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# ON THE CONVERGENCE OF AN APPROXIMATION SCHEME FOR THE VISCOSITY SOLUTIONS OF THE BELLMAN EQUATION ARISING IN A STOCHASTIC OPTIMAL CONTROL PROBLEM<sup>\*</sup>

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#### **TEODOR HAVÂRNEANU**

**Abstract.** The paper presents an Euler step approximation scheme for the viscosity solution of the dynamic programming equation for a stochastic control problem with infinite horizon. We established the rate of convergence of this scheme.

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### 1. Introduction and preliminary results

At the beginning of '80, using the viscosity solutions theory, several approximation results for the Bellman equation in the deterministic case were obtained (see [1],[2],[4],[5]). In [6] we proposed an Euler step approximation scheme (similar to those from [1],[3],[5]) for the viscosity solution of the Bellman equation of a stochastic optimal control problem and we proved its convergence. In the present paper we give a result concerning the rate of convergence for the approximation scheme presented in [6] and also we improve the convergence result there.

Let us consider the following stationary second order Bellman partial differential equation

(1.1) 
$$\lambda u(x) + \sup_{a \in A} \left[ -\nabla u(x) \cdot b(x, a) - f(x, a) \right] - \frac{\varepsilon}{2} \Delta u(x) = 0, \ x \in \mathbb{R}^n,$$

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where  $u(\cdot)$  is the unknown function,  $A \subset \mathbb{R}^m$  is a compact set,  $b : \mathbb{R}^n \times A \to \mathbb{R}^n$  and  $f : \mathbb{R}^n \times A \to \mathbb{R}^n$  are two given functions,  $\lambda > 0$  and  $\varepsilon > 0$  are two given constants, and "." stands for the inner product in  $(\mathbb{R}^n, \|\cdot\|)$ .

We suppose that the functions b and f satisfy the following hypotheses: There exist the positive constants  $L_1 \in (0, \lambda)$ ,  $L_2$ , and M such that

(i) 
$$||b(x,a) - b(y,a)|| \le L_1 ||x - y||, x, y \in \mathbb{R}^n, a \in A$$

(H) (ii) 
$$|f(x,a) - f(y,a)| \le L_2 ||x - y||, x, y \in \mathbb{R}^n, a \in A$$
  
(iiii)  $||b(x,a)|| \le M, |f(x,a)| \le M, x \in \mathbb{R}^n, a \in A.$ 

In [6] we have proved that equation (1.1) has a unique viscosity solution which is the value functions of the following stochastic control problem with infinite horizon.

Let  $(\Omega, \mathcal{F}, P, (\mathcal{F}_t)_{t\geq 0})$  be a fixed complete stochastic basis which means that  $(\Omega, \mathcal{F}, P)$  is a complete probability space and  $(\mathcal{F}_t)_{t\geq 0}$  is an increasing family of sub- $\sigma$ -algebras of  $\mathcal{F}$  such that  $\mathcal{F}_0$  contains every P-negligible subset of  $\Omega$ . Also we consider a standard n-dimensional  $\mathcal{F}_t$ -adapted Wiener process  $\{w(t), t \geq 0\}$  and

(1.2) 
$$\mathcal{A} = \{a : [0, +\infty) \times \Omega \to A; a(\cdot, \cdot) \text{ is measurable}, and a(t, \cdot) \text{ is } \mathcal{F}_t - \text{measurable}, t \ge 0\}.$$

**Remark 1.1.** In the following we shall consider that the  $\sigma$ -algebra  $\mathcal{F}_t$  is generated by  $\{w(s), 0 \leq s \leq t\}$  for  $t \geq 0$ .

Now we present the optimal control problem: Minimize

(1.3) 
$$J(x,a) = E\left[\int_0^\infty f(x(t),a(t))e^{-\lambda t}dt\right],$$

where E stands for the average,  $a \in \mathcal{A}$  and (x(t), a(t)) satisfies the state equation

(1.4) 
$$\begin{cases} dx(t) = b(x(t), a(t))dt + \sqrt{\varepsilon}dw(t), \ \varepsilon > 0\\ x(0) = x \in \mathbb{R}^n. \end{cases}$$

We define

(1.5) 
$$J(x) = \inf_{a \in \mathcal{A}} J(x, a), \ x \in \mathbb{R}^n$$

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**Definition 1.1** ([7]). The function  $u \in C(\mathbb{R}^n)$  is called viscosity subsolution (supersolution) of (1.1) iff for every  $\varphi \in C^2(\mathbb{R}^n)$  for which  $(u - \varphi)$ has a (strict) local maximum (minimum) in  $x_0$  we have

$$\lambda u(x_0) + \sup_{a \in A} \left[ -\nabla \varphi(x_0) \cdot b(x_0, a) - f(x_0, a) \right] - \frac{\varepsilon}{2} \Delta \varphi(x_0) \le 0 \quad (\ge 0).$$

A function  $u(\cdot)$  which is at the same time a viscosity subsolution and a viscosity supersolution is called viscosity solution.

In the next we present the approximation scheme ([6]).

Let h > 0 and  $t_j = jh$ , j = 0, 1, 2, ... Corresponding to each h > 0 we consider the set of discrete controls:

$$\mathcal{A}_{h} = \{ a \in \mathcal{A}; \ a \text{ is constant on } [t_{j}, t_{j+1}), \ j = 0, 1, 2, \dots, \text{ i.e.}$$

$$(1.6) \qquad a(s) = a_{j}(\cdot) \in L^{2}(\Omega, A), \ \text{ for } s \in [t_{j}, t_{j+1}), \ j = 0, 1, 2, \dots \}.$$

Now we define recursively the sequence  $\{x_j\}_{j\geq 0}$  (the approximation of (1.4))

(1.7) 
$$\begin{cases} x_0 = x \\ x_{j+1} = x_j + hb(x_j, a_j) + \sqrt{\varepsilon}(w(t_{j+1}) - w(t_j)), j \in \mathbb{N} \end{cases}$$

where  $a \in \mathcal{A}_h$   $(a_j = a(s), s \in [t_j, t_{j+1}))$ . Also we consider the following discretization of the functional (1.3)

(1.8) 
$$J_h(x,a) = hE \sum_{j=0}^{\infty} (1 - \lambda h)^j f(x_j, a_j),$$

were  $a \in \mathcal{A}_h$  and  $\{x_j\}_{j\geq 0}$  verifies (1.7).

Now we define the approximate solution of (1.1) as

(1.9) 
$$v_h(x) = \inf_{a \in \mathcal{A}_h} J_h(x, a) = \inf_{a \in \mathcal{A}_h} E \sum_{j=0}^{\infty} (1 - \lambda h)^j f(x_j, a_j).$$

For  $h \in (0; \frac{1}{\lambda}]$  we consider the following approximate equation of (1.1)

(1.10) 
$$u_h(x) + \sup_{a \in A} E[-(1-\lambda h)u_h(x+hb(x,a) + \sqrt{\varepsilon}w(h)) + f(x,a)] = 0,$$
$$x \in \mathbb{R}^n.$$

In the next we shall prove a dynamic programming principle for the discrete optimal control problem (1.9), (1.7).

**Theorem 1.1.** For each  $x \in \mathbb{R}^n$  we have

(1.11) 
$$v_h(x) = \inf_{a_0} E[hf(x, a_0) + (1 - \lambda h)v_h(x_1)],$$

where  $a_0 \in L^2(\Omega, A)$  is  $\mathcal{F}_0$ -measurable and  $x_1$  is given by (1.7) with  $x_0 = x$ .

**Proof.** Using (1.9) and (1.8) we obtain that for each  $\nu > 0$  there exists  $a^{\nu} \in \mathcal{A}_h$  such that

(1.12) 
$$v_h(x) + \nu \ge J_h(x, a^{\nu}) = hEf(x, a^{\nu}_0) + (1 - \lambda h)hE$$
$$\sum_{j=0}^{\infty} (1 - \lambda h)^j f(x^{\nu}_{j+1}, a^{\nu}_{j+1}) \ge hEf(x, a^{\nu}_0) + (1 - \lambda h)Ev_h(x^{\nu}_1),$$

where  $\{x_j^{\nu}\}_{j\in\mathbb{N}}$  and  $\{a_j^{\nu}\}_{j\in\mathbb{N}}$  verify (1.7). Taking the infimum in (1.12) we obtain

(1.13) 
$$v_h(x) + \nu \ge \inf_{a \in \mathcal{A}_h} E[hEf(x, a_0) + (1 - \lambda h)v_h(x_1)],$$

for every  $\nu > 0$ . For every  $\nu > 0$  there exist  $\tilde{a}^{\nu}, \hat{a}^{\nu} \in \mathcal{A}_h$  such that

(1.14)  

$$\inf_{a \in \mathcal{A}_{h}} E\left[hf(x,a) + (1-\lambda h)v_{h}(x_{1})\right] + \nu$$

$$\geq E\left[hf(x,\tilde{a}_{0}^{\nu}) + (1-\lambda h)v_{h}(\tilde{x}_{1}^{\nu})\right]$$

$$\geq E\left[hf(x,\tilde{a}_{0}^{\nu}) + (1-\lambda h)h\sum_{j=0}^{\infty}(1-\lambda h)^{j}f(\tilde{x}_{j}^{\nu},\tilde{a}_{j}^{\nu})\right] - \nu,$$

where  $\{\widetilde{x}_{j}^{\nu}\}_{j\in\mathbb{N}}, \{\widetilde{a}_{j}^{\nu}\}_{j\in\mathbb{N}}$  satisfy (1.7) with  $\widetilde{x}_{0}^{\nu} = x$ , and  $\{\widehat{x}_{j}^{\nu}\}_{j\in\mathbb{N}}, \{\widehat{a}_{j}^{\nu}\}_{j\in\mathbb{N}}$ satisfy (1.7) with  $\widehat{x}_{0}^{\nu} = \widetilde{x}_{1}^{\nu}$ . Let  $\overline{a}^{\nu} = \{\overline{a}_{j}^{\nu}\}_{j\in\mathbb{N}}$  and  $\overline{x}^{\nu} = \{\overline{x}_{j}^{\nu}\}_{j\in\mathbb{N}}$  defined by

(1.15) 
$$\begin{cases} \overline{a}_{0}^{\nu} = \widetilde{a}_{0}^{\nu}, \\ \overline{a}_{j}^{\nu} = \widehat{a}_{j-1}^{\nu}, \quad j \ge 1 \end{cases}, \qquad \begin{cases} \overline{x}_{0}^{\nu} = \widetilde{x}_{0}^{\nu} = x, \\ \overline{x}_{j}^{\nu} = \widehat{x}_{j-1}^{\nu}, \quad j \ge 1 \end{cases}$$

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Taking into account (1.13), (1.14), (1.15) and (1.8) it results

(1.16) 
$$\inf_{a \in \mathcal{A}_h} E[hf(x, a_0) + (1 - \lambda h)v_h(x_1)] + 2\nu \ge J_h(\overline{a}^\nu, x) \ge v_h(x).$$

Because  $\nu > 0$  is arbitrarily chosen, the relations (1.14) and (1.16) imply (1.11).

In the next, for the reader's convenience, we shall present some results (see also [6]) which we need in evaluation of the rate of convergence of our scheme.

**Theorem 1.2** ([6]). In the hypotheses (H) the  $v_h$  given by (1.9) is the unique bounded solution of (1.10).

Sketch of the proof. Using Theorem 1.1 we get that  $v_h$  verifies

(1.17) 
$$v_h(x) + \sup_{a_0} E[-(1-\lambda h)v_h(x+hb(x,a_0)) + \sqrt{\varepsilon}w(h) - hf(x,a_0)] = 0, \quad x \in \mathbb{R}^n,$$

where  $a_0 \in L^2(\Omega, A)$  and  $a_0$  is  $\mathcal{F}_0$ -measurable (see the definitions (1.2) and (1.6)).

Taking into account the fact that  $\mathcal{F}_0$  and  $\mathcal{F}_h$  are independent (see Remark 1.1) we may write the average in (1.14) as a double integral and so we obtain

(1.18) 
$$v_h(x) + \sup_{a_0 \in L^2(\Omega_1, A)} \int_{\Omega_1} \left\{ \int_{\Omega} [-(1 - \lambda h) v_h(x + b(x, a_0(\omega_1)) + \sqrt{\varepsilon}\omega(h)(\omega)) - hf(x, a_0(\omega_1))] dP(\omega) \right\} dP(\omega_1) = 0,$$

where  $\Omega_1$  is  $\Omega$  but we put subscript 1 to be more clear. Using boundness of  $v_h$  (which is evident from (1.8)), from the relation (1.18) we obtain that  $v_h$  is a solution of (1.10).

To prove the uniqueness, we suppose that there exist two solutions  $u_1$ and  $u_2$  of (1.10) and subtracting the two relations (1.10) which they verify, using their boundness and some calculus, it results

$$|u_1(x) - u_2(x)| \le (1 - \lambda h) \sup_{y \in \mathbb{R}^n} |u_1(y) - u_2(y)|, x \in \mathbb{R}^n,$$

relation which proves the uniqueness because  $0 < h \leq \frac{1}{\lambda}$ .

**Theorem 1.3** ([6]). In the hypotheses (H) the function  $v_h(\cdot)$  converges to  $J(\cdot)$  locally uniformly on  $\mathbb{R}^n$  as h tends to O.

Sketch of the proof. Using (1.9) and the hypotheses on b, f, and  $\lambda$ , it results

(1.19) 
$$\sup_{x \in \mathbb{R}^n} |v_h(x)| \le \frac{M}{\lambda} \\ |v_h(x) - v_h(x')| \le \frac{L_2}{\lambda - L_1} ||x - x'||, \text{ for } x, x' \in \mathbb{R}^n \text{ and } h \in (0, \frac{1}{\lambda}].$$

From the Ascoli–Arzela theorem we obtain that there exists a bounded Lipschitz continuous function  $u(\cdot)$  such that  $v_h(\cdot)$  converges to  $u(\cdot)$  locally uniformly on  $\mathbb{R}^n$  as h tends to 0.

In the following we shall prove that  $u(\cdot)$  is a viscosity subsolution of (1.1) (in the same manner one can prove that it is also a viscosity supersolution). We shall present this proof in more details because we consider that it is interesting and useful for the reader.

Let  $\varphi \in C_b^2(\mathbb{R}^n)$  (a bounded function of class  $C^2$ ),  $x_0$  a strict local maximum point of  $(u - \varphi)$ ,  $\overline{B}(x_0, r) = \{x \in \mathbb{R}^n \mid ||x - x_0|| \leq r\}$  and  $x_0^h$ a global maximum point of  $(v_h - \varphi)$  on  $\overline{B}(x_0, r)$ . The sequence  $\{x_0^h\}_{h>0}$ converges to  $x_0$  as h tends to 0 (see [3]) and so we can choose  $h_1 > 0$  such that for every  $h \in (0, h_1)$  we have

(1.20) 
$$||x_0^h - x_0|| \le \frac{r}{3}, \ hM \le \frac{r}{3}.$$

Let

(1.21) 
$$\Omega_r = \left\{ \omega \in \Omega; \ \|w(h)(\omega)\| \le \frac{r}{3} \right\}.$$

It is well-known (see [2]) that

(1.22) 
$$P(\Omega \setminus \Omega_r) \le \frac{E \|w(h)\|^4}{\frac{r^4}{81}} \le c \frac{h^2}{r^4},$$

where c > 0 is a constant independent of h and r.

Using now (1.10), (1.19), (1.20) and the fact that  $x_0^h$  is a maximum point

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of  $(v_h - \varphi)$  on  $\overline{B}(x_0, r)$  we obtain

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$$\begin{split} 0 &= v_h(x_0^h) + \sup_{a \in A} \{ \int_{\Omega_r} [-(1 - \lambda h) v_h(x_0^h + hb(x_0^h, a) + \sqrt{\varepsilon}w(h)) \\ &- hf(x_0^h, a)] dP + \int_{\Omega \setminus \Omega_r} [-(1 - \lambda h) v_h(x_0^h + hb(x_0^h, a) + \sqrt{\varepsilon}w(h)) \\ (1.23) &- hf(x_0^h, a)] dP \geq \sup_{a \in A} \{ \int_{\Omega_r} [\varphi(x_0^h) - \varphi(x_0^h + hb(x_0^h, a) \\ &+ \sqrt{\varepsilon}w(h)) + \lambda hv_h(x_0^h + hb(x_0^h, a) + \sqrt{\varepsilon}w(h))] dP \\ &+ \int_{\Omega \setminus \Omega_r} [v_h(x_0^h) - (1 - \lambda h) v_h(x_0^h + hb(x_0^h, a) \\ &+ \sqrt{\varepsilon}w(h))] dP - hf(x_0^h, a) \}. \end{split}$$

Taking into account (1.21) and the boundness of  $\varphi$  and  $v_h$ , the relation (1.22) implies

(1.24) 
$$0 \ge \sup_{a \in A} E[\varphi(x_0^h) - \varphi(x_0^h + hb(x_0^h, a) + \sqrt{\varepsilon}w(h)) \\ + \lambda hv_h(x_0^h + hb(x_0^h, a) + \sqrt{\varepsilon}w(h)) - hf(x_0^h, a)] - c_1 \frac{h^2}{r^4},$$

where  $c_1 > 0$  is a constant independent of h. Because  $\varphi \in C^2(\mathbb{R}^n)$ , using Itô formula, the relation (1.23) gives

$$0 \ge \sup_{a \in A} E\{-\int_0^h [\nabla \varphi(x_0^h + sb(x_0^h, a) + \sqrt{\varepsilon}w(s)) \cdot b(x_0^h, a)$$

$$(1.25) \qquad -\frac{\varepsilon}{2} \Delta \varphi(x_0^h + sb(x_0^h, a) + \sqrt{\varepsilon}w(s))]ds - \int_0^h \nabla \varphi(x_0^h + sb(x_0^h, a)$$

$$+ \sqrt{\varepsilon}w(s))dw(s) + \lambda hv_h(x_0^h + hb(x_0^h, a) + \sqrt{\varepsilon}w(h))$$

$$- hf(x_0^h, a)\} - c_1 \frac{h^2}{r^4}.$$

Dividing (1.24) by h and using the fact that  $\varphi \in C^2(\mathbb{R}^n)$ ,  $v_h \to u$  (locally uniformly as  $h \to 0$ ), and  $x_0^h \to x_0$  as  $h \to 0$ , it results

$$\lambda u(x_0) + \sup_{a \in A} \left[ -\nabla \varphi(x_0) \cdot b(x_0, a) - f(x_0, a) \right] - \frac{\varepsilon}{2} \Delta \varphi(t_0) \le 0,$$

i.e.  $u(\cdot)$  is a viscosity subsolution of (1.1) (see Definition (1.1)).

In a similar way one can prove that  $u(\cdot)$  is also a viscosity supersolution of (1.1). In our hypotheses on  $\lambda, b, f$  the equation (1.1) has an unique viscosity solution given by  $J(\cdot)$  (see [7]). So  $u(\cdot) = J(\cdot)$  and Theorem 1.3 is proved.

## 2. The rate of convergence of the approximation scheme

In this section we shall prove a result with respect to the rate of convergence of the approximate solutions  $v_h(\cdot)$  to the viscosity solution  $J(\cdot)$  of equation (1.1) and also we shall improve the result in Theorem 1.3. Such results for the deterministic case were given in [1], [3], [5].

**Theorem 2.1.** In the hypotheses (H) we have

$$\sup_{x \in \mathbb{R}^n} |J(x) - v_h(x)| \le ch, \ 0 < h \le \frac{1}{\lambda},$$

where c > 0 is a constant independent of h.

**Proof.** Let  $\delta, \rho \in (0, 1)$  be fixed. We define the auxiliary function

(2.1) 
$$\varphi(x,y) = v_h(x) - J(y) - \frac{\delta ||x-y||^2}{\rho^2}, \ (x,y) \in \mathbb{R}^n \times \mathbb{R}^n, 0 < h \le \frac{1}{\lambda}.$$

Using the fact that  $v_h$  and J are bounded (see (1.19) and (1.3)) it results that for each  $\gamma \in (0, 1)$  there exists  $(x_1, y_1) \in \mathbb{R}^n \times \mathbb{R}^n$  such that

(2.2) 
$$\varphi(x_1, y_1) > \sup_{(x,y) \in \mathbb{R}^n \times \mathbb{R}^n} \varphi(x, y) - \gamma.$$

Let's consider a smooth function  $\xi \in C_0^{\infty}(\mathbb{R}^n \times \mathbb{R}^n)$  which satisfies

(2.3) 
$$\xi(x_1, y_1) = 1, \ 0 \le \xi \le 1, \ \|\nabla \xi\| \le 1.$$

Define

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(2.4) 
$$\psi(x,y) = \varphi(x,y) + \gamma \xi(x,y), \ (x,y) \in \mathbb{R}^n \times \mathbb{R}^n.$$

From (2.2) and (2.4) we obtain that there exists  $(x_0, y_0) \in \mathbb{R}^n \times \mathbb{R}^n$  such that

(2.5) 
$$\psi(x_0, y_0) \ge \psi(x, y), \ (x, y) \in \mathbb{R}^n \times \mathbb{R}^n.$$

So, the function  $-\psi(x_0, y) = J(y) - (v_h(x_0) - \delta \frac{\|x_0 - y\|^2}{\rho^2} + \gamma \xi(x_0, y))$  has a global minimum point in  $y_0$ . Using the fact that  $J(\cdot)$  is a viscosity supersolution of (1.1) it results (see Definition 1.1)

(2.6) 
$$\lambda J(y_0) + \sup_{a \in A} \left[ -\frac{2\delta}{\rho^2} (x_0 - y_0) \cdot b(y_0, a) - \gamma (\nabla_y \xi(x_0, y_0)) \cdot b(y_0, a) - f(y_0, a) \right] + \varepsilon \frac{\delta}{\rho^2} - \frac{\varepsilon}{2} \gamma \Delta_y \xi(x_0, y_0) \ge 0.$$

The set A being compact it results that there exists  $a_0 \in A$  such that (see (2.6))

(2.7) 
$$\lambda J(y_0) - \frac{2\delta}{\rho^2} (x_0 - y_0) \cdot b(y_0, a_0) - \gamma (\nabla_y \xi(x_0, y_0)) \cdot b(y_0, a_0) - f(y_0, a_0) + \varepsilon \frac{\delta}{\rho^2} - \frac{\varepsilon}{2} \gamma \Delta_y \xi(x_0, y_0) \ge 0.$$

The function  $v_h(\cdot)$  verifies the relation (1.17) and so we have

(2.8) 
$$v_h(x_0) - (1 - \lambda h) E v_h(x_0 + hb(x_0, a_0) + \sqrt{\varepsilon} w(h)) - hf(x_0, a_0) \le 0.$$

Using (2.5) we obtain

$$\begin{aligned} v_h(x_0) - J(y_0) - \frac{\delta \|x_0 - y_0\|^2}{\rho^2} + \gamma \xi(x_0, y_0) \\ &\geq E[v_h(x_0 + hb(x_0, a_0) + \sqrt{\varepsilon}w(h)) \\ &- \frac{\delta \|x_0 + hb(x_0, a_0) + \sqrt{\varepsilon}w(h) - y_0\|^2}{\rho^2} \\ &+ \gamma \xi(x_0 + hb(x_0, a_0) + \sqrt{\varepsilon}(h), y_0)] - J(y_0), \end{aligned}$$

i.e.

(2.9)  

$$Ev_{h}(x_{0} + hb(x_{0}, a_{0}) + \sqrt{\varepsilon}w(h))$$

$$\leq v_{h}(x_{0}) + \gamma\xi(x_{0}, y_{0}) - \frac{\delta ||x_{0} - y_{0}||^{2}}{\rho^{2}}$$

$$+ \frac{\delta}{\rho^{2}}E||x_{0} - y_{0} + hb(x_{0}, a_{0}) + \sqrt{\varepsilon}w(h)||^{2}$$

$$- \gamma E\xi(x_{0} + hb(x_{0}, a_{0}) + \sqrt{\varepsilon}w(h), y_{0}).$$

The relations (2.8) and (2.9) imply

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$$\lambda h v_h(x_0) - (1 - \lambda h) \gamma \xi(x_0, y_0) + \frac{(1 - \lambda h)\delta}{\rho^2} \|x_0 - y_0\|^2$$

$$(2.10) \quad -\frac{(1 - \lambda h)\delta}{\rho^2} E \|x_0 - y_0 + hb(x_0, a_0) + \sqrt{\varepsilon}w(h)\|^2$$

$$+ (1 - \lambda h)\delta E \xi(x_0 + hb(x_0, a_0) + \sqrt{\varepsilon}w(h), y_0) - hf(x_0, a_0) \le 0.$$

Dividing (2.10) by h we obtain

$$\lambda v_h(x_0) - \frac{1 - \lambda h}{h} \gamma \xi(x_0, y_0) + \frac{(1 - \lambda h)\delta}{h\rho^2} ||x_0 - y_0||^2$$

$$(2.11) - \frac{(1 - \lambda h)\delta}{h\rho^2} E ||x_0 - y_0 + hb(x_0, a_0) + \sqrt{\varepsilon}\omega(h)||^2$$

$$+ \frac{(1 - \lambda h)\delta}{h} E \xi(x_0 + hb(x_0, a_0) + \sqrt{\varepsilon}w(h), y_0) - f(x_0, a_0) \le 0.$$

Using the hypotheses (H) and also the properties of  $\xi$  and w(h) (Ew(h) = 0,  $E||w(h)||^2 = h$ ), the inequality (2.11) implies

(2.12) 
$$\lambda v_h(x_0) - \frac{\gamma}{h} - \frac{(1-\lambda h)\delta h M^2}{\rho^2} - \frac{(2-\lambda h)\delta}{\rho^2}\varepsilon - \frac{2\delta}{\rho^2}(x_0 - y_0) \cdot b(x_0, a_0) - \frac{2\lambda h\delta}{\rho^2}M \|x_0 - y_0\| - \frac{h\delta}{\rho^2} - f(x_0, a_0) \le 0.$$

Subtracting (2.7) from (2.12) it results

(2.13) 
$$\lambda(v_h(x_0) - J(y_0)) + \gamma(\nabla_y \xi(x_0, y_0))b(y_0, a_0) - \frac{\gamma}{h} \\ - \frac{(1 - \lambda h)}{\rho^2} \delta h M^2 - \frac{(3 - \lambda h)\delta}{\rho^2} \varepsilon - \frac{2\lambda h\delta}{\rho^2} M \|x_0 - y_0\| \\ - \frac{h\delta}{\rho^2} + \frac{\varepsilon}{2} \gamma \Delta_y \xi(x_0, y_0) \le 0.$$

Making  $x = y = x_0$  in relation (2.5) and using the properties of  $v_h$  and  $\xi$  (see (1.19) and (2.3)) it results

(2.14) 
$$\delta \|x_0 - y_0\| \le \left(\gamma + \frac{L_2}{\lambda - L_1}\right) \rho^2.$$

Taking again into account hypotheses (H) and (2.3), the relations (2.13) and (2.14) imply

(2.15) 
$$v_h(x_0) - J(y_0) \le c \left(\gamma + \frac{\gamma}{h} + \frac{h\delta}{\rho^2} + \frac{\delta}{\rho^2} + h\right),$$

where c > 0 is a constant independent of  $h, \rho, \gamma$ , and  $\delta$ .

Now taking  $\gamma = h^2, \delta = h^3, \rho = h$ , the relation (2.15) gives

(2.16) 
$$v_h(x_0) - J(y_0) \le ch \text{ for } 0 < h \le \frac{1}{\lambda}.$$

Putting in the relation (2.5) y = x, it results

$$v_h(x) - J(x) + \gamma \xi(x, x) \le v_h(x_0) - J(y_0) - \frac{\delta ||x_0 - y_0||^2}{\rho^2} + \gamma \xi(x_0, y_0)$$

from where, using (2.16), we obtain

(2.17) 
$$v_h(x) - J(x) \le ch, \ x \in \mathbb{R}^n, 0 < h \le \frac{1}{\lambda},$$

where c > 0 is a constant independent of h.

To prove the opposite inequality we consider the auxiliary function

$$\widetilde{\varphi}(x,y) = J(y) - v_h(x) + \frac{\delta ||x-y||^2}{\rho^2}$$

which is bounded from below.

Using, this time, the fact that  $J(\cdot)$  is a viscosity subsolution of (1.1) and making a similar calculus as before we obtain the desired result.

**Remark 2.1.** Theorem 2.1 established that the approximate solution  $v_h(\cdot)$  converges globally uniformly on  $\mathbb{R}^n$  to the viscosity solution  $J(\cdot)$  and so we have obtained an improvement of the result in Theorem 1.3 (see [6]).

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Department of Mathematics, University "Al.I. Cuza", 11, Bd. Carol I, 700506, Iaşi, ROMANIA andInstitute of Mathematics "Octav Mayer", Romanian Academy, Iaşi Branch, ROMANIA havi@uaic.ro