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# Randomized dynamic programming principle and Feynman-Kac representation for optimal control of McKean-Vlasov dynamics

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## Abstract

We analyze a stochastic optimal control problem, where the state process follows a McKean-Vlasov dynamics and the diffusion coefficient can be degenerate. We prove that its value function  $V$  admits a nonlinear Feynman-Kac representation in terms of a class of forward-backward stochastic differential equations, with an autonomous forward process. We exploit this probabilistic representation to rigorously prove the dynamic programming principle (DPP) for  $V$ . The Feynman-Kac representation we obtain has an important role beyond its intermediary role in obtaining our main result: in fact it would be useful in developing probabilistic numerical schemes for  $V$ . The DPP is important in obtaining a characterization of the value function as a solution of a non-linear partial differential equation (the so-called Hamilton-Jacobi-Bellman equation), in this case on the Wasserstein space of measures. We should note that the usual way of solving these equations is through the Pontryagin maximum principle, which requires some convexity assumptions. There were attempts in using the dynamic programming approach before, but these works assumed a priori that the controls were of Markovian feedback type, which helps write the problem only in terms of the distribution of the state process (and the control problem becomes a deterministic problem). In this paper, we will consider open-loop controls and derive the dynamic programming principle in this most general case. In order to obtain the Feynman-Kac representation and the randomized dynamic programming principle, we implement the so-called randomization method, which consists in formulating a new McKean-Vlasov control problem, expressed in weak form taking the supremum over a family of equivalent probability measures. One of the main results of the paper is the proof that this latter control problem has the same value function  $V$  of the original control problem.

**Keywords:** Controlled McKean-Vlasov stochastic differential equations, dynamic programming principle, randomization method, forward-backward stochastic differential equations.

**AMS 2010 subject classification:** 49L20, 93E20, 60K35, 60H10, 60H30.

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# 1 Introduction

In the present paper we study a stochastic optimal control problem of McKean-Vlasov type. More precisely, let  $T > 0$  be a finite time horizon,  $(\Omega, \mathcal{F}, \mathbb{P})$  a complete probability space,  $B = (B_t)_{t \geq 0}$  a  $d$ -dimensional Brownian motion defined on  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  $\mathbb{F}^B = (\mathcal{F}_t^B)_{t \geq 0}$  the  $\mathbb{P}$ -completion of the filtration generated by  $B$ , and  $\mathcal{G}$  a sub- $\sigma$ -algebra of  $\mathcal{F}$  independent of  $B$ . Let also  $\mathcal{P}_2(\mathbb{R}^n)$  denote the set of all probability measures on  $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  with a finite second-order moment. We endow  $\mathcal{P}_2(\mathbb{R}^n)$  with the 2-Wasserstein metric  $\mathcal{W}_2$ , and assume that  $\mathcal{G}$  is rich enough in the sense that  $\mathcal{P}_2(\mathbb{R}^n) = \{\mathbb{P}_\xi : \xi \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)\}$ , where  $\mathbb{P}_\xi$  denotes the law of  $\xi$  under  $\mathbb{P}$ . Then, the controlled state equations are given by

$$X_s^{t,\xi,\alpha} = \xi + \int_t^s b(r, X_r^{t,\xi,\alpha}, \mathbb{P}_{X_r^{t,\xi,\alpha}}, \alpha_r) dr + \int_t^s \sigma(r, X_r^{t,\xi,\alpha}, \mathbb{P}_{X_r^{t,\xi,\alpha}}, \alpha_r) dB_r, \quad (1.1)$$

$$X_s^{t,x,\xi,\alpha} = x + \int_t^s b(r, X_r^{t,x,\xi,\alpha}, \mathbb{P}_{X_r^{t,\xi,\alpha}}, \alpha_r) dr + \int_t^s \sigma(r, X_r^{t,x,\xi,\alpha}, \mathbb{P}_{X_r^{t,\xi,\alpha}}, \alpha_r) dB_r, \quad (1.2)$$

for all  $s \in [t, T]$ , where  $(t, x, \xi) \in [0, T] \times \mathbb{R}^n \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , and  $\alpha$  is an admissible control process, namely an  $\mathbb{F}^B$ -progressive process  $\alpha: \Omega \times [0, T] \rightarrow A$ , with  $A$  Polish space. We denote by  $\mathcal{A}$  the set of admissible control processes. On the coefficients  $b: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A \rightarrow \mathbb{R}^n$  and  $\sigma: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A \rightarrow \mathbb{R}^{n \times d}$  we impose standard Lipschitz and linear growth conditions, which guarantee existence and uniqueness of a pair  $(X_s^{t,\xi,\alpha}, X_s^{t,x,\xi,\alpha})_{s \in [t, T]}$  of continuous  $(\mathcal{F}_s^B \vee \mathcal{G})$ -adapted processes solution to equations (1.1)-(1.2). Notice that  $X^{t,x,\xi,\alpha}$  depends on  $\xi$  only through its law  $\pi := \mathbb{P}_\xi$ . Therefore, we define  $X^{t,x,\pi,\alpha} := X^{t,x,\xi,\alpha}$ .

The control problem consists in maximizing over all admissible control processes  $\alpha \in \mathcal{A}$  the following functional

$$J(t, x, \pi, \alpha) = \mathbb{E} \left[ \int_t^T f(s, X_s^{t,x,\pi,\alpha}, \mathbb{P}_{X_s^{t,\xi,\alpha}}, \alpha_s) ds + g(X_T^{t,x,\pi,\alpha}, \mathbb{P}_{X_T^{t,\xi,\alpha}}) \right],$$

for any  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ , where  $f: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A \rightarrow \mathbb{R}$  and  $g: \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  satisfy suitable continuity and growth conditions, see Assumptions **(A1)** and **(A2)**. We define the value function

$$V(t, x, \pi) = \sup_{\alpha \in \mathcal{A}} J(t, x, \pi, \alpha), \quad (1.3)$$

for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ . We will show in Proposition 2.2 that the mapping  $V$  is the disintegration of the value function

$$V_{\text{MKV}}(t, \xi) = \sup_{\alpha \in \mathcal{A}_\xi} \mathbb{E} \left[ \int_t^T f(s, X_s^{t,\xi,\alpha}, \mathbb{P}_{X_s^{t,\xi,\alpha}}^\xi, \alpha_s) ds + g(X_T^{t,\xi,\alpha}, \mathbb{P}_{X_T^{t,\xi,\alpha}}^\xi) \right], \quad (1.4)$$

for every  $(t, \xi) \in [0, T] \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , where  $\mathcal{A}_\xi$  denotes the set of  $A$ -valued  $(\mathcal{F}_s^B \vee \sigma(\xi))$ -progressive processes, and  $\mathbb{P}_{X_s^{t,\xi,\alpha}}^\xi$  denotes the regular conditional distribution of the random variable  $X_s^{t,\xi,\alpha}: \Omega \rightarrow \mathbb{R}^n$  with respect to  $\sigma(\xi)$ . That is,

$$V_{\text{MKV}}(t, \xi) = \int V(t, x, \pi) \pi(dx). \quad (1.5)$$

Notice that at time  $t = 0$ , when  $\xi = x_0$  is a constant, then  $V_{\text{MKV}}(0, x_0)$  is the natural formulation of the McKean-Vlasov control problem as in [13].

Optimal control of McKean-Vlasov dynamics is a new type of stochastic control problem related to, but different from, what is well-known as mean field games (MFG), and which has attracted a surge of interest in the stochastic control community since the lectures by P.L. Lions at Collège de France, see [25] and [10], and the recent books [6] and [11]. Both of these problems describe equilibria states of large population of weakly interacting symmetric players and we refer to [14] for a discussion pointing out the differences between the two frameworks: In a nutshell MFGs describe Nash equilibrium in large populations and the optimal control of McKean-Vlasov dynamics describes the Pareto optimality, as heuristically shown in [14], and recently proved in [23]. As an example we mention the model of systemic risk due to [15], where, using our notation,  $X^{t,\xi,\alpha}$  (as well as the auxiliary process  $X^{t,x,\xi,\alpha}$ ) represents the log-reserve of the representative bank, and  $\alpha$  is the rate of borrowing/lending to a central bank.

In the literature McKean-Vlasov control problem is tackled by two different approaches: On the one hand, the stochastic Pontryagin maximum principle allows one to characterize solutions to the controlled McKean-Vlasov systems in terms of an adjoint backward stochastic differential equation (BSDE) coupled with a forward SDE: see [1], [8] in which the state dynamics depend upon moments of the distribution, and [13] for a deep investigation in a more general setting. On the other hand, the dynamic programming (DP) method (also called Bellman principle), which is known to be a powerful tool for standard Markovian stochastic control problem and does not require any convexity assumption usually imposed in Pontryagin principle, was first used in [24] and [5] for a specific McKean-Vlasov SDE and cost functional, depending only upon statistics like the mean of the distribution of the state variable. These papers assume a priori that the state variables marginals at all times have a density. Recently, [26] managed to drop the density assumption, but still restricted the admissible controls to be of closed-loop (a.k.a. feedback) type, i.e., deterministic and Lipschitz functions of the current value of the state, which is somewhat restrictive. This feedback form on the class of controls allows one to reformulate the McKean-Vlasov control problem (1.4) as a deterministic control problem in an infinite dimensional space with the marginal distribution as the state variable. In this paper we will consider the most general case and allow the controls to be open-loop. In this case reformulation mentioned above is no more possible. We will instead work with a proper disintegration of the value function, which we described in (1.4). The disintegration formula (1.5) was pointed out heuristically in [12], see their formulae (40) and (41), but the value function  $V$  was not identified. The idea of formulating the McKean-Vlasov control problem as in (1.3) (rather than as in (1.4)) is inspired by [9], where the uncontrolled case is addressed. We will then generalize the randomization approach developed by [21] to the McKean-Vlasov control problem corresponding to  $V$ .

The DPP that we will prove is the so-called *randomized dynamic programming principle* (see [4]), which is the dynamic programming principle for an intensity control problem for a Poisson random measure whose marks leave in a subclass of control processes which is dense with respect to the Krylov metric (see Definition 3.2.3 in [22]). See (3.8) for the definition of the randomized control problem, Theorem 3.1 for the equivalence to  $V$  (in itself is one of the main technical contributions), and Theorem 5.1, which is our main result, for the statement of the randomized dynamic programming principle. Although, the approach of replacing the original control problem with a randomized version is also taken in [4] and [17], our contribution here is in identifying the correct randomization that corresponds to the McKean-Vlasov problem. The McKean-Vlasov nature of the control problem makes this task rather difficult and as a result

the marks of the Poisson random measure live in an abstract space of processes. We should also emphasize that another relevant issue resolved in this paper concerns the flow properties for the solutions to equations (1.1) and (1.2), see Section 5.1. The importance of the flow properties is to prove an identification formula (Lemma 5.3) between  $V$  and the solution to the BSDE, which in turn allows to derive the randomized dynamic programming principle for  $V$ . Our aim is then to use the randomized dynamic programming principle to characterize  $V$  through a Hamilton-Jacobi-Bellman equation on the Wasserstein space  $\mathcal{P}_2(\mathbb{R}^n)$ , using the recent notion of Lions' differentiability.

Although it is an intermediary step in deriving the randomized DPP, we see Theorem 4.1 as the second main result of our paper. Here we derive the nonlinear Feynman-Kac representation of the value function  $V$  in terms of a class of forward-backward stochastic differential equations with constrained jumps, where the forward process is autonomous. This representation has been derived in [21] for the case of classical stochastic optimal control problems and here we are generalizing it to McKean-Vlasov control problems. The importance of this representation, beyond its intermediary role, is that it would be useful in developing probabilistic numerical schemes for  $V$  (see [20] for the case treated in [21]).

The rest of the paper is organized as follows. Section 2 is devoted to the formulation of the McKean-Vlasov control problem, and its continuity properties. In Section 3 we introduce the randomized McKean-Vlasov control problem and we prove the fundamental equivalence result between  $V$  and  $V^{\mathcal{R}}$  (Theorem 3.1). In Section 4 we prove the nonlinear Feynman-Kac representation for  $V$  in terms of the so-called randomized equation, namely BSDE (4.1). In Section 5 we derive the randomized dynamic programming principle, proving the flow properties (Lemma 5.2) and the identification between  $V$  and the solution to the BSDE (Lemma 5.3). Finally, in the Appendix we prove some convergence results with respect to the 2-Wasserstein metric  $\mathcal{W}_2$  (Appendix A), we report the proofs of the measurability Lemmata 3.1 and 3.2 (Appendix B), we state and prove a stability result with respect to the Krylov metric  $\tilde{\rho}$  (Appendix C), we consider an alternative randomization McKean-Vlasov control problem, more similar to the randomized problems studied for instance in [4, 16, 17, 21] (Appendix D).

## 2 Formulation of the McKean-Vlasov control problem

### 2.1 Notations

Consider a complete probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and a  $d$ -dimensional Brownian motion  $B = (B_t)_{t \geq 0}$  defined on it. Let  $\mathbb{F}^B = (\mathcal{F}_t^B)_{t \geq 0}$  denote the  $\mathbb{P}$ -completion of the filtration generated by  $B$ . Fix a finite time horizon  $T > 0$  and a Polish space  $A$ , endowed with a metric  $\rho$ . We suppose, without loss of generality, that  $\rho < 1$  (if this is not the case, we replace  $\rho$  with the equivalent metric  $\rho/(1 + \rho)$ ). We indicate by  $\mathcal{B}(A)$  the family of Borel subsets of  $A$ .

Let  $\mathcal{P}_2(\mathbb{R}^n)$  denote the set of all probability measures on  $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  with a finite second-order moment. We endow  $\mathcal{P}_2(\mathbb{R}^n)$  with the 2-Wasserstein metric  $\mathcal{W}_2$  defined as follows:

$$\mathcal{W}_2(\pi, \pi') = \inf \left\{ \left( \int_{\mathbb{R}^n \times \mathbb{R}^n} |x - x'|^2 \pi(dx, dx') \right)^{1/2} : \pi \in \mathcal{P}_2(\mathbb{R}^n \times \mathbb{R}^n) \text{ with marginals } \pi \text{ and } \pi' \right\},$$

for all  $\pi, \pi' \in \mathcal{P}_2(\mathbb{R}^n)$ . We recall from Theorem 6.18 in [31] that  $(\mathcal{P}_2(\mathbb{R}^n), \mathcal{W}_2)$  is a complete

separable metric space. Notice that

$$\mathcal{W}_2(\mathbb{P}_\xi, \mathbb{P}_{\xi'}) \leq (\mathbb{E}[|\xi - \xi'|^2])^{1/2}, \quad \text{for every pair } \xi, \xi' \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^n), \quad (2.1)$$

where  $\mathbb{P}_\xi$  denotes the law under  $\mathbb{P}$  of the random variable  $\xi: \Omega \rightarrow \mathbb{R}^n$ . We also denote by  $\|\pi\|_2$  the square root of the second-order moment of  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ :

$$\mathcal{W}_2(\pi, \delta_0) = \|\pi\|_2 = \left( \int_{\mathbb{R}^n} |x|^2 \pi(dx) \right)^{\frac{1}{2}}, \quad \text{for all } \pi \in \mathcal{P}_2(\mathbb{R}^n), \quad (2.2)$$

where  $\delta_0$  is the Dirac measure on  $\mathbb{R}^n$  concentrated at the origin. We denote  $\mathcal{B}(\mathcal{P}_2(\mathbb{R}^n))$  the Borel  $\sigma$ -algebra on  $\mathcal{P}_2(\mathbb{R}^n)$  induced by the 2-Wasserstein metric  $\mathcal{W}_2$ .

We assume that there exists a sub- $\sigma$ -algebra  $\mathcal{G} \subset \mathcal{F}$  such that  $B$  is independent of  $\mathcal{G}$  and  $\mathcal{P}_2(\mathbb{R}^n) = \{\mathbb{P}_\xi: \xi \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)\}$ .

Finally, we denote  $C_2(\mathbb{R}^n)$  the set of real-valued continuous functions with at most quadratic growth, and  $\mathcal{B}_2(\mathbb{R}^n)$  the set of real-valued Borel measurable functions with at most quadratic growth.

**Remark 2.1** For every  $\varphi \in C_2(\mathbb{R}^n)$ , let  $\Lambda_\varphi: \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  be given by

$$\Lambda_\varphi(\pi) = \int_{\mathbb{R}^n} \varphi(x) \pi(dx), \quad \text{for every } \pi \in \mathcal{P}_2(\mathbb{R}^n).$$

We notice that (as remarked on pages 6-7 in [18])  $\mathcal{B}(\mathcal{P}_2(\mathbb{R}^n))$  coincides with the  $\sigma$ -algebra generated by the family of maps  $\Lambda_\varphi$ ,  $\varphi \in C_2(\mathbb{R}^n)$ . As a consequence, we observe that, given a measurable space  $(E, \mathcal{E})$  and a map  $F: E \rightarrow \mathcal{P}_2(\mathbb{R}^n)$ , then  $F$  is measurable if and only if  $\Lambda_\varphi \circ F$  is measurable, for every  $\varphi \in C_2(\mathbb{R}^n)$ . Finally, we notice that if  $\varphi \in \mathcal{B}_2(\mathbb{R}^n)$  then the map  $\Lambda_\varphi$  is  $\mathcal{B}(\mathcal{P}_2(\mathbb{R}^n))$ -measurable. This latter property can be proved using a monotone class argument, noting that  $\Lambda_\varphi$  is  $\mathcal{B}(\mathcal{P}_2(\mathbb{R}^n))$ -measurable whenever  $\varphi \in C_2(\mathbb{R}^n)$ .  $\square$

## 2.2 Optimal control of McKean-Vlasov dynamics

Let  $\mathcal{A}$  denote the set of admissible control processes, which are  $\mathbb{F}^B$ -progressive processes  $\alpha: \Omega \times [0, T] \rightarrow A$ . Given  $(t, x, \xi) \in [0, T] \times \mathbb{R}^n \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$  and  $\alpha \in \mathcal{A}$ , the controlled state equations are given by:

$$dX_s^{t, \xi, \alpha} = b(s, X_s^{t, \xi, \alpha}, \mathbb{P}_{X_s^{t, \xi, \alpha}}, \alpha_s) ds + \sigma(s, X_s^{t, \xi, \alpha}, \mathbb{P}_{X_s^{t, \xi, \alpha}}, \alpha_s) dB_s, \quad X_t^{t, \xi, \alpha} = \xi, \quad (2.3)$$

$$dX_s^{t, x, \xi, \alpha} = b(s, X_s^{t, x, \xi, \alpha}, \mathbb{P}_{X_s^{t, x, \xi, \alpha}}, \alpha_s) ds + \sigma(s, X_s^{t, x, \xi, \alpha}, \mathbb{P}_{X_s^{t, x, \xi, \alpha}}, \alpha_s) dB_s, \quad X_t^{t, x, \xi, \alpha} = x, \quad (2.4)$$

for all  $s \in [t, T]$ . The coefficients  $b: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A \rightarrow \mathbb{R}^n$  and  $\sigma: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A \rightarrow \mathbb{R}^{n \times d}$  are assumed to be Borel measurable. Recall that  $\mathbb{P}_{X_s^{t, \xi, \alpha}}$  denotes the law under  $\mathbb{P}$  of the random variable  $X_s^{t, \xi, \alpha}: \Omega \rightarrow \mathbb{R}^n$ . Notice that  $(\mathbb{P}_{X_s^{t, \xi, \alpha}})_{s \in [t, T]}$  depends on  $\xi$  only through its law  $\pi = \mathbb{P}_\xi$ , and  $\pi$  is an element of  $\mathcal{P}_2(\mathbb{R}^n)$ . As a consequence,  $X^{t, x, \xi, \alpha} = (X_s^{t, x, \xi, \alpha})_{s \in [t, T]}$  depends on  $\xi$  only through  $\pi$ . Therefore, we denote  $X^{t, x, \xi, \alpha}$  simply by  $X^{t, x, \pi, \alpha}$ , whenever  $\pi = \mathbb{P}_\xi$ .

Our aim is to maximize, over all  $\alpha \in \mathcal{A}$ , the following functional

$$J(t, x, \pi, \alpha) = \mathbb{E} \left[ \int_t^T f(s, X_s^{t, x, \pi, \alpha}, \mathbb{P}_{X_s^{t, x, \pi, \alpha}}, \alpha_s) ds + g(X_T^{t, x, \pi, \alpha}, \mathbb{P}_{X_T^{t, x, \pi, \alpha}}) \right], \quad (2.5)$$

where  $f: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A \rightarrow \mathbb{R}$  and  $g: \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  are Borel measurable. We impose the following assumptions.

**(A1)**

(i) For every  $t \in [0, T]$ ,  $b(t, \cdot)$ ,  $\sigma(t, \cdot)$  and  $f(t, \cdot)$  are continuous on  $\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A$ , and  $g$  is continuous on  $\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ .

(ii) For every  $(t, x, x', \pi, \pi', a) \in [0, T] \times \mathbb{R}^n \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times \mathcal{P}_2(\mathbb{R}^n) \times A$ ,

$$|b(t, x, \pi, a) - b(t, x', \pi', a)| + |\sigma(t, x, \pi, a) - \sigma(t, x', \pi', a)| \leq L(|x - x'| + \mathcal{W}_2(\pi, \pi')),$$

$$|b(t, 0, \delta_0, a)| + |\sigma(t, 0, \delta_0, a)| \leq L,$$

$$|f(t, x, \pi, a)| + |g(x, \pi)| \leq h(\|\pi\|_2)(1 + |x|^p),$$

for some positive constants  $L$  and  $p$ , and some continuous function  $h: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ .

Under Assumption **(A1)**, and recalling property (2.1), it can be proved by standard arguments that there exists a unique (up to indistinguishability) pair  $(X_s^{t, \xi, \alpha}, X_s^{t, x, \pi, \alpha})_{s \in [t, T]}$  of continuous  $(\mathcal{F}_s^B \vee \mathcal{G})_s$ -adapted processes solution to equations (2.3)-(2.4), satisfying

$$\sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ \sup_{s \in [t, T]} (|X_s^{t, \xi, \alpha}|^2 + |X_s^{t, x, \pi, \alpha}|^q) \right] < \infty, \quad (2.6)$$

for all  $q \geq 1$ . The estimate  $\sup_{\alpha \in \mathcal{A}} \mathbb{E}[\sup_{s \in [t, T]} |X_s^{t, \xi, \alpha}|^q] < \infty$  holds whenever  $|\xi|^q$  is integrable. Notice that  $(X_s^{t, x, \pi, \alpha})_{s \in [t, T]}$  is  $\mathbb{F}^B$ -adapted.

Recalling  $\mathcal{P}_2(\mathbb{R}^n) = \{\mathbb{P}_\xi: \xi \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)\}$ , we see that  $J(t, x, \pi, \alpha)$  is defined for every quadruple  $(t, x, \pi, \alpha) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times \mathcal{A}$ . The value function of our stochastic control problem is the function  $V$  on  $[0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$  defined as

$$V(t, x, \pi) = \sup_{\alpha \in \mathcal{A}} J(t, x, \pi, \alpha), \quad (2.7)$$

for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ .

From estimate (2.6), we see that  $\|\mathbb{P}_{X_s^{t, \xi, \alpha}}\|_2 \leq M$ , for some positive constant  $M$  independent of  $\alpha \in \mathcal{A}$  and  $s \in [t, T]$ . It follows from the continuity of  $h$  that the quantity  $h(\|\mathbb{P}_{X_s^{t, \xi, \alpha}}\|_2)$  is bounded uniformly with respect to  $\alpha$  and  $s$ . Therefore, by the polynomial growth condition on  $f$  and  $g$  in Assumption **(A1)**(ii), we deduce that the value function  $V$  in (2.7) is always a finite real number on its domain  $[0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ , namely  $V: [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$ . In particular, it is easy to see that, under Assumption **(A1)**,  $V$  satisfies the following growth condition:

$$|V(t, x, \pi)| \leq \psi(\|\pi\|_2)(1 + |x|^p), \quad (2.8)$$

for some continuous function  $\psi: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ .

We now study the continuity of  $V$ . Firstly, we impose the following additional assumption.

**(A2)** For every  $t \in [0, T]$  and  $R > 0$ , the map  $(x, \pi) \mapsto f(t, \cdot, \cdot, a)$  is uniformly continuous and bounded on  $\{(x, \pi) \in \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n): |x|, \|\pi\|_2 \leq R\}$ , uniformly with respect to  $a \in A$ . For every  $R > 0$ , the map  $g$  is uniformly continuous and bounded on  $\{(x, \pi) \in \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n): |x|, \|\pi\|_2 \leq R\}$ .



**Proposition 2.1** *Under Assumptions (A1) and (A2), for every  $t \in [0, T]$  the map  $(x, \pi) \mapsto V(t, x, \pi)$  is continuous on  $\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ .*

**Proof.** We begin noting that, as a consequence of Assumption (A2), for every  $t \in [0, T]$  and  $R > 0$ , there exists a modulus of continuity  $\delta_t^R: [0, \infty) \rightarrow [0, \infty)$  such that, for  $t \in [0, T]$ ,

$$|f(t, x, \pi, a) - f(t, x', \pi', a)| \leq \delta_t^R(|x - x'| + \mathcal{W}_2(\pi, \pi')),$$

and, for  $t = T$ ,

$$|f(T, x, \pi, a) - f(T, x', \pi', a)| + |g(x, \pi) - g(x', \pi)| \leq \delta_T^R(|x - x'| + \mathcal{W}_2(\pi, \pi')),$$

for all  $(x, \pi), (x', \pi') \in \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ ,  $a \in A$ , with  $|x|, |x'|, \|\pi\|_2, \|\pi'\|_2 \leq R$ . Recall that, by definition (see for instance [2], page 406), the modulus of continuity  $\delta_s^R$  is nondecreasing and  $\lim_{\varepsilon \rightarrow 0+} \delta_s^R(\varepsilon) = 0$ . Moreover, by Assumption (A2), we see that  $\delta_s^R$  can be taken bounded. In particular,  $\limsup_{\varepsilon \rightarrow +\infty} \delta_s^R(\varepsilon)/\varepsilon = 0$ . Therefore, without loss of generality, we can suppose that  $\delta_s^R$  is concave (see for instance Theorem 1, page 406, in [2]; we refer, in particular, to the concave modulus of continuity constructed in the proof of Theorem 1 and given by formula (1.6) at page 407). Then, we notice that  $\delta_s^R$  is also subadditive.

Now, fix  $t \in [0, T]$  and  $(x, \pi), (x_m, \pi_m) \in \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ , with  $|x_m - x| \rightarrow 0$  and  $\mathcal{W}_2(\pi_m, \pi) \rightarrow 0$  as  $m$  goes to infinity. Our aim is to prove that

$$V(t, x_m, \pi_m) \xrightarrow{m \rightarrow \infty} V(t, x, \pi). \quad (2.9)$$

By Lemma A.1 we know that there exist random variables  $\xi, \xi_m \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$  such that  $\pi = \mathbb{P}_\xi$  and  $\pi_m = \mathbb{P}_{\xi_m}$  under  $\mathbb{P}$ , moreover  $\xi_m$  converges to  $\xi$  pointwise  $\mathbb{P}$ -a.s. and in  $L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ . In particular,  $\sup_m \mathbb{E}[|\xi_m|^2] < \infty$ . Then, by standard arguments, we have

$$\max \left\{ \sup_{s \in [t, T], \alpha \in \mathcal{A}} \|\mathbb{P}_{X_s^{t, \xi, \alpha}}\|_2, \sup_m \sup_{s \in [t, T], \alpha \in \mathcal{A}} \|\mathbb{P}_{X_s^{t, \xi_m, \alpha}}\|_2 \right\} =: \bar{R},$$

for some constant  $\bar{R} \geq 0$ . For every  $R > \bar{R}$  and  $\alpha \in \mathcal{A}$ , define the set  $E_\alpha \in \mathcal{F}$  as

$$E_\alpha := \left\{ \omega \in \Omega : \sup_{s \in [t, T]} |X_s^{t, x, \pi, \alpha}(\omega)|, \sup_m \sup_{s \in [t, T]} |X_s^{t, x_m, \pi_m, \alpha}(\omega)| \leq R \right\}.$$

Then, we have

$$\begin{aligned} & |V(t, x, \pi) - V(t, x_m, \pi_m)| \\ & \leq \sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ 1_{E_\alpha} \int_t^T \delta_s^R(|X_s^{t, x, \pi, \alpha} - X_s^{t, x_m, \pi_m, \alpha}|) ds + 1_{E_\alpha^c} \delta_T^R(|X_T^{t, x, \pi, \alpha} - X_T^{t, x_m, \pi_m, \alpha}|) \right] \\ & \quad + \sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ 1_{E_\alpha} \int_t^T \delta_s^R(\mathcal{W}_2(\mathbb{P}_{X_s^{t, \xi, \alpha}}, \mathbb{P}_{X_s^{t, \xi_m, \alpha}})) ds + 1_{E_\alpha^c} \delta_T^R(\mathcal{W}_2(\mathbb{P}_{X_T^{t, \xi, \alpha}}, \mathbb{P}_{X_T^{t, \xi_m, \alpha}})) \right] \\ & \quad + \sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ 1_{E_\alpha^c} |g(X_T^{t, x, \pi, \alpha}, \mathbb{P}_{X_T^{t, \xi, \alpha}}) - g(X_T^{t, x_m, \pi_m, \alpha}, \mathbb{P}_{X_T^{t, \xi_m, \alpha}})| \right] \\ & \quad + 1_{E_\alpha^c} \int_t^T |f(s, X_s^{t, x, \pi, \alpha}, \mathbb{P}_{X_s^{t, \xi, \alpha}}, \alpha_s) - f(s, X_s^{t, x_m, \pi_m, \alpha}, \mathbb{P}_{X_s^{t, \xi_m, \alpha}}, \alpha_s)| ds \Big] \\ & \leq \sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ \int_t^T \delta_s^R(|X_s^{t, x, \pi, \alpha} - X_s^{t, x_m, \pi_m, \alpha}|) ds + \delta_T^R(|X_T^{t, x, \pi, \alpha} - X_T^{t, x_m, \pi_m, \alpha}|) \right] \end{aligned}$$

$$\begin{aligned}
& + \sup_{\alpha \in \mathcal{A}} \left( \int_t^T \delta_s^R (\mathcal{W}_2(\mathbb{P}_{X_s^{t,\xi,\alpha}}, \mathbb{P}_{X_s^{t,\xi_m,\alpha}})) ds + \delta_T^R (\mathcal{W}_2(\mathbb{P}_{X_T^{t,\xi,\alpha}}, \mathbb{P}_{X_T^{t,\xi_m,\alpha}})) \right) \\
& + C(1 + |x|^p + |x_m|^p) \sup_{\alpha \in \mathcal{A}} \mathbb{P}(E_\alpha^c),
\end{aligned} \tag{2.10}$$

for some positive constant  $C$ , depending only on  $\bar{R}$ ,  $T$ , the constants  $L$ ,  $p$  in Assumption **(A1)**(ii), and the maximum  $\max_{0 \leq r \leq \bar{R}} h(r)$ , where the function  $h$  was introduced in Assumption **(A1)**(ii). Recalling that  $\mathcal{W}_2(\mathbb{P}_{X_s^{t,\xi,\alpha}}, \mathbb{P}_{X_s^{t,\xi_m,\alpha}}) \leq \mathbb{E}[|X_s^{t,\xi,\alpha} - X_s^{t,\xi_m,\alpha}|^2]$  and  $\delta_s^R$  is nondecreasing, we find

$$\delta_s^R (\mathcal{W}_2(\mathbb{P}_{X_s^{t,\xi,\alpha}}, \mathbb{P}_{X_s^{t,\xi_m,\alpha}})) \leq \delta_s^R \left( \mathbb{E}[|X_s^{t,\xi,\alpha} - X_s^{t,\xi_m,\alpha}|^2]^{1/2} \right). \tag{2.11}$$

Now, recall the standard estimate

$$\sup_{\alpha \in \mathcal{A}} \mathbb{E}[|X_s^{t,\xi,\alpha} - X_s^{t,\xi_m,\alpha}|^2]^{1/2} \leq \hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2}, \tag{2.12}$$

for some positive constant  $\hat{c}$ , depending only on  $T$  and  $L$ . Therefore, from (2.11) we obtain

$$\delta_s^R (\mathcal{W}_2(\mathbb{P}_{X_s^{t,\xi,\alpha}}, \mathbb{P}_{X_s^{t,\xi_m,\alpha}})) \leq \delta_s^R \left( \hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right). \tag{2.13}$$

On the other hand, from the concavity of  $\delta_s^R$ , we get

$$\mathbb{E}[\delta_s^R (|X_s^{t,x,\pi,\alpha} - X_s^{t,x_m,\pi_m,\alpha}|)] \leq \delta_s^R (\mathbb{E}[|X_s^{t,x,\pi,\alpha} - X_s^{t,x_m,\pi_m,\alpha}|]). \tag{2.14}$$

By standard arguments, we have

$$\sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ \sup_{s \in [t, T]} |X_s^{t,x,\pi,\alpha} - X_s^{t,x_m,\pi_m,\alpha}| \right] \leq c \left( |x - x_m| + \sup_{\alpha \in \mathcal{A}} \sup_{s \in [t, T]} \mathcal{W}_2(\mathbb{P}_{X_s^{t,\xi,\alpha}}, \mathbb{P}_{X_s^{t,\xi_m,\alpha}}) \right),$$

where  $c$  is a positive constant, depending only on  $T$  and  $L$ . Therefore, by (2.12), we obtain

$$\sup_{\alpha \in \mathcal{A}} \mathbb{E} \left[ \sup_{s \in [t, T]} |X_s^{t,x,\pi,\alpha} - X_s^{t,x_m,\pi_m,\alpha}| \right] \leq c \left( |x - x_m| + \hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right). \tag{2.15}$$

Since  $\delta_s^R$  is nondecreasing, from (2.14) and (2.15), we find

$$\sup_{\alpha \in \mathcal{A}} \mathbb{E}[\delta_s^R (|X_s^{t,x,\pi,\alpha} - X_s^{t,x_m,\pi_m,\alpha}|)] \leq \delta_s^R \left( c|x - x_m| + c\hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right). \tag{2.16}$$

Concerning  $\mathbb{P}(E_\alpha^c)$ , we have

$$\begin{aligned}
\mathbb{P}(E_\alpha^c) & \leq \mathbb{P} \left( \sup_{s \in [t, T]} |X_s^{t,x,\pi,\alpha}| > R \right) + \mathbb{P} \left( \sup_{s \in [t, T]} |X_s^{t,x_m,\pi_m,\alpha}| > R \right) \\
& \leq \frac{1}{R^2} \mathbb{E} \left[ \sup_{s \in [t, T]} |X_s^{t,x,\pi,\alpha}|^2 \right] + \frac{1}{R^2} \mathbb{E} \left[ \sup_{s \in [t, T]} |X_s^{t,x_m,\pi_m,\alpha}|^2 \right] \leq \frac{c_0}{R^2} (1 + |x|^2 + |x_m|^2),
\end{aligned} \tag{2.17}$$

for some positive constant  $c_0$ , depending only on  $T$ ,  $L$ ,  $\bar{R}$ . In conclusion, plugging (2.13)-(2.16)-(2.17) into (2.10), we get

$$\begin{aligned}
& |V(t, x, \pi) - V(t, x_m, \pi_m)| \\
& \leq \int_t^T \delta_s^R \left( c|x - x_m| + c\hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right) ds + \delta_T^R \left( c|x - x_m| + c\hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right)
\end{aligned}$$

$$\begin{aligned}
& + \int_t^T \delta_s^R \left( \hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right) ds + \delta_T^R \left( \hat{c} \mathbb{E}[|\xi - \xi_m|^2]^{1/2} \right) \\
& + \frac{c_0 C}{R^2} (1 + |x|^2 + |x_m|^2) (1 + |x|^p + |x_m|^p).
\end{aligned} \tag{2.18}$$

Taking the  $\limsup_{m \rightarrow \infty}$  in the above inequality, we find

$$\limsup_{m \rightarrow \infty} |V(t, x, \pi) - V(t, x_m, \pi_m)| \leq \frac{c_0 C}{R^2} (1 + 2|x|^2) (1 + 2|x|^p).$$

Letting  $R \rightarrow \infty$ , we deduce that  $\limsup_{n \rightarrow \infty} |V(t, x, \pi) - V(t, x_m, \pi_m)| = 0$ , therefore (2.9) holds.  $\square$

We end this section showing that the value function  $V : [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  given by (2.7) is the disintegration of the value function  $V_{\text{MKV}} : [0, T] \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n) \rightarrow \mathbb{R}$  given by:

$$V_{\text{MKV}}(t, \xi) = \sup_{\alpha \in \mathcal{A}_\xi} \mathbb{E} \left[ \int_t^T f(s, X_s^{t, \xi, \alpha}, \mathbb{P}_{X_s^{t, \xi, \alpha}}^\xi, \alpha_s) ds + g(X_T^{t, \xi, \alpha}, \mathbb{P}_{X_T^{t, \xi, \alpha}}^\xi) \right], \tag{2.19}$$

for every  $(t, \xi) \in [0, T] \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , where  $\mathcal{A}_\xi$  denotes the set of  $A$ -valued  $(\mathcal{F}_s^B \vee \sigma(\xi))$ -progressive processes,  $(X_s^{t, \xi, \alpha})_{s \in [t, T]}$  is the solution to the following equation:

$$dX_s^{t, \xi, \alpha} = b(s, X_s^{t, \xi, \alpha}, \mathbb{P}_{X_s^{t, \xi, \alpha}}^\xi, \alpha_s) ds + \sigma(s, X_s^{t, \xi, \alpha}, \mathbb{P}_{X_s^{t, \xi, \alpha}}^\xi, \alpha_s) dB_s, \quad X_t^{t, \xi, \alpha} = \xi,$$

for all  $s \in [t, T]$ , with  $\alpha \in \mathcal{A}_\xi$ , and  $\mathbb{P}_{X_s^{t, \xi, \alpha}}^\xi$  denotes the regular conditional distribution of the random variable  $X_s^{t, \xi, \alpha} : \Omega \rightarrow \mathbb{R}^n$  with respect to  $\sigma(\xi)$ , whose existence is guaranteed for instance by Theorem 6.3 in [19].

**Proposition 2.2** *Under Assumptions (A1) and (A2), for every  $(t, \xi) \in [0, T] \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\mathbb{P}$ , we have*

$$V_{\text{MKV}}(t, \xi) = \mathbb{E}[V(t, \xi, \pi)],$$

or, equivalently,

$$V_{\text{MKV}}(t, \xi) = \int_{\mathbb{R}^n} V(t, x, \pi) \pi(dx).$$

**Proof.** Fix  $t \in [0, T]$ . Recall from Proposition 2.1 that the map  $(x, \pi) \mapsto V(t, x, \pi)$  is continuous on  $\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ . Proceeding as in the proof of Proposition 2.1, we can also prove that the map  $\xi \mapsto V_{\text{MKV}}(t, \xi)$  is continuous on  $L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ . As a consequence, it is enough to prove the Proposition for  $\xi \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$  taking only a finite number of values, the general result being proved by approximation. In other words, we suppose that

$$\xi = \sum_{k=0}^K x_k 1_{E_k},$$

for some  $K \in \mathbb{N}$ ,  $x_k \in \mathbb{R}^n$ ,  $E_k \in \sigma(\xi)$ , with  $(E_k)_{k=1, \dots, K}$  being a partition of  $\Omega$ . Notice that  $\alpha \in \mathcal{A}_\xi$  if and only if

$$\alpha = \sum_{k=0}^K \alpha_k 1_{E_k}, \tag{2.20}$$

for some  $\alpha_k \in \mathcal{A}$ . We also observe that

$$X_s^{t,\xi,\alpha} = \sum_{k=0}^K X_s^{t,x_k,\alpha_k} 1_{E_k}, \quad \mathbb{P}_{X_s^{t,\xi,\alpha}}^\xi = \sum_{k=0}^K \mathbb{P}_{X_s^{t,x_k,\alpha_k}} 1_{E_k}.$$

Then, the stochastic processes  $(X_s^{t,\xi,\alpha})_{s \in [t,T]}$  and  $(X_s^{t,x_1,\delta_{x_1},\alpha_1} 1_{E_1} + \dots + X_s^{t,x_K,\delta_{x_K},\alpha_K} 1_{E_K})_{s \in [t,T]}$  are indistinguishable, since they solve the same equation. Therefore

$$\begin{aligned} V_{\text{MKV}}(t, \xi) &= \sup_{\alpha \in \mathcal{A}_\xi} \mathbb{E} \left[ \int_t^T f(s, X_s^{t,\xi,\alpha}, \mathbb{P}_{X_s^{t,\xi,\alpha}}^\xi, \alpha_s) ds + g(X_T^{t,\xi,\alpha}, \mathbb{P}_{X_T^{t,\xi,\alpha}}^\xi) \right] \\ &= \sup_{\alpha \in \mathcal{A}_\xi} \mathbb{E} \left[ \sum_{k=0}^K \left( \int_t^T f(s, X_s^{t,x_k,\delta_{x_k},\alpha_k}, \mathbb{P}_{X_s^{t,x_k,\alpha_k}}^\xi, (\alpha_k)_s) ds + g(X_T^{t,x_k,\delta_{x_k},\alpha_k}, \mathbb{P}_{X_T^{t,x_k,\alpha_k}}^\xi) \right) 1_{E_k} \right]. \end{aligned} \quad (2.21)$$

Since  $\xi$  is independent of  $X^{t,x_k,\delta_{x_k},\alpha_k}$  and of  $\alpha_k$ , we can write the last quantity in (2.21) as

$$V_{\text{MKV}}(t, \xi) = \sup_{\alpha \in \mathcal{A}_\xi} \mathbb{E} \left[ \sum_{k=0}^K \mathbb{E} \left[ \int_t^T f(s, X_s^{t,x_k,\delta_{x_k},\alpha_k}, \mathbb{P}_{X_s^{t,x_k,\alpha_k}}^\xi, (\alpha_k)_s) ds + g(X_T^{t,x_k,\delta_{x_k},\alpha_k}, \mathbb{P}_{X_T^{t,x_k,\alpha_k}}^\xi) \right] 1_{E_k} \right].$$

From (2.20), we conclude that

$$\begin{aligned} V_{\text{MKV}}(t, \xi) &= \mathbb{E} \left[ \sum_{k=0}^K \sup_{\alpha_k \in \mathcal{A}} \mathbb{E} \left[ \int_t^T f(s, X_s^{t,x_k,\delta_{x_k},\alpha_k}, \mathbb{P}_{X_s^{t,x_k,\alpha_k}}^\xi, (\alpha_k)_s) ds + g(X_T^{t,x_k,\delta_{x_k},\alpha_k}, \mathbb{P}_{X_T^{t,x_k,\alpha_k}}^\xi) \right] 1_{E_k} \right] \\ &= \mathbb{E} \left[ \sum_{k=0}^K V(t, x_k, \delta_{x_k}) 1_{E_k} \right] = \mathbb{E}[V(t, \xi, \pi)]. \end{aligned}$$

□

### 3 The randomized McKean-Vlasov control problem

Following Definition 3.2.3 in [22], we define on  $\mathcal{A}$  the metric  $\tilde{\rho}$  given by:

$$\tilde{\rho}(\alpha, \beta) := \mathbb{E} \left[ \int_0^T \rho(\alpha_t, \beta_t) dt \right], \quad (3.1)$$

where we recall that  $\rho$  is a metric on  $A$  satisfying  $\rho < 1$ . Notice that convergence with respect to  $\tilde{\rho}$  is equivalent to convergence in  $d\mathbb{P} dt$ -measure. We also observe that  $(\mathcal{A}, \tilde{\rho})$  is a metric space (identifying processes  $\alpha$  and  $\beta$  which are equal  $d\mathbb{P} dt$ -a.e. on  $\Omega \times [0, T]$ ). Moreover, since  $A$  is a Polish space, it turns out that  $(\mathcal{A}, \tilde{\rho})$  is also a Polish space (separability follows from Lemma 3.2.6 in [22], completeness follows from the completeness of  $A$  and the fact that a  $\tilde{\rho}$ -limit of  $\mathbb{F}^B$ -progressive processes is still  $\mathbb{F}^B$ -progressive). We denote by  $\mathcal{B}(\mathcal{A})$  the family of Borel subsets of  $\mathcal{A}$ .

Following [22], we introduce the following subset of admissible control processes.

**Definition 3.1** *For every  $t \in [0, T]$ , let  $(E_\ell^t)_{\ell \geq 1} \in \mathcal{F}$  be a countable class of subsets of  $\Omega$  which generates  $\sigma(B_s, s \in [0, t])$ . Fix a countable dense subset  $(a_m)_{m \geq 1}$  of  $A$ . Fix also, for every*

integer  $k \geq 1$ , a subdivision  $I_k := \{0 =: t_0 < t_1 < \dots < t_k := T\}$  of the interval  $[0, T]$ , with the diameter  $\max_{i=1, \dots, k} (t_i - t_{i-1})$  of the subdivision  $I_k$  going to zero as  $k \rightarrow \infty$ . Then, we denote

$$\begin{aligned} \mathcal{A}_{\text{step}} := & \left\{ \alpha \in \mathcal{A} : \text{there exist } k \geq 1, M \geq 1, L \geq 1, \text{ such that, for every } i = 0, \dots, k-1, \right. \\ & \alpha_{t_i} : \Omega \rightarrow (a_m)_{m=1, \dots, M}, \text{ with } \alpha_{t_i} \text{ constant on the sets of the partition} \\ & \text{generated by } E_1^{t_i}, \dots, E_L^{t_i}, \text{ and, for every } t \in [0, T], \\ & \left. \alpha_t = \alpha_{t_0} 1_{[t_0, t_1)}(t) + \dots + \alpha_{t_{k-1}} 1_{[t_{k-1}, t_k)}(t) + \alpha_{t_k} 1_{\{t_k\}}(t) \right\}. \end{aligned}$$

**Remark 3.1** Notice that  $\mathcal{A}_{\text{step}}$  depends (even if we omit to write explicitly this dependence) on the two sequences  $(a_m)_{m \geq 1}$  and  $(I_k)_{k \geq 1}$ , which are supposed to be fixed throughout the paper. The set  $\mathcal{A}_{\text{step}}$ , with  $\alpha_{t_i}$  being  $\sigma(B_s, s \in [0, t_i])$ -measurable, is introduced in the proof of Lemma 3.2.6 in [22], where it is proved that it is dense in  $\mathcal{A}$  with respect to the metric  $\tilde{\rho}$  defined in (3.1). It can be shown (proceeding as in the proof of Lemma C.1) that the map  $\alpha \mapsto J(t, x, \pi, \alpha)$  is continuous with respect to  $\tilde{\rho}$ , so that we could define  $V(t, x, \pi)$  in the following equivalent way:

$$V(t, x, \pi) = \sup_{\alpha \in \mathcal{A}_{\text{step}}} J(t, x, \pi, \alpha). \quad (3.2)$$

Finally, we observe that  $\mathcal{A}_{\text{step}}$  is a countable set, so that it is a Borel subset of  $\mathcal{A}$ , namely  $\mathcal{A}_{\text{step}} \in \mathcal{B}(\mathcal{A})$ .  $\square$

Now, in order to implement the randomization method, it is better to reformulate the original McKean-Vlasov control problem as follows. Let  $\mathcal{A}_{\text{step}}$  be the following set:

$$\mathcal{A}_{\text{step}} := \{ \alpha : [0, T] \rightarrow \mathcal{A}_{\text{step}} : \alpha \text{ is Borel-measurable, càdlàg, and piecewise constant} \}.$$

It is easy to see that, for every  $\alpha \in \mathcal{A}_{\text{step}}$ , the stochastic process  $((\alpha_s)_s)_{s \in [0, T]}$  is an element of  $\mathcal{A}$ . Vice versa, for every element  $\hat{\alpha} \in \mathcal{A}_{\text{step}}$ , there exists  $\hat{\alpha} \in \mathcal{A}_{\text{step}}$  such that  $((\hat{\alpha}_s)_s)_{s \in [0, T]}$  coincides with  $\hat{\alpha}$  (take  $\hat{\alpha}_s = \hat{\alpha}$ , for every  $s \in [0, T]$ ). Hence, by (3.2),

$$V(t, x, \pi) = \sup_{\alpha \in \mathcal{A}_{\text{step}}} J(t, x, \pi, ((\alpha_s)_s)_{s \in [0, T]}).$$

On the right-hand side of the above identity we have an optimization problem with class of admissible control processes given by  $\{((\alpha_s)_s)_{s \in [0, T]} : \alpha \in \mathcal{A}_{\text{step}}\}$ . We now randomize this latter control problem.

Consider another complete probability space  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ . We denote by  $\mathbb{E}^1$  the  $\mathbb{P}^1$ -expected value. We suppose that a Poisson random measure  $\mu$  on  $\mathbb{R}_+ \times \mathcal{A}$  is defined on  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ . The random measure  $\mu$  has compensator  $\lambda(d\alpha) dt$ , for some finite positive measure  $\lambda$  on  $\mathcal{A}$ , with full topological support given by  $\mathcal{A}_{\text{step}}$ . We denote  $\tilde{\mu}(dt d\alpha) := \mu(dt d\alpha) - \lambda(d\alpha) dt$  the compensated martingale measure associated to  $\mu$ . We introduce  $\mathbb{F}^\mu = (\mathcal{F}_t^\mu)_{t \geq 0}$ , which is the  $\mathbb{P}^1$ -completion of the filtration generated by  $\mu$ , given by:

$$\mathcal{F}_t^\mu = \sigma(\mu((0, s] \times \mathcal{A}') : s \in [0, t], \mathcal{A}' \subset \mathcal{A}_{\text{step}}) \vee \mathcal{N}^1,$$

for all  $t \geq 0$ , where  $\mathcal{N}^1$  is the class of  $\mathbb{P}^1$ -null sets of  $\mathcal{F}^1$ . We also denote  $\mathcal{P}(\mathbb{F}^\mu)$  the predictable  $\sigma$ -algebra on  $\Omega^1 \times \mathbb{R}_+$  corresponding to  $\mathbb{F}^\mu$ .

We recall that  $\mu$  is associated to a marked point process  $(T_n, \mathcal{A}_n)_{n \geq 1}$  on  $\mathbb{R}_+ \times \mathcal{A}$  by the formula  $\mu = \sum_{n \geq 1} \delta_{(T_n, \mathcal{A}_n)}$ , where  $\delta_{(T_n, \mathcal{A}_n)}$  is the Dirac measure concentrated at the random point  $(T_n, \mathcal{A}_n)$ . We recall that every  $T_n$  is an  $\mathbb{F}^\mu$ -stopping time and every  $\mathcal{A}_n$  is  $\mathcal{F}_{T_n}^\mu$ -measurable.

Let  $\bar{\Omega} = \Omega \times \Omega^1$ , and let  $\bar{\mathcal{F}}$  be the  $\mathbb{P} \otimes \mathbb{P}^1$ -completion of  $\mathcal{F} \otimes \mathcal{F}^1$ , and  $\bar{\mathbb{P}}$  the extension of  $\mathbb{P} \otimes \mathbb{P}^1$  to  $\bar{\mathcal{F}}$ . We denote by  $\bar{\mathcal{G}}, \bar{B}, \bar{\mu}$  the canonical extensions of  $\mathcal{G}, B, \mu$ , to  $\bar{\Omega}$ , given by:  $\bar{\mathcal{G}} := \{G \times \Omega^1 : G \in \mathcal{G}\}$ ,  $\bar{B}(\omega, \omega^1) := B(\omega)$ ,  $\bar{\mu}(\omega, \omega^1; dt d\alpha) := \mu(\omega^1; dt d\alpha)$ . Let  $\bar{\mathbb{F}}^B = (\bar{\mathcal{F}}_t^B)_{t \geq 0}$  (resp.  $\bar{\mathbb{F}}^\mu = (\bar{\mathcal{F}}_t^\mu)_{t \geq 0}$ ) denote the  $\bar{\mathbb{P}}$ -completion of the filtration generated by  $\bar{B}$  (resp.  $\bar{\mu}$ ). Notice that  $\bar{\mathcal{F}}_\infty^B$  and  $\bar{\mathcal{F}}_\infty^\mu$  are independent.

Let  $\bar{\mathbb{F}}^{B, \mu} = (\bar{\mathcal{F}}_t^{B, \mu})_{t \geq 0}$  denote the  $\bar{\mathbb{P}}$ -completion of the filtration generated by  $\bar{B}$  and  $\bar{\mu}$ . Notice that  $\bar{B}$  is a Brownian motion with respect to  $\bar{\mathbb{F}}^{B, \mu}$  and the  $\bar{\mathbb{F}}^{B, \mu}$ -compensator of  $\bar{\mu}$  is given by  $\lambda(d\alpha) dt$ . We define the  $A$ -valued piecewise constant process  $\bar{I} = (\bar{I}_t)_{t \geq 0}$  on  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$  as follows:

$$\bar{I}_t(\omega, \omega^1) = \sum_{n \geq 0} (\mathcal{A}_n(\omega^1))_{t \wedge T}(\omega) 1_{[T_n(\omega^1), T_{n+1}(\omega^1))}(t), \quad \text{for all } t \geq 0, \quad (3.3)$$

where  $T_0 := 0$  and  $\mathcal{A}_0 := \bar{\alpha}$ , for some deterministic and arbitrary control process  $\bar{\alpha} \in \mathcal{A}_{\text{step}}$ , which will remain fixed throughout the paper. Notice that  $\bar{I}$  is  $\bar{\mathbb{F}}^{B, \mu}$ -adapted.

Randomizing the control in (2.3)-(2.4), we are led to consider the following equations on  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$ , for every  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ :

$$d\bar{X}_s^{t, \bar{\xi}} = b(s, \bar{X}_s^{t, \bar{\xi}}, \mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_s) ds + \sigma(s, \bar{X}_s^{t, \bar{\xi}}, \mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_s) d\bar{B}_s, \quad \bar{X}_t^{t, \bar{\xi}} = \bar{\xi}, \quad (3.4)$$

$$d\bar{X}_s^{t, x, \pi} = b(s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_s) ds + \sigma(s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_s) d\bar{B}_s, \quad \bar{X}_t^{t, x, \pi} = x, \quad (3.5)$$

for all  $s \in [t, T]$ , where  $\mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}$  denotes the regular conditional distribution of the random variable  $\bar{X}_s^{t, \bar{\xi}} : \bar{\Omega} \rightarrow \mathbb{R}^n$  with respect to  $\bar{\mathcal{F}}_s^\mu$ , whose existence is guaranteed for instance by Theorem 6.3 in [19]. Notice that  $\mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}$  depends on  $\bar{\xi}$  only through its law  $\pi$ , so that equation (3.5) depends only on  $\pi$ . Under Assumption **(A1)**, it follows by standard arguments that there exists a unique (up to indistinguishability) pair  $(\bar{X}_s^{t, \bar{\xi}}, \bar{X}_s^{t, x, \pi})_{s \in [t, T]}$  of continuous  $(\bar{\mathcal{F}}_s^{B, \mu} \vee \bar{\mathcal{G}})_s$ -adapted processes solution to equations (3.4)-(3.5), satisfying

$$\bar{\mathbb{E}} \left[ \sup_{s \in [t, T]} (|\bar{X}_s^{t, \bar{\xi}}|^2 + |\bar{X}_s^{t, x, \pi}|^q) \right] < \infty, \quad (3.6)$$

for all  $q \geq 1$ , where  $\bar{\mathbb{E}}$  denotes the  $\bar{\mathbb{P}}$ -expected value. Moreover,  $(\bar{X}_s^{t, x, \pi})_{s \in [t, T]}$  is  $\bar{\mathbb{F}}^{B, \mu}$ -adapted.

We now prove two technical results concerning the process  $(\mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu})_{s \in [t, T]}$ . In particular, the first result (Lemma 3.1) concerns a particular version of  $(\mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu})_{s \in [t, T]}$ , which will be used in the proof of Lemma 3.2. This latter proves the existence of another version of  $(\mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu})_{s \in [t, T]}$ , which will be used throughout the paper.

**Lemma 3.1** *Under Assumption **(A1)**, for every  $(t, \pi) \in [0, T] \times \mathcal{P}_2(\mathbb{R}^n)$ , there exists a  $\mathcal{P}_2(\mathbb{R}^n)$ -valued  $\mathbb{F}^\mu$ -predictable stochastic process  $(\hat{\mathbb{P}}_s^{t, \pi})_{s \in [t, T]}$  which is a version of  $(\mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu})_{s \in [t, T]}$ , with  $\bar{\xi} \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  such that  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ . For all  $s \in [t, T]$ ,  $\hat{\mathbb{P}}_s^{t, \pi}$  is given by*

$$\hat{\mathbb{P}}_s^{t, \pi}(\omega^1)[\varphi] = \mathbb{E}[\varphi(\bar{X}_s^{t, \bar{\xi}}(\cdot, \omega^1))], \quad (3.7)$$

for every  $\omega^1 \in \Omega^1$  and  $\varphi \in \mathcal{B}_2(\mathbb{R}^n)$ .

**Proof.** See Appendix B.  $\square$

**Lemma 3.2** *Under Assumption (A1), for every  $t \in [0, T]$ , there exists a measurable map  $\mathbb{P}^{t, \pi}: (\Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n), \mathcal{F}^1 \otimes \mathcal{B}([t, T]) \otimes \mathcal{B}(\mathcal{P}_2(\mathbb{R}^n))) \rightarrow (\mathcal{P}_2(\mathbb{R}^n), \mathcal{B}(\mathcal{P}_2(\mathbb{R}^n)))$  such that*

$$\mathbb{P}_s^{t, \pi} = \mathbb{P}_{\bar{X}_s^{t, \xi}}^{\bar{\mathcal{F}}_s^\mu},$$

*$\mathbb{P}^1$ -a.s., for every  $s \in [t, T]$ ,  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ , where  $\bar{\xi} \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  has law  $\pi$  under  $\bar{\mathbb{P}}$ . In other words, for every  $s \in [t, T]$  and  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ ,  $(\mathbb{P}_s^{t, \pi})_{s \in [t, T]}$  is a version of  $(\mathbb{P}_{\bar{X}_s^{t, \xi}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$ .*

**Proof.** See Appendix B.  $\square$

From now on, we will always suppose that  $(\mathbb{P}_{\bar{X}_s^{t, \xi}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$  stands for the stochastic process  $(\mathbb{P}_s^{t, \pi})_{s \in [t, T]}$  introduced in Lemma 3.2.

Let us now formulate the randomized McKean-Vlasov control problem. An admissible control is a  $\mathcal{P}(\mathbb{F}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable map  $\nu: \Omega^1 \times \mathbb{R}_+ \times \mathcal{A} \rightarrow (0, \infty)$ , which is both bounded away from zero and bounded from above:  $0 < \inf_{\Omega^1 \times \mathbb{R}_+ \times \mathcal{A}} \nu \leq \sup_{\Omega^1 \times \mathbb{R}_+ \times \mathcal{A}} \nu < \infty$ . We denote by  $\mathcal{V}$  the set of admissible controls. Given  $\nu \in \mathcal{V}$ , we define  $\mathbb{P}^\nu$  on  $(\Omega^1, \mathcal{F}^1)$  as  $d\mathbb{P}^\nu = \kappa_t^\nu d\mathbb{P}^1$ , where  $\kappa^\nu = (\kappa_t^\nu)_{t \in [0, T]}$  is the Doléans exponential process on  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$  defined as

$$\begin{aligned} \kappa_t^\nu &= \mathcal{E}_t \left( \int_0^t \int_{\mathcal{A}} (\nu_s(\alpha) - 1) \tilde{\mu}(ds d\alpha) \right) \\ &= \exp \left( \int_0^t \int_{\mathcal{A}} \ln \nu_s(\alpha) \mu(ds d\alpha) - \int_0^t \int_{\mathcal{A}} (\nu_s(\alpha) - 1) \lambda(d\alpha) ds \right), \quad \text{for all } t \in [0, T]. \end{aligned}$$

Notice that  $\kappa^\nu$  is an  $\mathbb{F}^\mu$ -martingale under  $\mathbb{P}^1$ , so that  $\mathbb{P}^\nu$  is a probability measure on  $(\Omega^1, \mathcal{F}^1)$ . We denote by  $\mathbb{E}^\nu$  the  $\mathbb{P}^\nu$ -expected value. Observe that, by the Girsanov theorem, the  $\mathbb{F}^\mu$ -compensator of  $\mu$  under  $\mathbb{P}^\nu$  is given by  $\nu_t(\alpha) \lambda(d\alpha) dt$ . Let  $\bar{\mathbb{P}}^\nu$  denote the extension of  $\mathbb{P} \otimes \mathbb{P}^\nu$  to  $(\bar{\Omega}, \bar{\mathcal{F}})$ . Then  $d\bar{\mathbb{P}}^\nu = \bar{\kappa}_T^\nu d\bar{\mathbb{P}}$ , where  $\bar{\kappa}_t^\nu(\omega, \omega^1) := \kappa_t^\nu(\omega^1)$ , for all  $t \in [0, T]$ . Using again the Girsanov theorem, we see that the  $\bar{\mathbb{F}}^{B, \mu}$ -compensator of  $\bar{\mu}$  under  $\bar{\mathbb{P}}^\nu$  is  $\bar{\nu}_t(\alpha) \lambda(d\alpha) dt$ , where  $\bar{\nu}_t(\omega, \omega^1, \alpha) := \nu_t(\omega^1, \alpha)$  is the canonical extension of  $\nu$  to  $\bar{\Omega} \times \mathbb{R}_+ \times \mathcal{A}$ .

Notice that a  $\bar{\mathcal{G}}$ -measurable  $\bar{\xi}: \bar{\Omega} \rightarrow \mathbb{R}^n$  has law  $\pi$  under  $\bar{\mathbb{P}}$  if and only if it has the same law under  $\bar{\mathbb{P}}^\nu$ . In particular,  $\bar{\xi} \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  if and only if  $\bar{\xi} \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}^\nu; \mathbb{R}^n)$ . As a consequence, the following generalization of estimate (3.6) holds ( $\bar{\mathbb{E}}^\nu$  denotes the  $\bar{\mathbb{P}}^\nu$ -expected value):

$$\sup_{\nu \in \mathcal{V}} \bar{\mathbb{E}}^\nu \left[ \sup_{s \in [t, T]} (|\bar{X}_s^{t, \xi}|^2 + |\bar{X}_s^{t, x, \pi}|^q) \right] < \infty,$$

for all  $q \geq 1$ , for every  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$  (or, equivalently, under  $\bar{\mathbb{P}}^\nu$ ).

Let  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ , and  $\nu \in \mathcal{V}$ , then the gain functional for the randomized McKean-Vlasov control problem is given by:

$$J^{\mathcal{R}}(t, x, \pi, \nu) = \bar{\mathbb{E}}^\nu \left[ \int_t^T f(s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_{\bar{X}_s^{t, \xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\bar{X}_T^{t, x, \pi}, \mathbb{P}_{\bar{X}_T^{t, \xi}}^{\bar{\mathcal{F}}_T^\mu}) \right].$$

As for the functional (2.5), the quantity  $J^{\mathcal{R}}(t, x, \pi, \nu)$  is defined for every  $(t, x, \pi, \nu) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times \mathcal{V}$ , since by assumption  $\mathcal{P}_2(\mathbb{R}^n) = \{\mathbb{P}_\xi: \xi \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)\}$ . Then, we can define the value function of the randomized McKean-Vlasov control problem as

$$V^{\mathcal{R}}(t, x, \pi) = \sup_{\nu \in \mathcal{V}} J^{\mathcal{R}}(t, x, \pi, \nu), \quad (3.8)$$

for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ .

**Remark 3.2** Let  $\hat{\mathcal{V}}$  be the set of  $\mathcal{P}(\mathbb{F}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable maps  $\hat{\nu}: \Omega^1 \times \mathbb{R}_+ \times \mathcal{A} \rightarrow (0, \infty)$ , which are bounded from above  $\sup_{\Omega^1 \times \mathbb{R}_+ \times \mathcal{A}} \hat{\nu} < \infty$ , but not necessarily bounded away from zero. For every  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ , we define

$$\hat{V}^{\mathcal{R}}(t, x, \pi) = \sup_{\hat{\nu} \in \hat{\mathcal{V}}} J^{\mathcal{R}}(t, x, \pi, \hat{\nu})$$

In [4] the randomized control problem is formulated over  $\hat{\mathcal{V}}$ . Here we considered  $\mathcal{V}$  because this set is more convenient for the proof of Theorem 3.1. However, notice that

$$V^{\mathcal{R}}(t, x, \pi) = \hat{V}^{\mathcal{R}}(t, x, \pi). \quad (3.9)$$

Indeed, clearly we have  $\mathcal{V} \subset \hat{\mathcal{V}}$ , so that  $V^{\mathcal{R}}(t, x, \pi) \leq \hat{V}^{\mathcal{R}}(t, x, \pi)$ . On the other hand, let  $\hat{\nu} \in \hat{\mathcal{V}}$  and define  $\nu^\varepsilon = \hat{\nu} \vee \varepsilon$ , for every  $\varepsilon \in (0, 1)$ . Observe that  $\nu^\varepsilon \in \mathcal{V}$  and  $\bar{\kappa}_T^{\nu^\varepsilon}$  converges pointwise  $\bar{\mathbb{P}}$ -a.s. to  $\bar{\kappa}_T^{\hat{\nu}}$ . Then, it is easy to see that

$$J^{\mathcal{R}}(t, x, \pi, \nu^\varepsilon) = \mathbb{E} \left[ \bar{\kappa}_T^{\nu^\varepsilon} \left( \int_t^T f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \right) \right] \xrightarrow{\varepsilon \rightarrow 0^+} J^{\mathcal{R}}(t, x, \pi, \hat{\nu}).$$

This implies that  $J^{\mathcal{R}}(t, x, \pi, \hat{\nu}) \leq \sup_{\nu \in \mathcal{V}} J^{\mathcal{R}}(t, x, \pi, \nu)$ , from which we get the other inequality  $\hat{V}^{\mathcal{R}}(t, x, \pi) \leq V^{\mathcal{R}}(t, x, \pi)$ , and identity (3.9) follows.  $\square$

We can now prove one of the main results of the paper, namely the equivalence of the two value functions  $V$  and  $V^{\mathcal{R}}$ .

**Theorem 3.1** *Under Assumption (A1), the value function  $V$  in (2.7) of the McKean-Vlasov control problem coincides with the value function  $V^{\mathcal{R}}$  in (3.8) of the randomized problem:*

$$V(t, x, \pi) = V^{\mathcal{R}}(t, x, \pi),$$

for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ .

**Remark 3.3** As an immediate consequence of Theorem 3.1, we see that  $V^{\mathcal{R}}$  does not depend on  $a_0$  and  $\lambda$ , since  $V$  does not depend on them.  $\square$

**Proof (of Theorem 3.1).** Fix  $(t, x, \xi) \in [0, T] \times \mathbb{R}^n \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\mathbb{P}$ . Set  $\bar{\xi}(\omega, \omega^1) := \xi(\omega)$ , then  $\bar{\xi} \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  and  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ . We split the proof of the equality  $V(t, x, \pi) = V^{\mathcal{R}}(t, x, \pi)$  into three steps, that we now summarize:

- I) In step I we prove that the value of the randomized problem does not change if we formulate the randomized McKean-Vlasov control problem on a new probability space.
- II) Step II is devoted to the proof of the first inequality  $V(t, x, \pi) \geq V^{\mathcal{R}}(t, x, \pi)$ .
  - 1) In order to prove it, we construct in substep 1 a new probability space  $(\check{\Omega}, \check{\mathcal{F}}, \check{\mathbb{P}})$  for the randomized problem, which is a product space of  $(\Omega, \mathcal{F}, \mathbb{P})$  and a canonical space supporting the Poisson random measure. Step I guarantees that the value of the new randomized problem is still given by  $V^{\mathcal{R}}(t, x, \pi)$ .



- 2) In substep 2 we prove that the value of the original McKean-Vlasov control problem is still equal to  $V(t, x, \pi)$  if we enlarge the class of admissible controls, taking all  $\check{\alpha}: \check{\Omega} \times [0, T] \rightarrow A$  which are progressive with respect to the filtration  $\check{\mathbb{F}}^{B, \mu_\infty}$ . The new class of admissible controls is denoted  $\check{\mathcal{A}}^{B, \mu_\infty}$ .
- 3) In substep 3 we conclude the proof of the inequality  $V(t, x, \pi) \geq V^{\mathcal{R}}(t, x, \pi)$ , proving that for every  $\check{\nu} \in \check{\mathcal{V}}$  there exists  $\check{\alpha}^{\check{\nu}} \in \check{\mathcal{A}}^{B, \mu_\infty}$  such that  $\check{J}^{\mathcal{R}}(t, x, \pi, \check{\nu}) = \check{J}(t, x, \pi, \check{\alpha}^{\check{\nu}})$ . From substep 2, we immediately deduce that  $V(t, x, \pi) \geq V^{\mathcal{R}}(t, x, \pi)$ .

III) Step III is devoted to the proof of the other inequality  $V(t, x, \pi) \leq V^{\mathcal{R}}(t, x, \pi)$ . In few words, we prove that the set  $\{\check{\alpha}^{\check{\nu}}: \check{\nu} \in \check{\mathcal{V}}\}$  is dense in  $\check{\mathcal{A}}^{B, \mu_\infty}$  with respect to the distance  $\tilde{\rho}$  in (3.1). Then, the claim follows from the stability Lemma C.1.

**Step I.** *Value of the randomized McKean-Vlasov control problem.* Consider another probabilistic setting for the randomized problem, defined starting from  $(\Omega, \mathcal{F}, \mathbb{P})$ , along the same lines as in Section 3, where the objects  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ ,  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$ ,  $\bar{\mathcal{G}}, \bar{B}, \bar{\mu}, T_n, \mathcal{A}_n, \bar{I}, \bar{X}^{t, \xi}, \bar{X}^{t, x, \pi}, \mathcal{V}, J^{\mathcal{R}}(t, x, \pi, \nu), V^{\mathcal{R}}(t, x, \pi)$  are replaced respectively by  $(\check{\Omega}^1, \check{\mathcal{F}}^1, \check{\mathbb{P}}^1)$ ,  $(\check{\Omega}, \check{\mathcal{F}}, \check{\mathbb{P}})$ ,  $\check{\mathcal{G}}, \check{B}, \check{\mu}, \check{T}_n, \check{\mathcal{A}}_n, \check{I}, \check{X}^{t, \xi}, \check{X}^{t, x, \pi}, \check{\mathcal{V}}, \check{J}^{\mathcal{R}}(t, x, \pi, \check{\nu}), \check{V}^{\mathcal{R}}(t, x, \pi)$ , with  $\check{\xi}(\omega, \check{\omega}^1) := \xi(\omega)$ , so that  $\check{\xi} \in L^2(\check{\Omega}, \check{\mathcal{G}}, \check{\mathbb{P}}; \mathbb{R}^n)$  and  $\pi = \mathbb{P}_\xi$  under  $\check{\mathbb{P}}$ .

We claim that  $V^{\mathcal{R}}(t, x, \pi) = \check{V}^{\mathcal{R}}(t, x, \pi)$ . Let us prove  $V^{\mathcal{R}}(t, x, \pi) \leq \check{V}^{\mathcal{R}}(t, x, \pi)$ , the other inequality can be proved in a similar way. We begin noting that  $V^{\mathcal{R}}(t, x, \pi) \leq \check{V}^{\mathcal{R}}(t, x, \pi)$  follows if we prove that for every  $\nu \in \mathcal{V}$  there exists  $\check{\nu} \in \check{\mathcal{V}}$  such that  $J^{\mathcal{R}}(t, x, \pi, \nu) = \check{J}^{\mathcal{R}}(t, x, \pi, \check{\nu})$ . Observe that

$$J^{\mathcal{R}}(t, x, \pi, \nu) = \mathbb{E} \left[ \bar{\kappa}_T^\nu \left( \int_t^T f(s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_{\bar{X}_s^{t, \xi}}^{\bar{\mu}}) ds + g(\bar{X}_T^{t, x, \pi}, \mathbb{P}_{\bar{X}_T^{t, \xi}}^{\bar{\mu}}) \right) \right].$$

The quantity  $J^{\mathcal{R}}(t, x, \pi, \nu)$  depends only on the joint law of  $\bar{\kappa}_T^\nu, \bar{X}^{t, x, \pi}, \mathbb{P}_{\bar{X}^{t, \xi}}^{\bar{\mu}}, \bar{I}$  under  $\bar{\mathbb{P}}$ , which in turn depends on the joint law of  $\bar{B}, \bar{\mu}, \bar{\nu}$  under  $\bar{\mathbb{P}}$ .

Recall that  $\bar{\nu}_t(\omega, \omega^1, \alpha) := \nu_t(\omega^1, \alpha)$  and  $\nu$  is  $\mathcal{P}(\mathbb{F}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable. Then, we can suppose, using a monotone class argument, that  $\nu$  is given by

$$\nu_s(\alpha) = k(\alpha) 1_{(T_n, T_{n+1}]}(s) \Psi(s, T_1, \dots, T_n, \mathcal{A}_1, \dots, \mathcal{A}_n),$$

for some bounded and positive Borel-measurable maps  $k$  and  $\Psi$ . We then see that  $\check{\nu}$  defined by

$$\check{\nu}_s(\alpha) := k(\alpha) 1_{(\check{T}_n, \check{T}_{n+1}]}(s) \Psi(s, \check{T}_1, \dots, \check{T}_n, \check{\mathcal{A}}_1, \dots, \check{\mathcal{A}}_n)$$

is such that  $J^{\mathcal{R}}(t, x, \pi, \nu) = \check{J}^{\mathcal{R}}(t, x, \pi, \check{\nu})$ .

**Step II.** *Proof of the inequality  $V(t, x, \pi) \geq V^{\mathcal{R}}(t, x, \pi)$ .* We shall exploit Proposition 4.1 in [4], for which we need to introduce a specific probabilistic setting for the randomized problem.

**Substep 1.** *Canonical probabilistic setting for the randomized McKean-Vlasov control problem.* Recall that the Polish space  $\mathcal{A}$  can be countable or uncountable, and in this latter case it is Borel-isomorphic to  $\mathbb{R}$  (see Corollary 7.16.1 in [7]). Then, in both cases, it can be proved (see the beginning of Section 4.1 in [4]) that there exists a surjective measurable map  $\iota: \mathbb{R} \rightarrow \mathcal{A}$  and a finite positive measure  $\lambda'$  on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  with full topological support, such that  $\lambda = \lambda' \circ \iota^{-1}$  and  $\lambda'$  is diffuse, namely  $\lambda'(\{r\}) = 0$  for every  $r \in \mathbb{R}$ .

Now, consider the canonical probability space  $(\Omega', \mathcal{F}', \mathbb{P}')$  of a marked point process on  $\mathbb{R}_+ \times \mathbb{R}$  associated to a Poisson random measure with compensator  $\lambda'(dr) dt$ . In other words,  $\omega' \in \Omega'$  is a double sequence  $\omega' = (t_n, r_n)_{n \geq 1} \subset (0, \infty) \times \mathbb{R}$ , with  $t_n < t_{n+1} \nearrow \infty$ . We denote by  $(T'_n, R'_n)_{n \geq 1}$  the canonical marked point process defined as  $(T'_n(\omega'), R'_n(\omega')) = (t_n, r_n)$ , and by  $\zeta' = \sum_{n \geq 1} \delta_{(T'_n, R'_n)}$  the canonical random measure.  $\mathcal{F}'$  is the  $\sigma$ -algebra generated by the sequence  $(T'_n, R'_n)_{n \geq 1}$ .  $\mathbb{P}'$  is the unique probability on  $\mathcal{F}'$  under which  $\zeta'$  has compensator  $\lambda'(dr) ds$ . Finally, we complete  $(\Omega', \mathcal{F}', \mathbb{P}')$  and, to simplify the notation, we still denote its completion by  $(\Omega', \mathcal{F}', \mathbb{P}')$ .

Set  $\mathcal{A}'_n = \iota(R'_n)$  and  $\mu' = \sum_{n \geq 1} \delta_{(T'_n, \mathcal{A}'_n)}$ . Then  $\mu'$  is a Poisson random measure on  $(\Omega', \mathcal{F}', \mathbb{P}')$  with compensator  $\lambda(d\alpha) ds$ . Proceeding along the same lines as in Section 3, we define, starting from  $(\Omega, \mathcal{F}, \mathbb{P})$  and  $(\Omega', \mathcal{F}', \mathbb{P}')$ , a new setting for the randomized problem where the objects  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ ,  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$ ,  $\bar{\mathcal{G}}, \bar{B}, \bar{\mu}, \bar{\mathbb{F}}^B = (\bar{\mathcal{F}}_s^B)_{s \geq 0}$ ,  $\bar{\mathbb{F}}^\mu = (\bar{\mathcal{F}}_s^\mu)_{s \geq 0}$ ,  $\bar{\mathbb{F}}^{B, \mu} = (\bar{\mathcal{F}}_s^{B, \mu})_{s \geq 0}$ ,  $(T_n, \mathcal{A}_n)_{n \geq 1}$ ,  $\bar{I}, \bar{X}^{t, \xi}, \bar{X}^{t, x, \pi}, \bar{\nu}, \bar{\mathbb{P}}^\nu, \bar{J}^R(t, x, \pi, \nu), \bar{V}^R(t, x, \pi)$  are replaced respectively by  $(\Omega', \mathcal{F}', \mathbb{P}')$ ,  $(\check{\Omega}, \check{\mathcal{F}}, \check{\mathbb{P}})$ ,  $\check{\mathcal{G}}, \check{B}, \check{\mu}, \check{\mathbb{F}}^B = (\check{\mathcal{F}}_s^B)_{s \geq 0}$ ,  $\check{\mathbb{F}}^{\mu'} = (\check{\mathcal{F}}_s^{\mu'})_{s \geq 0}$ ,  $\check{\mathbb{F}}^{B, \mu} = (\check{\mathcal{F}}_s^{B, \mu})_{s \geq 0}$ ,  $(\check{T}_n, \check{\mathcal{A}}_n)_{n \geq 1}$ ,  $\check{I}, \check{X}^{t, \xi}, \check{X}^{t, x, \pi}, \check{\nu}, \check{\mathbb{P}}^{\check{\nu}}, \check{\mathbb{P}}^{\check{\nu}}, \check{J}^R(t, x, \pi, \check{\nu}), \check{V}^R(t, x, \pi)$ , with  $\check{\xi}(\omega, \omega') := \xi(\omega)$ , so that  $\check{\xi} \in L^2(\check{\Omega}, \check{\mathcal{G}}, \check{\mathbb{P}}; \mathbb{R}^n)$  and  $\pi = \mathbb{P}_{\check{\xi}}$  under  $\check{\mathbb{P}}$ .

**Substep 2.** *Value of the original McKean-Vlasov control problem.*  $\check{\mathbb{F}}^{B, \mu_\infty} = (\check{\mathcal{F}}_s^{B, \mu_\infty})_{s \geq 0}$  be the  $\check{\mathbb{P}}$ -completion of the filtration  $(\mathcal{F}_s^B \otimes \mathcal{F}')_{s \geq 0}$ , and  $\check{\mathcal{F}}'$  the canonical extension of  $\mathcal{F}'$  to  $\check{\Omega}$ . We define the set  $\check{\mathcal{A}}^{B, \mu_\infty}$  of all  $\check{\mathbb{F}}^{B, \mu_\infty}$ -progressive processes  $\check{\alpha}: \check{\Omega} \times [0, T] \rightarrow A$ . For every  $\check{\alpha} \in \check{\mathcal{A}}^{B, \mu_\infty}$ , we denote  $(\check{X}_s^{t, \xi, \check{\alpha}}, \check{X}_s^{t, x, \pi, \check{\alpha}})_{s \in [t, T]}$  the unique continuous  $(\check{\mathcal{F}}_s^{B, \mu_\infty} \vee \check{\mathcal{G}})_s$ -adapted solution to the following system of equations:

$$d\check{X}_s^{t, \xi, \check{\alpha}} = b(s, \check{X}_s^{t, \xi, \check{\alpha}}, \mathbb{P}_{\check{X}_s^{t, \xi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}, \check{\alpha}_s) ds + \sigma(s, \check{X}_s^{t, \xi, \check{\alpha}}, \mathbb{P}_{\check{X}_s^{t, \xi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}, \check{\alpha}_s) d\check{B}_s, \quad \check{X}_t^{t, \xi, \check{\alpha}} = \check{\xi}, \quad (3.10)$$

$$d\check{X}_s^{t, x, \pi, \check{\alpha}} = b(s, \check{X}_s^{t, x, \pi, \check{\alpha}}, \mathbb{P}_{\check{X}_s^{t, x, \pi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}, \check{\alpha}_s) ds + \sigma(s, \check{X}_s^{t, x, \pi, \check{\alpha}}, \mathbb{P}_{\check{X}_s^{t, x, \pi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}, \check{\alpha}_s) d\check{B}_s, \quad \check{X}_t^{t, x, \pi, \check{\alpha}} = x, \quad (3.11)$$

for all  $s \in [t, T]$ , where  $\mathbb{P}_{\check{X}_s^{t, \xi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}$  denotes the regular conditional distribution of the random variable  $\check{X}_s^{t, \xi, \check{\alpha}}: \check{\Omega} \rightarrow \mathbb{R}^n$  with respect to  $\check{\mathcal{F}}_s^\mu$ . We also define ( $\check{\mathbb{E}}$  denotes the  $\check{\mathbb{P}}$ -expected value)

$$\check{J}(t, x, \pi, \check{\alpha}) = \check{\mathbb{E}} \left[ \int_t^T f(s, \check{X}_s^{t, x, \pi, \check{\alpha}}, \mathbb{P}_{\check{X}_s^{t, x, \pi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}, \check{\alpha}_s) ds + g(\check{X}_T^{t, x, \pi, \check{\alpha}}, \mathbb{P}_{\check{X}_T^{t, x, \pi, \check{\alpha}}}^{\check{\mathcal{F}}_T^\mu}) \right],$$

and

$$\check{V}(t, x, \pi) = \sup_{\check{\alpha} \in \check{\mathcal{A}}^{B, \mu_\infty}} \check{J}(t, x, \pi, \check{\alpha}).$$

Let us prove that  $V(t, x, \pi) = \check{V}(t, x, \pi)$ .

The inequality  $V(t, x, \pi) \leq \check{V}(t, x, \pi)$  is obvious. Indeed, every  $\alpha \in \mathcal{A}$  admits an obvious extension  $\check{\alpha}(\omega, \omega') := \alpha(\omega)$  to  $\check{\Omega}$ . Notice that  $\check{\alpha} \in \check{\mathcal{A}}^{B, \mu_\infty}$ . We also observe that  $\check{X}_s^{t, \xi, \check{\alpha}}(\omega, \omega') = X_s^{t, \xi, \alpha}(\omega)$ , for  $\check{\mathbb{P}}$ -almost every  $(\omega, \omega') \in \check{\Omega}$ . Therefore  $\mathbb{P}_{\check{X}_s^{t, \xi, \check{\alpha}}}^{\check{\mathcal{F}}_s^\mu}$  is equal  $\check{\mathbb{P}}$ -a.s. to  $\mathbb{P}_{X_s^{t, \xi, \alpha}}$ . Then,  $\check{X}_s^{t, x, \pi, \check{\alpha}}(\omega, \omega') = X_s^{t, x, \pi, \alpha}(\omega)$ , for  $\check{\mathbb{P}}$ -almost every  $(\omega, \omega') \in \check{\Omega}$ . As a consequence, we see that  $\check{J}(t, x, \pi, \check{\alpha}) = J(t, x, \pi, \alpha)$ .

To prove the other inequality, let  $\check{\alpha} \in \check{\mathcal{A}}^{B, \mu_\infty}$ . Then, there exists an  $A$ -valued  $(\mathcal{F}_s^B \otimes \mathcal{F}')_{s \geq 0}$ -progressive process  $\check{\alpha}: \check{\Omega} \times [0, T] \rightarrow A$  satisfying  $\check{\alpha} = \check{\alpha}$ ,  $d\check{\mathbb{P}} ds$ -a.e., so that  $\check{J}(t, x, \pi, \check{\alpha}) = J(t, x, \pi, \check{\alpha})$ . Moreover, for every  $\omega' \in \Omega'$  the process  $\alpha^{\omega'}$ , given by  $\alpha_s^{\omega'}(\omega) := \check{\alpha}_s(\omega, \omega')$ , is  $\mathbb{F}^B$ -progressive.

Now, for every  $\omega' \in \Omega'$ , consider the solution  $(X_s^{t,\xi,\alpha^{\omega'}}, X_s^{t,x,\pi,\alpha^{\omega'}})_{s \in [t,T]}$  to (2.3)-(2.4) with  $\alpha$  replaced by  $\alpha^{\omega'}$ , namely

$$\begin{aligned} dX_s^{t,\xi,\alpha^{\omega'}} &= b(s, X_s^{t,\xi,\alpha^{\omega'}}, \mathbb{P}_{X_s^{t,\xi,\alpha^{\omega'}}}, \alpha_s^{\omega'}) ds + \sigma(s, X_s^{t,\xi,\alpha^{\omega'}}, \mathbb{P}_{X_s^{t,\xi,\alpha^{\omega'}}}, \alpha_s^{\omega'}) dB_s, \\ dX_s^{t,x,\pi,\alpha^{\omega'}} &= b(s, X_s^{t,x,\pi,\alpha^{\omega'}}, \mathbb{P}_{X_s^{t,\xi,\alpha^{\omega'}}}, \alpha_s^{\omega'}) ds + \sigma(s, X_s^{t,x,\pi,\alpha^{\omega'}}, \mathbb{P}_{X_s^{t,\xi,\alpha^{\omega'}}}, \alpha_s^{\omega'}) dB_s. \end{aligned}$$

On the other hand, since  $(\check{X}_s^{t,\xi,\check{\alpha}}, \check{X}_s^{t,x,\pi,\check{\alpha}})_{s \in [t,T]}$  is the solution to (3.10)-(3.11), we have, for  $\mathbb{P}'$ -a.e.  $\omega' \in \Omega'$ ,

$$\begin{aligned} d\check{X}_s^{t,\xi,\check{\alpha}}(\cdot, \omega') &= b(s, \check{X}_s^{t,\xi,\check{\alpha}}(\cdot, \omega'), \mathbb{P}_{\check{X}_s^{t,\xi,\check{\alpha}}}^{\check{\mu}}(\cdot, \omega'), \check{\alpha}_s(\cdot, \omega')) ds \\ &\quad + \sigma(s, \check{X}_s^{t,\xi,\check{\alpha}}(\cdot, \omega'), \mathbb{P}_{\check{X}_s^{t,\xi,\check{\alpha}}}^{\check{\mu}}(\cdot, \omega'), \check{\alpha}_s(\cdot, \omega')) dB_s, \\ d\check{X}_s^{t,x,\pi,\check{\alpha}}(\cdot, \omega') &= b(s, \check{X}_s^{t,x,\pi,\check{\alpha}}(\cdot, \omega'), \mathbb{P}_{\check{X}_s^{t,\xi,\check{\alpha}}}^{\check{\mu}}(\cdot, \omega'), \check{\alpha}_s(\cdot, \omega')) ds \\ &\quad + \sigma(s, \check{X}_s^{t,x,\pi,\check{\alpha}}(\cdot, \omega'), \mathbb{P}_{\check{X}_s^{t,\xi,\check{\alpha}}}^{\check{\mu}}(\cdot, \omega'), \check{\alpha}_s(\cdot, \omega')) dB_s. \end{aligned}$$

Notice that, for  $\mathbb{P}'$ -a.e.  $\omega' \in \Omega'$  we have that  $\mathbb{P}_{\check{X}_s^{t,\xi,\check{\alpha}}}^{\check{\mu}}(\cdot, \omega')$  is equal  $\mathbb{P}$ -a.s. to  $\mathbb{P}_{\check{X}_s^{t,\xi,\check{\alpha}}(\cdot, \omega')}$ , the law under  $\mathbb{P}$  of the random variable  $\check{X}_s^{t,\xi,\check{\alpha}}(\cdot, \omega') : \Omega \rightarrow \mathbb{R}^n$

Recalling the identity  $\alpha_s^{\omega'} = \check{\alpha}_s(\cdot, \omega')$ , we see that, for  $\mathbb{P}'$ -a.e.  $\omega' \in \Omega'$ ,  $(X_s^{t,\xi,\alpha^{\omega'}}, X_s^{t,x,\pi,\alpha^{\omega'}})_{s \in [t,T]}$  and  $(\check{X}_s^{t,\xi,\check{\alpha}}(\cdot, \omega'), \check{X}_s^{t,x,\pi,\check{\alpha}}(\cdot, \omega'))_{s \in [t,T]}$  solve the same system of equations. Then, by path-wise uniqueness, for  $\mathbb{P}'$ -a.e.  $\omega' \in \Omega'$ , we have  $X_s^{t,\xi,\alpha^{\omega'}}(\omega) = \check{X}_s^{t,\xi,\check{\alpha}}(\omega, \omega')$  and  $X_s^{t,x,\pi,\alpha^{\omega'}}(\omega) = \check{X}_s^{t,x,\pi,\check{\alpha}}(\omega, \omega')$ , for all  $s \in [t, T]$ ,  $\mathbb{P}(d\omega)$ -almost surely. Therefore, by Fubini's theorem,

$$\begin{aligned} \check{J}(t, x, \pi, \check{\alpha}) &= \int_{\Omega'} \mathbb{E} \left[ \int_t^T f(s, X_s^{t,x,\pi,\alpha^{\omega'}}, \mathbb{P}_{X_s^{t,\xi,\alpha^{\omega'}}}, \alpha_s^{\omega'}) ds + g(X_T^{t,x,\pi,\alpha^{\omega'}}, \mathbb{P}_{X_T^{t,\xi,\alpha^{\omega'}}}) \right] \mathbb{P}'(d\omega') \\ &= \int_{\Omega'} J(t, x, \pi, \alpha^{\omega'}) \mathbb{P}'(d\omega') \leq V(t, x, \pi). \end{aligned}$$

Recalling that  $\check{J}(t, x, \pi, \check{\alpha}) = \check{J}(t, x, \pi, \check{\alpha})$ , we deduce that  $\check{J}(t, x, \pi, \check{\alpha}) \leq V(t, x, \pi)$ . Taking the supremum over  $\check{\alpha} \in \check{\mathcal{A}}^{B,\mu\infty}$ , we conclude that  $\check{V}(t, x, \pi) \leq V(t, x, \pi)$ .

**Substep 3.** *Proof of the inequality  $V(t, x, \pi) \geq V^R(t, x, \pi)$ .* Let  $\check{\nu} \in \check{\mathcal{V}}$ . By Lemma 4.3 in [4] there exists a sequence  $(\check{T}_n^{\check{\nu}}, \check{\mathcal{A}}_n^{\check{\nu}})_{n \geq 1}$  on  $(\Omega', \mathcal{F}', \mathbb{P}')$  such that:

- $(\check{T}_n^{\check{\nu}}, \check{\mathcal{A}}_n^{\check{\nu}})$  takes values in  $(0, \infty) \times \mathcal{A}$ ;
- $\check{T}_n^{\check{\nu}} < \check{T}_{n+1}^{\check{\nu}} \nearrow \infty$ ;
- $\check{T}_n^{\check{\nu}}$  is an  $\mathbb{F}^{\mu'}$ -stopping time and  $\check{\mathcal{A}}_n^{\check{\nu}}$  is  $\mathcal{F}_{\check{T}_n^{\check{\nu}}}^{\mu'}$ -measurable;
- the law of  $(\check{T}_n^{\check{\nu}}, \check{\mathcal{A}}_n^{\check{\nu}})_{n \geq 1}$  under  $\mathbb{P}'$  coincides with the law of  $(\check{T}_n, \check{\mathcal{A}}_n)_{n \geq 1}$  under  $\mathbb{P}^{\check{\nu}}$ .

Let  $\check{\alpha}^{\check{\nu}} : \check{\Omega} \times [0, T] \rightarrow A$  be given by ( $\bar{\alpha}$  was introduced in (3.3))

$$\check{\alpha}_s^{\check{\nu}}(\omega, \omega') = \bar{\alpha}_s(\omega) 1_{[0, \check{T}_1^{\check{\nu}}(\omega'))}(s) + \sum_{n \geq 1} (\check{\mathcal{A}}_n^{\check{\nu}}(\omega'))_{s \wedge T}(\omega) 1_{[\check{T}_n^{\check{\nu}}(\omega'), \check{T}_{n+1}^{\check{\nu}}(\omega'))}(s).$$

Notice that  $\check{\alpha}^{\check{\nu}} \in \check{\mathcal{A}}^{B,\mu\infty}$ . For every  $n \geq 1$ , set  $\check{\alpha}_{n,s}(\omega, \omega') := (\check{\mathcal{A}}_n(\omega'))_s(\omega)$  and  $\check{\alpha}_{n,s}^{\check{\nu}}(\omega, \omega') := (\check{\mathcal{A}}_n^{\check{\nu}}(\omega'))_s(\omega)$ , for all  $s \in [0, T]$ . Notice that the law of  $(\check{\alpha}_{n,s})_{s \in [0, T]}$  under  $\mathbb{P}^{\check{\nu}}$  coincides with the

law of  $(\check{\alpha}_{n,s}^{\check{\nu}})_{s \in [0,T]}$  under  $\check{\mathbb{P}}$  (to see this, we can suppose, by an approximation argument, that the  $\mathcal{A}$ -valued random variables  $\check{\mathcal{A}}_n$  and  $\check{\mathcal{A}}_n^{\check{\nu}}$  take only a finite number of values). It follows that the law of  $\check{I}$  under  $\check{\mathbb{P}}^{\check{\nu}}$  coincides with the law of  $\check{\alpha}^{\check{\nu}}$  under  $\check{\mathbb{P}}$ .

More generally, for every  $n \geq 1$ , the law of  $(\check{\xi}, \check{B}, \check{\alpha}_n, \cdot)$  under  $\check{\mathbb{P}}^{\check{\nu}}$  is equal to the law of  $(\check{\xi}, \check{B}, \check{\alpha}_n^{\check{\nu}}, \cdot)$  under  $\check{\mathbb{P}}$ . Therefore, the law of  $(\check{\xi}, \check{B}, \check{I})$  under  $\check{\mathbb{P}}^{\check{\nu}}$  coincides with the law of  $(\check{\xi}, \check{B}, \check{\alpha}^{\check{\nu}})$  under  $\check{\mathbb{P}}$ . This implies that the law of  $(\check{X}^{t,\check{\xi}}, \check{X}^{t,x,\pi}, \check{I})$  under  $\check{\mathbb{P}}^{\check{\nu}}$  is equal to the law of  $(\check{X}^{t,\check{\xi},\check{\alpha}^{\check{\nu}}}, \check{X}^{t,x,\pi,\check{\alpha}^{\check{\nu}}}, \check{\alpha}^{\check{\nu}})$  under  $\check{\mathbb{P}}$ . It follows that  $\check{J}^{\mathcal{R}}(t, x, \pi, \check{\nu}) = \check{J}(t, x, \pi, \check{\alpha}^{\check{\nu}})$ . In particular, we have

$$\sup_{\check{\nu} \in \check{\mathcal{V}}} \check{J}^{\mathcal{R}}(t, x, \pi, \check{\nu}) = \sup_{\substack{\check{\alpha}^{\check{\nu}} \\ \check{\nu} \in \check{\mathcal{V}}}} \check{J}(t, x, \pi, \check{\alpha}^{\check{\nu}}).$$

Since the left-hand side is equal to  $\check{V}^{\mathcal{R}}(t, x, \pi)$ , while the right-hand side is clearly less than or equal to  $\check{V}(t, x, \pi)$ , we get  $\check{V}^{\mathcal{R}}(t, x, \pi) \leq \check{V}(t, x, \pi)$ . Recalling from step I that  $V^{\mathcal{R}}(t, x, \pi) = \check{V}^{\mathcal{R}}(t, x, \pi)$  and from substep 2 that  $\check{V}(t, x, \pi) = V(t, x, \pi)$ , we conclude  $V^{\mathcal{R}}(t, x, \pi) \leq V(t, x, \pi)$ .

**Step III.** *Proof of the inequality  $V(t, x, \pi) \leq V^{\mathcal{R}}(t, x, \pi)$ .* The proof of this step is based on Proposition A.1 in [4] (notice, however, that we will need to use some results from the proof of this Proposition, not only from its statement). More precisely, the set  $\Omega$  appearing in Proposition A.1 of [4] is the empty set  $\Omega = \emptyset$  in our context, so that the product probability space  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  coincides with  $(\Omega', \mathcal{F}', \mathbb{P}')$ , which is some suitably defined probability space (see Appendix A in [4] for the definition of  $(\Omega', \mathcal{F}', \mathbb{P}')$ ; here, we do not need to know the structure of  $(\Omega', \mathcal{F}', \mathbb{P}')$ ). Fix  $\hat{\alpha} \in \mathcal{A}$  and denote by  $\alpha: [0, T] \rightarrow \mathcal{A}$  the map  $\alpha_s = \hat{\alpha}$ , for every  $s \in [0, T]$ . By Proposition A.1 in [4] we have that, for every  $\ell \in \mathbb{N} \setminus \{0\}$ , there exists a marked point process  $(T_n^\ell, \mathcal{A}_n^\ell)_{n \geq 1}$  on  $(\Omega', \mathcal{F}', \mathbb{P}')$  such that ( $\bar{\alpha}$  was introduced in (3.3))

$$T_0^\ell = 0, \quad \mathcal{A}_0^\ell = \bar{\alpha}, \quad \mathcal{I}_s^\ell(\omega') = \sum_{n \geq 0} \mathcal{A}_n^\ell(\omega') 1_{[T_n^\ell(\omega'), T_{n+1}^\ell(\omega'))}(s), \quad \text{for all } s \geq 0$$

and

$$\mathbb{E}' \left[ \int_0^T \tilde{\rho}(\mathcal{I}_s^\ell, \alpha_s) ds \right] < \frac{1}{\ell}, \quad (3.12)$$

where  $\mathbb{E}'$  denotes the  $\mathbb{P}'$ -expected value. Set  $\mu_\ell = \sum_{n \geq 1} \delta_{(T_n^\ell, \mathcal{A}_n^\ell)}$  the random measure associated to  $(T_n^\ell, \mathcal{A}_n^\ell)_{n \geq 1}$ , and denote  $\mathbb{F}^{\mu_\ell} = (\mathcal{F}_s^{\mu_\ell})_{s \geq 0}$  the filtration generated by  $\mu_\ell$ . Then, by Proposition A.1 of [4] we have that the  $\mathbb{F}^{\mu_\ell}$ -compensator of  $\mu_\ell$  under  $\mathbb{P}'$  is given by  $\nu_s^\ell(\alpha) \lambda(d\alpha) ds$  for some  $\mathcal{P}(\mathbb{F}^{\mu_\ell}) \otimes \mathcal{B}(\mathcal{A})$ -measurable map  $\nu^\ell: \Omega' \times \mathbb{R}_+ \times \mathcal{A} \rightarrow \mathbb{R}_+$  satisfying

$$0 < \inf_{\Omega' \times [0, T] \times \mathcal{A}} \nu^\ell \leq \sup_{\Omega' \times [0, T] \times \mathcal{A}} \nu^\ell < \infty. \quad (3.13)$$

Noting that the definition of  $\nu^\ell$  on  $\Omega' \times (T, \infty) \times \mathcal{A}$  is not relevant in order to guarantee (3.12), we can assume that  $\nu^\ell \equiv 1$  on  $\Omega' \times (T, \infty) \times \mathcal{A}$ .

Observe that

$$\mathbb{E}' \left[ \int_0^T \tilde{\rho}(\mathcal{I}_s^\ell, \alpha_s) ds \right] = \sum_{n \geq 0} \mathbb{E}' \left[ 1_{\{T_n^\ell < T\}} \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \mathbb{E} \left[ \int_0^T \rho((\mathcal{A}_n^\ell)_r, \hat{\alpha}_r) dr \right] ds \right] < \frac{1}{\ell}.$$

On the other hand, let

$$\tilde{I}_s^\ell(\omega, \omega') = \sum_{n \geq 0} (\mathcal{A}_n^\ell(\omega'))_{s \wedge T}(\omega) 1_{[T_n^\ell(\omega'), T_{n+1}^\ell(\omega'))}(s), \quad \text{for all } s \geq 0.$$

Our aim is to prove that

$$\tilde{\rho}^{\mathbb{Q}}(\tilde{I}^\ell, \hat{\alpha}) := \mathbb{E}' \left[ \mathbb{E} \left[ \int_0^T \rho(\tilde{I}_r^\ell, \hat{\alpha}_r) dr \right] \right] \xrightarrow{\ell \rightarrow \infty} 0. \quad (3.14)$$

**Digression.** *Estimate for the series  $\sum_{n \geq 0} \mathbb{P}'(T_n^\ell < T)$ .* We recall from the proof of Proposition A.1 in [4] that the sequence  $(T_n^\ell)_{n \geq 0}$  is the disjoint union of  $(R_n^m)_{n \geq 1}$  and  $(T_n^k)_{n \geq 0}$  (we refer to the proof of Proposition A.1 in [4] for all unexplained notations), namely

$$\sum_{n \geq 0} \mathbb{P}'(T_n^\ell < T) = \sum_{n \geq 1} \mathbb{P}'(R_n^m < T) + \sum_{n \geq 0} \mathbb{P}'(T_n^k < T). \quad (3.15)$$

We also recall that  $T_n^k - T_{n-1}^k$  has an exponential distribution with parameter  $k^{-1}\lambda(\mathcal{A})$ . Then, it is easy to prove by induction on  $n$ , the estimate

$$\mathbb{P}'(T_n^k < T) \leq (1 - e^{-k^{-1}\lambda(\mathcal{A})T})^n. \quad (3.16)$$

On the other hand, concerning the sequence  $(R_n^m)_{n \geq 1}$ , we begin noting that since  $\alpha$  is constant and identically equal to  $\hat{\alpha}$ , the sequence of deterministic times  $(t_n)_{n \geq 0}$  appearing in the proof of Proposition A.1 in [4] can be taken as follows:  $t_0 = 0$ ,  $t_1 \in (0, \frac{1}{3\ell} \wedge T)$ , and  $t_n = T + n - 2$  for every  $n \geq 2$ . Therefore  $R_n^m \geq T$  for all  $n \geq 2$ , while  $R_1^m = t_1 + V_1^m$ , where  $V_1^m$  is an exponential random variable with parameter  $\lambda_{1m} > m$ . In particular, we have

$$\mathbb{P}'(R_1^m < T) = \mathbb{P}'(V_1^m < T - t_1) = 1 - e^{-\lambda_{1m}(T-t_1)} \leq 1. \quad (3.17)$$

Plugging (3.16) and (3.17) into (3.15), we obtain

$$\sum_{n \geq 0} \mathbb{P}'(T_n^\ell < T) \leq 1 + \sum_{n \geq 0} (1 - e^{-k^{-1}\lambda(\mathcal{A})T})^n \leq 1 + e^{k^{-1}\lambda(\mathcal{A})T} \leq 1 + e^{\lambda(\mathcal{A})T}. \quad (3.18)$$

**Continuation of the proof of Step III.** We can now prove (3.14). In particular, we have, using (3.18),

$$\begin{aligned} \tilde{\rho}^{\mathbb{Q}}(\tilde{I}^\ell, \hat{\alpha}) &= \mathbb{E}' \left[ \mathbb{E} \left[ \int_0^T \rho(\tilde{I}_r^\ell, \hat{\alpha}_r) dr \right] \right] = \sum_{n \geq 0} \mathbb{E}' \left[ 1_{\{T_n^\ell < T\}} \mathbb{E} \left[ \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \rho((\mathcal{A}_n^\ell)_r, \hat{\alpha}_r) dr \right] \right] \\ &= \sum_{n \geq 0} \mathbb{E}' \left[ 1_{\{T_n^\ell < T\}} \frac{1}{T_{n+1}^\ell \wedge T - T_n^\ell} \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \mathbb{E} \left[ \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \rho((\mathcal{A}_n^\ell)_r, \hat{\alpha}_r) dr \right] ds \right] \\ &= \sum_{n \geq 0} \mathbb{E}' \left[ 1_{\{T_{n+1}^\ell \wedge T - T_n^\ell \geq 1/\sqrt{\ell}\}} 1_{\{T_n^\ell < T\}} \frac{1}{T_{n+1}^\ell \wedge T - T_n^\ell} \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \mathbb{E} \left[ \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \rho((\mathcal{A}_n^\ell)_r, \hat{\alpha}_r) dr \right] ds \right] \\ &\quad + \sum_{n \geq 0} \mathbb{E}' \left[ 1_{\{T_{n+1}^\ell \wedge T - T_n^\ell < 1/\sqrt{\ell}\}} 1_{\{T_n^\ell < T\}} \frac{1}{T_{n+1}^\ell \wedge T - T_n^\ell} \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \mathbb{E} \left[ \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \rho((\mathcal{A}_n^\ell)_r, \hat{\alpha}_r) dr \right] ds \right] \\ &\leq \sqrt{\ell} \sum_{n \geq 0} \mathbb{E}' \left[ 1_{\{T_n^\ell < T\}} \int_{T_n^\ell}^{T_{n+1}^\ell \wedge T} \mathbb{E} \left[ \int_0^T \rho((\mathcal{A}_n^\ell)_r, \hat{\alpha}_r) dr \right] ds \right] + \frac{1}{\sqrt{\ell}} \sum_{n \geq 0} \mathbb{P}'(T_n^\ell < T) \\ &= \sqrt{\ell} \mathbb{E}' \left[ \int_0^T \tilde{\rho}(\mathcal{I}_s^\ell, \alpha_s) ds \right] + \frac{1}{\sqrt{\ell}} \sum_{n \geq 0} \mathbb{P}'(T_n^\ell < T) \end{aligned}$$

$$\leq \sqrt{\ell} \mathbb{E} \left[ \int_0^T \tilde{\rho}(\mathcal{I}_s^\ell, \alpha_s) ds \right] + \frac{1 + e^{\lambda(\mathcal{A})T}}{\sqrt{\ell}} \leq \frac{2 + e^{\lambda(\mathcal{A})T}}{\sqrt{\ell}},$$

which yields (3.14).

We consider now the product probability space  $(\Omega \times \Omega', \mathcal{F} \otimes \mathcal{F}', \mathbb{P} \otimes \mathbb{P}')$ , which we still denote  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  (by an abuse of notation, since according to Proposition A.1 in [4],  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  coincides with  $(\Omega', \mathcal{F}', \mathbb{P}')$ ). We complete the probability space  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  and, to simplify the notation, we still denote by  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  its completion. Let  $\tilde{\xi}$ ,  $\tilde{B}$ ,  $\tilde{\nu}^\ell$  be the canonical extensions of  $\xi$ ,  $B$ ,  $\nu^\ell$  to  $\tilde{\Omega}$ . On the other hand, we still denote by  $\mu_\ell$  the extension of  $\mu_\ell$  to  $\tilde{\Omega}$ . We denote by  $\tilde{\mu}_\ell(ds d\alpha) = \mu_\ell(ds d\alpha) - \tilde{\nu}_s^\ell(\alpha) \lambda(d\alpha) ds$  the compensated martingale measure associated to  $\mu^\ell$ . We also denote by  $\tilde{\mathbb{F}}^{B, \mu_\ell} = (\tilde{\mathcal{F}}_s^{B, \mu_\ell})_{s \geq 0}$  (resp.  $\tilde{\mathbb{F}}^{\mu_\ell} = (\tilde{\mathcal{F}}_s^{\mu_\ell})_{s \geq 0}$ ) the  $\mathbb{Q}$ -completion of the filtration generated by  $\tilde{B}$  and  $\mu_\ell$  (resp.  $\mu_\ell$ ). For every  $\ell \in \mathbb{N} \setminus \{0\}$ , we define the Doléans exponential

$$\tilde{\kappa}_s^\ell = \mathcal{E}_s \left( \int_0^\cdot \int_A (\tilde{\nu}_r^\ell(\alpha)^{-1} - 1) \tilde{\mu}_\ell(dr d\alpha) \right), \quad \text{for all } s \in [0, T].$$

By (3.13) we see that  $(\tilde{\kappa}_s^\ell)_{s \in [0, T]}$  is an  $\tilde{\mathbb{F}}^{B, \mu_\ell}$ -martingale under  $\mathbb{Q}$ , so that we can define on  $(\tilde{\Omega}, \tilde{\mathcal{F}})$  a probability  $\tilde{\mathbb{P}}_\ell$  equivalent to  $\mathbb{Q}$  by  $d\tilde{\mathbb{P}}_\ell = \tilde{\kappa}_T^\ell d\mathbb{Q}$ . By the Girsanov theorem,  $\mu_\ell$  has  $\tilde{\mathbb{F}}^{B, \mu_\ell}$ -compensator given by  $\lambda(d\alpha) ds$  under  $\tilde{\mathbb{P}}_\ell$ . Moreover,  $\tilde{B}$  remains a Brownian motion under  $\tilde{\mathbb{P}}_\ell$ , and  $\pi = \mathbb{P}_{\tilde{\xi}}$  under  $\tilde{\mathbb{P}}_\ell$ .

Let  $\tilde{\mathcal{G}}$  be the canonical extension of  $\mathcal{G}$  to  $\tilde{\Omega}$  and denote  $(\tilde{X}_s^{t, \tilde{\xi}, \ell}, \tilde{X}_s^{t, x, \pi, \ell})_{s \in [t, T]}$  the unique continuous  $(\tilde{\mathcal{F}}_s^{B, \mu_\ell} \vee \tilde{\mathcal{G}})$ -adapted solution to equations (3.4)-(3.5) on  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}}_\ell)$  with  $\tilde{\xi}$ ,  $\tilde{B}$ ,  $\tilde{I}$ ,  $\tilde{\mathcal{F}}_s^\mu$  replaced by  $\tilde{\xi}$ ,  $\tilde{B}$ ,  $\tilde{I}^\ell$ ,  $\tilde{\mathcal{F}}_s^{\mu_\ell}$ . Finally, we define in an obvious way the following objects:  $\tilde{\mathcal{V}}_\ell$ ,  $\tilde{\mathbb{P}}_\ell^\nu$ ,  $\tilde{\mathbb{E}}_\ell^\nu$ ,  $\tilde{J}_\ell^{\mathcal{R}}(t, x, \pi, \tilde{\nu})$ ,  $\tilde{V}_\ell^{\mathcal{R}}(t, x, \pi)$ .

For every  $\ell$  we have constructed a new probabilistic setting for the randomized problem, where the objects  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ ,  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  $\bar{\mathcal{G}}$ ,  $\bar{B}$ ,  $\bar{\mu}$ ,  $\bar{I}$ ,  $\bar{X}^{t, \bar{\xi}}$ ,  $\bar{X}^{t, x, \pi}$ ,  $\mathcal{V}$ ,  $J^{\mathcal{R}}(t, x, \pi, \nu)$ ,  $V^{\mathcal{R}}(t, x, \pi)$  are replaced respectively by  $(\Omega', \mathcal{F}', \mathbb{P}')$ ,  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}}_\ell)$ ,  $\tilde{\mathcal{G}}$ ,  $\tilde{B}$ ,  $\mu^\ell$ ,  $\tilde{I}^\ell$ ,  $\tilde{X}^{t, \tilde{\xi}, \ell}$ ,  $\tilde{X}^{t, x, \pi, \ell}$ ,  $\tilde{\mathcal{V}}_\ell$ ,  $\tilde{J}_\ell^{\mathcal{R}}(t, x, \pi, \tilde{\nu})$ ,  $\tilde{V}_\ell^{\mathcal{R}}(t, x, \pi)$ .

Now, let us prove that  $\tilde{J}_\ell^{\mathcal{R}}(t, x, \pi, \tilde{\nu}^\ell) \rightarrow J(t, x, \pi, \hat{\alpha})$  as  $\ell \rightarrow \infty$ . To this end, notice that  $\tilde{\mathbb{P}}_\ell^{\tilde{\nu}^\ell} \equiv \mathbb{Q}$ . Therefore  $\tilde{J}_\ell^{\mathcal{R}}(t, x, \pi, \tilde{\nu}^\ell)$  can be written in terms of  $\mathbb{E}^\mathbb{Q}$  as follows:

$$\tilde{J}_\ell^{\mathcal{R}}(t, x, \pi, \tilde{\nu}^\ell) = \mathbb{E}^\mathbb{Q} \left[ \int_t^T f(s, \tilde{X}_s^{t, x, \pi, \ell}, \mathbb{P}_{\tilde{X}_s^{t, \tilde{\xi}, \ell}}^{\tilde{\mathcal{F}}_s^{\mu_\ell}}, \tilde{I}_s^\ell) ds + g(\tilde{X}_T^{t, x, \pi, \ell}, \mathbb{P}_{\tilde{X}_T^{t, \tilde{\xi}, \ell}}^{\tilde{\mathcal{F}}_T^{\mu_\ell}}) \right].$$

On the other hand, let  $\tilde{\mathbb{F}}^B = (\tilde{\mathcal{F}}_s^B)_{s \geq 0}$  be the  $\mathbb{Q}$ -completion of the filtration generated by  $\tilde{B}$ , and  $\tilde{\alpha}$  the canonical extension of  $\hat{\alpha}$  to  $\tilde{\Omega}$ . Then, we denote by  $(\tilde{X}_s^{t, \tilde{\xi}, \tilde{\alpha}}, \tilde{X}_s^{t, x, \pi, \tilde{\alpha}})_{s \in [t, T]}$  the unique continuous  $(\tilde{\mathcal{F}}_s^B \vee \tilde{\mathcal{G}})$ -adapted solution to equations (2.3)-(2.4) on  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  with  $\xi$ ,  $B$ ,  $\alpha$  replaced by  $\tilde{\xi}$ ,  $\tilde{B}$ ,  $\tilde{\alpha}$ . Notice that  $(\tilde{X}_s^{t, \tilde{\xi}, \tilde{\alpha}}, \tilde{X}_s^{t, x, \pi, \tilde{\alpha}})_{s \in [t, T]}$  coincides with the obvious extension of  $(X_s^{t, \xi, \hat{\alpha}}, X_s^{t, x, \pi, \hat{\alpha}})_{s \in [t, T]}$  to  $\tilde{\Omega}$ . Hence, we have

$$J(t, x, \pi, \hat{\alpha}) = \mathbb{E}^\mathbb{Q} \left[ \int_t^T f(s, \tilde{X}_s^{t, x, \pi, \tilde{\alpha}}, \mathbb{P}_{\tilde{X}_s^{t, \tilde{\xi}, \tilde{\alpha}}}^{\tilde{\mathcal{F}}_s^{\mu_\ell}}, \tilde{\alpha}_s) ds + g(\tilde{X}_T^{t, x, \pi, \tilde{\alpha}}, \mathbb{P}_{\tilde{X}_T^{t, \tilde{\xi}, \tilde{\alpha}}}^{\tilde{\mathcal{F}}_T^{\mu_\ell}}) \right].$$

Then, it follows that  $\tilde{J}_\ell^{\mathcal{R}}(t, x, \pi, \tilde{\nu}^\ell) \rightarrow J(t, x, \pi, \hat{\alpha})$  as  $\ell \rightarrow \infty$ . Indeed, this is a direct consequence of Lemma C.1, with  $\tilde{\mathbb{F}}^{\mu_0} := (\{\emptyset, \tilde{\Omega}\})_{s \geq 0}$  being the trivial filtration,  $\tilde{\mathbb{F}}^\ell := (\tilde{\mathcal{F}}_s^{B, \mu_\ell} \vee \tilde{\mathcal{G}})_{s \geq 0}$  for every  $\ell \geq 1$ ,  $\tilde{\mathbb{F}}^0 := (\tilde{\mathcal{F}}_s^B \vee \tilde{\mathcal{G}})_{s \geq 0}$ ,  $\tilde{I}^0 := \tilde{\alpha}$ ,  $\tilde{X}^{t, \tilde{\xi}, 0} := \tilde{X}^{t, \tilde{\xi}, \tilde{\alpha}}$ , and  $\tilde{X}^{t, x, \pi, 0} := \tilde{X}^{t, x, \pi, \tilde{\alpha}}$ .

We conclude that for every  $\varepsilon > 0$  there exists some  $L_\varepsilon \in \mathbb{N}$  such that, for every  $\ell > L_\varepsilon$ , we have

$$J(t, x, \pi, \hat{\alpha}) - \varepsilon \leq \tilde{J}_\ell^\mathcal{R}(t, x, \pi, \tilde{\nu}^\ell) \leq \sup_{\tilde{\nu} \in \tilde{\mathcal{V}}_\ell} \tilde{J}_\ell^\mathcal{R}(t, x, \pi, \tilde{\nu}) =: \tilde{V}_\ell^\mathcal{R}(t, x, \pi) \stackrel{\text{Step I}}{\downarrow} V^\mathcal{R}(t, x, \pi).$$

From the arbitrariness of  $\varepsilon$ , we see that  $J(t, x, \pi, \hat{\alpha}) \leq V^\mathcal{R}(t, x, \pi)$ . The claim follows taking the supremum over  $\hat{\alpha} \in \mathcal{A}$ .  $\square$

**Remark 3.4** Let  $\mathcal{V}_{1,t} \subset \mathcal{V}$  be the set of  $\nu \in \mathcal{V}$  such that  $\nu \equiv 1$  on  $\Omega \times [0, t) \times \mathcal{A}$ . Then

$$V(t, x, \pi) = \sup_{\nu \in \mathcal{V}_{1,t}} J^\mathcal{R}(t, x, \pi, \nu), \quad (3.19)$$

for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ . Indeed, by step II of the proof of Theorem 3.1, we have  $V(t, x, \pi) \geq V^\mathcal{R}(t, x, \pi) \geq \sup_{\nu \in \mathcal{V}_{1,t}} J^\mathcal{R}(t, x, \pi, \nu)$ . Let us prove the other inequality. We begin noting that in Lemma C.1, the convergence  $\mathbb{E}^\mathbb{Q}[\int_t^T \tilde{\rho}(\tilde{I}_s^\ell, \tilde{I}_s^0) ds] \rightarrow 0$  as  $\ell \rightarrow \infty$  is needed, rather than  $\mathbb{E}^\mathbb{Q}[\int_0^T \tilde{\rho}(\tilde{I}_s^\ell, \tilde{I}_s^0) ds] \rightarrow 0$ . In other words, the behavior of  $(\tilde{I}_s^\ell)_{s \in [0, t]}$  on the interval  $[0, t)$  is not relevant. Therefore, proceeding as in step III of the proof of Theorem 3.1, we see that we can take  $\tilde{\nu}^\ell \equiv 1$  on  $\tilde{\Omega} \times [0, t) \times \mathcal{A}$ , in order to guarantee the convergence  $\mathbb{E}^\mathbb{Q}[\int_t^T \tilde{\rho}(\tilde{I}_s^\ell, \hat{\alpha}_s) ds] \rightarrow 0$  as  $\ell \rightarrow \infty$ . Then, from the same proof of Lemma C.1, we conclude that  $\tilde{J}^\mathcal{R}(t, x, \pi, \tilde{\nu}^\ell) \rightarrow J(t, x, \pi, \hat{\alpha})$  as  $\ell \rightarrow \infty$ . This implies the validity of the other inequality  $V(t, x, \pi) \leq \sup_{\nu \in \mathcal{V}_{1,t}} J^\mathcal{R}(t, x, \pi, \nu)$  and proves (3.19).  $\square$

## 4 Feynman-Kac representation: randomized equation

In the present section we introduce, for every  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , a forward-backward stochastic differential system of equations, which provides a probabilistic representation for the value  $V(t, x, \pi)$ , with  $\pi = \mathbb{P}_\xi$  under  $\bar{\mathbb{P}}$ . In other words, we derive a nonlinear Feynman-Kac formula for the value function  $V$  in (2.7) of the McKean-Vlasov control problem.

We firstly introduce the following spaces, for every  $t \in [0, T]$ .

- $\mathcal{S}^2(t, T)$ , the set of real-valued càdlàg  $\mathbb{F}^\mu$ -adapted processes  $Y = (Y_s)_{s \in [t, T]}$ , with  $Y: \Omega^1 \times [t, T] \rightarrow \mathbb{R}$ , satisfying  $\|Y\|_{\mathcal{S}^2(t, T)}^2 := \mathbb{E}^1[\sup_{t \leq s \leq T} |Y_s|^2] < \infty$ .
- $L_\mu^2(t, T)$ , the set of real-valued  $\mathcal{P}(\mathbb{F}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable maps  $U = (U_s(\alpha))_{s \in [t, T], \alpha \in \mathcal{A}}$ , with  $U: \Omega^1 \times [t, T] \times \mathcal{A} \rightarrow \mathbb{R}$ , satisfying  $\|U\|_{L_\mu^2(t, T)}^2 := \mathbb{E}^1[\int_t^T \int_{\mathcal{A}} |U_s(\alpha)|^2 \lambda(d\alpha) ds] < \infty$ .
- $\mathcal{K}^2(t, T)$ , the set of nondecreasing  $\mathbb{F}^\mu$ -predictable processes  $K = (K_s)_{s \in [t, T]}$ , with  $K: \Omega^1 \times [t, T] \rightarrow \mathbb{R}_+$ , satisfying  $K \in \mathcal{S}^2(t, T)$  and  $K_t = 0$ .

Given  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\bar{\mathbb{P}}$ , consider on  $(\Omega^1, \mathcal{F}^1, \mathbb{F}^\mu, \mathbb{P}^1)$  the following backward stochastic differential equation with constrained jumps over  $[t, T]$ :

$$\begin{cases} Y_s = \mathbb{E}[g(\bar{X}_T^{t, x, \pi}, \mathbb{P}_{\bar{X}_T^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu})] + \int_s^T \mathbb{E}[f(r, \bar{X}_r^{t, x, \pi}, \mathbb{P}_{\bar{X}_r^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_r)] dr + K_T - K_s \\ \quad - \int_s^T \int_{\mathcal{A}} U_r(\alpha) \mu(dr d\alpha), & s \in [t, T], \\ U_s(\alpha) \leq 0, & d\mathbb{P}^1 ds \lambda(d\alpha)\text{-a.e. on } \Omega^1 \times [t, T] \times \mathcal{A}. \end{cases} \quad (4.1)$$

Notice that  $\mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})]$ , as well as  $\mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)]$ , is a random variable on  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ .

Equations (3.3)-(3.4)-(3.5)-(4.1) constitute a forward-backward stochastic differential system of equations. We also observe that equation (4.1) depends on  $\xi$  only through its law  $\pi = \mathbb{P}_\xi$ . We now prove that there exists a unique solution  $(Y^{t,x,\pi}, U^{t,x,\pi}, K^{t,x,\pi}) \in \mathcal{S}^2(t, T) \times L_\mu^2(t, T) \times \mathcal{K}^2(t, T)$  to (4.1), which is minimal in the following sense: if  $(\bar{Y}, \bar{U}, \bar{K}) \in \mathcal{S}^2(t, T) \times L_\mu^2(t, T) \times \mathcal{K}^2(t, T)$  is another solution to (4.1), then the inequality  $Y^{t,x,\pi} \leq \bar{Y}$  holds on  $\Omega^1 \times [t, T]$ , up to a  $\mathbb{P}^1$ -evanescent set.

**Theorem 4.1** *Under Assumption (A1), for every  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\bar{\mathbb{P}}$ , there exists a unique minimal solution  $(Y^{t,x,\pi}, U^{t,x,\pi}, K^{t,x,\pi}) \in \mathcal{S}^2(t, T) \times L_\mu^2(t, T) \times \mathcal{K}^2(t, T)$  to (4.1), with  $Y_t^{t,x,\pi}$  equal  $\mathbb{P}^1$ -a.s. to a constant. In addition,  $V$  admits the Feynman-Kac representation*

$$V(t, x, \pi) = Y_t^{t,x,\pi} \quad (4.2)$$

$\mathbb{P}^1$ -a.s., for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ . Moreover, we have

$$\begin{aligned} Y_t^{t,x,\pi} &= \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right] \\ &= \sup_{\nu \in \mathcal{V}} \bar{\mathbb{E}}^\nu \left[ \int_t^s f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r) dr + Y_s^{t,x,\pi} \right], \end{aligned} \quad (4.3)$$

$\mathbb{P}^1$ -a.s., for all  $s \in [t, T]$ .

**Proof.** *Existence and uniqueness of the minimal solution to (4.1).* Fix  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\bar{\mathbb{P}}$ . Consider, for every  $n \in \mathbb{N}$ , the following unconstrained backward stochastic differential equation on  $[t, T]$ :

$$\begin{aligned} Y_s &= \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] + \int_s^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + n \int_s^T \int_{\mathcal{A}} (U_r(\alpha))_+ \lambda(d\alpha) dr \\ &\quad - \int_s^T \int_{\mathcal{A}} U_r(\alpha) \mu(dr d\alpha). \end{aligned} \quad (4.4)$$

By Lemma 2.4 in [30], there exists a unique solution  $(Y^{n,t,x,\pi}, U^{n,t,x,\pi}) \in \mathcal{S}^2(t, T) \times L_\mu^2(t, T)$  to the above equation.

For every  $n \in \mathbb{N}$ , let  $\hat{\mathcal{V}}^n$  denote the set of  $\mathcal{P}(\mathbb{F}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable maps  $\hat{\nu}: \Omega^1 \times \mathbb{R}_+ \times \mathcal{A} \rightarrow (0, n]$ , which are not necessarily bounded away from zero. Then, let us prove the following formula:

$$Y_{\bar{t}}^{n,t,x,\pi} = \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}^n} \mathbb{E}^{\hat{\nu}} \left[ \int_{\bar{t}}^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{n,t,x,\pi} \middle| \mathcal{F}_{\bar{t}}^\mu \right], \quad (4.5)$$

for all  $\bar{t}, s \in [t, T]$ , with  $\bar{t} \leq s$ . Let  $\hat{\nu} \in \hat{\mathcal{V}}$  (see Remark 3.2 for the definition of  $\hat{\mathcal{V}}$ ). Then, considering (4.4) between  $\bar{t}$  and  $s$ , and taking the  $\mathbb{P}^{\hat{\nu}}$ -conditional expectation with respect to  $\mathcal{F}_{\bar{t}}^\mu$ , we obtain

$$\begin{aligned} Y_{\bar{t}}^{n,t,x,\pi} &= \mathbb{E}^{\hat{\nu}} \left[ \int_{\bar{t}}^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{n,t,x,\pi} \right. \\ &\quad \left. + \int_{\bar{t}}^s \int_{\mathcal{A}} [n(U_r(\alpha))^{n,t,x,\pi}_+ - U_r^{n,t,x,\pi}(\alpha) \hat{\nu}_r(\alpha)] \lambda(d\alpha) dr \middle| \mathcal{F}_{\bar{t}}^\mu \right]. \end{aligned} \quad (4.6)$$



Since  $\hat{\nu}_r(\alpha) \in (0, n]$ , the last term inside the expectation is nonnegative. Therefore

$$Y_{\bar{t}}^{n,t,x,\pi} \geq \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}^n} \mathbb{E}^{\hat{\nu}} \left[ \int_{\bar{t}}^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{n,t,x,\pi} \middle| \mathcal{F}_{\bar{t}}^\mu \right]. \quad (4.7)$$

To prove the other inequality, define, for every  $\varepsilon \in (0, n]$ , the map  $\hat{\nu}^{n,\varepsilon}$  as

$$\hat{\nu}^{n,\varepsilon}(\alpha) = n \mathbf{1}_{\{U_r^{n,t,x,\pi}(\alpha) \geq 0\}} + \varepsilon \mathbf{1}_{\{-1 \leq U_r^{n,t,x,\pi}(\alpha) < 0\}} + \frac{\varepsilon}{|U_r^{n,t,x,\pi}(\alpha)|} \mathbf{1}_{\{U_r^{n,t,x,\pi}(\alpha) < -1\}},$$

on  $\Omega^1 \times [t, T] \times \mathcal{A}$ , and  $\hat{\nu}^{n,\varepsilon} \equiv 1$  on  $\Omega^1 \times ([0, t) \cup (T, \infty)) \times \mathcal{A}$ . Notice that  $\hat{\nu}^{n,\varepsilon}$  belongs to  $\hat{\mathcal{V}}^n$ , and it is not necessarily bounded away from zero. Taking  $\hat{\nu}$  equal to  $\hat{\nu}^{n,\varepsilon}$  in (4.6), we obtain

$$\begin{aligned} Y_{\bar{t}}^{n,t,x,\pi} &\leq \mathbb{E}^{\hat{\nu}^{n,\varepsilon}} \left[ \int_{\bar{t}}^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{n,t,x,\pi} \middle| \mathcal{F}_{\bar{t}}^\mu \right] + \varepsilon(T-t)\lambda(\mathcal{A}) \\ &\leq \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}^n} \left[ \int_{\bar{t}}^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{n,t,x,\pi} \middle| \mathcal{F}_{\bar{t}}^\mu \right] + \varepsilon(T-t)\lambda(\mathcal{A}). \end{aligned} \quad (4.8)$$

From the arbitrariness of  $\varepsilon$  we get the reverse inequality of (4.7), from which we deduce the validity of (4.5). In particular, when  $s = T$  in (4.5), we obtain

$$Y_{\bar{t}}^{n,t,x,\pi} = \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}^n} \mathbb{E}^{\hat{\nu}} \left[ \int_{\bar{t}}^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \middle| \mathcal{F}_{\bar{t}}^\mu] \right], \quad (4.9)$$

for all  $\bar{t} \in [t, T]$ . Then, it is easy to see that the following estimate holds:

$$\sup_n Y_{\bar{t}}^{n,t,x,\pi} < \infty, \quad \text{for all } \bar{t} \in [t, T]. \quad (4.10)$$

Hence, the existence and uniqueness of the minimal solution to equation (4.1) follows from Theorem 2.1 in [21] (apart from the fact that  $K_t^{t,x,\pi} = 0$ , as required in the definition of  $\mathcal{K}^2(t, T)$ , which will be proved later). Indeed, (4.1) can be seen as an equation on the entire interval  $[0, T]$ , with terminal condition  $\mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})]$  and generator  $\mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] \mathbf{1}_{[t,T]}(r)$ . Assumption (H0) in [21] holds under Assumption **(A1)**. Moreover, Assumption (H1) in [21] is imposed only to guarantee the validity of (4.10), which in our case follows directly from formula (4.9), since  $f$  does not depend on  $Y^{n,t,x,\pi}$ ,  $U^{n,t,x,\pi}$ . It only remains to prove that  $K_t^{t,x,\pi} = 0$ . This is clearly true if we show that  $Y_t^{t,x,\pi}$  is equal  $\mathbb{P}^1$ -a.s. to a constant (as a matter of fact, if  $Y_t^{t,x,\pi}$  is equal  $\mathbb{P}^1$ -a.s. to a constant, then, by uniqueness,  $Y_s^{t,x,\pi} = Y_t^{t,x,\pi}$  on  $[0, t]$ , so that  $K_s^{t,x,\pi}$  is also constant on  $[0, t]$ , and, in particular, equal to  $K_0^{t,x,\pi} = 0$ ). This latter property is proved below. Finally, for later use, we notice that, according to Theorem 2.1 in [21], the sequence  $(Y_{\bar{t}}^{n,t,x,\pi})_{n \geq 0}$  is nondecreasing (this is a direct consequence of formula (4.9), since  $\hat{\mathcal{V}}^n \subset \hat{\mathcal{V}}^{n+1}$ ) and converges pointwise  $\mathbb{P}^1$ -a.s. to  $Y_{\bar{t}}^{t,x,\pi}$ , for all  $\bar{t} \in [t, T]$ .

*Proof of (4.2), in particular  $Y_t^{t,x,\pi}$  is equal  $\mathbb{P}^1$ -a.s. to a constant.* Notice that  $Y_t^{t,x,\pi}$  is  $\mathcal{F}_t^\mu$ -measurable, therefore it is not a priori clear that it is  $\mathbb{P}^1$ -a.s. a constant. For every  $n \in \mathbb{N}$ , consider (4.5) with  $\bar{t} = t$  and  $s = T$ :

$$Y_t^{n,t,x,\pi} = \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}^n} \mathbb{E}^{\hat{\nu}} \left[ \int_t^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\xi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \middle| \mathcal{F}_t^\mu] \right].$$

Letting  $n \rightarrow \infty$ , recalling that  $Y_t^{n,t,x,\pi} \nearrow Y_t^{t,x,\pi}$ ,  $\mathbb{P}^1$ -a.s., and noting that  $\hat{\mathcal{V}}_n \subset \hat{\mathcal{V}}_{n+1} \subset \cup_n \hat{\mathcal{V}}_n = \hat{\mathcal{V}}$ , we obtain

$$Y_t^{t,x,\pi} = \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}} \mathbb{E}^{\hat{\nu}} \left[ \int_t^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,x,\pi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right]. \quad (4.11)$$

Reasoning as in Remark 3.2, we can show that the right-hand side of (4.11) does not change if we take the supremum over  $\mathcal{V}$ . In other words, (4.11) can be equivalently written as follows:

$$Y_t^{t,x,\pi} = \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,x,\pi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right]. \quad (4.12)$$

From Corollary D.1 it follows that the right-hand side of (4.12) is equal  $\mathbb{P}^1$ -a.s. to  $V(t, x, \pi)$ , which yields  $Y_t^{t,x,\pi} = V(t, x, \pi)$ ,  $\mathbb{P}^1$ -a.s..

*Proof of formula (4.3).* Let  $\nu \in \mathcal{V}$ . Consider (4.1) between  $t$  and  $s$ , and take the expectation with respect to  $\mathbb{E}^\nu$ , then (recalling that  $K^{t,x,\pi}$  is nondecreasing and  $U^{t,x,\pi}$  is nonpositive)

$$Y_t^{t,x,\pi} \geq \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right]. \quad (4.13)$$

From the arbitrariness of  $\nu \in \mathcal{V}$ , we get the first inequality. To prove the reverse inequality, considering (4.8) with  $\bar{t} = t$ , and taking the expectation  $\mathbb{E}^{\hat{\nu}^{n,\varepsilon}}$ , we obtain

$$\begin{aligned} \mathbb{E}^{\hat{\nu}^{n,\varepsilon}} [Y_t^{n,t,x,\pi}] &\leq \mathbb{E}^{\hat{\nu}^{n,\varepsilon}} \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{n,t,x,\pi} \right] + \varepsilon(T-t)\lambda(\mathcal{A}) \\ &\leq \sup_{\hat{\nu} \in \hat{\mathcal{V}}} \mathbb{E}^{\hat{\nu}} \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right] + \varepsilon(T-t)\lambda(\mathcal{A}) \\ &= \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right] + \varepsilon(T-t)\lambda(\mathcal{A}), \end{aligned}$$

where the last equality can be proved arguing as in Remark 3.2. From the definition of  $\hat{\nu}^{n,\varepsilon}$ , we see that  $\kappa_t^{\hat{\nu}^{n,\varepsilon}} = 1$ , therefore  $\mathbb{E}^{\hat{\nu}^{n,\varepsilon}} [Y_t^{n,t,x,\pi}] = \mathbb{E}^1 [Y_t^{n,t,x,\pi}]$ . Hence

$$\mathbb{E}^1 [Y_t^{n,t,x,\pi}] \leq \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right] + \varepsilon(T-t)\lambda(\mathcal{A}).$$

Recall that the sequence  $(Y_t^{n,t,x,\pi})_{n \geq 0}$  is nondecreasing and converges pointwise  $\mathbb{P}^1$ -a.s. to  $Y_t^{t,x,\pi}$ . In particular,  $Y_t^{0,t,x,\pi} \leq Y_t^{n,t,x,\pi} \leq Y_t^{t,x,\pi}$ , for every  $n \in \mathbb{N}$ . Therefore, letting  $n \rightarrow \infty$  and using Lebesgue's dominated convergence theorem, we obtain

$$Y_t^{t,x,\pi} = \mathbb{E}^1 [Y_t^{t,x,\pi}] \leq \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right] + \varepsilon(T-t)\lambda(\mathcal{A}).$$

Sending  $\varepsilon \rightarrow 0$ , we get

$$Y_t^{t,x,\pi} \leq \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,x,\pi}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + Y_s^{t,x,\pi} \right],$$

which, together with (4.13), gives formula (4.3) and concludes the proof.  $\square$

## 5 Randomized dynamic programming principle

The present section is devoted to the proof of the dynamic programming principle for  $V$  in the randomized framework. Firstly, we prove the flow properties of  $\bar{X}^{t,\bar{\xi}}$  and  $\bar{X}^{t,x,\pi}$ . These in turn imply the identification  $\mathbb{E}[V(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^\mu)] = Y_s^{t,x,\pi}$ ,  $\mathbb{P}^1$ -a.s., for all  $s \in [t, T]$ . Then, (4.3) allows to derive the randomized dynamic programming principle for  $V$ .

### 5.1 Flow properties

We begin considering the solution to system (3.4)-(3.5) with more general initial conditions. More precisely, concerning equation (3.4), for every  $(t, \bar{\eta}) \in [0, T] \times L^2(\bar{\Omega}, \bar{\mathcal{F}}_t^{B,\mu} \vee \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , consider the following equation:

$$d\bar{X}_s^{t,\bar{\eta}} = b(s, \bar{X}_s^{t,\bar{\eta}}, \mathbb{P}_{\bar{X}_s^{t,\bar{\eta}}}^\mu, \bar{I}_s) ds + \sigma(s, \bar{X}_s^{t,\bar{\eta}}, \mathbb{P}_{\bar{X}_s^{t,\bar{\eta}}}^\mu, \bar{I}_s) d\bar{B}_s, \quad \bar{X}_t^{t,\bar{\eta}} = \bar{\eta}, \quad (5.1)$$

for all  $s \in [t, T]$ . Concerning equation (3.5), we begin recalling that  $(\mathbb{P}_{\bar{X}_s^{t,\bar{\eta}}}^\mu)_{s \in [t, T]}$  stands for the stochastic process  $(\mathbb{P}_s^{t,\pi})_{s \in [t, T]}$  introduced in Lemma 3.2, with  $\pi = \mathbb{P}_{\bar{\eta}}$  under  $\bar{\mathbb{P}}$ . In the sequel, when considering equation (3.5), it is more convenient to adopt the notation  $\mathbb{P}_s^{t,\pi}$  instead of  $\mathbb{P}_{\bar{X}_s^{t,\bar{\eta}}}^\mu$ . For every  $(t, \bar{\eta}) \in [0, T] \times L^2(\bar{\Omega}, \bar{\mathcal{F}}_t^{B,\mu} \vee \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  and  $\bar{\Pi}: \bar{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$ , with  $\bar{\Pi}$  measurable with respect to  $\bar{\mathcal{F}}_t^\mu$  and such that  $\bar{\mathbb{E}}[\|\bar{\Pi}\|_2^2] < \infty$ , consider the following equation:

$$d\bar{X}_s^{t,\bar{\eta},\bar{\Pi}} = b(s, \bar{X}_s^{t,\bar{\eta},\bar{\Pi}}, \mathbb{P}_s^{t,\bar{\Pi}}, \bar{I}_s) ds + \sigma(s, \bar{X}_s^{t,\bar{\eta},\bar{\Pi}}, \mathbb{P}_s^{t,\bar{\Pi}}, \bar{I}_s) d\bar{B}_s, \quad \bar{X}_t^{t,\bar{\eta},\bar{\Pi}} = \bar{\eta}, \quad (5.2)$$

for all  $s \in [t, T]$ , where

$$\mathbb{P}_s^{t,\bar{\Pi}}(\bar{\omega}) := \mathbb{P}_s^{t,\bar{\Pi}(\bar{\omega})}(\omega^1), \quad \text{for all } (\bar{\omega}, s) = (\omega, \omega^1, s) \in \bar{\Omega} \times [t, T]. \quad (5.3)$$

Notice that, thanks to Lemma 3.2, the stochastic process  $(\mathbb{P}_s^{t,\bar{\Pi}})_{s \in [t, T]}$  is well-defined. In particular, for every  $s \in [t, T]$ ,  $\mathbb{P}_s^{t,\bar{\Pi}}$  is  $\bar{\mathcal{F}}_s^\mu$ -measurable. Under Assumption **(A1)**, we have the following result, whose standard proof is not reported.

**Lemma 5.1** *Under Assumption **(A1)**, for every  $(t, \bar{\eta}) \in [0, T] \times L^2(\bar{\Omega}, \bar{\mathcal{F}}_t^{B,\mu} \vee \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  and  $\bar{\Pi}: \bar{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$ , with  $\bar{\Pi}$  measurable with respect to  $\bar{\mathcal{F}}_t^\mu$  and such that  $\bar{\mathbb{E}}[\|\bar{\Pi}\|_2^2] < \infty$ , there exists a unique (up to indistinguishability) pair  $(\bar{X}_s^{t,\bar{\eta}}, \bar{X}_s^{t,\bar{\eta},\bar{\Pi}})_{s \in [t, T]}$  of continuous  $(\bar{\mathcal{F}}_s^{B,\mu,t} \vee \bar{\mathcal{G}} \vee \sigma(\bar{\eta}, \bar{\Pi}))_s$ -adapted processes solution to equations (5.1)-(5.2), satisfying*

$$\bar{\mathbb{E}} \left[ \sup_{s \in [t, T]} (|\bar{X}_s^{t,\bar{\eta}}|^2 + |\bar{X}_s^{t,\bar{\eta},\bar{\Pi}}|^2) \right] < \infty.$$

Moreover, there exists a positive constant  $C$  such that

$$\bar{\mathbb{E}} \left[ \sup_{s \in [t, T]} |\bar{X}_s^{t,\bar{\eta},\bar{\Pi}} - \bar{X}_s^{t,\bar{\eta}',\bar{\Pi}'}|^2 \right] \leq C(\bar{\mathbb{E}}[|\bar{\eta} - \bar{\eta}'|^2] + \bar{\mathbb{E}}[\mathcal{W}_2(\bar{\Pi}, \bar{\Pi}')^2]), \quad (5.4)$$

for every  $t \in [0, T]$ ,  $\bar{\eta}, \bar{\eta}' \in L^2(\bar{\Omega}, \bar{\mathcal{F}}_t^{B,\mu} \vee \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , and any  $\bar{\Pi}, \bar{\Pi}': \bar{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$ , with  $\bar{\Pi}, \bar{\Pi}'$  measurable with respect to  $\bar{\mathcal{F}}_t^\mu$  and such that  $\bar{\mathbb{E}}[\|\bar{\Pi}\|_2^2], \bar{\mathbb{E}}[\|\bar{\Pi}'\|_2^2] < \infty$ .

**Proof.** The proof of the existence and uniqueness of  $(\bar{X}_s^{t,\bar{\eta}}, \bar{X}_s^{t,\bar{\eta},\bar{\Pi}})_{s \in [t, T]}$  is standard under Assumption **(A1)**, and can be done as usual by a fixed point argument. Concerning estimate (5.4), the proof can be done proceeding as in Lemma 3.1 in [9].  $\square$

**Remark 5.1** When in equation (5.2) the random variables  $\bar{\eta}$  and  $\bar{\Pi}$  are equal  $\bar{\mathbb{P}}$ -a.s. to some  $x \in \mathbb{R}^n$  and  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ , respectively, then  $(\bar{X}_s^{t,\bar{\eta},\bar{\Pi}})_{s \in [t,T]}$  coincides (up to indistinguishability) with the stochastic process  $(\bar{X}_s^{t,x,\pi})_{s \in [t,T]}$  defined in Section 3. Indeed,  $(\bar{X}_s^{t,\bar{\eta},\bar{\Pi}})_{s \in [t,T]}$  and  $(\bar{X}_s^{t,x,\pi})_{s \in [t,T]}$  solve the same equation, therefore the claim follows from the uniqueness of the solution.  $\square$

**Remark 5.2** Suppose that  $\bar{\eta}$  and  $\bar{\Pi}$  in Lemma 5.1 takes only a finite number of values, namely

$$\bar{\eta} = \sum_{k=0}^K x_k 1_{E_k}, \quad \bar{\Pi} = \sum_{k=0}^K \pi_k 1_{E_k},$$

for some  $K \in \mathbb{N}$ ,  $x_k \in \mathbb{R}^n$ ,  $\pi_k \in \mathcal{P}_2(\mathbb{R}^n)$ ,  $E_k \in \bar{\mathcal{F}}_t^{B,\mu} \vee \bar{\mathcal{G}}$ , with  $(E_k)_{k=1,\dots,K}$  being a partition of  $\bar{\Omega}$ . Then, by definition of  $\mathbb{P}_s^{t,\bar{\Pi}}$  (formula (5.3)), we have  $\mathbb{P}_s^{t,\bar{\Pi}} = \mathbb{P}_s^{t,\pi_0} 1_{E_0} + \dots + \mathbb{P}_s^{t,\pi_K} 1_{E_K}$ . Therefore, the stochastic processes  $(\bar{X}_s^{t,x_0,\pi_0} 1_{E_0} + \dots + \bar{X}_s^{t,x_K,\pi_K} 1_{E_K})_{s \in [t,T]}$  and  $(\bar{X}_s^{t,\bar{\eta},\bar{\Pi}})_{s \in [t,T]}$  are indistinguishable, since they solve the same stochastic differential equation.  $\square$

**Lemma 5.2** Under Assumption (A1), for every  $(t, s, x, \bar{\xi}) \in [0, T] \times [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $t \leq s$  and  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ , we have the flow properties:

$$\bar{X}_r^{s,\bar{X}_s^{t,\bar{\xi}}} = \bar{X}_r^{t,\bar{\xi}}, \quad (5.5)$$

$$\bar{X}_r^{s,\bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu}} = \bar{X}_r^{t,x,\pi}, \quad (5.6)$$

for all  $r \in [s, T]$ ,  $\bar{\mathbb{P}}$ -almost surely.

**Proof.** Flow property (5.5). Consider the process  $(\bar{X}_r^{s,\bar{X}_s^{t,\bar{\xi}}})_{r \in [s,T]}$  solution to equation (5.1) with initial conditions  $t = s$  and  $\bar{\eta} = \bar{X}_s^{t,\bar{\xi}}$ . Since  $(\bar{X}_r^{t,\bar{\xi}})_{r \in [s,T]}$  solves the same equation, by pathwise uniqueness we deduce that  $(\bar{X}_r^{s,\bar{X}_s^{t,\bar{\xi}}})_{r \in [s,T]}$  and  $(\bar{X}_r^{t,\bar{\xi}})_{r \in [s,T]}$  are indistinguishable, namely (5.5) holds.

Flow property (5.6). Recall that  $(\mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t,T]}$  stands for the stochastic process  $(\mathbb{P}_s^{t,\pi})_{s \in [t,T]}$  introduced in Lemma 3.2. In the present proof it is more convenient to adopt the notation  $\mathbb{P}_s^{t,\pi}$  instead of  $\mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu}$ . Notice that, by (5.5), we have  $\mathbb{P}_r^{t,\pi} = \mathbb{P}_r^{s,\mathbb{P}_s^{t,\pi}}$ , for all  $r \in [s, T]$ ,  $\bar{\mathbb{P}}$ -almost surely. Therefore

$$\begin{aligned} \bar{X}_r^{t,x,\pi} &= \bar{X}_s^{t,x,\pi} + \int_s^r b(u, \bar{X}_u^{t,x,\pi}, \mathbb{P}_u^{t,\pi}, \bar{I}_u) du + \int_s^r \sigma(u, \bar{X}_u^{t,x,\pi}, \mathbb{P}_u^{t,\pi}, \bar{I}_u) d\bar{B}_u \\ &= \bar{X}_s^{t,x,\pi} + \int_s^r b(u, \bar{X}_u^{t,x,\pi}, \mathbb{P}_u^{s,\mathbb{P}_s^{t,\pi}}, \bar{I}_u) du + \int_s^r \sigma(u, \bar{X}_u^{t,x,\pi}, \mathbb{P}_u^{s,\mathbb{P}_s^{t,\pi}}, \bar{I}_u) d\bar{B}_u, \end{aligned}$$

for all  $r \in [s, T]$ ,  $\bar{\mathbb{P}}$ -a.s.. On the other hand, consider the process  $(\bar{X}_r^{s,\bar{X}_s^{t,x,\pi}, \mathbb{P}_s^{t,\pi}})_{r \in [s,T]}$  solution to equation (5.2) with initial conditions  $t = s$ ,  $\bar{\eta} = \bar{X}_s^{t,x,\pi}$ ,  $\bar{\Pi} = \mathbb{P}_s^{t,\pi}$ . Then, we see that  $(\bar{X}_r^{s,\bar{X}_s^{t,x,\pi}, \mathbb{P}_s^{t,\pi}})_{r \in [s,T]}$  and  $(\bar{X}_r^{t,x,\pi})_{r \in [s,T]}$  solve the same equation. It follows that they are indistinguishable, namely (5.6) holds.  $\square$

## 5.2 Randomized dynamic programming principle

We begin proving the following identification result between  $V$  and  $Y^{t,x,\pi}$ .

**Lemma 5.3** *Under Assumptions (A1) and (A2), for every  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ , we have*

$$\mathbb{E}[V(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})] = Y_s^{t,x,\pi},$$

$\mathbb{P}^1$ -a.s., for all  $s \in [t, T]$ .

**Proof.** Fix  $(t, s, x, \bar{\xi}) \in [0, T] \times [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $t \leq s$  and  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ . Using the same notations as in the proof of Theorem 4.1, let us consider, for every  $n \in \mathbb{N}$ , formula (4.5) with  $\bar{t}$  and  $s$  replaced respectively by  $s$  and  $T$ :

$$Y_s^{n,t,x,\pi} = \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}^n} \mathbb{E}^{\hat{\nu}} \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_s^\mu \right].$$

Letting  $n \rightarrow \infty$ , we obtain

$$Y_s^{t,x,\pi} = \operatorname{ess\,sup}_{\hat{\nu} \in \hat{\mathcal{V}}} \mathbb{E}^{\hat{\nu}} \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_s^\mu \right].$$

Reasoning as in Remark 3.2, we can show that the right-hand side of (4.11) does not change if we take the supremum over  $\mathcal{V}$ . In other words, (4.11) can be equivalently written as follows:

$$Y_s^{t,x,\pi} = \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_s^\mu \right].$$

Then, we see that the claim follows if we prove the following equality:  $\mathbb{P}^1$ -a.s.

$$\mathbb{E}[V(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})] = \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{t,x,\pi}, \mathbb{P}_{\bar{X}_r^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_r^\mu}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_s^\mu \right]. \quad (5.7)$$

As in the proof of Lemma 5.2, it is more convenient to adopt the notation  $\mathbb{P}_s^{t,\pi}$  instead of  $\mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu}$  (recall that  $(\mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$  stands for the stochastic process  $(\mathbb{P}_s^{t,\pi})_{s \in [t, T]}$  introduced in Lemma 3.2). Then, from the flow properties (5.5) and (5.6), we have

$$Y_s^{t,x,\pi} = \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_s^{t,\pi}}, \mathbb{P}_r^{s, \mathbb{P}_s^{t,\pi}}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_s^{t,\pi}}, \mathbb{P}_T^{s, \mathbb{P}_s^{t,\pi}})] \middle| \mathcal{F}_s^\mu \right]. \quad (5.8)$$

Now, notice that  $\bar{X}_s^{t,x,\pi} \in L^2(\bar{\Omega}, \bar{\mathcal{F}}_s^{B,\mu}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , so that it is the  $L^2$ -limit (and also pointwise  $\bar{\mathbb{P}}$ -a.s.) of a sequence  $(\bar{X}_m)_{m \geq 0} \subset L^2(\bar{\Omega}, \bar{\mathcal{F}}_s^{B,\mu}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , where each  $\bar{X}_m$  takes only a finite number of values. Similarly,  $\mathbb{P}_s^{t,\pi}$  is a random variable  $\mathbb{P}_s^{t,\pi}: \bar{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  such that  $\bar{\mathbb{E}}[\|\mathbb{P}_s^{t,\pi}\|_2^2] < \infty$ . Therefore, by Lemma A.3 there exists a sequence  $(\mathbb{P}_m)_{m \geq 0}$  of  $\bar{\mathcal{F}}_s^{B,\mu}$ -measurable maps  $\mathbb{P}_m: \bar{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$ , with  $\bar{\mathbb{E}}[\|\mathbb{P}_m\|_2^2] < \infty$  and each  $\mathbb{P}_m$  takes only a finite number values, such that  $\bar{\mathbb{E}}[\mathcal{W}_2(\mathbb{P}_m, \mathbb{P}_s^{t,\pi})^2] \rightarrow 0$  as  $m$  goes to infinity (and also  $\mathcal{W}_2(\mathbb{P}_m, \mathbb{P}_s^{t,\pi}) \rightarrow 0$  pointwise  $\bar{\mathbb{P}}$ -a.s.). In particular, for every  $m \geq 0$ , we have

$$\bar{X}_m = \sum_{k=0}^{K_m} x_{m,k} 1_{E_{m,k}}, \quad \mathbb{P}_m = \sum_{k=0}^{K_m} \pi_{m,k} 1_{E_{m,k}},$$

for some  $K_m \in \mathbb{N}$ ,  $x_{m,k} \in \mathbb{R}^n$ ,  $\pi_{m,k} \in \mathcal{P}_2(\mathbb{R}^n)$ ,  $E_{m,k} \in \bar{\mathcal{F}}_s^{B,\mu}$ , with  $(E_{m,k})_k$  being a partition of  $\bar{\Omega}$ . For every  $m \geq 0$ , consider the process  $(\bar{X}_r^{s, \bar{X}_m, \mathbb{P}_m})_{r \in [s, T]}$ , solution to equation (5.2) with

initial conditions  $t = s$ ,  $\bar{\eta} = \bar{X}_m$ ,  $\bar{\Pi} = \mathbb{P}_m$ . Recall from Remark 5.2, we have that the stochastic processes  $(\bar{X}_r^{s, \bar{X}_m, \mathbb{P}_m})_{r \in [s, T]}$  and  $(\sum_{k=0}^{K_m} \bar{X}_r^{s, x_{m,k}, \pi_{m,k}} 1_{E_{m,k}})_{r \in [s, T]}$  are indistinguishable.

Notice that, for every  $\nu \in \mathcal{V}$ , we have, from Corollary D.1,  $\mathbb{P}^1$ -a.s.,

$$\begin{aligned} \mathbb{E}[V(s, \bar{X}_m, \mathbb{P}_m)] &= \sum_{k=0}^{K_m} \mathbb{E}[V(s, x_{m,k}, \pi_{m,k}) 1_{E_{m,k}}] \\ &= \sum_{k=0}^{K_m} \mathbb{E} \left[ 1_{E_{m,k}} \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{s, x_{m,k}, \pi_{m,k}}, \mathbb{P}_r^{s, \pi_{m,k}}, \bar{I}_r)] dr \right. \right. \\ &\quad \left. \left. + \mathbb{E}[g(\bar{X}_T^{s, x_{m,k}, \pi_{m,k}}, \mathbb{P}_T^{s, \pi_{m,k}})] \middle| \mathcal{F}_s^\mu \right] \right] \\ &= \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{s, \bar{X}_m, \mathbb{P}_m}, \mathbb{P}_r^{s, \mathbb{P}_m}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{s, \bar{X}_m, \mathbb{P}_m}, \mathbb{P}_T^{s, \mathbb{P}_m})] \middle| \mathcal{F}_s^\mu \right]. \end{aligned} \quad (5.9)$$

From the continuity of the map  $(y, \gamma) \mapsto V(s, y, \gamma)$  stated in Proposition 2.1, and the growth condition (2.8), we see that

$$\mathbb{E}[V(s, \bar{X}_m, \mathbb{P}_m)] \xrightarrow[\mathbb{P}^1\text{-a.s.}]{m \rightarrow \infty} \mathbb{E}[V(s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_s^{t, \pi})]. \quad (5.10)$$

On the other hand, using estimate (5.4) and proceeding as in the proof of inequality (2.18) in Proposition 2.1, we can prove the following convergence:

$$\begin{aligned} &\operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{s, \bar{X}_m, \mathbb{P}_m}, \mathbb{P}_r^{s, \mathbb{P}_m}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{s, \bar{X}_m, \mathbb{P}_m}, \mathbb{P}_T^{s, \mathbb{P}_m})] \middle| \mathcal{F}_s^\mu \right] \\ &\xrightarrow[\mathbb{P}^1\text{-a.s.}]{m \rightarrow \infty} \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_s^T \mathbb{E}[f(r, \bar{X}_r^{s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_s^{t, \pi}}, \mathbb{P}_r^{s, \mathbb{P}_s^{t, \pi}}, \bar{I}_r)] dr + \mathbb{E}[g(\bar{X}_T^{s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_s^{t, \pi}}, \mathbb{P}_T^{s, \mathbb{P}_s^{t, \pi}})] \middle| \mathcal{F}_s^\mu \right]. \end{aligned} \quad (5.11)$$

Hence, by (5.10) and (5.11), together with equalities (5.8) and (5.9), we see that (5.7) holds, therefore the claim follows.  $\square$

We can now state the main result of this section.

**Theorem 5.1** *Suppose that Assumptions (A1) and (A2) hold. Then, for every  $(t, s, x, \bar{\xi}) \in [0, T] \times [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $t \leq s$  and  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ , we have*

$$V(t, x, \pi) = \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t, x, \pi}, \mathbb{P}_{\bar{X}_r^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_r)] dr + \mathbb{E}[V(s, \bar{X}_s^{t, x, \pi}, \mathbb{P}_{\bar{X}_s^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu})] \right].$$

**Proof.** Fix  $(t, s, x, \bar{\xi}) \in [0, T] \times [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $t \leq s$  and  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ . Recall that by (4.3) we have,  $\mathbb{P}^1$ -a.s.,

$$Y_t^{t, x, \pi} = \sup_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^s \mathbb{E}[f(r, \bar{X}_r^{t, x, \pi}, \mathbb{P}_{\bar{X}_r^{t, \bar{\xi}}}^{\bar{\mathcal{F}}^\mu}, \bar{I}_r)] dr + Y_s^{t, x, \pi} \right].$$

Then, the claim follows from Lemma 5.3.  $\square$

**Remark 5.3** *Hamilton-Jacobi-Bellman equation for  $V$  and  $V_{\text{MKV}}$ .* Let us derive, in a formal way, the dynamic programming equation for the value function  $V$ . We proceed as usual, starting from the dynamic programming principle of Theorem 5.1 and applying Itô's formula (see the

Appendix in [12]) to the process  $V(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{F}_s^\mu})$ , supposing that  $V$  is smooth enough. Then, it is easy to see that the Hamilton-Jacobi-Bellman equation for  $V$  takes the following form (see Section 6 of [10] for the definition of  $\partial_\pi$ ):

$$\begin{aligned} \partial_t V(t, x, \pi) + \sup_{a \in A} \left\{ f(t, x, \pi, a) + b(t, x, \pi, a) \cdot \left[ \partial_x V(t, x, \pi) + \int_{\mathbb{R}^n} \partial_\pi V(t, x', \pi)(x) \pi(dx') \right] \right. \\ \left. + \frac{1}{2} \text{tr}(\sigma \sigma^\top(t, x, \pi, a) \left[ \partial_x^2 V(t, x, \pi) + \int_{\mathbb{R}^n} \partial_x \partial_\pi V(t, x', \pi)(x) \pi(dx') \right] \right\} = 0, \end{aligned}$$

for all  $(t, x, \pi) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n)$ , with terminal condition

$$V(T, x, \pi) = g(x, \pi), \quad \text{for all } (x, \pi) \in \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n).$$

We can also derive the Hamilton-Jacobi-Bellman equation for the value function  $V_{\text{MKV}}$  defined by (2.19). From Proposition 2.2, we have

$$V_{\text{MKV}}(t, \xi) = \mathbb{E}[V(t, \xi, \pi)] = \int_{\mathbb{R}^n} V(t, x, \pi) \pi(dx),$$

for all  $(t, \xi) \in [0, T] \times L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\mathbb{P}$ . From the above formula we see that  $V_{\text{MKV}}$  depends on  $\xi$  only through its law  $\pi$ . In other words,  $V_{\text{MKV}}(t, \xi) = V_{\text{MKV}}(t, \xi')$  whenever  $\xi$  and  $\xi'$  have the same law  $\pi$ . Then, by an abuse of notation, we suppose that  $V_{\text{MKV}}$  is defined on  $[0, T] \times \mathcal{P}_2(\mathbb{R}^n)$  with  $V_{\text{MKV}}(t, \pi)$  given by  $V_{\text{MKV}}(t, \xi)$ , for some  $\xi$  such that  $\pi = \mathbb{P}_\xi$ . Now, recalling the definition of the derivative  $\partial_\pi$ , we obtain

$$\begin{aligned} \partial_t V_{\text{MKV}}(t, \pi) &= \mathbb{E}[\partial_t V(t, \xi, \pi)], \\ \partial_\pi V_{\text{MKV}}(t, \pi)(x) &= \partial_x V(t, x, \pi) + \mathbb{E}[\partial_\pi V(t, \xi, \pi)(x)], \\ \partial_x \partial_\pi V_{\text{MKV}}(t, \pi)(x) &= \partial_x^2 V(t, x, \pi) + \mathbb{E}[\partial_x \partial_\pi V(t, \xi, \pi)(x)]. \end{aligned}$$

Integrating with respect to  $\pi$  in the Hamilton-Jacobi-Bellman equation of  $V$ , we obtain the following dynamic programming equation for  $V_{\text{MKV}}$ :

$$\begin{aligned} \partial_t V_{\text{MKV}}(t, \pi) + \int_{\mathbb{R}^n} \sup_{a \in A} [f(t, x, \pi, a) + b(t, x, \pi, a) \cdot \partial_\pi V_{\text{MKV}}(t, \pi)(x) \\ + \frac{1}{2} \text{tr}(\sigma \sigma^\top(t, x, \pi, a) \partial_x \partial_\pi V_{\text{MKV}}(t, \pi)(x))] \pi(dx) = 0, \end{aligned} \quad (5.12)$$

for all  $(t, \pi) \in [0, T] \times \mathcal{P}_2(\mathbb{R}^n)$ , with terminal condition

$$V_{\text{MKV}}(T, \pi) = \int_{\mathbb{R}^n} g(x, \pi) \pi(dx), \quad \text{for all } \pi \in \mathcal{P}_2(\mathbb{R}^n).$$

Notice that if the supremum inside the integral in (5.12) is attained at some  $\hat{a}(x)$ , for some map  $\hat{a}: \mathbb{R}^n \rightarrow A$  Lipschitz continuous in  $x$ , then the above equation can be written as (we denote by  $L(\mathbb{R}^n; A)$  the set of Lipschitz continuous maps from  $\mathbb{R}^n$  into  $A$ )

$$\begin{aligned} \partial_t V_{\text{MKV}}(t, \pi) + \sup_{\tilde{\alpha} \in L(\mathbb{R}^n; A)} \int_{\mathbb{R}^n} [f(t, x, \pi, \tilde{\alpha}(x)) + b(t, x, \pi, \tilde{\alpha}(x)) \cdot \partial_\pi V_{\text{MKV}}(t, \pi)(x) \\ + \frac{1}{2} \text{tr}(\sigma \sigma^\top(t, x, \pi, \tilde{\alpha}(x)) \partial_x \partial_\pi V_{\text{MKV}}(t, \pi)(x))] \pi(dx) = 0. \end{aligned}$$

This latter is the Hamilton-Jacobi-Bellman equation obtained in [26] under the assumption that the optimization in the McKean-Vlasov control problem is performed only over the class of Lipschitz continuous closed-loop controls.  $\square$

## A Some convergence results with respect to the 2-Wasserstein metric $\mathcal{W}_2$

**Lemma A.1 (Skorohod's representation theorem for  $\mathcal{W}_2$ -convergence)** *Let  $(\pi_m)_m$  be a sequence in  $\mathcal{P}_2(\mathbb{R}^n)$  such that  $\mathcal{W}_2(\pi_m, \pi) \rightarrow 0$ , for some  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ . Then, there exists a sequence of random variables  $(\xi_m)_m \subset L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , with  $\mathbb{P}_{\xi_m} = \pi_m$ , converging pointwise  $\mathbb{P}$ -a.s. and in  $L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$  to some  $\xi \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , with  $\mathbb{P}_\xi = \pi$ .*

**Proof.** By Theorem 6.9 and point (i) of Definition 6.8 in [31], we have that  $\mathcal{W}_2(\pi_m, \pi) \rightarrow 0$  is equivalent to:

$$\pi_m \xrightarrow[\text{weakly}]{m \rightarrow \infty} \pi \quad \text{and} \quad \int_{\mathbb{R}^n} |x|^2 \pi_m(dx) \xrightarrow{m \rightarrow \infty} \int_{\mathbb{R}^n} |x|^2 \pi(dx). \quad (\text{A.1})$$

Then, by the classical Skorohod representation theorem for weak convergence, there exist random variables  $\xi_m, \xi \in L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ , with  $\mathbb{P}_{\xi_m} = \pi_m$  and  $\mathbb{P}_\xi = \pi$ , such that  $\xi_m$  converges pointwise  $\mathbb{P}$ -a.s. to  $\xi$ . It remains to prove the convergence in  $L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ . To this end, we notice that (A.1) implies  $\mathbb{E}[|\xi_m|^2] \rightarrow \mathbb{E}[|\xi|^2]$ . Therefore, by Theorem II.6.5 in [28], the sequence  $(|\xi_m|^2)_m$  is uniformly integrable. Then, it follows that  $\xi_m \rightarrow \xi$  in  $L^2(\Omega, \mathcal{G}, \mathbb{P}; \mathbb{R}^n)$ .  $\square$

**Lemma A.2** *There exists a countable convergence determining class  $(\varphi_k)_{k \geq 1} \subset C_2(\mathbb{R}^n)$  for the  $\mathcal{W}_2$ -convergence. In other words, given  $\pi_1, \pi_2, \dots, \pi \in \mathcal{P}_2(\mathbb{R}^n)$ , we have:*

$$\mathcal{W}_2(\pi_m, \pi) \xrightarrow{m \rightarrow \infty} 0 \quad \text{if and only if} \quad \int_{\mathbb{R}^n} \varphi_k(x) \pi_m(dx) \xrightarrow{m \rightarrow \infty} \int_{\mathbb{R}^n} \varphi_k(x) \pi(dx), \quad \text{for all } k.$$

**Proof.** Let  $\pi_1, \pi_2, \dots, \pi \in \mathcal{P}_2(\mathbb{R}^n)$ . We recall from Theorem 6.9 and point (i) of Definition 6.8 in [31] that

$$\mathcal{W}_2(\pi_m, \pi) \xrightarrow{m \rightarrow \infty} 0 \quad \text{if and only if} \quad \pi_m \xrightarrow[\text{weakly}]{m \rightarrow \infty} \pi \quad \text{and} \quad \int_{\mathbb{R}^n} |x|^2 \pi_m(dx) \xrightarrow{m \rightarrow \infty} \int_{\mathbb{R}^n} |x|^2 \pi(dx).$$

Now, it is well-known that there exists a countable convergence determining class  $(\psi_h)_{h \geq 1} \subset C_b(\mathbb{R}^n)$  (the set of real-valued continuous and bounded functions) for the weak convergence (see, for instance, Theorem 2.18 in [3]). In other words, we have

$$\pi_m \xrightarrow[\text{weakly}]{m \rightarrow \infty} \pi \quad \text{if and only if} \quad \int_{\mathbb{R}^n} \psi_h(x) \pi_m(dx) \xrightarrow{m \rightarrow \infty} \int_{\mathbb{R}^n} \psi_h(x) \pi(dx), \quad \text{for all } h.$$

Then, the claim follows taking  $\varphi_1(x) := |x|^2$ , for every  $x \in \mathbb{R}^n$ , and  $\varphi_k := \psi_{k-1}$ , for every  $k \geq 2$ .  $\square$

**Lemma A.3** *Let  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$  be a probability space and let  $\Pi: \tilde{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  be a measurable map. Suppose that  $(\tilde{\mathbb{E}}$  denotes the  $\tilde{\mathbb{P}}$ -expected value)*

$$\tilde{\mathbb{E}}[\|\Pi\|_2^2] < +\infty. \quad (\text{A.2})$$

*Then, there exists a sequence  $(\Pi_m)_{m \geq 1}$  of measurable maps  $\Pi_m: \tilde{\Omega} \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  such that:*

$$\mathcal{W}_2(\Pi_m(\tilde{\omega}), \Pi(\tilde{\omega})) \xrightarrow{m \rightarrow \infty} 0, \quad \tilde{\mathbb{P}}(d\tilde{\omega})\text{-a.s.}, \quad \text{and} \quad \tilde{\mathbb{E}}[\mathcal{W}_2(\Pi_m, \Pi)^2] \xrightarrow{m \rightarrow \infty} 0,$$



where, for every  $m \geq 1$ ,

$$\Pi_m(\tilde{\omega}) = \sum_{k=1}^{K_m} \pi_{m,k} 1_{E_{m,k}}(\tilde{\omega}), \quad \text{for every } \tilde{\omega} \in \tilde{\Omega},$$

for some finite integer  $K_m \geq 1$ ,  $\pi_{m,k} \in \mathcal{P}_2(\mathbb{R}^n)$ ,  $E_{m,k} \in \tilde{\mathcal{F}}$ , with  $(E_{m,k})_{k=1,\dots,K_m}$  being a partition of  $\tilde{\Omega}$ .

**Proof.** Recall from Theorem 6.18 in [31] that  $(\mathcal{P}_2(\mathbb{R}^n), \mathcal{W}_2)$  is a complete separable metric space. Then, there exists a sequence  $(\pi_h)_{h \geq 1}$  dense in  $\mathcal{P}_2(\mathbb{R}^n)$ . Now, for every  $\ell, h \geq 1$ , define the measurable set  $\bar{B}_{\ell,h} \in \tilde{\mathcal{F}}$  by

$$\bar{B}_{\ell,h} := \{\tilde{\omega} \in \tilde{\Omega} : \mathcal{W}_2(\Pi(\tilde{\omega}), \pi_h) \leq 1/\ell\}.$$

We also define the disjoint measurable sets:  $B_{\ell,1} := \bar{B}_{\ell,1}$  and  $B_{\ell,h} := \bar{B}_{\ell,h} \setminus (\bar{B}_{\ell,1} \cup \dots \cup \bar{B}_{\ell,h-1})$ , for any  $h \geq 2$ . Notice that  $\tilde{\Omega} = \cup_{h \geq 1} B_{\ell,h}$ . In particular, for every  $\ell \geq 1$ , there exists  $K_\ell \geq 1$  such that  $\tilde{\mathbb{P}}(\cup_{h \geq K_\ell+1} B_{\ell,h}) \leq 1/\ell^2$ . Finally, we set

$$\bar{\Pi}_\ell(\tilde{\omega}) := \sum_{h=1}^{K_\ell} \pi_h 1_{B_{\ell,h} \cap A_\ell}(\tilde{\omega}) + \delta_0 (1_{(\cup_{h \geq K_\ell+1} B_{\ell,h}) \cap A_\ell}(\tilde{\omega}) + 1_{A_\ell^c}(\tilde{\omega})), \quad \text{for every } \tilde{\omega} \in \tilde{\Omega},$$

where

$$A_\ell := \{\tilde{\omega} \in \tilde{\Omega} : \|\Pi(\tilde{\omega})\|_2^2 \leq \ell\}.$$

Then, we see that (recall from (2.2) that  $\mathcal{W}_2(\delta_0, \Pi(\tilde{\omega})) = \|\Pi(\tilde{\omega})\|_2$ )

$$\mathcal{W}_2(\bar{\Pi}_\ell(\tilde{\omega}), \Pi(\tilde{\omega})) \leq \frac{1}{\ell} 1_{(\cup_{h=1}^{K_\ell} B_{\ell,h}) \cap A_\ell}(\tilde{\omega}) + \|\Pi(\tilde{\omega})\|_2 (1_{(\cup_{h \geq K_\ell+1} B_{\ell,h}) \cap A_\ell}(\tilde{\omega}) + 1_{A_\ell^c}(\tilde{\omega})),$$

for all  $\tilde{\omega} \in \tilde{\Omega}$ . Therefore (recalling that  $\tilde{\mathbb{P}}(\cup_{h \geq K_\ell+1} B_{\ell,h}) \leq 1/\ell^2$ )

$$\begin{aligned} \tilde{\mathbb{E}}[\mathcal{W}_2(\bar{\Pi}_\ell, \Pi)^2] &\leq \frac{1}{\ell^2} + \tilde{\mathbb{E}}[\|\Pi(\tilde{\omega})\|_2^2 1_{(\cup_{h \geq K_\ell+1} B_{\ell,h}) \cap A_\ell}] + \tilde{\mathbb{E}}[\|\Pi(\tilde{\omega})\|_2^2 1_{A_\ell^c}] \\ &\leq \frac{1}{\ell^2} + \ell \tilde{\mathbb{P}}((\cup_{h \geq K_\ell+1} B_{\ell,h}) \cap A_\ell) + \tilde{\mathbb{E}}[\|\Pi(\tilde{\omega})\|_2^2 1_{A_\ell^c}] \\ &\leq \frac{1}{\ell^2} + \ell \frac{1}{\ell^2} + \tilde{\mathbb{E}}[\|\Pi(\tilde{\omega})\|_2^2 1_{A_\ell^c}] \xrightarrow{\ell \rightarrow \infty} 0, \end{aligned}$$

where the convergence  $\tilde{\mathbb{E}}[\|\Pi(\tilde{\omega})\|_2^2 1_{A_\ell^c}] \rightarrow 0$  follows from the Lebesgue dominated convergence theorem, using (A.2) and noting that  $1_{A_\ell^c}$  converges pointwise  $\tilde{\mathbb{P}}$ -a.s. to zero.

Let  $Y_\ell : \tilde{\Omega} \rightarrow [0, \infty)$  be the nonnegative random variable given by  $Y_\ell := \mathcal{W}_2(\bar{\Pi}_\ell, \Pi)$ . We know that  $Y_\ell \rightarrow 0$ , as  $\ell \rightarrow \infty$ , in  $L^2(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$ . Then, it is well-known that this implies the existence of a subsequence  $(Y_{\ell_m})_{m \geq 1}$  such that  $Y_{\ell_m} = \mathcal{W}_2(\bar{\Pi}_{\ell_m}, \Pi) \rightarrow 0$ , as  $m \rightarrow \infty$ , pointwise  $\tilde{\mathbb{P}}$ -a.s. and in  $L^2(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$ . Then,  $(\Pi_m)_{m \geq 1}$ , with  $\Pi_m := \bar{\Pi}_{\ell_m}$ , is the desired sequence.  $\square$

## B Proofs of Lemma 3.1 and Lemma 3.2

**Proof of Lemma 3.1.** Recall that, by construction, the map  $\bar{X}^{t,\tilde{\xi}} : ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \tilde{\mathcal{F}}) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. Therefore, up to indistinguishability, we can suppose that  $\bar{X}^{t,\tilde{\xi}} : ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \mathcal{F} \otimes \mathcal{F}^1) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. Since  $(\bar{X}_s^{t,\tilde{\xi}})_{s \in [t, T]}$  is also

$(\bar{\mathcal{F}}_s^{B,\mu} \vee \bar{\mathcal{G}})_s$ -adapted, we deduce that, for every  $s \in [t, T]$ , the map  $\bar{X}_s^{t,\bar{\xi}}: (\Omega \times \Omega^1, (\mathcal{G} \vee \mathcal{F}_s^B) \otimes \mathcal{F}_s^\mu) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. Therefore, by estimate (3.6) and Fubini's theorem, we see that, for every  $\varphi \in \mathcal{B}_2(\mathbb{R}^n)$ , the map

$$\omega^1 \longmapsto \mathbb{E}[\varphi(\bar{X}_s^{t,\bar{\xi}}(\cdot, \omega^1))],$$

from  $\Omega^1$  into  $\mathbb{R}$ , is  $\mathcal{F}_s^\mu$ -measurable. In particular, when  $\varphi \in C_2(\mathbb{R}^n)$ , the continuous process  $(\mathbb{E}[\varphi(\bar{X}_s^{t,\bar{\xi}})])_{s \in [t, T]}$  is  $\mathbb{F}^\mu$ -predictable. Then, by Remark 2.1 it follows that the process  $(\hat{\mathbb{P}}_s^{t,\pi})_{s \in [t, T]}$  is  $\mathbb{F}^\mu$ -predictable.

Finally, we observe that

$$\hat{\mathbb{P}}_s^{t,\pi}(\omega^1)[\varphi] = \mathbb{E}[\varphi(\bar{X}_s^{t,\bar{\xi}}(\cdot, \omega^1))] = \bar{\mathbb{E}}[\varphi(\bar{X}_s^{t,\bar{\xi}}) | \bar{\mathcal{F}}_s^\mu](\omega^1) = \mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu}(\omega^1)[\varphi],$$

$\mathbb{P}^1(d\omega^1)$ -a.s., for every  $\varphi \in \mathcal{B}_2(\mathbb{R}^n)$ . Let  $(\varphi_k)_k \subset \mathcal{B}_2(\mathbb{R}^n)$  be a countable separating class of continuous functions, whose existence is guaranteed for instance by Theorem 2.18 in [3] ( $\varphi_k$  can be taken even bounded). Then, there exists a unique  $\mathbb{P}^1$ -null set  $N^1 \in \mathcal{F}^1$  such that

$$\hat{\mathbb{P}}_s^{t,\pi}(\omega^1)[\varphi_k] = \mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu}(\omega^1)[\varphi_k], \quad \text{for every } k,$$

whenever  $\omega^1 \notin N^1$ . Since  $(\varphi_k)_k$  is separating, we conclude that  $\hat{\mathbb{P}}_s^{t,\pi}$  coincides with  $\mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu}$  on  $\Omega^1 \setminus N^1$ . In other words,  $(\hat{\mathbb{P}}_s^{t,\pi})_{s \in [t, T]}$  is a version of  $(\mathbb{P}_{\bar{X}_s^{t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$ .  $\square$

**Proof of Lemma 3.2.** Fix  $t \in [0, T]$  and consider a generic  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ . Let  $\bar{\xi} \in L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$  be such that  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ . We construct  $\bar{X}^{t,\bar{\xi}}$  using Picard's iterations. More precisely, we define recursively a sequence of  $\mathbb{R}^n$ -valued processes  $(\bar{X}^{m,t,\bar{\xi}})_m$  on  $\bar{\Omega} \times [t, T]$  as follows.

**Recursive construction of the sequence  $(\bar{X}^{m,t,\bar{\xi}})_m$ .** *Definition of  $\bar{X}^{0,t,\bar{\xi}}$ .* We set  $\bar{X}^{0,t,\bar{\xi}} \equiv 0$ . Defining  $\hat{\mathbb{P}}^{0,t,\bar{\xi}}$  by formula (3.7) with  $\bar{X}^{0,t,\bar{\xi}}$  in place of  $\bar{X}^{t,\bar{\xi}}$ , we see that  $\hat{\mathbb{P}}_s^{0,t,\pi} \equiv \delta_0$ , the Dirac delta at zero, for all  $s \in [t, T]$ . In other words, up to a version,  $(\mathbb{P}_{\bar{X}_s^{0,t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$  is identically equal to  $\delta_0$ .

*Definition of  $\bar{X}^{1,t,\bar{\xi}}$ .* The process  $\bar{X}^{1,t,\bar{\xi}}$  is given by:

$$\bar{X}_s^{1,t,\bar{\xi}} = \bar{\xi} + \int_t^s b(r, 0, \delta_0, \bar{I}_r) dr + \int_t^s \sigma(r, 0, \delta_0, \bar{I}_r) d\bar{B}_r,$$

for all  $s \in [t, T]$ . Notice that, by construction, the map  $\bar{X}^{1,t,\bar{\xi}}: ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \bar{\mathcal{F}}) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. Up to indistinguishability, we can suppose that  $\bar{X}^{1,t,\bar{\xi}}: ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \mathcal{F} \otimes \mathcal{F}^1) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. As a consequence, by Fubini's theorem, we can define the  $\mathcal{P}_2(\mathbb{R}^n)$ -valued  $\mathbb{F}^\mu$ -predictable stochastic process  $(\hat{\mathbb{P}}_s^{1,t,\pi})_{s \in [t, T]}$  by formula (3.7) with  $\bar{X}^{1,t,\bar{\xi}}$  in place of  $\bar{X}^{t,\bar{\xi}}$ . Notice that  $(\hat{\mathbb{P}}_s^{1,t,\pi})_{s \in [t, T]}$  is a version of  $(\mathbb{P}_{\bar{X}_s^{1,t,\bar{\xi}}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$ . Moreover, from (3.7), we see that (using the definition of  $\bar{X}_s^{1,t,\bar{\xi}}$ , and the independence of  $\bar{\mathcal{G}}$  and  $\bar{\mathcal{F}}_\infty^B$ )

$$\hat{\mathbb{P}}_s^{1,t,\pi}(\omega^1)[\varphi] = \mathbb{E}[\varphi(\bar{X}_s^{1,t,\bar{\xi}}(\cdot, \omega^1))] = \int_{\mathbb{R}^n} \Phi_{1,\varphi}(\omega^1, s, x) \pi(dx),$$

for every  $\omega^1 \in \Omega^1$  and  $\varphi \in \mathcal{B}_2(\mathbb{R}^n)$ , where  $\Phi_{1,\varphi}: \Omega^1 \times [t, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$  is measurable, with at most quadratic growth in  $x$  uniformly with respect to  $(\omega^1, s)$ , and it is given by

$$\Phi_{1,\varphi}(\omega^1, s, x) := \mathbb{E} \left[ \varphi \left( x + \int_t^s b(r, 0, \delta_0, \bar{I}_r(\cdot, \omega^1)) dr + \int_t^s \sigma(r, 0, \delta_0, \bar{I}_r(\cdot, \omega^1)) d\bar{B}_r \right) \right].$$

Then, we see that the map  $\hat{\mathbb{P}}^{1,t,\cdot}[\varphi]: \Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  is measurable. Indeed, when  $\Phi_{1,\varphi}(\omega^1, s, x) = \ell(\omega^1, s)h(x)$ , for some measurable functions  $\ell$  and  $h$ , with  $\ell$  bounded and  $h$  with at most quadratic growth (namely  $h \in \mathcal{B}_2(\mathbb{R}^n)$ ), the result follows from Remark 2.1. The general case can be proved by a monotone class argument.

Using again Remark 2.1, we conclude that the map  $\hat{\mathbb{P}}^{1,t,\cdot}: \Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  is measurable.

*Definition of  $\bar{X}^{m+1,t,\bar{\xi}}$ , for every integer  $m \geq 1$ .* We define  $\bar{X}^{m+1,t,\bar{\xi}}$  recursively, assuming that  $\bar{X}^{m,t,\bar{\xi}}$  has already been defined. We also assume that the map  $\bar{X}^{m,t,\bar{\xi}}: ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \mathcal{F} \otimes \mathcal{F}^1) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable and that  $(\hat{\mathbb{P}}_s^{m,t,\pi})_{s \in [t, T]}$  is the  $\mathcal{P}_2(\mathbb{R}^n)$ -valued  $\mathbb{F}^\mu$ -predictable stochastic process given by formula (3.7) with  $\bar{X}^{m,t,\bar{\xi}}$  in place of  $\bar{X}^{t,\bar{\xi}}$ . Moreover, we suppose that the map  $\hat{\mathbb{P}}^{m,t,\cdot}: \Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  is measurable. Notice that  $(\hat{\mathbb{P}}_s^{m,t,\pi})_{s \in [t, T]}$  is a version of  $(\mathbb{P}_{\bar{X}^{m,t,\bar{\xi}}}^{\bar{\mathbb{P}}_s^{\mu}})_{s \in [t, T]}$ .

Then, we define  $\bar{X}^{m+1,t,\bar{\xi}}$  as follows:

$$\bar{X}_s^{m+1,t,\bar{\xi}} = \bar{\xi} + \int_t^s b(r, \bar{X}_r^{m,t,\bar{\xi}}, \hat{\mathbb{P}}_r^{m,t,\pi}, \bar{I}_r) dr + \int_t^s \sigma(r, \bar{X}_r^{m,t,\bar{\xi}}, \hat{\mathbb{P}}_r^{m,t,\pi}, \bar{I}_r) d\bar{B}_r,$$

for all  $s \in [t, T]$ . Notice that, by construction, the map  $\bar{X}^{m+1,t,\bar{\xi}}: ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \bar{\mathcal{F}}) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. Therefore, up to indistinguishability, we can suppose that  $\bar{X}^{m+1,t,\bar{\xi}}: ([t, T] \times \Omega \times \Omega^1, \mathcal{B}([t, T]) \otimes \mathcal{F} \otimes \mathcal{F}^1) \rightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n))$  is measurable. Then, by Fubini's theorem, we can define the  $\mathcal{P}_2(\mathbb{R}^n)$ -valued  $\mathbb{F}^\mu$ -predictable stochastic process  $(\hat{\mathbb{P}}_s^{m+1,t,\pi})_{s \in [t, T]}$  by formula (3.7) with  $\bar{X}^{m+1,t,\bar{\xi}}$  in place of  $\bar{X}^{t,\bar{\xi}}$ , namely

$$\hat{\mathbb{P}}_s^{m+1,t,\pi}(\omega^1)[\varphi] = \mathbb{E}[\varphi(\bar{X}_s^{m+1,t,\bar{\xi}}(\cdot, \omega^1))],$$

for every  $\omega^1 \in \Omega^1$ ,  $\varphi \in \mathcal{B}_2(\mathbb{R}^n)$ ,  $s \in [t, T]$ . In particular, we have

$$\begin{aligned} \hat{\mathbb{P}}_s^{m+1,t,\pi}(\omega^1)[\varphi] &= \mathbb{E} \left[ \varphi \left( \bar{\xi} + \int_t^s b(r, \bar{\xi} + \dots, \hat{\mathbb{P}}_r^{m,t,\pi}(\omega^1), \bar{I}_r(\cdot, \omega^1)) dr \right. \right. \\ &\quad \left. \left. + \int_t^s \sigma(r, \bar{\xi} + \dots, \hat{\mathbb{P}}_r^{m,t,\pi}(\omega^1), \bar{I}_r(\cdot, \omega^1)) d\bar{B}_r \right) \right] \\ &= \int_{\mathbb{R}^n} \Phi_{m+1,\varphi}(\omega^1, s, x, \pi) \pi(dx), \end{aligned}$$

for some measurable  $\Phi_{m+1,\varphi}: \Omega^1 \times [t, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$ , with at most quadratic growth in  $(x, \pi)$  uniformly with respect to  $(\omega^1, s)$  (the dependence of  $\Phi_{m+1,\varphi}$  on  $\pi$  is due to the presence of  $\hat{\mathbb{P}}_r^{m,t,\pi}$ ). Then, we see that the map  $\hat{\mathbb{P}}^{m+1,t,\cdot}[\varphi]: \Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  is measurable, as it can be deduced using a monotone class argument, first taking  $\Phi_{m+1,\varphi}$  of the form  $\Phi_{m+1,\varphi}(\omega^1, s, x, \pi) = \ell(\omega^1, s, \pi)h(x)$ , for some  $h \in \mathcal{B}_2(\mathbb{R}^n)$ , and some measurable function  $\ell$  with at most quadratic growth in  $\pi$  uniformly with respect to  $(\omega^1, s)$ . Then, by Remark 2.1, we see that the map  $\hat{\mathbb{P}}^{m+1,t,\cdot}: \Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  is measurable.

**End of the proof of Lemma 3.2.** Now that we have constructed the sequence  $(\bar{X}^{m,t,\bar{\xi}})_m$ , we notice that it can be proved (proceeding for instance along the same lines as in the proof of Theorem IX.2.1 in [27]) that

$$\sup_{s \in [t, T]} |\bar{X}_s^{m,t,\bar{\xi}} - \bar{X}_s^{t,\bar{\xi}}| \xrightarrow[m \rightarrow \infty]{\bar{\mathbb{P}}} 0, \quad (\text{B.1})$$

where the convergence holds in probability. Fix  $s \in [t, T]$  and let us prove that (B.1) implies the following convergence in probability:

$$\mathcal{W}_2(\hat{\mathbb{P}}_s^{m,t,\pi}, \hat{\mathbb{P}}_s^{t,\pi}) \xrightarrow[m \rightarrow \infty]{\mathbb{P}^1} 0. \quad (\text{B.2})$$

In order to prove (B.2), it is enough to show that every subsequence  $(\hat{\mathbb{P}}_s^{m_\ell,t,\pi})_\ell$  admits a subsubsequence  $(\hat{\mathbb{P}}_s^{m_{\ell_h},t,\pi})_h$  for which (B.2) holds. Let us fix a subsequence  $(\hat{\mathbb{P}}_s^{m_\ell,t,\pi})_\ell$ . We begin noting that, by (B.1), we have, for every  $\varphi \in C_2(\mathbb{R}^n)$ ,

$$\hat{\mathbb{P}}_s^{m_\ell,t,\pi}[\varphi] \xrightarrow[\ell \rightarrow \infty]{\mathbb{P}^1} \hat{\mathbb{P}}_s^{t,\pi}[\varphi].$$

Let  $(\varphi_k)_k \subset C_2(\mathbb{R}^n)$  be a countable convergence determining class for the  $\mathcal{W}_2$ -convergence, whose existence follows from Lemma A.2. Then, there exists a unique  $\mathbb{P}^1$ -null set  $N^1 \in \mathcal{F}^1$  and a subsubsequence  $(\hat{\mathbb{P}}_s^{m_{\ell_h},t,\pi})_h$  such that, for all  $\omega^1 \in \Omega^1 \setminus N^1$ ,

$$\hat{\mathbb{P}}_s^{m_{\ell_h},t,\pi}(\omega^1)[\varphi_k] \xrightarrow{h \rightarrow \infty} \hat{\mathbb{P}}_s^{t,\pi}(\omega^1)[\varphi_k], \quad \text{for every } k.$$

By Theorem 6.9 in [31] it follows that, for all  $\omega^1 \in \Omega^1 \setminus N^1$ ,

$$\mathcal{W}_2(\hat{\mathbb{P}}_s^{m_{\ell_h},t,\pi}(\omega^1), \hat{\mathbb{P}}_s^{t,\pi}(\omega^1)) \xrightarrow{h \rightarrow \infty} 0.$$

In particular, the above convergence holds in probability. This concludes the proof of (B.2).

Notice that convergence (B.2) holds for every  $s \in [t, T]$  and  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ . Moreover, for every  $m \in \mathbb{N}$ ,  $\hat{\mathbb{P}}_s^{m,t,\cdot}$  is jointly measurable with respect to  $(\omega^1, s, \pi)$ . Then, we deduce (proceeding for instance as in the first item of Exercise IV.5.17 in [27] or as in Proposition 1 of [29]) that there exists a measurable map  $\mathbb{P}_s^{t,\cdot} : \Omega^1 \times [t, T] \times \mathcal{P}_2(\mathbb{R}^n) \rightarrow \mathcal{P}_2(\mathbb{R}^n)$  such that

$$\mathcal{W}_2(\hat{\mathbb{P}}_s^{m,t,\pi}, \mathbb{P}_s^{t,\pi}) \xrightarrow[m \rightarrow \infty]{\mathbb{P}^1} 0,$$

for every  $s \in [t, T]$  and  $\pi \in \mathcal{P}_2(\mathbb{R}^n)$ . This implies that  $\mathbb{P}_s^{t,\pi}$  coincides  $\mathbb{P}^1$ -a.s. with  $\hat{\mathbb{P}}_s^{t,\pi}$ . By Lemma 3.1 we conclude that  $(\mathbb{P}_s^{t,\pi})_{s \in [t, T]}$  is a version of  $(\mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu})_{s \in [t, T]}$ .  $\square$

## C Stability lemma

For the proof of Theorem 3.1, we need the following stability result.

**Lemma C.1** *Suppose that Assumption (A1) holds.*

- Let  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  be a probability space, on which a  $d$ -dimensional Brownian motion  $\tilde{B} = (\tilde{B}_t)_{t \geq 0}$  is defined.
- For every  $\ell \in \mathbb{N}$ , let  $\tilde{\mathbb{F}}^\ell = (\tilde{\mathcal{F}}_s^\ell)_{s \geq 0}$  be a filtration on  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  such that  $\tilde{B}$  is a Brownian motion with respect to  $\tilde{\mathbb{F}}^\ell$ .
- For every  $\ell \in \mathbb{N}$ , let  $\tilde{\mathbb{F}}^{\mu_\ell} = (\tilde{\mathcal{F}}_s^{\mu_\ell})_{s \geq 0}$ , with  $\tilde{\mathcal{F}}_s^{\mu_\ell} \subset \tilde{\mathcal{F}}_s^\ell$ , be a filtration on  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q})$  independent of  $\tilde{B}$ .
- Let  $(t, x, \tilde{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\tilde{\Omega}, \tilde{\mathcal{F}}, \mathbb{Q}; \mathbb{R}^n)$ , where  $\tilde{\xi}$  is  $\tilde{\mathcal{F}}_t^\ell$ -measurable for every  $\ell \in \mathbb{N}$  and  $\pi = \mathbb{P}_{\tilde{\xi}}$  under  $\mathbb{Q}$ .

For every  $\ell \in \mathbb{N}$ , consider the system of equations:

$$\begin{aligned} d\tilde{X}_s^{t,\tilde{\xi},\ell} &= b(s, \tilde{X}_s^{t,\tilde{\xi},\ell}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) ds + \sigma(s, \tilde{X}_s^{t,\tilde{\xi},\ell}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) d\tilde{B}_s, & \tilde{X}_t^{t,\tilde{\xi},\ell} &= \tilde{\xi}, \\ d\tilde{X}_s^{t,x,\pi,\ell} &= b(s, \tilde{X}_s^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) ds + \sigma(s, \tilde{X}_s^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) d\tilde{B}_s, & \tilde{X}_t^{t,x,\pi,\ell} &= x, \end{aligned}$$

for all  $s \in [t, T]$ , where  $(\tilde{I}_s^\ell)_{s \in [t, T]}$  is an  $A$ -valued  $\tilde{\mathbb{F}}^\ell$ -progressive process. Then

$$\begin{aligned} &\mathbb{E}^\mathbb{Q} \left[ \int_t^T f(s, \tilde{X}_s^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) ds + g(\tilde{X}_T^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_T^{t,x,\pi,\ell}}^{\tilde{\mathcal{F}}_T^{\mu\ell}}) \right] \\ &\xrightarrow{\ell \rightarrow \infty} \mathbb{E}^\mathbb{Q} \left[ \int_t^T f(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^0) ds + g(\tilde{X}_T^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_T^{t,x,\pi,0}}^{\tilde{\mathcal{F}}_T^{\mu 0}}) \right]. \end{aligned}$$

whenever  $\tilde{\rho}^\mathbb{Q}(\tilde{I}^\ell, \tilde{I}^0) := \mathbb{E}^\mathbb{Q}[\int_0^T \rho(\tilde{I}_s^\ell, \tilde{I}_s^0) ds] \rightarrow 0$  as  $\ell \rightarrow \infty$ .

**Proof.** We begin noting that, by standard arguments (based on the Burkholder-Davis-Gundy and Gronwall inequalities), we have

$$\sup_{\ell \in \mathbb{N}} \mathbb{E}^\mathbb{Q} \left[ \sup_{s \in [t, T]} (|\tilde{X}_s^{t,\tilde{\xi},\ell}|^2 + |\tilde{X}_s^{t,x,\pi,\ell}|^q) \right] < \infty, \quad (\text{C.1})$$

for all  $q \geq 1$ . We also have

$$\begin{aligned} \mathbb{E}^\mathbb{Q} \left[ \sup_{s \in [t, T]} |\tilde{X}_s^{t,\tilde{\xi},\ell} - \tilde{X}_s^{t,\tilde{\xi},0}|^2 \right] &\leq C \mathbb{E}^\mathbb{Q} \left[ \int_t^T (|b(s, \tilde{X}_s^{t,\tilde{\xi},0}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^\ell) - b(s, \tilde{X}_s^{t,\tilde{\xi},0}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^0)|^2 \right. \\ &\quad \left. + |\sigma(s, \tilde{X}_s^{t,\tilde{\xi},0}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^\ell) - \sigma(s, \tilde{X}_s^{t,\tilde{\xi},0}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^0)|^2) ds \right], \quad (\text{C.2}) \end{aligned}$$

for some positive constant  $C$ , independent of  $\ell$ . Now, we notice that  $\tilde{\rho}^\mathbb{Q}(\tilde{I}^\ell, \tilde{I}^0) \rightarrow 0$  implies  $\tilde{I}^\ell \rightarrow \tilde{I}^0$  in  $d\mathbb{Q} ds$ -measure, which in turn implies the convergence to zero in  $d\mathbb{Q} ds$ -measure of the integrand in the right-hand side of (C.2). By uniform integrability (which follows from (C.1) and Assumption **(A1)**(ii)), we deduce

$$\mathcal{W}_2(\mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},0}}^{\tilde{\mathcal{F}}_s^{\mu 0}})^2 \leq \mathbb{E}^\mathbb{Q} \left[ |\tilde{X}_s^{t,\tilde{\xi},\ell} - \tilde{X}_s^{t,\tilde{\xi},0}|^2 \bigg| \bigvee_{\ell \in \mathbb{N}} \tilde{\mathcal{F}}_\infty^{\mu\ell} \right] \xrightarrow{\ell \rightarrow \infty} 0,$$

$\mathbb{Q}$ -a.s., for all  $s \in [t, T]$ . Moreover

$$\sup_{s \in [t, T]} \mathcal{W}_2(\mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},\ell}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \mathbb{P}_{\tilde{X}_s^{t,\tilde{\xi},0}}^{\tilde{\mathcal{F}}_s^{\mu 0}})^2 \leq \mathbb{E}^\mathbb{Q} \left[ \sup_{s \in [t, T]} |\tilde{X}_s^{t,\tilde{\xi},\ell} - \tilde{X}_s^{t,\tilde{\xi},0}|^2 \bigg| \bigvee_{\ell \in \mathbb{N}} \tilde{\mathcal{F}}_\infty^{\mu\ell} \right] \xrightarrow{\ell \rightarrow \infty} 0. \quad (\text{C.3})$$

Similarly, we have

$$\begin{aligned} \mathbb{E}^\mathbb{Q} \left[ \sup_{s \in [t, T]} |\tilde{X}_s^{t,x,\pi,\ell} - \tilde{X}_s^{t,x,\pi,0}|^2 \right] &\leq C \mathbb{E}^\mathbb{Q} \left[ \int_t^T (|b(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,0}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) \right. \\ &\quad \left. - b(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^0)|^2 + |\sigma(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,0}}^{\tilde{\mathcal{F}}_s^{\mu\ell}}, \tilde{I}_s^\ell) - \sigma(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,x,\pi,0}}^{\tilde{\mathcal{F}}_s^{\mu 0}}, \tilde{I}_s^0)|^2) ds \right]. \end{aligned}$$

Then, by (C.3), the convergence  $\tilde{I}^\ell \rightarrow \tilde{I}^0$  in  $d\mathbb{Q} ds$ -measure, estimate (C.1), and Assumption **(A1)**(ii), we obtain

$$\mathbb{E}^\mathbb{Q} \left[ \sup_{s \in [t, T]} |\tilde{X}_s^{t,x,\pi,\ell} - \tilde{X}_s^{t,x,\pi,0}|^2 \right] \xrightarrow{\ell \rightarrow \infty} 0. \quad (\text{C.4})$$

Then, by (C.3) and (C.4), we see that  $f(s, \tilde{X}_s^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_s^{t,\xi,\ell}}^{\tilde{\mathcal{F}}_s^{\mu,\ell}}, \tilde{I}_s^\ell) \rightarrow f(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,\xi,0}}^{\tilde{\mathcal{F}}_s^{\mu,0}}, \tilde{I}_s^0)$  as  $\ell \rightarrow \infty$  in  $d\mathbb{Q}ds$ -measure. Therefore, by uniform integrability (which follows from estimate (C.1) and Assumption **(A1)**(ii)), we deduce

$$\mathbb{E}^\mathbb{Q} \left[ \int_t^T f(s, \tilde{X}_s^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_s^{t,\xi,\ell}}^{\tilde{\mathcal{F}}_s^{\mu,\ell}}, \tilde{I}_s^\ell) ds \right] \xrightarrow{\ell \rightarrow \infty} \mathbb{E}^\mathbb{Q} \left[ \int_t^T f(s, \tilde{X}_s^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_s^{t,\xi,0}}^{\tilde{\mathcal{F}}_s^{\mu,0}}, \tilde{I}_s^0) ds \right].$$

Using again (C.3) and (C.4), we obtain the  $\mathbb{Q}$ -a.s. pointwise convergence  $g(\tilde{X}_T^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_T^{t,\xi,\ell}}^{\tilde{\mathcal{F}}_T^{\mu,\ell}}) \rightarrow g(\tilde{X}_T^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_T^{t,\xi,0}}^{\tilde{\mathcal{F}}_T^{\mu,0}})$  as  $\ell \rightarrow \infty$ . By estimate (2.6) together with the polynomial growth condition of  $g$  in Assumption **(A1)**(ii), we can apply Lebesgue's dominated convergence theorem and obtain

$$\mathbb{E}^\mathbb{Q} \left[ g(\tilde{X}_T^{t,x,\pi,\ell}, \mathbb{P}_{\tilde{X}_T^{t,\xi,\ell}}^{\tilde{\mathcal{F}}_T^{\mu,\ell}}) \right] \xrightarrow{\ell \rightarrow \infty} \mathbb{E}^\mathbb{Q} \left[ g(\tilde{X}_T^{t,x,\pi,0}, \mathbb{P}_{\tilde{X}_T^{t,\xi,0}}^{\tilde{\mathcal{F}}_T^{\mu,0}}) \right],$$

which concludes the proof.  $\square$

## D On a different randomization of the control

In the present appendix we introduce, following [21], a different kind of randomization, which in our paper turns out to be useful in the proof of Theorem 4.1. More precisely, for every  $t \in [0, T]$ ,  $a_0 \in A$ , consider the  $A$ -valued piecewise constant process  $\bar{I}^{t,a_0} = (\bar{I}_s^{t,a_0})_{s \geq t}$  on  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$  given by:

$$\bar{I}_s^{t,a_0}(\omega, \omega^1) = \sum_{\substack{n \geq 0 \\ t < T_{n+1}(\omega^1)}} (a_0 1_{\{T_n(\omega^1) < t\}} + (\mathcal{A}_n(\omega^1))_{s \wedge T}(\omega) 1_{\{t \leq T_n(\omega^1)\}}) 1_{[T_n(\omega^1), T_{n+1}(\omega^1))}(s), \quad (\text{D.1})$$

for all  $s \geq t$ , where we recall that  $T_0 = 0$  and  $\mathcal{A}_0 = \bar{\alpha}$ . The process  $\bar{I} = (\bar{I}_s)_{s \geq 0}$  defined in (3.3) corresponds to  $\bar{I}^{0,a_0} = (\bar{I}_s^{0,a_0})_{s \geq 0}$ , for any  $a_0 \in A$  (when  $t = 0$ ,  $a_0$  plays no role in (D.1)).

Let  $\bar{\mathbb{F}}^{B,t} = (\bar{\mathcal{F}}_s^{B,t})_{s \geq t}$  (resp.  $\bar{\mathbb{F}}^{\mu,t} = (\bar{\mathcal{F}}_s^{\mu,t})_{s \geq t}$ ) be the  $\mathbb{P}$ -completion of the filtration generated by  $(\bar{B}_s - \bar{B}_t)_{s \geq t}$  (resp.  $\bar{\mu} 1_{(t,\infty) \times \mathcal{A}}$ ), and let  $\bar{\mathbb{F}}^{B,\mu,t} = (\bar{\mathcal{F}}_s^{B,\mu,t})_{s \geq t}$  denote the  $\mathbb{P}$ -completion of the filtration generated by  $(\bar{B}_s - \bar{B}_t)_{s \geq t}$  and  $\bar{\mu} 1_{(t,\infty) \times \mathcal{A}}$ . If we randomize the control in (2.3)-(2.4) by means of the process  $\bar{I}^{t,a_0}$ , we obtain, for every  $(x, \bar{\xi}) \in \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\bar{\mathbb{P}}$ :

$$d\bar{X}_s^{t,\bar{\xi},a_0} = b(s, \bar{X}_s^{t,\bar{\xi},a_0}, \mathbb{P}_{\bar{X}_s^{t,\bar{\xi},a_0}}^{\bar{\mathcal{F}}_s^{\mu,t}}, \bar{I}_s^{t,a_0}) ds + \sigma(s, \bar{X}_s^{t,\bar{\xi},a_0}, \mathbb{P}_{\bar{X}_s^{t,\bar{\xi},a_0}}^{\bar{\mathcal{F}}_s^{\mu,t}}, \bar{I}_s^{t,a_0}) d\bar{B}_s, \quad (\text{D.2})$$

$$d\bar{X}_s^{t,x,\pi,a_0} = b(s, \bar{X}_s^{t,x,\pi,a_0}, \mathbb{P}_{\bar{X}_s^{t,\xi,a_0}}^{\bar{\mathcal{F}}_s^{\mu,t}}, \bar{I}_s^{t,a_0}) ds + \sigma(s, \bar{X}_s^{t,x,\pi,a_0}, \mathbb{P}_{\bar{X}_s^{t,\xi,a_0}}^{\bar{\mathcal{F}}_s^{\mu,t}}, \bar{I}_s^{t,a_0}) d\bar{B}_s, \quad (\text{D.3})$$

for all  $s \in [t, T]$ , with  $\bar{X}_t^{t,\bar{\xi},a_0} = \bar{\xi}$  and  $\bar{X}_t^{t,x,\pi,a_0} = x$ . Under Assumption **(A1)**, there exists a unique (up to indistinguishability) pair  $(\bar{X}_s^{t,\bar{\xi},a_0}, \bar{X}_s^{t,x,\pi,a_0})_{s \in [t, T]}$  of continuous  $(\bar{\mathcal{F}}_s^{B,\mu,t} \vee \bar{\mathcal{G}})_s$ -adapted processes solution to equations (D.2)-(D.3), satisfying

$$\bar{\mathbb{E}} \left[ \sup_{s \in [t, T]} (|\bar{X}_s^{t,\bar{\xi},a_0}|^2 + |\bar{X}_s^{t,x,\pi,a_0}|^q) \right] < \infty,$$

for all  $q \geq 1$ .

Let  $\mathbb{F}^{\mu,t} = (\mathcal{F}_s^{\mu,t})_{s \geq t}$  be the  $\mathbb{P}^1$ -completion of the filtration generated by  $\mu 1_{(t,\infty) \times \mathcal{A}_{\text{step}}}$ , and denote by  $\mathcal{P}(\mathbb{F}^{\mu,t})$  the predictable  $\sigma$ -algebra on  $\Omega^1 \times [t, \infty)$  corresponding to  $\mathbb{F}^{\mu,t}$ . Then, we

define  $\mathcal{V}_t$  as the set of  $\mathcal{P}(\mathbb{F}^{\mu,t}) \otimes \mathcal{B}(\mathcal{A})$ -measurable maps  $\nu: \Omega^1 \times [t, \infty) \times \mathcal{A} \rightarrow (0, \infty)$ , with  $0 < \inf_{\Omega^1 \times [t, \infty) \times \mathcal{A}} \nu \leq \sup_{\Omega^1 \times [t, \infty) \times \mathcal{A}} \nu < \infty$ . Given  $\nu \in \mathcal{V}_t$ , we define  $\nu^* \in \mathcal{V}$  as  $\nu^* = 1_{\Omega^1 \times [0, t) \times \mathcal{A}} + \nu 1_{\Omega^1 \times [t, \infty) \times \mathcal{A}}$ . We denote  $\mathbb{P}^\nu$  (resp.  $\bar{\mathbb{P}}^\nu$ ) the probability  $\mathbb{P}^{\nu^*}$  (resp.  $\bar{\mathbb{P}}^{\nu^*}$ ), and  $\mathbb{E}^\nu$  (resp.  $\bar{\mathbb{E}}^\nu$ ) the expectation  $\mathbb{E}^{\nu^*}$  (resp.  $\bar{\mathbb{E}}^{\nu^*}$ ). Then, for every  $\nu \in \mathcal{V}_t$ , we define the gain functional (notice that  $J^\mathcal{R}(t, x, \pi, a_0, \nu)$  does not depend on the value of  $\nu^*$  on  $\Omega^1 \times [0, t) \times \mathcal{A}$ )

$$J^\mathcal{R}(t, x, \pi, a_0, \nu) = \bar{\mathbb{E}}^\nu \left[ \int_t^T f(s, \bar{X}_s^{t,x,\pi,a_0}, \mathbb{P}_{\bar{X}_s^{t,\bar{\varepsilon},a_0}}^{\bar{\mathcal{F}}_s^{\mu,t}}, \bar{I}_s^{t,a_0}) ds + g(\bar{X}_T^{t,x,\pi,a_0}, \mathbb{P}_{\bar{X}_T^{t,\bar{\varepsilon},a_0}}^{\bar{\mathcal{F}}_T^{\mu,t}}) \right]$$

and the value function

$$V^\mathcal{R}(t, x, \pi, a_0) = \sup_{\nu \in \mathcal{V}_t} J^\mathcal{R}(t, x, \pi, a_0, \nu).$$

Finally, let  $\mathbb{F}^{B,t} = (\mathcal{F}_s^{B,t})_{s \geq t}$  be the  $\mathbb{P}$ -completion of the filtration generated by  $(B_s - B_t)_{s \geq t}$ , and let  $\mathcal{A}_t$  denote the set of  $\mathbb{F}^{B,t}$ -progressive processes  $\alpha: \Omega \times [t, T] \rightarrow A$ . Given  $\alpha \in \mathcal{A}_t$ , we define  $\alpha^* \in \mathcal{A}$  as  $\alpha^* = \bar{a} 1_{\Omega \times [0, t)} + \alpha 1_{\Omega \times [t, T]}$ , for some deterministic and fixed point  $\bar{a} \in A$ . Then, we denote  $J(t, x, \pi, \alpha^*)$  simply by  $J(t, x, \pi, \alpha)$  (notice that  $J(t, x, \pi, \alpha^*)$  does not depend on the value of  $\alpha^*$  on  $\Omega \times [0, t)$ , namely on  $\bar{a}$ ).

**Theorem D.1** *Under Assumption (A1), we have the following identities:*

$$\begin{aligned} V(t, x, \pi) &:= \sup_{\alpha \in \mathcal{A}} J(t, x, \pi, \alpha) = \sup_{\alpha \in \mathcal{A}_t} J(t, x, \pi, \alpha) = \sup_{\nu \in \mathcal{V}_t} J^\mathcal{R}(t, x, \pi, a_0, \nu) =: V^\mathcal{R}(t, x, \pi, a_0) \\ &= \sup_{\nu \in \mathcal{V}} J^\mathcal{R}(t, x, \pi, \nu) =: V^\mathcal{R}(t, x, \pi), \end{aligned} \quad (\text{D.4})$$

for all  $(t, x, \pi, a_0) \in [0, T] \times \mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^n) \times A$ .

**Remark D.1** From Theorem 3.1 we conclude that the function  $V^\mathcal{R}(t, x, \pi, a_0)$  does not depend on  $a_0 \in A$  and coincides with the function  $V^\mathcal{R}(t, x, \pi)$  defined in (3.8).  $\square$

**Proof.** When  $t = 0$ , we see that, for every  $a_0 \in A$ , we have  $\bar{I}^{0,a_0} = \bar{I}$ ,  $\mathcal{A}_0 = \mathcal{A}$ , and  $\mathcal{V}_0 = \mathcal{V}$ . Therefore,  $V^\mathcal{R}(0, x, \pi, a_0)$  coincides with  $V^\mathcal{R}(0, x, \pi)$ , so the result follows from Theorem 3.1. When  $t > 0$ , we proceed along the same lines as in the proof of Theorem 3.1 for the case  $t = 0$ , with  $(\bar{B}_s)_{s \geq 0}$ ,  $\bar{\mathbb{F}}^B = (\bar{\mathcal{F}}_s^B)_{s \geq 0}$ ,  $\mathcal{A}$ ,  $\bar{\mu}$ ,  $\bar{\mathbb{F}}^{B,\mu} = (\bar{\mathcal{F}}_s^{B,\mu})_{s \geq 0}$ ,  $\mathcal{V}$  replaced respectively by  $(\bar{B}_s - \bar{B}_t)_{s \geq t}$ ,  $\bar{\mathbb{F}}^{B,t} = (\bar{\mathcal{F}}_s^{B,t})_{s \geq t}$ ,  $\mathcal{A}_t$ ,  $\bar{\mu} 1_{(t, \infty) \times \mathcal{A}}$ ,  $\bar{\mathbb{F}}^{B,\mu,t} = (\bar{\mathcal{F}}_s^{B,\mu,t})_{s \geq t}$ ,  $\mathcal{V}_t$ . Then, we obtain

$$\sup_{\alpha \in \mathcal{A}_t} J(t, x, \pi, \alpha) = \sup_{\nu \in \mathcal{V}_t} J^\mathcal{R}(t, x, \pi, a_0, \nu).$$

This implies that  $V^\mathcal{R}(t, x, \pi, a_0)$  does not depend on  $a_0 \in A$ , since the left-hand side of the above inequality does not depend on it.

By Theorem 3.1, equivalence (D.4) follows if we prove the following inequalities

$$V(t, x, \pi) \geq \sup_{\alpha \in \mathcal{A}_t} J(t, x, \pi, \alpha), \quad \sup_{\nu \in \mathcal{V}_t} J^\mathcal{R}(t, x, \pi, a_0, \nu) \geq V^\mathcal{R}(t, x, \pi). \quad (\text{D.5})$$

Since for every  $\alpha \in \mathcal{A}_t$  we have, by definition,  $J(t, x, \pi, \alpha) = J(t, x, \pi, \alpha^*)$ , where  $\alpha^* = \bar{a} 1_{\Omega \times [0, t)} + \alpha 1_{\Omega \times [t, T]}$ , we see that  $\sup_{\alpha \in \mathcal{A}_t} J(t, x, \pi, \alpha) \leq \sup_{\alpha \in \mathcal{A}} J(t, x, \pi, \alpha) = V(t, x, \pi)$ . Therefore, the first inequality in (D.5) is proved.

In order to establish the second inequality in (D.5), we fix  $(t, x, \bar{\xi}, \pi) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n) \times \mathcal{P}_2(\mathbb{R}^n)$ , with  $\pi = \mathbb{P}_\xi$  under  $\bar{\mathbb{P}}$ , and we take a particular probabilistic setting for the randomized McKean-Vlasov control problem. More precisely, we first consider another probabilistic framework for randomized problem, where the objects  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  $(\Omega^1, \mathcal{F}^1, \mathbb{P}^1)$ ,  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$ ,  $\bar{B}$ ,  $\bar{\mu}$ ,  $(T_n, \mathcal{A}_n)$ ,  $\bar{I}$  are replaced respectively by  $(\Omega^0, \mathcal{F}^0, \mathbb{P}^0)$ ,  $(\bar{\Omega}^1, \bar{\mathcal{F}}^1, \bar{\mathbb{P}}^1)$ ,  $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$ ,  $\bar{B}$ ,  $\bar{\mu}$ ,  $(\bar{T}_n, \bar{\mathcal{A}}_n)$ ,  $\bar{I}$ .

Let  $\hat{\Omega} = \bar{\Omega} \times \bar{\Omega}$ ,  $\hat{\mathcal{F}}$  the  $\bar{\mathbb{P}} \otimes \bar{\mathbb{P}}$ -completion of  $\bar{\mathcal{F}} \otimes \bar{\mathcal{F}}$ ,  $\hat{\mathbb{P}}$  the extension of  $\bar{\mathbb{P}} \otimes \bar{\mathbb{P}}$  to  $\hat{\mathcal{F}}$ , and  $\hat{\mathbb{E}}$  the  $\hat{\mathbb{P}}$ -expected value. Let also  $\hat{\mathcal{G}}$  be the canonical extension of  $\bar{\mathcal{G}}$  to  $\hat{\Omega}$ . Define  $\hat{\xi}(\bar{\omega}, \bar{\omega}) := \bar{\xi}(\bar{\omega})$  and

$$\begin{aligned} \hat{B}_s(\bar{\omega}, \bar{\omega}) &:= \bar{B}_s(\bar{\omega}) 1_{\{s \leq t\}} + (\bar{B}_s(\bar{\omega}) - \bar{B}_t(\bar{\omega}) + \bar{B}_t(\bar{\omega})) 1_{\{s > t\}}, \\ \hat{\mu}(\bar{\omega}, \bar{\omega}; ds d\alpha) &:= \bar{\mu}(\bar{\omega}; ds d\alpha) 1_{\{s \leq t\}} + \bar{\mu}(\bar{\omega}; ds d\alpha) 1_{\{s > t\}}. \end{aligned}$$

Notice that  $\pi = \mathbb{P}_\xi$  under  $\hat{\mathbb{P}}$ ,  $\hat{B} = (\hat{B}_s)_{s \geq 0}$  is a Brownian motion on  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{\mathbb{P}})$ , and  $\hat{\mu}$  is a Poisson random measure with compensator  $\lambda(d\alpha) ds$  under  $\hat{\mathbb{P}}$ , with respect to its natural filtration. We also define as in (3.3) the  $A$ -valued piecewise constant process  $\hat{I} = (\hat{I}_s)_{s \geq 0}$  associated to  $\hat{\mu}$ , which in the present case takes the following form:

$$\begin{aligned} \hat{I}_s(\bar{\omega}, \bar{\omega}) &= \bar{I}_s(\bar{\omega}) 1_{\{s \leq t\}} \\ &+ \sum_{\substack{n \geq 0 \\ t < T_{n+1}(\omega^1)}} (\bar{I}_s(\bar{\omega}) 1_{\{T_n(\omega^1) < t\}} + (\mathcal{A}_n(\omega^1))_{s \wedge T}(\omega) 1_{\{t \leq T_n(\omega^1)\}}) 1_{[T_n(\omega^1), T_{n+1}(\omega^1))}(s) 1_{\{s > t\}}. \end{aligned}$$

In particular,  $\hat{I}_t = \bar{I}_t$ . We define  $\hat{\mathbb{F}}^{B, \mu} = (\hat{\mathcal{F}}_s^{B, \mu})_{s \geq 0}$  (resp.  $\hat{\mathbb{F}}^\mu = (\hat{\mathcal{F}}_s^\mu)_{s \geq 0}$ ) as the  $\hat{\mathbb{P}}$ -completion of the filtration generated by  $\hat{B}$  and  $\hat{\mu}$  (resp.  $\hat{\mu}$ ). We denote  $(\hat{X}_s^{t, \hat{\xi}}, \hat{X}_s^{t, x, \pi})_{s \in [t, T]}$  the unique (up to indistinguishability) continuous  $(\hat{\mathcal{F}}_s^{B, \mu} \vee \hat{\mathcal{G}})_s$ -adapted solution to equations (3.4)-(3.5) on  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{\mathbb{P}})$  with  $\bar{\xi}$ ,  $\bar{B}$ ,  $\bar{I}$ ,  $\bar{\mathcal{F}}^\mu$  replaced respectively by  $\hat{\xi}$ ,  $\hat{B}$ ,  $\hat{I}$ ,  $\hat{\mathcal{F}}^\mu$ . For later use, we also consider, for every  $\bar{\omega} \in \bar{\Omega}$ , the unique (up to indistinguishability) continuous  $(\bar{\mathcal{F}}_s^{B, \mu, t} \vee \bar{\mathcal{G}})_s$ -adapted solution  $(\bar{X}_s^{t, \hat{\xi}, \bar{I}_t(\bar{\omega})}, \bar{X}_s^{t, x, \pi, \bar{I}_t(\bar{\omega})})_{s \in [t, T]}$  to equations (D.2)-(D.3) with  $a_0$  replaced by  $\bar{I}_t(\bar{\omega})$ . Then, we see that, for  $\bar{\mathbb{P}}$ -a.e.  $\bar{\omega} \in \bar{\Omega}$ ,  $(\hat{X}_s^{t, \hat{\xi}}(\bar{\omega}, \cdot), \hat{X}_s^{t, x, \pi}(\bar{\omega}, \cdot))_{s \in [t, T]}$  and  $(\bar{X}_s^{t, \hat{\xi}, \bar{I}_t(\bar{\omega})}, \bar{X}_s^{t, x, \pi, \bar{I}_t(\bar{\omega})})_{s \in [t, T]}$  solve the same system of equations. Therefore, by pathwise uniqueness, for  $\bar{\mathbb{P}}$ -a.e.  $\bar{\omega} \in \bar{\Omega}$ , we have  $\hat{X}_s^{t, \hat{\xi}}(\bar{\omega}, \bar{\omega}) = \bar{X}_s^{t, \hat{\xi}, \bar{I}_t(\bar{\omega})}(\bar{\omega})$  and  $\hat{X}_s^{t, x, \pi}(\bar{\omega}, \bar{\omega}) = \bar{X}_s^{t, x, \pi, \bar{I}_t(\bar{\omega})}(\bar{\omega})$ , for all  $s \in [t, T]$ ,  $\bar{\mathbb{P}}(d\bar{\omega})$ -almost surely.

Let  $\mathcal{P}(\hat{\mathbb{F}}^\mu)$  be the predictable  $\sigma$ -algebra on  $\hat{\Omega} \times \mathbb{R}_+$  corresponding to  $\hat{\mathbb{F}}^\mu$ . In order to define the randomized McKean-Vlasov control problem on  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{\mathbb{P}})$ , we introduce the set  $\hat{\mathcal{V}}$  of all  $\mathcal{P}(\hat{\mathbb{F}}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable maps  $\hat{\nu}: \hat{\Omega} \times \mathbb{R}_+ \times \mathcal{A} \rightarrow (0, \infty)$ , satisfying  $0 < \inf_{\hat{\Omega} \times \mathbb{R}_+ \times \mathcal{A}} \hat{\nu} \leq \sup_{\hat{\Omega} \times \mathbb{R}_+ \times \mathcal{A}} \hat{\nu} < \infty$ . Then, we define in an obvious way  $\kappa^{\hat{\nu}}$ ,  $\hat{\mathbb{P}}^{\hat{\nu}}$ ,  $\hat{\mathbb{E}}^{\hat{\nu}}$ ,  $\hat{J}^{\mathcal{R}}(t, x, \pi, \hat{\nu})$ , and the corresponding value function  $\hat{V}^{\mathcal{R}}(t, x, \pi)$ . We recall from step I of the proof of Theorem 3.1 that  $\hat{V}^{\mathcal{R}}(t, x, \pi) = V^{\mathcal{R}}(t, x, \pi)$ .

We can now prove the second inequality in (D.5), namely

$$V^{\mathcal{R}}(t, x, \pi) = \hat{V}^{\mathcal{R}}(t, x, \pi) := \sup_{\hat{\nu} \in \hat{\mathcal{V}}} \hat{J}^{\mathcal{R}}(t, x, \pi, \hat{\nu}) \leq \sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu). \quad (\text{D.6})$$

Fix  $\hat{\nu} \in \hat{\mathcal{V}}$ . We begin noting that, since  $\hat{\nu}$  is  $\mathcal{P}(\hat{\mathbb{F}}^\mu) \otimes \mathcal{B}(\mathcal{A})$ -measurable, up to a  $\hat{\mathbb{P}}$ -null set,  $\hat{\nu}$  depends only  $(\bar{\omega}^1, \omega^1)$ . Now, by a monotone class argument, we see that there exists a  $\bar{\mathbb{P}}^1$ -null set  $\bar{N}^1 \in \bar{\mathcal{F}}^1$  such that  $\nu^{\bar{\omega}^1} = \nu_s^{\bar{\omega}^1}(\omega^1, \alpha): \Omega^1 \times [t, \infty) \times \mathcal{A} \rightarrow (0, \infty)$ , given by

$$\nu_s^{\bar{\omega}^1}(\omega^1, \alpha) := \hat{\nu}_s(\bar{\omega}^1, \omega^1, \alpha), \quad \text{for all } (\bar{\omega}^1, \omega^1, s, \alpha) \in \bar{\Omega}^1 \times \Omega^1 \times [t, \infty) \times \mathcal{A},$$



is an element of  $\mathcal{V}_t$ , for every  $\tilde{\omega}^1 \notin \tilde{N}^1$ . In other words, for every  $\tilde{\omega}^1 \notin \tilde{N}^1$ ,  $\nu^{\tilde{\omega}^1}$  is a  $\mathcal{P}(\mathbb{F}^{\mu,t}) \otimes \mathcal{B}(\mathcal{A})$ -measurable map satisfying  $0 < \inf_{\Omega^1 \times [t,\infty) \times \mathcal{A}} \nu^{\tilde{\omega}^1} \leq \sup_{\Omega^1 \times [t,\infty) \times \mathcal{A}} \nu^{\tilde{\omega}^1} < \infty$ . Therefore, by Fubini's theorem,

$$\begin{aligned} \hat{J}^{\mathcal{R}}(t, x, \pi, \hat{\nu}) &= \hat{\mathbb{E}} \left[ \kappa_T^{\hat{\nu}} \left( \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \right) \right] \\ &= \int_{\tilde{\Omega}} \hat{\mathbb{E}} \left[ \kappa_T^{\nu^{\tilde{\omega}^1}} \left( \int_t^T f(s, \bar{X}_s^{t,x,\pi}, \bar{I}_s(\tilde{\omega}), \mathbb{P}_{\bar{X}_s^{t,\xi}, \bar{I}_s(\tilde{\omega})}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s(\tilde{\omega})) ds + g(\bar{X}_T^{t,x,\pi}, \bar{I}_t(\tilde{\omega}), \mathbb{P}_{\bar{X}_T^{t,\xi}, \bar{I}_t(\tilde{\omega})}^{\bar{\mathcal{F}}_T^\mu}) \right) \right] \tilde{\mathbb{P}}(d\tilde{\omega}) \\ &= \int_{\tilde{\Omega}} J^{\mathcal{R}}(t, x, \pi, \bar{I}_t(\tilde{\omega}), \nu^{\tilde{\omega}^1}) \tilde{\mathbb{P}}(d\tilde{\omega}) \leq \sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu), \end{aligned}$$

for any  $a_0 \in A$  (recall that  $\sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu)$  does not depend on  $a_0 \in A$ ). From the arbitrariness of  $\hat{\nu} \in \hat{\mathcal{V}}$ , we deduce that  $\sup_{\hat{\nu} \in \hat{\mathcal{V}}} \hat{J}^{\mathcal{R}}(t, x, \pi, \hat{\nu}) \leq \sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu)$ , hence establishing (D.6), and consequently the second inequality in (D.5).  $\square$

**Corollary D.1** *Under Assumption (A1), we have*

$$V(t, x, \pi) = \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right], \quad (\text{D.7})$$

$\mathbb{P}^1$ -a.s., for all  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ .

**Proof.** Fix  $(t, x, \bar{\xi}) \in [0, T] \times \mathbb{R}^n \times L^2(\bar{\Omega}, \bar{\mathcal{G}}, \bar{\mathbb{P}}; \mathbb{R}^n)$ , with  $\pi = \mathbb{P}_{\bar{\xi}}$  under  $\bar{\mathbb{P}}$ . We have

$$\begin{aligned} &\mathbb{E}^1 \left[ \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \right] \\ &\geq \mathbb{E}^1 \left[ \operatorname{ess\,sup}_{\nu \in \mathcal{V}_{1,t}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \right] \\ &\geq \sup_{\nu \in \mathcal{V}_{1,t}} \mathbb{E}^1 \left[ \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \right]. \end{aligned}$$

By the Bayes formula, and recalling that  $\kappa_t^\nu = 1$  whenever  $\nu \in \mathcal{V}_{1,t}$ , we obtain

$$\begin{aligned} &\sup_{\nu \in \mathcal{V}_{1,t}} \mathbb{E}^1 \left[ \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \right] \\ &= \sup_{\nu \in \mathcal{V}_{1,t}} \mathbb{E}^1 \left[ \mathbb{E}^1 \left[ \kappa_T^\nu \left( \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \right) \middle| \mathcal{F}_t^\mu \right] \right] \\ &= \sup_{\nu \in \mathcal{V}_{1,t}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \right] = V(t, x, \pi), \end{aligned}$$

where the last equality follows from Remark 3.4. Then, we conclude that

$$\mathbb{E}^1 \left[ \operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \right] \geq V(t, x, \pi). \quad (\text{D.8})$$

Let us now prove the following inequality: for every  $\nu \in \mathcal{V}$ ,  $\mathbb{P}^1$ -a.s.,

$$\mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \leq V(t, x, \pi). \quad (\text{D.9})$$

Suppose we have already proved (D.9). Hence,  $\mathbb{P}^1$ -a.s.,

$$\operatorname{ess\,sup}_{\nu \in \mathcal{V}} \mathbb{E}^\nu \left[ \int_t^T \mathbb{E}[f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s)] ds + \mathbb{E}[g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu})] \middle| \mathcal{F}_t^\mu \right] \leq V(t, x, \pi).$$

From the above inequality and (D.8), it is then easy to see that equality (D.7) holds. It remains to prove (D.9). To this end, we notice that (D.9) holds if and only if the following inequality holds: for every  $\nu \in \mathcal{V}$ ,  $\bar{\mathbb{P}}$ -a.s.,

$$\bar{\mathbb{E}}^\nu \left[ \int_t^T f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \middle| \bar{\mathcal{F}}_t^\mu \right] \leq V(t, x, \pi). \quad (\text{D.10})$$

Now, consider the same probabilistic setting introduced in the proof of Theorem D.1:  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{\mathbb{P}})$ ,  $\hat{\mathcal{G}}, \hat{B}, \hat{\mu}, \hat{\mathbb{F}}^{B,\mu} = (\hat{\mathcal{F}}_s^{B,\mu})_{s \geq 0}, \hat{\mathbb{F}}^\mu = (\hat{\mathcal{F}}_s^\mu)_{s \geq 0}, \hat{I}, \hat{X}^{t,\xi}, \hat{X}^{t,x,\pi}, \hat{\mathcal{V}}, \hat{\mathcal{V}}_{1,t}, \hat{\mathbb{P}}^{\hat{\nu}}, \hat{\mathbb{E}}^{\hat{\nu}}, \hat{\mathcal{J}}^{\mathcal{R}}(t, x, \pi, \hat{\nu}), \hat{\mathcal{V}}^{\mathcal{R}}(t, x, \pi)$ . Observe that (D.10) holds if and only if the following inequality holds: for every  $\hat{\nu} \in \hat{\mathcal{V}}$ ,  $\hat{\mathbb{P}}$ -a.s.,

$$\hat{\mathbb{E}}^{\hat{\nu}} \left[ \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\hat{\mathcal{F}}_s^\mu}, \hat{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\hat{\mathcal{F}}_T^\mu}) \middle| \hat{\mathcal{F}}_t^\mu \right] \leq V(t, x, \pi). \quad (\text{D.11})$$

Indeed, let us prove that if (D.11) holds then (D.10) holds as well (the other implication has a similar proof). Fix  $\nu \in \mathcal{V}$ . Then, proceeding as in step I of the proof of Theorem 3.1, we see that there exists  $\hat{\nu} \in \hat{\mathcal{V}}$  such that

$$\frac{\bar{\kappa}_T^\nu}{\bar{\kappa}_t^\nu} \left( \int_t^T f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \right), \bar{B}, \bar{\mu}$$

and

$$\frac{\kappa_T^{\hat{\nu}}}{\kappa_t^{\hat{\nu}}} \left( \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\hat{\mathcal{F}}_s^\mu}, \hat{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\hat{\mathcal{F}}_T^\mu}) \right), \hat{B}, \hat{\mu}$$

have the same joint law. As a consequence,

$$\bar{\mathbb{E}}^\nu \left[ \int_t^T f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \middle| \bar{\mathcal{F}}_t^\mu \right]$$

and

$$\hat{\mathbb{E}}^{\hat{\nu}} \left[ \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\hat{\mathcal{F}}_s^\mu}, \hat{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\hat{\mathcal{F}}_T^\mu}) \middle| \hat{\mathcal{F}}_t^\mu \right]$$

have the same law. In particular, we have

$$\begin{aligned} & \bar{\mathbb{P}} \left( \bar{\mathbb{E}}^\nu \left[ \int_t^T f(s, \bar{X}_s^{t,x,\pi}, \mathbb{P}_{\bar{X}_s^{t,\xi}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s) ds + g(\bar{X}_T^{t,x,\pi}, \mathbb{P}_{\bar{X}_T^{t,\xi}}^{\bar{\mathcal{F}}_T^\mu}) \middle| \bar{\mathcal{F}}_t^\mu \right] \leq V(t, x, \pi) \right) \\ &= \hat{\mathbb{P}} \left( \hat{\mathbb{E}}^{\hat{\nu}} \left[ \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\hat{\mathcal{F}}_s^\mu}, \hat{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\hat{\mathcal{F}}_T^\mu}) \middle| \hat{\mathcal{F}}_t^\mu \right] \leq V(t, x, \pi) \right) = 1, \end{aligned}$$

where the last equality follows from the assumption that (D.11) holds. This implies that (D.10) also holds for  $\nu$ . Since  $\nu$  was arbitrary, the claim follows.

Let us now prove that (D.11) holds. For every  $\hat{\nu} \in \hat{\mathcal{V}}$ , by the Bayes formula, and proceeding as in the proof of Theorem D.1, we find

$$\hat{\mathbb{E}}^{\hat{\nu}} \left[ \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\hat{\mathcal{F}}_s^\mu}, \hat{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\hat{\mathcal{F}}_T^\mu}) \middle| \hat{\mathcal{F}}_t^\mu \right]$$

$$\begin{aligned}
&= \mathbb{E} \left[ \frac{\kappa_T^{\nu'}}{\kappa_t^{\nu'}} \left( \int_t^T f(s, \hat{X}_s^{t,x,\pi}, \mathbb{P}_{\hat{X}_s^{t,\xi}}^{\hat{\mathcal{F}}_s^\mu}, \hat{I}_s) ds + g(\hat{X}_T^{t,x,\pi}, \mathbb{P}_{\hat{X}_T^{t,\xi}}^{\hat{\mathcal{F}}_T^\mu}) \right) \middle| \hat{\mathcal{F}}_t^\mu \right] \\
&= \mathbb{E} \left[ \frac{\kappa_T^{\nu'}}{\kappa_t^{\nu'}} \left( \int_t^T f(s, \bar{X}_s^{t,x,\pi, \check{I}_t}, \mathbb{P}_{\bar{X}_s^{t,\xi, \check{I}_t}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s^{\check{I}_t}) ds + g(\bar{X}_T^{t,x,\pi, \check{I}_t}, \mathbb{P}_{\bar{X}_T^{t,\xi, \check{I}_t}}^{\bar{\mathcal{F}}_T^\mu}) \right) \middle| \hat{\mathcal{F}}_t^\mu \right].
\end{aligned}$$

Then, by the freezing lemma (see for instance Proposition 10.1.2 in [32]), we obtain

$$\begin{aligned}
&\hat{\mathbb{E}} \left[ \frac{\kappa_T^{\nu'}}{\kappa_t^{\nu'}} \left( \int_t^T f(s, \bar{X}_s^{t,x,\pi, \check{I}_t}, \mathbb{P}_{\bar{X}_s^{t,\xi, \check{I}_t}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s^{\check{I}_t}) ds + g(\bar{X}_T^{t,x,\pi, \check{I}_t}, \mathbb{P}_{\bar{X}_T^{t,\xi, \check{I}_t}}^{\bar{\mathcal{F}}_T^\mu}) \right) \middle| \hat{\mathcal{F}}_t^\mu \right] \\
&= \mathbb{E} \left[ \kappa_T^{\nu^{\check{\omega}^1}} \left( \int_t^T f(s, \bar{X}_s^{t,x,\pi, \check{I}_t(\check{\omega})}, \mathbb{P}_{\bar{X}_s^{t,\xi, \check{I}_t(\check{\omega})}}^{\bar{\mathcal{F}}_s^\mu}, \bar{I}_s^{\check{I}_t(\check{\omega})}) ds + g(\bar{X}_T^{t,x,\pi, \check{I}_t(\check{\omega})}, \mathbb{P}_{\bar{X}_T^{t,\xi, \check{I}_t(\check{\omega})}}^{\bar{\mathcal{F}}_T^\mu}) \right) \right] \\
&= J^{\mathcal{R}}(t, x, \pi, \check{I}_t(\check{\omega}), \nu^{\check{\omega}^1}) \leq \sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu),
\end{aligned}$$

$\hat{\mathbb{P}}$ -a.s., for any  $a_0 \in A$  (recall from Theorem D.1 that  $\sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu)$  does not depend on  $a_0 \in A$ ). Then, since by Theorem D.1 we have that  $\sup_{\nu \in \mathcal{V}_t} J^{\mathcal{R}}(t, x, \pi, a_0, \nu) = V(t, x, \pi)$ , we deduce that (D.11) holds, which concludes the proof.  $\square$

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