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## ON LIMIT PROPERTIES OF THE REWARD FROM A MARKOV REPLACEMENT PROCESS

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This paper is a close continuation of [7] and extends the validity of assertions proved there on replacement processes.

### 1. BASIC DEFINITIONS AND NOTATIONS

Let a homogeneous Markov process with rewards  $\{X_t, t \geq 0\}$  describing the evolution of a system in a state space  $I = \{1, \dots, r\}$  be defined by exit intensities  $(\mu(1), \dots, \mu(r))$ ,  $0 < \mu(j) \leq \infty$ ,  $j = 1, \dots, r$  and by a matrix  $\mathbf{P} = \|p(i, j)\|_{i, j=1}^r$  of transition probabilities in the moment of exit. Let us denote by  $\mathbf{M} = \|\mu(i, j)\|_{i, j=1}^r$  the matrix of transition intensities of the process, where

$$\mu(i, j) = \mu(i)p(i, j) \text{ for } i \neq j, \mu(i, i) = -\mu(i) = -\sum_{j \neq i} \mu(i, j).$$

Consider a situation, where the development of the process may be influenced by an action called *replacement*. According to [5] we mean under a replacement of type  $(i, +j)$  the instantaneous shift of the system from state  $i$  into state  $j$ . The complete history of this process is given by the following sequence

$$\omega = \{i_0, t_0, \delta_0; i_1, t_1, \delta_1; \dots; i_n, t_n, \delta_n; \dots\},$$

where  $i_0, i_1, \dots, i_n, \dots$  are the states visited,  $t_0, t_1, \dots, t_n, \dots$  the corresponding sojourn times and  $\delta_0, \delta_1, \dots, \delta_n, \dots$  is the sequence of zeros and units, where  $\delta_n = 0$  in case of  $i_n \rightarrow i_{n+1}$  without interference and  $\delta_n = 1$  in case of  $i_n \rightarrow i_{n+1}$  being the replacement. We use in accordance with [5] the notation

$$\omega_n = \{i_0, \dots; i_{n-1}, t_{n-1}, \delta_{n-1}; i_n\}$$

for the history up to the  $n$ -th state change.

A *replacement policy* (see [5]) is a decision for all possible sequences  $\omega_n$  for how long time the system will be left in  $i_n$  without shifting (maximal sojourn time) and

in what state it is to be shifted. Since we do not to exclude the random choice of these quantities, we identify a replacement policy with a sequence of functions

$$F = \{^n F_k(t/\omega_n)\}, \quad k = 1, \dots, r; n = 1, 2, \dots$$

where  $^n F_k(t/\omega_n)$  is a probability that the maximal sojourn time in  $i_n$  will be less than  $t$  and the eventual shift will be into  $k \neq i_n$ .

**Assumption 1.**

Consider such replacement policies  $F$  only, where

- a) there exists only a finite number of replacements in every finite interval,
- b) there are neither two or more replacements in the same moment, with probability 1.

According to Assumption 1 there is assigned a trajectory  $\{Y_t, t \geq 0\}$  not left continuous at the time of transition and not right continuous at the time of replacement to almost every  $\omega$ .

In what follows we denote by

$\sigma_0 = 0, \sigma_1, \sigma_2, \dots$  such moments in which the trajectory is discontinuous,

$$Y_t^- = Y_{t-}, t > 0; Y_0^- = Y_0; Y_t^+ = Y_{t+}, t \geq 0;$$

$\mathcal{B}_t = \sigma a \{(Y_s = j), j \in I, s \in \langle 0, t \rangle; \text{events of zero probability}\}$ ,

$$\mathcal{B}_t^+ = \bigcap_{s>t} \mathcal{B}_s,$$

$E^F$  a mathematical expectation in a replacement process under the replacement policy  $F$ ,

$D$  a set of couples  $(i, +j)$  meaning admissible replacements,

$$D_i = \{j: (i, +j) \in D\}.$$

The reward from the process is defined by the following sets of numbers:

$\varrho(i), i \in I$ , the reward per a time unit in state  $i$ ,

$r(i, j), i, j \in I$ , the reward from the transition  $(i, j)$ ; we set  $r(i, i) = 0$ ,

$v(i, j), i, j \in I$ , the reward from the replacement  $(i, +j)$ ; we set  $v(i, i) = 0$ .

A stationary replacement policy  $f$  is given by a function  $f(j)$  defined on a subset  $I_f \subset I$  and taking values in  $I$  such that  $f(j) \in D_j$  for  $j \in I_f, f(j) \neq j$ . The replacement policy  $f$  is the prescription to realize instantaneously the replacement  $j \rightarrow f(j)$  whenever there occurs a transition in state  $j$ . No replacements occur in states  $j \notin I_f$ .

**Assumption 2.**

$$(i, +j) \in D, (j, +k) \in D \Rightarrow (i, +k) \in D \quad \text{or} \quad i = k,$$

$$v(i, j) + v(j, k) \leq v(i, k).$$

Let  $R_T$  be a reward from the process up to the time  $T$ . In accordance with our previous definitions

$$R_T = \int_0^T \varrho(Y_t) dt + \sum_{n=0}^N [r(Y_{\sigma_n}^-, Y_{\sigma_n}) + v(Y_{\sigma_n}, Y_{\sigma_n}^+)], \quad \sigma_N \leq T < \sigma_{N+1}.$$

## 2. LIMIT PROPERTIES OF A REWARD

We demonstrate first some auxiliary assertions.

### Lemma 1.

Let  $g(i, k)$  be a function defined on  $I \times I$ ,  $g(i, i) = 0$ ,  $i \in I$ . Let

$$G_T = \sum_{n=1}^N g(Y_{\sigma_n}^-, Y_{\sigma_n}), \quad \sigma_N \leq T < \sigma_{N+1},$$

introduce

$$\gamma(i) = \sum_{k \neq i} \mu(i, k) g(i, k), \quad \gamma_2(i) = \sum_{k \neq i} \mu(i, k) (g(i, k))^2.$$

Then it holds under an arbitrary replacement policy  $F$  for  $0 \leq t \leq T$

$$E^F \{G_T - G_t / \mathcal{B}_t^+\} = E^F \left\{ \int_t^T \gamma(Y_s) ds / \mathcal{B}_t^+ \right\}, \quad (1)$$

$$E^F \left\{ (G_T - G_t - \int_t^T \gamma(Y_s) ds)^2 / \mathcal{B}_t^+ \right\} = E^F \left\{ \int_t^T \gamma_2(Y_s) ds / \mathcal{B}_t^+ \right\}. \quad (2)$$

Proof: a) Since the conditional distribution describes a Markov replacement process under common replacement policy, the proof of (1) reduces to the verification of

$$E^{F'}(G_T) = E^{F'} \left\{ \int_0^T \gamma(Y_s) ds \right\}, \quad T \geq 0,$$

for an arbitrary initial probability distribution and an arbitrary policy  $F'$ .

The proof of the above assertion proceeds similarly to that of Lemma 1 in [6].

b) Taking instead of  $g(i, k)$  the function  $g^2(i, k)$  throughout the proof of (1) we show that

$$E^{F'} \left( \sum_{n=1}^N g^2(Y_{\sigma_n}^-, Y_{\sigma_n}) \right) = E^{F'} \left( \int_0^T \gamma_2(Y_s) ds \right), \quad \sigma_N \leq T < \sigma_{N+1}.$$

Then (2) will be established by proving

$$E^{F'} \left( G_T - \int_0^T \gamma(Y_s) ds \right)^2 = E^{F'} \left( \sum_{n=1}^N g^2(Y_{\sigma_n}^-, Y_{\sigma_n}) \right),$$

under an arbitrary policy  $F'$  and an arbitrary initial distribution. The proof proceeds analogous to that of Corollary 1 in [6].

### Lemma 2.

There exist constants  $K_{mT}$  such that

$$E^F |G_T|^m \leq K_{mT} \left[ \max_{i, j \in I} (|g(i, j)|) \right]^m, \quad m = 1, 2, \dots, \quad (3)$$

for an arbitrary replacement policy  $F$ .

Proof: We denote by  $\mu = \max(\mu(1), \dots, \mu(r))$ ,  $\sigma'_n$  the moment of the  $n$ -th transition (the  $n$ -th left discontinuity of the trajectory). We prove by induction

$$P^F(\sigma'_n \leq t) \leq H^{(n)}(t), \quad (4)$$

where  $H^{(n)}(t)$  is the  $n$ -multiple convolution  $H^{(1)}(t) = 1 - e^{-\mu t}$ . We denote by  $N'_T$  the number of transitions in  $\langle 0, T \rangle$ . According to (4) it holds

$$\begin{aligned} E^F(N'_T)^m &= \sum_{n=1}^{\infty} n^m [P^F(\sigma'_n \leq T) - P^F(\sigma'_{n+1} \leq T)] \leq \\ &\leq \sum_{n=1}^{\infty} (n^m - (n-1)^m) H^{(n)}(T) = \sum_{n=1}^{\infty} (n^m - (n-1)^m) \frac{1}{(n-1)!} \int_0^{\mu T} x^{n-1} e^{-x} dx = \\ &= \sum_{n=1}^{\infty} n^m \frac{(\mu T)^n}{n!} e^{-\mu T} = K_{mT}. \end{aligned}$$

Thus

$$\begin{aligned} E^F |G_T|^m &= E^F(|\sum_{n=1}^N g(Y_{\sigma_n}^-, Y_{\sigma_n})|^m) = E^F(|\sum_{j=1}^{N'_T} g(Y_{\sigma_j}^-, Y_{\sigma_j})|^m) \leq \\ &\leq E^F[(N'_T)^m (\max_{i,k \in I} \{ |g(i,k)| \})^m] \leq (\max_{i,k \in I} \{ |g(i,k)| \})^m \cdot K_{mT}. \square \end{aligned}$$

Let  $f$  be a fixed chosen stationary replacement policy such that under it exists one recurrent class and eventually a transient class only. Let the constant  $\Theta$ ,  $w(1), \dots, w(r)$  be defined by the following equations

$$\begin{aligned} v(i, f(i)) + w(f(i)) - w(i) &= 0, \quad i \in I_f, \quad (5) \\ \varrho(i) + \sum_{k \neq i} \mu(i, k) [r(i, k) + w(k) - w(i)] - \Theta &= 0, \quad i \notin I_f. \end{aligned}$$

According to [2] the system (5) uniquely determines the number  $\Theta$  ( $\Theta$  is the mean reward per a time unit from the process in using the replacement policy  $f$ ),  $w(1), \dots, w(r)$  except for adding an arbitrary constant.

Denote for  $i \in I$

$$\begin{aligned} \varphi(i) &= \varrho(i) + \sum_{k \neq i} \mu(i, k) [r(i, k) + w(k) - w(i)] - \Theta, \\ \psi_1(i) &= \sum_{k \neq i} \mu(i, k) [r(i, k) + w(k) - w(i)], \\ \psi_2(i) &= \sum_{k \neq i} \mu(i, k) [r(i, k) + w(k) - w(i)]^2. \end{aligned}$$

Let us introduce an auxiliary random process (see [3])

$$\begin{aligned} M_T &= R_T - \Theta T + w(Y_T^+) - w(Y_0) - \int_0^T \varphi(Y_t) dt - \\ &- \sum_{n=0}^N [v(Y_{\sigma_n}, Y_{\sigma_n}^+) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})], \quad T \geq 0, \sigma_N \leq T < \sigma_{N+1}. \end{aligned}$$

**Lemma 3.**

$\{M_T, T \geq 0\}$  is a martingale with respect to  $\{\mathcal{B}_T^+, T \geq 0\}$  under an arbitrary policy  $F$ . It holds for  $0 \leq t \leq T$

$$E^F\{(M_T - M_t)^2/\mathcal{B}_t^+\} = E^F\left\{\int_t^T \psi_2(Y_s) ds/\mathcal{B}_t^+\right\} \quad F\text{-almost everywhere.}$$

Proof: By substituting instead of  $R_T$  and  $w(Y_T^+) - w(Y_0) = \sum_{n=0}^N [w(Y_{\sigma_n}) - w(Y_{\sigma_n}^-) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})]$ ,  $\sigma_N \leq T < \sigma_{N+1}$ , into the expression for  $M_T$  we obtain

$$M_T = -\int_0^T \psi_1(Y_t) dt + \sum_{n=0}^N [r(Y_{\sigma_n}^-, Y_{\sigma_n}) + w(Y_{\sigma_n}) - w(Y_{\sigma_n}^-)].$$

The substitution of  $g(i, k) = r(i, k) + w(k) - w(i)$  in (1) of Lemma 1 gives

$$E^F\{M_T - M_t/\mathcal{B}_t^+\} = E^F\{G_T - G_t - \int_t^T \gamma(Y_s) ds/\mathcal{B}_t^+\} = 0, \quad t \leq T,$$

and thus

$$E^F\{M_T/\mathcal{B}_t^+\} = E^F\{M_t/\mathcal{B}_t^+\} = M_t \quad \text{for all } t \leq T.$$

The other assertion proved follows analogous from (2), Lemma 1.  $\square$

**Corollary.**

Under an arbitrary replacement policy  $F$

$$\lim_{n \rightarrow \infty} \frac{1}{T} M_T = 0 \quad F\text{-almost everywhere.} \quad (6)$$

Proof: 1. We can write  $M_n = \sum_{k=1}^n (M_k - M_{k-1})$ . According to Lemma 3

$$\begin{aligned} & \sum_{n=1}^{\infty} \frac{1}{n^2} E(M_n - M_{n-1})^2 = \\ & = \sum_{n=1}^{\infty} \frac{1}{n^2} E\left(\int_{n-1}^n \psi_2(Y_s) ds\right) \leq \sum_{n=1}^{\infty} \frac{1}{n^2} (\max_{i \in I} \{\psi_2(i)\}), \end{aligned}$$

and  $\{M_n, n = 1, 2, \dots\}$  being a martingale, it is by [4], page 407

$$\lim_{n \rightarrow \infty} \frac{1}{n} M_n = 0 \quad F\text{-almost everywhere.} \quad (7)$$

2. Let  $n \leq T < n + 1$ , then

$$\left| \frac{1}{T} M_T \right| \leq \frac{1}{n} \sup_{n \leq T < n+1} |M_T - M_n| + \frac{1}{n} |M_n|.$$

According to (7) it suffices to prove that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sup_{n \leq T < n+1} |M_T - M_n| = 0 \quad F\text{-almost everywhere.} \quad (8)$$

Denote by

$$c = \max_{i \in I} \{\psi_1(i)\}, \quad k = \max_{i, j \in I} \{|r(i, j) + w(j) - w(i)|\},$$

$X_n$  the number of transitions during the time  $\langle n, n + 1 \rangle$ . Then

$$\sup_{n \leq T < n+1} |M_T - M_n| \leq c + kX_n. \quad (9)$$

As the series  $E(\sum_{n=1}^{\infty} \frac{1}{n^2} X_n^2)$  converges, it is  $\lim_{n \rightarrow \infty} \frac{1}{n^2} X_n^2 = 0$   $F$ -almost everywhere.

Hence

$$\lim_{n \rightarrow \infty} \frac{1}{n} (c + kX_n) = 0 \quad F\text{-almost everywhere.}$$

This due to (9) proves (8).  $\square$

**Theorem 1.**

Let the optimality equation (see [2]) for the replacement policy  $f$  hold, i.e.

$$\max_{j \in I} \{v(j, k) + w(k) - w(j); \varrho(j) + \sum_{k \neq j} \mu(j, k) [r(j, k) + w(k) - w(j)] - \Theta\} = 0, \quad (10)$$

Then under an arbitrary policy  $F$

$$\limsup_{n \rightarrow \infty} \frac{1}{T} R_T \leq \Theta \quad F\text{-almost everywhere.}$$

**Proof:**

It follows from assumption (10) that  $\varphi(j) \leq 0$  for all  $j \in I$ , i.e.

$$-\int_0^T \varphi(Y_t) dt \geq 0.$$

Likewise, we have from (10)

$$-\sum_{n=0}^N [v(Y_{\sigma_n}, Y_{\sigma_n}^+) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})] \geq 0.$$

Thus

$$M_T \geq R_T - \Theta T + w(Y_T^+) - w(Y_0).$$

Since

$$\lim_{T \rightarrow \infty} \frac{1}{T} [w(Y_T^+) - w(Y_0)] = 0, \quad (11)$$

it holds

$$\limsup_{T \rightarrow \infty} \frac{1}{T} M_T \geq \limsup_{T \rightarrow \infty} \frac{1}{T} R_T - \Theta \quad F\text{-almost everywhere}$$

whence the statement follows from Corollary 3.  $\square$

**Definitions.**

We call the state  $i \in I$  consistent with the policy  $f$ , if  $\varphi(i) = 0$ . We call the replacement  $i \rightarrow k$  consistent with  $f$ , if  $v(i, k) + w(k) - w(i) = 0$ .

Denote by

- $Q_T$  the whole sojourn time in the inconsistent states in  $\langle 0, T \rangle$ ,
- $\bar{Q}_T$  the whole sojourn time in states  $I_f$  in the interval  $\langle 0, T \rangle$ ,
- $O_T$  the whole number of inconsistent replacements in  $\langle 0, T \rangle$ ,
- $\bar{O}_T$  the whole number of replacements different from  $i \rightarrow f(i)$  in  $\langle 0, T \rangle$ .

Obviously

$$\bar{Q}_T \geq Q_T, \quad \bar{O}_T \geq O_T.$$

**Theorem 2.**

Let  $F$  be a replacement policy. If

$$\lim_{T \rightarrow \infty} \frac{1}{T} Q_T = \lim_{T \rightarrow \infty} \frac{1}{T} O_T = 0 \quad F\text{-almost everywhere (F-in probability)} \quad (12)$$

then

$$\lim_{T \rightarrow \infty} \frac{1}{T} R_T = \Theta \quad F\text{-almost everywhere (F-in probability)}. \quad (13)$$

If the equation of optimality (10) is valid, then (12) is necessary for the validity of (13) as well.

Proof:

$$M_T = R_T - \Theta T + w(Y_T^+) - w(Y_0) - \int_0^T \varphi(Y_t) dt - \sum_{n=0}^N [v(Y_{\sigma_n}, Y_{\sigma_n}^+) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})], \quad \sigma_N \leq T < \sigma_{N+1}.$$

a) The function  $\varphi(\cdot)$  is constant in any interval  $\langle \sigma_{j-1}, \sigma_j \rangle$ . If  $i$  is a consistent state with  $f$ , then  $\varphi(i) = 0$  and thus

$$\min_{i \in I} \{\varphi(i)\} Q_T \leq \int_0^T \varphi(Y_t) dt \leq \max_{i \in I} \{\varphi(i)\} Q_T.$$

There are nonzero addends in the last sum of the expression  $M_T$  in those moments  $\sigma_n$  only, where an inconsistent replacement with  $f$  occurs, hence

$$\begin{aligned} \min_{i, j \in I} \{v(i, j) + w(j) - w(i)\} O_T &\leq \sum_{n=0}^N [v(Y_{\sigma_n}, Y_{\sigma_n}^+) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})] \leq \\ &\leq \max_{i, j \in I} \{v(i, j) + w(j) - w(i)\} O_T. \end{aligned}$$

The above relations prove together with (6) and (11) that (12) follows from (13).

b) Let (13) hold and let  $f$  fulfil (10). If  $i$  is the state consistent with  $f$ , then  $\varphi(i) = 0$ . In the opposite case then  $i \in I_f$  and according to (10)  $\varphi(i) < 0$ .



Denote by  $I_0$  the set of inconsistent states with  $f$ . According to (13)

$$0 \geq \max_{i \in I_0} \{\varphi(i)\} \frac{Q_T}{T} \geq \frac{1}{T} \int_0^T \varphi(Y_t) dt \rightarrow 0 \quad \text{for } T \rightarrow \infty.$$

The nonzero expressions are in the sum

$$\sum_{n=0}^N [v(Y_{\sigma_n}, Y_{\sigma_n}^+) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})]$$

in those moments  $\sigma_n$ , if there is in  $F$  a transition or a replacement consistent with  $f$ . If (10), (13) hold, then

$$\begin{aligned} 0 &\geq \max_{i \rightarrow k \text{ replacements inconsistent with } f} \{v(i, k) + w(k) - w(i)\} \frac{O_T}{T} \geq \\ &\geq \frac{1}{T} \sum_{n=0}^N [v(Y_{\sigma_n}, Y_{\sigma_n}^+) + w(Y_{\sigma_n}^+) - w(Y_{\sigma_n})] \rightarrow 0 \quad \text{for } T \rightarrow \infty. \end{aligned}$$

Hence, if (10) holds, then (12) is necessary for (13) to be fulfilled.  $\square$

### Theorem 3.

Let  $F$  be a replacement policy. Let

$$\lim_{T \rightarrow \infty} \frac{1}{\sqrt{T}} \bar{Q}_T = 0 = \lim_{T \rightarrow \infty} \frac{1}{\sqrt{T}} \bar{O}_T \quad F\text{-in probability} \quad (14)$$

then

$$\frac{R_T - \Theta T}{\sqrt{T}}$$

has for  $T \rightarrow \infty$  asymptotically normal distribution  $N(0, \zeta)$ , where  $\zeta$  is determined by equations

$$\begin{aligned} w_2(f(i)) - w_2(i) &= 0, \quad i \in I_f, \\ \psi_2(i) + \sum_{k \neq i} \mu(i, k) [w_2(k) - w_2(i)] - \zeta &= 0, \quad i \notin I_f, \end{aligned}$$

containing auxiliary constants  $w_2(\bar{1}), \dots, w_2(r)$ .

Proof: We prove this theorem in several steps.

I. We prove first that it follows from (14)

$$\lim_{n \rightarrow \infty} \left( \frac{R_n - \Theta n}{\sqrt{n}} - \frac{M_n}{\sqrt{n}} \right) = 0 \quad f\text{-in probability.}$$

According to definition

$$\begin{aligned} M_n &= R_n - \Theta n + w(Y_n^+) - w(Y_0) - \int_0^n \varphi(Y_t) dt - \\ &- \sum_{j=0}^N [v(Y_{\sigma_j}, Y_{\sigma_j}^+) + w(Y_{\sigma_j}^+) - w(Y_{\sigma_j})], \quad \sigma_N \leq n < \sigma_{N+1}. \end{aligned}$$

Obviously

$$\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} [w(Y_n^+) - w(Y_0)] = 0.$$

Since

$$\begin{aligned} \min_{i \in I} \{\varphi(i)\} Q_n &\leq \int_0^n \varphi(Y_t) dt \leq \max_{i \in I} \{\varphi(i)\} Q_n, \\ \min_{i, k \in I} \{v(i, k) + w(k) - w(i)\} O_n &\leq \sum_{j=0}^N [v(Y_{\sigma_j}, Y_{\sigma_j}^+) + w(Y_{\sigma_j}^+) - w(Y_{\sigma_j})] \leq \\ &\leq \max_{i, k \in I} \{v(i, k) + w(k) - w(i)\} O_n, \end{aligned}$$

(see the proof of Theorem 2)

assertion I follows from (14) by using  $Q_n \leq \bar{Q}_n$ ,  $O_n \leq \bar{O}_n$ .

II.  $\frac{M_n}{\sqrt{n}}$  has for  $n \rightarrow \infty$  asymptotically normal distribution  $N(0, \zeta)$ .

The proof of the above statement lies in the verification of assumptions of the central limit theorem for martingales below (see [1], [7]):

Let  $\{M_n = \sum_{m=0}^{n-1} Y_m, n = 1, 2, \dots\}$  be a martingale with respect to the class of  $\sigma$ -algebras  $\{\mathcal{F}_n, n = 1, 2, \dots\}$ . Let

- (i)  $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=0}^{n-1} E\{Y_m^2 \cdot \chi_{\{|Y_m| \geq \varepsilon/\sqrt{n}\}} / \mathcal{F}_m\} = 0$  in probability for all  $\varepsilon > 0$ ,
- (ii)  $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=0}^{n-1} E\{Y_m^2 / \mathcal{F}_m\} = \zeta$  in probability, where  $\zeta$  is a constant,

then  $\frac{M_n}{\sqrt{n}}$  is asymptotically normal  $N(0, \zeta)$  for  $n \rightarrow \infty$ .

In our case we have  $M_n = \sum_{m=0}^{n-1} (M_{m+1} - M_m)$ . By Lemma 3  $\{M_n, n = 1, 2, \dots\}$  is a martingale with respect to the class of  $\sigma$ -algebras  $\{\mathcal{B}_n^+, n = 1, 2, \dots\}$ .

1. Let  $\varepsilon > 0$  be an arbitrary number. Then

$$E\{(M_{m+1} - M_m)^2 \chi_{\{|M_{m+1} - M_m| \geq \varepsilon/\sqrt{n}\}} / \mathcal{B}_m^+\} \leq \frac{1}{\varepsilon \sqrt{n}} E\{|M_{m+1} - M_m|^3 / \mathcal{B}_m^+\}.$$

To the proof of

$$E\{|M_{m+1} - M_m|^3 / \mathcal{B}_m^+\} \leq c, \quad c \text{ constant}, \quad (15)$$

it is sufficient to show that under an arbitrary replacement policy  $F'$  and under an arbitrary distribution

$$E^{F'}(|M_1 - M_0|^3) \leq c.$$

As  $M_0 = 0$ , we have (using the notation of the proof in Lemma 3)

$$E^{F'}(|M_1|^3) = E^{F'}(|G_1 - \int_0^1 \gamma(Y_s) ds|^3) \leq K_{1,3} + 3kK_{1,2} + 3k^2K_{1,1} + k^3 = c,$$

where  $k = \max_{i \in I} \{|\gamma(i)|\}$  and where according to Lemma 2  $E^F(|G_1|^m) \leq K_{1,m}$ ,  $m = 1, 2, \dots$ . The realization of (i) follows then from (15).

2. Let the numbers  $w_2(1), \dots, w_2(r), \zeta$  be solutions of the system of equations from the statement of the theorem. Let us define to the verification of (ii)

$$\varphi_2(i) = \psi_2(i) + \sum_{k \neq i} \mu(i, k) [w_2(k) - w_2(i)] - \zeta, \quad i \in I.$$

a) We prove that under an arbitrary policy  $F$

$$U_T = \int_0^T \psi_2(Y_t) dt - \zeta T + w_2(Y_T^+) - w_2(Y_0) - \int_0^T \varphi_2(Y_t) dt - \sum_{n=0}^N [w_2(Y_{\sigma_n}^+) - w_2(Y_{\sigma_n})],$$

$$T \geq 0, \quad \sigma_N \leq T < \sigma_{N+1},$$

is a martingale with respect to  $\{\mathcal{B}_T^+, T \geq 0\}$  satisfying the law of large numbers. Denote

$$\xi_1(i) = \sum_{k \neq i} \mu(i, k) [w_2(k) - w_2(i)], \quad i \in I.$$

On substituting and modifying we get

$$U_T = \sum_{n=0}^N [w_2(Y_{\sigma_n}) - w_2(Y_{\sigma_n}^-)] - \int_0^T \xi_1(Y_t) dt.$$

Using Lemma 1 for  $\gamma(i, k) = w_2(k) - w_2(i)$  gives

$$E(U_T - U_t | \mathcal{B}_t^+) = E(G_T - G_t - \int_t^T \gamma(Y_s) ds | \mathcal{B}_t^+) = 0, \quad t \leq T, \quad (16)$$

$$E\{(U_T - U_t)^2 | \mathcal{B}_t^+\} = E\left\{\int_t^T \xi_2(Y_s) ds | \mathcal{B}_t^+\right\}, \quad t \leq T, \quad (17)$$

where

$$\xi_2(i) = \sum_{k \neq i} \mu(i, k) [w_2(k) - w_2(i)]^2, \quad i \in I.$$

It follows from (16) that  $\{U_T, T \geq 0\}$  is a martingale with respect to  $\{\mathcal{B}_t^+, t \geq 0\}$  and thus from (17) in the same manner as in the Corollary of Lemma 3

$$\lim_{T \rightarrow \infty} \frac{1}{T} U_T = 0, \quad F\text{-almost everywhere.} \quad (18)$$

We can prove similarly as in the proof of Theorem 2 that under the validity of (14)

$$\lim_{T \rightarrow \infty} \frac{1}{T} \left\{ w_2(Y_T^+) - w_2(Y_0) - \int_0^T \varphi_2(Y_t) dt - \sum_{n=0}^N [w_2(Y_{\sigma_n}^+) - w_2(Y_{\sigma_n})] \right\} = 0,$$

$F$ -in probability (19)

and thus from (18) and from the definition of  $U_T$

$$\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T \psi_2(Y_t) dt = \zeta, \quad F\text{-in probability.} \quad (20)$$

b) We shall prove further that  $\{D_n, n = 1, 2, \dots\}$ , where (see [7])

$$D_n = \int_0^n \psi_2(Y_t) dt - \sum_{m=0}^{n-1} E\{(M_{m+1} - M_m)^2 / \mathcal{B}_m^+\},$$

is a martingale with respect to  $\{\mathcal{B}_n^+, n = 1, 2, \dots\}$ , for which the law of large numbers holds.

According to Lemma 3 we can write

$$D_n = \int_0^n \psi_2(Y_t) dt - \sum_{m=0}^{n-1} E\left\{ \int_m^{m+1} \psi_2(Y_t) dt / \mathcal{B}_m^+ \right\}.$$

For each  $m \leq n$  natural numbers

$$E\{D_n / \mathcal{B}_m^+\} = \int_0^m \psi_2(I_t) dt - \sum_{j=0}^{m-1} E\left\{ \int_j^{j+1} \psi_2(Y_t) dt / \mathcal{B}_j^+ \right\} = D_m.$$

If we denote

$$Y_m = \int_m^{m+1} \psi_2(Y_t) dt - E\left\{ \int_m^{m+1} \psi_2(Y_t) dt / \mathcal{B}_m^+ \right\},$$

then

$$D_n = \sum_{m=0}^{n-1} Y_m.$$

As for arbitrary  $m = 0, 1, \dots$

$$EY_m^2 \leq E\left( \int_m^{m+1} \psi_2(Y_t) dt \right)^2 \leq c^2,$$

where

$$c = \max_{i \in I} \{\psi_2(i)\},$$

is the series

$$\sum_{m=0}^{\infty} \frac{EY_m^2}{(m+1)^2}$$

convergent and by [4], page 407

$$\lim_{n \rightarrow \infty} \frac{1}{n} D_n = 0 \quad F\text{-almost everywhere.} \quad (21)$$

It is obvious from (20) and (21) that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=0}^{n-1} E\{(M_{m+1} - M_m)^2 / \mathcal{B}_m^+\} = \zeta \quad F\text{-in probability.}$$

it means the assumption (ii) for martingale  $\{M_n, n = 1, 2, \dots\}$  is valid.

In parts I and II of the proof we have proved the following assertion: Let (14) be valid, then  $\frac{R_n - \Theta n}{\sqrt{n}}$  has for  $n \rightarrow \infty$  asymptotically normal distribution  $N(0, \zeta)$ .

III. Analogous to part I of this proof we can verify that

$$\lim_{T \rightarrow \infty} \left( \frac{R_T - \Theta T}{\sqrt{T}} - \frac{M_T}{\sqrt{T}} \right) = 0 \quad F\text{-in probability.}$$

IV. To conclude the proof we establish  $\frac{M_T}{\sqrt{T}}$  having for  $T \rightarrow \infty$  asymptotically normal distribution  $N(0, \zeta)$ .

Let  $n \leq T < n + 1$ . We know (see the proof of Lemma 3) that

$$E(M_T - M_n)^2 \leq \max_{i \in I} \{\psi_2(i)\} = c,$$

and thus

$$EM_T^2 \leq cT.$$

Hence

$$\begin{aligned} E \left( \frac{M_T}{\sqrt{T}} - \frac{M_n}{\sqrt{n}} \right)^2 &= E \left[ M_T \left( \frac{1}{\sqrt{T}} - \frac{1}{\sqrt{n}} \right) + \frac{1}{\sqrt{n}} (M_T - M_n) \right]^2 \leq \\ &\leq 2 \left[ \frac{1}{T} \left( 1 - \sqrt{\frac{T}{n}} \right)^2 cT + \frac{c}{n} \right] \leq 2c \left[ \left( 1 - \sqrt{1 - \frac{1}{T}} \right)^2 + \frac{1}{T-1} \right], \end{aligned}$$

and thus

$$\lim_{T \rightarrow \infty} E \left( \frac{M_T}{\sqrt{T}} - \frac{M_n}{\sqrt{n}} \right)^2 = 0.$$

Using Chebyshev inequality we get

$$\lim_{T \rightarrow \infty} \left( \frac{M_T}{\sqrt{T}} - \frac{M_n}{\sqrt{n}} \right) = 0 \quad F\text{-in probability}$$

and the assertion IV. follows from assertion II.

Theorem 3 is proved by III. and IV.  $\square$

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SOUHRN

## LIMITNÍ VLASTNOSTI VÝNOSU Z MARKOVOVA PROCESU S OBNOVAMI

PAVLA KUNDEROVÁ

Článek úzce navazuje na [7] a rozšiřuje platnost tam uvedených tvrzení pro procesy s obnovami (viz [5]). Nechť  $R_T$  je výnos z procesu za dobu  $\langle 0, T \rangle$ ,  $\Theta$  průměrný výnos na jednotku času při užití stacionární strategie  $f$  při níž existuje pouze jedna třída rekurentních stavů. Je dokázáno (věta 1), že je-li  $f$  optimální (viz [2]), je při libovolné strategii obnovy  $F$

$$\limsup_{T \rightarrow \infty} \frac{1}{T} R_T \leq \Theta \quad F \text{ — skoro všude.}$$

Zavádí se pojem souhlasné obnovy a souhlasného stavu se strategií  $f$ . Věta 2 uvádí podmínky postačující resp. nutné k tomu, aby

$$\lim_{T \rightarrow \infty} \frac{1}{T} R_T = \Theta \quad F \text{ — skoro všude (} F \text{ — podle pravděpodobnosti).}$$

Jsou formulovány podmínky (věta 3), za nichž má  $\frac{R_T - \Theta T}{\sqrt{T}}$  pro  $T \rightarrow \infty$  asymptoticky normální rozdělení  $N(0, \zeta)$ , kde  $\zeta$  je jistá konstanta.

РЕЗЮМЕ

## ПРЕДЕЛЬНЫЕ КАЧЕСТВА ДОХОДА ИЗ ПРОЦЕССА МАРКОВА С ВОССТАНОВЛЕНИЯМИ

ПАВЛА КУНДЕРОВА

В работе обобщаются теоремы сформулированные в [7] для управляемых процессов Маркова. Пусть  $R_T$  доход из процесса в течение интервала  $\langle 0, T \rangle$ ,  $\Theta$  средний доход за единицу времени, когда множество состояний процесса при использовании стационарной стратегии  $f$  имеет единственный класс возвратных состояний. Показано достаточное условие для того, чтобы для любой стратегии  $F$

$$\limsup_{T \rightarrow \infty} \frac{1}{T} R_T \leq \Theta$$

$F$  — почти наверное (теорема 1). Определены согласное восстановление и согласное состояние со стратегией  $f$ . Решение проблемы об асимптотическом распределении  $\frac{R_T - \Theta T}{\sqrt{T}}$  при  $T \rightarrow \infty$  находится в теореме 3. Теорема 2 устанавливает условия для того чтобы

$$\lim_{T \rightarrow \infty} \frac{1}{T} R_T = \Theta \quad F \text{ — почти наверное (} F \text{ — по вероятности).}$$