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MARKOV DECISION PROGRAMMING – THE MOMENT OPTIMAL PROBLEM FOR THE FIRST-PASSAGE MODEL

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Abstract

In this paper, we discuss MDP-the moment optimal problem for the first-passage model. A policy improvement iteration algorithm is given for finding the k-moment optimal stationary policy.

1. Introduction

Allowing for the risk factor Jaquette [5, 6] posed a moment optimality model for a discounted Markov decision process. Sobel [15] presented a formula for the *k*-th moment of the total discounted return. A minimal variance problem (that is, a two-moment optimal problem) in optimal policies for the discounted MDP was discussed in [2, 12]. A moment optimality model in which the discount factor is dependent on history was discussed in [10]. For other works in the field see also Baykal-Gürsoy and Ross [1], Filar, Kallenberg and Lee [3], Filar and Lee [4], Kawai [7], Chung [8, 9], Sobel [13, 14] and White [16].

This paper discusses the moment optimal problem for the first-passage model on the basis of [11]. The first-passage model is also of practical interest. In particular, the model can be applied to solve optimal control problems of reliability and queueing systems and other controlled stochastic systems.

A k-moment is defined in Section 2. Some formulas for k-moments are given by Theorem 2.1 in Section 2. Sufficient and necessary conditions for a policy π to be a k-moment optimal policy are given by Theorem 2.6. Theorems 2.7 and 2.8 state that the problems of the existence and calculation of a k-moment optimal policy (or a moment-optimal policy) in the space of general policies can be changed into the same problems in the space of deterministic stationary policies. Theorem 2.9 states that there exists a stationary policy which is moment optimal if A is nonempty and

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finite. An algorithm of policy-improvement type is given in Section 3 for finding the k-moment optimal stationary policy.

The first-passage model with denumerable state space is $\{S, A, q, r, V_k\}$, where the state space S and action set A are nonempty and countable. Let $S = \{0, 1, 2, ...\}$, $S_0 = \{1, 2, 3, ...\}$. A one-step reward r satisfies $|r(i, a)| \le M$ and r(0, a) = 0, $i \in S, a \in A$. The symbol q denotes the family of stationary one-step transition laws: when the system is in state i and we take an action a, the system moves to a new state j selected according to the conditional probability q(j|i, a), where q satisfies $q(0|0, a) = 1, a \in A$. A definition of criterion V_k is given in Section 2.

The set of general policies $\pi = (\pi_0, \pi_1, \pi_2, ...)$ is denoted by Π . A mapping $f: S \to A$ is called a deterministic decision rule. Let F denote the set of all deterministic decision rules f. For $f \in F$, $f^{\infty} = (f, f, ...)$ is called a stationary policy. Π_s^d denotes the set of all stationary policies. Obviously $\Pi_s^d \subset \Pi$.

At any stage $t \ge 0$, X_t and Δ_t denote respectively a state of the system and an action taken in that state.

ASSUMPTION A. There exists a real number $\alpha > 0$ and a positive integer N such that $P_{\pi}\{x_N = 0 | x_0 = i\} \ge \alpha$ for $\forall \pi \in \Pi, \forall i \in S_0$.

In the following, we assume that Assumption A is always true.

Let $X_0 = i_0$, $\Delta_0 = a_0$, $X_1 = i_1$, $\Delta_1 = a_1$, ..., $X_n = i_n$. The sequence $h_n = (i_0, a_0, i_1, a_1, ..., i_n)$ is called a history up to stage n and $H_n (n \ge 0)$ denotes the set of all h_n .

Let $\pi = (\pi_0, \pi_1, \pi_2, \dots) \in \Pi$, $h_n = (i_0, a_0, i_1, a_1, \dots, i_n) \in H_n (n \ge 1)$. The policy $\pi' = (\pi'_0, \pi'_1, \dots) \in \Pi$ is defined as follows. For $\forall t \ge 0, \forall h_t \in H_t$, define

$$\pi'_{t}(a|h_{t}) = \pi_{n+t}(a|i_{0}, a_{0}, i_{1}, a_{1}, \ldots, a_{n-1}, h_{t}), \qquad a \in A.$$

Write $\pi' = \pi(i_0, a_0, \dots, i_{n-1}, a_{n-1})$ or $\pi' = \pi(\overline{h}_n)$.

The following facts stated here without proof are derived in [11].

LEMMA 1.1. Let $n \ge N$, $i_0 \in S_0$, $\pi \in \Pi$, then

$$\sum_{i\in\mathcal{S}_0} P_{\pi}\{X_n = i | X_0 = i_0\} \le (1-\alpha)^{[n/N]},$$

where [X] denotes the greatest integer which does not exceed X.

LEMMA 1.2.

$$\sum_{t=0}^{\infty}\sum_{j\in S_0}P_{\pi}\{X_t=j|X_0=i\}\leq \frac{N}{\alpha} \quad for \,\forall i\in S_0, \; \forall \pi\in \Pi.$$

PROOF. This follows immediately from the proof of Lemma 2.2 in [11].

Suppose $X_0 = i$ and let τ denote the smallest integer t such that $X_t = 0$. Let

$$V(\pi, i) = E_{\pi}\left[\sum_{t=0}^{\tau} r(X_t, \Delta_t) | X_0 = i\right], \qquad \pi \in \Pi, \ i \in S$$

 $V(\pi, i)$ is the expected total reward obtained using the policy π starting from *i*. Let $V^*(i) = \sup_{\pi \in \Pi} V(\pi, i), i \in S.$

THEOREM 1.1 (Optimality equation).

$$V^{*}(i) = \sup_{a \in A} \left\{ r(i, a) + \sum_{j \in S_{0}} q(j|i, a) V^{*}(j) \right\}, \qquad i \in S.$$

Let $\pi \in \Pi$, $h_n = (i_0, a_0, i_1, a_1, \dots, i_n) \in H_n$. If $P_{\pi}\{X_0 = i_0, \Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \dots, X_n = i_n | X_0 = i_0 \} > 0$, then h_n is called a realizable history under the policy π .

Let

$$A^{*}(i) = \left\{ a \in A | r(i, a) + \sum_{j \in S_{0}} q(j|i, a) V^{*}(j) = V^{*}(i) \right\}, \qquad i \in S.$$

THEOREM 1.2. Let $i \in S$, $\pi \in \Pi$. Then a necessary and sufficient condition that $V(\pi, i) = V^*(i)$ is that for $\forall n \ge 0$, if $h_n = (i, a_0, \dots, i_n)$ is a realizable history under the policy π and $\pi_n(a|h_n) > 0$, then $a \in A^*(i_n)$.

PROOF. Similar to the proof of Theorem 2.4 in [11].

By Theorem 1.2 we have

- COROLLARY 1.1. (1) If $f(i) \in A^*(i)$ for all $i \in S$, then $V(f^{\infty}, i) = V^*(i)$ for all $i \in S$.
- (2) Let $i \in S$, $\pi = (\pi_0, \pi_1, ...) \in \Pi$ and $V(\pi, i) = V^*(i)$. If $\pi_0(a|i) > 0$, then $a \in A^*(i)$.

COROLLARY 1.2 (Bellman's optimality principle). Let $i \in S$, $\pi \in \Pi$ and $V(\pi, i) = V^*(i)$. If $h_n = (i, a_0, i_1, a_1, \dots, i_n)$ $(n \ge 1)$ is a realizable history under the policy π , then $V(\pi(\bar{h}_n), i_n) = V^*(i_n)$.

PROOF. Let $\pi(\overline{h}_n) = (\pi'_0, \pi'_1, \pi'_2, ...) \forall m \ge 0$. Let $\tilde{h}_m = (i_n, \tilde{a}_0, \tilde{i}_1, \tilde{a}_1, ..., \tilde{i}_m) \in H_m$ be a realizable history under the policy $\pi(\overline{h}_n)$ and $\pi'_m(a|\tilde{h}_m) > 0$. It is easy to see, $(i, a_0, i_1, a_1, ..., i_n, \tilde{a}_0, \tilde{i}_1, \tilde{a}_1, ..., \tilde{i}_m)$ is a realizable history under policy π . By the definition of $\pi(\overline{h}_n)$,

$$\pi_{n+m}(a|i, a_0, i_1, a_1, \dots, i_{n-1}, a_{n-1}, h_m) = \pi'_m(a|h_m) > 0,$$

by Theorem 1.2 (necessity), $a \in A^*(\tilde{i}_m)$. So, by Theorem 1.2 (sufficiency), $V(\pi(\bar{h}_n), i_n) = V^*(i_n)$.

THEOREM 1.3. If f^{∞} is optimal in Π_s^d (that is, $V(f^{\infty}, i) \ge V(g^{\infty}, i)$ for $\forall i \in S$, $\forall g^{\infty} \in \Pi_s^d$.), then f^{∞} is also optimal in Π (that is, $V(f^{\infty}, i) \ge V(\pi, i)$ for $\forall i \in S$, $\forall \pi \in \Pi$).

LEMMA 1.3. Let S be finite, $f \in F$. If a set of numbers $\{V(i) : i \in S_0\}$ satisfies

$$V(i) = \sum_{j \in S_0} q(j|i, f(i))V(j), \qquad i \in S_0,$$

then $V(i) \equiv 0, i \in S_0$.

Let $V_1, V_2 \in \mathbb{R}^n (n \ge 1), V_i = (V_i(1), V_i(2), \dots, V_i(n)), i = 1, 2$. Define

$$V_1 \ge V_2 \iff V_1(i) \ge V_2(i)$$
 for $i = 1, 2, ..., n$.
 $V_1 > V_2 \iff V_1 \ge V_2$ and $V_1 \ne V_2$.

2. The moment optimal problem

By the Cauchy criterion, we know that $\sum_{n=N+1}^{\infty} n^p (1-\alpha)^{[n-1/N]}$ is convergent for $p = 1, 2, \dots$ Let

$$D(\alpha, N, p) = \left[\sum_{n=N+1}^{\infty} n^{p} (1-\alpha)^{[n-1/N]}\right] + N^{p}, \qquad p = 1, 2, \dots$$

LEMMA 2.1. Let $i \in S_0, \pi \in \Pi, p = 1, 2,$ Then

$$E_{\pi}[\tau^{p}|X_{0}=i] \leq D(\alpha, N, p).$$

PROOF. By Lemma 1.1,

$$\begin{split} E_{\pi}[\tau^{p}|X_{0}=i] &= \sum_{n=1}^{\infty} n^{p} P_{\pi}\{\tau=n|X_{0}=i\} \\ &= \sum_{n=1}^{N} n^{p} P_{\pi}\{\tau=n|X_{0}=i\} + \sum_{n=N+1}^{\infty} n^{p} P_{\pi}\{\tau=n|X_{0}=i\} \\ &\leq N^{p} \sum_{n=1}^{N} P_{\pi}\{\tau=n|X_{0}=i\} + \sum_{n=N+1}^{\infty} n^{p} P_{\pi}\{X_{n-1}\neq 0|X_{0}=i\} \\ &\leq N^{p} P_{\pi}\{\tau\leq N|X_{0}=i\} + \sum_{n=N+1}^{\infty} n^{p} (1-\alpha)^{[n-1/N]} \\ &\leq D(\alpha, N, p). \end{split}$$

So, by Lemma 2.1, when $i \in S_0, \pi \in \Pi, p = 1, 2, ...,$

$$E_{\pi} \left[\left| \sum_{t=0}^{\tau} r(X_{t}, \Delta_{t}) \right|^{p} |X_{0} = i \right] \leq E_{\pi} [M^{p} (\tau + 1)^{p} |X_{0} = i]$$
$$\leq (2M)^{p} E_{\pi} [\tau^{p} |X_{0} = i]$$
$$\leq (2M)^{p} D(\alpha, N, p).$$
(2.1)

DEFINITION 2.1. Let

$$V_k(\pi, i) = E_{\pi} \left\{ \left[\sum_{t=0}^{\tau} r(X_t, \Delta_t) \right]^k | X_0 = i \right\}, \ i \in S, \pi \in \Pi, \ k = 1, 2, \dots$$

Let $V_0(\pi, i) \equiv 1$, $i \in S$, $\pi \in \Pi$.

It is easy to see, $V_k(\pi, 0) = 0, \pi \in \Pi, k = 1, 2, \dots$ Because r(0, a) = 0 and q(0|0, a) = 1, we have

$$V_k(\pi, i) = E_{\pi} \left\{ \left[\sum_{t=0}^{\infty} r(X_t, \Delta_t) \right]^k | X_0 = i \right\}, \ i \in S, \pi \in \Pi, \ k = 1, 2, \dots$$

THEOREM 2.1. Let $\pi = (\pi_0, \pi_1, ...) \in \Pi$, $i \in S, k = 1, 2,$ Then

$$V_{k}(\pi, i) = \sum_{a \in A} \pi_{0}(a|i) \left\{ R_{k}(i, a, \pi) + \sum_{j \in S} q(j|i, a) V_{k}(\pi(i, a), j) \right\},\$$

[6] Markov decision programming – the moment optimal problem For the first-passage model 547 *where*

$$R_{k}(i, a, \pi) = \sum_{p=0}^{k-1} C_{k}^{p} r^{k-p}(i, a) \sum_{j \in S} q(j|i, a) V_{p}(\pi(i, a), j),$$
$$r^{k-p}(i, a) \equiv [r(i, a)]^{k-p}.$$

The definition of $\pi(i, a)$ can be found in Section 1.

PROOF. Let $i \in S_0$, k = 1, 2, ... By the total mathematical expectation formula,

$$\begin{aligned} V_{k}(\pi, i) &= E_{\pi} \left\{ \left[\sum_{t=0}^{\infty} r(X_{t}, \Delta_{t}) \right]^{k} | X_{0} = i \right\} \\ &= \sum_{a \in A} \pi_{0}(a|i) E_{\pi} \left\{ \left[\sum_{t=0}^{\infty} r(X_{t}, \Delta_{t}) \right]^{k} | X_{0} = i, \Delta_{0} = a \right\} \\ &= \sum_{a \in A} \pi_{0}(a|i) E_{\pi} \left\{ \left[r(i, a) + \sum_{t=1}^{\infty} r(X_{t}, \Delta_{t}) \right]^{k} | X_{0} = i, \Delta_{0} = a \right\} \\ &= \sum_{a \in A} \pi_{0}(a|i) \left[\sum_{p=0}^{k-1} C_{k}^{p} r^{k-p}(i, a) \sum_{j \in S} q(j|i, a) V_{p}(\pi(i, a), j) + \sum_{j \in S} q(j|i, a) V_{k}(\pi(i, a), j) \right] \\ &= \sum_{a \in A} \pi_{0}(a|i) \left[R_{k}(i, a, \pi) + \sum_{j \in S} q(j|i, a) V_{k}(\pi(i, a), j) \right]. \end{aligned}$$

The proposition is obviously true for i=0.

Let $M_l(\pi) = (-1)^{l+1} V_l(\pi), \pi \in \Pi, l = 0, 1, 2, \dots$, where $V_l(\pi)$ is a vector and its *i*-th component is $V_l(\pi, i), i \in S$.

Let $M^k(\pi) = (M_0(\pi), M_1(\pi), \dots, M_k(\pi)), \pi \in \Pi, \ k = 1, 2, \dots$

DEFINITION 2.2. Let $k \ge 1$, $\pi_1, \pi_2 \in \Pi$. $M^k(\pi_1) > M^k(\pi_2) \iff \exists n, 1 \le n \le k$, such that $M_l(\pi_1) = M_l(\pi_2)$ for l < n and $M_n(\pi_1) > M_n(\pi_2)$.

$$M^k(\pi_1) \ge M^k(\pi_2) \iff M^k(\pi_1) > M^k(\pi_2)$$
 or $M^k(\pi_1) = M^k(\pi_2)$.

DEFINITION 2.3. Let $k \ge 1$, $\pi^* \in \Pi$. If $M^k(\pi^*) \ge M^k(\pi)$ for $\forall \pi \in \Pi$, then π^* is called a k-moment optimal policy in Π .

If π^* is a k-moment optimal policy in Π for all $k \ge 1$, then π^* is called a moment-optimal policy in Π .

The set of the k-moment optimal policies in Π is denoted by $\Pi(k)(k \ge 1)$. Let $\Pi(0) = \Pi$. The set of the moment optimal policy in Π is denoted by $\Pi(\infty)$. Obviously, $\Pi(\infty) = \bigcap_{k=1}^{\infty} \Pi(k)$. It is easy to see by the definition that $\Pi(k) \subset \Pi(k-1)$, $k \ge 1$.

DEFINITION 2.4. Let $M_0^*(i) \equiv -1$, $\Pi(0, i) \equiv \Pi$, $i \in S$ and define $M_n^*(i)$ and $\Pi(n, i)(i \in S, n \ge 1)$ as follows. If $\Pi(n - 1, i) \ne \emptyset$, then

$$M_n^*(i) = \sup_{\pi \in \Pi(n-1,i)} M_n(\pi, i),$$

$$\Pi(n, i) = \{\pi \in \Pi(n-1, i) | M_n(\pi, i) = M_n^*(i) \},$$

where $M_n(\pi, i) = (-1)^{n+1} V_n(\pi, i)$.

It is easy to see that $\Pi(n, 0) \equiv \Pi, n = 0, 1, 2, ... By (2.1)$,

$$|M_n^*(i)| \le (2M)^n D(\alpha, N, n), \qquad i \in S, \ n = 1, 2, \dots$$
 (2.2)

DEFINITION 2.5. Let

$$R_n(i,a) = (-1)^{n+1} \sum_{k=0}^{n-1} C_n^k (-1)^{k+1} r^{n-k}(i,a) \sum_{j \in S} q(j|i,a) M_k^*(j),$$

$$i \in S, \ a \in A, \ n = 1, 2, \dots$$

Let $A_0^*(i) \equiv A$, $i \in S$ and define $A_n^*(i)$ $(i \in S, n \ge 1)$ as follows. If $A_{n-1}^*(i) \neq \emptyset$ and $\Pi(n-1, j) \neq \emptyset$ for all $j \in S$, then

$$A_{n}^{*}(i) = \left\{ a \in A_{n-1}^{*}(i) | R_{n}(i,a) + \sum_{j \in S} q(j|i,a) M_{n}^{*}(j) \right.$$
$$= \sup_{\tilde{a} \in A_{n-1}^{*}(i)} \left[R_{n}(i,\tilde{a}) + \sum_{j \in S} q(j|i,\tilde{a}) M_{n}^{*}(j) \right] \right\}.$$

It is easy to see that $R_n(0, a) \equiv 0$, $a \in A$, n = 1, 2, ...; and $A_n^*(0) \equiv A$, n = 0, 1, 2, ...

THEOREM 2.2. Let $k \ge 1$. (1) Let $A_{k-1}^*(i) \ne \emptyset$ for all $i \in S$, then

$$\sup_{a\in A_{k-1}^*(i)}\left\{R_k(i,a)+\sum_{j\in S}q(j|i,a)M_k^*(j)\right\}=M_k^*(i) \quad for \ all \quad i\in S.$$

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- (2) If $f(i) \in A_k^*(i)$ for all $i \in S$, then $f^{\infty} \in \bigcap_{i \in S} \prod(k, i)$.
- (3) Let $A_{k-1}^*(j) \neq \emptyset$ for all $j \in S$. Let $i \in S$, $\pi \in \Pi(k, i)$. If $\pi_0(a|i) > 0$, then $a \in A_k^*(i)$.
- (4) Let $A_{k-1}^*(j) \neq \emptyset$ for all $j \in S$. Let $i \in S, \pi \in \Pi(k, i)$. If $h_n = (i, a_0, i_1, a_1, \ldots, i_n) \in H_n (n \ge 1)$ is a realizable history under the policy π , then $\pi(\overline{h}_n) \in \Pi(k, i_n)$.

PROOF. (Apply induction to k). We know that proposition (Theorem 2.2) is true for k = 1 by Theorem 1.1, Corollary 1.1 and Corollary 1.2.

Inductive hypothesis I: the proposition (Theorem 2.2) is true for $1 \le k \le l - 1$.

(1) Let $A_{l-1}^*(i) \neq \emptyset$ for all $i \in S$. We take $f(i) \in A_{l-1}^*(i)$ for $\forall i \in S$. By the inductive hypothesis I and (2) in Theorem 2.2, $f^{\infty} \in \bigcap_{i \in S} \prod (l-1, i)$. So $\prod (l-1, i) \neq \emptyset$ for all $i \in S$.

For $\forall i \in S, \forall \pi \in \Pi(l-1, i)$, by Theorem 2.1,

$$M_{l}(\pi, i) = \sum_{a \in A} \pi_{0}(a|i) \left\{ (-1)^{l+1} R_{l}(i, a, \pi) + \sum_{j \in S} q(j|i, a) M_{l}(\pi(i, a), j) \right\}.$$

By the inductive hypothesis I and (4) in Theorem 2.2, $\pi(i, a) \in \Pi(l-1, j)$ when $\pi_0(a|i)q(j|i, a) > 0$. So

$$\begin{split} &\sum_{\substack{a \in A \\ \pi_0(a|i) > 0}} \pi_0(a|i)(-1)^{l+1} R_l(i, a, \pi) \\ &= \sum_{\substack{a \in A \\ \pi_0(a|i) > 0}} \pi_0(a|i)(-1)^{l+1} \sum_{p=0}^{l-1} C_l^p r^{l-p}(i, a) \sum_{\substack{j \in S \\ q(j|i,a) > 0}} q(j|i, a) M_p(\pi(i, a), j)(-1)^{p+1} \\ &= \sum_{\substack{a \in A \\ \pi_0(a|i) > 0}} \pi_0(a|i)(-1)^{l+1} \sum_{p=0}^{l-1} C_l^p (-1)^{p+1} r^{l-p}(i, a) \sum_{\substack{j \in S \\ q(j|i,a) > 0}} q(j|i, a) M_p^*(j) \\ &= \sum_{\substack{a \in A \\ \pi_0(a|i) > 0}} \pi_0(a|i) R_l(i, a), \end{split}$$

and

$$\sum_{\substack{a \in A \\ \pi_0(a|i) > 0 \\ q(j|i,a) > 0}} \pi_0(a|i) \sum_{\substack{j \in S \\ q(j|i,a) > 0 \\ q(j|i,a) > 0}} q(j|i,a) M_l(\pi(i,a), j) \leq \sum_{\substack{a \in A \\ \pi_0(a|i) > 0 \\ \pi_0(a|i) > 0 \\ q(j|i,a) > 0}} \pi_0(j|i,a) M_l^*(j).$$

That is,

$$M_l(\pi,i) \leq \sum_{a \in A} \pi_0(a|i) \left\{ R_l(i,a) + \sum_{j \in S} q(j|i,a) M_l^*(j) \right\}.$$

By the inductive hypothesis I and (3) in Theorem 2.2, $a \in A_{l-1}^*(i)$ when $\pi_0(a|i) > 0$. Therefore we have

$$M_l(\pi, i) \leq \sup_{a \in A_{l-1}^*(i)} \left\{ R_l(i, a) + \sum_{j \in S} q(j|i, a) M_l^*(j) \right\}.$$

By definition,

$$M_{l}^{*}(i) \leq \sup_{a \in A_{l-1}^{*}(i)} \left\{ R_{l}(i,a) + \sum_{j \in S} q(j|i,a) M_{l}^{*}(j) \right\}, \quad i \in S.$$
 (2.3)

For each $\epsilon > 0$, we take $f(i) \in A_{l-1}^*(i)$ for $\forall i \in S$ such that

$$R_{l}(i,f(i)) + \sum_{j \in S} q(j|i,f(i)) M_{l}^{*}(j) \geq \sup_{a \in A_{l-1}^{*}(i)} \left\{ R_{l}(i,a) + \sum_{j \in S} q(j|i,a) M_{l}^{*}(j) \right\} - \frac{\epsilon \alpha}{N}$$
$$\geq M_{l}^{*}(i) - \frac{\epsilon \alpha}{N}.$$
(2.4)

PROPOSITION A1. Let $i \in S_0$. Then

$$\sum_{n=0}^{m-1} \sum_{i_n \in S_0} P_{f^{\infty}} \{ X_n = i_n | X_0 = i \} R_l(i_n, f(i_n)) + \sum_{i_m \in S_0} P_{f^{\infty}} \{ X_m = i_m | X_0 = i \} M_l^*(i_m)$$

$$\geq M_l^*(i) - \frac{\epsilon \alpha}{N} \sum_{n=0}^{m-1} \sum_{i_n \in S_0} P_{f^{\infty}} \{ X_n = i_n | X_0 = i \}, \qquad m = 1, 2, \dots.$$

PROOF OF PROPOSITION A1. This follows immediately on applying induction to m (or see the proof of (2.2) in [11]).

PROPOSITION A2. If $g(i) \in A_{l-1}^*(i)$ for all $i \in S$, then

$$M_{l}(g^{\infty}, i) = \sum_{n=0}^{m-1} \sum_{i_{n} \in S} P_{g^{\infty}} \{X_{n} = i_{n} | X_{0} = i\} R_{l}(i_{n}, g(i_{n}))$$

+
$$\sum_{i_{m} \in S} P_{g^{\infty}} \{X_{m} = i_{m} | X_{0} = i\} M_{l}(g^{\infty}, i_{m}) \qquad i \in S, \quad m = 1, 2, \dots.$$

PROOF OF PROPOSITION A2. By inductive hypothesis I and (2) in Theorem 2.2, $g^{\infty} \in \bigcap_{i \in S} \prod(l-1, i)$. By Theorem 2.1,

$$M_{l}(g^{\infty}, i) = (-1)^{l+1} R_{l}(i, g(i), g^{\infty}) + \sum_{j \in S} q(j|i, g(i)) M_{l}(g^{\infty}, j)$$

= $R_{l}(i, g(i)) + \sum_{j \in S} q(j|i, g(i)) M_{l}(g^{\infty}, j), \quad i \in S.$ (2.5)

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By (2.5), we can prove that Proposition A2 is true by applying induction to m. By Propositions A1, A2 and Lemma 1.2,

$$M_{l}(f^{\infty}, i) \geq M_{l}^{*}(i) - \epsilon + \sum_{i_{m} \in S_{0}} P_{f^{\infty}} \{X_{m} = i_{m} | X_{0} = i\} (M_{l}(f^{\infty}, i_{m}) - M_{l}^{*}(i_{m})),$$

$$i \in S_{0}, \quad m = 1, 2, \dots$$

By Lemma 1.1 and (2.1), (2.2)

$$M_l(f^{\infty}, i) \ge M_l^*(i) - \epsilon - 2(1-\alpha)^{[m/N]}(2M)^l D(\alpha, N, l), \quad i \in S_0, m = N, N+1, \dots$$

Let $m \to \infty$. We have $M_l(f^{\infty}, i) \ge M_l^*(i) - \epsilon, i \in S_0$. So, by (2.5), (2.4)

$$\begin{split} M_{l}^{*}(i) &\geq M_{l}(f^{\infty}, i) = R_{l}(i, f(i)) + \sum_{j \in S} q(j|i, f(i)) M_{l}(f^{\infty}, j) \\ &\geq R_{l}(i, f(i)) + \sum_{j \in S} q(j|i, f(i)) [M_{l}^{*}(j) - \epsilon] \\ &= R_{l}(i, f(i)) + \sum_{j \in S} q(j|i, f(i)) M_{l}^{*}(j) - \epsilon \\ &\geq \sup_{a \in A_{l-1}^{*}(i)} \left\{ R_{l}(i, a) + \sum_{j \in S} q(j|i, a) M_{l}^{*}(j) \right\} - \frac{\epsilon \alpha}{N} - \epsilon, \quad i \in S. \end{split}$$

That is,

$$M_l^*(i) \geq \sup_{a \in A_{l-1}^*(i)} \left\{ R_l(i,a) + \sum_{j \in S} q(j|i,a) M_l^*(j) \right\} - \frac{\epsilon \alpha}{N} - \epsilon, \qquad i \in S.$$

If we let $\epsilon \to 0$, we see that (1) is true for k = l combining (2.3).

(2) Let $f(i) \in A_l^*(i)$ for all $i \in S$. Obviously $f(i) \in A_{l-1}^*(i)$ for all $i \in S$. By the definition of $A_l^*(i)$,

$$R_{l}(i,f(i)) + \sum_{j \in S} q(j|i,f(i))M_{l}^{*}(j) = \sup_{a \in A_{l-1}^{*}(i)} \left\{ R_{l}(i,a) + \sum_{j \in S} q(j|i,a)M_{l}^{*}(j) \right\}, \quad i \in S.$$

We have from the above proof of (1) that

$$M_l(f^{\infty}, i) \ge M_l^*(i), \qquad i \in S.$$
(2.6)

By inductive hypothesis I and (2) in Theorem 2.2, $f^{\infty} \in \bigcap_{i \in S} \Pi(l-1, i)$. So $M_l(f^{\infty}, i) \leq M_l^*(i), i \in S$. From (2.6) we have $f^{\infty} \in \bigcap_{i \in S} \Pi(l, i)$, that is, (2) is true for k = l.

(3) Let $A_{l-1}^*(j) \neq \emptyset$ for all $j \in S$. Let $i \in S, \pi \in \Pi(l, i)$. Obviously $\pi \in \Pi(l-1, i)$. By inductive hypothesis I and (3) in Theorem 2.2, $a \in A_{l-1}^*(i)$ when $\pi_0(a|i) > 0$. So

$$\pi_{0}(a|i)\left\{R_{l}(i,a) + \sum_{j\in S}q(j|i,a)M_{l}^{*}(j)\right\} \leq \pi_{0}(a|i)\sup_{\overline{a}\in A_{l-1}^{*}(i)}\left\{R_{l}(i,\overline{a}) + \sum_{j\in S}q(j|i,\overline{a})M_{l}^{*}(j)\right\}, \quad a\in A.$$
(2.7)

We know from the above proof of (1) that

$$\begin{split} M_{l}^{*}(i) &= M_{l}(\pi, i) \leq \sum_{a \in A} \pi_{0}(a|i) \left\{ R_{l}(i, a) + \sum_{j \in S} q(j|i, a) M_{l}^{*}(j) \right\} \\ &\leq \sum_{a \in A} \pi_{0}(a|i) \sup_{\bar{a} \in A_{l-1}^{*}(i)} \left\{ R_{l}(i, \bar{a}) + \sum_{j \in S} q(j|i, \bar{a}) M_{l}^{*}(j) \right\} \\ &= \sup_{a \in A_{l-1}^{*}(i)} \left\{ R_{l}(i, a) + \sum_{j \in S} q(j|i, a) M_{l}^{*}(j) \right\} = M_{l}^{*}(i). \end{split}$$

So

$$\sum_{a \in A} \pi_0(a|i) \left\{ R_l(i,a) + \sum_{j \in S} q(j|i,a) M_l^*(j) \right\}$$

= $\sum_{a \in A} \pi_0(a|i) \sup_{\overline{a} \in A_{l-1}^*(i)} \left\{ R_l(i,\overline{a}) + \sum_{j \in S} q(j|i,\overline{a}) M_l^*(j) \right\}.$ (2.8)

By (2.8) and (2.7),

$$\pi_0(a|i) \left\{ R_l(i,a) + \sum_{j \in S} q(j|i,a) M_l^*(j) \right\}$$
$$= \pi_0(a|i) \sup_{\overline{a} \in A_{l-1}^*(i)} \left\{ R_l(i,\overline{a}) + \sum_{j \in S} q(j|i,\overline{a}) M_l^*(j) \right\}, \qquad a \in A.$$

Therefore, when $\pi_0(a|i) > 0$, we have $a \in A^*_{l-1}(i)$ and

$$R_l(i,a) + \sum_{j \in S} q(j|i,a) M_l^*(j) = \sup_{\overline{a} \in A_{l-1}^*(i)} \left\{ R_l(i,\overline{a}) + \sum_{j \in S} q(j|i,\overline{a}) M_l^*(j) \right\},$$

that is, $a \in A_l^*(i)$. So (3) is true for k = l.

(4) Let $A_{l-1}^*(j) \neq \emptyset$ for all $j \in S$. Let $i \in S, \pi \in \Pi(l, i)$ and $h_n = (i, a_0, i_1, a_1, \ldots, i_n)$ $(n \ge 1)$ be a realizable history under the policy π . We shall prove that $\pi(\overline{h_n}) \in \Pi(l, i_n)$.

(Applying induction to *n*). Let n = 1 and $h_1 = (i, a_0, i_1)$ be a realizable history under the policy π . Obviously $\pi \in \Pi(l-1, i)$. We have from the above proofs of (1) and (3),

$$\begin{split} M_{l}^{*}(i) &= M_{l}(\pi, i) = \sum_{a \in A} \pi_{0}(a|i) \left\{ R_{l}(i, a) + \sum_{j \in S} q(j|i, a) M_{l}(\pi(i, a), j) \right\} \\ &\leq \sum_{a \in A} \pi_{0}(a|i) \left\{ R_{l}(i, a) + \sum_{j \in S} q(j|i, a) M_{l}^{*}(j) \right\} \\ &\leq M_{l}^{*}(i). \end{split}$$

Therefore

$$\sum_{a \in A} \pi_0(a|i) \sum_{j \in S} q(j|i, a) M_l(\pi(i, a), j) = \sum_{a \in A} \pi_0(a|i) \sum_{j \in S} q(j|i, a) M_l^*(j).$$
(2.9)

By inductive hypothesis I and (4) in Theorem 2.2, $\pi(i, a) \in \Pi(l - 1, j)$ when $\pi_0(a|i)q(j|i, a) > 0$. So

$$\pi_0(a|i)q(j|i,a)M_l(\pi(i,a),j) \le \pi_0(a|i)q(j|i,a)M_l^*(j), \quad a \in A, j \in S.$$
(2.10)

By (2.9) and (2.10),

$$\pi_0(a|i)q(j|i,a)M_l(\pi(i,a),j) = \pi_0(a|i)q(j|i,a)M_l^*(j), \ a \in A, j \in S$$

So, when $\pi_0(a_0|i)q(i_1|i, a_0) > 0$, we have $\pi(i, a_0) \in \Pi(l-1, i_1)$ and $M_l(\pi(i, a_0), i_1) = M_l^*(i_1)$, that is, $\pi(\overline{h}_1) \in \Pi(l, i_1)$. The proposition is true for n = 1.

Suppose the proposition is true for *n*. Let $h_{n+1} = (i, a_0, i_1, a_1, \dots, i_{n+1})$ be a realizable history under the policy π . It is easy to see that $h_n = (i, a_0, i_1, a_1, \dots, i_n)$ is also a realizable history under the policy π . By the supposition that $\pi(\overline{h_n}) \in \Pi(l, i_n)$, it is also easy to see that $\pi_n(a_n|h_n)q(i_{n+1}|i_n, a_n) > 0$, that is, (i_n, a_n, i_{n+1}) is a realizable history under the policy $\pi(\overline{h_n})$. Applying the result for n = 1, we have $\pi(\overline{h_{n+1}}) = \pi(\overline{h_n})(i_n, a_n) \in \Pi(l, i_{n+1})$, that is, the proposition is also true for n + 1. So (4) is true for k = l.

COROLLARY 2.1. Let $k \ge 1$, $A_{k-1}^*(j) \ne \emptyset$ for all $j \in S$. Let $i \in S$, $\Pi(k, i) \ne \emptyset$. Then $A_k^*(i) \ne \emptyset$.

PROOF. This follows immediately from Theorem 2.2(3).

COROLLARY 2.2. Let $k \ge 1$. If $A_k^*(i) \ne \emptyset$ for all $i \in S$, then $\bigcap_{i \in S} \Pi(k, i) \ne \emptyset$.

PROOF. This follows immediately from Theorem 2.2(2).

COROLLARY 2.3. Let $n \ge 1$. Then

$$\Pi(n, j) \neq \emptyset \text{ for all } j \in S \iff A_n^*(j) \neq \emptyset \text{ for all } j \in S.$$

PROOF. (\Leftarrow) This follows immediately from Corollary 2.2.

 (\Rightarrow) (Apply induction to *n*). The proposition is true for n = 1 by Corollary 2.1.

Suppose it is true for *n*. Let $\Pi(n+1, j) \neq \emptyset$ for all $j \in S$. Obviously $\Pi(n, j) \neq \emptyset$ for all $j \in S$. So $A_n^*(j) \neq \emptyset$ for all $j \in S$. By Corollary 2.1, $A_{n+1}^*(j) \neq \emptyset$ for all $j \in S$. That is, the proposition is also true for n + 1.

THEOREM 2.3. Let $k \ge 0$, $A_k^*(i) \ne \emptyset$ for all $i \in S$. Then $\forall \epsilon > 0$, $\exists f^{\infty}$ such that $f(i) \in A_k^*(i)$ for all $i \in S$ and

$$M_{k+1}(f^{\infty}, i) \ge M_{k+1}^*(i) - \epsilon, \qquad i \in S.$$

PROOF. The case for k = 0 corresponds to Theorem 2.2 in [11]. We know that the proposition is true for $k \ge 1$ from the proof of Theorem 2.2(1).

THEOREM 2.4. Let $k \ge 1$, $A_{k-1}^*(j) \ne \emptyset$ for all $j \in S$. Let $i \in S$. Then $\pi \in \Pi(k, i) \iff \forall n \ge 0$, if $h_n = (i, a_0, i_1, a_1, \dots, i_n)$ is a realizable history under the policy π and $\pi_n(a|h_n) > 0$, then $a \in A_k^*(i_n)$.

PROOF. (\Rightarrow) Let $n \ge 1$. By Theorem 2.2(4), $\pi(\overline{h}_n) \in \Pi(k, i_n)$. Let $\pi(\overline{h}_n) = (\pi'_0, \pi'_1, \pi'_2, \ldots)$. It is easy to see that $\pi'_0(a|i_n) = \pi_n(a|h_n)$, $a \in A$. By Theorem 2.2(3), $a \in A_k^*(i_n)$ when $\pi_n(a|h_n) > 0$.

Let n = 0. By Theorem 2.2(3), $a \in A_k^*(i)$ when $\pi_0(a|i) > 0$.

(\Leftarrow) (Apply induction to k). The proposition is true for k = 1 by Theorem 1.2. Suppose the proposition is true for $1 \le k \le l - 1$. We consider the case that k = l.

Let $A_{l-1}^*(j) \neq \emptyset$ for all $j \in S$ and let $i \in S$. By the inductive hypothesis and the sufficiency supposition, $\pi \in \Pi(l-1, i)$. We have from the proof of Theorem 2.2(1) that

$$M_{l}(\pi, i) = \sum_{a \in A} \pi_{0}(a|i) \left\{ R_{l}(i, a) + \sum_{j \in S} q(j|i, a) M_{l}(\pi(i, a), j) \right\}.$$
 (2.11)

Let $m \ge 0$. By Theorem 2.2(4), when $P_{\pi}\{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \dots, X_{m+1} = i_{m+1} | X_0 = i \} > 0$, we have $\pi(i, a_0, i_1, a_1, \dots, i_m, a_m) \in \Pi(l-1, i_{m+1})$. So, by (2.11),

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when $P_{\pi}\{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \dots, X_{m+1} = i_{m+1} | X_0 = i \} > 0$, we have

$$M_{l}(\pi(i, a_{0}, i_{1}, a_{1}, \dots, i_{m}, a_{m}), i_{m+1})$$

$$= \sum_{a_{m+1} \in A} \pi_{m+1}(a_{m+1}|i, a_{0}, i_{1}, a_{1}, \dots, i_{m+1}) \left\{ R_{l}(i_{m+1}, a_{m+1}) + \sum_{i_{m+2} \in S} q(i_{m+2}|i_{m+1}, a_{m+1}) M_{l}(\pi(i, a_{0}, i_{1}, a_{1}, \dots, i_{m}, a_{m})(i_{m+1}, a_{m+1}), i_{m+2}) \right\}.$$

Therefore, we have

$$\sum_{\substack{a_{0} \in A, i_{1} \in S, \\a_{1} \in A, \dots, i_{m+1} \in S}} P_{\pi} \{\Delta_{0} = a_{0}, X_{1} = i_{1}, \Delta_{1} = a_{1}, \dots, X_{m+1} = i_{m+1} | X_{0} = i \} \times M_{l}(\pi(i, a_{0}, i_{1}, a_{1}, \dots, i_{m}, a_{m}), i_{m+1})$$

$$= \sum_{\substack{a_{0} \in A, i_{1} \in S, \\a_{1} \in A, \dots, i_{m+1} \in S}} P_{\pi} \{\Delta_{0} = a_{0}, X_{1} = i_{1}, \Delta_{1} = a_{1}, \dots, X_{m+1} = i_{m+1} | X_{0} = i \} \times \left\{ \sum_{\substack{a_{0} \in A, i_{1} \in S, \\a_{1} \in A, \dots, i_{m+1} \in S}} \pi_{m+1}(a_{m+1} | i, a_{0}, i_{1}, a_{1}, \dots, i_{m+1}) R_{l}(i_{m+1}, a_{m+1}) + \sum_{\substack{a_{m+1} \in A, \\i_{m+2} \in S}} \pi_{m+1}(a_{m+1} | i, a_{0}, i_{1}, a_{1}, \dots, i_{m+1}) q(i_{m+2} | i_{m+1}, a_{m+1}) \times M_{l}(\pi(i, a_{0}, i_{1}, a_{1}, \dots, i_{m}, a_{m}, i_{m+1}, a_{m+1}), i_{m+2}) \right\}$$

$$= \sum_{\substack{i_{m+1} \in A \\a_{m+1} \in A}} P_{\pi} \{X_{m+1} = i_{m+1}, \Delta_{m+1} = a_{m+1} | X_{0} = i \} R_{l}(i_{m+1}, a_{m+1}) + \sum_{\substack{a_{0} \in A, i_{1} \in S, \\a_{1} \in A, \dots, i_{m+2} \in S}} P_{\pi} \{\Delta_{0} = a_{0}, X_{1} = i_{1}, \Delta_{1} = a_{1}, \dots, X_{m+2} = i_{m+2} | X_{0} = i \} \times M_{l}(\pi(i, a_{0}, i_{1}, a_{1}, \dots, i_{m+1}, a_{m+1}), i_{m+2}), m \ge 0.$$

$$(2.12)$$

By (2.11) and (2.12), it is easy to prove by induction that

$$M_{l}(\pi, i) = \sum_{n=0}^{m} \sum_{\substack{i_{n} \in S, a_{n} \in A \\ a_{1} \in A, \dots, i_{m+1} \in S}} P_{\pi} \{X_{n} = i_{n}, \Delta_{n} = a_{n} | X_{0} = i\} R_{l}(i_{n}, a_{n})$$

$$+ \sum_{\substack{a_{0} \in A, i_{1} \in S \\ a_{1} \in A, \dots, i_{m+1} \in S}} P_{\pi} \{\Delta_{0} = a_{0}, X_{1} = i_{1}, \Delta_{1} = a_{1}, \dots, X_{m+1} = i_{m+1} | X_{0} = i\} \times M_{l}(\pi(i, a_{0}, i_{1}, a_{1}, \dots, i_{m}, a_{m}), i_{m+1}), \qquad m = 0, 1, 2, \dots$$

By the sufficiency supposition, $a \in A_l^*(i)$ when $\pi_0(a|i) > 0$. So, by Theorem 2.2(1),

$$M_{l}^{*}(i) = \sum_{a \in A} \pi_{0}(a|i) \left\{ R_{l}(i,a) + \sum_{j \in S} q(j|i,a) M_{l}^{*}(j) \right\},$$
(2.13)

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Let $m \ge 0$. By the sufficiency supposition, when $P_{\pi}\{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \ldots, X_{m+1} = i_{m+1} | X_0 = i \} > 0$, if $\pi_{m+1}(a_{m+1}|i, a_0, i_1, a_1, \ldots, i_{m+1}) > 0$, then $a_{m+1} \in A_l^*(i_{m+1})$. So, by Theorem 2.2(1), when $P_{\pi}\{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \ldots, X_{m+1} = i_{m+1} | X_0 = i \} > 0$, we have

$$M_l^*(i_{m+1}) = \sum_{a_{m+1} \in A} \pi_{m+1}(a_{m+1}|i, a_0, i_1, a_1, \dots, i_{m+1}) \left\{ R_l(i_{m+1}, a_{m+1}) + \sum_{j \in S} q(j|i_{m+1}, a_{m+1}) M_l^*(j) \right\}.$$

Therefore, we have

 $\sum_{\substack{a_0 \in A, i_1 \in S, \\ a_1 \in A, \dots, i_{m+1} \in S}} P_{\pi} \{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \dots, X_{m+1} = i_{m+1} | X_0 = i\} M_i^*(i_{m+1})$ $= \sum_{\substack{a_0 \in A, i_1 \in S, \\ a_1 \in A, \dots, i_{m+1} \in S}} P_{\pi} \{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \dots, X_{m+1} = i_{m+1} | X_0 = i\}$ $+ \sum_{\substack{a_{m+1} \in A}} \pi_{m+1}(a_{m+1} | i, a_0, i_1, a_1, \dots, i_{m+1}) \left\{ R_i(i_{m+1}, a_{m+1}) + \sum_{\substack{j \in S}} q(j | i_{m+1}, a_{m+1}) M_i^*(j) \right\}$ $= \sum_{\substack{i_{m+1} \in S \\ a_{m+1} \in A}} P_{\pi} \{X_{m+1} = i_{m+1}, \Delta_{m+1} = a_{m+1} | X_0 = i\} R_i(i_{m+1}, a_{m+1})$ $+ \sum_{\substack{a_0 \in A, i_1 \in S, \\ a_1 \in A, \dots, i_{m+2} \in S}} P_{\pi} \{\Delta_0 = a_0, X_1 = i_1, \Delta_1 = a_1, \dots, X_{m+2} = i_{m+2} | X_0 = i\} M_i^*(i_{m+2}),$ $m \ge 0.$ (2.14)

By (2.13) and (2.14), it is easy to prove by induction that

$$M_{l}^{*}(i) = \sum_{n=0}^{m} \sum_{\substack{i_{n} \in S, a_{n} \in A \\ a_{0} \in A, i_{1} \in S, \\ a_{1} \in A, \dots, i_{m+1} \in S}} P_{\pi} \{X_{n} = i_{n}, \Delta_{n} = a_{n} | X_{0} = i\} R_{l}(i_{n}, a_{n})$$

$$+ \sum_{\substack{a_{0} \in A, i_{1} \in S, \\ a_{1} \in A, \dots, i_{m+1} \in S}} P_{\pi} \{\Delta_{0} = a_{0}, X_{1} = i_{1}, \Delta_{1} = a_{1}, \dots, X_{m+1} = i_{m+1} | X_{0} = i\} M_{l}^{*}(i_{m+1}),$$

$$m=0, 1, 2, \ldots$$

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So, when $i \in S_0$, by (2.1), (2.2) and Lemma 1.1,

$$\begin{split} |M_{l}(\pi, i) - M_{l}^{*}(i)| &\leq \sum_{\substack{a_{0} \in A, i_{1} \in S, \\ a_{1} \in A, \dots, i_{m} \in S, a_{m} \in A, i_{m+1} \in S_{0}}} P_{\pi} \{\Delta_{0} = a_{0}, X_{1} = i_{1}, \Delta_{1} = a_{1}, \dots, \\ X_{m+1} = i_{m+1} | X_{0} = i \} 2(2M)^{l} D(\alpha, N, l) \\ &= \sum_{i_{m+1} \in S_{0}} P_{\pi} \{X_{m+1} = i_{m+1} | X_{0} = i \} 2(2M)^{l} D(\alpha, N, l) \\ &\leq (1 - \alpha)^{[m+1/N]} 2(2M)^{l} D(\alpha, N, l), \qquad m = N, N + 1, \dots. \end{split}$$

Let $m \to \infty$. We have $M_l(\pi, i) = M_l^*(i)$. So $\pi \in \Pi(l, i)$ (if i = 0, then $\pi \in \Pi = \Pi(l, 0)$ obviously). The proposition is also true for k = l.

Obviously Theorem 2.4 is an extension of Theorem 1.2.

THEOREM 2.5. Let $k \ge 0$. Then $\Pi(k) = \bigcap_{i \in S} \Pi(k, i)$.

PROOF. (Apply induction to k.) The proposition is true for k = 0 obviously. Suppose the proposition is true for $0 \le k \le l - 1$.

Let $\pi \in \Pi(l)$. It is easy to see that $\pi \in \Pi(l-1)$. By the inductive hypothesis, $\pi \in \bigcap_{i \in S} \Pi(l-1, i)$. By Corollary 2.3, $A_{l-1}^*(i) \neq \emptyset$ for all $i \in S$. By Theorem 2.3, $\forall \epsilon > 0, \exists f^{\infty}$ such that $f(i) \in A_{l-1}^*(i)$ for all $i \in S$ and

$$M_l(f^{\infty}, i) \ge M_l^*(i) - \epsilon, \qquad i \in S.$$

By Theorem 2.2(2) and the inductive hypothesis, $f^{\infty} \in \Pi(l-1)$. Since π , $f^{\infty} \in \Pi(l-1)$, therefore $M^{l-1}(\pi) = M^{l-1}(f^{\infty})$. Since $\pi \in \Pi(l)$, therefore $M^{l}(\pi) \ge M^{l}(f^{\infty})$. Hence $M_{l}(\pi, i) \ge M_{l}(f^{\infty}, i)$ for all $i \in S$, that is,

$$M_l(\pi, i) \ge M_l^*(i) - \epsilon, \qquad i \in S.$$

Let $\epsilon \to 0$. We have $M_l(\pi, i) = M_l^*(i)$ for all $i \in S$. So $\pi \in \bigcap_{i \in S} \Pi(l, i)$, that is, $\Pi(l) \subset \bigcap_{i \in S} \Pi(l, i)$.

Let $\pi \in \bigcap_{i \in S} \prod(l, i)$. It is easy to see that $\pi \in \bigcap_{i \in S} \prod(l-1, i)$. By the inductive hypothesis, $\pi \in \prod(l-1)$. Choose any $\tilde{\pi} \in \prod$. Obviously $M^{l-1}(\pi) \ge M^{l-1}(\tilde{\pi})$. If $M^{l-1}(\pi) > M^{l-1}(\tilde{\pi})$, then

$$M^{l}(\pi) > M^{l}(\tilde{\pi}). \tag{2.15}$$

If $M^{l-1}(\pi) = M^{l-1}(\tilde{\pi})$, then $\tilde{\pi} \in \Pi(l-1)$. By the inductive hypothesis, $\tilde{\pi} \in \bigcap_{i \in S} \Pi(l-1, i)$. Since $\pi \in \bigcap_{i \in S} \Pi(l, i)$, we have $M_l(\pi) \ge M_l(\tilde{\pi})$. Hence

$$M^{l}(\pi) \ge M^{l}(\tilde{\pi}). \tag{2.16}$$

By (2.15) and (2.16), $M^{l}(\pi) \ge M^{l}(\tilde{\pi})$. Therefore $\pi \in \Pi(l)$, that is, $\bigcap_{i \in S} \Pi(l, i) \subset \Pi(l)$. To sum up, we know that the proposition is true for k = l.

THEOREM 2.6. Let $k \ge 1$. Then $\pi \in \Pi(k) \iff \forall n \ge 0$ if $h_n = (i_0, a_0, i_1, a_1, \dots, i_n)$ is a realizable history under the policy π and $\pi_n(a|h_n) > 0$, then $a \in A_k^*(i_n)$.

PROOF. (\Rightarrow) Let $\pi \in \Pi(k)$. By Theorem 2.5, $\pi \in \bigcap_{i \in S} \Pi(k, i)$. By Corollary 2.3, $A_k^*(i) \neq \emptyset$ for all $i \in S$. Obviously $\pi \in \Pi(k, i_0)$. By Theorem 2.4, if $h_n = (i_0, a_0, i_1, a_1, \dots, i_n)$ $(n \ge 0)$ is a realizable history under the policy π and $\pi_n(a|h_n) > 0$, then $a \in A_k^*(i_n)$.

(\Leftarrow) Choose any $i \in S$. We take $a \in A$ such that $\pi_0(a|i) > 0$. By the sufficiency supposition, $a \in A_k^*(i)$. So $A_k^*(j) \neq \emptyset$ for all $j \in S$. By the sufficiency supposition and Theorem 2.4, $\pi \in \Pi(k, i)$ for all $i \in S$. By Theorem 2.5, $\pi \in \Pi(k)$.

Obviously this theorem is an extension of Theorem 2.4 in [11].

COROLLARY 2.4. $\pi \in \Pi(\infty) \iff \forall n \ge 0$, if $h_n = (i_0, a_0, i_1, a_1, \dots, i_n)$ is a realizable history under the policy π and $\pi_n(a|h_n) > 0$, then $a \in \bigcap_{n=1}^{\infty} A_k^*(i_n)$.

PROOF. This follows immediately from Theorem 2.6.

THEOREM 2.7. (1) Let $k \ge 1$. If $\Pi(k) \ne \emptyset$, then $\exists f^{\infty} \in \Pi(k)$. (2) If $\Pi(\infty) \ne \emptyset$, then $\exists f^{\infty} \in \Pi(\infty)$.

PROOF. (1) By Theorem 2.5 and Corollary 2.3, $A_k^*(i) \neq \emptyset$ for all $i \in S$. We take $f(i) \in A_k^*(i)$ for all $i \in S$. By Theorem 2.2(2) and Theorem 2.5, $f^{\infty} \in \Pi(k)$.

(2) We take $\pi \in \Pi(\infty)$ and $\forall i \in S$ take $a \in A$ such that $\pi_0(a|i) > 0$. By Corollary 2.4, $a \in \bigcap_{k=1}^{\infty} A_k^*(i)$. That is, $\bigcap_{k=1}^{\infty} A_k^*(i) \neq \emptyset$ for all $i \in S$. We take $f(i) \in \bigcap_{k=1}^{\infty} A_k^*(i)$ for all $i \in S$. By Corollary 2.4, $f^{\infty} \in \Pi(\infty)$.

THEOREM 2.8. (1) Let $k \ge 1$. If f^{∞} is a k-moment optimal policy in Π_s^d (that is, $M^k(f^{\infty}) \ge M^k(g^{\infty})$ for all $g^{\infty} \in \Pi_s^d$), then $f^{\infty} \in \Pi(k)$. (2) If f^{∞} is a moment optimal policy in Π_s^d , then $f^{\infty} \in \Pi(\infty)$.

PROOF. (1) (Apply induction to k.) The proposition is true for k = 1 by Theorem 1.3 and Theorem 2.5. Suppose the proposition is true for $1 \le k \le l - 1$.

Let f^{∞} be a *l*-moment optimal policy in Π_s^d . It is easy to see that f^{∞} is a (l-1)-moment optimal policy in Π_s^d . By the inductive hypothesis and Theorem 2.5,

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 $f^{\infty} \in \Pi(l-1) = \bigcap_{i \in S} \Pi(l-1, i)$. By Corollary 2.3, $A_{l-1}^{*}(i) \neq \emptyset$ for all $i \in S$. By Theorem 2.3, $\forall \epsilon > 0$, $\exists g^{\infty}$ such that $g(i) \in A_{l-1}^{*}(i)$ for all $i \in S$ and

$$M_l(g^{\infty}, i) \ge M_l^*(i) - \epsilon, \qquad i \in S.$$

By Theorem 2.2(2) and Theorem 2.5, $g^{\infty} \in \Pi(l-1)$. So $M^{l-1}(g^{\infty}) = M^{l-1}(f^{\infty})$. By the supposition, $M^{l}(f^{\infty}) \ge M^{l}(g^{\infty})$. So $M_{l}(f^{\infty}, i) \ge M_{l}(g^{\infty}, i), i \in S$. Hence

$$M_l(f^{\infty}, i) \ge M_l^*(i) - \epsilon, \qquad i \in S.$$

Let $\epsilon \to 0$. We have $M_l(f^{\infty}, i) = M_l^*(i), i \in S$. By Theorem 2.5, $f^{\infty} \in \bigcap_{i \in S} \Pi(l, i) = \Pi(l)$. That is, the proposition is true for k = l. The proof of (1) is complete.

(2) This follows immediately from (1).

Theorems 2.7 and 2.8 state that the problems of the existence and calculation of a k-moment optimal policy (or a moment optimal policy) in Π can be changed into the same problems in Π_s^d .

THEOREM 2.9. If A is nonempty and finite, then $\exists f^{\infty} \in \Pi(\infty)$.

PROOF. Let A be nonempty and finite. By the definition of $A_k^*(i)$ and Corollary 2.3, $A_k^*(i) \neq \emptyset$ for $\forall i \in S, \forall k \ge 1$. Because A is finite and $A_k^*(i) \subset A_{k-1}^*(i), i \in S, k \ge 1$, it is easy to see that $\bigcap_{k=1}^{\infty} A_k^*(i) \neq \emptyset$ for all $i \in S$. We take $f(i) \in \bigcap_{k=1}^{\infty} A_k^*(i)$ for all $i \in S$. By Corollary 2.4, $f^{\infty} \in \Pi(\infty)$.

THEOREM 2.10. For $k \ge 1$, let $f^{\infty} \in \Pi(k-1)$. If

$$M_k(f^{\infty},i) = \sup_{a \in A_{k-1}^*(i)} \left\{ R_k(i,a) + \sum_{j \in S} q(j|i,a) M_k(f^{\infty},j) \right\} \text{ for all } i \in S,$$

then $f^{\infty} \in \Pi(k)$.

PROOF. By Theorem 2.5 and Corollary 2.3, $A_{k-1}^*(i) \neq \emptyset$ for all $i \in S$. By Theorem 2.3, $\forall \epsilon > 0, \exists g^{\infty}$ such that $g(i) \in A_{k-1}^*(i)$ for all $i \in S$ and

$$M_k(g^{\infty}, i) \geq M_k^*(i) - \epsilon, \ i \in S.$$

By the supposition,

$$R_k(i,g(i)) + \sum_{j\in S} q(j|i,g(i)) M_k(f^{\infty},j) \le M_k(f^{\infty},i), \qquad i\in S.$$

Imitating the proof of Theorem 2.2(1), we have

$$M_k(f^{\infty}, i) \ge M_k(g^{\infty}, i), \qquad i \in S,$$

that is,

$$M_k(f^{\infty}, i) \ge M_k^*(i) - \epsilon, \qquad i \in S.$$

Let $\epsilon \to 0$. We have

$$M_k(f^{\infty},i) \geq M_k^*(i), \qquad i \in S.$$

By Theorem 2.5, $f^{\infty} \in \bigcap_{i \in S} \Pi(k-1, i)$. So, by Theorem 2.5, $f^{\infty} \in \bigcap_{i \in S} \Pi(k, i) = \Pi(k)$.

3. Algorithm

We shall now give an algorithm of policy-improvement type for finding a kmoment optimal stationary policy. In this section we suppose that S and A are finite. By Theorem 2.9, there exists a f^{∞} which is a moment-optimal policy. Obviously, f^{∞} is also a $k(\geq 1)$ -moment optimal policy.

THEOREM 3.1. Let $k \ge 1$, $f^{\infty} \in \Pi(k-1)$. The equation

$$R_k(i, f(i)) + \sum_{j \in S_0} q(j|i, f(i))V(j) = V(i), \qquad i \in S_0,$$
(3.1)

possesses a unique solution $V(i) = M_k(f^{\infty}, i), i \in S_0$.

PROOF. By Theorem 2.1 and 2.5, $\{M_k(f^{\infty}, i) : i \in S_0\}$ is a solution of (3.1). By Lemma 1.3, the solution of (3.1) is unique.

By solving (3.1), we can find $M_k(f^{\infty}, i), i \in S$.

THEOREM 3.2 (Policy improvement). For $k \ge 1$, let $f^{\infty} \in \Pi(k-1)$. If $g(i) \in A_{k-1}^{*}(i)$ for all $i \in S$ and

$$R_k(i, g(i)) + \sum_{j \in S} q(j|i, g(i)) M_k(f^{\infty}, j) \ge M_k(f^{\infty}, i) \text{ for all } i \in S,$$

then $M_k(g^{\infty}) \geq M_k(f^{\infty})$.

PROOF. The proof is similar to that of Theorem 2.2(1). Note that, by Theorem 2.5 and Corollary 2.3, $A_{k-1}^*(i) \neq \emptyset$ for all $i \in S$.

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Let $k \ge 1$. By Theorem 2.9, $\exists f^{\infty} \in \Pi(k-1)$. We take $f_0^{\infty} \in \Pi(k-1)$. By Theorem 2.5 and Corollary 2.3, $A_{k-1}^*(i) \ne \emptyset$ for all $i \in S$. $f_n^{\infty}(n = 1, 2, ...)$ is defined as follows: $\forall i \in S$, we take $f_n(i) \in A_{k-1}^*(i)$ such that

$$\max_{a \in A_{k-1}^{*}(i)} \left\{ R_{k}(i,a) + \sum_{j \in S} q(j|i,a) M_{k}(f_{n-1}^{\infty},j) \right\}$$

= $R_{k}(i,f_{n}(i)) + \sum_{j \in S} q(j|i,f_{n}(i)) M_{k}(f_{n-1}^{\infty},j).$ (3.2)

THEOREM 3.3. Let $k \ge 1$. For f_n^{∞} (n = 0, 1, 2, ...) defined above, we have

- (1) $M_k(f_n^{\infty}) \geq M_k(f_{n-1}^{\infty}), n = 1, 2, ...$
- (2) $\exists n_0 \geq 0$ such that $M_k(f_{n_0}^{\infty}) = M_k(f_{n_0+1}^{\infty})$.
- (3) If $M_k(f_{n_0}^{\infty}) = M_k(f_{n_0+1}^{\infty})$, then $f_{n_0}^{\infty} \in \Pi(k)$.
- PROOF. (1) By Theorem 2.2(2) and Theorem 2.5, $f_n^{\infty} \in \Pi(k-1), n \ge 0$. By Theorem 2.6, $f_n(i) \in A_{k-1}^*(i), i \in S, n \ge 0$. By Theorem 3.1 and 3.2, (1) is true.
- (2) Because S and A are finite, Π_s^d is finite. Condition (2) is true from (1).
- (3) From Theorem 3.1 and Theorem 2.10, (3) is true.

Let $k \ge 1$. An iteration algorithm for finding a k-moment optimal stationary policy is stated as follows:

- (1) $l \leftarrow 1$. Choose any $f_0^{\infty} \in \Pi_s^d$.
- (2) By (3.2), with the policy improvement iteration starting from f_0^{∞} (replace k by l in (3.2)), we can find $g^{\infty} \in \Pi(l)$ (see Theorem 3.3). By Theorem 2.5, $M_l(g^{\infty}, i) = M_l^*(i), i \in S$.
- (3) If l = k, then stop. We have $g^{\infty} \in \Pi(k)$. If l < k, then go to (4).
- (4) By the definition of $A_i^*(i)$, we find $A_i^*(i)$, $i \in S$. Obviously $A_i^*(i) \neq \emptyset$, $i \in S$.
- (5) $l \leftarrow l + 1$. Let $f_0 = g$. Go to (2).

By the above algorithm, we can find $A_k^*(i)$, $i \in S$, $k \ge 1$. We take $f(i) \in \bigcap_{k=1}^{\infty} A_k^*(i)$ for all $i \in S$, then $f^{\infty} \in \Pi(\infty)$ (see the proof of Theorem 2.9).

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