

# Chapter 9

## Viscosity solutions for the dynamic programming equation

### 9.1 Viscosity solutions for parabolic PDEs

Let  $F$  be an operator from  $[0, T] \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R} \times \mathbb{R}^n \times \mathbb{S}^n$  into  $\mathbb{R}$  where  $\mathbb{S}^n$  stands for the set of  $n$ -dimensional symmetric matrices. In this chapter, we will be mostly interested by the case

$$F(t, x, u, q, p, M) = -\sup_{a \in A} \{q + \mathcal{L}^a[t, x, u, p, M] + f(x, a)\}$$

where

$$\mathcal{L}^a[t, x, u, p, M] = b(x, a) \cdot p + \frac{1}{2} \text{Tr}(\sigma \sigma^\top(x, a) M)$$

for  $(t, x, u, q, p, M) \in [0, T] \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R} \times \mathbb{R}^n \times \mathbb{S}^n$  where the coefficients  $b, \sigma$  and  $f$  and the set  $A$  are those of the previous chapters and satisfy the related assumptions. This case is called the HJB case.

In the sequel, we assume that  $F$  is an elliptic operator, that is,  $F$  is nonincreasing in the variable  $M$ . We are interested in the PDE

$$F(t, x, u(t, x), \partial_t u(t, x), \nabla u(t, x), \nabla^2 u(t, x)) = 0 \quad (9.1.1)$$

for  $(t, x) \in [0, T] \times \mathbb{R}^n$ , together with the terminal condition

$$u(T, x) = g(x), \quad x \in \mathbb{R}^n. \quad (9.1.2)$$

Our objective is to give a notion of weak solution to such a PDE.

**Definition 9.1.2** (i) A continuous function  $u : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$  is called a viscosity subsolution to (9.1.1) if for any function  $\varphi \in C^{1,2}([0, T] \times \mathbb{R}^n)$  and  $(t, x) \in [0, T] \times \mathbb{R}^n$  such that

$$0 = (u - \varphi)(t, x) = \max_{[0, T] \times \mathbb{R}^n} (u - \varphi)$$

we have

$$F(t, x, \varphi(t, x), \partial_t \varphi(t, x), \nabla \varphi(t, x), \nabla^2 \varphi(t, x)) \leq 0.$$

(ii) A continuous function  $u : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$  is called a viscosity supersolution to (9.1.1) if for any function  $\varphi \in C^{1,2}([0, T] \times \mathbb{R}^n)$  and  $(t, x) \in [0, T] \times \mathbb{R}^n$  such that

$$0 = (u - \varphi)(t, x) = \min_{[0, T] \times \mathbb{R}^n} (u - \varphi)$$

we have

$$F(t, x, \varphi(t, x), \partial_t \varphi(t, x), \nabla \varphi(t, x), \nabla^2 \varphi(t, x)) \geq 0.$$

(iii) A function  $u : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$  is called a viscosity solution to (9.1.1) if it is a viscosity subsolution to (9.1.1) and a viscosity supersolution to (9.1.1).

## 9.2 Viscosity properties of the value function

We consider the HJB case for PDE (9.1.1) and the value function  $v$  defined by (7.1.2).

**Theorem 9.2.7** The value function  $v$  is a viscosity solution to (9.1.1) and satisfies (9.1.2).

**Proof.** The fact that  $v$  satisfies (9.1.2) comes from its definition. We turn to the viscosity properties w.r.t. equation to (9.1.1).

**Step 1.** Subsolution property. Let  $(\hat{t}, \hat{x}) \in [0, T] \times \mathbb{R}^n$  and  $\varphi \in C^{1,2}([0, T] \times \mathbb{R}^n)$  be such that

$$0 = (v - \varphi)(\hat{t}, \hat{x}) = \max_{[0, T] \times \mathbb{R}^n} (v - \varphi). \quad (9.2.3)$$

We proceed by contradiction and assume that

$$-\sup_{a \in A} \{(\partial_t + \mathcal{L}^a)\varphi(\hat{t}, \hat{x}) + f(\hat{x}, a)\} > 0.$$

By continuity of the coefficients, we can find  $r \in (T - \hat{t}, T)$  such that

$$-\sup_{a \in A} \{(\partial_t + \mathcal{L}^a)\varphi(t, x) + f(x, a)\} > 0 \quad (9.2.4)$$

for  $(t, x) \in B := B(\hat{t}, r) \times B(\hat{x}, r)$ . Without loss of generality, we can assume that the minimum in (9.2.5), is strict and that

$$\max_{\partial B} (v - \varphi) \leq -\eta.$$

for some  $\eta > 0$ . Let  $\nu$  be the first exit time of  $(s, X_s^{t,x,\hat{\alpha}})_{s \geq t}$  from  $B$ . From the DPP, there exists a control  $\hat{\alpha} \in \mathcal{A}_{\hat{t}}$  such that

$$v(\hat{t}, \hat{x}) \leq \mathbb{E} \left[ v(\theta, X_\theta^{t,x,\hat{\alpha}}) + \int_t^\theta f(X_s^{t,x,\hat{\alpha}}, \hat{\alpha}_s) ds \right] + \frac{\eta}{2}.$$

From the previous inequalities, we get

$$\varphi(\hat{t}, \hat{x}) \leq \mathbb{E} \left[ \varphi(\theta, X_\theta^{t,x,\hat{\alpha}}) - \eta + \int_t^\theta f(X_s^{t,x,\hat{\alpha}}, \hat{\alpha}_s) ds \right] + \frac{\eta}{2}.$$

Applying Itô's Lemma to  $\varphi$ , we get

$$0 \leq -\frac{\eta}{2} + \mathbb{E} \left[ \int_t^\theta ((\partial_t + \mathcal{L}^{\hat{\alpha}_s})\varphi(s, X_s^{t,x,\hat{\alpha}}) + f(X_s^{t,x,\hat{\alpha}}, \hat{\alpha}_s)) ds \right].$$

Using (9.2.4) we get

$$0 \leq -\frac{\eta}{2}.$$

**Step 2.** Supersolution property. Let  $(t, x) \in [0, T] \times \mathbb{R}^n$  and  $\varphi \in C^{1,2}([0, T] \times \mathbb{R}^n)$  be such that

$$0 = (v - \varphi)(t, x) = \min_{[0, T] \times \mathbb{R}^n} (v - \varphi). \quad (9.2.5)$$

Fix  $a \in A$  and denote by  $\hat{\alpha}$  the control constant equal to  $a$ . From the dynamic programming principle we have

$$v(t, x) \geq \mathbb{E}\left[v(\theta, X_\theta^{t,x,\hat{\alpha}}) + \int_t^{t+h} f(X_s^{t,x,\hat{\alpha}}, a) ds\right]$$

for  $h \in (0, T - t)$ . Using (9.2.5), we get

$$\varphi(t, x) \geq \mathbb{E}\left[\varphi(t+h, X_{t+h}^{t,x,\hat{\alpha}}) + \int_t^{t+h} f(X_s^{t,x,\hat{\alpha}}, a) ds\right]$$

Applying Itô's formula and diving by  $h > 0$ , we get

$$0 \geq \frac{1}{h} \mathbb{E}\left[\int_t^{t+h} [(\partial_t + \mathcal{L}^a)\varphi(X_s^{t,x,\hat{\alpha}}, a) + f(X_s^{t,x,\hat{\alpha}}, a)] ds\right]$$

Sending  $h$  to  $0+$ , we get by the mean value theorem

$$(\partial_t + \mathcal{L}^a)\varphi(t, x) + f(x, a) \leq 0.$$

Since this holds for all  $a \in A$ , we get the supersolution property.  $\square$

### 9.3 Comparison and uniqueness

We first have the following result which allow to compare a supersolution and a subsolution as soon as we can compare their values at the terminal time  $T$ .

**Theorem 9.3.8** *Let  $U$  (resp.  $V$ )  $\in C^{1,2}([0, T] \times \mathbb{R}^n) \cap C^0([0, T] \times \mathbb{R}^n)$  be a subsolution (resp. supersolution) with polynomial growth to (9.1.1) in the HJB case. If  $U(T, .) \leq V(T, .)$  on  $\mathbb{R}^n$ , then  $U \leq V$  on  $[0, T] \times \mathbb{R}^n$ .*

**Proof.** We present the proof in the regular case. The proof for viscosity solutions uses an additional result called Ishii's lemma which gives a similar condition to second order optimality condition. We proceed in 3 steps.

**Step 1.** Let  $\tilde{U}(t, x) = e^{\lambda t} U(t, x)$  and  $\tilde{V}(t, x) = e^{\lambda t} V(t, x)$ . Then a straightforward calculation shows that  $\tilde{U}$  (resp.  $\tilde{V}$ ) is a subsolution (resp. supersolution) to

$$\tilde{F}(t, x, w, \partial_t w(t, x), \nabla w(t, x), \nabla^2 w(t, x)) = 0 \quad (9.3.6)$$

with

$$\begin{aligned} F(t, x, u, q, p, M) &= -\sup_{a \in A} \{q + \tilde{\mathcal{L}}^a[t, x, u, p, M] + \tilde{f}(t, x, a)\}, \\ \tilde{\mathcal{L}}^a[t, x, u, p, M] &= b(x, a).p + \frac{1}{2}\text{Tr}(\sigma\sigma^\top(x, a)M) - \lambda u, \\ \tilde{f}(t, x, a) &= e^{\lambda t} f(x, a) \end{aligned}$$

for  $(t, x, u, q, p, M) \in [0, T] \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R} \times \mathbb{R}^n \times \mathbb{S}^n$  and  $a \in A$ .

**Step 2.** From the polynomial growth of  $U$  and  $V$ , we may choose an integer  $p \geq 1$  such that

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^n} \frac{|\tilde{U}(t, x)| + |\tilde{V}(t, x)|}{1 + |x|^p} < +\infty,$$

and we consider the function  $\phi(t, x) = e^{-\lambda t}(1 + |x|^{2p}) =: e^{-\lambda t}\psi(x)$ . From the linear growth condition on  $b$  and  $\sigma$ , a straightforward computation shows that there exists some constant  $c > 0$  s.t.

$$\begin{aligned} &-\partial_t \phi(t, x) + \lambda \phi(t, x) - \sup_{a \in A} b(x, a).\nabla \phi(t, x) + \frac{1}{2}\text{Tr}(\sigma\sigma^\top(x, a)\nabla^2 \phi(t, x)) \\ &\geq e^{-\lambda t}\psi(x)(\lambda - c). \end{aligned}$$

By taking  $\lambda \geq c$ , the function  $\tilde{V}_\varepsilon := \tilde{V} + \varepsilon\phi$  is a supersolution to (9.1.1) for any  $\varepsilon > 0$ . Moreover, from the growth conditions on  $U, V$  and  $\phi$ , we have

$$\lim_{|x| \rightarrow +\infty} \sup_{t \in [0, T]} (\tilde{U} - \tilde{V}_\varepsilon)(t, x) = -\infty. \quad (9.3.7)$$

for all  $\varepsilon > 0$ .

**Step 3.** We finally argue by contradiction to show that  $\tilde{U} - \tilde{V}_\varepsilon \leq 0$  on  $[0, T] \times \mathbb{R}^n$  for all  $\varepsilon > 0$ , which gives the required result by sending  $\varepsilon$  to 0. On the contrary, by continuity of  $\tilde{U} - \tilde{V}_\varepsilon$ , and from (9.3.7), there exists  $\varepsilon > 0$ ,  $(t, x) \in [0, T] \times \mathbb{R}^n$  such that

$$\sup_{[0, T] \times \mathbb{R}^n} (\tilde{U} - \tilde{V}_\varepsilon) = (\tilde{U} - \tilde{V}_\varepsilon)(t, x) > 0. \quad (9.3.8)$$

Since  $(\tilde{U} - \tilde{V}_\varepsilon)(T, .) \leq (\tilde{U} - \tilde{V})(T, .) \leq 0$  on  $\mathbb{R}^n$ , we have  $t < T$ . Therefore, the first and second order optimality conditions (9.3.8) give

$$\begin{aligned} \partial_t(\tilde{U} - \tilde{V}_\varepsilon)(t, x) &\leq 0, \\ \nabla(\tilde{U} - \tilde{V}_\varepsilon)(t, x) &= 0, \\ \nabla^2(\tilde{U} - \tilde{V}_\varepsilon)(t, x) &\leq 0. \end{aligned}$$

By writing that  $\tilde{U}$  (resp.  $\tilde{V}_\varepsilon$ ) is a subsolution (resp. supersolution) to (9.3.6), and recalling that  $\tilde{F}$  is nondecreasing in its last argument, we then deduce that

$$\begin{aligned} &\lambda(\tilde{U} - \tilde{V}_\varepsilon)(t, x) \\ &\leq \tilde{F}(t, x, \partial_t \tilde{U}(t, x), \nabla \tilde{U}(t, x), \nabla^2 \tilde{U}(t, x)) \\ &\quad - \tilde{F}(t, x, \partial_t \tilde{V}_\varepsilon(t, x), \nabla \tilde{V}_\varepsilon(t, x), \nabla^2 \tilde{V}_\varepsilon(t, x)) \\ &\leq 0 \end{aligned}$$

which contradicts (9.3.8).  $\square$

**Corollary 9.3.1** *The value function  $v$  is the unique viscosity solution to (9.1.1) in the HJB case satisfying (9.1.1) and having polynomial growth.*

**Proof.** We simply need to prove that  $v$  has polynomial growth. It actually follows from the polynomial growth assumption on  $f$  and  $g$  and Proposition (7.1.3).  $\square$