

# A NOTE ON THE GEOMETRIC ERGODICITY OF A MARKOV CHAIN

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## Abstract

It is known that if an irreducible and aperiodic Markov chain satisfies a ‘drift’ condition in terms of a non-negative measurable function  $g(x)$ , it is geometrically ergodic. See, e.g. Nummelin (1984), p. 90. We extend the analysis to show that the distance between the  $n$ th-step transition probability and the invariant probability measure is bounded above by  $\rho^n(a + bg(x))$  for some constants  $a, b > 0$  and  $\rho < 1$ . The result is then applied to obtain convergence rates to the invariant probability measures for an autoregressive process and a random walk on a half line.

AUTOREGRESSIVE PROCESS; DRIFT CONDITION; RANDOM WALK

## 1. Introduction

Let  $(R, \mathcal{E})$  be a measurable space,  $\mathcal{E}$  being assumed countably generated. Let  $\phi$  be a non-trivial  $\sigma$ -finite measure on  $(E, \mathcal{E})$ . An  $E$ -valued Markov chain  $(X_n)_{n \in \mathbb{N}}$  is said to be  $\phi$ -irreducible if

$$(1.1) \quad \sum_{0 \leq n < \infty} P^n(x, A) > 0 \text{ for all } A \in \mathcal{E} \text{ with } \phi(A) > 0,$$

where  $P(\cdot, \cdot)$  denotes the transition probability. In Sections 1 and 2,  $(X_n)$  is assumed to be a  $\phi$ -irreducible Markov chain.  $(X_n)$  is said to be geometrically ergodic if it admits an invariant probability measure  $\pi$ , a  $\pi$ -integrable function  $M$  and a constant  $\rho < 1$  such that

$$(1.2) \quad \|P^n(x, \cdot) - \pi\| \leq M(x)\rho^n, \quad \forall x \in E, \quad n \geq 0,$$

where  $\|\cdot\|$  denotes the total variation norm.

It is known that if  $(X_n)$  is aperiodic and satisfies a ‘drift’ condition in terms of a well-behaved non-negative measurable function  $g(\cdot)$ , then it is geometrically ergodic. We extend the analysis to show that  $M(x)$  in (1.2) can be taken as  $a + bg(x)$  for some positive constants  $a$  and  $b$ . We follow the notation adopted in Nummelin (1984) and refer the reader to it for any unexplained notation.

## 2. Main result

We now state the main theorem.

*Theorem 1.* Suppose that  $(X_n)$  is aperiodic and that for some small set  $C$ , a non-negative measurable function  $g$ , a constant  $r > 1$  such that

$$(2.1a) \quad \sup_{x \in C^c} \mathbb{E}(rg(X_{n+1}) - g(X_n) \mid X_n = x) \equiv \gamma < 0;$$

$$(2.1b) \quad \sup_{x \in C} \mathbb{E}(g(X_{n+1}); X_{n+1} \in C^c \mid X_n = x) \equiv B < \infty;$$

$$(2.1c) \quad g(x) \text{ is bounded away from } 0 \text{ and } +\infty \text{ on } C.$$

Then  $(X_n)$  is geometrically ergodic and  $M(x)$  in (1.2) can be taken as  $a + bg(x)$  for some constants  $a$  and  $b$ .

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Before we prove the theorem, first some remarks. Suppose that condition (2.1) hold. Then Proposition 5.21 in Nummelin (1984) shows that  $(X_n)$  is geometrically recurrent. Thus if  $(X_n)$  is aperiodic, then it is geometrically ergodic. It also follows from Theorem 4 in Tweedie (1983) that  $g(\cdot)$  is  $\pi$ -integrable. Here, we extend the analysis to show that actually  $M(x)$  in (1.2) can be taken as  $a + bg(x)$  for some positive constants  $a$  and  $b$ .

Condition (2.1) is referred to as a kind of ‘drift’ condition in the literature.

*Proof of Theorem 1.* It follows from the smallness of  $C$  that there exist an integer  $m_0$ , a constant  $1 \cong \beta > 0$ , a probability measure  $\nu$  such that

$$(2.2) \quad P^{m_0}(x, A) \cong \beta 1_C(x) \nu(A), \quad \forall x \in E, \quad A \in \mathcal{E}.$$

Let  $\beta 1_C(x)$  be denoted by  $s(x)$ . The pair  $(s, \nu)$  is called an atom and is denoted by  $\alpha$ . Since  $\|P^n(x, \cdot) - \pi\|$  is non-increasing in  $n$ , with no loss of generality, we can assume that  $m_0 = 1$  lest we work with  $(X_{nm_0})$ .

Arguing as in the proof of Theorem 6.14 in Nummelin (1984), we have

$$(2.3) \quad \sum_n r^n \|P^n(x, \cdot) - \pi\| \leq G_\alpha^{(r)} 1(x) + r G_\alpha^{(r)} s(x) \sum_m r^m |u_m - M_b^{-1}| \nu G_\alpha^{(r)} 1 + r(r-1)^{-1} \pi(s) \nu G_\alpha^{(r)} 1 (G_\alpha^{(r)} s(x) + 1),$$

all summations being from 0 to  $\infty$ .

It follows from condition (2.1) and the proof of Proposition 5.21 in Nummelin (1984) that, after some arrangement,

$$(2.4) \quad \mathbb{E}_x(r^{S_C}) \leq a_1 + b_1 g(x)$$

where  $a_1$  and  $b_1$  can be chosen as  $r((r-1)B/\gamma + 1) + 1 + r$  and  $(r-1)/\gamma$  respectively.

From (2.2), we see that there is a probability  $\beta > 0$  that  $X_n \in \alpha$  given that  $X_n \in C$ . Thus, in view of the smallness of  $C$ , it follows from Lemma 5.6 in Nummelin that there exists a constant  $a_2$  such that

$$(2.5) \quad \mathbb{E}_x(r^{S_\alpha}) \leq a_2 \mathbb{E}_x(r^{S_C}) = a_1 a_2 + b_1 a_2 g(x).$$

Now,

$$(2.6) \quad G_\alpha^{(r)} s(x) = \mathbb{E}_x(r^{T_\alpha}) \leq \mathbb{E}_x(r^{S_\alpha}).$$

Applying Lemma 6.2 in Nummelin (1984) with  $\lambda$  there chosen as  $\epsilon_x$ , the probability measure with all its mass at  $x$ , we have

$$(2.7) \quad G_\alpha^{(r)} 1(x) < (r-1)^{-1} G_\alpha^{(r)} s(x).$$

Now arguing again as in the proof of Theorem 6.14 in Nummelin (1984), by decreasing  $r > 1$  if necessary, we have both  $\sum_m r^m |u_m - M_b^{-1}|$  and  $\nu G_\alpha^{(r)} 1$  being finite. Combining (2.3), (2.5), (2.6) and (2.7), it is readily seen that there exist positive constants  $a$  and  $b$  such that  $\sum_n r^n \|P^n(x, \cdot) - \pi\| \leq a + bg(x)$ . So, by taking  $\rho = r^{-1}$ , the proof of the theorem is completed.

### 3. Examples

Theorem 1 provides an upper bound on the convergence rate to the invariant probability measure in the form of  $\rho^n(a + bg(x))$  with  $\rho < 1$ . We now consider two examples in which  $g(x)$  may be chosen as linear in  $|x|$  and exponential in  $x$  respectively. It is also noted that the convergence rate in  $x$  thus obtained is exact for some special cases.

*Example 1.* Let  $(X_n)$  be the stable first-order autoregressive process, i.e.,

$$(3.1) \quad X_n = \phi X_{n-1} + a_n, \quad n = 1, 2, 3, \dots$$

where  $|\phi| < 1$ ,  $(a_n)$  i.i.d. with finite first absolute moment and  $a_n$  independent of  $X_{n-1}, X_{n-2}, \dots, X_0$ . It is assumed that  $a_1$  has an absolutely continuous component which admits a density positive over some open interval about 0. Then  $(X_n)$  is irreducible and

aperiodic. Let  $g(x) = |x| + 1$ . Then condition (2.1) holds with  $C$  chosen as  $[-c, c]$  for some  $c > 0$ . It follows from Theorem 1 that an upper bound on the rate of convergence to the invariant probability measure is linear in  $|x|$ . In the case of  $a_n$  being Gaussian, it can be directly verified that the rate of convergence is indeed linear in  $|x|$ .

*Example 2.* Let  $(X_n)$  be a random walk on  $\mathbb{R}^+$ , i.e.,

$$(3.2) \quad X_{n+1} = (X_n + a_n)^+, \quad n = 1, 2, 3, \dots$$

where  $(a_n)$  is i.i.d.;  $a_n$  independent of  $X_{n-1}, X_{n-2}, \dots, X_0$ ;  $E(a_1) < 0$  and, for some  $M < \infty$  and  $\beta > 0$ ,  $\Pr(a_1 > y) \leq M \exp(-\beta y)$ ,  $\forall y > 0$ . Then  $(X_n)$  is aperiodic and irreducible. It is shown in Nummelin and Tuominen (1982) that condition (2.1) holds with  $g(x) = \exp(tx) + 1$  for some positive  $t$  and  $C$  chosen as  $[0, c]$  for some  $c > 0$ . It follows from Theorem 1 that an upper bound on the convergence rate to the invariant probability measure is exponential in  $x$ . In the special case when

$$(3.3) \quad a_1 = \begin{cases} -1 & \text{with probability } p \\ 0 & \text{with probability } q \end{cases}$$

with  $p + q = 1$  and  $0 < p < 1$ , it is readily seen that the rate of convergence is indeed exponential in  $x$ .

## References

- NUMMELIN, E. (1984) *General Irreducible Markov Chains and Non-negative Operators*. Cambridge University Press.
- NUMMELIN, E. AND TUOMINEN, P. (1982) Geometric ergodicity of Harris recurrent Markov chains with applications to renewal theory. *Stoch. Proc. Appl.* **12**, 187–202.
- TWEEDIE, R. L. (1983) The existence of moments for stationary Markov chains. *J. Appl. Prob.* **20**, 191–196.