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## Renewal theorem and Markov chains

by

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RÉSUMÉ. — Une nouvelle démonstration des théorèmes de renouvellement de Orey-Feller-Blackwell est donnée; elle utilise les propriétés des fonctions harmoniques d'un processus markovien *ad hoc*.

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We will explore a remark by Feller [1], a remark unfortunately not initially remarked by the author, concerning a Markov process associated with the renewal process. From considerations of the asymptotic properties of this Markov process we obtain the Orey-Feller-Blackwell renewal theorems in a unified « simple » way.

Consider a sequence of iid random variables  $T_1, T_2, \dots$  with distribution functions  $F$  concentrated on  $(0, \infty)$ , that is  $F(0) = 0$ . Following Feller [2] we may have a non-negative variable  $S_0$  with a proper distribution  $G$  and we put

$$S_n = S_0 + T_1 + \dots + T_n$$

The renewal process  $S_n$  is called pure if  $S_0 = 0$  and delayed otherwise (for brevity a delayed process with starting distribution  $G$  will be called a  $G$ -process as opposed to a pure process). Also for any  $t > 0$  there is a unique subscript  $N_t$  [3] such that  $S_{N_t} < t \leq S_{N_t+1}$  (note this definition for the stopping time  $N_t+1$  differs slightly from Feller's). We define the excess (waiting) time at  $t$  as  $S_{N_t+1} - t$  and we define

$Y^t(x)$  = the probability the excess time at  $t$  for the pure process  $\leq x$ ,  
 $Y_G^t(x)$  = the probability the excess time at  $t$  for the  $G$ -process  $\leq x$ ,  
 $u(t)$  = expected number of renewals for the pure process by  $t$ ,

and

$V_G(t)$  = expected number of renewals for the G-process by  $t$ .

We now consider  $R_+ = [0, \infty]$  as the state space of a Markov process where the transition probabilities are defined as follows. For  $A \in \mathcal{B}(R_+)$ , the Borel sets on  $R_+$ ,  $t > 0$  and  $x \in R_+$

$$\begin{aligned} P_t(x, A) &= \Pr(\text{the excess time at } t \text{ for a } \delta_x\text{-process} \in A) && \text{if } x < t \\ P_t(x, A) &= 1 && \text{if } x \in t + A \\ P_t(x, A) &= 0 && \text{if } x \notin t + A \\ P_0(x, A) &= 1 && \text{if } x \in A \\ P_0(x, A) &= 0 && \text{if } x \notin A. \end{aligned}$$

(The renewal  $\delta_x$ -process gives  $S_0$  distribution  $\delta_x$  where  $\delta_x$  means the distribution where all the probability is concentrated at  $x$ ). In more picturesque language we say  $P_t(x, A)$  is the probability of starting at  $x$  and with variable steps of length  $T$  finally jumping right over  $t$  and into the set  $t + A$ . Of course if  $x \geq t$  we are already passed  $t$  and we only ask if we are already in  $t + A$ . Also we note that  $P_t(x, A) = \Pr(\text{the excess time for the pure process at time } t - A)$  since these probabilities are invariant under translation.

It is quickly seen that for all  $x \in R_+$ ,  $t \geq 0$ ,  $A \rightarrow P_t(x, A)$  is a probability on  $\mathcal{B}(R_+)$ . Also for all  $A \in \mathcal{B}(R_+)$ ,  $x \rightarrow P_t(x, A)$  is measurable. Thus we need only establish the Chapman-Kolmogorov relation to show that  $(P_t)_{t \geq 0}$  is a semi-group of transition probabilities.

LEMMA 1.

$$P_{t_1+t_2}(x, A) = \int_0^\infty P_{t_1}(x, ds)P_{t_2}(s, A).$$

*Proof.* — For  $x < t_1$ .

$P_{t_1+t_2}(x, A) = \Pr(\text{the excess time for the } \delta_x\text{-process at } t_1 + t_2 \in A)$ . Conditioning on the excess time at  $t_1$  we have

$$\begin{aligned} P_{t_1+t_2}(x, A) &= \int_0^\infty \Pr(\text{excess time for } \delta_x\text{-process at } t_1 + t_2 \in A \mid \text{excess time} \\ &\hspace{15em} \text{for } \delta_x\text{-process at } t_1 = s) dY^{t_1-x}(s) \\ &= \int_0^{t_2} \Pr(\text{excess time for } \delta_s\text{-process at } t_2 \in A) dY^{t_1-x}(s) \\ &\quad + \int_{t_2}^\infty \chi_{\{s \in t_2 + A\}} dY^{t_1-x}(s) \text{ by definition of } P_{t_2}(s, A). \end{aligned}$$

Finally we notice that if  $I_y = [0, y]$  then  $P_t(x, I_y) = Y^{t-x}(y)$ .  $x < t$ . Thus for  $x < t$

$$P_{t_1+t_2}(x, A) = \int_0^\infty P_{t_2}(s, A)P_{t_1}(x, ds).$$

For  $t_1 \leq x < t_1 + t_2$ .

$$\begin{aligned} P_{t_1+t_2}(x, A) &= \Pr(\text{the excess time for } \delta_{x-t_1}\text{-process at } t_2 \in A) \\ &= P_{t_2}(x - t_1, A) \\ &= \int_0^\infty \delta_{x-t_1}(ds)P_{t_2}(s, A). \end{aligned}$$

Now for  $x \geq t_1$ ,  $P_{t_1}(x, x - t_1) = 1$  by definition and we see that for  $t_1 \leq x < t_1 + t_2$

$$P_{t_1+t_2}(x, A) = \int_0^\infty P_{t_2}(s, A)P_{t_1}(x, ds).$$

Lastly for  $t_1 + t_2 \leq x$ .

$$\begin{aligned} P_{t_1+t_2}(x, A) &= 1 && \text{if } x - t_1 - t_2 \in A \\ &= 0 && \text{if } x - t_1 - t_2 \notin A. \end{aligned}$$

Now  $P_{t_1}(x, ds) = \delta_{x-t_1}(ds)$ , which gives

$$\int_0^\infty P_{t_2}(s, A)P_{t_1}(x, ds) = P_{t_2}(x - t_1, A).$$

Since  $x - t_1 \geq t_2$ , we have

$$\begin{aligned} P_{t_2}(x - t_1, A) &= 1 && \text{if } (x - t_1) - t_2 \in A \\ &= 0 && \text{if } (x - t_1) - t_2 \notin A. \end{aligned}$$

This is precisely  $P_{t_1+t_2}(x, A)$ . Thus for  $x \geq t_1 + t_2$

$$P_{t_1+t_2}(x, A) = \int_0^\infty P_{t_2}(s, A)P(x, ds),$$

and we have the result for all  $x$ .

Given the initial distribution on the state space (or delay distribution if you like) we may now construct a Markov process  $(X_t)_{t \geq 0}$  defined on a probability space  $(\Omega, \mathcal{F}, P_G)$ , where  $G$  is the delay distribution, such that

$$P_G(X_{t_n} \in A / X_{t_0} = x_0, \dots, X_{t_k} = x_k, t_0 < \dots, t_k < t_n) = P_{t_n - t_k}(x_{t_k}, A).$$

We note in passing that

$$P_G(X_t \in [0, y]) = Y_G^t(y).$$

We are interested in the limits of these distributions as  $t$  goes to infinity.

We approach this problem by looking at the properties of the harmonic

functions on the related space-time process. We pick an arbitrary sequence  $I = \{t_n\}_{n=0}^\infty$   $0 = t_0 < t_1 < t_2 < \dots < t_n \dots t_n \rightarrow \infty$ . The random variables  $X_{t_n}$  are by construction all measurable with respect to the measure space  $(\Omega, \mathcal{F}, P_G)$  and we define  $\mathcal{F}^n$  to be the  $\sigma$ -field generated by  $\{X_{t_m}, m \geq n\}$ ,

and let  $\mathcal{F}^\infty = \bigcap_{n=0}^\infty \mathcal{F}^n$ . A random variable  $Y$  is (following [4]) a tail random variable if there exists a sequence  $(f_n)$  of  $\mathcal{F}$ -measurable functions on  $\Omega$  such that  $Y = f_n(X_{t_n}, X_{t_{n+1}}, \dots)$ . If  $f$  can be chosen independently of  $n$  so that

$$Y = f(X_{t_n}, X_{t_{n+1}}, \dots)$$

then  $Y$  is called invariant. If  $Y = \chi_A$  where  $Y$  is a tail random variable (respectively an invariant random variable) then  $A$  is called a tail event (respectively an invariant event). The class of all tail events (invariant events) form a  $\sigma$ -field called the tail  $\sigma$ -field (the invariant  $\sigma$ -field). We will call the pair  $(X_{t_n}, t_n)_{n=0}^\infty$ , the space-time chain associated with  $(X_{t_n})$  and we see that we can construct a Markov chain on  $\mathcal{R}_+ \times I$  with transition probabilities

$$\tilde{P}((x, t_n), \{A, t_{n+1}\}) = P_{t_{n+1}-t_n}(x, A), \quad x \in \mathcal{R}_+, t_n, t_{n+1} \in I, A \in \mathcal{B}(\mathcal{R}_+).$$

A real valued measurable function  $g$  on  $\mathcal{R} \times I$  is called space-time harmonic if

$$g(x, t_n) = \int P_{t_{n+1}-t_n}(x, ds) g(s, t_{n+1}) \quad x \in \mathcal{R}_+, t_n, t_{n+1} \in I.$$

We should note that the restriction of our space-time process to the times  $t$  is not obligatory and is only done to avoid certain measurability questions and to apply more easily the theorems in [4].

We now examine the space-time harmonic functions which are bounded.

LEMMA 2. — If  $h$  is a bounded space-time harmonic function then

$$h(x, t_n) = h(x - (t_{n+1} - t_n), t_{n+1}), \quad x \geq t_{n+1} - t_n.$$

*Proof.* — A space-time harmonic function satisfies

$$\begin{aligned} h(x, t_n) &= \int_0^\infty h(s, t_{n+1}) \tilde{P}((x, t_n), ds \times t_{n+1}) \\ &= \int_0^\infty h(s, t_{n+1}) P_{t_{n+1}-t_n}(x, ds). \end{aligned}$$

Now for  $t_{n+1} - t_n \leq x$ ,  $P_{t_{n+1}-t_n}(x, ds) = \delta_{x-(t_{n+1}-t_n)}(ds)$ . Therefore

$$h(x, t_n) = h(x - (t_{n+1} - t_n), t_{n+1}) \quad \text{for} \quad t_{n+1} - t_n \leq x.$$

We may extend this equality right along the diagonal.

We need now only consider  $h(x, 0)$  to establish results on all space-time.

LEMMA 3. — (a) If  $F$  is periodic with period  $p$   $h(x, 0)$  is constant on the periods of  $F$ . That is  $h(x, 0) = h(x + p, 0) \forall x \geq 0$ .

(b) If  $F$  is aperiodic  $h(x, 0)$  is almost surely constant with respect to Lebesgue measure.

(c) If  $F$  is aperiodic and there exists some convolution  $F^{*n}$  of  $F$  which is not singular with respect to Lebesgue measure then  $h(x, 0)$  is constant.

*Proof.*

$$\begin{aligned} h(0, t_n) &= \int_0^\infty \tilde{P} \{ (0, t_n), ds \times t_{n+1} \} h(s, t_{n+1}) \\ &= \int_0^\infty P_{t_{n+1}-t_n}(0, ds) h(s, t_{n+1}). \end{aligned}$$

From the equality along the diagonals we have

$$h(t_n, 0) = \int_0^\infty P_{t_{n+1}-t_n}(0, ds) h(s + t_{n+1}, 0).$$

Letting  $q(x) = h(x, 0)$  we have

$$q(t_n) = \int_0^\infty P_{t_{n+1}-t_n}(0, ds) q(s + t_{n+1})$$

where  $q(t_n)$  is measurable and bounded. Since the  $t_n$  were chosen arbitrarily we have in general that

$$q(x) = \int_0^\infty P_t(0, ds) q(x + t + s) \quad \forall t, x \geq 0. \quad (1)$$

(a) Now if  $F$  has period  $p$  we may set  $t = p$  to get  $q(x) = \int_0^\infty q(x+s) dF(s)$  and by Choquet-Deny [5] we have  $q(x)$  is constant on the periods of  $F$ .

(b) If  $F$  is aperiodic we remark that by regularization of (Eq. 1) we have functions  $q^\varepsilon(x)$  which are bounded, uniformly continuous solutions of (Eq. 1) and which tend almost surely to  $q(x)$  (w. r. t. Lebesgue measure) as  $\varepsilon \rightarrow 0$ . Also letting  $t \rightarrow 0$  in  $q^\varepsilon(x) = \int_0^\infty P_t(0, ds) q^\varepsilon(x + t + s)$  we have  $q^\varepsilon(x) = \int_0^\infty q^\varepsilon(x + s) dF(s)$  and again by [5]  $q^\varepsilon(x)$  is constant. Thus  $q(x)$  is constant almost surely.

(c) We note that if  $F^{*n}$  is not singular w. r. t. Lebesgue measure we can pick a  $\bar{t}$  sufficiently big so that  $F_e^{*n}(\bar{t}) > 0$  ( $F_e^{*n}$  is the absolutely conti-

nuous part of  $F^{**}$ ). Hence  $P_{\tau}(0, ds)$  is not singular and we may employ the method in Note 1 to prove that  $q(x)$  is constant.

We remark that the aperiodic distribution  $F$  giving mass  $\frac{1}{2}$  to 1 and  $\sqrt{2}$  and the function

$$\begin{aligned} h(x, 0) &= 1 & \text{if } x &= p + q\sqrt{2} \text{ } p, q \text{ positive integers} \\ &= 0 & \text{if } x &\neq p + q\sqrt{2} \text{ for any } p, q. \end{aligned}$$

generate a counter example to an extension of Lemma 3 (b).

It is now useful to distinguish between the cases when  $F$  is arithmetic (without loss of generality having period 1) and when  $F$  is non-arithmetic. Henceforth if  $F$  is arithmetic the  $t_n$  and  $x$ 's will be restricted to the positive integers. Moreover for the arithmetic case we will define the measure  $m$  to be the counting measure on the positive integers while in the non-arithmetic case  $m$  will be Lebesgue measure. With this in mind we note that the distribution,

$$\begin{aligned} E(r) &= \frac{1}{\mu} \int_0^r (1 - F(y))m(dy) \\ \mu &= \text{mean of } F \end{aligned}$$

called the equilibrium distribution, gives a stationary delayed renewal process [6]. Hence  $Y_E^t(r) = E(r) \forall t \geq 0$ , and see  $X^t$  is stationary w. r. t.  $P_E$ .

The utility of the space-time chain is seen in the following.

PROPOSITION 1. — The following conditions are equivalent.

(a) For all probability measures  $\mu$  and  $\nu$  on  $\mathcal{B}(\mathcal{R}_+)$

$$\lim_{n \rightarrow \infty} \|\mu P_{t_n} - \nu P_{t_n}\| \rightarrow 0. \quad \text{Where } \mu P(A) = \int P(x, A)\mu(dx) \text{ for } A \in \mathcal{B},$$

and  $\|\nu\|$  is the total variation of  $\nu$ .

(b) The only bounded space-time harmonic functions are constants.

*Proof.* — The proof is an adaptation of the proof for Proposition 4.3 in [4].

THEOREM 1. — If (a)  $F$  is arithmetic and  $\alpha$  is any probability measure on the positive integers or if (b)  $F^{**}$  is not singular w. r. t. Lebesgue measure for some  $n$  and  $\alpha$  is a probability measure on  $[0, \infty)$  then  $\lim_{t \rightarrow \infty} \|\alpha P_t - e\| = 0$  (in the arithmetic case  $t$  is an integer), where  $e$  is the equilibrium measure having distribution  $E(r)$ .

*Proof.* — In the arithmetic case we restrict attention to the positive integers. Hence by Lemmas 3 (a) and 3 (c) respectively we see the bounded space-time harmonic functions are constant. Hence by Proposition 1, we have for probability measures  $\alpha$  and  $e$

$$\lim_{t \rightarrow \infty} \|\alpha P_t - e P_t\| = 0.$$

However

$$\begin{aligned} e P_t(A) &= \int_0^\infty P_t(x, A) dE(x) \\ &= e(A) \end{aligned}$$

and we have the result. Q. E. D.

COROLLARY 1 (Feller). — If  $F$  is arithmetic with period 1 and  $F(0) = 0$  then

$$\lim_{n \rightarrow \infty} \Pr \{ \text{renewal at } n \} = 1/\mu (\mu \text{ is mean of } F)$$

*Proof.*

$$\Pr \{ \text{renewal at } n \} = P_n(0, 0).$$

Thus

$$\lim_{n \rightarrow \infty} \{ \text{renewal at } n \} = \frac{1}{\mu} e \{ 0 \} = 1/\mu, \text{ from Theorem 1.}$$

If  $F$  is aperiodic but not absolutely continuous w. r. t. Lebesgue measure we must be a little more subtle.

PROPOSITION 2. — If  $F$  is aperiodic and if  $G$  is a probability measure which is absolutely continuous with respect to Lebesgue measure ( $G \ll m$ ) then the tail field of  $\{X_{t_n}\}$  is trivial w. r. t.  $P_G$ .

*Proof.* — Let  $A$  be a tail event of  $\{X_{t_n}\}$ . Then by Proposition 4.1 of [4]  $A$  is an invariant event of  $\{X_{t_n}, t_n\}$ . Consider the function defined on space-time by

$$h_A(Z) = \tilde{E}_Z(A) \quad (Z \text{ of the form } Z = (x, t_n)).$$

By Proposition 4.2 of [4]  $h_A(Z)$  is bounded and space-time harmonic. Also

$$h_A(X_n, t_n) = \tilde{E}_{[X_n, t_n]}[A] = \tilde{E}_\eta[\theta^n A \mid \mathcal{F}_n] = \tilde{E}_\eta[A \mid \mathcal{F}_n]$$

(where again  $\eta$  is the initial distribution given by

$$\eta \{ A x t_0 \} = G(A).$$

Thus  $h_A(X_n, t_n)$  converges a. s. w. r. t.  $\tilde{P}_\eta$  to  $A$ . Moreover by Lemma 3 (b)



we know  $h_A(Z) = C$  (constant) a. s. w. r. t. Lebesgue measure.

$$\text{Let } B = \{ (x, t) \mid h(x, t) \neq C \}.$$

Since  $G \ll m$  we have  $\tilde{P}_\eta \{ (X_n, t_n) \in B \} = 0$ . Thus a. s.  $\tilde{P}_\eta h_A(X_n, t_n)$  converges to  $C$  and hence  $A$  is trivial w. r. t.  $\tilde{P}_\eta$ . Thus  $A$  is trivial w. r. t.  $P_G$ .

**THEOREM 2.** — If  $F$  is aperiodic and if  $\alpha$  is a probability measure which is absolutely continuous w. r. t. Lebesgue measure then

$$\lim_{t_n \rightarrow \infty} \| \alpha P_{t_n} - e \| = 0.$$

*Proof.* — By Proposition 2, the tail field of  $\{ X_{t_n} \}$  is trivial w. r. t.  $P_e$  (since the equilibrium distribution  $e \ll m$ ). Hence as in Theorem 4.1 of [4].

$$\lim_{t_n \rightarrow \infty} \sup_{A \in \mathcal{F}^n} | P_e(A \cap B) - P_e(A)P_e(B) | = 0 \quad \text{for every } B \in \mathcal{F}. \quad (2)$$

$$\text{Let } A = \{ X_{t_n} \in F \}; \quad B = \{ X_{t_0} \in G \}.$$

Thus we have

$$\lim_{t_n \rightarrow \infty} \sup_F \left| \int_G P_{t_n}(x, F) e(dx) - e(F)e(G) \right| = 0.$$

Now  $\alpha \ll m$ . Also  $m \ll e$  on  $\{ x \mid F(x) < 1 \}$  by the construction of  $e$ . Moreover by truncation there exists a probability measure  $\tilde{\alpha} \ll m$  concentrated on  $[0, T]$  such that  $\| \tilde{\alpha} - \alpha \| < \varepsilon$  for any  $\varepsilon$ . Hence

$$\| \tilde{\alpha} P_{t_n} - \alpha P_{t_n} \| < \varepsilon.$$

Also by the construction of our chain we see  $\tilde{\alpha} P_T$  is concentrated on  $\{ x \mid F(x) < 1 \}$ , and hence for  $t_n$  sufficiently large  $\tilde{\alpha} P_{t_n} \ll e$ . We may therefore consider the case  $\alpha \ll e$ . Let  $\mathcal{F}(x) = \frac{d\alpha}{de}(x)$  and hence  $\int \mathcal{F} de = 1$ . Let  $\mathcal{F}_k(x)$  be step functions such that  $\mathcal{F}_k \uparrow \mathcal{F}$ . Thus by Eq. 2 we have

$$\lim_{t_n \rightarrow \infty} \sup_F \left| \int P_{t_n}(x, F) \mathcal{F}_k(x) e(dx) - \int \mathcal{F}_k(x) e(dx) \cdot e(F) \right| = 0.$$

However we remark that

$$\begin{aligned} & \left| \int P_{t_n}(x, F) \mathcal{F}_k(x) e(dx) - \int P_{t_n}(x, F) \mathcal{F}(x) e(dx) \right| \\ & \leq \int P_{t_n}(x, F) | \mathcal{F}_k(x) - \mathcal{F}(x) | e(dx) \leq \int | \mathcal{F}_k(x) - \mathcal{F}(x) | e(dx). \end{aligned}$$

This uniform bound implies

$$\lim_{t_n \rightarrow \infty} \sup_F \left| \int P_{t_n}(x, F) \mathcal{F}(x) e(dx) - \int \mathcal{F}(x) e(dx) \cdot e(F) \right| = 0.$$

and 
$$\lim_{t_n \rightarrow \infty} \sup_F \left| \int P_{t_n}(x, F) \alpha(dx) - e(F) \right| = 0$$

and 
$$\lim_{t_n \rightarrow \infty} \sup_F | \alpha P_{t_n}(F) - e(F) | = 0$$

and hence

$$\lim_{n \rightarrow \infty} \| \alpha P_{t_n} - e \| = 0.$$

COROLLARY 2. — For F aperiodic  $Y^t(x)$  converges weakly to  $E(x)$ .

*Proof.* — For any interval  $[x, y]$  we pick  $0 < \varepsilon < x$  and by considering the possible trajectories of our process, as well as the translation invariance of our transition probabilities we see

$$P_{t_n}(s, [x + \varepsilon, y]) \leq P_{t_n}(0, [x, y]) \leq P_{t_n}(s, [x, y + \varepsilon]) + P_{t_n}(s, [0, \varepsilon])$$

for  $0 \leq s \leq \varepsilon$ . Consider the uniform probability measure  $u_\varepsilon$  on  $[0, \varepsilon]$ . Integrating our inequality we have

$$u_\varepsilon P_{t_n}[x + \varepsilon, y] \leq P_{t_n}(0, [x, y]) \leq u_\varepsilon P_{t_n}[x, y + \varepsilon] + u_\varepsilon P_{t_n}[0, \varepsilon].$$

Hence by Theorem 2. We have for all  $\varepsilon < x$

$$e[x + \varepsilon, y] \leq \underline{\lim} P_{t_n}(0, [x, y]) \leq \overline{\lim} P_{t_n}(0, [x, y]) \leq e[x, y + \varepsilon] + e[0, \varepsilon].$$

Thus by the continuity of  $E(x)$  we have  $\lim_{t_n \rightarrow \infty} P_{t_n}(0, [x, y]) = e[x, y]$  which implies

$$\lim_{t_n \rightarrow \infty} Y^{t_n}(x) = E(x).$$

COROLLARY 3 (Blackwell). — If F is aperiodic then  $\lim_{t \rightarrow \infty} u(t+h) - u(t) = h/\mu$ .

*Proof.* —  $u(t+h) - u(t)$  is the expected number of renewals in  $[t, t+h]$ . If we condition on the excess time at  $t$  we have

$$u(t+h) - u(t) = \int_0^{h-} \{1 + u(h-s)\} dY^t(s)$$

(if we step over  $t$  and land at  $t+s < t+h$  then we restart the process at  $t+s$  and we take on the average  $u(h-s)$  more steps before  $t+h$ ). Next we remark that  $1 + u(h-s)$  is a decreasing function of  $s$  and hence

has a countable number of discontinuities on  $[0, h]$ . Moreover by Corollary 2,  $Y^t$  converges weakly to  $E$  and  $E \ll m$ . Hence the set of discontinuities of  $1 + u(h - s)$  has measure 0 w. r. t.  $E$  and we have

$$\lim_{t \rightarrow \infty} u(t + h) - u(t) = \int_0^h (1 + u(h - s)) dE(s)$$

(see Theorem 5.2 (iii) in [7] for example).

However since  $E$  is the equilibrium measure

$$\int_0^h (1 + u(h - s)) dE(s) = V_E(h) = h/\mu \quad (\text{see } [6]).$$

*Remarks.* — If  $F$  has infinite mean we have no equilibrium probability measure. We still have however an invariant measure  $e$  with distribution

$$E(x) = \int_0^x (1 - F(s)) m(ds).$$

We have for every  $\gamma > 0$  and every  $x \in R_+$

$$\frac{P_{t_n}(x, F)}{e(F) + \gamma} \rightarrow 0 \quad \text{uniformly in } F \in \mathcal{B}(R_+) \quad \text{as } t_n \rightarrow \infty.$$

(See Theorem 7.3 in [4]). We have all the obvious extensions of Corollaries 1, 2 and 3.

If our distribution  $F$  is not concentrated on the half line but has  $\mu > 0$  we can still consider the excess time at  $t \in R_+$  (since the walk drifts to the right) and we can prove our theorems based on the strict ladder distribution [8].

NOTE 1. — If  $h$  is measurable and bounded and satisfies

$$h(x) = \int_0^\infty h(x + y) dF(y) \tag{3}$$

and if  $F^{*n}$  is not singular (w. r. t. Lebesgue measure), then  $h(x)$  is constant.

*Proof.* — By Choquet and Deny's lemma  $h(x)$  is almost surely constant (say  $C$ ). Subtracting  $C$  from both sides of (Eq. 3) we have

$$g(x) = \int_0^\infty g(x + y) dF(y) \tag{4}$$

where  $g(x) = h(x) - C$  is almost surely 0.

By convolution of (Eq. 4) we have

$$g(x) = \int_0^\infty g(x + y) dF^{*n}(y).$$

Now  $F^{*n} = G = pG_e + qG_d$  where  $G_e$  generates a measure absolutely continuous with respect to Lebesgue measure and  $G_d$  generates a measure singular with respect to Lebesgue measure [7] and  $p > 0, q > 0, p + q = 1$ .

Now

$$\begin{aligned} g(x) &= p \int_0^\infty g(x+y) dG_e(y) + q \int_0^\infty g(x+y) dG_d(y) \\ &= q \int_0^\infty g(x+y) dG_d(y). \end{aligned}$$

since  $g(x) = 0$ . a. e.

Again using

$$g(x+y) = \int_0^\infty g(x+y+y_1) dG(y_1)$$

we have

$$\begin{aligned} g(x) &= q^2 \int_0^\infty g(x+y) dG_d^{*2}(y) \\ &= q^n \int_0^\infty g(x+y) dG_d^{*n}(y) \downarrow 0. \end{aligned}$$

Thus  $g(x) = 0$  everywhere and hence  $h(x) = C$  everywhere.

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#### BIBLIOGRAPHY

- [1] W. FELLER, *An introduction to probability theory and its applications*. Vol. 2, p. 372.
- [2] *Ibid.*, p. 368.
- [3] *Ibid.*, p. 188.
- [4] F. OREY, *Lectures notes on limit theorems for Markov chain transition probabilities*. Van Nostrand Reinhold, 1971.
- [5] G. CHOQUET and J. DENY, *C. R. Acad. Sci. Paris*, Vol. 250, 1960, p. 799-801.
- [6] W. FELLER, Vol. 2, p. 369.
- [7] P. BILLINGSLEY, *Convergence of probability measures*, p. 30-31.
- [8] W. FELLER, Vol. 2, p. 391.

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