

Asymptotic Properties in Dynamic Programming

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Abstract: In the framework of dynamic programming we provide two results:

- An example where uniform convergence of the T -stage value does not imply equality of the limit and the lower infinite value.
- Generalized Tauberian theorems, that relate uniform convergence of the T -stage value to uniform convergence of values associated with a general distribution on stages.

1 Introduction

Let S be a state space. For each $s \in S$ let $\emptyset \neq \Gamma(s) \subseteq S$, and let f be a real bounded function on S . Consider the dynamic programming problem where the decision maker on day t , at stage s_t , has to choose a new state $s_{t+1} \in \Gamma(s_t)$, and receives a payoff $f(s_t)$. A *play* at $s \in S$ is a sequence $(s_t)_{t=0}^{\infty}$ with $s_0 = s$ and $s_{t+1} \in \Gamma(s_t)$ for all $t \geq 0$. One traditionally considers the λ -discounted value $V_\lambda(s)$:

$$V_\lambda(s) = \sup_{(s_t)_{t=0}^{\infty}} (1-\lambda) \sum_{t=0}^{\infty} \lambda^t f(s_t),$$

or the T -stage value $V_T(s)$:

$$V_T(s) = \sup_{(s_t)_{t=0}^{\infty}} \frac{1}{T+1} \sum_{t=0}^T f(s_t),$$

where in both cases the supremum ranges over all plays at s .

One can also consider other evaluations: Let $\theta = (\theta(t))_{t=0}^{\infty}$ be a probability on the set of non-negative integers and define:

$$V_\theta(s) = \sup_{(s_t)_{t=0}^{\infty}} \sum_{t=0}^{\infty} \theta(t) f(s_t).$$

Lehrer and Sorin (1992) proved that if either one of the limits $\lim_{\lambda \rightarrow 1} V_\lambda(s)$, or $\lim_{T \rightarrow \infty} V_T(s)$ exists uniformly in $s \in S$, then the other limit also exists uniformly, and the limit functions coincide.

In Section 3 we give sufficient conditions on linearly ordered families $(\Theta, <)$ of probabilities on the integers to get analogous results for $(V_\theta)_{\theta \in \Theta}$ and $(V_T)_{T \geq 0}$.

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There are other natural ways of evaluating streams of payoffs in dynamic programming (except for those discussed above):

The lower (long-run average) value,

$$\underline{V}(s) = \sup_{(s_t)_{t=0}^{\infty}} \liminf_{T \rightarrow \infty} \frac{1}{T+1} \sum_{t=0}^T f(s_t),$$

and the upper (long-run average) value,

$$\overline{V}(s) = \sup_{(s_t)_{t=0}^{\infty}} \limsup_{T \rightarrow \infty} \frac{1}{T+1} \sum_{t=0}^T f(s_t),$$

where, again, the supremum is taken on all plays at s .

Lehrer and Monderer (1989) proved that uniform convergence of $(V_\lambda)_{\lambda \in [0,1]}$ to some V implies $V = \overline{V}$, and showed in an example that it does not imply the equality $V = \underline{V}$. If one allows the decision maker to use mixed strategies, i.e., to choose a play in random, and then defines the payoff of each state as the expectation, one obtains new evaluations. It is clear that the evaluations V_λ , V_T , V_θ , and \overline{V} will not change by allowing mixed strategies, but \underline{V} will change in general. Let

$$\underline{U}(s) = \sup_{\mu \in \Delta} \liminf_{T \rightarrow \infty} E_\mu \left(\frac{1}{T+1} \sum_{t=0}^T f(s_t) \right),$$

where Δ is the set of all probabilities on the set of plays, endowed with the cylinder σ -field, and E_μ stands for the expectation operator with respect to μ .

Obviously $\underline{U} \geq \underline{V}$. As for the relationship between \underline{U} and the limit V of the discounted value functions, Mertens and Neyman (1981) provided sufficient conditions, stronger than the uniform convergence of $(V_\lambda)_{\lambda \in [0,1]}$ (and satisfied in every finite setup), that ensure the equality $\underline{U} = V$ (even for stochastic games). In Section 2 we show that uniform convergence alone is not sufficient by providing a counter example. See Mertens (1987) for related conjectures, hints, and comments. Other type of necessary conditions, for specific types of dynamic programming problems, are discussed in Dutta (1991).

2 The Counter Example

Every rooted directed tree without terminal nodes naturally defines a dynamic programming problem when we attach payoffs to the nodes. Our dynamic programming problem will be defined as a tree, constructed inductively in the spirit of Lehrer and Monderer (1989).

Given two decreasing vanishing sequences $(\varepsilon_n)_{n=1}^{\infty}$ and $(\delta_n)_{n=1}^{\infty}$, define for every real number x the tree $T(x)$ as follows:

Every node of $T(x)$ except for the root has an outdegree one, and the root itself has countably many branches. On the n^{th} branch of the root the payoff, $g(s)$, is 0 until node $[\varepsilon_n n] + 1$, it then equals $x - \delta_n$ until node n , and from then on it equals 0. Define a valuation φ at each node s different from the root as follows: $\varphi(s) = x - \delta_n$ for every s in the n^{th} branch appearing before the n^{th} node in this branch, and $\varphi(s) = 0$ for every node thereafter. Set $T_1 = T(1)$. T_2 is obtained from T_1 by attaching the tree $T(\varphi(s))$ to each node s of T_1 , different from the root, and keeping the old payoff of s (i.e., its payoff in T_1). One can continue naturally and define inductively the trees T_3, T_4, \dots and finally define $T = \bigcup_{n=1}^{\infty} T_n$. Denote the root of T by s_0 , and the payoff function by g .

Note that although g is bounded from above by 1, it is not necessarily bounded from below. Therefore we replace g with a new bounded payoff function f , defined by: $f(s) = \max(g(s), 0)$ for every node s of T .

It is clear that $\lim_{T \rightarrow \infty} V_T(s) = \varphi(s)$ uniformly on all nodes s of T . In particular $V(s_0) = 1$.

We will show that for a specific choice of the sequences $(\varepsilon_n)_{n=1}^{\infty}$ and $(\delta_n)_{n=1}^{\infty}$, $\underline{U}(s_0) = 0$.

Let then $\alpha > 0$ and let us prove that $\underline{U}(s_0) < \alpha$. Assume in negation that there exists $\mu \in \Delta$ such that for some integer M , $T \geq M$ implies

$$E_{\mu} \left(\frac{1}{T+1} \sum_{t=0}^T f(s_t) \right) \geq \alpha. \quad (2.1)$$

We remark that we can assume that all plays in the support of μ belong to the following set Ω :

If a play in Ω is on the n^{th} branch of some $T(\cdot)$, it remains in this branch until exactly node n . In fact, if some play leaves the branch before node $[\varepsilon_n n] + 1$, the decision maker will increase his payoff by leaving the branch at its root, and if a play leaves the n^{th} branch after node n , it is better for the decision maker to leave it at precisely node n . In particular, a play in Ω never remains in a branch of some $T(\cdot)$ and is thus characterized by a sequence $(m_i)_{i=1}^{\infty}$ of integers inducing the path: Branch m_1 of $T(1)$ until the m_1 th node, s_{m_1} , branch m_2 of $T(\varphi(s_{m_1}))$ until node m_2 of this branch (with the valuation $1 - \delta_{m_1} - \delta_{m_2}$), etc. ... Finally, for every play in Ω ,

$$\sum_{i=1}^{\infty} \delta_{m_i} \leq 1. \quad (2.2)$$

Having done the above reduction, we can now replace any strictly positive payoff on any play in Ω by 1.

The basic idea of the proof is to choose a sequence $(\delta_n)_{n=1}^{\infty}$ converging very slowly to zero, implying by (2.2), that for every play in Ω , for a set of integers i with positive density, $\varepsilon_{m_i} m_i$ is much larger than $\sum_{k < i} m_k$. Hence, every play in Ω has "many" large blocks of zeros.

More precisely, let $M_1 = 2$, and define inductively $n_i = \sum_{k \leq i} M_k$ and $M_{i+1} = n_i^4$ for every $i \geq 1$. Define $\delta_{n_i} = \frac{1}{i}$ for all i and extend δ by monotonicity to all other in-

tegers. Choose $\varepsilon_n = \frac{1}{\sqrt{n}}$. We say that a play w is good in the i^{th} block $I_i = [n_{i-1}, n_i]$ if a sequence of ones starts in this block. That is, if w is determined by m_1, m_2, \dots , there exists m_k adapted to I_i in the sense that

$$\sum_{j < k} m_j + \varepsilon_{m_k} m_k \in I_i. \quad (2.3)$$

Set $S_n(w) = \frac{1}{n} \sum_{k=1}^n w_k$. We claim that there exists i_0 such that for every $i > i_0$ and for every $w \in \Omega$, if $S_{n_i}(w) \geq \alpha$, then w is good in the i^{th} block. Otherwise, denote by k the largest integer such that the k^{th} sequence of ones in w starts before the i^{th} block. Then $\varepsilon_{m_k} m_k \leq n_{i-1}$, and hence $m_k \leq n_{i-1}^2 = \frac{1}{n_{i-1}^2} M_i$. This implies that this sequence of ones ends very early in the i^{th} block, and that $w_t = 0$ for $\left(1 - \frac{1}{n_{i-1}^2}\right) M_i$ t 's in this block. As $\frac{n_{i-1}}{M_i} \rightarrow 0$ as $i \rightarrow \infty$, then $S_{n_i}(w)$ must be very small contradicting our assumption.

Define $J_i(w)$ to be one if $S_{n_i}(w) \geq \alpha$ and 0 otherwise. If $J_i(w) = 1$, one can by the above claim, define $k(w, i)$ as the smallest k that satisfy (2.3). Denote $\theta_i(w) = \delta_{k(w, i)}$ if $J_i(w) = 1$ and 0 otherwise.

Using the monotone convergence theorem we have:

$$1 \geq E_\mu \left(\sum_{i \geq i_0} J_i(w) \theta_i(w) \right) \geq \sum_{i \geq i_0} E_\mu(J_i(w) \theta_i(w)) \geq \sum_{i \geq i_0} E_\mu(J_i(w)) \delta_{n_i}.$$

Since (2.1) at n_i implies $E_\mu(J_i(w)) \geq \alpha$, we obtain, recalling that $\delta_{n_i} = \frac{1}{i}$,

$$1 \geq \left(\sum_{i \geq i_0} \frac{1}{i} \right) \alpha,$$

a contradiction. ■

3 Uniform Convergence

We first establish a few notations. Let D denote the set of all probability distributions θ on the set $N = \{0, 1, 2, \dots\}$ of non-negative integers, that are non-increasing. That is,

$$\theta(t+1) \leq \theta(t) \quad \text{for all } t \in N. \quad (A)$$

For real numbers $\alpha \leq \beta$ and for a distribution θ ,

$$\theta[\alpha, \beta] = \sum_{\alpha \leq t \leq \beta} \theta(t).$$

For $\theta \in D$, define $\hat{\theta}$ on N as follows:

$$\hat{\theta}(t) = (\theta(t) - \theta(t+1))(t+1) \quad \text{for all } t \in N. \quad (3.1)$$

Note that

$$\sum_{t=0}^T \hat{\theta}(t) = \sum_{t=0}^T \theta(t) - (T+1)\theta(T+1) \quad \text{for all } T \geq 0. \quad (3.2)$$

Because of (A), $\lim_{t \rightarrow \infty} t\theta(t) = 0$, and therefore $\hat{\theta}$ is a probability distribution on N .

Let $a = (a_t)_{t=0}^{\infty}$ be a bounded sequence. For every $T \geq 0$, denote

$$S_T(a) = \frac{1}{T+1} \sum_{t=0}^T a_t,$$

and denote $S(a) = (S_t(a))_{t=0}^{\infty}$. For every probability θ , set,

$$S_{\theta}(a) = \sum_{t=0}^{\infty} \theta(t) a_t.$$

Observe that by (3.1), similarly to the way (3.2) was obtained, we have $S_{\theta}(a) = S_{\hat{\theta}}(S(a))$ for all sequences a and probabilities θ , that is,

$$\sum_{t=0}^{\infty} \theta(t) a_t = \sum_{t=0}^{\infty} \hat{\theta}(t) S_t(a). \quad (3.3)$$

We consider linearly ordered families $(\Theta, >)$, where $\Theta \subseteq D$, and “ $>$ ” is a linear (complete) order on Θ , satisfying:

$$\forall \varepsilon > 0, \forall N \geq 0, \exists \theta_0 \in \Theta, \text{ such that } \forall \theta > \theta_0, \sum_{t=0}^N \theta(t) < \varepsilon, \quad (B)$$

which is obviously equivalent to:

$$\forall \varepsilon > 0, \exists \theta_0 \in \Theta, \text{ such that } \forall \theta > \theta_0, \theta(0) < \varepsilon. \quad (B^*)$$

Note that Condition (B) implies that for every $\theta \in \Theta$, there exists $\hat{\theta} \in \Theta$, with $\theta < \hat{\theta}$. Therefore, the notions of \lim , $\lim \inf$, $\lim \sup$, etc. ... are naturally defined for real-valued function on Θ . An increasing sequence $(\theta_n)_{n=0}^{\infty}$ in Θ , is *increasing* to ∞ , if for every $\theta \in \Theta$, there exists an integer N such that $\theta_n > \theta$ for all $n \geq N$. For the equivalence results we will need the next properties:

(C) $\exists \varepsilon_0 > 0$ and $\varphi: (0, \varepsilon_0) \rightarrow (0, 1)$ such that $\forall \varepsilon < \varepsilon_0$, $\exists J(\varepsilon)$, and a sequence $(\theta_{n,\varepsilon})_{n=J(\varepsilon)}^{\infty}$, that increases to ∞ and satisfies:

$$\hat{\theta}_{n,\varepsilon}[(1-\varepsilon)n, n] > \varphi(\varepsilon) \quad \text{for all } n \geq J(\varepsilon).$$

(D) There exists a sequence $(\hat{\theta}_n)_{n=0}^\infty$, that increases to ∞ , and $\exists \varepsilon_0 > 0$ and $\psi: (0, \varepsilon_0) \rightarrow (0, 1)$ such that $\forall \varepsilon < \varepsilon_0, \exists I(\varepsilon)$,

$$\hat{\theta}_n[\psi(\varepsilon)n, n] \geq 1 - \varepsilon \quad \text{for all } n \geq I(\varepsilon).$$

3.1 Preliminary Results

We will assume without loss of generality that the payoff function in our dynamic programming satisfies $0 \leq f \leq 1$.

Lemma 3.1. $\forall \varepsilon > 0, \forall N, \exists \theta_0$ such that $\forall \theta > \theta_0, \forall s_0 \in S, \exists n \geq N$ satisfying $V_n(s_0) \geq V_\theta(s_0) - \varepsilon$.

Proof: By condition (B) and by (3.2), there exists θ_0 , such that $\sum_{t=0}^N \hat{\theta}(t) < \frac{\varepsilon}{2}$ for all $\theta > \theta_0$. Let $\theta > \theta_0$, and let $s_0 \in S$. Let $s = (s_t)_{t=0}^\infty$ be an $\frac{\varepsilon}{2}$ -optimal play for θ in s_0 . Then by (3.3),

$$\sum_{t=N+1}^{\infty} \hat{\theta}(t) S_t(f(s)) \geq V_\theta(s_0) - \varepsilon,$$

where $f(s) = (f(s_t))_{t=0}^\infty$.

As $\sum_{t=N+1}^{\infty} \hat{\theta}(t) \leq 1$, the above inequality implies that a convex combination of $\{S_t(f(s)) \mid t \geq N+1\}$ is greater or equals $V_\theta(s_0) - \varepsilon$. Therefore there exists $t \geq N+1$ with $S_t(f(s)) \geq V_\theta(s_0)$, implying $V_t(s_0) \geq V_\theta(s_0) - \varepsilon$. ■

Corollary 3.2.

$$\limsup_{n \rightarrow \infty} V_n \geq \limsup_{\theta \rightarrow \infty} V_\theta.$$

Lemma 3.3. $\limsup V_\theta$ is non-increasing in plays. That is,

$$\limsup V_\theta(s_0) \geq \limsup V_\theta(s_1) \quad \text{for every } s_1 \in \Gamma(s_0).$$

Proof: Note that if $(s_t)_{t=1}^\infty$ is ε -optimal in s_1 for θ , then $s = (s_t)_{t=0}^\infty$ is a play in s_0 . Hence, it suffices to prove that for every $\varepsilon > 0$, for sufficiently large θ ,

$$\sum_{t=0}^{\infty} \theta(t) f(s_{t+1}) - f(s_t) < \varepsilon.$$

By rearranging terms and by (3.3), the last inequality can be proved by showing that

$$\sum_{t=0}^{\infty} \hat{\theta}(t) \frac{f(s_{t+1}) - f(s_0)}{t+1} < \varepsilon.$$

Hence, it suffices to prove that for every $\varepsilon > 0$, for sufficiently large θ ,

$$\sum_{t=0}^{\infty} \hat{\theta}(t) \frac{1}{t+1} < \varepsilon,$$

which follows easily from Condition (B). ■

Lemma 3.4 (Lehrer and Sorin (1992)). $\forall \varepsilon > 0, \forall n > \frac{2}{\varepsilon}$, and $\forall s_0 \in S$, there exist a play $s = (s_t)_{t=0}^{\infty}$ and a stage L such that

$$\frac{1}{T+1} \sum_{t=0}^T f(s_{L+t}) \geq V_n(s_0) - \varepsilon \quad \text{for every } 0 \leq T \leq \frac{\varepsilon}{2} n.$$

3.2 From V_{θ} to V_n

Proposition 1. Assume $\lim_{\theta \rightarrow \infty} V_{\theta} = V$, uniformly.

$\forall \varepsilon > 0 \exists N$, such that $\forall n \geq N, V_n \leq V + \varepsilon$.

Proof: Set $\varepsilon_1 = \frac{\varepsilon}{3}$. By the uniform convergence assumption, there exists θ_0 , such that

$$|V_{\theta}(s_0) - V(s_0)| < \varepsilon_1 \quad \text{for all } s_0 \in S. \quad (3.4)$$

Let M be an integer satisfying

$$\sum_{t=0}^M \hat{\theta}_0(t) > 1 - \varepsilon_1, \quad (3.5)$$

and let N be an integer satisfying $N > \frac{2}{\varepsilon_1}$. We now show that N satisfies the assertion of the proposition. Indeed, let $n \geq N$, and let $s_0 \in S$. By Lemma 3.4, there exists a play $s = (s_t)_{t=0}^{\infty}$ and an integer L that satisfy the assertion of Lemma 3.4 for ε_1 . By (3.3) and (3.5), this implies, $V_{\theta_0}(s_L) \geq V_n(s_0) - 2\varepsilon_1$. Therefore $V(s_L) \geq V_n(s_0) - 3\varepsilon_1$, by (3.4). Hence, by Lemma 3.3, and because $3\varepsilon_1 = \varepsilon$,

$$V(s_0) \geq V_n(s_0) - \varepsilon. \quad \blacksquare$$

Proposition 2. Assume $(\Theta, >)$ satisfies Condition (C), and uniform convergence of $(V_\theta)_{\theta \in \Theta}$ to V .

$$\forall \varepsilon > 0, \exists N, \text{ such that } \forall n \geq N, V_n \geq V - \varepsilon.$$

Proof: Otherwise, there exists $\varepsilon > 0$ such that for every N , there exists $n \geq N$ and $s_0 \in \mathcal{S}$ with $V_n(s_0) < V(s_0) - \varepsilon$. We now choose a particular integer N as follows: set

$\varepsilon_1 = \varepsilon_2 = \frac{\varepsilon}{2}$, and choose $\varepsilon_3, \varepsilon_4, \varepsilon_5$ in a way that will be described later. Choose an

integer K satisfying the following 4 properties.

(1) K is large enough such that at every play $s = (s_t)_{t=0}^\infty$, $\forall n \geq K$, if $V_n(s_0) < V(s_0) - \varepsilon$, then

$$S_T(f(s)) \leq V(s_0) - \varepsilon_1 \quad \text{for all } (1 - \varepsilon_1)n \leq T \leq n.$$

(2) Let $J(\varepsilon_2)$ and the sequence $(\theta_n, \varepsilon_2)_{n \geq J(\varepsilon_2)}$ satisfy the property stated in Condition (C). Choose $K \geq J(\varepsilon_2)$. That is,

$$\hat{\theta}_n[(1 - \varepsilon_2)n, n] > \varphi(\varepsilon_2) \quad \text{for every } n \geq K,$$

where $\theta_n = \theta_{n, \varepsilon_2}$.

(3) As $(\theta_n)_{n=K}^\infty$ is increasing to ∞ , and $V_\theta \rightarrow V$, we can choose K large enough such that

$$-\varepsilon_4 \leq V_{\theta_n} - V \leq \varepsilon_4 \quad \text{for all } n \geq K.$$

(4) By Proposition 1, we can choose K large enough, such that for every $n \geq K$,

$$V_n \leq V + \varepsilon_3 \quad \text{for all } n \geq K.$$

Finally, choose $N > K$ satisfying

$$\sum_{t=0}^K \hat{\theta}_n(t) < \varepsilon_5 \quad \text{for all } n \geq N.$$

By our initial assumption there exists $n \geq N$ and s_0 with $V_n(s_0) < V(s_0) - \varepsilon$. Let $s = (s_t)_{t=0}^\infty$ be any play at s_0 . Set $a_t = \hat{\theta}_n(t) S_t(f(s))$. Then

$$S_{\theta_n}(f(s)) = \sum_{t=0}^K a_t + \sum_{K < t < (1 - \varepsilon_2)n} a_t + \sum_{(1 - \varepsilon_2)n \leq t \leq n} a_t + \sum_{t > n} a_t.$$

Therefore, by the way we chose N ,

$$S_{\theta_n}(f(s)) \leq V(s_0) + \Delta,$$

where

$$\Delta = \varepsilon_3 + \varepsilon_5 - \varphi(\varepsilon_2) \varepsilon_1.$$

As the last inequality holds for every play at s_0 , then

$$V_{\theta_n}(s_0) \leq V(s_0) + \Delta.$$

Hence, by property (3), satisfied by K and hence by N , and recalling that $\varepsilon_1 = \frac{\varepsilon}{2}$, we have

$$\varphi\left(\frac{\varepsilon}{2}\right) \frac{\varepsilon}{2} \leq \varepsilon_3 + \varepsilon_4 + \varepsilon_5.$$

Thus we can have a contradiction by choosing ε_i , $i=3, 4, 5$, to be less than $\frac{1}{3} \varphi\left(\frac{\varepsilon}{2}\right) \frac{\varepsilon}{2}$. ■

3.3 From V_n to V_θ

Proposition 3. Assume $\lim_{n \rightarrow \infty} V_n = W$ uniformly.

$$\forall \varepsilon > 0, \exists \theta_0, \text{ such that } \forall \theta > \theta_0, V_\theta \leq W + \varepsilon.$$

Proof: The proof is an immediate consequence of Lemma 3.1. ■

Lemma 3.5 (Lehrer and Sorin (1992)). Assume $\lim_{n \rightarrow \infty} V_n = W$ uniformly. Then for every ε small enough, there exists an integer N , such that for every $n \geq N$ and $s_0 \in S$, there is a play $s = (s_i)_{i=0}^\infty$ at s_0 satisfying:

$$\frac{1}{T+1} \sum_{t=0}^T f(s_t) \geq W(s_0) - \varepsilon \quad \text{for every } \varepsilon n \leq T \leq (1-\varepsilon)n.$$

Proposition 4. Assume $(\Theta, >)$ satisfies condition (D), and $\lim_{n \rightarrow \infty} V_n = W$ uniformly.

$$\forall \varepsilon > 0, \exists N, \text{ such that } \forall n \geq N, V_{\theta_n} \geq W - \varepsilon,$$

where $(\bar{\theta}_n)_{n=0}^\infty$ is defined in Condition (D).

Proof: Let $\varepsilon > 0$. Let $\delta > 0$ satisfies $\frac{\delta}{1-\delta} < \min(\psi(\varepsilon), \varepsilon)$. Then by Lemma 3.5 there exists N such that for every $n \geq N$ and $s_0 \in S$, there is a play $s = (s_t)_{t=0}^\infty$ at s_0 satisfying:

$$\frac{1}{T+1} \sum_{t=0}^T f(s_t) \geq W(s_0) - \delta \quad \text{for every } \delta n \leq T \leq (1-\delta)n.$$

Without loss of generality we can choose $N \geq I(\varepsilon)$. Note that if $m \geq N$ (assuming that N was chosen large enough), there exists $n \geq N$, with

$$[\psi(\varepsilon)m, m] \subseteq [\delta n, (1-\delta)n].$$

Hence, $\hat{\theta}_m[\psi(\varepsilon)m, m] \geq 1 - \varepsilon$, and $S_T(f(s)) \geq 1 - \delta \geq 1 - \varepsilon$, for $T \in [\psi(\varepsilon)m, m]$. Therefore,

$$V_{\theta_m}(s_0) \geq W(s_0) - 2\varepsilon \quad \text{for all } m \geq N \text{ and all } s_0 \in S. \quad \blacksquare$$

Remark 1.

If the sequence $(\hat{\theta}_n)_{n=0}^\infty$, given in Condition (D) is dense in $(\Theta, >)$ (in the sense that its uniform convergence implies the uniform convergence of $(V_\theta)_{\theta \in \Theta}$), then under conditions (C) and (D), uniform convergence of $(V_n)_{n=0}^\infty$ implies uniform convergence of $(V_\theta)_{\theta \in \Theta}$ to the same limit function. As it was proved in Lehrer and Sorin (1992), such is the case when $\Theta = \{\theta_\lambda: \lambda \in [0, 1)\}$, where $\theta_\lambda(t) = (1-\lambda)\lambda^t$, and “ $>$ ” is the natural order on real numbers.

Remark 2.

Let $(\Theta, >)$ be a linearly ordered set of distributions on N satisfying (B), (C*), and (D*), where (C*) and (D*) are obtained from (C) and (D) respectively, by replacing $\hat{\theta}$ with θ everywhere. Define,

$$U_\theta(s_0) = \sup_{(s_t)_{t=0}^\infty} \sum_{t=0}^\infty \theta(t) S_t(f(s)).$$

It is obvious that our proofs yield the equivalence theorem for this solution concept as well. E.g., for every $0 < \lambda < 1$ define

$$U_\lambda(s_0) = \sup_{(s_t)_{t=0}^\infty} (1-\lambda) \sum_{t=0}^\infty \lambda^t S_t(f(s)).$$

Then (U_λ) converges uniformly if and only if (V_n) converges uniformly, and both share the same limit function.

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