

# On semi-Markov games

By

Kensuke TANAKA and Kazuyoshi WAKUTA

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## 1. Introduction

This paper is a continuation of our paper [4] and is concerned with semi-Markov games with some criterion. In the Markov games, time until the transition from a state to a next state occurs is a unit time, but it does not seem general enough. For this reason, in the paper, we shall consider the semi-Markov game which time until the transition occurs is a known random variable.

However, so far as we know, such the games have not been tried up to the present. Hence, at first, we shall give the formulation of semi-Markov game with the criterion of long-run average reward as the game proceeds over the infinite future. Then, we shall show that the game has a value and there exist the optimal stationary strategies for both players under this criterion and some assumptions. Moreover, we shall give a sufficient condition for some important assumption.

This paper consists of four sections. In Section 2, we shall give the formulation of the problem treated by us in this paper. In Section 3, we shall show the existence of optimal stationary strategies and, in Section 4, we shall give a sufficient condition.

## 2. The formulation of the problem

In this paper, we determine "*semi-Markov game*" by six objects  $(S, A, B, q, F, r)$ . Here,  $S$  is a non-empty Borel subset of a Polish space, the set of states of a system;  $A$  is a non-empty Borel subset of a Polish space, the set of actions available to player I;  $B$  is a non-empty Borel subset of a Polish space, the set of actions available to player II;  $q$  is the law of motion of the system, it associates Borel measurably with each triple  $(s, a, b) \in S \times A \times B$  a probability measure  $q(\cdot | s, a, b)$  on the Borel measurable space  $(S, \mathfrak{B}(S))$ , where  $\mathfrak{B}(S)$  is the  $\sigma$ -field generated by the metric on  $S$ ;  $F(\cdot | s, a, b, s')$  is a distribution of time until the transition from  $s$  to  $s'$  occurs, given that the next state is  $s'$ ;  $r$ , the reward function, is a bounded Borel measurable function on  $S \times A \times B \times R$ , where  $R$  is a real line. At successive random times, player I and player II observe the current state  $s$  of the system and choose actions  $a$  and  $b$ , respectively, according to the full knowledge of the history of

the system as it has evolved to the present. As a consequence of the actions chosen by the players and the duration time of  $s$ , player II pays player I  $r(s, a, b, t)$  units of money, and the system moves to a new state  $s'$  according to the conditional distribution  $q(\cdot | s, a, b)$  after some duration of state  $s$ . Then the whole process is repeated from the state  $s'$ . Here, our optimization problem is to maximize the limit of expected reward of player I gained during the first  $n$  transitions divided by the expected length of the first  $n$  transitions as the game proceeds over the infinite future and to minimize the limit of expected loss of player II incurred during the first  $n$  transitions divided by the expected length of the first  $n$  transitions.

A strategy  $\pi$  for player I is a sequence of  $\pi_1, \pi_2, \dots$ , where  $\pi_n$  specifies the  $n$ th action to be chosen by player I by associating Borel measurably with each history  $h_n = (s_1, a_1, b_1, t_1, \dots, s_{n-1}, a_{n-1}, b_{n-1}, t_{n-1}, s_n)$  of the system a probability distribution  $\pi_n(\cdot | h_n)$  on  $(A, \mathfrak{B}(A))$ , where  $s_i, a_i, b_i$ , and  $t_i$  are the  $i$ th state, the  $i$ th action chosen by player I, the  $i$ th action chosen by player II and the  $i$ th duration time, respectively. A strategy  $\pi$  is, particularly, said to be *stationary* if there is a Borel measurable map  $f$  from  $S$  to  $P_A$ , where  $P_A$  is the set of all probability measures on  $(A, \mathfrak{B}(A))$ , such that  $\pi_n = f$  for all  $n$  and in this case,  $\pi$  is denoted by  $f^\infty$ .  $\Pi$  denotes the class of all strategies for player I. Strategies and stationary strategies for player II are defined analogously.  $\Gamma$  denotes the class of all strategies for player II.

In order to ensure that the transitions do not take place too quickly, we shall need to assume the following:

ASSUMPTION 1. There exists  $\delta > 0, \varepsilon > 0$  such that

$$\int_S F(\delta | s, a, b, s') dq(s' | s, a, b) < 1 - \varepsilon \quad (2.1)$$

for all  $s, a$  and  $b$ .

DEFINITION 1. A strategy  $\pi^*$  is optimal for player I if for each  $\sigma \in \Gamma$  and  $s_1 \in S$ ,

$$\inf_{\sigma \in \Gamma} \sup_{\pi \in \Pi} \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i | s_1 \right]} \quad (2.2)$$

$$\leq \lim_{n \rightarrow \infty} \frac{E_{\pi^*, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi^*, \sigma} \left[ \sum_{i=1}^n t_i | s_1 \right]},$$

where  $E_{\pi, \sigma}$  denotes the expectation by the pair  $(\pi, \sigma)$  of strategies for player I and player II.

A strategy  $\sigma^*$  is optimal for player II if for each  $\pi \in \Pi$  and  $s_1 \in S$ ,

$$\begin{aligned} & \sup_{\pi \in \Pi} \inf_{\sigma \in \Gamma} \lim_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i \mid s_1 \right]} & (2.3) \\ & \geq \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \sigma^*} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma^*} \left[ \sum_{i=1}^n t_i \mid s_1 \right]} \end{aligned}$$

DEFINITION 2. Semi-Markov game has a value if for each  $s_1 \in S$ ,

$$\begin{aligned} & \inf_{\sigma \in \Gamma} \sup_{\pi \in \Pi} \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i \mid s_1 \right]} & (2.4) \\ & = \sup_{\pi \in \Pi} \inf_{\sigma \in \Gamma} \lim_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i \mid s_1 \right]} \end{aligned}$$

In the case semi-Markov game has a *value*, the quantity

$$\inf_{\sigma \in \Gamma} \sup_{\pi \in \Pi} \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i \mid s_1 \right]},$$

as a function on  $S$ , is called the value function.

### 3. Existence of optimal strategies

In this section, we are concerned with the existence of optimal stationary strategies for our semi-Markov game. We have to impose some assumptions on  $S, A, B, q, F$  and  $r$  to ensure that there exist optimal strategies.

ASSUMPTION 2. (i)  $A, B$  and  $S$  are compact metric spaces, (ii) whenever  $s_n \rightarrow s_0, a_n \rightarrow a_0$  and  $b_n \rightarrow b_0, q(\cdot \mid s_n, a_n, b_n)$  converges weakly to  $q(\cdot \mid s_0, a_0, b_0)$ .

ASSUMPTION 3. (i)  $\int_0^\infty t dF(t \mid s, a, b, s') \equiv \tau(s, a, b, s')$  is a continuous function on  $S \times A \times B \times S$ , (ii)

$\int_0^\infty r(s, a, b, t) dF(t \mid s, a, b, s') \equiv r(s, a, b, s')$  is a continuous function on  $S \times A \times B \times S$ .

From these assumptions we can show the following lemma.

LEMMA 3.1.  $\bar{\tau}(s, a, b)$  and  $\bar{r}(s, a, b)$  are bounded continuous functions on  $S \times A \times B$ , where

$$\bar{\tau}(s, a, b) = \int_S \tau(s, a, b, s') dq(s' | s, a, b)$$

and

$$\bar{r}(s, a, b) = \int_S r(s, a, b, s') dq(s' | s, a, b).$$

PROOF. It holds that

$$\begin{aligned} & |\bar{\tau}(s_n, a_n, b_n) - \bar{\tau}(s_0, a_0, b_0)| \\ & \leq \int_S |\tau(s_n, a_n, b_n, s') - \tau(s_0, a_0, b_0, s')| dq(s' | s_n, a_n, b_n) \\ & \quad + \left| \int_S \tau(s_0, a_0, b_0, s') dq(s' | s_n, a_n, b_n) \right. \\ & \quad \left. - \int_S \tau(s_0, a_0, b_0, s') dq(s' | s_0, a_0, b_0) \right|. \end{aligned} \tag{3.1}$$

Then, from Assumption 2 (i), (ii), Assumption 3 (i) and (3.1), we can prove that  $\bar{\tau}(s, a, b)$  is a bounded continuous function on  $S \times A \times B$ . Similarly,  $\bar{r}(s, a, b)$  is a bounded continuous function on  $S \times A \times B$ .

ASSUMPTION 4. There exist a continuous function  $u(s)$  on  $S$  and a constant  $d$  such that for each  $s \in S$ ,

$$u(s) = \sup_{\mu \in P_A} \inf_{\lambda \in P_B} \{ \bar{r}(s, \mu, \lambda) + \int_S u(s') dq(s' | s, \mu, \lambda) - d \bar{\tau}(s, \mu, \lambda) \}, \tag{3.2}$$

where for each  $\mu \in P_A$ ,  $\lambda \in P_B$  and  $E \in \mathfrak{G}(S)$ ,

$$\bar{r}(s, \mu, \lambda) \equiv \int_A \int_B \bar{r}(s, a, b) d\lambda(b) d\mu(a),$$

$$\bar{\tau}(s, \mu, \lambda) \equiv \int_A \int_B \bar{\tau}(s, a, b) d\lambda(b) d\mu(a),$$

and

$$q(E | s, \mu, \lambda) \equiv \int_A \int_B q(E | s, a, b) d\lambda(b) d\mu(a).$$

Since  $A$  and  $B$  are compact metric spaces,  $P_A$  and  $P_B$ , endowed with weak topology, are compact metric spaces.

LEMMA 3.2. Let  $u(s, a, b)$  be a continuous, real-valued function on  $S \times A \times B$ . Then under Assumption 2 (i),

$$u(s, \mu, \lambda) = \int_A \int_B u(s, a, b) d\lambda(b) d\mu(a),$$

$s \in S$ ,  $\mu \in P_A$ ,  $\lambda \in P_B$  is continuous function on  $S \times P_A \times P_B$ .

LEMMA 3.3. Let  $u$  be a bounded, continuous function on  $X \times Y$ , where  $X$  is a Borel subset of a Polish space and  $Y$  is a compact metric space. Then,  $u^*(x) = \max_{y \in Y} u(x, y)$  is continuous. Moreover,  $u_*(x) = \min_{y \in Y} u(x, y)$  is also continuous.

LEMMA 3.4. Let  $u$  be a bounded, continuous function on  $X \times Y$ , where  $X$  is a Borel subset of a Polish space and  $Y$  is a compact metric space. Then, there exist Borel measurable maps  $f$  and  $g$  from  $X$  into  $Y$  such that  $u(x, f(x)) = \max_{y \in Y} u(x, y)$  and  $u(x, g(x)) = \min_{y \in Y} u(x, y)$ ,  $x \in X$ .

The proofs of these lemmas are given in Lemma 2.1, Lemma 2.2 and Lemma 2.3 of [1].

Let  $C(S)$  denote the family of all bounded, continuous functions on  $S$ . For  $u \in C(S)$  we define  $\|u\| = \sup_{s \in S} |u(s)|$ . Then  $(C(S), d)$  is a complete metric space, where  $d(u, v) = \|u - v\|$  for each  $u, v \in C(S)$ .

Now, for  $u \in C(S)$  and  $d$  in Assumption 4, we define

$$K(s, \mu, \lambda) \equiv \bar{r}(s, \mu, \lambda) + \int_S u(s') dq(s' | s, \mu, \lambda) - d\bar{\tau}(s, \mu, \lambda). \quad (3.3)$$

Then, by virtue of Lemma 3.2,  $K(s, \mu, \lambda)$  is a continuous function on  $S \times P_A \times P_B$ .  $K(s, \mu, \lambda)$ ,  $P_A$  and  $P_B$  satisfy the conditions of Sion's minimax theorem (Theorem 3.4 of [3]) because of its bilinearity in  $(\mu, \lambda)$  and, consequently,

$$\sup_{\mu \in P_A} \inf_{\lambda \in P_B} K(s, \mu, \lambda) = \inf_{\lambda \in P_B} \sup_{\mu \in P_A} K(s, \mu, \lambda), \quad s \in S. \quad (3.4)$$

Moreover, since  $K(s, \mu, \lambda)$  is continuous on  $S \times P_A \times P_B$  and  $P_A, P_B$  are compact, sup and inf can be replaced by max and min, respectively. Thus, (3.2) can be written as follows:

$$u(s) = \max_{\mu \in P_A} \min_{\lambda \in P_B} K(s, \mu, \lambda) = \min_{\lambda \in P_B} \max_{\mu \in P_A} K(s, \mu, \lambda), \quad s \in S. \quad (3.5)$$

LEMMA 3.5. There exist Borel measurable maps  $\mu_*$  and  $\lambda_*$  from  $S$  into  $P_A$  and  $P_B$ , respectively, such that

$$\begin{aligned} \min_{\lambda \in P_B} K(s, \mu_*, \lambda) &= \max_{\mu \in P_A} \min_{\lambda \in P_B} K(s, \mu, \lambda) \\ &= \min_{\lambda \in P_B} \max_{\mu \in P_A} K(s, \mu, \lambda) \\ &= \max_{\mu \in P_A} K(s, \mu, \lambda_*), \quad s \in S. \end{aligned} \quad (3.6)$$

The proof is immediate from Lemma 3.3 and Lemma 3.4.

THEOREM 3.1. Under the Assumptions 1, 2, 3, 4, semi-Markov game has a value and both players have optimal stationary strategies.

PROOF. By Assumption 4 and Lemma 3.5, there exists a Borel measurable map  $\mu_*$  from  $S$  into  $P_A$  such that

$$\begin{aligned}
u(s) &= \max_{\mu \in P_A} \min_{\lambda \in P_B} \{ \bar{r}(s, \mu, \lambda) + \int_S u(s') dq(s' | s, \mu, \lambda) - d\bar{\tau}(s, \mu, \lambda) \} \\
&= \max_{\mu \in P_A} \min_{\lambda \in P_B} K(s, \mu, \lambda) \\
&= \min_{\lambda \in P_B} K(s, \mu_*, \lambda).
\end{aligned}$$

For a stationary strategy  $\mu_*^\infty = (\mu_*, \mu_*, \dots)$  for player I and any strategy  $\sigma$  for player II, we have

$$E_{\mu_*^\infty, \sigma} \left[ \sum_{t=2}^{n+1} \{ u(s_t) - E_{\mu_*^\infty, \sigma} [ u(s_t) | h_{t-1} ] \} \right] = 0. \quad (3.7)$$

But, for each  $t$ , it holds that

$$\begin{aligned}
E_{\mu_*^\infty, \sigma} [ u(s_t) | h_{t-1} ] &= \int_S u(s') dq(s' | s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \\
&= \{ \bar{r}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) + \int_S u(s') dq(s' | s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \\
&\quad - d\bar{\tau}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \} - \{ \bar{r}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \\
&\quad - d\bar{\tau}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \} \geq \min_{\lambda \in P_B} \{ r(s_{t-1}, \mu_*(s_{t-1}), \lambda) \\
&\quad + \int_S u(s') dq(s' | s_{t-1}, \mu_*(s_{t-1}), \lambda) - d\bar{\tau}(s_{t-1}, \mu_*(s_{t-1}), \lambda) \} \\
&\quad - \{ \bar{r}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) - d\bar{\tau}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \} \\
&= u(s_{t-1}) - \{ \bar{r}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) - d\bar{\tau}(s_{t-1}, \mu_*(s_{t-1}), \lambda_{t-1}) \},
\end{aligned} \quad (3.8)$$

where  $\lambda_{t-1}$  denotes a probability measure on  $(B, \mathfrak{B}(B))$  determined by  $\sigma_{t-1}(\cdot | h_{t-1})$ .

Hence, from (3.7) and (3.8),

$$0 \leq E_{\mu_*^\infty, \sigma} \left[ \sum_{t=2}^{n+1} \{ u(s_t) - (u(s_{t-1}) - \bar{r}(s_{t-1}, \mu_*, \lambda_{t-1}) + d\bar{\tau}(s_{t-1}, \mu_*, \lambda_{t-1})) \} \right] \quad (3.9)$$

$$= E_{\mu_*^\infty, \sigma} \left[ u(s_{n+1}) - u(s_1) + \sum_{t=2}^{n-1} ( \bar{r}(s_{t-1}, \mu_*, \lambda_{t-1}) - d\bar{\tau}(s_{t-1}, \mu_*, \lambda_{t-1}) ) \right],$$

or

$$d \leq \frac{E_{\mu_*^\infty, \sigma} \left[ \sum_{t=2}^{n+1} \bar{r}(s_{t-1}, \mu_*, \lambda_{t-1}) \right]}{E_{\mu_*^\infty, \sigma} \left[ \sum_{t=2}^{n+1} \bar{\tau}(s_{t-1}, \mu_*, \lambda_{t-1}) \right]} + \frac{E_{\mu_*^\infty, \sigma} [ u(s_{n+1}) - u(s_1) ]}{E_{\mu_*^\infty, \sigma} \left[ \sum_{t=2}^{n+1} \bar{\tau}(s_{t-1}, \mu_*, \lambda_{t-1}) \right]} \quad (3.10)$$

$$= \frac{E_{\mu_{*}, \sigma}^{\infty} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\mu_{*}, \sigma}^{\infty} \left[ \sum_{i=1}^n t_i | s_1 \right]} + \frac{E_{\mu_{*}, \sigma}^{\infty} [u(s_{n+1}) - u(s_1)]}{E_{\mu_{*}, \sigma}^{\infty} \left[ \sum_{i=1}^n t_i | s_1 \right]} .$$

By Assumption 1, it is easy to see that

$$E_{\mu_{*}, \sigma}^{\infty} \left[ \sum_{i=1}^n t_i | s_1 \right] \geq n \varepsilon \delta \longrightarrow \infty \text{ as } n \longrightarrow \infty. \quad (3.11)$$

Since  $u$  is bounded, from (3.10) and (3.11), we obtain

$$d \leq \lim_{n \rightarrow \infty} \frac{E_{\mu_{*}, \sigma}^{\infty} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\mu_{*}, \sigma}^{\infty} \left[ \sum_{i=1}^n t_i | s_1 \right]} \quad \text{for any } \sigma \in \Gamma \quad (3.12)$$

Thus, from (3.12), it holds that for any  $s_1 \in S$

$$d \leq \sup_{\pi \in \Pi} \inf_{\sigma \in \Gamma} \lim_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i | s_1 \right]} . \quad (3.13)$$

Similarly, from Assumption 4 and Lemma 3.5, it holds that, for any  $\pi \in \Pi$  and  $s_1 \in S$

$$\begin{aligned} d &\geq \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \lambda_{*}}^{\infty} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \lambda_{*}}^{\infty} \left[ \sum_{i=1}^n t_i | s_1 \right]} \quad (3.14) \\ &\geq \inf_{\sigma \in \Gamma} \sup_{\pi \in \Pi} \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i | s_1 \right]} . \end{aligned}$$

On the other hand, generally we have

$$\begin{aligned} &\sup_{\pi \in \Pi} \inf_{\sigma \in \Gamma} \lim_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i | s_1 \right]} \quad (3.15) \\ &\leq \inf_{\sigma \in \Gamma} \sup_{\pi \in \Pi} \overline{\lim}_{n \rightarrow \infty} \frac{E_{\pi, \sigma} \left[ \sum_{i=1}^n r(s_i, a_i, b_i, t_i) \right]}{E_{\pi, \sigma} \left[ \sum_{i=1}^n t_i | s_1 \right]} \quad \text{for any } s_1 \in S. \end{aligned}$$

By (3. 13), (3. 14) and (3. 15), our semi-Markov game has a value  $d$  and  $\mu_*^\infty$  and  $\lambda_*^\infty$  are stationary optimal strategies for player I and player II, respectively. Thus the proof is complete.

#### 4. Sufficient condition

In this section, we shall give the sufficient condition to ensure Assumption 4.

First we define an operator  $T_\alpha: C(S) \rightarrow C(S)$  as follows: for  $\alpha > 0$

$$(T_\alpha u)(s) = \max_{\mu \in P_A} \min_{\lambda \in P_B} \left[ \int_A \int_B e^{-\alpha \bar{\tau}(s, a, b)} \{ \bar{r}(s, a, b) + \int_S u(s') dq(s' | s, a, b) \} d\lambda(b) d\mu(a) \right]. \quad (4. 1)$$

LEMMA 4. 1. *The operator  $T_\alpha$  is a contraction mapping on  $C(S)$  for any  $\alpha > 0$ .*

PROOF. It is easily proved that, for any  $u, v \in C(S)$ ,  $\|T_\alpha u - T_\alpha v\|$

$$\leq \max_{\mu \in P_A} \min_{\lambda \in P_B} \left[ \int_A \int_B e^{-\alpha \bar{\tau}(s, a, b)} \|u - v\| d\lambda(b) d\mu(a) \right] \quad (4. 2)$$

$$\text{and, by Assumption 1,} \quad \bar{\tau}(s, a, b) \geq \delta \varepsilon. \quad (4. 3)$$

From (4. 2) and (4. 3), we have

$$\|T_\alpha u - T_\alpha v\| \leq e^{-\alpha \delta \varepsilon} \|u - v\|.$$

This completes the proof.

Since  $C(S)$  is a complete metric space,  $T_\alpha$  has a unique fixed point in  $C(S)$ , by virtue of the Banach fixed point theorem. Let  $u_\alpha^*$  be the unique fixed point of  $T_\alpha$ . Then it holds that, for each  $s \in S$ ,

$$u_\alpha^*(s) = \max_{\mu \in P_A} \min_{\lambda \in P_B} \left[ \int_A \int_B e^{-\alpha \bar{\tau}(s, a, b)} \{ \bar{r}(s, a, b) + \int_S u_\alpha^*(s') dq(s' | s, a, b) \} d\lambda(b) d\mu(a) \right]. \quad (4. 4)$$

Now, fix some state 0 and let

$$f_\alpha(s) = u_\alpha^*(s) - u_\alpha^*(0). \quad (4. 5)$$

From (4. 4) and (4. 5), we have

$$u_\alpha^*(0) + f_\alpha(s) = \max_{\mu \in P_A} \min_{\lambda \in P_B} \left[ \int_A \int_B e^{-\alpha \bar{\tau}(s, a, b)} \{ \bar{r}(s, a, b) + \int_S f_\alpha(s') dq(s' | s, a, b) \} d\lambda(b) d\mu(a) \right]$$

$$+u_{\alpha^*}(0)\int_A\int_B e^{-\alpha\bar{\tau}(s,a,b)} d\lambda(b) d\mu(a)]. \quad (4.6)$$

But

$$\int_A\int_B e^{-\alpha\bar{\tau}(s,a,b)} d\lambda(b) d\mu(a) = 1 - \alpha\bar{\tau}(s, \mu, \lambda) + \int_A\int_B 0(\alpha) d\lambda(b) d\mu(a). \quad (4.7)$$

Hence, from (4.6) and (4.7),

$$\begin{aligned} f_{\alpha}(s) = & \max_{\mu \in P_A} \min_{\lambda \in P_B} \left[ \int_A\int_B e^{-\alpha\bar{\tau}(s,a,b)} \{\bar{r}(s, a, b) \right. \\ & + \int_S f_{\alpha}(s') dq(s'|s, a, b)\} d\lambda(b) d\mu(a) - \alpha u_{\alpha^*}(0) \bar{\tau}(s, \mu, \lambda) \\ & \left. + u_{\alpha^*}(0) \Sigma \right], \end{aligned} \quad (4.8)$$

where

$$\Sigma = \int_A\int_B 0(\alpha) d\lambda(b) d\mu(a).$$

**THEOREM 4.1.** *If  $\{f_{\alpha}(s), 0 < \alpha < c\}$  is a uniformly bounded, equi-continuous family of functions on  $S$  for some  $0 < c < \infty$ . Then, Assumption 4 holds.*

**PROOF.** By Ascoli-Arzelà's theorem there exist a sequence  $\alpha_n \rightarrow 0$  and a continuous  $u(s)$  such that  $f_{\alpha_n}(s)$  converges uniformly to  $u(s)$  on  $S$ .

Now we show that  $\{\alpha u_{\alpha^*}(0), 0 < \alpha < c\}$  is bounded. By virtue of Lemma 3.2 and Lemma 3.3, for a fixed point  $u_{\alpha^*}$ , there exist Borel measurable maps  $\mu_*$  and  $\lambda_*$  from  $S$  into  $P_A$  and  $P_B$  such that for each  $s \in S$

$$\begin{aligned} u_{\alpha^*}(s) = & \int_A\int_B e^{-\alpha\bar{\tau}(s,a,b)} \{\bar{r}(s, a, b) \\ & + \int_S u_{\alpha^*}(s') dq(s'|s, a, b)\} d\lambda_*(b) d\mu_*(a). \end{aligned} \quad (4.9)$$

From (4.9), it is easy to see that for each  $s_1 \in S$

$$u_{\alpha^*}(s_1) = E_{\mu_*^{\infty}, \lambda_*^{\infty}} \left[ \sum_{n=1}^{\infty} e^{-\sum_{k=1}^n \alpha\bar{\tau}(s_k, a_k, b_k)} \bar{r}(s_n, a_n, b_n) \right]. \quad (4.10)$$

Then, since  $|\bar{r}| \leq M$  and  $|\bar{\tau}| \geq \varepsilon \delta$ , from (4.10)  $|\alpha u_{\alpha^*}|$  is bounded. Hence, we can require that  $\alpha_n u_{\alpha_n^*}(0)$  converges to  $d$  as  $\alpha_n \rightarrow 0$  and we can show that  $u_{\alpha^*}(0) \Sigma$  converges uniformly to zero in  $\mu$  and  $\lambda$  as  $\alpha_n \rightarrow 0$ . Thus, from (4.9), we get

$$u(s) = \max_{\mu \in P_A} \min_{\lambda \in P_B} \{\bar{r}(s, \mu, \lambda) + \int_S u(s') dq(s'|s, \mu, \lambda) - d\bar{\tau}(s, \mu, \lambda)\}.$$

This completes the proof.

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