#### **ORIGINAL ARTICLE**



# **Risk-sensitive discounted cost criterion for continuous-time Markov decision processes on a general state space**

**Subrata Golui<sup>1</sup> · Chandan Pal[1](http://orcid.org/0000-0002-4684-0481)**

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### **Abstract**

In this paper, we consider risk-sensitive discounted control problem for continuoustime jump Markov processes taking values in general state space. The transition rates of underlying continuous-time jump Markov processes and the cost rates are allowed to be unbounded. Under certain Lyapunov condition, we establish the existence and uniqueness of the solution to the Hamilton–Jacobi–Bellman equation. Also, we prove the existence of optimal risk-sensitive control in the class of Markov control and completely characterized the optimal control.

**Keywords** Continuous-time Markov decision process · History-dependent control · General state space · Risk-sensitive discounted criterion · HJB equation · Optimal control

## **1 Introduction**

In this paper, we study the risk-sensitive discounted criterion for continuous-time Markov decision processes (CTMDPs) with Borel state space. In the risk-neutral criterion, the controller wants to optimize the expected value of the total payoff. But in the risk-sensitive criterion, controller considers the expected value of the exponential of the total payoff and so, the risk-sensitive criterion gives a better protection from the risk. Therefore, the risk-sensitive or exponential of integral is a very popular cost criterion due to its applications in many areas such as queueing systems and finance, for more details see Bauerle and Riede[r](#page-28-0) [\(2014\)](#page-28-0) and Whittl[e](#page-28-1) [\(1990](#page-28-1)) and the references therein. In the literature, risk-sensitive control problems for CTMDPs are an important

B Chandan Pal cpal@iitg.ac.in Subrata Golui golui@iitg.ac.in

<sup>&</sup>lt;sup>1</sup> Department of Mathematics, Indian Institute of Technology Guwahati, Guwahati, Assam, India

class of stochastic optimal control problems and have been widely studied under different sets of conditions. Finite horizon risk-sensitive CTMDPs for countable state space were studied in Ghosh and Sah[a](#page-28-2) [\(2014\)](#page-28-2), Guo et al[.](#page-28-3) [\(2019](#page-28-3)) and We[i](#page-28-4) [\(2016\)](#page-28-4) and for infinite horizon risk-sensitive CTMDPs we refer to Ghosh and Sah[a](#page-28-2) [\(2014](#page-28-2)), Guo and Zhan[g](#page-28-5) [\(2020](#page-28-5)), Kumar and Pa[l](#page-28-6) [\(2013,](#page-28-6) [2015\)](#page-28-7), Pal and Pradha[n](#page-28-8) [\(2019](#page-28-8)) and Zhan[g](#page-28-9) [\(2017\)](#page-28-9). For important contributions to the risk-sensitive control of discrete-time MDP on a general state space, see Masi and Stettne[r](#page-28-10) [\(2000\)](#page-28-10) and Masi and Stettne[r](#page-28-11) [\(2007](#page-28-11)). Although risk-sensitive control of CTMDPs on a countable state space have been studied extensively, but the corresponding literature in the context of risk-sensitive control of CTMDPs on a general state space is rather limited. Some exceptions are (Guo and Zhan[g](#page-28-12) [2019](#page-28-12), [2020;](#page-28-5) Pal and Pradha[n](#page-28-8) [2019\)](#page-28-8).

In the paper (Pal and Pradha[n](#page-28-8) [2019](#page-28-8)), the authors studied risk-sensitive control of pure jump processes on a general state space. They considered bounded transition and cost rates and all controls are Markovian. In Pal and Pradha[n](#page-28-8) [\(2019\)](#page-28-8), the authors proved the HJB characterization of the optimal risk-sensitive control. The boundedness assumption on the transition and cost rates plays a key role in the proof of the existence of the optimal risk-sensitive control in Pal and Pradha[n](#page-28-8) [\(2019\)](#page-28-8). This boundedness requirement, however, imposes some restrictions in applications, for instance in queueing control and population processes, where the transition and reward/cost rates are usually unbounded. Also, there are many real-life situations where the state space may be uncountable, for example, the chemical reaction model, Gaussian model, etc. One can see Guo and Zhan[g](#page-28-12) [\(2019](#page-28-12)), Piunovskiy and Zhan[g](#page-28-13) [\(2020](#page-28-13)) and references therein for the real-life examples. In Guo and Zhan[g](#page-28-12) [\(2019\)](#page-28-12), the author considered the finite-horizon risk-sensitive control problem for CTMDPs on a Borel state space with unbounded transition and cost rates and proved the existence of optimal control via the HJB equation.

In this paper, we study a much more general risk-sensitive control problem for CTMDP with general state space. To the best of our knowledge, this is the first work which deals with infinite horizon discounted risk-sensitive control for CTMDPs on a general state space with unbounded cost and transition rates and the controls can be history-dependent. The main objective of this work is to prove the existence of the solution of the HJB equation and the characterization of the optimal risk-sensitive control. In particular (1) We prove that the HJB equations has a unique solution  $\varphi_{\alpha} \in L^{\infty}_{\mathcal{V}}([0, 1] \times S)$  satisfying the bounds as in Eq. [\(3.1\)](#page-6-0) below, where  $L^{\infty}_{\mathcal{V}}([0, 1] \times S)$ is described below. (2) We prove that any measurable minimizer of the HJB equation is optimal, and conversely any optimal control in the class of Markov controls is a minimizer of the HJB Eq.  $(3.1)$  below. We first consider for bounded transition and cost rates and establish the existence of a solution to the corresponding HJB equation by Banach's fixed point theorem as in Pal and Pradha[n](#page-28-8) [\(2019](#page-28-8)). Then we will relax the bounded hypothesis and extend this result to unbounded transition and cost rates. We characterize the value function via the HJB equation. Also, we prove the existence of an optimal control in the class of Markov control and the HJB characterization of the optimal risk-sensitive control and prove its complete characterization. In Corollary [5.1,](#page-25-0) we prove that if the cost and transition rates are bounded, then an optimal control exists for our model.

The rest of this article is structured as follows. Section [2](#page-2-0) deals with the description of the problem, required notations, some Assumptions, and preliminary results. In Sect. [3,](#page-6-1) we give a continuity-compactness Assumption and prove the stochastic representation of the solution of the HJB Eq.  $(3.1)$ . In Sect. [4,](#page-11-0) we truncate our transition and cost rates and prove the existence of the unique solution to the HJB equation. A complete characterization of optimal control is proven in Sect. [5.](#page-23-0) In Sect. [6,](#page-26-0) we illustrate our theory and assumptions by an illustrative example.

#### <span id="page-2-0"></span>**2 The control problems**

The model of CTMDP is a five-tuple which consists of the following elements:

$$
\mathbb{M} := \{ S, A, (A(x) \subset A, x \in S), c(x, a), q(\cdot | x, a) \},
$$

- a Borel space *S*, called the state space, whose elements are referred to as states of the system and the corresponding Borel  $\sigma$ -algebra is  $\mathcal{B}(S)$ . (Throughout the whole paper we consider that for any Borel space *X*, the corresponding Borel  $\sigma$ -algebra is  $\mathcal{B}(X)$ .)
- *A* is the action set, which is assumed to be Borel space with the Borel  $\sigma$ -algebra *B*(*A*).
- for each  $x \in S$ ,  $A(x) \in B(A)$  denotes the set of admissible actions for state *x*. Let  $K := \{(x, a) | x \in S, a \in A(x)\}$ , which is a Borel subset of  $S \times A$ .
- the measurable function  $c: K \to \mathbb{R}_+$  denotes the cost rate function. We require the cost function  $c(x, a)$  to measure (or evaluate) the utility of taking action *a* at state *x*.
- given any  $(x, a) \in K$ , the transition rate  $q(\cdot|x, a)$  is a Borel measurable signed kernel on *S* given *K*. That is,  $q(\cdot|x, a)$  satisfies countable additivity;  $q(D|x, a) \ge 0$ where  $(x, a) \in K$ ,  $x \notin D$  and  $D \in B(S)$ . Moreover, we assume that  $q(\cdot|x, a)$ satisfies the following conservative and stable conditions: for any  $x \in S$ ,

$$
q(S|x, a) \equiv 0 \text{ and}
$$
  

$$
q^*(x) := \sup_{a \in A(x)} q_x(a) < \infty,
$$

where  $q_x(a) := -q(\{x\}|x, a) \geq 0$ . We need the transition rates to specify the random dynamic evolution of the system.

Next, we give an informal description of the evolution of the CTMDPs as follows. The controller observes continuously the current state of the system. When the system is in state  $x \in S$  at time  $t \geq 0$ , he/she chooses action  $a_t \in A(x)$  according to some control. As a consequence of this, the following happens:

- the controller incurs an immediate cost at rate  $c(x, a_t)$ ; and
- after a random sojourn time (i.e., the holding time at state  $x$ ), the system jumps to a set  $B \in \mathcal{B}(S)$  ( $x \notin B$ ) of states with the transition probability  $\frac{q(B|x, a_t)}{q(B|x, a_t)}$

 $q_x(a_t)$ 

determined by the transition rates  $q(dy|x, a_t)$ . The distribution function of the sojourn time is  $(1-e^{-\int_t^{t+x} q_x(a_s)ds})$ . (see Proposition B.8 in [ Guo and Hernandez-Lerm[a](#page-28-14) [\(2009](#page-28-14)), p. 205] for details).

When the state of the system transits to a new state  $y \neq x$ , the above procedure is repeated. Thus, the controller tries to minimize his/her costs with respect to some performance criterion  $\mathcal{J}_{\alpha}(\cdot, \cdot, \cdot)$ , which in our present case is defined by [\(2.2\)](#page-4-0), below. To formalize what is described above, below we describe the construction of continuous time Markov decision processes (CTMDPs) under possibly history-dependent controls. To construct the underlying CTMDPs (as in Guo and Piunovski[y](#page-28-15) [2011](#page-28-15); Kitae[v](#page-28-16) [1995](#page-28-16); Piunovskiy and Zhan[g](#page-28-17) [2011](#page-28-17), [2020](#page-28-13)) we introduce some notations: let  $S_{\Delta}$  := *S* ∪ { $\Delta$ } (with some "isolated" state  $\Delta \notin S$ ),  $\Omega_0 := (S \times (0, \infty))^{\infty}$ ,  $\Omega_k := (S \times (0, \infty))^k \times S \times (\{\infty\} \times \{\Delta\})^{\infty}$  for  $k \ge 1$  and  $\Omega := \cup_{k=0}^{\infty} \Omega_k$ . Let *F* be the Borel  $\sigma$ -algebra on  $\Omega$ . Then we obtain the measurable space  $(\Omega, \mathcal{F})$ . For some  $k \ge 1$ , and sample  $\omega := (x_0, \theta_1, x_1, \dots, \theta_k, x_k, \dots) \in \Omega$ , define

$$
T_0(\omega) := 0, \ T_k(\omega) := T_{k-1}(\omega) + \theta_k, \ T_{\infty}(\omega) := \lim_{k \to \infty} T_k(\omega).
$$

Using  ${T_k}$ , we define the state process  ${\xi_t}_{t\geq0}$  as

<span id="page-3-0"></span>
$$
\xi_t(\omega) := \sum_{k \ge 0} I_{\{T_k \le t < T_{k+1}\}} x_k + I_{\{t \ge T_\infty\}} \Delta, \text{ for } t \ge 0 \text{ (with } T_0 := 0). \tag{2.1}
$$

Here,  $I_E$  denotes the indicator function of a set  $E$ , and we use the convention that  $0+z =: z$  and  $0z =: 0$  for all  $z \in S_\Delta$ . Obviously,  $\xi_t(\omega)$  is right-continuous on  $[0, \infty)$ . We denote  $\xi_{t-}(\omega) := \liminf_{s \to t^-} \xi_s(\omega)$ . From Eq. [\(2.1\)](#page-3-0), we see that  $T_k(\omega)$  ( $k \ge 1$ ) denotes the *k*-th jump moment of  $\{\xi_t, t \geq 0\}$ ,  $x_{k-1}$  is the state of the process on  $[T_{k-1}(\omega), T_k(\omega)), \theta_k = T_k(\omega) - T_{k-1}(\omega)$  plays the role of sojourn time at state  $x_{k-1}$ , and the sample path  $\{\xi_t(\omega), t \geq 0\}$  has at most denumerable states  $x_k (k = 0, 1, \dots)$ . The process after  $T_{\infty}$  is regarded to be absorbed in the state  $\Delta$ . Thus, let  $q(\cdot|\Delta, a_{\Delta}) :=$  $0, A_{\Delta} := A \cup \{a_{\Delta}\}, A(\Delta) := \{a_{\Delta}\}, c(\Delta, a) := 0$  for all  $a \in A_{\Delta}$ , where  $a_{\Delta}$  is an isolated point.

To precisely define the criterion, we need to introduce the concept of a control as in Guo et al[.](#page-28-18) [\(2012](#page-28-18)), Guo and Piunovski[y](#page-28-15) [\(2011\)](#page-28-15) and Kitaev and Ryko[v](#page-28-19) [\(1995\)](#page-28-19). Take the rightcontinuous  $\sigma$ -algebras  $\{\mathcal{F}_t\}_{t>0}$  with  $\mathcal{F}_t := \sigma(\{T_k \leq s, \xi_{T_k} \in S\} : 0 \leq s \leq t, k \geq 0).$ For all  $t \geq 0$ ,  $\mathcal{F}_{s-} =: \bigvee_{0 \leq t \leq s} \mathcal{F}_t$ . Now define a  $\sigma$ -algebra  $\mathcal{P} := \sigma(A \times \{0\}, B \times$  $(s, \infty)$ :  $A \in \mathcal{F}_0, B \in \mathcal{F}_{s-}$ , which denotes the  $\sigma$ -algebra of predictable sets on  $\Omega \times [0, \infty)$  related to  $\{\mathcal{F}_t\}_{t \geq 0}$ . To complete the specification of a stochastic optimal control problem, we need, of course, to introduce an optimality criterion. This requires to define the class of controls as below.

**Definition 2.1** A history-dependent policy  $\pi := {\pi_t(\omega)}_{t>0}$  is a measurable map from  $(\Omega \times [0, \infty), \mathcal{P})$  onto  $(A_{\Delta}, \mathcal{B}(A_{\Delta}))$  satisfying  $\pi_t(\omega) \in A(\xi_t(\omega))$  for all  $\omega \in \Omega$  and  $t \geq 0$ . For notational simplicity, we denote a history-dependent control as  $\{\pi_t\}_{t>0}$ . The set of all history-dependent controls is denoted by  $\Pi$ . A control  $\pi \in \Pi$ , is called a Markov if  $\pi_t(\omega) = \pi_t(\xi_{t-1}(w))$  for every  $w \in \Omega$  and  $t \geq 0$ , where  $\xi_{t-1}(w) :=$  $\lim_{s \uparrow t} \xi_s(w)$ . We denote by  $\Pi^m$  the family of all Markov controls.

For any compact metric space *Y* , let *P*(*Y* ) denote the space of probability measures on *Y* with Prohorov topology. Under Assumption [2.1](#page-4-1) below, for any initial state  $x \in S$ and any control  $\pi \in \Pi$ , Theorem 4.27 in Kitae[v](#page-28-19) and Rykov [\(1995](#page-28-19)) yields the existence of a unique probability measure denoted by  $P_{x}^{\pi}$  on  $(\Omega, \mathcal{F})$ . Let  $E_{x}^{\pi}$  be the expectation operator with respect to  $P_{x}^{\pi}$ . Fix any discounted factor  $\alpha > 0$ . For any  $\pi \in \Pi$  and  $x \in S$ , the risk-sensitive discounted criterion is defined as

$$
\mathcal{J}_{\alpha}(\theta, x, \pi) := \frac{1}{\theta} \log \bigg\{ E_x^{\pi} \bigg[ \exp \bigg( \theta \int_0^{\infty} e^{-\alpha t} c(\xi_t, \pi_t) dt \bigg) \bigg] \bigg\}, \tag{2.2}
$$

provided that the integral is well defined, where  $\{\xi_t\}_{t\geq 0}$  is the Markov process corresponding to  $\pi = {\pi_t}_{t\geq 0} \in \Pi$  and  $\theta \in (0, 1]$  denotes a risk-sensitive parameter and the limiting case of  $\theta \rightarrow 0$  is the risk-neutral case. For each  $x \in S$ , let

<span id="page-4-0"></span>
$$
\mathcal{J}_{\alpha}^{*}(\theta, x) = \inf_{\pi \in \Pi} \mathcal{J}_{\alpha}(\theta, x, \pi).
$$

A control  $\pi^* \in \Pi$  is said to be optimal if  $\mathcal{J}_\alpha(\theta, x, \pi^*) = \mathcal{J}_\alpha^*(\theta, x)$  for all  $x \in S$ . The objective of this paper is to provide conditions for the existence of optimal control and introduce a HJB characterization of such control.

Since the logarithm is an increasing function, instead of studying  $\mathcal{J}_{\alpha}(\theta, x, \pi)$ , we will consider  $J_\alpha(\theta, x, \pi)$  on  $[0, 1] \times S \times \Pi$  defined by

<span id="page-4-2"></span>
$$
\tilde{J}_{\alpha}(\theta, x, \pi) := E_x^{\pi} \bigg[ \exp \bigg( \theta \int_0^{\infty} e^{-\alpha t} c(\xi_t, \pi_t) dt \bigg) \bigg]. \tag{2.3}
$$

Obviously,  $J_\alpha(\theta, x, \pi) \ge 1$  for  $(\theta, x) \in [0, 1] \times S$  and  $\pi \in \Pi$ , and we have  $\pi^*$  is optimal if and only if  $\inf_{\pi \in \Pi} J_{\alpha}(\theta, x, \pi) = J_{\alpha}(\theta, x, \pi^*) =: J_{\alpha}^*(\theta, x) \forall x \in S$ . Since the rates  $q(dy|x, a)$  and costs  $c(x, a)$  are allowed to be unbounded, we next give conditions for the non-explosion of  $\{\xi_t, t \geq 0\}$  and finiteness of  $\mathcal{J}_\alpha(\theta, x, \pi)$ , which had been widely used in CTMDPs; see, for instance, (Guo and Hernandez-Lerm[a](#page-28-14) [2009;](#page-28-14) Guo et al. [2012](#page-28-18); Guo and Lia[o](#page-28-20) [2019;](#page-28-20) Guo and Piunovski[y](#page-28-15) [2011;](#page-28-15) Prieto-Rumeau and Hernandez-Lerm[a](#page-28-21) [2012](#page-28-21)) and references therein.

<span id="page-4-1"></span>**Assumption 2.1** There exists a real-valued Borel measurable function  $V > 1$  on *S* and constants  $\rho_0 > 0$ ,  $M_0 > 0$ ,  $L_0 \ge 0$  and  $0 < \rho_1 < \min{\{\alpha, \rho_0^{-1} \alpha^2\}}$  such that

(i)  $\int_S V(y)q(dy|x, a) \leq \rho_0 V(x) \quad \forall (x, a) \in K;$ 

(ii)  $\sup_{a \in A(x)} q_x(a) \leq M_0 \mathcal{V}(x) \quad \forall x \in S;$ 

(iii)  $\sup_{a \in A(x)} c(x, a) \le \rho_1 \log V(x) + L_0 \quad \forall x \in S.$ 

- *Remark 2.1* (a) Note that, when the transition rates are bounded i.e., sup<sub>*x*∈*S*</sub>  $q^*(x)$  <  $\infty$ , Assumptions [2.1](#page-4-1) (i) and (ii) are satisfied by taking a suitable constant value of  $V(x)$ .
- <span id="page-4-3"></span>(b) Under Assumption [2.1](#page-4-1) (iii) the criterion [\(2.3\)](#page-4-2) is well defined and finite; see Proposition  $2.1(c)$  $2.1(c)$  below.

**Proposition [2.1](#page-4-1)** *Grant Assumption 2.1. Then for any control*  $\pi \in \Pi$  *and*  $(\theta, x) \in$  $[0, 1] \times S$ , the following results are true:

- (a)  $P_x^{\pi}(T_{\infty} = \infty) = 1$ ,  $P_x^{\pi}(\xi_0 = x) = 1$ , and  $P_x^{\pi}(\xi_t \in S) = 1$  for all  $t \ge 0$ ;
- (b)  $E_x^{\pi}[V(\xi_t)] \leq e^{\rho_0 t} V(x)$  *for all t*  $\geq 0$ ;
- (c) *We have*

$$
\tilde{J}_{\alpha}(\theta, x, \pi) \leq \frac{\alpha^2}{\alpha^2 - \rho_0 \rho_1 \theta} e^{\theta L_0/\alpha} [\mathcal{V}(x)]^{\frac{\rho_1 \theta}{\alpha}} \leq \frac{\alpha^2}{\alpha^2 - \rho_0 \rho_1} e^{L_0/\alpha} \mathcal{V}(x).
$$

*Also, we get*

$$
\mathcal{J}_{\alpha}^*(\theta, x) \le \log\left(\frac{\alpha^2}{\alpha^2 - \rho_0 \rho_1}\right) + \frac{L_0}{\alpha} + \frac{\rho_1}{\alpha} \log \mathcal{V}(x) \ \forall \theta \in (0, 1], x \in S. \tag{2.4}
$$

*Proof* For parts(*a*) and (*b*), see, Guo et al[.](#page-28-18) [\(2012\)](#page-28-18) and Guo and Piunovskiy [\(2011,](#page-28-15)Theorem 3.1).

Proof of part (c): Observe that  $d(-e^{-\alpha t})$  is a probability measure on [0, ∞). For any  $\pi \in \Pi$  and  $(\theta, x) \in [0, 1] \times S$ , by [\(2.3\)](#page-4-2) and Jensen's inequality we have

<span id="page-5-0"></span>
$$
\tilde{J}_{\alpha}(\theta, x, \pi) = E_{x}^{\pi} \left[ \exp \left( \int_{0}^{\infty} \frac{\theta}{\alpha} c(\xi_{t}, \pi_{t}) d(-e^{-\alpha t}) \right) \right]
$$
\n
$$
\leq E_{x}^{\pi} \left[ \int_{0}^{\infty} \exp \left( \frac{\theta}{\alpha} c(\xi_{t}, \pi_{t}) \right) d(-e^{-\alpha t}) \right].
$$

By Assumption [2.1](#page-4-1) and part (b) we obtain

$$
\tilde{J}_{\alpha}(\theta, x, \pi) \leq E_{x}^{\pi} \left[ \int_{0}^{\infty} \exp\left(\frac{\theta}{\alpha} (\rho_{1} \log \mathcal{V}(\xi_{t}) + L_{0})\right) d(-e^{-\alpha t})\right]
$$
\n
$$
= e^{\theta L_{0}/\alpha} \left[ \int_{0}^{\infty} E_{x}^{\pi} \left(\mathcal{V}(\xi_{t})^{\frac{\rho_{1}\theta}{\alpha}}\right) d(-e^{-\alpha t})\right]
$$
\n
$$
\leq e^{\theta L_{0}/\alpha} \left[ \int_{0}^{\infty} (E_{x}^{\pi} [\mathcal{V}(\xi_{t})])^{\frac{\rho_{1}\theta}{\alpha}} d(-e^{-\alpha t})\right] \text{ (since } \rho_{1}\theta < \alpha)
$$
\n
$$
\leq \alpha e^{\theta L_{0}/\alpha} [\mathcal{V}(x)]^{\frac{\rho_{1}\theta}{\alpha}} \left[ \int_{0}^{\infty} \exp\left(\frac{\rho_{0}\rho_{1}\theta t}{\alpha} - \alpha t\right) dt\right]
$$
\n
$$
= \frac{\alpha^{2}}{\alpha^{2} - \rho_{0}\rho_{1}\theta} e^{\theta L_{0}/\alpha} [\mathcal{V}(x)]^{\frac{\rho_{1}\theta}{\alpha}},
$$

where the last equality holds due to the fact that  $\rho_0 \rho_1 \theta < \alpha^2$ . Next observe that sup  $\theta \in [0,1]$  $J^*_{\alpha}(\theta, x) \leq$  $\alpha^2$  $\frac{\alpha}{\alpha^2 - \rho_0 \rho_1} e^{L_0/\alpha} \mathcal{V}(x)$ , and sup  $\sup_{\theta \in (0,1]} \mathcal{J}_{\alpha}^{*}(\theta, x) = \sup_{\theta \in (0,1]}$  $\frac{1}{\theta} \log \tilde{J}_{\alpha}^{*}(\theta, x) \leq \sup_{\theta \in (0,1]}$ 1 θ  $\left(\log \frac{\alpha^2}{2}\right)$  $\alpha^2 - \rho_0 \rho_1 \theta$  $\lambda$ +  $\frac{L_0}{\alpha} + \frac{\rho_1}{\alpha} \log \mathcal{V}(x)$ .

Also, doing a simple and direct calculation, we achieve  $(2.4)$ .

In Ghosh and Sah[a](#page-28-2) [\(2014](#page-28-2)) and Kumar and Pa[l](#page-28-6) [\(2013\)](#page-28-6), the authors used the Dynkin's formula within the class of Markov controls by using the Markov property of the state process  $\{\xi_t\}_{t>0}$ . But this Markov property may fail to hold when we study within the class of history-dependent controls, and consequently, here we can't directly apply the Dynkin formula. Hence we assume the following condition, so that we can apply the Dynkin's formula for a large enough class of functions, which had been widely used in CTMDPs; see, for instance, (Guo and Lia[o](#page-28-20) [2019](#page-28-20); Guo et al[.](#page-28-3) [2019;](#page-28-3) Guo and Zhan[g](#page-28-12) [2019\)](#page-28-12).

<span id="page-6-2"></span>**Assumption 2.2** The Borel measurable function  $V^2 \ge 1$  on *S* satisfies the following Lyapunov condition

$$
\int_{S} q(dy|x, a)\mathcal{V}^{2}(y) \leq \rho_2 \mathcal{V}^{2}(x) + b_0 \ \forall (x, a) \in K,
$$

for some constants  $0 < \rho_2 < \alpha$  and  $b_0 \ge 0$ . Here *V* is as in Assumption [2.1.](#page-4-1)

We now introduce some frequently used notations.

- $C_c^{\infty}(a, b)$  denotes the set of all infinitely differentiable functions on  $(a, b)$  with compact support.
- Let  $A_{as}([0, 1] \times S)$  denote the space of all functions which are real-valued and differentiable almost everywhere with respect to the first variable  $\theta \in [0, 1]$ . Given any real-valued function  $W > 1$  on *S* and any Borel set *X*, a real-valued function  $\varphi$  on *X* × *S* is called *W* bounded if  $\|\varphi\|_W^{\infty}$  :=  $W := \sup_{(\theta, x) \in X \times S} \frac{W(x)}{W(x)}$  $|\varphi(\theta, x)|$  $< \infty$ . Denote  $B_W(X \times S)$  the Banach space of all *W*-bounded functions. When  $W \equiv 1$ ,  $B_1([0, 1] \times S)$  is the space of all bounded functions on  $[0, 1] \times S$ . Now define  $L_W^{\infty}([0, 1] \times S) := \{ \varphi : [0, 1] \times S \to \mathbb{R} : \varphi \in B_W([0, 1] \times S) \cap A_{as}([0, 1] \times S) \}.$

#### <span id="page-6-1"></span>**3 Stochastic representation of a solution to the HJB equation**

In this section, we prove that if the HJB equation for the cost criterion [\(2.3\)](#page-4-2) has a solution then we will give a stochastic representation of that solution. Using dynamic programming heuristics, the HJB equations for the discounted cost criterion [\(2.3\)](#page-4-2) is given by

<span id="page-6-0"></span>
$$
\begin{cases}\n\alpha \theta \frac{\partial \varphi_{\alpha}}{\partial \theta}(\theta, x) = \inf_{a \in A(x)} \left[ \int_{S} q(dy|x, a) \varphi_{\alpha}(\theta, y) + \theta c(x, a) \varphi_{\alpha}(\theta, x) \right], \\
1 \leq \varphi_{\alpha}(\theta, x) \leq \frac{\alpha^{2}}{\alpha^{2} - \rho_{0} \rho_{1} \theta} e^{\theta L_{0}/\alpha} (\mathcal{V}(x)) \frac{\rho_{1} \theta}{\alpha} \text{ for } (\theta, x) \in [0, 1] \times S,\n\end{cases}
$$
\n(3.1)

for each  $x \in S$  and a.e.  $\theta \in [0, 1]$  where the upper bound of  $\varphi_{\alpha}(\theta, x)$  is motivated by Proposition [2.1.](#page-4-3)

*Remark 3.1* To prove the existence of an optimal control for bounded cost and transition rates, in Pal and Pradha[n](#page-28-8) [\(2019\)](#page-28-8), the authors studied the following HJB equation having a solution  $\phi_{\alpha}(\theta, x)$  on [0, 1]  $\times$  *S* such that

<span id="page-7-0"></span>
$$
\begin{cases}\n\alpha \theta \frac{\partial \phi_{\alpha}}{\partial \theta}(\theta, x) = \inf_{a \in A(x)} \left[ \int_{S} q(dy|x, a) \phi_{\alpha}(\theta, y) + \theta c(x, a) \phi_{\alpha}(\theta, x) \right], & \text{for } (\theta, x) \in [0, 1] \times S, \\
\lim_{\theta \to 0} \phi_{\alpha}(\theta, x) = 1 \text{ uniformly in } x \in S.\n\end{cases}
$$
\n(3.2)

From the arguments for the existence of a unique solution to the Eq.  $(3.2)$ , it is necessary to have  $\phi_{\alpha}(\theta, x)$  converges to 1 uniformly in *x* as  $\theta \to 0$ . But, it is not true in general when the cost and transition rates are unbounded; for more details see Example 3.2 in Guo and Lia[o](#page-28-20) [\(2019](#page-28-20)). In this article we replace the uniform convergence condition by the above new one.

<span id="page-7-1"></span>To ensure the existence of an optimal control, in addition to Assumptions [2.1](#page-4-1) and [2.2,](#page-6-2) we also need the following continuity and compactness conditions.

**Assumption 3.1** The following conditions hold:

- (i) for each  $x \in S$ , the set  $A(x)$  is compact;
- (ii) for any fixed  $x \in S$ , the function  $c(x, a)$  is continuous in  $a \in A(x)$ ;
- (iii) for any given  $x \in S$ , the function  $\int_S V(y)q(dy|x, a)$  is continuous in  $a \in A(x)$ , where  $V$  is introduced in Assumption [2.1.](#page-4-1)

*Remark 3.2* Assumptions [3.1](#page-7-1) (i)–(iii) are commonly used to find an optimal control for continuous-time MDP, see Guo and Hernandez-Lerm[a](#page-28-14) [\(2009](#page-28-14)), Guo and Lia[o](#page-28-20) [\(2019](#page-28-20)), Guo et al[.](#page-28-3) [\(2019](#page-28-3)), Guo and Piunovski[y](#page-28-15) [\(2011](#page-28-15)) and Guo and Zhan[g](#page-28-12) [\(2019](#page-28-12)). Also, note that if Assumption [3.1](#page-7-1) (iii) is satisfied, then for any given  $x \in S$ , the function  $\overline{1}$ *S* 

<span id="page-7-2"></span>In the next theorem we show that if the HJB equation has a solution then its stochastic representation is equal to the value function corresponding to the cost criterion [\(2.3\)](#page-4-2).

**Theorem 3.1** *Under Assumptions* [2.1](#page-4-1)*,* [2.2](#page-6-2)*, and* [3.1](#page-7-1) *suppose that the HJB Eq.*[\(3.1\)](#page-6-0) *has a solution*  $\varphi_{\alpha} \in L^{\infty}_{\mathcal{V}}([0, 1] \times S)$  *satisfying the bounds as in Eq.*[\(3.1\)](#page-6-0)*. Then, for all*<br> $(\alpha, \alpha) \in [0, 1] \cup S$  are known to analyzize any contribution of a set  $(\theta, x) \in [0, 1] \times S$ , we have the probabilistic representation of  $\varphi_{\alpha}$  as

<span id="page-7-3"></span>
$$
\varphi_{\alpha}(\theta, x) = \inf_{\pi \in \Pi} E_{x}^{\pi} \left[ \exp \left( \theta \int_{0}^{\infty} e^{-\alpha t} c(\xi_{t}, \pi_{t}) dt \right) \right]
$$
(3.3)

*i.e.,*  $\varphi_{\alpha}(\theta, x) = J_{\alpha}^{*}(\theta, x)$  *for all*  $(\theta, x) \in [0, 1] \times S$ .

*Proof* First, we see that

$$
\left[\theta c(x,a)\varphi_{\alpha}(\theta,x)+\int_{S}q(dy|x,a)\varphi_{\alpha}(\theta,y)\right]
$$

is continuous in  $a \in A(x)$  and  $A(x)$  is compact. So by measurable selection theorem, [Bertsekas and Shrev[e](#page-28-22) [\(1996\)](#page-28-22), Proposition 7.33], there exists a measurable function  $f^* : [0, 1] \times S \rightarrow A$  such that

$$
\inf_{a \in A(x)} \left[ \theta c(x, a) \varphi_{\alpha}(\theta, x) + \int_{S} q(dy|x, a) \varphi_{\alpha}(\theta, y) \right]
$$
  
= 
$$
\left[ \theta c(x, f^{*}(\theta, x)) \varphi_{\alpha}(\theta, x) + \int_{S} q(dy|x, f^{*}(\theta, x)) \varphi_{\alpha}(\theta, y) \right].
$$
 (3.4)

Let

<span id="page-8-2"></span> $\pi^*$ :  $S \times \mathbb{R}_+ \to A$ 

be defined by

<span id="page-8-1"></span><span id="page-8-0"></span>
$$
\pi_t^*(x) := f^*(\theta e^{-\alpha t}, x).
$$

Now we observe from Eq. [\(3.1\)](#page-6-0) that for any  $x \in S$ ,  $a \in A(x)$  and a.e.  $\theta \in [0, 1]$  that

$$
-\alpha \theta \frac{\partial \varphi_{\alpha}}{\partial \theta}(\theta, x) + \left[ \int_{S} q(dy|x, a) \varphi_{\alpha}(\theta, y) + \theta c(x, a) \varphi_{\alpha}(\theta, x) \right] \ge 0. \tag{3.5}
$$

For any history-dependent control  $\pi \in \Pi$  and  $\theta \in [0, 1]$ , let  $\{\xi_t, t \geq 0\}$  be the corresponding process, and define  $\theta(t) := \theta e^{-\alpha t}$ . Now for each  $\omega \in \Omega$ , by Eq. [\(3.5\)](#page-8-0), we get for a.e.  $s \geq 0$ ,

$$
-\alpha\theta(s)\frac{\partial\varphi_{\alpha}}{\partial\theta}(\theta(s),\xi_{s}) + \left[\int_{S} q(dy|\xi_{s},\pi_{s})\varphi_{\alpha}(\theta(s),y) + \theta(s)c(\xi_{s},\pi_{s})\varphi_{\alpha}(\theta(s),\xi_{s})\right] \geq 0.
$$
\n(3.6)

Define a function *g* :  $[0, \infty) \times S \times \Omega \rightarrow [0, \infty)$  by

$$
g(t, x, \omega) := \exp\left(\int_0^t \theta(s)c(\xi_s, \pi_s)ds\right) \varphi_\alpha(\theta(t), x).
$$

In view of Assumptions [2.1](#page-4-1) and [2.2,](#page-6-2) we have

$$
E_x^{\pi} \left[ \exp \left( \int_0^t 2e^{-\alpha s} c(\xi_s, \pi_s) ds \right) \right]
$$
  
\n
$$
\leq E_x^{\pi} \left[ \exp \left( \int_0^{\infty} \frac{2}{\alpha} c(\xi_s, \pi_s) d(-e^{-\alpha s}) \right) \right]
$$
  
\n
$$
\leq E_x^{\pi} \left[ \int_0^{\infty} \exp \left( \frac{2}{\alpha} c(\xi_s, \pi_s) \right) d(-e^{-\alpha s}) \right]
$$
  
\n
$$
\leq E_x^{\pi} \left[ \int_0^{\infty} \exp \left( \frac{2}{\alpha} (\rho_1 \log \mathcal{V}(\xi_s) + L_0) \right) d(-e^{-\alpha s}) \right]
$$

(by Assumptio[n2](#page-4-1).1)

<span id="page-9-2"></span>
$$
= e^{2L_0/\alpha} \left[ \int_0^\infty E_x^\pi \left( \mathcal{V}(\xi_s) \frac{2\rho_1}{\alpha} \right) d(-e^{-\alpha s}) \right]
$$
  
\n
$$
\leq \alpha e^{2L_0/\alpha} \left( \mathcal{V}^2(x) + \frac{b_0}{\rho_2} \right) \left[ \int_0^\infty e^{\rho_2 s - \alpha s} ds \right]
$$
  
\n
$$
= \frac{\alpha e^{2L_0/\alpha}}{\alpha - \rho_2} \left( \mathcal{V}^2(x) + \frac{b_0}{\rho_2} \right),
$$
\n(3.7)

where second inequality is obtained by using Jensen's inequality.

Hence  $E_x^{\pi}$  $\left[\exp\left(\int_0^t 2e^{-\alpha s}c(\xi_s, \pi_s)ds\right)\right] < \infty$  for all  $x \in S$  and  $t \in (0, \infty)$ . Thus, using the extension of Dynkin formula in Guo et al[.](#page-28-3) [\(2019\)](#page-28-3), Theorem[3.1](#page-7-2) to the function *g*, we have

$$
E_x^{\pi}[g(t, \xi_t, \omega)] - \varphi_{\alpha}(\theta, x)
$$
  
= 
$$
E_x^{\pi} \left\{ \int_0^t \exp \left( \int_0^s \theta(v) c(\xi_v, \pi_v) dv \right) \times \left[ -\alpha \theta(s) \frac{\partial \varphi_{\alpha}}{\partial \theta} (\theta(s), \xi_s) + \int_S q(dy|\xi_s, \pi_s) \varphi_{\alpha}(\theta(s), y) + \theta(s) c(\xi_s, \pi_s) \varphi_{\alpha}(\theta(s), \xi_s) \right] ds \right\}.
$$
 (3.8)

Now from  $(3.6)$  and  $(3.8)$ , we have

<span id="page-9-1"></span><span id="page-9-0"></span>
$$
\varphi_{\alpha}(\theta, x) \le E_{x}^{\pi} \Bigg[ \exp \Biggl( \int_{0}^{t} \theta(s) c(\xi_{s}, \pi_{s}) ds \Biggr) \varphi_{\alpha}(\theta(t), \xi_{t}) \Biggr]. \tag{3.9}
$$

Given any  $p > 1$ , let  $q > 1$  such that  $\frac{1}{p} + \frac{1}{q} = 1$ , by Holder's inequality we have

$$
\varphi_{\alpha}(\theta, x) \le E_{x}^{\pi} \left[ \exp \left( \int_{0}^{t} \theta(s) c(\xi_{s}, \pi_{s}) ds \right) \varphi_{\alpha}(\theta(t), \xi_{t}) \right]
$$
  
\n
$$
\le \left\{ E_{x}^{\pi} \left[ \exp \left( p \int_{0}^{t} \theta(s) c(\xi_{s}, \pi_{s}) ds \right) \right] \right\}^{1/p} \times \left\{ E_{x}^{\pi} [\varphi_{\alpha}^{q}(\theta(t), \xi_{t})] \right\}^{1/q}
$$
  
\n=:  $T_{1}(p, t) \cdot T_{2}(q, t).$  (3.10)

For  $T_2(q, t) := \{E_x^{\pi} [\varphi_{\alpha}^q(\theta(t), \xi_t)]\}^{1/q}$ , by the upper bound of  $\varphi_{\alpha}$  in [\(3.1\)](#page-6-0), we have

$$
\varphi_{\alpha}(\theta(t),\xi_t) = \varphi_{\alpha}(\theta e^{-\alpha t},\xi_t) \leq \frac{\alpha^2}{\alpha^2 - \theta e^{-\alpha t} \rho_0 \rho_1} \exp\left(\frac{\theta e^{-\alpha t} L_0}{\alpha}\right) [\mathcal{V}(\xi_t)]^{\frac{\rho_1 \theta e^{-\alpha t}}{\alpha}}.
$$

If  $t > \alpha^{-1} \log(\theta q \rho_1/\alpha)$  then  $\theta e^{-\alpha t} q \rho_1/\alpha < 1$ . Applying Jensen's inequality and Proposition [2.1\(](#page-4-3)b), we get

$$
T_2(q, t) \leq \left\{ E_x^{\pi} \left[ \left( \frac{\alpha^2}{\alpha^2 - \theta e^{-\alpha t} \rho_0 \rho_1} \right)^q \exp\left( \frac{q \theta e^{-\alpha t} L_0}{\alpha} \right) \left[ \mathcal{V}(\xi_t) \right]^{\frac{q \rho_1 \theta e^{-\alpha t}}{\alpha}} \right] \right\}^{1/q}
$$
  
\n
$$
= \frac{\alpha^2}{\alpha^2 - \theta e^{-\alpha t} \rho_0 \rho_1} \exp\left( \frac{\theta e^{-\alpha t} L_0}{\alpha} \right) \left[ E_x^{\pi} \left[ \mathcal{V}^{\frac{q \rho_1 \theta e^{-\alpha t}}{\alpha}}(\xi_t) \right] \right]^{\frac{1}{q}}
$$
  
\n
$$
\leq \frac{\alpha^2}{\alpha^2 - \theta e^{-\alpha t} \rho_0 \rho_1} \exp\left( \frac{\theta e^{-\alpha t} L_0}{\alpha} \right) \left[ E_x^{\pi} \left( \mathcal{V}(\xi_t) \right) \right]^{\frac{\rho_1 \theta e^{-\alpha t}}{\alpha}}
$$
  
\n
$$
\leq \frac{\alpha^2}{\alpha^2 - \theta e^{-\alpha t} \rho_0 \rho_1} \exp\left( \frac{\theta e^{-\alpha t}}{\alpha} (L_0 + \rho_0 \rho_1 t) \right) \mathcal{V}^{\frac{\theta e^{-\alpha t} \rho_1}{\alpha}}(x) =: T_3(t).
$$
\n(3.11)

Next take  $t \to \infty$  and get

$$
T_1(p,t) \to \left\{ E_x^{\pi} \bigg[ \exp \bigg( p \int_0^{\infty} \theta(s) c(\xi_s, \pi_s) ds \bigg) \bigg] \right\}^{1/p} \text{and } T_3(t) \to 1. \tag{3.12}
$$

By [\(3.10\)](#page-9-1), [\(3.11\)](#page-10-0) and [\(3.12\)](#page-10-1) we obtain

<span id="page-10-1"></span><span id="page-10-0"></span>
$$
\varphi_{\alpha}(\theta, x) \leq \left\{ E_x^{\pi} \left[ \exp \left( p\theta \int_0^{\infty} e^{-\alpha t} c(\xi_t, \pi_t) dt \right) \right] \right\}^{1/p}.
$$

Now, take the limit as  $p \downarrow 1$  and get the result

$$
\varphi_{\alpha}(\theta, x) \leq E_{x}^{\pi} \bigg[ \exp \bigg( \theta \int_{0}^{\infty} e^{-\alpha t} c(\xi_{t}, \pi_{t}) dt \bigg) \bigg].
$$

Since  $\pi \in \Pi$  is an arbitrary control, we have

$$
\varphi_{\alpha}(\theta, x) \le \inf_{\pi \in \Pi} E_{x}^{\pi} \bigg[ \exp \bigg( \theta \int_{0}^{\infty} e^{-\alpha t} c(\xi_{t}, \pi_{t}) dt \bigg) \bigg]. \tag{3.13}
$$

Using  $(3.1)$ ,  $(3.4)$  and  $(3.10)$ , we can show that

$$
E_x^{\pi^*} \bigg[ \exp \bigg( \int_0^t \theta(s) c(\xi_s, \pi_s^*(\xi_{s-})) ds \bigg) \varphi_\alpha(\theta(t), \xi_t) \bigg] = \varphi_\alpha(\theta, x). \tag{3.14}
$$

Now, using the lower bound of  $\varphi_{\alpha}$  in [\(3.1\)](#page-6-0) and Fatou's lemma, we obtain

$$
\liminf_{t \to \infty} E_x^{\pi^*} \Bigg[ \exp \Biggl( \int_0^t \theta(s) c(\xi_s, \pi_s^*(\xi_{s-})) ds \Biggr) \varphi_\alpha(\theta(t), \xi_t) \Biggr]
$$
\n
$$
\geq \liminf_{t \to \infty} E_x^{\pi^*} \Bigg[ \exp \Biggl( \int_0^t \theta(s) c(\xi_s, \pi_s^*(\xi_{s-})) ds \Biggr) \Biggr]
$$
\n
$$
\geq E_x^{\pi^*} \Bigg[ \liminf_{t \to \infty} \exp \Biggl( \int_0^t \theta(s) c(\xi_s, \pi_s^*(\xi_{s-})) ds \Biggr) \Biggr]
$$
\n
$$
= \tilde{J}_\alpha(\theta, x, \pi^*).
$$
\n(3.15)

<span id="page-10-4"></span><span id="page-10-3"></span><span id="page-10-2"></span><sup>2</sup> Springer

From  $(3.14)$  and  $(3.15)$ , we have

$$
\tilde{J}_{\alpha}(\theta, x, \pi^*) \leq \varphi_{\alpha}(\theta, x).
$$

Thus

$$
\inf_{\pi \in \Pi} \tilde{J}_{\alpha}(\theta, x, \pi) \le \tilde{J}_{\alpha}(\theta, x, \pi^*) \le \varphi_{\alpha}(\theta, x). \tag{3.16}
$$

From [\(3.13\)](#page-10-4) and [\(3.16\)](#page-11-1), we have [\(3.3\)](#page-7-3).

#### <span id="page-11-0"></span>**4 The existence of solution to the HJB equation**

In this Section, we prove that the Eq.  $(3.1)$  is the HJB equation for the  $\alpha$  discounted cost [\(2.3\)](#page-4-2) and the Eq. [\(3.1\)](#page-6-0) has a solution in  $L^{\infty}_V([0, 1] \times S)$ . We now proceed to make a rigorous analysis of the above. First, we prove a lemma about the existence of a solution for the HJB equation for bounded transition and cost rates; see Lemma [4.1](#page-11-2) below. Then in Theorem [4.1,](#page-14-0) we relax these boundedness condition and prove the existence of a solution to the HJB Eq.  $(3.1)$ . For that, we first truncate our transition and cost rates which plays a crucial role to derive the HJB equations and find the solution. Fix any  $n \ge 1, 0 < \delta < 1$ . For each  $n \ge 1, x \in S$ ,  $a \in A(x)$ , let  $A_n(x) := A(x)$ ,  $S_n := \{x \in S | \mathcal{V}(x) \leq n\}$ , and  $K_n := \{(x, a) | x \in S_n, a \in A_n(x)\}$ . Moreover for each  $x \in S$ ,  $a \in A_n(x)$  define

<span id="page-11-5"></span>
$$
q^{(n)}(dy|x,a) := \begin{cases} q(dy|x,a) & \text{if } x \in S_n, \\ 0 & \text{if } x \notin S_n \end{cases}
$$
 (4.1)

and

$$
c_n(x, a) := \begin{cases} c(x, a) \wedge \min\{n, \rho_1 \ln \mathcal{V}(x) + L_0\} & \text{if } x \in S_n, \\ 0 & \text{if } x \notin S_n. \end{cases}
$$
(4.2)

<span id="page-11-2"></span>**Lemma 4.1** *Grant Assumptions* [2.1](#page-4-1)*,* [2.2](#page-6-2) *and* [3.1](#page-7-1)*. Then, there exists a unique function*  $\varphi_\alpha^{(n,\delta)}$  (depending on n,  $\delta$ ) in  $L^\infty_\mathcal{V}([0,1]\times S)$  for which the followings are true :

(1)  $\varphi_{\alpha}^{(n,\delta)} \in B_1([0,1] \times S)$  *is a bounded solution to the following differential equations (DEs) for all*  $x \in S$  *and a.e.*  $\theta \in (\delta, 1]$ :

$$
\begin{cases}\n\alpha \theta \frac{\partial \varphi_{\alpha}^{(n,\delta)}}{\partial \theta}(\theta, x) = \inf_{a \in A(x)} \left[ \theta c_n(x, a) \varphi_{\alpha}^{(n,\delta)}(\theta, x) + \int_S q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n,\delta)}(\theta, y) \right] \\
\varphi_{\alpha}^{(n,\delta)}(\delta, x) = e^{n\delta/\alpha}.\n\end{cases}
$$
\n(4.3)

(2)  $\varphi_{\alpha}^{(n,\delta)}(\theta, x)$  has a stochastic representation as follows: for each  $x \in S$  and a.e.  $\theta \in (\delta, 1]$ ,

<span id="page-11-4"></span><span id="page-11-3"></span><span id="page-11-1"></span>

<span id="page-12-2"></span>
$$
\varphi_{\alpha}^{(n,\delta)}(\theta,x) = \inf_{\pi \in \Pi} E_x^{\pi} \bigg[ e^{n\delta/\alpha} \exp\bigg( \theta \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \bigg) \bigg], \qquad (4.4)
$$

*where*  $T_{\delta}(\theta) := \alpha^{-1} \log(\theta/\delta)$  *and*  $\{\xi_t^{(n)}\}_{t>0}$  *is the process corresponding to the*  $q^{(n)}(\cdot|x, a)$ .

*Proof* (1) Since  $S_n := \{x \in S | \mathcal{V}(x) \leq n\}$ , by Assumption [2.1\(](#page-4-1)ii), we see that  $q_x^{(n)}(a) := \int_{\mathcal{C}(a)} q^{(n)}(dy|x, a)$  is bounded. So we can use the Lyapunov function *S*/{*x*}  $V \equiv 1$  such that  $\int_{S} q^{(n)}(dy|x, a)V(y) \le \rho_0 V(x)$ , and  $\overline{q}^{(n)} := \sup_{(x, a) \in K} q_x^{(n)}(a)$  $\infty$ . Now let us define a nonlinear operator *T* on  $B_1([0, 1] \times S)$  as follows:

$$
Tu(\theta, x) = e^{\delta n/\alpha} + \frac{1}{\alpha} \int_{\delta}^{\theta} \inf_{a \in A(x)} \left[ \frac{1}{s} \int_{S} q^{(n)}(dy|x, a)u(s, y) + c_n(x, a)u(s, x) \right] ds,
$$

where  $u \in B_1([0, 1] \times S)$  and  $(\theta, x) \in [\delta, 1] \times S$ . By using the Assumption [2.1](#page-4-1) and the fact that  $c_n$  is bounded, we obtain

$$
\sup_{\theta \in [\delta, 1]} \sup_{x \in S} |Tu(\theta, x)|
$$
\n
$$
\leq e^{\delta n/\alpha} + \frac{1}{\alpha} \int_{\delta}^{1} \sup_{a \in A(x)} \left\{ \frac{1}{s} \sup_{x \in S} \left[ \int_{S} |q^{(n)}(dy|x, a)| |u(s, y)| \right] + n \sup_{x \in S} |u(s, x)| \right\} ds
$$
\n
$$
\leq e^{\delta n/\alpha} + \frac{\|u\|_{1}^{\infty}}{\alpha} \left\{ \int_{\delta}^{1} \sup_{a \in A(x)} \frac{1}{s} \sup_{x \in S} \left( 2q_{x}^{(n)}(a) \right) ds + n(1 - \delta) \right\}
$$
\n
$$
\leq e^{\delta n/\alpha} + \frac{1}{\alpha} \left[ (-2)\overline{q}^{(n)} \log \delta + n(1 - \delta) \right] \|u\|_{1}^{\infty}.
$$

Therefore, *T* is a nonlinear operator from  $B_1([0, 1] \times S)$  to  $B_1([0, 1] \times S)$ . For any  $g_1, g_2 \in B_1([0, 1] \times S)$  and  $\theta \in [\delta, 1]$ , we have

$$
\sup_{x \in S} |Tg_1(t, x) - Tg_2(t, x)| \le \frac{1}{\alpha} \int_{\delta}^{t} \left( 2\overline{q}^{(n)} / s + n \right) \sup_{x \in S} |g_1(s, x) - g_2(s, x)| ds
$$
  

$$
\le \frac{1}{\alpha} \left[ 2\overline{q}^{(n)} (\log t - \log \delta) + n(t - \delta) \right] \|g_1 - g_2\|_1^{\infty}.
$$
\n(4.5)

Now, we prove the following:

<span id="page-12-1"></span>
$$
\sup_{x \in S} |T^{l} g_1(t, x) - T^{l} g_2(t, x)|
$$
\n
$$
\leq \frac{\|g_1 - g_2\|_1^{\infty}}{\alpha^{l} \cdot l!} \left[ 2\overline{q}^{(n)} (\log t - \log \delta) + n(t - \delta) \right]^l \ \forall \ l \geq 1. \tag{4.6}
$$

<span id="page-12-0"></span> $\mathcal{D}$  Springer

By  $(4.5)$  and  $(4.6)$  we have

$$
\sup_{x \in S} |T^{l+1}g_1(t, x) - T^{l+1}g_2(t, x)|
$$
\n
$$
\leq \frac{1}{\alpha} \int_{\delta}^{t} \left(2\overline{q}^{(n)}/s + n\right) \sup_{x \in S} |T^l g_1(s, x) - T^l g_2(s, x)| ds
$$
\n
$$
\leq \frac{||g_1 - g_2||_1^{\infty}}{\alpha^{l+1} \cdot l!} \int_{\delta}^{t} \left(2\overline{q}^{(n)}/s + n\right) \left[2\overline{q}^{(n)}(\log s - \log \delta) + n(s - \delta)\right]^l ds
$$
\n
$$
= \frac{||g_1 - g_2||_1^{\infty}}{\alpha^{l+1} \cdot (l+1)!} \left[2\overline{q}^{(n)}(\log t - \log \delta) + n(t - \delta)\right]^{l+1}.
$$

Since  $\sum_{k\geq 1} \frac{1}{\alpha^k \cdot k!}$  $\left[-2\overline{q}^{(n)}\log\delta + n(1-\delta)\right]^{k}$  $< \infty$ , there exists some *m* such that  $\beta :=$  $rac{1}{\alpha^m \cdot m!}$  $\left[-2\overline{q}^{(n)}\log \delta + n(1-\delta)\right]^m < 1$ , which implies that  $\|T^m g_1 - T^m g_2\|_1^{\infty} \leq \beta \|g_1 - g_2\|_1^{\infty}$  $g_2\Vert_1^{\infty}$ . Therefore, T is a *m*-step contraction operator on  $B_1([0, 1] \times S)$ . So, by Banach fixed point theorem, there exists a unique bounded function  $\varphi_{\alpha}^{(n,\delta)} \in B_1([0, 1] \times S)$ (depending on  $(n, \delta)$ ) such that  $T \varphi_{\alpha}^{(n, \delta)} = \varphi_{\alpha}^{(n, \delta)}$ ; that is,

$$
\varphi_{\alpha}^{(n,\delta)}(\theta, x) = e^{\delta n/\alpha} + \frac{1}{\alpha} \int_{\delta}^{\theta} \inf_{a \in A(x)} \left[ \frac{1}{s} \int_{S} q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n,\delta)}(s, y) + c_n(x, a) \varphi_{\alpha}^{(n,\delta)}(s, x) \right] ds.
$$

Also note that  $\varphi_{\alpha}^{(n,\delta)}(\delta, x) = e^{\delta n/\alpha}$ . Hence by using [\(4.1\)](#page-11-3), [\(4.2\)](#page-11-4) and the above equation, we have  $\varphi_{\alpha}^{(n,\delta)} \in L^{\infty}_{\mathcal{V}}([0, 1] \times S)$  and it satisfies equation [\(4.3\)](#page-11-5). (2) First we see that

<span id="page-13-0"></span>
$$
\[ \theta c_n(x, a)\varphi_\alpha^{(n,\delta)}(\theta, x) + \int_S q^{(n)}(dy|x, a)\varphi_\alpha^{(n,\delta)}(\theta, y) \]
$$

is continuous in  $a \in A(x)$  and  $A(x)$  is compact. So by measurable selection theorem (Bertsekas and Shrev[e](#page-28-22) [1996](#page-28-22)), Proposition 7.33, there exists a measurable function  $f^{*\delta}: [0, 1] \times S \rightarrow A$  such that

$$
\inf_{a\in A(x)} \left[\theta c_n(x,a)\varphi_{\alpha}^{(n,\delta)}(\theta,x) + \int_S q^{(n)}(dy|x,a)\varphi_{\alpha}^{(n,\delta)}(\theta,y)\right]
$$
\n
$$
= \left[\theta c_n(x,f^{*\delta}(\theta,x))\varphi_{\alpha}^{(n,\delta)}(\theta,x) + \int_S q^{(n)}(dy|x,f^{*\delta}(\theta,x))\varphi_{\alpha}^{(n,\delta)}(\theta,y)\right].
$$
 (4.7)

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Let

$$
\pi^{*\delta}: S \times \mathbb{R}_+ \to A
$$

be defined by

<span id="page-14-1"></span>
$$
\pi_t^{*\delta}(x) := f^{*\delta}(\theta e^{-\alpha t}, x).
$$

Let  $\theta(t) := \theta e^{-\alpha t}$  for  $t \in [0, \infty)$ . Since  $c_n$  and  $\varphi_\alpha^{(n, \delta, k)}$  are bounded, by Dynkin's formula we get

$$
E_x^{\pi} \left[ \exp \left( \int_0^{T_\delta(\theta)} \theta(s) c_n(\xi_s^{(n)}, \pi_s) ds \right) \varphi_\alpha^{(n,\delta)} \left( \theta(T_\delta), \xi_{T_\delta}^{(n)} \right) \right] - \varphi_\alpha^{(n,\delta)}(\theta, x)
$$
  
\n
$$
= E_x^{\pi} \left\{ \int_0^{T_\delta(\theta)} \left[ -\alpha \theta(s) \frac{\partial \varphi_\alpha^{(n,\delta)}}{\partial \theta} (\theta(s), \xi_s^{(n)}) + \int_S q^{(n)}(dy | \xi_s^{(n)}, \pi_s) \varphi_\alpha^{(n,\delta)}(\theta(s), y) \right. \right. \\
\left. + \theta(s) c_n(\xi_s^{(n)}, \pi_s) \varphi_\alpha^{(n,\delta)}(\theta(s), \xi_s^{(n)}) \right] \times \exp \left( \int_0^s \theta(v) c_n(\xi_v^{(n)}, \pi_v) dv \right) ds \right\}.
$$
\n(4.8)

By using  $(4.3)$  and  $(4.8)$ , we obtain

$$
E_x^{\pi}\bigg[\exp\bigg(\int_0^{T_\delta(\theta)}\theta(s)c_n(\xi_s^{(n)},\pi_s)ds\bigg)\varphi_\alpha^{(n,\delta)}\bigg(\theta(T_\delta),\xi_{T_\delta}^{(n)}\bigg)\bigg]\geq \varphi_\alpha^{(n,\delta)}(\theta,x).
$$

Since  $\pi \in \Pi$  is an arbitrary control and  $\varphi_{\alpha}^{(n,\delta)}(\theta(T_{\delta}(\theta)), \xi_{T_{\delta}}^{(n)}) = e^{n\delta/\alpha}$ , we have

$$
\varphi_{\alpha}^{(n,\delta)}(\theta,x) \le \inf_{\pi \in \Pi} E_x^{\pi} \bigg[ e^{n\delta/\alpha} \exp \biggl( \int_0^{T_\delta(\theta)} \theta(s) c_n(\xi_s^{(n)}, \pi_s) ds \biggr) \biggr]. \tag{4.9}
$$

Using Eqs.  $(4.3)$ ,  $(4.7)$  and  $(4.8)$ , we can show that

$$
\varphi_{\alpha}^{(n,\delta)}(\theta,x) = E_x^{\pi^{*\delta}} \bigg[ e^{n\delta/\alpha} \exp \biggl( \int_0^{T_\delta(\theta)} \theta(s) c_n(\xi_s^{(n)}, \pi_s^{*\delta}(\xi_{s-}^{(n)})) ds \biggr) \biggr].
$$

Therefore

$$
\varphi_{\alpha}^{(n,\delta)}(\theta,x) \ge \inf_{\pi \in \Pi} E_{x}^{\pi} \bigg[ e^{n\delta/\alpha} \exp \biggl( \int_{0}^{T_{\delta}(\theta)} \theta(s) c_{n}(\xi_{s}^{(n)}, \pi_{s}) ds \biggr) \biggr]. \tag{4.10}
$$

Therefore, from [\(4.9\)](#page-14-2) and [\(4.10\)](#page-14-3), we obtain [\(4.4\)](#page-12-2). This completes the proof.  $\Box$ 

<span id="page-14-0"></span>**Theorem 4.1** *Grant Assumptions* [2.1](#page-4-1)*,* [2.2](#page-6-2) *and* [3.1](#page-7-1)*. Then the HJB Eq. [\(3.1\)](#page-6-0) has a unique solution*  $\varphi_{\alpha} \in L^{\infty}_{\mathcal{V}}([0, 1] \times S)$  *satisfying*  $1 \leq \varphi_{\alpha}(\theta, x) \leq \frac{\alpha^2 e^{\theta L_0/\alpha}}{\alpha^2 - \rho_0 \rho_1 \theta} (\mathcal{V}(x))^{\frac{\rho_1 \theta}{\alpha}}$  *for all*  $(\theta, x) \in [0, 1] \times S$ .

<span id="page-14-3"></span><span id="page-14-2"></span><sup>2</sup> Springer

*Proof* First note that  $\varphi_{\alpha}^{(n,\delta)}$  is the solution to the Eq. [\(4.3\)](#page-11-5), which depends on two parameters  $n$ ,  $\delta$ . We prove this theorem in two steps.

**Step 1:** In the first step, we construct a solution  $\varphi_{\alpha}^{(n)}(\cdot, x)$  from  $\varphi_{\alpha}^{(n,\delta)}(\cdot, x)$  by passing the limit as  $\delta \to 0$ , such that  $\varphi_{\alpha}^{(n)}(\cdot, x)$  is an absolutely continuous function and satisfies the following DEs:

$$
\begin{cases}\n\alpha \theta \frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) = \inf_{a \in A(x)} \left[ \int_{S} q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n)}(\theta, y) + \theta c_{n}(x, a) \varphi_{\alpha}^{(n)}(\theta, x) \right], \\
x \in S, \text{ a.e., } \theta \in [0, 1], \\
1 \leq \varphi_{\alpha}^{(n)}(\theta, x) \leq \frac{\alpha^{2} e^{\theta L_{0}/\alpha}}{\alpha^{2} - \rho_{0} \rho_{1} \theta} (\mathcal{V}(x))^{\frac{\rho_{1} \theta}{\alpha}} \quad \forall (\theta, x) \in [0, 1] \times S.\n\end{cases}
$$
\n(4.11)

Given  $0 < \delta < 1$  and  $1 \le n < \infty$  by [\(4.4\)](#page-12-2) and  $\sup_{x \in \mathcal{X}} c_n(x, a) \le n$ , we have (*x*,*a*)∈*K*  $\varphi_{\alpha}^{(n,\delta)}(\theta, x) \leq e^{2n/\alpha}, \quad x \in S, \theta \in [\delta, 1].$ 

Next, we extend the domain of  $\varphi_{\alpha}^{(n,\delta)}$  to [0, 1] × *S* by

<span id="page-15-0"></span>
$$
\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x) = \begin{cases} \varphi_{\alpha}^{(n,\delta)}(\theta, x), & \delta \le \theta \le 1 \,\forall x \in S \\ e^{n\delta/\alpha}, & 0 \le \theta < \delta \,\forall x \in S. \end{cases}
$$

We consider the following expression, for any given  $\pi \in \Pi$ ,  $x \in S$ ,  $\theta$ ,  $\theta_0 \in [\delta, 1]$ :

$$
\left| E_x^{\pi} \left[ e^{n\delta/\alpha} \exp \left( \theta \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \right) \right] - E_x^{\pi} \left[ e^{n\delta/\alpha} \exp \left( \theta_0 \int_0^{T_\delta(\theta_0)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \right) \right] \right|
$$
  
\$\leq P\_1 + P\_2\$,

where

$$
P_1 := \Big| E_x^{\pi} \Bigg[ e^{n\delta/\alpha} \exp \Big( \theta \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \Big) \Big] - E_x^{\pi} \Bigg[ e^{n\delta/\alpha} \exp \Big( \theta_0 \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \Big) \Big] \Big|,
$$

and

$$
P_2 := \left| E_x^{\pi} \left[ e^{n\delta/\alpha} \exp \left( \theta_0 \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \right) \right] - E_x^{\pi} \left[ e^{n\delta/\alpha} \exp \left( \theta_0 \int_0^{T_\delta(\theta_0)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \right) \right] \right|.
$$

Consider  $c \wedge d := min\{c, d\}$  and  $c \vee d := max\{c, d\}$ . Then for fix  $n \ge 1$ ; we have

$$
\int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \le n \int_0^{T_\delta(\theta)} e^{-\alpha t} dt \le \frac{n}{\alpha}
$$

and

$$
\int_{T_{\delta}(\theta \wedge \theta_0)}^{T_{\delta}(\theta \vee \theta_0)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \le n \int_{T_{\delta}(\theta \wedge \theta_0)}^{T_{\delta}(\theta \vee \theta_0)} e^{-\alpha t} dt
$$
  
= 
$$
\frac{n}{\alpha} [\exp(-\alpha T_{\delta}(\theta \wedge \theta_0)) - \exp(-\alpha T_{\delta}(\theta \vee \theta_0))] \le \frac{\delta n |\theta_0 - \theta|}{\alpha \theta \theta_0}.
$$

Using the above results and knowing the fact that  $e^{bz} - 1 \le (e^b - 1)z$  for all  $z \in [0, 1]$ and  $b > 0$ , we obtain

$$
P_1 = e^{n\delta/\alpha} E_x^{\pi} \bigg[ exp\bigg( (\theta \wedge \theta_0) \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \bigg) \times \bigg( exp\bigg( |\theta_0 - \theta| \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \bigg) - 1 \bigg) \bigg]
$$
  

$$
\leq e^{2n/\alpha} E_x^{\pi} \bigg[ exp\bigg( |\theta_0 - \theta| \int_0^{T_\delta(\theta)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \bigg) - 1 \bigg]
$$
  

$$
\leq e^{2n/\alpha} \bigg( exp\bigg( \frac{n}{\alpha} |\theta_0 - \theta| \bigg) - 1 \bigg)
$$
  

$$
\leq e^{2n/\alpha} \bigg( e^{n/\alpha} - 1 \bigg) |\theta_0 - \theta|.
$$

Similarly for  $P_2$  we have

$$
P_2 = e^{n\delta/\alpha} E_x^{\pi} \left[ \exp\left(\theta_0 \int_0^{T_\delta(\theta \wedge \theta_0)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt\right) \times \left( \exp\left(\theta_0 \int_{T_\delta(\theta \wedge \theta_0)}^{T_\delta(\theta \wedge \theta_0)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt\right) - 1 \right) \right]
$$
  

$$
\leq e^{2n/\alpha} E_x^{\pi} \left[ \exp\left(\theta_0 \int_{T_\delta(\theta \wedge \theta_0)}^{T_\delta(\theta \wedge \theta_0)} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt\right) - 1 \right]
$$
  

$$
\leq e^{2n/\alpha} \left( \exp\left(\frac{n\delta|\theta - \theta_0|}{\alpha \theta}\right) - 1 \right)
$$
  

$$
\leq e^{2n/\alpha} \left( e^{n/\alpha} - 1 \right) |\theta_0 - \theta|.
$$

Hence for all  $(\theta, x) \in [0, 1] \times S$ , we have

<span id="page-16-0"></span>
$$
|\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta_0, x) - \overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x)| \le 2e^{2n/\alpha}(e^{n/\alpha} - 1)|\theta - \theta_0|.
$$
 (4.12)

Now we want to show that  $\overline{\varphi}_{\alpha}^{(n,\delta)}$  is decreasing as  $\delta \to 0$  for any  $(\theta, x)$ . For a fixed  $\alpha > 0$  and  $\varepsilon > 0$  small enough, consider  $\overline{\varphi}_{\alpha}^{(n,\delta+\varepsilon)}(\theta, x) - \overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x)$  and assume that  $h_\delta := e^{\frac{n\delta}{\alpha}}$ . By measurable selection theorem we get the minimizer  $\pi^{*(\delta + \varepsilon)}$  like in Eq. [\(3.4\)](#page-8-2), corresponding to  $\overline{\varphi}_{\alpha}^{(n,\delta+\varepsilon)}$  such that the followings cases hold. *Case 1.* If  $\delta + \varepsilon < \theta$  then

$$
\overline{\varphi}_{\alpha}^{(n,\delta+\varepsilon)}(\theta, x) - \overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x)
$$
\n
$$
= E_{x}^{\pi^{*(\delta+\varepsilon)}} \bigg[ h_{\delta+\varepsilon} \exp \bigg( \theta \int_{0}^{T_{\delta+\varepsilon}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*(\delta+\varepsilon)}(\xi_{t-}^{(n)})) dt \bigg) \bigg]
$$
\n
$$
- \inf_{\pi \in \Pi} E_{x}^{\pi} \bigg[ h_{\delta} \exp \bigg( \theta \int_{0}^{T_{\delta}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \bigg) \bigg]
$$
\n
$$
\geq h_{\delta} E_{x}^{\pi^{*(\delta+\varepsilon)}} \bigg[ \exp \bigg( \theta \int_{0}^{T_{\delta+\varepsilon}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*(\delta+\varepsilon)}(\xi_{t-}^{(n)})) dt \bigg) \bigg]
$$
\n
$$
\times \bigg\{ h_{\varepsilon} - \exp \bigg( \theta \int_{T_{\delta+\varepsilon}}^{T_{\delta}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*(\delta+\varepsilon)}(\xi_{t-}^{(n)})) dt \bigg) \bigg\} \bigg]
$$
\n
$$
\geq h_{\delta} E_{x}^{\pi^{*(\delta+\varepsilon)}} \bigg[ \exp \bigg( \theta \int_{0}^{T_{\delta+\varepsilon}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*(\delta+\varepsilon)}(\xi_{t-}^{(n)})) dt \bigg) \bigg\{ h_{\varepsilon} - \exp \bigg( \theta \int_{T_{\delta+\varepsilon}}^{T_{\delta}} e^{-\alpha t} n dt \bigg) \bigg\} \bigg]
$$
\n
$$
= h_{\delta} E_{x}^{\pi^{*(\delta+\varepsilon)}} \bigg[ \exp \bigg( \theta \int_{0}^{T_{\delta+\varepsilon}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*(\delta+\varepsilon)}(\xi_{t-}^{(n)})) dt \bigg) \bigg\{ h_{\varepsilon} - \exp
$$

*Case 2.*  $\delta < \theta \leq \delta + \varepsilon$ 

$$
\overline{\varphi}_{\alpha}^{(n,\delta+\varepsilon)}(\theta, x) - \overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x)
$$
\n
$$
= h_{\delta+\varepsilon} - E_{x}^{\pi^{*\delta}} \bigg[ h_{\delta} \exp\bigg( \theta \int_{0}^{T_{\delta}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*\delta}(\xi_{t-}^{(n)})) dt \bigg) \bigg]
$$
\n
$$
= h_{\delta} \bigg[ h_{\varepsilon} - E_{x}^{\pi^{*\delta}} \bigg[ \exp\bigg( \theta \int_{0}^{T_{\delta}} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}^{*\delta}(\xi_{t-}^{(n)})) dt \bigg) \bigg] \bigg]
$$
\n
$$
\geq h_{\delta} \bigg[ h_{\varepsilon} - \exp\bigg( \theta \int_{0}^{T_{\delta}} e^{-\alpha t} n dt \bigg) \bigg]
$$
\n
$$
= h_{\delta} \bigg[ h_{\varepsilon} - e^{n\theta} \frac{(1 - e^{-\alpha T_{\delta}})}{\alpha} \bigg] \geq 0.
$$

*Case 3.*  $\theta < \delta$ 

$$
\overline{\varphi}_{\alpha}^{(n,\delta+\varepsilon)}(\theta,x)-\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta,x)=h_{\delta+\varepsilon}-h_{\delta}=h_{\delta}(h_{\varepsilon}-1)=h_{\delta}(e^{\frac{ne}{\alpha}}-1)\geq 0.
$$

Hence  $\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x)$  is increasing in  $\delta$  for any  $(\theta, x) \in [0, 1] \times S$ . Now from [\(4.12\)](#page-16-0), we know that for each  $x \in S$ ,  $\overline{\varphi}_{\alpha}^{(n,\delta)}(\cdot, x)$  is Lipschitz continous in  $\theta \in [0, 1]$ . Also,  $\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x)$  is increasing in  $\delta$  for any  $(\theta, x) \in [0, 1] \times S$  and bounded above (since  $\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta, x) \leq e^{2n/\alpha}, \quad x \in S, \theta \in [\delta, 1]$ , therefore there exists a function  $\varphi_{\alpha}^{(n)}$  on  $[0, 1] \times S$  that is continuous with respect to  $\theta \in [0, 1]$ , such that along a subsequence

 $\delta_m \to 0$ , we have  $\lim_{m\to\infty} \overline{\varphi}_{\alpha}^{(n,\delta_m)}(\theta, x) = \varphi_{\alpha}^{(n)}(\theta, x)$  and for any fixed  $x \in S$  this convergence is uniform in  $\theta \in [0, 1]$ .

Let  $\psi \in C_c^{\infty}(0, 1)$ , then we have

$$
-\int_{0}^{1} \alpha \frac{d(\theta \psi)}{d\theta}(\theta) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) d\theta = \int_{0}^{1} \alpha \theta \frac{\partial \overline{\varphi}_{\alpha}^{(n,\delta_{m})}}{\partial \theta}(\theta, x) \psi(\theta) d\theta
$$
  
\n
$$
=\int_{0}^{1} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) + \int_{S} q^{(n)}(dy | x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, y) \right] \psi(\theta) d\theta
$$
  
\n
$$
-\int_{0}^{\delta_{m}} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) + \int_{S} q^{(n)}(dy | x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, y) \right] \psi(\theta) d\theta
$$
  
\n
$$
=\int_{0}^{1} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) + \int_{S} q^{(n)}(dy | x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, y) \right] \psi(\theta) d\theta
$$
  
\n
$$
-\int_{0}^{\delta_{m}} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \right] \psi(\theta) d\theta.
$$
 (4.13)

Now take  $\tau(x) := M_0 V(x)$  and define

<span id="page-18-0"></span>
$$
Q^{(n)}(dy|x,a) := \delta_x(dy) + \frac{q^{(n)}(dy|x,a)}{\tau(x)}
$$

for all  $(x, a) \in K$  where  $\delta_x(\cdot)$  is the Dirac measure concentrated at *x*. We see that under Assumption [2.1,](#page-4-1)  $Q^{(n)}$  is a stochastic kernel on *S* given *K*. Then [\(4.13\)](#page-18-0) can be written as

$$
-\int_{0}^{1} \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta} \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) - \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \psi(\theta) \right\} d\theta
$$
  
\n
$$
= \int_{0}^{1} \inf_{a \in A(x)} \left[ \frac{\theta}{\tau(x)} c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) + \int_{S} Q^{(n)}(dy | x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, y) \right] \psi(\theta) d\theta
$$
  
\n
$$
- \frac{1}{\tau(x)} \int_{0}^{\delta_{m}} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \right] \psi(\theta) d\theta.
$$
 (4.14)

Now

$$
\begin{split}\n&\left|\inf_{a\in A(x)}\left[\frac{\theta}{\tau(x)}c_n(x,a)\overline{\varphi}_{\alpha}^{(n,\delta_m)}(\theta,x)+\int_{S}Q^{(n)}(dy|x,a)\overline{\varphi}_{\alpha}^{(n,\delta_m)}(\theta,y)\right]\psi(\theta)\right| \\
&\leq |\psi(\theta)|\sup_{a\in A(x)}\left[\frac{\theta}{\tau(x)}|c_n(x,a)||\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta,x)|+\int_{S}Q^{(n)}(dy|x,a)|\overline{\varphi}_{\alpha}^{(n,\delta)}(\theta,y)|\right] \\
&\leq \frac{\alpha^2}{\alpha^2-\rho_0\rho_1\theta}e^{\theta L_0/\alpha}\sup_{a\in A(x)}\left[\frac{\theta}{\tau(x)}n\mathcal{V}_{\alpha}^{\frac{\rho_1\theta}{\alpha}}(x)+\int_{S}Q^{(n)}(dy|x,a)\mathcal{V}_{\alpha}^{\frac{\rho_1\theta}{\alpha}}(y)\right]|\psi(\theta)| \\
&\leq \frac{\alpha^2}{\alpha^2-\rho_0\rho_1\theta}e^{\theta L_0/\alpha}\sup_{a\in A(x)}\left[\frac{\theta}{\tau(x)}n\mathcal{V}(x)+\int_{S}Q^{(n)}(dy|x,a)\mathcal{V}(y)\right]|\psi(\theta)|\n\end{split}
$$

$$
\leq \frac{\alpha^2}{\alpha^2 - \rho_0 \rho_1 \theta} e^{\theta L_0/\alpha} \left[ \frac{\theta}{\tau(x)} n \mathcal{V}(x) + \mathcal{V}(x) + \rho_0 \frac{\mathcal{V}(x)}{\tau(x)} \right] |\psi(\theta)|
$$
  
= 
$$
\frac{\alpha^2}{\alpha^2 - \rho_0 \rho_1 \theta} e^{\theta L_0/\alpha} \left[ \frac{\theta}{\tau(x)} n \mathcal{V}(x) + \mathcal{V}(x) + \frac{\rho_0}{M_0} \right] |\psi(\theta)|.
$$
 (4.15)

Since for each fixed  $x \in S$ ,  $A(x)$  is compact, there exist a subsequence of  $\{m\}$ , by abuse of notation, we denote the same sequence and  $a^* \in A(x)$  such that  $\lim_{m \to \infty} a_m^* = a^*$ . Now, from [\(4.13\)](#page-18-0), for any  $a \in A(x)$ , we have

$$
-\int_{0}^{1} \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta} \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) - \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \psi(\theta) \right\} d\theta
$$
  
\n
$$
= \int_{0}^{1} \left[ \frac{\theta}{\tau(x)} c_{n}(x, a_{m}^{*}) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) + \int_{S} Q^{(n)}(dy | x, a_{m}^{*}) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, y) \right] \psi(\theta) d\theta
$$
  
\n
$$
- \frac{1}{\tau(x)} \int_{0}^{\delta_{m}} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \right] \psi(\theta) d\theta.
$$
 (4.16)

So, by Lemma 8.3.7 in Hernandez-Lerma and Lassere (1999) (Hernandez-Lerma and Lass[e](#page-28-23)rre [1999\)](#page-28-23) taking limit as  $m \to \infty$  in [\(4.16\)](#page-19-0), we get

<span id="page-19-0"></span>
$$
- \int_0^1 \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta}(\theta) \varphi_\alpha^{(n)}(\theta, x) - \varphi_\alpha^{(n)}(\theta, x) \psi(\theta) \right\} d\theta
$$
  
\n
$$
\geq \int_0^1 \left[ \frac{\theta}{\tau(x)} c_n(x, a^*) \varphi_\alpha^{(n)}(\theta, x) + \int_S Q^{(n)}(dy | x, a^*) \varphi_\alpha^{(n)}(\theta, y) \right] \psi(\theta) d\theta.
$$

Hence

<span id="page-19-1"></span>
$$
-\int_{0}^{1} \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta}(\theta) \varphi_{\alpha}^{(n)}(\theta, x) - \varphi_{\alpha}^{(n)}(\theta, x) \psi(\theta) \right\} d\theta
$$
  
\n
$$
\geq \inf_{a \in A(x)} \int_{0}^{1} \left[ \frac{\theta}{\tau(x)} c_n(x, a) \varphi_{\alpha}^{(n)}(\theta, x) + \int_{S} Q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n)}(\theta, y) \right] \psi(\theta) d\theta.
$$
\n(4.17)

But

$$
- \int_{0}^{1} \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta} \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) - \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \psi(\theta) \right\} d\theta
$$
  
\n
$$
\leq \int_{0}^{1} \left[ \frac{\theta}{\tau(x)} c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) + \int_{S} Q^{(n)}(dy | x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, y) \right] \psi(\theta) d\theta
$$
  
\n
$$
- \frac{1}{\tau(x)} \int_{0}^{\delta_{m}} \inf_{a \in A(x)} \left[ \theta c_{n}(x, a) \overline{\varphi}_{\alpha}^{(n,\delta_{m})}(\theta, x) \right] \psi(\theta) d\theta.
$$

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By analogous arguments, we get

<span id="page-20-0"></span>
$$
-\int_{0}^{1} \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta}(\theta) \varphi_{\alpha}^{(n)}(\theta, x) - \varphi_{\alpha}^{(n)}(\theta, x) \psi(\theta) \right\} d\theta
$$
  

$$
\leq \inf_{a \in A(x)} \int_{0}^{1} \left[ \frac{\theta}{\tau(x)} c_n(x, a) \varphi_{\alpha}^{(n)}(\theta, x) + \int_{S} Q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n)}(\theta, y) \right] \psi(\theta) d\theta.
$$
  
(4.18)

From [\(4.17\)](#page-19-1) and [\(4.18\)](#page-20-0), we get

$$
-\int_{0}^{1} \left\{ \frac{\alpha}{\tau(x)} \frac{d(\theta \psi)}{d\theta}(\theta) \varphi_{\alpha}^{(n)}(\theta, x) - \varphi_{\alpha}^{(n)}(\theta, x) \psi(\theta) \right\} d\theta
$$
  
= 
$$
\inf_{a \in A(x)} \int_{0}^{1} \left[ \frac{\theta}{\tau(x)} c_n(x, a) \varphi_{\alpha}^{(n)}(\theta, x) + \int_{S} Q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n)}(\theta, y) \right] \psi(\theta) d\theta.
$$
  
(4.19)

Thus we obtain

$$
- \int_0^1 \alpha \frac{d(\theta \psi)}{d\theta} (\theta) \varphi_\alpha^{(n)}(\theta, x) d\theta
$$
  
= 
$$
\inf_{a \in A(x)} \int_0^1 \left[ \theta c_n(x, a) \varphi_\alpha^{(n)}(\theta, x) + \int_S q^{(n)}(dy | x, a) \varphi_\alpha^{(n)}(\theta, y) \right] \psi(\theta) d\theta.
$$

Hence

$$
\alpha \theta \frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) = \inf_{a \in A(x)} \left[ \theta c_n(x, a) \varphi_{\alpha}^{(n)}(\theta, x) + \int_S q^{(n)}(dy | x, a) \varphi_{\alpha}^{(n)}(\theta, y) \right] \text{a.e. } \theta \in [0, 1]
$$

in the sense of distribution. Now for  $\theta \in [\delta_m, 1]$ , by using [\(4.4\)](#page-12-2) and Proposition [2.1,](#page-4-3) we have

$$
\varphi_{\alpha}^{(n,\delta_{m})}(\theta, x) = \inf_{\pi \in \Pi} E_{x}^{\pi} \bigg[ e^{n\delta_{m}/\alpha} \exp\bigg(\theta \int_{0}^{T_{\delta_{m}}(\theta)} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \bigg) \bigg]
$$
  
\n
$$
\leq e^{n\delta_{m}/\alpha} \inf_{\pi \in \Pi} E_{x}^{\pi} \bigg[ \exp\bigg(\theta \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \bigg) \bigg]
$$
  
\n
$$
\leq e^{n\delta_{m}/\alpha} \inf_{\pi \in \Pi} E_{x}^{\pi} \bigg[ \exp\bigg(\theta \int_{0}^{\infty} e^{-\alpha t} c(\xi_{t}^{(n)}, \pi_{t}) dt \bigg) \bigg]
$$
  
\n
$$
\leq e^{n\delta_{m}/\alpha} \frac{\alpha^{2} e^{\theta L_{0}/\alpha}}{\alpha^{2} - \rho_{0} \rho_{1} \theta} (\mathcal{V}(x))^{\frac{\rho_{1} \theta}{\alpha}}
$$

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Note that  $\varphi_{\alpha}^{(n,\delta_m)} \to \varphi_{\alpha}^{(n)}$  as  $m \to \infty$ . Thus, letting  $m \to \infty$  in the above equation, we obtain

<span id="page-21-0"></span>
$$
1 \le \varphi_{\alpha}^{(n)}(\theta, x) \le \frac{\alpha^2 e^{\theta L_0/\alpha}}{\alpha^2 - \rho_0 \rho_1 \theta} (\mathcal{V}(x))^{\frac{\rho_1 \theta}{\alpha}}.
$$
 (4.20)

By using [\(4.1\)](#page-11-3), [\(4.2\)](#page-11-4), [\(4.20\)](#page-21-0), and the DE satisfied by  $\varphi_{\alpha}^{(n)}$  (that is just proven), we see that  $\varphi_n^{(n)} \in L^\infty_V([0, 1] \times S)$  and it is a solution of [\(4.11\)](#page-15-0). Thus by closely mimicking the arguments as in Theorem 3.1, one can essily get the stochastic pornectation of the arguments as in Theorem [3.1,](#page-7-2) one can easily get the stochastic representation of the solution  $\varphi_{\alpha}^{(n)}$ , that is

<span id="page-21-1"></span>
$$
\varphi_{\alpha}^{(n)}(\theta, x) = \inf_{\pi \in \Pi} E_x^{\pi} \bigg[ \exp \bigg( \theta \int_0^{\infty} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \bigg) \bigg]. \tag{4.21}
$$

**Step 2:** In this step we prove Theorem [4.1,](#page-14-0) by passing to the limit as  $n \to \infty$ . Now we will prove that for each  $x \in S$ ,  $\{\varphi_{\alpha}^{(n)}\}_{n>1}$  is equicontinuous on [0, 1]. We consider the following expression, for any given  $\pi \in \Pi$ ,  $x \in S$ ,  $\theta$ ,  $\theta_0 \in [0, 1]$ :

$$
\left| E_x^{\pi} \left[ \exp \left( \theta \int_0^{\infty} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \right) \right] - E_x^{\pi} \left[ \exp \left( \theta_0 \int_0^{\infty} e^{-\alpha t} c_n(\xi_t^{(n)}, \pi_t) dt \right) \right] \right|
$$
  
\$\leq K\_1\$,

where

$$
K_{1} = E_{x}^{\pi} \left[ \exp\left( (\theta \wedge \theta_{0}) \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) \times \left( \exp\left( |\theta_{0} - \theta| \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) - 1 \right) \right]
$$
  
\n
$$
\leq E_{x}^{\pi} \left[ \exp\left( (\theta \wedge \theta_{0}) \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) \times \left( \exp\left( \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) - 1 \right) |\theta_{0} - \theta| \right]
$$
  
\n
$$
\leq E_{x}^{\pi} \left[ \exp\left( \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) \times \left( \exp\left( \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) |\theta_{0} - \theta| \right) \right]
$$
  
\n
$$
= |\theta_{0} - \theta| \times E_{x}^{\pi} \left[ \exp\left( 2 \int_{0}^{\infty} e^{-\alpha t} c_{n}(\xi_{t}^{(n)}, \pi_{t}) dt \right) \right]
$$
  
\n
$$
\leq |\theta_{0} - \theta| \times \frac{\alpha e^{2L_{0}/\alpha}}{\alpha - \rho_{2}} \left( \mathcal{V}^{2}(x) + \frac{b_{0}}{\rho_{2}} \right).
$$

Here, the first inequality is according to  $e^{bz} - 1 \le (e^b - 1)z$  for all  $z \in [0, 1]$  and  $b > 0$  and the last inequality follows from  $(3.7)$ . Therefore, we have

$$
|\varphi_{\alpha}^{(n)}(\theta_0, x) - \varphi_{\alpha}^{(n)}(\theta, x)| \le \sup_{\pi \in \Pi} |\theta_0 - \theta| \times \frac{\alpha e^{2L_0/\alpha}}{\alpha - \rho_2} \left(\mathcal{V}^2(x) + \frac{b_0}{\rho_2}\right)
$$

$$
= |\theta_0 - \theta| \times \frac{\alpha e^{2L_0/\alpha}}{\alpha - \rho_2} \left(\mathcal{V}^2(x) + \frac{b_0}{\rho_2}\right). \tag{4.22}
$$

By measurable selection theorem, [Bertsekas and Shrev[e](#page-28-22) [\(1996\)](#page-28-22), Proposition 7.33], there exists a measurable function  $f^{*n}$  : [0, 1]  $\times S \rightarrow A$  such that

$$
\inf_{a \in A(x)} \left[ \theta c_n(x, a) \varphi_\alpha(\theta, x) + \int_S q^{(n)}(dy | x, a) \varphi_\alpha(\theta, y) \right]
$$
\n
$$
= \left[ \theta c_n(x, f^{*n}(\theta, x)) \varphi_\alpha(\theta, x) + \int_S q^{(n)}(dy | x, f^{*n}(\theta, x)) \varphi_\alpha(\theta, y) \right]. \tag{4.23}
$$

Let

<span id="page-22-2"></span>
$$
\pi^{*n}: S \times \mathbb{R}_+ \to A
$$

be defined by

$$
\pi_t^{*n}(x) := f^{*n}(\theta e^{-\alpha t}, x).
$$

Hence by Eq. [\(4.11\)](#page-15-0), we have a.e.  $\theta \in [0, 1]$  and  $\forall x \in S$ , we have

$$
\begin{cases}\n\alpha \theta \frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) = \left[ \int_{S} q^{(n)}(dy | x, f^{*n}(\theta, x)) \varphi_{\alpha}^{(n)}(\theta, y) + \theta c_n(x, f^{*n}(\theta, x)) \varphi_{\alpha}^{(n)}(\theta, x) \right] \\
1 \leq \varphi_{\alpha}^{(n)}(\theta, x) \leq \frac{\alpha^{2} e^{\theta L_0/\alpha}}{\alpha^{2} - \rho_0 \rho_1 \theta} (\mathcal{V}(x))^{\frac{\rho_1 \theta}{\alpha}} \quad \forall (\theta, x) \in [0, 1] \times S.\n\end{cases}
$$
\n(4.24)

Since  $c_n \ge 0$ , by [\(4.21\)](#page-21-1), we see  $\varphi_{\alpha}^{(n)}(\theta, x)$  is increasing in  $\theta$ . Also we know that  $\varphi_{\alpha}^{(n)}(\theta, x)$  is differentiable a.e. with respect to  $\theta \in [0, 1]$ . So

<span id="page-22-1"></span>
$$
\frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) \ge 0 \text{ for a.e. } \theta.
$$
 (4.25)

So, by [\(4.1\)](#page-11-3), [\(4.2\)](#page-11-4) and [\(4.24\)](#page-22-0), for all  $x \in S$  and for a.e.  $\theta$ , we have

$$
\begin{cases}\n-\alpha \theta \frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) + \left[ \int_{S} q^{(n-1)}(dy | x, f^{*n}(\theta, x)) \varphi_{\alpha}^{(n)}(\theta, y) + \theta c_{n-1}(x, f^{*n}(\theta, x)) \varphi_{\alpha}^{(n)}(\theta, x) \right] \\
\leq 0 \quad \text{if } x \in S_{n-1}\n\end{cases}
$$
\n(4.26)

<span id="page-22-0"></span> $\bigcirc$  Springer

and

$$
\begin{cases}\n-\alpha \theta \frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) + \left[ \int_{S} q^{(n-1)}(dy | x, f^{*n}(\theta, x)) \varphi_{\alpha}^{(n)}(\theta, y) + \theta c_{n-1}(x, f^{*n}(\theta, x)) \varphi_{\alpha}^{(n)}(\theta, x) \right] \\
= -\alpha \theta \frac{\partial \varphi_{\alpha}^{(n)}}{\partial \theta}(\theta, x) \le 0 \\
\text{if } x \notin S_{n-1} \text{ (by (4.25)).}\n\end{cases}
$$
\n(4.27)

So, by Dynkin formula, we get

$$
E_{x}^{\pi^{*n}} \bigg[ \exp \bigg( \theta \int_{0}^{\infty} e^{-\alpha t} c_{n-1}(\xi_{t}^{(n-1)}, \pi_{t}^{*n}(\xi_{t-}^{(n-1)})) dt \bigg) \bigg] \leq \varphi_{\alpha}^{(n)}(\theta, x) \text{ for all } (\theta, x) \in [0, 1] \times S. \tag{4.28}
$$

Also using  $(4.11)$  and Dynkin formula (see  $(3.7)$  and  $(3.13)$ ), we have

<span id="page-23-1"></span>
$$
\varphi_{\alpha}^{(n-1)}(\theta, x) \le E_{x}^{\pi^{*n}} \bigg[ \exp \bigg( \theta \int_{0}^{\infty} e^{-\alpha t} c_{n-1}(\xi_{t}^{(n-1)}, \pi_{t}^{*n}(\xi_{t-}^{(n-1)})) dt \bigg) \bigg]. \tag{4.29}
$$

By [\(4.28\)](#page-23-1) and [\(4.29\)](#page-23-2), we have  $\varphi_{\alpha}^{(n-1)}(\theta, x) \leq \varphi_{\alpha}^{(n)}(\theta, x)$ .

Hence  $\varphi_n^{(n)}(\theta, x)$  is increasing in *n* for any  $(\theta, x) \in [0, 1] \times S$ . Now from [\(4.22\)](#page-22-2), we know that for each  $x \in S$ ,  $\varphi^{(n)}(\cdot, x)$  is Lipschitz continuous in  $\theta \in [0, 1]$ . Also,  $\varphi_{\alpha}^{(n)}(\theta, x)$  is increasing as  $n \to \infty$  for any  $(\theta, x) \in [0, 1] \times S$  and bounded above (by [\(4.20\)](#page-21-0)), therefore there exists a function  $\varphi_{\alpha}$  on [0, 1] × *S* that is continuous with respect to  $\theta \in [0, 1]$ , such that along a subsequence  $n_k \to \infty$ , we have  $\lim_{n_k \to \infty} \varphi_\alpha^{(n_k)}(\theta, x) =$  $\varphi_{\alpha}(\theta, x)$  and this convergence is uniform in  $\theta \in [0, 1]$  for each fixed  $x \in S$ . Moreover, by  $(4.20)$ , we have

<span id="page-23-3"></span><span id="page-23-2"></span>
$$
1 \le \varphi_{\alpha}(\theta, x) \le \frac{\alpha^2 e^{\theta L_0/\alpha}}{\alpha^2 - \rho_0 \rho_1 \theta} (\mathcal{V}(x))^{\frac{\rho_1 \theta}{\alpha}}.
$$
\n(4.30)

As the proof of equation  $(4.11)$  in step 1 (starting from the first equality of  $(4.13)$ ), we see that  $\varphi_{\alpha}$  is a solution to the HJB Eq. [\(3.1\)](#page-6-0). Also by [\(4.30\)](#page-23-3), we can conclude that  $\varphi_{\alpha} \in L^{\infty}_{\mathcal{V}}([0, 1] \times S)$ . Finally, the uniqueness of  $\varphi_{\alpha}(\theta, x)$  follows from the stochastic proposes taking in Theorem 2.1. representation in Theorem [3.1.](#page-7-2)

#### <span id="page-23-0"></span>**5 The existence of optimal control**

<span id="page-23-4"></span>In this section, we present the main result of this article. Here we show the existence of an optimal control.

**Theorem 5.1** *Suppose that Assumptions* [2.1](#page-4-1)*,* [2.2](#page-6-2) *and* [3.1](#page-7-1) *are satisfied. Then, the following assertions hold.*

(1) *The HJB Eq.*[\(3.1\)](#page-6-0) *has a unique solution*  $\varphi_{\alpha} \in L^{\infty}_{\mathcal{V}}([0, 1] \times S)$  *and the solution admits the following representation admits the following representation*

$$
1 \leq \varphi_{\alpha}(\theta, x) = \inf_{\pi \in \Pi} E_{x}^{\pi} \left[ \exp \left( \theta \int_{0}^{\infty} e^{-\alpha t} c(\xi_{t}, \pi_{t}) dt \right) \right]
$$

$$
\leq \frac{\alpha^{2} e^{\theta L_{0}/\alpha}}{\alpha^{2} - \rho_{0} \rho_{1} \theta} (\mathcal{V}(x))^{\frac{\rho_{1} \theta}{\alpha}}.
$$

(2) *There exists a measurable function*  $f^* : [0, 1] \times S \rightarrow A$  *such that* 

$$
\alpha \theta \frac{\partial \varphi_{\alpha}}{\partial \theta}(\theta, x) = \left[ \int_{S} q(dy|x, f^{*}(\theta, x)) \varphi_{\alpha}(\theta, y) + \theta c(x, f^{*}(\theta, x)) \varphi_{\alpha}(\theta, x) \right]
$$
  
a.e.  $\theta \in [0, 1].$  (5.1)

(3) *Furthermore an optimal Markov control for the cost criterion* [\(2.2\)](#page-4-0) *exists and is given by*

<span id="page-24-0"></span>
$$
\tilde{\pi}_t^*(x) := f^*(\theta e^{-\alpha t}, x),
$$

*where*  $f^*$  *satisfies* [\(5.1\)](#page-24-0).

*Proof* Part (1) follows from Theorems [3.1](#page-7-2) and [4.1.](#page-14-0)

To prove (2), by Hernandez-Lerma and Lasserr[e](#page-28-23) [\(1999](#page-28-23)), we first observe that the function

$$
\int_{S} q(dy|x, a)\varphi_{\alpha}(\theta, y) + \theta c(x, a)\varphi_{\alpha}(\theta, x)
$$

is continuous in  $a \in A(x)$  for each given  $(\theta, x) \in [0, 1] \times S$ . Thus, by the measurable selection theorem (Bertsekas and Shrev[e](#page-28-22) [1996\)](#page-28-22), Proposition 7.33 there exists a measurable function  $f^*$  satisfying [\(5.1\)](#page-24-0), and so (2) follows. For part (3), take any  $f^*$  that satisfies [\(5.1\)](#page-24-0). Then by Theorem [3.1,](#page-7-2) we have  $\inf_{\pi \in \Pi} J_\alpha(\theta, x, \pi) =$  $J_{\alpha}(\theta, x, \tilde{\pi}^*) = \varphi_{\alpha}(\theta, x)$ , which together with [\(2.2\)](#page-4-0), [\(2.3\)](#page-4-2) and part (1), we have inf  $\inf_{\pi \in \Pi} \mathcal{J}_{\alpha}(\theta, x, \pi) = \mathcal{J}_{\alpha}(\theta, x, \tilde{\pi}^*) = \frac{1}{\theta} \ln \tilde{J}_{\alpha}(\theta, x, \tilde{\pi}^*) = \frac{1}{\theta} \ln \varphi_{\alpha}(\theta, x)$ . Hence  $\tilde{\pi}^*$ is an optimal Markov control.

Now we prove the converse of the Theorem [5.1.](#page-23-4)

**Theorem 5.2** *Grant Assumptions* [2.1](#page-4-1)*,* [2.2](#page-6-2) *and* [3.1](#page-7-1)*. Suppose there exists an optimal Markov control for the cost criterion* [\(2.2\)](#page-4-0) *and is given by*

$$
\hat{\pi}_t^*(x) := \tilde{f}^*(\theta e^{-\alpha t}, x),
$$

*for some measurable function*  $\tilde{f}^*$ *. Then we prove that*  $\tilde{f}^*$  *is a minimizing selector of* [\(3.1\)](#page-6-0)*.*

*Proof* sSince  $\hat{\pi}^*$  is optimal for the cost criterion [\(2.2\)](#page-4-0), therefore we have

<span id="page-25-3"></span><span id="page-25-1"></span>
$$
\inf_{\pi \in \Pi} \tilde{J}_{\alpha}(\theta, x, \pi) = \tilde{J}_{\alpha}(\theta, x, \hat{\pi}^*) = \tilde{J}_{\alpha}^*(\theta, x). \tag{5.2}
$$

Now for  $f^*$  by Theorem [4.1,](#page-14-0) there exists a unique solution  $\psi_\alpha \in L^\infty_\mathcal{V}([0, 1] \times S)$  for the equation the equation

$$
\alpha \theta \frac{\partial \psi_{\alpha}}{\partial \theta}(\theta, x) = \left[ \int_{S} q(dy|x, \tilde{f}^*(\theta, x)) \psi_{\alpha}(\theta, y) + \theta c(x, \tilde{f}^*(\theta, x)) \psi_{\alpha}(\theta, x) \right],
$$
\n(5.3)

for each  $x \in S$  and a.e.  $\theta \in [0, 1]$ , satisfying  $1 \leq \psi_{\alpha}(\theta, x) \leq \frac{\alpha^2 e^{\theta L_0/\alpha}}{\alpha^2 - \rho_0 \rho_1 \theta} (\mathcal{V}(x))^{\frac{\rho_1 \theta}{\alpha}}$  for all  $(\theta, x) \in [0, 1] \times S$ .

Now by Theorem [3.1,](#page-7-2) we know that

<span id="page-25-2"></span>
$$
1 \leq \psi_{\alpha}(\theta, x) = E_{x}^{\hat{\pi}^{*}} \bigg[ \exp\bigg(\theta \int_{0}^{\infty} e^{-\alpha t} c(\xi_{t}, \hat{\pi}_{t}^{*}(\xi_{t-})) dt \bigg) \bigg]
$$

$$
\leq \frac{\alpha^{2} e^{\theta L_{0}/\alpha}}{\alpha^{2} - \rho_{0} \rho_{1} \theta} (\mathcal{V}(x))^{\frac{\rho_{1} \theta}{\alpha}}.
$$
(5.4)

From  $(5.2)$  and  $(5.4)$ , we get

<span id="page-25-4"></span>
$$
\psi_{\alpha}(\theta, x) = \inf_{\pi \in \Pi} \tilde{J}_{\alpha}(\theta, x, \pi) = \tilde{J}_{\alpha}(\theta, x, \hat{\pi}^*) = \tilde{J}_{\alpha}^*(\theta, x) \text{ for } (\theta, x) \in [0, 1] \times S. \tag{5.5}
$$

So, in view of Theorem [3.1,](#page-7-2) by Eqs. [\(3.1\)](#page-6-0), [\(5.3\)](#page-25-3), and [\(5.5\)](#page-25-4), we conclude that  $\tilde{f}^*$  is a minimizing selector of  $(3.1)$ .

<span id="page-25-0"></span>When the transition and cost rates are bounded, the existence of an optimal control is ensured by Theorem [5.1.](#page-23-4)

**Corollary 5.1** *Grant Assumption* [3.1](#page-7-1) *((i)–(ii)). Also, assume that the transition and cost rates are bounded. Then, there exist a unique solution*  $\varphi_{\alpha}$  *and an optimal control for the HJB Eq.* [\(3.1\)](#page-6-0)*.* 

*Proof* Suppose there exist constants  $L_1$  and  $b_1$ , such that (*x*,*a*)∈*K*  $\sup q_x(a) \leq L_1$  and  $\sup_{x \to a} c(x, a) \leq b_1$ . First we take the Lyapunov function  $V(x) \equiv P$ , for all  $x \in C$ (*x*,*a*)∈*K*

*S*,  $P \ge 1$ , a constant. Now  $\int_S V(y)q(dy|x, a) = \int_S V^2(y)q(dy|x, a) = 0$ , for all  $(x, a) \in K$ . Now, take  $\rho_0 = \alpha$ ,  $M_0 = L_1$ , any real number,  $\rho_1 \in (0, \alpha)$ , and  $L_0 = b_1$ . Then Assumption [2.1](#page-4-1) is verified. Now for all  $x \in S$ , take any constants  $\rho_2 \in (0, \alpha)$ and  $b_0 \in (0, \infty)$ . Then Assumption [2.2](#page-6-2) holds. Also  $\int_S V(y)q(dy|x, a)$  is continuous in *a* ∈ *A*(*x*). So, Assumption [3.1](#page-7-1) is also true. Then, by Theorem [5.1,](#page-23-4) we have a unique solution  $\omega_{\alpha}$  and an optimal control for the HJB Eq. (3.1). solution  $\varphi_{\alpha}$  and an optimal control for the HJB Eq. [\(3.1\)](#page-6-0).

#### <span id="page-26-0"></span>**6 Application and example**

In this section, we verify the above assumptions with one example, where the transition and cost rates are unbounded.

*Example 6.1* **The Gaussian Model:** Suppose a hunter is hunting outside his house for his manager. Suppose the house is at state 0. A positive state represents the distance from the house to the right, and a negative state represents the distance from the house to the left. Let  $S = \mathbb{R}$ . If the current position is  $x \in S$ , the hunter takes an action  $a \in A(x)$ , then after an exponentially distributed travel time with rate  $\lambda(x, a) > 0$ , the hunter reaches the new position, and the travel distance follows the normal distribution with mean *x* and variance  $\sigma$ . (Or we can interpret  $\lambda(x, a)$  as the total jump intensity that is an arbitrary measurable positive-valued function on  $S \times A$ , and the distribution of the state after a jump from  $x \in S$  is normal with the variance  $\sigma$  and expectation *x*.) Also assume that the hunter receives a payoff  $c(x, a)$ from his manager for each unit of time he spends there. Let us consider the model as  $A_2 := \{S, (A, A(x), x \in S), c(x, a), q(dy|x, a)\}$ , where  $S = (-\infty, \infty)$ . For each  $D \in \mathcal{B}(S)$ , the transition rate is

$$
q(D|x,a) = \lambda(x,a) \left[ \int_{y \in D} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(y-x)^2}{2\sigma^2}} dy - \delta_x(D) \right], \ x \in S, a \in A(x), \sigma > 0.
$$
\n
$$
(6.1)
$$

To ensure the existence of an optimal Markov control for the model, we consider the following hypotheses.

- (I) For each fixed  $x \in S$ ,  $\lambda(x, a)$  is continuous in  $a \in A(x)$  and there exists a positive constant  $M_1$  such that  $0 < \text{sup}$  $\sup_{a \in A(x)} \lambda(x, a) \le M_1(x^2 + 1)$  and  $M_1 < \frac{\alpha}{6\sigma^2(\sigma^2 + 1)}$ .
- (II) For each  $x \in S$ , the cost rate  $c(x, a)$  is nonnegative and continuous in  $a \in A(x)$ and there exists a constant  $0 < \rho_1 < \min{\{\alpha, \frac{\alpha^2}{M_1 \sigma^2}\}}$  such that

$$
\sup_{a \in A(x)} c(x, a) \le \rho_1 \log(1 + x^2).
$$

(III) For each fixed  $x \in S$ ,  $A(x)$  is a compact subset of the Borel spaces A.

**Proposition 6.1** *Under conditions (I)–(III), the above controlled system satisfies the Assumptions* [2.1](#page-4-1)*,* [2.2](#page-6-2)*, and* [3.1](#page-7-1)*. Hence by Theorem* [5.1](#page-23-4)*, there exists an optimal Markov control for this model.*

**Proof** We know  $\frac{1}{\sqrt{2}}$  $\frac{1}{2\pi\sigma}\int_{-\infty}^{\infty}(y-x)^{2k+1}e^{-\frac{(y-x)^2}{2\sigma^2}}dy = 0$  and  $\frac{1}{\sqrt{2\pi}}$  $\frac{1}{2\pi\sigma}\int_{-\infty}^{\infty}(y$  $f(x) = \int_0^{x} \frac{(y-x)^2}{2\sigma^2} dy = 1 \cdot 3 \cdots (2k-1)\sigma^{2k}$  for all  $k = 0, 1 \cdots$ . We first verify Assumption [2.1.](#page-4-1) Let  $V(x) = x^2 + 1$ .

$$
\int_{S} \mathcal{V}(y)q(dy|x,a) = \lambda(x,a) \left[ \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} (y^2 + 1)e^{-\frac{(y-x)^2}{2\sigma^2}} dy - (x^2 + 1) \right]
$$

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<span id="page-27-0"></span>
$$
= \lambda(x, a)\sigma^2
$$
  
\n
$$
\leq M_1 \sigma^2 \mathcal{V}(x).
$$
 (6.2)

Let  $\rho_0 = M_1 \sigma^2$ . Then  $\int_S V(y)q(dy|x, a) \le \rho_0 V(x)$ . Now

$$
q^*(x) = \sup_{a \in A(x)} q_x(a) = \sup_{a \in A(x)} \lambda(x, a) \le M_1(x^2 + 1) = M_1 \mathcal{V}(x) \,\forall \, x \in S.
$$

Now by condition (II), we can write

$$
\sup_{a \in A(x)} c(x, a) \le \rho_1 \log(1 + x^2) + M_1.
$$

Observe that by condition (II),  $0 < \rho_1 < \min{\{\alpha, \rho_0^{-1}\alpha^2\}}$ . Hence Assumption [2.1](#page-4-1) is verified with  $M_0 = L_0 = M_1$ . Next we verify Assumption [2.2.](#page-6-2) For any  $x \in S$ ,  $a \in A(x)$ ,

$$
\int_{S} q(dy|x, a)\mathcal{V}^{2}(y) = \int_{S} q(dy|x, a)(1 + y^{2})^{2}
$$
\n
$$
= \lambda(x, a) \left[ \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} (y^{2} + 1)^{2} e^{-\frac{(y-x)^{2}}{2\sigma^{2}}} dy - (x^{2} + 1)^{2} \right]
$$
\n
$$
= \lambda(x, a)[1 \cdot 3\sigma^{4} + \sigma^{2}(2 + 6x^{2})]
$$
\n
$$
\leq M_{1}(x^{2} + 1)[3\sigma^{4} + \sigma^{2}(2 + 6x^{2})]
$$
\n
$$
= M_{1}\sigma^{2}(x^{2} + 1)(3\sigma^{2} + 2 + 6x^{2})
$$
\n
$$
\leq 6M_{1}\sigma^{2}(x^{2} + 1)(x^{2} + 1)(\sigma^{2} + 1)
$$
\n
$$
= 6M_{1}\mathcal{V}^{2}(x)\sigma^{2}(\sigma^{2} + 1)
$$
\n
$$
\leq \rho_{2}\mathcal{V}^{2}(x) + 1
$$

where  $\rho_2 = 6M_1\sigma^2(\sigma^2 + 1)$ , and  $b_0 = 1$ . Then by condition (I), we have  $0 < \rho_2 < \alpha$ . Hence, Assumption [2.2](#page-6-2) is verified. Now by conditions (I) and (II)  $c(x, a)$  is continuous in *a* ∈ *A*(*x*). Observe that by condition (I) and [\(6.2\)](#page-27-0),  $\int_S V(y)q(dy|x, a)$  is continuous in *a* ∈ *A*(*x*). Hence Assumption [3.1](#page-7-1) is also verified. So, by Theorem [5.1,](#page-23-4) we see that there exists an optimal Markov control for this model.  $\Box$ there exists an optimal Markov control for this model.

*Remark 6.1* As we mention in the introduction, there are many real-life applications, where the underlying system dynamic is modeled as a CTMDP, with a Borel state and action spaces as well as cost and transition rates are unbounded, see such a cash-flow problem in Guo and Zhan[g](#page-28-12) [\(2019](#page-28-12)), [p. 112, Piunovskiy and Zhan[g](#page-28-13) [\(2020\)](#page-28-13)]. Also, there are lots of real-life examples like infrastructure surveillance models [p. 115– 116, Piunovskiy and Zhan[g](#page-28-13) [\(2020](#page-28-13))], queueing model [p. 192, Piunovskiy and Zhan[g](#page-28-13) [\(2020\)](#page-28-13)], where we see that the state space is uncountable, can be formulated in our set-up.

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