

A NEW APPROACH TO THE LIMIT THEORY OF RECURRENT MARKOV CHAINS

BY

K. B. ATHREYA AND P. NEY

ABSTRACT. Let $\{X_n; n > 0\}$ be a Harris-recurrent Markov chain on a general state space. It is shown that there is a sequence of random times $\{N_i; i > 1\}$ such that $\{X_{N_i}; i > 1\}$ are independent and identically distributed. This idea is used to show that $\{X_n\}$ is equivalent to a process having a recurrence point, and to develop a regenerative scheme which leads to simple proofs of the ergodic theorem, existence and uniqueness of stationary measures.

1. Introduction. The purpose of this paper is to introduce a new tool for the study of the limit theory of a recurrent Markov chain (M.C.) $\{X_n; n \geq 0\}$ on a measurable space (S, \mathfrak{S}) . When there is a single point x_0 , called a *recurrence point*, which is visited infinitely often (i.o.) by $\{X_n\}$, a well-known approach to its limit theory is via the imbedded renewal process of returns to x_0 . The key idea in the present work is the observation that *a similar regenerative scheme exists for general recurrent chains*, in which the role of x_0 is played by a set which is *hit i.o. with the same distribution*. In fact, it is possible to adjoin a point Δ to S and to extend $\{X_n\}$ to $S \cup \Delta$ in such a way that Δ is a recurrence point for the new process, and that the two processes are equivalent.

We will use this idea to give an elementary proof of the ergodic theorem, and to establish the existence and uniqueness of stationary measures. Extensions to semi-Markov processes and applications to renewal theory will be treated in [1]. Comprehensive background discussions on recurrent chains are available in the books of Doob [3], Neveu [7], Orey [8] and Revuz [9].

2. Recurrence. An irreducible chain having a recurrence point x_0 is recurrent if it returns to x_0 with probability one. If such a point does not exist, the notion of recurrence can be captured by requiring that certain sets which are "nonnegligible" with respect to a reference measure are visited "often enough". This idea was introduced by Doeblin [2], in his proof of the ergodic theorem for $\{X_n\}$. He hypothesized that there exist a probability measure φ on (S, \mathfrak{S}) , numbers $\varepsilon > 0$, $\delta < 1$, and an integer $n_0 < \infty$, such that for all

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$x \in S, E \in \mathfrak{S}$

$$P^{n_0}(x, E) \geq \epsilon \text{ whenever } \varphi(E) \geq \delta. \tag{2.1}$$

A slightly stronger version was introduced by Doob [3]. Under such hypotheses it is possible to prove a strong (operator) convergence of $\{P^n\}$ (see e.g. Neveu [7, §5.3], Revuz [9, §6.3]). We will return to this point in §4.

A weaker condition, introduced by Harris [6], is the so-called φ -recurrence or Harris-recurrence, which requires the existence of a σ -finite measure φ on (S, \mathfrak{S}) such that $P_x\{X_n \in A \text{ for some } n\} = 1$ for all $A \in \mathfrak{S}$ with $\varphi(A) > 0$. Under this hypothesis much of the discrete state space theory has been carried over to the general case by Harris [6], Orey [8], and others.

Our regeneration technique becomes particularly transparent if we express recurrence in yet another form.

(2.2) DEFINITION. $\{X_n\}$ is $(A, \lambda, \varphi, n_0)$ -recurrent if there exist a set $A \in \mathfrak{S}$, a probability measure φ on A , a number $0 < \lambda < \infty$, and an integer $0 < n_0 < \infty$, such that

- (i) $P_x\{X_n \in A \text{ for some } n \geq 1\} = 1$ for all $x \in S$, and
- (ii) $P_x\{X_{n_0} \in E\} \geq \lambda\varphi(E)$ for all $x \in A$ and $E \subset A$.

REMARKS. (2.3) By making use of the existence of C -sets (see Orey [8]) it is easy to show that (2.2) is in fact equivalent to Harris-recurrence.

(2.4) Neveu [7, p. 185] has observed that if (ii) in (2.2) is satisfied with $A = S$, then the Doeblin condition is satisfied.

In this paper we wish to focus on the regeneration idea in its simplest setting, and to that end we will treat only the case $n_0 = 1$ (in (2.2)), in which case we call the chain *recurrent and strongly aperiodic*. The extension to the general case offers no serious difficulties, and involves typical periodicity and cyclic class arguments. We leave this to a more leisurely exposition of this subject which we are planning, which will also include some applications, discussion of related literature, and some other ramifications. *The set A and the measure φ are fixed throughout.*

3. Regeneration. Our main tool will be the following:

(3.1) REGENERATION LEMMA. *If $\{X_n\}$ is recurrent and strongly aperiodic then there exists a random time $N \geq 1$ such that $P_x(N < \infty) = 1$, and for all $x \in S, E \in \mathfrak{S}$*

$$P_x(X_n \in E, N = n) = \varphi(A \cap E)P_x(N = n). \tag{3.2}$$

REMARK. As will be clear from the proof, N is actually defined on a somewhat enlarged state space. It is nonanticipating for $\{X_n\}$ but is not a stopping time for this process.

Here is the idea of the proof. If the chain hits A at a point x , at time k randomize the next transition as follows: (i) with probability p ($0 < p \leq \lambda$) distribute X_{k+1} over A according to φ ; (ii) with probability $(1 - p)$ distribute

X_{k+1} over the entire state space S according to a transition function $Q(x, \cdot)$, chosen so that the overall transition probabilities for the chain remain unchanged. This is achieved by taking Q so that

$$P(x, E) = p\varphi(E \cap A) + (1 - p)Q(x, E), \quad x \in A, E \in \mathfrak{S}, \quad (3.3)$$

which is possible by (ii) of Definition 2.2. Since A is visited i.o., and each time there is (independent) probability $p > 0$ that at the next step “ A is entered according to φ ”, this event will ultimately occur at some time $N < \infty$ a.s. Here is a more formal

PROOF OF THE LEMMA. Let $S' = S \times \{0, 1\}$, $\mathfrak{S}' =$ the σ -algebra of subsets of S' induced by \mathfrak{S} , and P' be a transition function on $S' \times \mathfrak{S}' \rightarrow [0, 1]$ defined by

$$P'\{(x, 0), E \times \{\delta\}\} = \begin{cases} [p\delta + (1 - p)(1 - \delta)]P(x, E) & \text{if } x \notin A, \\ [p\delta + (1 - p)(1 - \delta)]Q(x, E) & \text{if } x \in A, \end{cases}$$

$$P'\{(x, 1), E \times \{\delta\}\} = \begin{cases} [p\delta + (1 - p)(1 - \delta)]P(x, E) & \text{if } x \notin A, \\ [p\delta + (1 - p)(1 - \delta)]\varphi(A \cap E) & \text{if } x \in A, \end{cases}$$

where $\delta = 0$ or 1 and $Q(\cdot, \cdot)$ is as in (3.3). Let $\{X'_n = (X_n, \delta_n): n = 0, 1, 2, \dots\}$ be a Markov chain on (S', δ') with transition function P' . Then one can check that

- (i) $\{X_n: n \geq 0\}$ is an M.C. on (S, \mathfrak{S}) with transition function P ;
- (ii) $\{\delta_n: n \geq 0\}$ is a sequence i.i.d. 0-1 random variables with $P(\delta_n = 1) = p$;
- (iii) $\{X_n\}_0^\infty$ and $\{\delta_n\}_0^\infty$ are independent.

Define $N = \inf\{n: n \geq 1, X'_{n-1} \in Ax\{1\}\}$, $= \infty$ if $\{-\}$ is empty. It remains only to observe that $P'_{(x,\delta)}\{N < \infty\} = 1$. To see this notice that

$$P'_{(x,\delta)}(N = \infty) = P'_{(x,\delta)}(N = \infty, X_n \in A \text{ for } \infty \text{ many } n) \\ = P'_{(x,\delta)}(\delta_{L_i} = 0 \text{ for all } i),$$

where L_1, L_2, \dots are the successive hitting times of A by $\{X_n\}$. The last probability $= 0$ due to the independence of $\{X_n\}$ and $\{\delta_n\}$. \square

REMARK. Note that the random time N is a stopping time for $\{X'_n\}$.

We recall that X_N is distributed over A according to φ . Reapplying the regeneration lemma, there exists a regeneration time M for $\{X_n\}$ initiated with $X_0 = X_N$, i.e. with distribution φ . Let $N = N_1, N + M = N_2$. Continuing in this way, we use the lemma to observe that under $(A, \lambda, \varphi, 1)$ -recurrence we have

(3.4) **COROLLARY.** *There exists a sequence of random times N_1, N_2, \dots for X_n such that $P_x(N_k < \infty) \equiv 1$ for all $k > 1$, and*

$$P_x(X_n \in E, N_k = n) = \varphi(E \cap A)P_x(N_k = n). \quad (3.5)$$

Here $P_x(N_1 = n) = P_x(N = n)$ as in (3.2), while for $k \geq 2$, $N_k = N_1 + M_2 + \dots + M_k$, where the M_i 's are i.i.d. (and independent of N_1), and $P_x(M_i = n) = P_\varphi(N = n)$, $x \in S$, $n \geq 1$.

Thus $\{X_{N_i}; i \geq 1\}$ are independently, identically distributed according to φ , and in this sense the process regenerates itself at $\{N_i\}$.

There is some similarity between the kind of randomization (or smoothing) which goes on in our lemma, and a coupling technique for MC.'s which goes back to Doeblin [2], and has been developed by Griffeath [5].

4. Ergodicity. Let \mathfrak{B} denote the Banach space of bounded measurable functions $f: S \rightarrow R$, with sup norm $\|f\|$. If μ is a finite measure on (S, \mathfrak{S}) and $f \in \mathfrak{B}$, let $\mu f = \int f(x) \mu(dx)$, $(\mu P)(x) = \int \mu(ds) P(s, x)$, $\|\mu\| =$ total variation norm of μ . P^n is defined as an operator on \mathfrak{B} by the formula $(P^n f)(x) = E_x f(X_n)$. Let $\|P\|$ denote its norm. Let T_B denote the first hitting time of any set $B \in \mathfrak{S}$ by $\{X_n\}$; $N =$ the random time in Lemma (3.1).

(4.1) THEOREM. (i) If $\{X_n\}$ is recurrent and strongly aperiodic, with invariant probability measure $\pi(\cdot)$, then for all $x \in S$

$$\|P^n(x, \cdot) - \pi(\cdot)\| \rightarrow 0 \text{ as } n \rightarrow \infty.$$

(ii) If furthermore $\sup_x P_x(T_A > t) < 1$ for some $t < \infty$, then $\|P^n - \pi\| \rightarrow 0$.

PROOF OF (i). Let $v_n = E_\varphi f(X_n)$, $a_n = E_\varphi \{f(X_n); N > n\}$, $p_n = P_\varphi(N = n)$, $m = E_\varphi N$. The condition $n_0 = 1$ implies that $\{p_n\}$ is aperiodic, and hence by the renewal theorem $u_n \equiv \sum_k p_n^{*k} \rightarrow m^{-1}$ as $n \rightarrow \infty$. But by the regeneration lemma, v_n satisfies the renewal equation $v_n = a_n + \sum_{j=0}^{n-1} p_j v_{n-j}$, and hence

$$v_n = \sum_{j=0}^n a_{n-j} u_j \rightarrow m^{-1} \sum_0^\infty a_k, \tag{4.2}$$

provided $\sum |a_k| < \infty$. To see this let $N = N_1, N_2, \dots$ denote the regeneration times, and $K_n = \sup\{k: N_k \leq n\}$. Since $(S_{K_n}/K_n) \leq (n/K_n) \leq (S_{K_n+1}/K_n)$, the law of large numbers implies that $K_n/n \rightarrow (E_\varphi N)^{-1}$ a.s., and since $0 < K_n/n \leq 1$, $E_\mu(K_n/n) \rightarrow (E_\varphi N)^{-1}$ for any initial distribution $\mu(\cdot)$. Now, K_n can be written as $\sum_{i=1}^n \delta_i$, where $\delta_i = 1$ or 0 depending on whether a regeneration occurs at i or not. Since $P_\pi(\delta_i = 1) = p P_\pi(X_{i-1} \in A) = p \pi(A)$, $E_\pi K_n = n p \pi(A)$. Furthermore the fact that $P_x(X_n \in A \text{ for some } n \geq 1) \equiv 1$ implies that $P_\pi(X_n \in A \text{ for some } n \geq 1) = 1$, and since $P_\pi(X_n \in A) \equiv \pi(A)$ for all n by stationarity, we see that $\pi(A)$ has to be nonzero. Thus, $E_\varphi(N) = (p \pi(A))^{-1} < \infty$. Thus also $\sum |a_n| \leq \|f\| \sum_n P_\varphi(N > n) < \infty$, proving (4.2).

Now let x be any initial state. Since $\lim v_n$ exists and

$$E_x f(X_n) = E_x \{f(X_n); N > n\} + E_x \{v_{n-N}; N \leq n\}, \tag{4.3}$$

we see that $\lim E_x f(X_n)$ exists, and hence by the bounded convergence theorem $\lim E_\mu f(X_n) = \lim v_n$ for any initial distribution μ . Choosing μ to be π , we conclude that

$$E_x f(X_n) \rightarrow m^{-1} \sum a_k = \pi f \quad \text{for all } x \in S. \tag{4.4}$$

We can strengthen the convergence to total variation norm. Let $\epsilon_K = \sup\{|u_j - m^{-1}|: j > K\}$, and note that $0 < u_n < 1$ and $0 < m^{-1} < 1$. Thus

$$\begin{aligned} |(\varphi P^n - \pi)f| &= |v_n - \pi f| < \sum_0^n |a_{n-j}| \cdot |u_j - m^{-1}| + m^{-1} \sum_{j>n} |a_j| \\ &< \sum_{j>n-K} |a_j| + \epsilon_K \sum_{j<n-K} |a_j| < \|f\| \left\{ \epsilon_K + \sum_{j>n-K} P_\varphi(N > j) \right\}. \end{aligned} \tag{4.5}$$

Let $\gamma_n = \sup\{|v_n - \pi f|/\|f\|: 0 < \|f\| < \infty\} = \|\varphi P^n - \pi\|$. Then

$$\gamma_n = \epsilon_K + \sum_{j>n-K} P_\varphi(N > j), \tag{4.6}$$

and taking K large we see that $\gamma_n \rightarrow 0$ as $n \rightarrow \infty$. Finally use (4.3) to observe that

$$|E_x f(X_n) - \pi f|/\|f\| < 2P_x(N > n) + E_x\{|v_{n-N} - \pi f|/\|f\|; N \leq n\}. \tag{4.7}$$

Hence $\|P^n(x, \cdot) - \pi(\cdot)\| < 2P_x(N > n) + E_x\{\gamma_{n-N}; N \leq n\}$, and the bounded convergence theorem then implies (i). \square

PROOF OF (ii). Suppose first that $\sup_x P_x(N > n) \rightarrow 0$ as $n \rightarrow \infty$, and for a given ϵ , choose n_0 so that $\sup_x P_x(N > n) < \epsilon$ and $\gamma_n < \epsilon$ for $n > n_0$. Let $|v_n - \pi f|/\|f\| = \xi_n(f)$. Then for $n > 2n_0$,

$$\begin{aligned} \sup_{x,f} E_x\{\xi_{n-N}(f); N \leq n\} &< \sup E_x\{\xi_{n-N}(f); N \leq n - n_0\} \\ &\quad + \sup E_x\{\xi_{n-N}(f); n - n_0 < N \leq n\} \\ &< \sup_{n>n_0} \gamma_n + \sup_x P_x(N > n_0) < 2\epsilon, \end{aligned}$$

and hence (by (4.7))

$$\begin{aligned} \sup_{x,f} \{|E_x f(X_n) - \pi f|/\|f\|\} \\ < 2 \sup_x P_x(N > n) + \sup_{x,f} E_x\{\xi_{n-N}(f); N \leq n\} < 4\epsilon. \end{aligned}$$

Thus $\|P^n - \pi\| \rightarrow 0$ as $n \rightarrow \infty$. It remains only to observe that the hypothesis of (ii) implies $\sup_x P_x(N > n) \rightarrow 0$. To see this, pick n_1 so that $\sup_x P_x(T_A > n_1) < 1 - \eta$ for some $\eta > 0$. Then

$$\begin{aligned} \delta_n &\equiv \sup_x P_x(N > nn_1) \\ &\leq \sup_x P_x \left\{ \sum_{i=1}^{nn_1} \chi_A(X_i) > k_n, N > nn_1 \right\} + \sup_x P_x \left\{ \sum_{i=1}^{nn_1} \chi_A(X_i) < k_n \right\} \\ &\leq (1 - p)^{k_n} + P\{B_{n,\eta} \leq k_n\}, \end{aligned}$$

where $B_{n,\eta}$ is binomial (n, η) . Now take $k_n = \epsilon n$, with ϵ small enough so that $\delta_n \rightarrow 0$. \square

REMARK. Note that actually $\delta_n \rightarrow 0$ geometrically fast, and hence so does $|u_n - m^{-1}| \rightarrow 0$ (Stone [10]). Thus the convergence $E_x f(X_n) \rightarrow \pi f$ is also geometric.

When an invariant probability measure does not exist, then one still has

(4.8) THEOREM (OREY). *If $\{X_n\}$ is recurrent and strongly aperiodic then for any two initial distributions μ and η ,*

$$\|\mu P^n - \nu P^n\| \rightarrow 0. \tag{4.9}$$

PROOF. Let $f: S \times N^+ \rightarrow R$ be a bounded space-time harmonic function for $\{X_n\}$, i.e. $f(s, m) = E\{f(X_1, m + 1) | X_0 = s\}$. Then $Y_n = f(X_n, m + n)$, $n > 0$, is a bounded martingale and letting $g(n) = E_\varphi f(X_0, n)$, and N be as before,

$$\begin{aligned} f(s, m) &= E_s Y_0 = E_s f(X_N, m + N) \\ &= \sum g(m + k) P_s(N = k) = E_s g(m + N). \end{aligned} \tag{4.10}$$

Integrating with respect to φ we get $g(m) = E_\varphi g(m + N)$. Since $g(\cdot)$ is bounded and N is aperiodic this implies that g is constant (see Feller [4, p. 351]), and hence by (4.9) f is constant. Orey [8, Proposition 4.3] has shown that this in turn implies (4.9). \square

5. Equivalence to a chain with a recurrence point. Adjoin a point Δ to S , and extend $\{X_n\}$ to a M.C. $\{\tilde{X}_n\}$ on $S \cup \Delta$, with transition function \tilde{P} defined by

$$\tilde{P}(x, E) = \begin{cases} P(x, E - \Delta) & \text{for } x \notin A \cup \Delta, \\ (1 - p)Q(x, E - \Delta) + p\chi_E(\Delta) & \text{for } x \in A, \\ (1 - p) \int_A Q(z, E - \Delta) \varphi(dz) + p\chi_E(\Delta) & \text{for } x = \Delta. \end{cases} \tag{5.1}$$

Note that Δ is a recurrence point since $\{\tilde{X}_n\}$ is in A i.o., and $\tilde{P}(x, \Delta) = p > 0$ for $x \in A$. For bounded measurable $f: S \rightarrow R$ define

$$\begin{aligned} \tilde{f}(x) &= f(x) \quad \text{for } x \in S, \\ \tilde{f}(\Delta) &= \int_A f(z) \varphi(dz) = E f(X_N). \end{aligned} \tag{5.2}$$

If $\pi(\cdot)$ is a measure on S , define the measure $\tilde{\pi}(\cdot)$ on $S \cup \Delta$ by

$$\begin{aligned} \tilde{\pi}(E) &= \pi(E) - p\pi(A)\varphi(A \cap E), \quad E \subset S, \\ \tilde{\pi}(\Delta) &= p\pi(A). \end{aligned} \tag{5.3}$$

(5.4) EQUIVALENCE THEOREM. (i) *Let $\{X_n\}$ be recurrent and strongly aperiodic, and $\{\tilde{X}_n\}$ be an M.C. with transition function \tilde{P} . Then for any initial distribution μ on S*

$$E_\mu f(X_n) = E_\mu \tilde{f}(X_n). \tag{5.5}$$

(ii) *If π is a stationary measure for P , then $\tilde{\pi}$ is a stationary measure for P , and conversely. If either π or $\tilde{\pi}$ is unique (up to multiplicative constants), then so is the other. If π or $\tilde{\pi}$ is a probability measure, so is the other.*

The proof is a straightforward verification of definitions and formulas, so we will leave it to the reader.

6. Invariant measures. Harris-recurrent chains are known to have unique, σ -finite, invariant measures (Harris [6], Orey [8]). (Uniqueness will always be understood to be up to a multiplicative constant.) The proofs have been quite involved, but our regeneration scheme now provides an easy alternative.

(6.1) THEOREM. *A recurrent, strongly aperiodic M.C. $\{X_n\}$ has a unique invariant measure. It is finite if and only if $E_\varphi N < \infty$.*

PROOF. By part (ii) of the Equivalence Theorem (5.4) it is sufficient to prove the theorem when $\{X_n\}$ has a recurrence point (say x_0). Note that in this case

$$\bigcup_n \{x: P^n(x, x_0) > 0\} = S. \tag{6.2}$$

Let $N = \inf\{n: n \geq 1, X_n = x_0\}$, and define

$$\nu(E) = E_{x_0} \sum_{i=0}^{N-1} \chi_E(X_i), \quad E \in \mathfrak{S}. \tag{6.3}$$

Clearly $\nu(\cdot)$ is finite iff $E_\varphi N < \infty$.

(i) $\nu(\cdot)$ is an invariant measure. To see this write

$$\begin{aligned} \nu(E) &= \chi_E(x_0) + \sum_{i=1}^{\infty} E_{x_0} \{ \chi_E(X_i); N > i - 1 \} - \sum_{i=1}^{\infty} E_{x_0} \{ \chi_E(X_i); N = i \} \\ &= \chi_E(x_0) + E_{x_0} \left\{ \sum_{i=1}^{\infty} P(X_{i-1}, E) \chi_{(N > i-1)} \right\} - E_{x_0} \chi_E(X_N). \end{aligned}$$

The first and last terms cancel, and the middle term

$$= E_{x_0} \sum_{i=0}^{N-1} P(X_i, E) = \int \nu(dx) P(x, E).$$

(ii) $\nu(\cdot)$ is unique (up to multiplicative constants). Suppose that $\lambda(\cdot)$ is

another invariant measure. Then

$$\begin{aligned} \lambda(E) &= \int_S \lambda(dx)P(x, E) = \lambda(x_0)P(x_0, E) + \int_{S-x_0} \lambda(dx)P(x, E) \\ &= \lambda(x_0)P_{x_0}(X_1 \in E) + \lambda(x_0)P_{x_0}(X_1 \neq x_0, X_2 \in E) \\ &\quad + \int_{S-x_0} \int_{S-x_0} \lambda(dz)P(z, dx)P(x, E). \end{aligned}$$

Iterating this relation, we get for all $n > 1$,

$$\begin{aligned} \lambda(E) &\geq \lambda(x_0)\{P_{x_0}(X_1 \in E) + P_{x_0}(X_1 \neq x_0, X_2 \in E) + \dots \\ &\quad + P_{x_0}(X_1 \neq x_0, \dots, X_{n-1} \neq x_0, X_n \in E)\}. \end{aligned}$$

Letting $n \rightarrow \infty$ and observing that $X_N = x_0$, yields

$$\lambda(E) \geq \lambda(x_0) \sum_{i=1}^{\infty} E_{x_0} \{ \chi_E(X_i); N > i \} = \lambda(x_0) E_{x_0} \sum_{i=1}^N \chi_E(X_i) = \lambda(x_0) \nu(E). \tag{6.4}$$

Thus, since $\nu(x_0) = 1$ by definition of $\nu(\cdot)$,

$$\lambda(x_0) = \int_S \lambda(dx)P^n(x, x_0) \geq \lambda(x_0) \int_S \nu(dx)P^n(x, x_0) = \lambda(x_0)\nu(x_0) = \lambda(x_0)$$

and hence

$$\int_S \lambda(dx)P^n(x, x_0) = \lambda(x_0) \int_S \nu(dx)P^n(x, x_0). \tag{6.5}$$

Let $D_n = \{x: P^n(x, x_0) > 0\}$. To complete the proof it is sufficient (by (6.2)) to show that

$$\lambda(E) = \lambda(x_0)\nu(E) \quad \text{for all } E \subset D_n, \quad n > 1. \tag{6.6}$$

If this is false, then by (6.4) there must be an $E_0 \subset D_n$ for some $n < \infty$ such that $\lambda(E_0) > \lambda(x_0)\nu(E_0)$. But then

$$\int_{E_0} P^n(x, x_0)\lambda(dx) > \lambda(x_0) \int_{E_0} P^n(x, x_0)\nu(dx),$$

while

$$\int_{S-E_0} P^n(x, x_0)\lambda(dx) \geq \lambda(x_0) \int_{S-E_0} P^n(x, x_0)\nu(dx).$$

This contradicts (6.5). \square

ADDENDUM. We have recently learned of concurrent work along similar lines by E. Nummelin, Helsinki University of Technology.

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DEPARTMENT OF APPLIED MATHEMATICS, INDIAN INSTITUTE OF SCIENCE, BANGALORE 560012, INDIA

DEPARTMENT OF MATHEMATICS, UNIVERSITY OF WISCONSIN, MADISON, WISCONSIN 53706