any C^1 function satisfying $w(t) = x$, we have

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

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

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

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 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 $[s, t]$.

Proof. 1. Fix $y \in \mathbb{R}^n$, $0 < s < t$ and choose $z \in \mathbb{R}^n$ so that

$$(23) \quad u(y, s) = sL \left(\frac{y-z}{s} \right) + g(z).$$

Now since L is convex and $\frac{x-z}{t} = (1 - \frac{s}{t}) \frac{x-y}{t-s} + \frac{s}{t} \frac{y-z}{s}$, we have

$$L \left(\frac{x-z}{t} \right) \leq \left(1 - \frac{s}{t}\right) L \left(\frac{x-y}{t-s} \right) + \frac{s}{t} L \left(\frac{y-z}{s} \right).$$

Thus

$$\begin{aligned} u(x, t) &\leq tL \left(\frac{x-z}{t} \right) + g(z) \leq (t-s)L \left(\frac{x-y}{t-s} \right) + sL \left(\frac{y-z}{s} \right) + g(z) \\ &= (t-s)L \left(\frac{x-y}{t-s} \right) + u(y, s), \end{aligned}$$

by (23). This inequality is true for each $y \in \mathbb{R}^n$. Therefore, since $y \mapsto u(y, s)$ is continuous (according to Lemma 2 below), we have

$$(24) \quad u(x, t) \leq \min_{y \in \mathbb{R}^n} \left\{ (t-s)L \left(\frac{x-y}{t-s} \right) + u(y, s) \right\}.$$

2. Now choose w such that

$$(25) \quad u(x, t) = tL \left(\frac{x-w}{t} \right) + g(w),$$

and set $y := \frac{s}{t}x + (1 - \frac{s}{t})w$. Then $\frac{x-y}{t-s} = \frac{x-w}{t} = \frac{y-w}{s}$. Consequently

$$\begin{aligned} (t-s)L \left(\frac{x-y}{t-s} \right) + u(y, s) &\leq (t-s)L \left(\frac{x-w}{t} \right) + sL \left(\frac{y-w}{s} \right) + g(w) \\ &= tL \left(\frac{x-w}{t} \right) + g(w) = u(x, t), \end{aligned}$$

by (25). Hence

$$(26) \quad \min_{y \in \mathbb{R}^n} \left\{ (t-s)L \left(\frac{x-y}{t-s} \right) + u(y, s) \right\} \leq u(x, t).$$

\square

LEMMA 2 (Lipschitz continuity). *The function u is Lipschitz continuous in $\mathbb{R}^n \times [0, \infty)$, and*

$$u = g \quad \text{on } \mathbb{R}^n \times \{t = 0\}.$$

Proof. 1. Fix $t > 0$, $x, \hat{x} \in \mathbb{R}^n$. Choose $y \in \mathbb{R}^n$ such that

$$(27) \quad tL\left(\frac{x-y}{t}\right) + g(y) = u(x, t).$$

Then

$$\begin{aligned} u(\hat{x}, t) - u(x, t) &= \inf_z \left\{ tL\left(\frac{\hat{x}-z}{t}\right) + g(z) \right\} - tL\left(\frac{x-y}{t}\right) - g(y) \\ &\leq g(\hat{x} - x + y) - g(y) \leq \text{Lip}(g)|\hat{x} - x|. \end{aligned}$$

Hence

$$u(\hat{x}, t) - u(x, t) \leq \text{Lip}(g)|\hat{x} - x|;$$

and, interchanging the roles of \hat{x} and x , we find

$$(28) \quad |u(x, t) - u(\hat{x}, t)| \leq \text{Lip}(g)|x - \hat{x}|.$$

2. Now select $x \in \mathbb{R}^n$, $t > 0$. Choosing $y = x$ in (21), we discover

$$(29) \quad u(x, t) \leq tL(0) + g(x).$$

Furthermore,

$$\begin{aligned} u(x, t) &= \min_{y \in \mathbb{R}^n} \left\{ tL\left(\frac{x-y}{t}\right) + g(y) \right\} \\ &\geq g(x) + \min_{y \in \mathbb{R}^n} \left\{ -\text{Lip}(g)|x-y| + tL\left(\frac{x-y}{t}\right) \right\} \\ &= g(x) - t \max_{z \in \mathbb{R}^n} \{ \text{Lip}(g)|z| - L(z) \} \quad (z = \frac{x-y}{t}) \\ &= g(x) - t \max_{w \in B(0, \text{Lip}(g))} \max_{z \in \mathbb{R}^n} \{ w \cdot z - L(z) \} \\ &= g(x) - t \max_{B(0, \text{Lip}(g))} H. \end{aligned}$$

This inequality and (29) imply

$$|u(x, t) - g(x)| \leq Ct$$

for

$$(30) \quad C := \max(|L(0)|, \max_{B(0, \text{Lip}(g))} |H|).$$

3. Finally select $x \in \mathbb{R}^n$, $0 < \hat{t} < t$. Then $\text{Lip}(u(\cdot, t)) \leq \text{Lip}(g)$ by (28) above. Consequently Lemma 1 and calculations like those employed in step 2 above imply

$$|u(x, t) - u(x, \hat{t})| \leq C|t - \hat{t}|$$

for the constant C defined by (30). □

Now Rademacher's Theorem (which we will prove later, in §5.8.3) asserts that a Lipschitz function is differentiable almost everywhere. Consequently in view of Lemma 2 our function u defined by the Hopf-Lax formula (21) is differentiable for a.e. $(x, t) \in \mathbb{R}^n \times (0, \infty)$. The next theorem asserts u in fact solves the Hamilton-Jacobi PDE wherever u is differentiable.

THEOREM 5 (Solving the Hamilton-Jacobi equation). *Suppose $x \in \mathbb{R}^n$, $t > 0$, and u defined by the Hopf-Lax formula (21) is differentiable at a point $(x, t) \in \mathbb{R}^n \times (0, \infty)$. Then*

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

Proof. 1. Fix $q \in \mathbb{R}^n$, $h > 0$. Owing to Lemma 1,

$$\begin{aligned} u(x + hq, t + h) &= \min_{y \in \mathbb{R}^n} \left\{ hL\left(\frac{x + hq - y}{h}\right) + u(y, t) \right\} \\ &\leq hL(q) + u(x, t). \end{aligned}$$

Hence

$$\frac{u(x + hq, t + h) - u(x, t)}{h} \leq L(q).$$

Let $h \rightarrow 0^+$, to compute

$$q \cdot Du(x, t) + u_t(x, t) \leq L(q).$$

This inequality is valid for all $q \in \mathbb{R}^n$, and so

$$(31) \quad u_t(x, t) + H(Du(x, t)) = u_t(x, t) + \max_{q \in \mathbb{R}^n} \{ q \cdot Du(x, t) - L(q) \} \leq 0.$$

The first equality holds since $H = L^*$.

2. Now choose z such that $u(x, t) = tL\left(\frac{x-z}{t}\right) + g(z)$. Fix $h > 0$ and set $s = t - h$, $y = \frac{s}{t}x + (1 - \frac{s}{t})z$. Then $\frac{x-z}{t} = \frac{y-z}{s}$, and thus

$$\begin{aligned} u(x, t) - u(y, s) &\geq tL\left(\frac{x-z}{t}\right) + g(z) - \left[sL\left(\frac{y-z}{s}\right) + g(z) \right] \\ &= (t-s)L\left(\frac{x-z}{t}\right). \end{aligned}$$

That is,

$$\frac{u(x, t) - u\left(\left(1 - \frac{h}{t}\right)x + \frac{h}{t}z, t - h\right)}{h} \geq L\left(\frac{x-z}{t}\right).$$