

Square summability of variations and convergence of the transfer operator

ANDERS JOHANSSON[†] and ANDERS ÖBERG[‡]

[†] *Sub-department of Mathematics and Statistics, University of Gävle,
SE-801 76 Gävle, Sweden*

(e-mail: ajj@hig.se)

[‡] *Department of Mathematics, Uppsala University, PO Box 480,
SE-751 06 Uppsala, Sweden*

(e-mail: anders@math.uu.se)

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Abstract. In this paper we study the one-sided shift operator on a state space defined by a finite alphabet. Using a scheme developed by Walters [P. Walters. *Trans. Amer. Math. Soc.* **353**(1) (2001), 327–347], we prove that the sequence of iterates of the transfer operator converges under square summability of variations of the g -function, a condition which gave uniqueness of a g -measure in our earlier work [A. Johansson and A. Öberg. *Math. Res. Lett.* **10**(5–6) (2003), 587–601]. We also prove uniqueness of the so-called G -measures, introduced by Brown and Dooley [G. Brown and A. H. Dooley. *Ergod. Th. & Dynam. Sys.* **11** (1991), 279–307], under square summability of variations.

1. Introduction

We consider the left one-sided shift map T on the state space $X^+ = S^{\mathbb{Z}^+}$, where S is a finite set. Thus T acts on elements x of X^+ , $x = (x_0, x_1, x_2, \dots)$, in the following way (each x_i belongs to S):

$$T(x_0, x_1, x_2, \dots) = (x_1, x_2, \dots).$$

A g -measure associated with a continuous function $g : S^{\mathbb{Z}^+} \rightarrow [0, 1]$ is a T -invariant measure μ in the space of Borel probability measures $\mathcal{P}X^+$, such that $g = d\mu/(d\mu \circ T)$, with $\sum_{y \in T^{-1}x} g(y) = 1$, for all $x \in X^+$. Such a function is referred to as a g -function and can be viewed as a probability transition function for the local inverses of T . Keane [6] introduced the concept of g -measures to ergodic theory.

Since the function g is continuous, existence of at least one g -measure follows if $g > 0$ and S is finite. If S is countably infinite, then existence is no longer automatic, as $S^{\mathbb{Z}^+}$ is not compact; however, a weak sufficient condition was given in [8].

Recent results concerning necessary conditions for the uniqueness of a g -measure have been given by Berger *et al* [1], and also by Hulse [5].

Necessary and sufficient conditions sometimes, but not always, focus on variations of the g -function over small parts of its domain. For $f : X^+ \rightarrow \mathbb{R}$, the variation is defined as

$$\text{var}_n f = \sup_{x \sim_n y} |f(x) - f(y)|,$$

where $x \sim_n y$ if $x, y \in X^+$ coincide in their first n coordinates.

Walters [12] proved uniqueness and rates of convergence under summable variations of g -functions, or rather their logarithms $\log g$, which amounts to the same thing if $\inf g > 0$.

When S is finite, a sufficient condition for uniqueness of a g -measure, *square summability of variations* of the g -function, was found by the present authors in [7]. This result was extended to *countable state shifts* in Johansson *et al* [8], where square summability of variations of a g -function means that

$$\sum_{n=1}^{\infty} (\text{var}_n f)^2 < \infty. \quad (1.1)$$

In this paper we prove convergence of the iterates of the transfer (transition) operator, which is defined by

$$\mathcal{L}f(x) = \sum_{y \in T^{-1}x} g(y)f(y) \quad (1.2)$$

for continuous functions f on X^+ .

More specifically, we prove that with finitely many symbols in S and under square summability of variations of a strictly positive g -function, we have $\mathcal{L}^n f(x) \rightarrow \int f d\mu$ uniformly in x , where μ is the unique g -measure.

To accomplish this, we use a theory developed by Walters in [13] to study the sequence of iterates of an operator P , defined as

$$P^n f(x) = (\mathcal{L}^n f)(T^n x) \quad (1.3)$$

for $n \geq 1$.

Walters introduces an adjoint sequence of operators $P^{(n)*} : \mathcal{P}X^+ \rightarrow \mathcal{P}X^+$, $n \geq 1$. Define $K_g^n = \{v \in \mathcal{P}X^+ \mid P^{(n)*} v = v\}$. Then [13, Theorem 2.1(v)] implies that $K_g^n = P^{(n)*} \mathcal{P}X^+$ and that $K_g^1 \supset K_g^2 \supset \dots$. Walters defined $K_g = \bigcap_{n=1}^{\infty} K_g^n$, which corresponds to the set of G -measures in the terminology of Brown and Dooley [2]; see also Fan [4] and Dooley and Stenflo [3] for some recent contributions to establishing uniqueness of G -measures. One of the motivations for studying G -measures has been to obtain a more general framework for studying Riesz products; see [4], for example.

Our strategy is to prove that there exists a *unique* measure in K_g (it is trivial that any g -measure must be an element of K_g). We then use [13, Theorem 2.9] to conclude that $\mathcal{L}^n f(x) \rightarrow \int f d\mu$, where μ is the unique element of K_g . This follows from a compactness argument; therefore, to prove the convergence of $\mathcal{L}^n f$, we have to assume that the symbol space S is finite.

We need to define a probability measure, a Markov chain on X^+ which we call a g -chain, in order to reverse the dynamics in the absence of stationarity; it is not known whether or when the measures in K_g are invariant under T .

To conclude the proof, we use the same method as in [7] to show that any two extremal measures in K_g must be absolutely continuous, and hence that there can exist only one such measure. This method was developed further in [8] to cover the case of countable state shifts for g -measures (thus improving on the results of [7]), using more sophisticated methods to prove absolute continuity of measures. We note that these methods had been employed much earlier by Shiryaev and co-authors; see, for example, the survey by Shiryaev [11].

We end the paper by stating two open questions.

2. Results and proofs

THEOREM 2.1. Assume that we have a finite symbol set S . Let $g > 0$ satisfy

$$\sum_{n=1}^{\infty} (\text{var}_n \log g)^2 \asymp \sum_{n=1}^{\infty} (\text{var}_n g)^2 < \infty$$

and let μ be the corresponding unique g -measure. Then $K_g = \{\mu\}$ or, in other words, there exists a unique G -measure which is the same as the unique g -measure.

The following theorem is, in view of [13], a corollary of Theorem 2.1. It is based on [13, Theorem 2.9], and we reproduce part of the proof here to explain why we need to assume that there are only finitely many symbols: this is because one has to have a compact state space, or at least some kind of tightness property, so that a certain convergent subsequence will exist for a sequence of measures.

THEOREM 2.2. Assume that we have a finite symbol set S . Let $g > 0$ be a g -function with square summable variation and let μ be the corresponding unique g -measure. Then $\sup_x |\mathcal{L}^n f(x) - \int f d\mu| \rightarrow 0$.

Proof. (Borrowed from [13, Theorem 2.9]) Under these same assumptions, Theorem 2.1 gives us a unique measure in K_g . From this we can show that $\sup_x |P^n f(x) - \int f d\mu| \rightarrow 0$, because otherwise there would exist a continuous function f , an $\epsilon > 0$ and sequences $\{n_k\}$ and $\{x_k\}$ such that

$$\left| \int f d(P^{(n_k)^*} \delta_{x_k}) - \int f d\mu \right| \geq \epsilon \quad \text{for all } k \geq 1.$$

Owing to compactness, we can then pick a subsequence $\{k_j\}$ of $\{k\}$ such that the sequence of measures $\{P^{(n_{k_j})^*} \delta_{x_{k_j}}\}$ converges to $\nu \in \mathcal{P}X^+$; however, since $K_g^{(n)} = P^{(n)*} \mathcal{P}X^+$, we would also have $\nu \in K_g$, which is a contradiction. Since $P^{(n)} f(x) = (\mathcal{L}^n f)(T^n x)$, we have $\sup_x |\mathcal{L}^n f(x) - \int f d\mu| \rightarrow 0$. □

2.1. *Proof of Theorem 2.1.* We begin with some preliminary terminology; see [8] for a more thorough exposition. For a pair of probability measures $\nu, \tilde{\nu} \in \mathcal{P}X$ and some filtration $\{\mathcal{F}_n\}$, let $Z_n(x) = Z_n(x; \tilde{\nu}, \nu, \mathcal{F}_n)$ be the likelihood-ratio martingale

$$Z_n(x; \tilde{\nu}, \nu, \{\mathcal{F}_n\}) = \frac{d\tilde{\nu}|_{\mathcal{F}_n}}{d\nu|_{\mathcal{F}_n}}(x)$$

on \mathcal{F}_n , where we assume that $\tilde{\nu}$ and ν are locally absolutely continuous on \mathcal{F}_n .

Now let $X = S^{\mathbb{Z}}$ and $X^+ = S^{\mathbb{Z}_+}$, and extend the one-sided shift T on X^+ to a two-sided shift on X . For $a, b \in \mathbb{Z}$, let $\Pi_{a,b}$ be the mapping that takes $x \in X$ to $(x_a, x_{a+1}, \dots, x_b) \in S^{b-a+1}$, and define the natural projection $\Pi_+ : X \rightarrow X^+$ taking bi-infinite sequences x in X to one-sided ones in X^+ by $\Pi_+((x_i)_{i=-\infty}^{\infty}) = (x_0, x_1, \dots)$. Also let \mathcal{F}_a^b be the algebra generated by $\Pi_{a,b}$ and define $\mathcal{F}_n^- := \mathcal{F}_{-n}^{-1}$, the ‘forward’ algebra, and $\mathcal{F}_n^+ := \mathcal{F}_0^{n-1}$, the ‘backward’ algebra. A *cylinder set* is a set of the form $[x]_a^b = \Pi_{a,b}^{-1}\Pi_{a,b}(x)$.

We now define a certain Markov chain that will enable us to go forward in time, since we may not assume that the measures in K_g defined by Walters are invariant under T .

Definition 1. A *g-chain* on $X = S^{\mathbb{Z}}$ is a probability measure $\nu \in \mathcal{P}X$ such that, for all $n \in \mathbb{Z}$,

$$\nu(x_n | x_{n+1}, x_{n+2}, \dots) = g(x_n, x_{n+1}, x_{n+2}, \dots). \tag{2.1}$$

A *forward g-chain* is a probability measure ν on X satisfying (2.1) for $n \leq -1$.

The distribution under a *g-chain* ν of the process $x^{(t)} \in X^+, t \in \mathbb{Z}$ defined by

$$x^{(t)} := \Pi_+(T^{-t}x) = (x_{-t}, x_{-t+1}, \dots)$$

is that of a Markov chain such that the transition probabilities are given by g (and the transition operator of the chain is \mathcal{L}). That is, for all $t \in \mathbb{Z}$,

$$\nu(x^{(t)} | x^{(t-1)}) = g(x^{(t)}).$$

The same holds for $t \geq 1$ in the case of a forward *g-chain* ν .

LEMMA 2.3. *A probability measure $\nu \in \mathcal{P}X$ is a g-chain only if $\nu \circ \Pi_+^{-1} \in \mathcal{P}X^+$ is an element of the set K_g of eigen-measures as defined by Walters in [13]. Conversely, any ν in K_g corresponds, by extension, to a unique g-chain.*

Proof. Suppose that we have $P^*\nu = \nu$. This is equivalent to $\mathcal{L}^*(\nu \circ T^{-1}) = \nu$. By interpreting the conditional probability $\nu(x_0 | x_1, x_2, \dots)$ as the transition probability $\nu(x^{(0)} | x^{(-1)})$ from the Markov chain defined above, we reach the conclusion that

$$\nu(x_0 | x_1, x_2, \dots) = g(x_0, x_1, \dots).$$

Thus, if $P^{(n)*}\nu = \nu$, we have, for $0 \leq k \leq n - 1$,

$$\nu(x_k | x_{k+1}, x_{k+2}, \dots) = g(x_k, x_{k+1}, x_{k+2}, \dots). \tag{2.2}$$

Since $K_g^n = P^{(n)*}\mathcal{P}X^+$ and $K_g = \bigcap_{n=1}^{\infty} K_g^n$, we know (2.2) holds for all $k \geq 0$. Using measures in the non-empty set K_g as initial distributions, we may define the full *g-chain* by making a unique canonical extension (see Neveu [9, p. 83, the corollary]) so that (2.2) is true for $k \leq -1$. Hence (2.2) holds for all $k \in \mathbb{Z}$. □

From now on, we shall not distinguish between a *g-chain* ν and its one-sided restriction $\nu \circ \Pi_+^{-1}$.

LEMMA 2.4. *The set K_g is a non-empty convex subset with mutually singular extreme points.*

Proof. See Walters [13, Theorem 2.11 and its proof]. □

We work under the assumption that $g > 0$ and hence that any two g -chains will be locally absolutely continuous on any of the algebras \mathcal{F}_a^b , where a and b are finite. Given two g -chains $\nu, \tilde{\nu} \in K_g$, let

$$\xi_n(x) := Z_n(x; \tilde{\nu}, \nu, \{\mathcal{F}_n^+\}). \tag{2.3}$$

Note that ξ_n is the likelihood-ratio martingale and, since it is a positive $L_1(\nu)$ -bounded ν -martingale, we know that it converges ν -almost surely. We want to show that it is uniformly integrable (UI) with respect to ν , which gives L_1 -density between $\tilde{\nu}$ and ν on $\mathcal{F}^+ = \lim \mathcal{F}_n^+$ and thus, according to Lemma 2.4, a contradiction if ν and $\tilde{\nu}$ are chosen to be distinct and extremal in K_g .

It was shown in [7, p. 595] that verifying the UI property for a likelihood-ratio martingale such as ξ_n amounts to showing that

$$\lim_{K \rightarrow \infty} \sup_n \tilde{\nu}(\log \xi_n > K) = 0. \tag{2.4}$$

To see (2.4), we note that, for a fixed value of m , a translation by m steps to the left of both the point and the measure gives

$$\xi_m(x) = Z_m(T^m x; \tilde{\nu} \circ T^{-m}, \nu \circ T^{-m}, \{\mathcal{F}_m^-\}). \tag{2.5}$$

Hence the law of ξ_m under $\tilde{\nu}$, namely $\tilde{\nu} \circ \xi_m^{-1}$, equals the distribution $(\tilde{\nu} \circ T^{-m}) \circ \zeta_{m,m}^{-1}$ where $\{\zeta_{m,n}(x) : n \in \mathbb{Z}_+\}$ is the forward likelihood-ratio martingale

$$\zeta_{m,n}(x) = Z_n(x; \tilde{\nu} \circ T^{-m}, \nu \circ T^{-m}, \{\mathcal{F}_n^-\}).$$

This means that we start with two extremal measures $\nu, \tilde{\nu} \in K_g$ translated m times to the left, i.e. $\nu \circ T^{-m}$ and $\tilde{\nu} \circ T^{-m}$; these can be extended to well-defined g -chains and we may then go forward along \mathcal{F}_n^- .

Thus, it is enough to prove the following lemma which states that the forward likelihood ratios are uniformly tight in a strong sense. The estimates have to be uniform in all g -chains, since our substitutions $\mu = \nu \circ T^{-m}$ and $\tilde{\mu} = \tilde{\nu} \circ T^{-m}$ are valid only for a fixed m .

LEMMA 2.5. Assume that $\text{var}_n g$ is square summable and that g is bounded away from zero. For all $\epsilon > 0$, there exists a $K = K(\epsilon)$ such that

$$\sup_{\tilde{\mu}, \mu} \sup_n \tilde{\mu}(\log Z_n(x; \tilde{\mu}, \mu, \{\mathcal{F}_n^-\}) > K) < \epsilon, \tag{2.6}$$

where $\tilde{\mu}$ and μ are chosen among all pairs of forward g -chains.

Proof. We adapt part of the proof in [7, pp. 597–598]. One could also use the more advanced theory of Shiryaev [11] and his co-authors, as was done in [8].

Given μ and $\tilde{\mu}$, write (using the notation from [7]) $M_n(x) = Z_n(x; \tilde{\mu}, \mu, \{\mathcal{F}_n^-\})$. We show that $\log M_n$ has a Doob decomposition

$$\log M_n = A_n + \eta_n$$

where A_n is pre-visible with the uniform bound $A_n \leq C_1 \sum (\text{var}_n g)^2$ and, moreover, η_n is a $\tilde{\mu}$ -martingale and uniformly bounded in $L_2(\tilde{\mu})$ with $\tilde{\mu}(\eta_n^2) \leq C_2 \sum (\text{var}_n g)^2$ (see, for

example, [14, pp. 120–121]). It is the uniformity of the estimates that makes it possible to conclude the strong formulation of (2.6). We define

$$P_n(x) = \frac{\mu[x_{-n}, x_{-n+1}, \dots, x_{-1}]}{\mu[x_{-n+1}, \dots, x_{-1}]},$$

$$\tilde{P}_n(x) = \frac{\tilde{\mu}[x_{-n}, x_{-n+1}, \dots, x_{-1}]}{\tilde{\mu}[x_{-n+1}, \dots, x_{-1}]},$$

and note that $|\tilde{P}_n - P_n| \leq \text{var}_n g$.

The proof then proceeds exactly as in [7, pp. 597–598]. \square

3. Open questions

In this section we present two questions which the authors find interesting, as well as challenging, in the light of the present investigation.

3.1. Question 1. Does uniqueness of a g -measure imply

$$\mathcal{L}^n f \rightarrow \int f d\mu ?$$

We assume that we have finitely many symbols for the left-shift map T and a continuous and strictly positive g -function.

In this paper we used the fact that a unique measure in K_g (in Walters' notation) or, equivalently, a unique G -measure (as defined by Brown and Dooley [2], Fan [4] and others) implies that $\mathcal{L}^n f(x) \rightarrow \int f d\mu$ uniformly in $x \in X^+$. It is natural to ask: if K_g only contains the g -measure, then would this also be the unique member of K_g ? (Recall that a g -measure is always a member of K_g .) It would then follow that uniqueness of a g -measure (under the assumptions given in our question) implies convergence of the iterates of \mathcal{L} , without imposing any stronger regularity on g than continuity.

3.2. Question 2. Under square summability of variations of the g -function, what is the rate of convergence of $\mathcal{L}^n f(x)$ in the supremum-norm? Rates of convergence in the case of *summable variations* are available; see Pollicott [10].

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