Journal of Probability

APPROXIMATING VALUE FUNCTIONS FOR CONTROLLED DEGENERATE DIFFUSION PROCESSES BY USING PIECE-WISE CONSTANT POLICIES

N.V. KRYLOV

ABSTRACT. It is shown that value functions for controlled degenerate diffusion processes can be approximated with error of order $h^{1/3}$ by using policies which are constant on intervals $[kh^2, (k+1)h^2)$.

1

1. Introduction

Piece-wise constant policies play an important role in the theory of controlled diffusion processes. For instance, in [4] they are used for proving Bellman's principle in its most general form. These policies are also important from the point of view of numerical computations since one knows how to approximate diffusion processes without control and on each interval of time where a policy is constant we are dealing with usual diffusion processes. In this connection an interesting question arises as to how good the gain for piece-wise constant policies approximates the value function.

In the framework of controlled diffusion processes this question was first addressed in article [6], in which it is shown that if the intervals, where policies are constant, have length h^2 and the controlled process can degenerate and has coefficients independent of time and space variable, then the rate of approximation is not less than $h^{1/3}$. In the general case of variable coefficients the rate is shown to be not less than $h^{1/6}$. The general situation in [6] allows coefficients to be discontinuous in time although we still need them to be Hölder continuous with respect to some integral norms. The main result of the present article, stated as Theorem 2.3 in Section 2, is that the rate is still $h^{1/3}$ if the coefficients are Hölder $1/2$ in time variable. In the proof of this result we use an idea from [7], which consists of "shaking" the coefficients and in order to be sure that after "shaking" we get the value functions which are close to initial ones, we assume usual Hölder continuity instead of Hölder continuity in integral norms.

Results of that kind allow one (see, for instance, [7]) to analyze the rate of convergence of finite-difference approximations for Bellman's equations. In Section 5 we apply the results of Section 2 to improve some of the results of [7]. However the best we can do for finite-difference approximations is still $h^{1/21}$ (with h being the mesh size) and this is why in Section 5 we discuss other methods of approximating the value function which can be used for numerical computations and are of order of accuracy $h^{1/3}$.

Apart from the above mentioned sections, there are also Sections 3 and 6 containing some auxiliary results and Section 4 containing the proofs of our main results.

Throughout the article $\mathbb{R}^d$ is a d-dimensional Euclidean space, A is a separable metric space, $T \in (0, \infty)$, $K \in [1, \infty)$, and $\delta_0 \in (0, 1]$ are some fixed constants. By N we denote various constants depending only on K, d, and d_1, where d_1 is introduced in the next section.

2. Main results

Let $(\Omega, \mathcal{F}, P)$ be a complete probability space, $\{\mathcal{F}_t; t \geq 0\}$ be an increasing filtration of σ–algebras $\mathcal{F}_t \subset \mathcal{F}$ which are complete with respect to $\mathcal{F}, P$. Assume that on $(\Omega, \mathcal{F}, P)$ a d_1–dimensional Wiener process w_t is defined for $t \geq 0$. We suppose that w_t is a Wiener process with respect to $\{\mathcal{F}_t\}$, or in other terms, that $\{w_t, \mathcal{F}_t\}$ is a Wiener process.

Definition 2.1. An A–valued process $\alpha_t = \alpha_t(\omega)$ defined for all $t \geq 0$ and $\omega \in \Omega$ is called *a policy* if it is $\mathcal{F} \otimes B([0, \infty))$–measurable with respect to (ω, t) and $\mathcal{F}_t$–measurable with respect to ω for each $t \geq 0$. The set of all policies is denoted by $\mathfrak{A}$. For any $h \in (0, 1]$ let $\mathfrak{A}_h$ be the subset of $\mathfrak{A}$ consisting of all processes α_t which are constant on intervals $[0, h^2)$, $[h^2, 2h^2)$, and so on.

Fix an integer $d \geq 1$ and suppose that on $A \times [0, \infty) \times \mathbb{R}^d$ we are given a $d \times d_1$ matrix–valued function $\sigma(\alpha, t, x)$, an $\mathbb{R}^d$–valued function $b(\alpha, t, x)$, and real-valued functions $c^\alpha(t, x) \geq 0$, $f^\alpha(t, x)$, and $g(x)$. We assume that these functions are Borel measurable.

Assumption 2.2. Let $u(\alpha, t, x)$ stand for $\sigma(\alpha, t, x)$, $b(\alpha, t, x)$, $c^\alpha(t, x)$, $f^\alpha(t, x)$, and $g(x)$.
  (i) The functions $u(\alpha, t, x)$ are continuous with respect to α.
  (ii) For any $\alpha \in A$, $t \in [0, \infty)$, and $x, y \in \mathbb{R}^d$

$$|u(\alpha, t, x)| \leq K, \quad |u(\alpha, t, x) - u(\alpha, t, y)| \leq K|x - y|.$$

  (iii) For any $\alpha \in A$, $s, t \in [0, \infty)$, and $x \in \mathbb{R}^d$ we have

$$|u(\alpha, t, x) - u(\alpha, s, x)| \leq K|t - s|^{\delta_0/2}.$$

By Itô's theorem for any $\alpha \in \mathfrak{A}$, $s \in [0, T]$, and $x \in \mathbb{R}^d$, there exists a unique solution $x_t = x_t^{\alpha, s, x}$, $t \in [0, \infty)$, of the following equation

$$x_t = x + \int_0^t \sigma(\alpha_r, s + r, x_r)\, dw_r + \int_0^t b(\alpha_r, s + r, x_r)\, dr. \qquad (2.1)$$

For $\alpha \in \mathfrak{A}$, $s \in [0, T]$, and $x \in \mathbb{R}^d$ define

$$\varphi_t^{\alpha, s, x} = \int_0^t c^{\alpha_r}(s + r, x_r^{\alpha, s, x})\, dr,$$

$$v^\alpha(s, x) = E_{s,x}^\alpha \Big[\int_0^{T-s} f^{\alpha_t}(s + t, x_t) e^{-\varphi_t}\, dt + g(x_{T-s}) e^{-\varphi_{T-s}} \Big],$$

$$v(s, x) = \sup_{\alpha \in \mathfrak{A}} v^\alpha(s, x), \quad v_h(s, x) = \sup_{\alpha \in \mathfrak{A}_h} v^\alpha(s, x),$$

where, as usual, the indices α, s, x accompanying the symbol of expectation mean that one has to put them inside the expectation at appropriate places.

Here is our first main result.

Theorem 2.3. *For any $s \in [0, T]$, $x \in \mathbb{R}^d$, and $h \in (0, 1]$, we have*
$$|v(s, x) - v_h(s, x)| \leq N e^{NT} h^{\delta_0^2/(2\delta_0+1)},$$
where the constant N depends only on K, d, and d_1.

Remark 2.4. The largest value of $\delta_0^2/(2\delta_0 + 1)$ for $\delta_0 \in (0, 1]$ is $1/3$ which is achieved if $\delta_0 = 1$.

One may try to use Theorem 2.3 for approximating v in real-world problems. At least theoretically v_h can be found in the following way. The dynamic programming principle says that, for $s + h^2 \leq T$,
$$v_h(s, x) = G_{s,s+h^2} v_h(s + h^2, \cdot)(x), \tag{2.2}$$
where
$$G_{s,t} z(x) = \sup_{\alpha \in A} G_{s,t}^\alpha z(x),$$
$$G_{s,t}^\alpha z(x) = E_{s,x}^\alpha \Big[\int_0^{t-s} f^{\alpha_r}(s + r, x_r) e^{-\varphi_r} \, dr + z(x_{t-s}) e^{-\varphi_{t-s}} \Big]$$
and the last expectation is evaluated for the constant policy $\alpha_t \equiv \alpha$. Therefore, if one knows how to compute $G_{s,s+h^2}^\alpha$, one can use (2.2) in order to find v_h from its boundary value $v_h(T, x) = g(x)$ by backward iteration.

However, finding $G_{s,s+h^2}^\alpha$ requires finding distributions of solutions of (2.1) even though only for constant α_t. Since $G_{s,s+h^2}^\alpha$ only depends on solutions of (2.1) on a small time interval, it is natural to think that one can "freeze" not only α in (2.1) but also $s+r$ and x_r. Then one can avoid solving equation (2.1) and deal with Gaussian random variables with known mean and variance. The following theorem is an example of what one can get along these lines.

Theorem 2.5. *For $\alpha \in A$, $s, t \geq 0$, and $x \in \mathbb{R}^d$ define*
$$y_t^{\alpha,s,x} = x + \sigma(\alpha, s, x) w_t + b(\alpha, s, x) t,$$
$$\bar{G}_{s,s+t}^\alpha z(x) = f^\alpha(s, x) t + e^{-c^\alpha(s,x)t} E z(y_t^{\alpha,s,x}),$$
$$\bar{G}_{s,s+t} z = \sup_{\alpha \in A} \bar{G}_{s,s+t}^\alpha z.$$
Define recursively
$$\begin{aligned}
\bar{v}_h(s, x) &= g(x) \quad \text{if} \quad s \in (T - h^2, T], \\
\bar{v}_h(s, x) &= \bar{G}_{s,s+h^2} \bar{v}_h(s + h^2, \cdot)(x) \quad \text{if} \quad s \leq T - h^2.
\end{aligned} \tag{2.3}$$

Then, for any $s \in [0, T]$, $x \in \mathbb{R}^d$, and $h \in (0, 1]$, we have

$$|v(s, x) - \bar{v}_h(s, x)| \le N e^{NT} h^{\delta_0^2/(2\delta_0 + 1)},$$

where the constant N depends only on K, d, and d_1.

3. Auxiliary results

Define $B_1 = \{x \in \mathbb{R}^d : |x| < 1\}$ and

$$B = A \times \{(\tau, \xi) : \tau \in (-1, 0), \xi \in B_1\}. \tag{3.1}$$

As in the example $\sigma(\alpha, t, x) = \sigma(\alpha, 0, x)$, extend σ, b, c, and f for negative t and for a fixed $\varepsilon \in (0, 1]$ and any $\beta = (\alpha, \tau, \xi) \in B$ let

$$\sigma(\beta, t, x) = \sigma_\varepsilon(\beta, t, x) = \sigma(\alpha, t + \varepsilon^2 \tau, x + \varepsilon \xi) \tag{3.2}$$

and similarly define $b(\beta, t, x)$, $c^\beta(t, x)$, and $f^\beta(t, x)$. We denote by $\mathfrak{B}$ the set of all measurable $\mathcal{F}_t$–adapted B–valued processes. As usual starting with these objects defined on B, for any $\beta \in \mathfrak{B}$, $s \ge 0$, and $x \in \mathbb{R}^d$, we define a controlled diffusion process $x_t^{\beta, s, x}$ and the value functions

$$u^\beta(s, x) = E_{s,x}^\beta \Big[\int_0^{S-s} f^{\beta_t}(s + t, x_t) e^{-\varphi_t} \, dt + g(x_{S-s}) e^{-\varphi_{S-s}} \Big],$$

$$u(s, x) = \sup_{\beta \in \mathfrak{B}} u^\beta(s, x),$$

which we consider for $s \le S$, where $S = T + \varepsilon^2$. We will keep in mind that $x_t^{\beta, s, x}$, $u^\beta(s, x)$, and $u(s, x)$ also depend on ε which is not explicitly shown just for simplicity of notation. Also let $\mathfrak{B}_h$ be the set of functions from $\mathfrak{B}$ which are constant on the intervals $[0, h^2)$, $[h^2, 2h^2)$, and so on and let

$$u_h(s, x) = \sup_{\beta \in \mathfrak{B}_h} u^\beta(s, x).$$

Lemma 3.1. *Take $s, r \in [0, S]$, $x, y \in \mathbb{R}^d$, and $\beta = (\alpha, \tau, \xi) \in \mathfrak{B}$. Then*

$$E \sup_{t \le S} |x_t^{\beta, s, x} - x_t^{\alpha, s, x}|^2 \le N e^{NT} \varepsilon^{2\delta_0}, \tag{3.3}$$

$$E \sup_{t \le S} |x_t^{\beta, s, x} - x_t^{\beta, s, y}|^2 \le N e^{NT} |x - y|^2, \tag{3.4}$$

$$E \sup_{t \le S} |x_t^{\beta, s, x} - x_t^{\beta, r, x}|^2 \le N e^{NT} |s - r|^{\delta_0}. \tag{3.5}$$

Proof. By comparing the equations defining $x_t^{\beta,s,x}$ and $x_t^{\alpha,s,x}$ one can easily get (see, for instance, Theorem 2.5.9 in [4]) that the left-hand side of (3.3) is less than

$$Ne^{NT} \sup\{|\sigma(\alpha, t + \varepsilon^2\tau, x + \varepsilon\xi) - \sigma(\alpha, t, x)|^2$$
$$+ |b(\alpha, t + \varepsilon^2\tau, x + \varepsilon\xi) - b(\alpha, t, x)|^2\}, \quad (3.6)$$

where the sup is taken over

$$\alpha \in A, \quad t \leq S, \quad \tau \in (-1, 0), \quad x \in \mathbb{R}^d, \quad \xi \in B_1.$$

In turn (3.6) is less than the right-hand side of (3.3) by Assumption 2.2.

One proves (3.4) and (3.5) similarly on the basis of the same Theorem 2.5.9 in [4]. The lemma is proved.

Corollary 3.2. *(i) For $s \in [0, T]$ and $x \in \mathbb{R}^d$,*

$$|u^\beta(s, x) - v^\alpha(s, x)| \leq Ne^{NT}\varepsilon^{\delta_0}, \quad |u(s, x) - v(s, x)| \leq Ne^{NT}\varepsilon^{\delta_0},$$

$$|u_h(s, x) - v_h(s, x)| \leq Ne^{NT}\varepsilon^{\delta_0}.$$

(ii) For $h \in (0, 1]$, $s, r \in [0, S]$, and $x, y \in \mathbb{R}^d$, we have

$$|u_h(s, x) - u_h(r, y)| + |u(s, x) - u(r, y)| \leq Ne^{NT}(|x - y| + |s - r|^{\delta_0/2}).$$

Proof. (i) It suffices to prove the first inequality. By Lemma 3.1 and Hölder's inequality

$$E \sup_{t \leq T-s} |x_t^{\beta,s,x} - x_t^{\alpha,s,x}| \leq Ne^{NT}\varepsilon^{\delta_0},$$

$$E \int_0^{S-s} |f^{\beta_t}(s + t, x_t^{\beta,s,x}) - f^{\alpha_t}(s + t, x_t^{\alpha,s,x})|\, dt \leq N(1 + T)\varepsilon^{\delta_0}$$

$$+ E \int_0^{T-s} |f^{\alpha_t}(s + t, x_t^{\beta,s,x}) - f^{\alpha_t}(s + t, x_t^{\alpha,s,x})|\, dt \leq Ne^{NT}\varepsilon^{\delta_0},$$

$$E \int_0^{S-s} |c^{\beta_t}(s + t, x_t^{\beta,s,x}) - c^{\alpha_t}(s + t, x_t^{\alpha,s,x})|\, dt \leq Ne^{NT}\varepsilon^{\delta_0},$$

$$E|g(x_{S-s}^{\beta,s,x}) - g(x_{T-s}^{\alpha,s,x})| \leq KE|x_{S-s}^{\beta,s,x} - x_{T-s}^{\alpha,s,x}| \leq N\varepsilon + Ne^{NT}\varepsilon^{\delta_0}.$$

It only remains to use

$$|f_1e^{-c_1} - f_2e^{-c_2}| \leq |f_1 - f_2| + (|f_1| + |f_2|)|c_1 - c_2|, \quad c_1, c_2 \geq 0.$$

In the same way one proves (ii). The corollary is proved.

The next lemma bears on the dynamic programming principle, in which the assertion regarding u is a particular case of Theorem 3.1.6 in [4] and the assertion regarding u_h is a particular case of Exercise 3.2.1 in [4], a solution to which can be easily obtained from Lemma 3.2.14 and the proof of Lemma 3.3.1 of [4].

Lemma 3.3. *(i) For any $x \in \mathbb{R}^d$ and $0 \le s \le t \le S$,*

$$u(s,x) = \sup_{\beta \in \mathfrak{B}} E_{s,x}^{\beta} \Big[\int_0^{t-s} f^{\beta_r}(s+r, x_r) e^{-\varphi_r} \, dr + u(t, x_{t-s}) e^{-\varphi_{t-s}} \Big].$$

(ii) For any $x \in \mathbb{R}^d$ and $0 \le s \le t \le S$ such that $(t-s)/h^2$ is an integer,

$$u_h(s,x) = \sup_{\beta \in \mathfrak{B}_h} E_{s,x}^{\beta} \Big[\int_0^{t-s} f^{\beta_r}(s+r, x_r) e^{-\varphi_r} \, dr + u_h(t, x_{t-s}) e^{-\varphi_{t-s}} \Big].$$

This lemma allows us to improve part of assertions in Corollary 3.2.

Corollary 3.4. *(i) For any $x \in \mathbb{R}^d$ and $0 \le s \le t \le S$,*

$$|u(s,x) - u(t,x)| \le N e^{NT} |t-s|^{1/2},$$

$$|u_h(s,x) - u_h(t,x)| \le N e^{NT} (h^{\delta_0} + |t-s|^{1/2}).$$

(ii) For any $x \in \mathbb{R}^d$ and $0 \le s \le t \le S$ such that $(t-s)/h^2$ is an integer,

$$|u_h(s,x) - u_h(t,x)| \le N e^{NT} |t-s|^{1/2}.$$

Indeed, the second inequality in (i) follows trivially from (ii) and Corollary 3.2 (ii). To prove the first assertion in (i) it suffices to notice that

$$|u(s,x) - u(t,x)| \le \sup_{\beta \in \mathfrak{B}} E_{s,x}^{\beta} \Big[\int_0^{t-s} |f^{\beta_r}(s+r, x_r)| \, dr + |u(t, x_{t-s}) e^{-\varphi_{t-s}} - u(t,x)| \Big]$$

$$\le K(t-s) + \sup |u| K(t-s) + \sup_{\beta \in \mathfrak{B}} E_{s,x}^{\beta} |u(t, x_{t-s}) - u(t,x)|$$

$$\le N(1+T)(t-s) + N e^{NT} \sup_{\beta \in \mathfrak{B}} E_{s,x}^{\beta} |x_{t-s} - x| \le N e^{NT} |t-s|^{1/2}.$$

Similarly one proves assertion (ii).

Next, take a nonnegative function $\zeta \in C_0^{\infty}((-1,0) \times B_1)$ with unit integral, and for $\varepsilon > 0$ define $\zeta_\varepsilon(t,x) = \varepsilon^{-d-2} \zeta(t/\varepsilon^2, x/\varepsilon)$. Also use the notation

$$z^{(\varepsilon)}(t,x) = z(t,x) * \zeta_\varepsilon(t,x).$$

The knowledge of Hölder continuity of u_h allows us to estimate $C^{2+\delta_0}$-norm of $u_h^{(\varepsilon)}$. For a function $z(t,x)$ defined on $[0,T] \times \mathbb{R}^d$ we denote

$$|z|_{2+\delta_0} = |z|_0 + |z_x|_0 + |z_{xx}|_0 + |z_t|_0 + [z_{xx}]_{\delta_0} + [z_t]_{\delta_0},$$

where z_x, z_{xx}, z_t are the gradient of z in x, the Hessian matrix of z in x, and the derivative in t, respectively, and

$$|z|_0 = \sup_{[0,T] \times \mathbb{R}^d} |z(t,x)|, \quad [z]_{\delta_0} = \sup_{(t,x),(s,y) \in [0,T] \times \mathbb{R}^d} \frac{|z(t,x) - z(s,y)|}{|t-s|^{\delta_0/2} + |x-y|^{\delta_0}}.$$

The space $C^{2+\delta_0}$ is defined as the collection of all functions z with finite norm $|z|_{2+\delta_0}$.

Lemma 3.5. *If $\varepsilon \geq h^{\delta_0}$, then*

$$|u_h^{(\varepsilon)}|_{2+\delta_0} \leq Ne^{NT}\varepsilon^{-1-\delta_0}, \quad |u_h^{(\varepsilon)} - u_h|_0 \leq Ne^{NT}\varepsilon. \qquad (3.7)$$

Proof. Observe that, owing to Corollaries 3.2 and 3.4 and the inequality $\varepsilon \geq h^{\delta_0}$, for $s \in [0,T]$ and $x \in \mathbb{R}^d$, we have

$$\left|\frac{\partial}{\partial s}u_h^{(\varepsilon)}(s,x)\right| = \varepsilon^{-2}\left|\int_{\mathbb{R}^{d+1}} u_h(s - \varepsilon^2\tau, x - \varepsilon\xi)\frac{\partial}{\partial s}\zeta(\tau,\xi)\,d\tau d\xi\right|$$

$$= \varepsilon^{-2}\left|\int_{\mathbb{R}^{d+1}} [u_h(s - \varepsilon^2\tau, x - \varepsilon\xi) - u_h(s,x)]\frac{\partial}{\partial s}\zeta(\tau,\xi)\,d\tau d\xi\right|$$

$$\leq Ne^{NT}\varepsilon^{-2}(h^{\delta_0} + \varepsilon) \leq Ne^{NT}\varepsilon^{-1}.$$

Similarly,

$$\left|\frac{\partial}{\partial x^i}u_h^{(\varepsilon)}\right|_0 \leq Ne^{NT} \leq Ne^{NT}\varepsilon^{-1}, \quad \left|\frac{\partial^2}{\partial x^i \partial x^j}u_h^{(\varepsilon)}\right|_0 \leq Ne^{NT}\varepsilon^{-1},$$

$$\left|\frac{\partial^2}{\partial s^2}u_h^{(\varepsilon)}\right|_0 + \left|\frac{\partial^3}{\partial x^i \partial x^j \partial s}u_h^{(\varepsilon)}\right|_0 \leq Ne^{NT}\varepsilon^{-3},$$

$$\left|\frac{\partial^2}{\partial s \partial x^i}u_h^{(\varepsilon)}\right|_0 + \left|\frac{\partial^3}{\partial x^i \partial x^j \partial x^k}u_h^{(\varepsilon)}\right|_0 \leq Ne^{NT}\varepsilon^{-2}.$$

It follows that

$$\left|\frac{\partial}{\partial s}u_h^{(\varepsilon)}(s,x) - \frac{\partial}{\partial s}u_h^{(\varepsilon)}(t,y)\right| \leq |t - s|\left|\frac{\partial^2}{\partial s^2}u_h^{(\varepsilon)}\right|_0 + N|x - y|\left|\frac{\partial^2}{\partial x^i \partial s}u_h^{(\varepsilon)}\right|_0$$

$$\leq Ne^{NT}|t - s|\varepsilon^{-3} + Ne^{NT}|x - y|\varepsilon^{-2}.$$

As easy to see, the last expression is less than $Ne^{NT}\varepsilon^{-1-\delta_0}(|t-s|^{\delta_0/2} + |x-y|^{\delta_0})$ if $|t-s|^{1/2}+|x-y| \leq \varepsilon$. On the other hand, if $|t-s|^{1/2}+|x-y| \geq \varepsilon$, then

$$\left|\frac{\partial}{\partial s}u_h^{(\varepsilon)}(s,x) - \frac{\partial}{\partial s}u_h^{(\varepsilon)}(t,y)\right| \leq 2\left|\frac{\partial}{\partial s}u_h^{(\varepsilon)}\right|_0$$

$$\leq Ne^{NT}\varepsilon^{-1} \leq Ne^{NT}\varepsilon^{-1-\delta_0}(|t - s|^{\delta_0/2} + |x - y|^{\delta_0}).$$

Hence the inequality between extreme terms holds for all $t, s \leq T$. In the same way one gets that

$$\left|\frac{\partial^2}{\partial x^i \partial x^j}u_h^{(\varepsilon)}(s,x) - \frac{\partial^2}{\partial x^i \partial x^j}u_h^{(\varepsilon)}(t,y)\right| \leq Ne^{NT}\varepsilon^{-1-\delta_0}(|t-s|^{\delta_0/2}+|x-y|^{\delta_0}),$$

and this proves the first inequality in (3.7).

To prove the second one, it suffices to notice that

$$|u_h^{(\varepsilon)}(s,x) - u_h(s,x)| \leq \int_{\mathbb{R}^{d+1}} |u_h(s - \varepsilon^2\tau, x - \varepsilon\xi) - u_h(s,x)|\zeta(\tau,\xi)\,d\tau d\xi$$

$$\leq Ne^{NT}(h^{\delta_0} + \varepsilon) \leq Ne^{NT}\varepsilon,$$

The lemma is proved.

4. THE PROOFS OF THEOREMS 2.3 AND 2.5

4.1. **Proof of Theorem 2.3.** Since obviously $v_h \leq v$, it suffices to prove

$$v(s, x) \leq v_h(s, x) + Ne^{NT}h^{\delta_0^2/(2\delta_0+1)}. \tag{4.1}$$

Furthermore, if $s \in [T - h^2, T]$, then, for any $\alpha \in \mathfrak{A}$,

$$|v^\alpha(s, x) - g(x)| \leq Kh^2 + E_{s,x}^\alpha|g(x_{T-s})e^{-\varphi_{T-s}} - g(x)|$$

$$\leq Nh^2 + E_{s,x}^\alpha|g(x_{T-s}) - g(x)| \leq N(h^2 + E_{s,x}^\alpha|x_{T-s} - x|)$$

$$\leq Nh \leq Ne^{NT}h^{\delta_0^2/(2\delta_0+1)}.$$

Hence we only need prove (4.1) for $s \leq T - h^2$, assuming without loss of generality that $T \geq h^2$.

Denote

$$\varepsilon = h^{\delta_0/(2\delta_0+1)}. \tag{4.2}$$

By comparing the equations for $x_t^{\beta,s,x}$ and $x_t^{\alpha,s+\varepsilon^2\tau,x+\varepsilon\xi}$, one easily sees that, if $\beta = (\alpha, \tau, \xi) \in \mathfrak{B}_h$ and $t \leq h^2$, then

$$x_t^{\beta,s,x} = x_t^{\alpha,s+\varepsilon^2\tau,x+\varepsilon\xi} - \varepsilon\xi, \quad \varphi_t^{\beta,s,x} = \varphi_t^{\alpha,s+\varepsilon^2\tau,x+\varepsilon\xi}.$$

This and Lemma 3.3 (ii) imply that, for any $\beta = (\alpha, \tau, \xi) \in B$, $s \in [0, S - h^2]$, and $x \in \mathbb{R}^d$,

$$u_h(s, x) \geq E[u_h(s + h^2, x_{h^2}^{\beta,s,x})e^{-\varphi_{h^2}^{\beta,s,x}} + \int_0^{h^2} f^\beta(s + t, x_t^{\beta,s,x})e^{-\varphi_t^{\beta,s,x}} dt]$$

$$= E_{s+\varepsilon^2\tau,x+\varepsilon\xi}^\alpha[u_h(s + h^2, x_{h^2} - \varepsilon\xi)e^{-\varphi_{h^2}} + \int_0^{h^2} f^\alpha(s + \varepsilon^2\tau + t, x_t)e^{-\varphi_t} dt]$$

We restate this in the following way

$$u_h(s - \varepsilon^2\tau, x - \varepsilon\xi) \geq E_{s,x}^\alpha[u_h(s - \varepsilon^2\tau + h^2, x_{h^2} - \varepsilon\xi)e^{-\varphi_{h^2}}$$

$$+ \int_0^{h^2} f^\alpha(s + t, x_t)e^{-\varphi_t} dt] \tag{4.3}$$

if $\alpha \in A$, $\tau \in (-1, 0)$, $\xi \in B_1$, $s \leq S - h^2 + \varepsilon^2\tau$, and $x \in \mathbb{R}^d$. It follows from (4.3) that, for any constant policy α, $s \leq T - h^2$, and $x \in \mathbb{R}^d$, we have

$$u_h^{(\varepsilon)}(s, x) \geq E_{s,x}^\alpha[\int_0^{h^2} f^\alpha(s + t, x_t)e^{-\varphi_t} dt + u_h^{(\varepsilon)}(s + h^2, x_{h^2})e^{-\varphi_{h^2}}].$$

By Itô's formula we further infer that

$$0 \geq h^{-2} E_{s,x}^{\alpha} \int_0^{h^2} [f^{\alpha}(s+t, x_t) + L^{\alpha} u_h^{(\varepsilon)}(s+t, x_t)] e^{-\varphi_t} \, dt,$$

$$[L^{\alpha} u_h^{(\varepsilon)}(s, x) + f^{\alpha}(s, x)] h^{-2} E_{s,x}^{\alpha} \int_0^{h^2} e^{-\varphi_t} \, dt \leq I^{\alpha}(s, x), \qquad (4.4)$$

where

$$L^{\alpha} z(s, x) := a^{ij}(\alpha, s, x) z_{x^i x^j}(s, x) + b^i(\alpha, s, x) z_{x^i}(s, x)$$

$$- c^{\alpha}(s, x) z(s, x) + z_s(s, x), \quad (a^{ij}) := \frac{1}{2} \sigma \sigma^*, \quad (4.5)$$

$$I^{\alpha}(s, x) := h^{-2} E_{s,x}^{\alpha} \int_0^{h^2} [|f^{\alpha}(s+t, x_t) - f^{\alpha}(s, x)|$$

$$+ |L^{\alpha} u_h^{(\varepsilon)}(s+t, x_t) - L^{\alpha} u_h^{(\varepsilon)}(s, x)|] \, dt$$

$$\leq (2K + N^{\alpha}) h^{-2} \int_0^{h^2} (t^{\delta_0/2} + E_{s,x}^{\alpha} |x_t - x|^{\delta_0}) \, dt \leq N(1 + N^{\alpha}) h^{\delta_0}$$

with N^{α} being the sum of the Hölder $\delta_0/2$-constant with respect to s and the Hölder δ_0-constant with respect to x of the function $L^{\alpha} u_h^{(\varepsilon)}(s, x)$.

To estimate N^{α} we remember that the Hölder continuity of the co-efficients of L^{α} is one of our assumptions and we use Lemma 3.5. Then we easily get that $N^{\alpha} \leq N e^{NT} \varepsilon^{-1-\delta_0}$, so that (4.4) implies

$$L^{\alpha} u_h^{(\varepsilon)}(s, x) + f^{\alpha}(s, x) \leq N e^{NT} \varepsilon^{-1-\delta_0} h^{\delta_0} \qquad (4.6)$$

for any $\alpha \in A$, $s \leq T - h^2$, and $x \in \mathbb{R}^d$.

Next, from (4.6) by Itô's formula, for any $\alpha \in \mathfrak{A}$, $s \leq T - h^2$, and $x \in \mathbb{R}^d$, we obtain

$$E_{s,x}^{\alpha} u_h^{(\varepsilon)}(T - h^2, x_{T-h^2-s}) e^{-\varphi_{T-h^2-s}} \leq u_h^{(\varepsilon)}(s, x)$$

$$- E_{s,x}^{\alpha} \int_0^{T-s-h^2} f^{\alpha_t}(s+t, x_t) e^{-\varphi_t} \, dt + N e^{NT} \varepsilon^{-1-\delta_0} h^{\delta_0},$$

which by Lemma 3.5 allows us to conclude that

$$E_{s,x}^{\alpha} u_h(T - h^2, x_{T-h^2-s}) e^{-\varphi_{T-h^2-s}} \leq u_h(s, x)$$

$$- E_{s,x}^{\alpha} \int_0^{T-s-h^2} f^{\alpha_t}(s+t, x_t) e^{-\varphi_t} \, dt + N e^{NT} (\varepsilon + \varepsilon^{-1-\delta_0} h^{\delta_0}). \quad (4.7)$$

Furthermore, notice that

$$|u_h(T - h^2, x) - g(x)| \leq |u_h(S, x) - g(x)| + |u_h(S, x) - u_h(T - h^2, x)|$$

$$= |u_h(S, x) - u_h(T - h^2, x)| \leq N e^{NT} (h^{\delta_0} + \varepsilon) \leq N e^{NT} \varepsilon.$$

Hence, from (4.7) and the inequality $u_h \leq v_h + Ne^{NT}\varepsilon^{\delta_0}$ (see Corollary 3.2 (i)), we obtain

$$E_{s,x}^{\alpha} g(x_{T-s})e^{-\varphi_{T-s}} \leq v_h(s,x)$$

$$-E_{s,x}^{\alpha} \int_0^{T-s} f^{\alpha_t}(s+t,x_t)e^{-\varphi_t}\,dt + Ne^{NT}[\varepsilon^{-1-\delta_0}h^{\delta_0} + h^2 + \varepsilon^{\delta_0}].$$

Upon remembering that $\varepsilon = h^{\delta_0/(2\delta_0+1)}$, we finally get

$$v^{\alpha}(s,x) \leq v_h(s,x) + Ne^{NT}h^{\delta_0^2/(2\delta_0+1)}$$

and this proves (4.1) for $s \in [0, T-h^2]$.

4.2. **Proof of Theorem 2.5.** Fix $h \in (0,1]$ and for $\alpha \in \mathfrak{A}_h$ define $\bar{x}_t = \bar{x}_t^{\alpha,s,x}$ recursively by

$$\bar{x}_0 = x, \quad \bar{x}_t = \bar{x}_{nh^2} + \sigma(\alpha_{nh^2}, s+nh^2, \bar{x}_{nh^2})(w_t - w_{nh^2})$$

$$+ b(\alpha_{nh^2}, s+nh^2, \bar{x}_{nh^2})(t - nh^2)$$

for $nh^2 \leq t \leq (n+1)h^2$. Of course, $\bar{x}_t^{\alpha,s,x}$ also depends on h, dependence on which is not shown explicitly just for simplicity of notation. It is easy to see that $\bar{x}_t^{\alpha,s,x}$ satisfies

$$\bar{x}_t = x + \int_0^t \sigma(\alpha_r, s+\kappa_h(r), \bar{x}_{\kappa_h(r)})\,dw_r + \int_0^t b(\alpha_r, s+\kappa_h(r), \bar{x}_{\kappa_h(r)})\,dr, \tag{4.8}$$

where $\kappa_h(t) = h^2[th^{-2}]$. Next, let

$$\bar{\varphi}_t^{\alpha,s,x} = \int_0^t c^{\alpha_r}(s+\kappa_h(r), \bar{x}_{\kappa_h(r)}^{\alpha,s,x})\,dr,$$

$$\bar{v}_h^{\alpha}(s,x) = E_{s,x}^{\alpha}\Big[\int_0^{\kappa_h(T-s)} f^{\alpha_t}(s+\kappa_h(t), \bar{x}_{\kappa_h(t)})e^{-\bar{\varphi}_{\kappa_h(t)}}\,dt$$

$$+ g(\bar{x}_{\kappa_h(T-s)})e^{-\bar{\varphi}_{\kappa_h(T-s)}}\Big].$$

Notice that equation (2.3) is a dynamic programming equation for the problem of maximizing $\bar{v}_h^{\alpha}(s,x)$ over $\alpha \in \mathfrak{A}_h$. Therefore,

$$\bar{v}_h = \sup_{\alpha \in \mathfrak{A}_h} \bar{v}_h^{\alpha}(s,x).$$

Now we see that, owing to Theorem 2.3, to prove the present theorem, it suffices to prove that, for any $\alpha \in \mathfrak{A}_h$, $s \in [0,T]$, and $x \in \mathbb{R}^d$,

$$|v^{\alpha}(s,x) - \bar{v}_h^{\alpha}(s,x)| \leq Ne^{NT}h^{\delta_0}. \tag{4.9}$$

This inequality is similar to the inequalities from Corollary 3.2 and we prove it by using again Theorem 2.5.9 of [4]. We rewrite (4.8) as

$$\bar{x}_t = x + \int_0^t \bar{\sigma}_r(\bar{x}_r)\,dw_r + \int_0^t \bar{b}_r(\bar{x}_r)\,dr,$$

where
$$\bar{\sigma}_t(y) = \sigma(\alpha_t, s + \kappa_h(t), y + \bar{x}^{\alpha,s,x}_{\kappa_h(t)} - \bar{x}^{\alpha,s,x}_t),$$
$$\bar{b}_t(y) = b(\alpha_t, s + \kappa_h(t), y + \bar{x}^{\alpha,s,x}_{\kappa_h(t)} - \bar{x}^{\alpha,s,x}_t).$$

Then by Theorem 2.5.9 of [4] we get

$$E^\alpha_{s,x} \sup_{t \le T} |\bar{x}_t - x_t|^2 \le N e^{NT} E^\alpha_{s,x} \int_0^T [|\sigma(\alpha_t, s + t, x_t) - \bar{\sigma}_t(x_t)|^2$$
$$+ |b(\alpha_t, s + t, x_t) - \bar{b}_t(x_t)|^2] \, dt$$
$$\le N e^{NT} \int_0^T [h^{2\delta_0} + E^\alpha_{s,x} |\bar{x}_{\kappa_h(t)} - \bar{x}_t|^2] \, dt \le N e^{NT} h^{2\delta_0}.$$

It follows that
$$E^\alpha_{s,x} \sup_{t \le T} |\bar{x}_t - x_t| \le N e^{NT} h^{\delta_0},$$

$$E^\alpha_{s,x} \int_0^T |f^{\alpha_t}(s + \kappa_h(t), \bar{x}_{\kappa_h(t)}) - f^{\alpha_t}(s + t, x_t)| \, dt \le N e^{NT} h^{\delta_0}$$
$$+ K \sup_{t \le T} E^\alpha_{s,x} |x_{\kappa_h(t)} - x_t| \le N e^{NT} h^{\delta_0},$$

$$E^\alpha_{s,x} \int_0^T |c^{\alpha_t}(s + \kappa_h(t), \bar{x}_{\kappa_h(t)}) - c^{\alpha_t}(s + t, x_t)| \, dt \le N e^{NT} h^{\delta_0},$$

$$E^\alpha_{s,x} |g(\bar{x}_{\kappa_h(T-s)}) - g(x_{T-s})|$$
$$\le N e^{NT} h^{\delta_0} + K E^\alpha_{s,x} |x_{\kappa_h(T-s)} - x_{T-s}| \le N e^{NT} h^{\delta_0}.$$

It is easy to see that (4.9) follows from the above estimates. The theorem is proved.

5. Other methods of approximating value functions

Remember that the operator L^α is introduced in (4.5) and define
$$F[u] = \sup_{\alpha \in A} [L^\alpha u + f^\alpha].$$

By definition v is a probabilistic solution of the problem
$$F[u] = 0 \quad \text{in} \quad (0, T) \times \mathbb{R}^d, \quad u(T, \cdot) = g. \tag{5.1}$$

The function v is also a viscosity solution of (5.1) (see, for instance, [3]).

Next, we describe the approximating scheme for solving (5.1) introduced in [7]. Let $\mathbb{R}^{d+1}_+ = [0, \infty) \times \mathbb{R}^d$ and $\mathcal{B} = \mathcal{B}(\mathbb{R}^{d+1}_+)$ be the set of all bounded functions on $\mathbb{R}^{d+1}_+$. For any $h \in (0, 1]$ let a number $p_h \in [1, \infty)$ and an operator $F_h : u \in \mathcal{B} \to F_h[u] \in \mathcal{B}$ be defined.

We need the space $C^{2+\delta_0}([t, t+h^2])$ provided with norm $|\cdot|_{2+\delta_0, [t, t+h^2]}$. These objects are introduced in the same way as the space $C^{2+\delta_0}$ in

Section 3 before Lemma 3.5 only this time we consider functions defined on $[t, t + h^2] \times \mathbb{R}^d$.

Assumption 5.1. (i) F_h is an h^2–local operator in t, that is, for any $t \in [0, \infty)$ and $u_1, u_2 \in \mathcal{B}$, we have $F_h[u_1](t, x) = F_h[u_2](t, x)$ for all $x \in \mathbb{R}^d$ whenever $u_1(s, y) = u_2(s, y)$ for all $s \in [t, t + h^2]$ and $y \in \mathbb{R}^d$;

(ii) F_h is locally consistent with F in the sense that, for any $t \in [0, \infty)$, $x \in \mathbb{R}^d$, and $u \in C^{2+\delta_0}([t, t + h^2])$, we have

$$|F_h[u](t, x) - F[u](t, x)| \leq K h^{\delta_0} |u|_{2+\delta_0, [t, t+h^2]};$$

(iii) the operator $u \to \Phi_h[u] := F_h[u] + p_h u$ is monotone, by which we mean that if $u_1, u_2 \in \mathcal{B}$ and $u_1 \geq u_2$, then $\Phi_h[u_1] \geq \Phi_h[u_2]$; moreover,

(iv) let $\ell := \ell(t) := e^{-2t}$, then for any constant $M \geq 0$ and $u_1, u_2 \in \mathcal{B}$ such that $u_1 + M\ell \geq u_2$, we have

$$\Phi_h[u_1] + M\ell(p_h - 1) \geq \Phi_h[u_1 + M\ell] \geq \Phi_h[u_2].$$

Remark 5.2. The reader can find in [7] several examples of operators F_h satisfying Assumption 5.1. In particular, these examples include implicit and explicit finite-difference schemes. Furthermore, as easy to see the operators

$$h^{-2}[G_{s,s+h^2} u(s + h^2, \cdot)(x) - u(s, x)], \quad h^{-2}[\bar{G}_{s,s+h^2} u(s + h^2, \cdot)(x) - u(s, x)] \tag{5.2}$$

with $p_h = h^{-2}$ satisfy Assumption 5.1.

By Lemma 1.7 of [7] there exists a unique bounded function $\bar{v}_h$ defined on $[0, T + h^2] \times \mathbb{R}^d$ and solving the problem

$$F_h[\bar{v}_h](t, x) = 0 \quad \forall t \in [0, T], x \in \mathbb{R}^d;$$
$$\bar{v}_h = g \quad \forall t \in (T, T + h^2], x \in \mathbb{R}^d.$$

Theorem 1.9 in [7] in our particular situation of Lipschitz continuous c, f, g reads as follows.

Theorem 5.3. *For any $h \in (0, 1]$, in $[0, T] \times \mathbb{R}^d$ we have*

$$\bar{v}_h \leq v + N e^{NT} h^{\delta_0^2 / (2\delta_0 + 1)}.$$

The following is an improvement of Theorem 1.11 of [7].

Theorem 5.4. *For any $h \in (0, 1]$, in $[0, T] \times \mathbb{R}^d$ we have*

$$v \leq \bar{v}_h + N e^{NT} h^{\delta_1},$$

where $\delta_1 = \delta_0^4 (2\delta_0 + 1)^{-1} (4\delta_0 + 2 + \delta_0^2)^{-1}$, which is $1/21$ if $\delta_0 = 1$.

Proof. By Lemma 3.2 of [7] for any integer $n \geq 1$ and $h^2 \leq 1/n$ we have

$$v_{1/\sqrt{n}} \leq \bar{v}_h + N e^{NT} n h^{\delta_0^2/(2\delta_0+1)}.$$

Hence, by Theorem 2.3

$$v \leq \bar{v}_h + N e^{NT} [n h^{\delta_0^2/(2\delta_0+1)} + n^{-\delta_0^2/(4\delta_0+2)}].$$

By taking n of order $h^{-2\delta_0^2/(4\delta_0+2+\delta_0^2)}$, we get the result. The theorem is proved.

Remark 5.5. If $\sigma \equiv 0$, Bellman's equation becomes the Hamilton-Jacobi equation. In this case better results can be found in Appendix A, written by M. Falcone, in [1] and in the references therein. Also notice that in the case of $\delta_0 = 1$ and σ, b independent of (t, x) we have $|v - \bar{v}_h| \leq N e^{NT} h^{1/3}$ as can be seen from [5].

Remark 5.6. The rate of convergence in Theorem 1.11 of [7] is $1/39$ if $\delta_0 = 1$. Improving it to $1/21$ is of course a step forward. However, we still do not know what kind of additional conditions are needed in order to get the rate $1/3$ for more or less general approximating operators F_h. Even more than that, we do not know what is the real rate of convergence in the case when $d = 2$ and

$$F[u](s, x) = \sigma_1^2(x)(u_{x^1 x^1})_+ + \sigma_2^2(x)(u_{x^2 x^2})_+ + u_s,$$

$$\begin{aligned}
h^2 F_h[u](s, x) \\
= \sigma_1^2(x) \big(u(s + h^2, x + h e_1) - u(s + h^2, x) + u(s + h^2, x - e_1 h) \big)_+ \\
+ \sigma_2^2(x) \big(u(s + h^2, x + h e_2) - u(s + h^2, x) + u(s + h^2, x - e_2 h) \big)_+ \\
+ u(s + h^2, x) - u(s, x), \quad (5.3)
\end{aligned}$$

where e_i are unit basis vectors and σ_i are smooth, say small functions having zeros. Theorem 5.4 only says that the rate is not smaller than $1/21$ if g is bounded and Lipschitz.

Approximations (5.2) with the same time step as in (5.3) are of order at least $1/3$ by Theorems 2.3 and 2.5. However computing the operators in (5.2) requires computing some integrals at all points of $\mathbb{R}^d$, whereas in (5.3) one only meets simplest sums and one can restrict oneself to points on a grid.

Below we give some conditions which allow one to construct approximations like (5.2) with other type of random variables involved, say only taking finitely many values on a grid but still not as specified as in (5.3). As everywhere in the article, the conditions of Section 2 are assumed to hold.

Fix an $h \in (0, 1]$ and assume that we are given an $\mathbb{R}^{d_1}$-valued random variable $\hat{w}$ such that

$$E\hat{w} = 0, \quad |E\hat{w}^i \hat{w}^j - h^2 \delta^{ij}| \leq K h^{2+\delta_0}, \quad E|\hat{w}|^{2+\delta_0} \leq K h^{2+\delta_0}. \quad (5.4)$$

Theorem 5.7. *For $\alpha \in A$, $s \geq 0$, and $x \in \mathbb{R}^d$ define*

$$\hat{y}^{\alpha,s,x} = x + \sigma(\alpha, s, x)\hat{w} + b(\alpha, s, x)h^2,$$

$$\hat{G}^\alpha_{s,s+h^2} z(x) = f^\alpha(s, x)h^2 + e^{-c^\alpha(s,x)h^2} Ez(\hat{y}^{\alpha,s,x}),$$

$$\hat{G}_{s,s+h^2} z = \sup_{\alpha \in A} \hat{G}^\alpha_{s,s+h^2} z.$$

Define recursively

$$\hat{v}_h(s, x) = g(x) \quad if \quad s \in (T - h^2, T],$$

$$\hat{v}_h(s, x) = \hat{G}_{s,s+h^2} \hat{v}_h(s + h^2, \cdot)(x) \quad if \quad s \leq T - h^2. \quad (5.5)$$

Then, for any $s \in [0, T]$ and $x \in \mathbb{R}^d$, we have

$$|v(s, x) - \hat{v}_h(s, x)| \leq N e^{NT} h^{\delta_0^2/(2\delta_0+1)}.$$

The proof of this theorem is based on several auxiliary results. Take a sequence of i.i.d. $\mathbb{R}^{d_1}$-valued random variables $\hat{w}_n$, $n = 0, 1, 2...$, having the same distribution as $\hat{w}$. Let $\hat{\mathcal{F}}_t$ be the σ-field generated by $\hat{w}_n$ for $n \leq [th^{-2}]$ and let $\hat{\mathfrak{A}}_h$ be the set of all A-valued $\hat{\mathcal{F}}_t$-adapted processes. For $\alpha \in \hat{\mathfrak{A}}_h$ define $\hat{x}_{nh^2} = \hat{x}^{\alpha,s,x}_{nh^2}$ recursively by

$$\hat{x}_0 = x, \quad \hat{x}_{(n+1)h^2} = \hat{x}_{nh^2} + \sigma(\alpha_{nh^2}, s + nh^2, \hat{x}_{nh^2})\hat{w}_n$$

$$+ b(\alpha_{nh^2}, s + nh^2, \hat{x}_{nh^2})h^2.$$

Again we do not include h in the notation $\hat{x}^{\alpha,s,x}_{nh^2}$ just for simplicity. Also let

$$\hat{\varphi}^{\alpha,s,x}_t = \int_0^t c^{\alpha_r}(s + \kappa_h(r), \hat{x}^{\alpha,s,x}_{\kappa_h(r)})\, dr,$$

$$\hat{v}^\alpha_h(s, x) = E^\alpha_{s,x}\Big[\int_0^{\kappa_h(T-s)} f^{\alpha_t}(s + \kappa_h(t), \hat{x}_{\kappa_h(t)})e^{-\hat{\varphi}_{\kappa_h(t)}}\, dt$$

$$+ g(\hat{x}_{\kappa_h(T-s)})e^{-\hat{\varphi}_{\kappa_h(T-s)}}\Big].$$

Notice that equation (5.5) is the dynamic programming equation for the problem of maximizing $\hat{v}^\alpha_h(s, x)$ over $\alpha \in \hat{\mathfrak{A}}_h$. Therefore,

$$\hat{v}_h(s, x) = \sup_{\alpha \in \hat{\mathfrak{A}}_h} \hat{v}^\alpha_h(s, x).$$

Next we consider "shaken" coefficients, fixing an $\varepsilon \in (0, 1]$. We use again objects (3.1) and (3.2) and we define $b(\beta, t, x)$, $c^\beta(t, x)$, and $f^\beta(t, x)$ similarly. We denote by $\hat{\mathfrak{B}}_h$ the set of all measurable $\hat{\mathcal{F}}_t$-adapted B-valued processes.

By using the functions $\sigma(\beta, t, x)$, $b(\beta, t, x)$, $c^\beta(t, x)$, and $f^\beta(t, x)$, in an obvious way, we define $\hat{x}_t^{\beta,s,x}$ and $\hat{\varphi}_t^{\beta,s,x}$ and let

$$\hat{u}^\beta(s, x) = E_{s,x}^\beta \Big[\int_0^{\kappa_h(S-s)} f^{\beta_t}(s + \kappa_h(t), \hat{x}_{\kappa_h(t)}) e^{-\hat{\varphi}_{\kappa_h(t)}} \, dt$$

$$+ g(\hat{x}_{\kappa_h(S-s)}) e^{-\hat{\varphi}_{\kappa_h(S-s)}} \Big], \quad \hat{u}_h(s, x) = \sup_{\beta \in \hat{\mathfrak{B}}_h} \hat{u}^\beta(s, x),$$

where, as before, $S = T + \varepsilon^2$.

Lemma 5.8. *(i) For $s \in [0, T]$ and $x \in \mathbb{R}^d$, we have $|\hat{u}_h(s, x) - \hat{v}_h(s, x)| \leq N e^{NT} \varepsilon^{\delta_0}$.*
   (ii) For $s, r \in [0, S]$ and $x, y \in \mathbb{R}^d$, we have

$$|\hat{u}_h(s, x) - \hat{u}_h(s, y)| \leq N e^{NT} |x - y|,$$

$$|\hat{u}_h(s, x) - \hat{u}_h(t, x)| \leq N e^{NT} (h^{\delta_0} + |t - s|^{1/2}).$$

We prove this lemma in Section 6. Our next lemma is the dynamic programming principle.

Lemma 5.9. *For any $x \in \mathbb{R}^d$ and $0 \leq s \leq t \leq S$ such that $(t - s)/h^2$ is an integer,*

$$\hat{u}_h(s, x) = \sup_{\beta \in \hat{\mathfrak{B}}_h} E_{s,x}^\beta \Big[\int_0^{t-s} f^{\beta_r}(s + \kappa_h(r), \hat{x}_{\kappa_h(r)}) e^{-\hat{\varphi}_{\kappa_h(r)}} \, dr$$

$$+ \hat{u}_h(t, \hat{x}_{t-s}) e^{-\hat{\varphi}_{t-s}} \Big].$$

In particular, for $t = s + h^2 \leq S$, we have

$$\hat{u}_h(s, x) = \sup_{(\alpha, \tau, \xi) \in B} [f^\alpha(s + \varepsilon^2\tau, x + \varepsilon\xi) h^2$$

$$+ e^{-c^\alpha(s + \varepsilon^2\tau, x + \varepsilon\xi) h^2} E\hat{u}_h(s + h^2, \hat{y}^{\alpha, s + \varepsilon^2\tau, x + \varepsilon\xi} - \varepsilon\xi)]. \quad (5.6)$$

Finally, we need the following fact which is easily proved by using Taylor's formula and assumption (5.4).

Lemma 5.10. *If $z \in C^{2+\delta_0}[0, T]$, then, for any $\alpha \in A$, $s \in [0, T - h^2]$, and $x \in \mathbb{R}^d$, we have*

$$|L^\alpha z(s, x) - h^{-2}[e^{-c^\alpha(s,x)h^2} Ez(s + h^2, \hat{y}^{\alpha,s,x}) - z(s, x)]| \leq N h^{\delta_0} |z|_{2+\delta_0, [0,T]}.$$

Proof of Theorem 5.7. First we prove that

$$v \leq \hat{v}_h + N e^{NT} h^{\delta_0^2/(2\delta_0 + 1)}. \quad (5.7)$$

Take ε from (4.2) and proceed as in Subsection 4.1 observing that (5.6) implies that

$$\hat{u}_h(s - \varepsilon^2 \tau, x - \varepsilon\xi) \geq f^\alpha(s, x)h^2$$
$$+ e^{-c^\alpha(s,x)h^2} E\hat{u}_h(s - \varepsilon^2\tau + h^2, \hat{y}^{\alpha,s,x} - \varepsilon\xi)$$

provided $\alpha \in A$, $\tau \in (-1, 0)$, $\xi \in B_1$, $s \leq S - h^2 + \varepsilon^2\tau$, and $x \in \mathbb{R}^d$. Hence for any $\alpha \in A$, $s \leq T - h^2$, and $x \in \mathbb{R}^d$, we have

$$\hat{u}_h^{(\varepsilon)}(s, x) \geq f^\alpha(s, x)h^2 + e^{-c^\alpha(s,x)h^2} E\hat{u}_h^{(\varepsilon)}(s + h^2, \hat{y}^{\alpha,s,x}).$$

Now Lemma 5.10 implies that $L^\alpha \hat{u}_h^{(\varepsilon)} + f^\alpha \leq Nh^{\delta_0}|\hat{u}_h^{(\varepsilon)}|_{2+\delta_0,[0,T]}$ in $[0, T - h^2] \times \mathbb{R}^d$. Furthermore, we notice that the proof of Lemma 3.5 carries over to the function $\hat{u}_h^{(\varepsilon)}$ without any change owing to Lemma 5.8 (ii). Hence (4.6) holds with $\hat{u}_h^{(\varepsilon)}$ in place of $u_h^{(\varepsilon)}$ and we can finish the proof of (5.7) in exactly the same way as the proof of Theorem 2.3.

Now we prove that

$$\hat{v}_h \leq v + Ne^{NT}h^{\delta_0^2/(2\delta_0+1)}. \tag{5.8}$$

We are going to use the following lemma which is a particular case of Theorem 2.1 in [7].

Lemma 5.11. *For any $\varepsilon \in (0, 1]$ there exists a function u defined in $[0, T + \varepsilon^2] \times \mathbb{R}^d$ such that*

$$|u(t, x) - g(x)| \leq N\varepsilon \quad for \quad t \in [T, T + \varepsilon^2]; \tag{5.9}$$

$$\sup_{\alpha \in A}[L^\alpha u + f^\alpha] \leq 0, \quad |u - v| \leq Ne^{NT}\varepsilon^{\delta_0} \quad in \quad [0, T] \times \mathbb{R}^d, \tag{5.10}$$

$$|u|_{2+\delta_0,[0,T+\varepsilon^2]} \leq Ne^{NT}\varepsilon^{-1-\delta_0}. \tag{5.11}$$

Now, again take $\varepsilon \ (\geq h)$ from (4.2) and u from Lemma 5.11. Observe that by (5.10) and Lemma 5.10 we have

$$\hat{G}^\alpha_{s,s+h^2}u(s+h^2, \cdot) - u(s, \cdot) \leq N|u|_{2+\delta_0,[0,T-h^2+\varepsilon^2]}h^{2+\delta_0} \leq Ne^{NT}h^{2+\delta_0}\varepsilon^{-1-\delta_0}$$

in $[0, T] \times \mathbb{R}^d$. In addition, if $s \in [T - h^2, T]$, then by (5.9) and (5.11)

$$u(s, x) \geq u(T, x) - Ne^{NT}h^2\varepsilon^{-1-\delta_0} \geq g(x) - Ne^{NT}h^{\delta_0}\varepsilon^{-1-\delta_0} - N\varepsilon.$$

From here and from (5.5), it follows easily by induction that $\hat{v}_h \leq u + N\varepsilon + Ne^{NT}h^{\delta_0}\varepsilon^{-1-\delta_0}$. This and the second equation in (5.10) leads us to $\hat{v}_h \leq v + Ne^{NT}(h^{\delta_0}\varepsilon^{-1-\delta_0} + \varepsilon^{\delta_0})$ and this is exactly (5.8). The theorem is proved.

6. Proof of Lemma 5.8

We need the following counterpart of Lemma 3.1.

Lemma 6.1. *Take $t, s, r \in [0, S]$, $x, y \in \mathbb{R}^d$, and $\beta = (\alpha, \tau, \xi) \in \hat{\mathfrak{B}}_h$. Then*

$$E_{s,x}^{\beta} |\hat{x}_{\kappa_h(t)} - \hat{x}_{\kappa_h(r)}|^2 \leq N e^{NT} |t - r|, \tag{6.1}$$

$$\sup_{t \leq S} E |\hat{x}_{\kappa_h(t)}^{\beta,s,x} - \hat{x}_{\kappa_h(t)}^{\beta,s,y}|^2 \leq N e^{NT} |x - y|^2, \tag{6.2}$$

$$\sup_{t \leq S} E |\hat{x}_{\kappa_h(t)}^{\beta,s,x} - \hat{x}_{\kappa_h(t)}^{\beta,r,x}|^2 \leq N e^{NT} |s - r|^{\delta_0}, \tag{6.3}$$

$$\sup_{t \leq S} E |\hat{x}_{\kappa_h(t)}^{\beta,s,x} - \hat{x}_{\kappa_h(t)}^{\alpha,s,x}|^2 \leq N e^{NT} \varepsilon^{2\delta_0}. \tag{6.4}$$

Proof. These inequalities are absolutely standard and may be claimed to be well known. Say, (6.1) and (6.2) appeared probably for the first time as Lemmas 2.2 and 2.3 in [2], where the theory of stochastic differential equations is applied to proving the solvability of the Cauchy problem for degenerate parabolic equations (a new result at that time). For completeness we outline the proofs of (6.3) and (6.4) following [2].

We have

$$I_{n+1} := E |\hat{x}_{(n+1)h^2}^{\beta,s,x} - \hat{x}_{(n+1)h^2}^{\beta,r,x}|^2 = I_n$$

$$+ 2h^2 E[\hat{x}_{nh^2}^{\beta,s,x} - \hat{x}_{nh^2}^{\beta,r,x}][b(u) - b(v)] + E|b(u) - b(v)|^2 h^4$$

$$+ E|(\sigma(u) - \sigma(v))\hat{w}_n|^2 := I_n + J_1 + J_2 + J_3,$$

where

$$u = (\alpha_{nh^2}, s + nh^2 + \varepsilon^2 \tau_{nh^2}, \hat{x}_{nh^2}^{\beta,s,x} + \varepsilon \xi_{nh^2}),$$

$$v = (\alpha_{nh^2}, r + nh^2 + \varepsilon^2 \tau_{nh^2}, \hat{x}_{nh^2}^{\beta,r,x} + \varepsilon \xi_{nh^2}).$$

It follows by Hölder's inequality and by our assumptions that

$$J_1 \leq 2I_n^{1/2} K h^2 \big(E(|s - r|^{\delta_0/2} + |\hat{x}_{nh^2}^{\beta,s,x} - \hat{x}_{nh^2}^{\beta,r,x}|)^2 \big)^{1/2}$$

$$\leq N I_n^{1/2} h^2 (|s - r|^{\delta_0/2} + I_n)^{1/2} \leq N h^2 (|s - r|^{\delta_0} + I_n).$$

Also

$$J_2 \leq K^2 h^4 E(|s - r|^{\delta_0/2} + |\hat{x}_{nh^2}^{\beta,s,x} - \hat{x}_{nh^2}^{\beta,r,x}|)^2 \leq N h^2 (|s - r|^{\delta_0} + I_n),$$

$$J_3 \leq N h^2 E(|s - r|^{\delta_0/2} + |\hat{x}_{nh^2}^{\beta,s,x} - \hat{x}_{nh^2}^{\beta,r,x}|)^2 \leq N h^2 (|s - r|^{\delta_0} + I_n).$$

Hence

$$I_{n+1} \leq (1 + N h^2) I_n + N h^2 |s - r|^{\delta_0},$$

$$I_{n+1} \leq (1 + N h^2)^{n+1} I_0 + N h^2 |s - r|^{\delta_0} \sum_{i=0}^{n} (1 + N h^2)^i.$$

Since $I_0 = 0$ and $(1 + Nh^2)^{n+1} \leq Ne^{NT}$ for $nh^2 \leq S$, we obtain (6.3).

To prove (6.4) we proceed similarly defining this time

$$I_n = E|\hat{x}_{nh^2}^{\beta,s,x} - \hat{x}_{nh^2}^{\alpha,s,x}|^2$$

and noticing that

$$I_{n+1} = I_n + 2h^2 E[\hat{x}_{nh^2}^{\beta,s,x} - \hat{x}_{nh^2}^{\alpha,s,x}][b(u) - b(v)] + E|b(u) - b(v)|^2 h^4$$
$$+ E|(\sigma(u) - \sigma(v))\hat{w}_n|^2,$$

where

$$u = (\alpha_{nh^2}, s + nh^2 + \varepsilon^2 \tau_{nh^2}, \hat{x}_{nh^2}^{\beta,s,x} + \varepsilon \xi_{nh^2}), \ v = (\alpha_{nh^2}, s + nh^2, \hat{x}_{nh^2}^{\alpha,s,x}).$$

After that an obvious modification of the above argument leads us to (6.4) thus proving the lemma.

On the basis of this lemma and Lemma 5.9, Lemma 5.8 is proved by repeating the proofs of Corollaries 3.2 and 3.4.

Acknowledgment

References

[1] M. Bardi and I. Cappuzo-Dolcetta (1997), Optimal control and viscosity solutions of Hamilton-Jacobi-Bellman equations, Birkhäuser, Boston.

[2] I.I. Gihman, *Certain differential equations with random functions*, Ukrainski Mat. Zh., **2**, 4 (1950), 37-63 and **3**, 3 (1951), 317-339 (in Russian).

[3] W. Fleming and M. Soner (1993), Controlled Markov processes and viscosity solutions, Springer.

[4] N.V. Krylov (1977), Controlled diffusion processes, Nauka, Moscow, in Russian; English translation: Springer, 1980.

[5] N.V. Krylov (1997), *On the rate of convergence of finite–difference approximations for Bellman's equations*, Алгебра и Анализ, St. Petersburg Math. J., **9**, 3, 245-256.

[6] N.V. Krylov, *Mean value theorems for stochastic integrals*, submitted to the Annals of Probability.

[7] N.V. Krylov, *On the rate of convergence of finite–difference approximations for Bellman's equations with variable coefficients*, submitted to Probability Theory and Related Fields.

127 Vincent Hall, University of Minnesota, Minneapolis, MN, 55455
E-mail address: krylov@math.umn.edu