

# Law of large numbers and central limit theorem for ergodic quantum processes

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## Abstract

A discrete quantum process is represented by a sequence of quantum operations, which are completely positive maps that are not necessarily trace preserving. We consider quantum processes that are obtained by repeated iterations of a quantum operation with noise. Such ergodic quantum processes generalize independent quantum processes. An ergodic theorem describing convergence to equilibrium for a general class of such processes was recently obtained by Movassagh and Schenker. Under irreducibility and mixing conditions we obtain a central limit type theorem describing fluctuations around the ergodic limit.

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## 1 Introduction and main results

A *quantum channel* (QC) is a linear, completely positive, and trace preserving map on the trace class operators, where the state of the system is represented by a non-negative operator of trace one — a *density matrix*. Such maps can describe the evolution of an open quantum system over a discrete unit of time, including averaged effects of measurements and environmental noise. More generally, one introduces *quantum operations* (QOs) — completely positive and trace non-increasing maps — to describe processes with loss or which happen only with a certain probability. A *quantum process* is a sequence of QOs describing the evolution of the system over a consecutive sequence of time intervals. Quantum processes represent the most general description of the average evolution of an open quantum system neglecting memory effects in the environment.

In a pair of recent papers [22, 21], Movassagh and the second author formulated the notion of an *ergodic quantum process* in which the individual QOs are obtained by sampling a QO valued function along a trajectory of an ergodic dynamical system. For processes on a finite dimensional Hilbert space and satisfying a physically natural decoherence condition, they proved convergence of the density matrix to a stationary, ergodic sequence of density matrices as time goes to infinity. This theorem of [21] generalizes a result of Hennion [14] on products of non-negative random matrices and is closely related Oseledec's multiplicative ergodic theorem [23].

The results of [21] require essentially only decoherence and ergodicity. In the present paper, we examine processes that satisfy stronger integrability and mixing conditions. We prove a law of large numbers and a central limit theorem for the expectation values of observables in states evolving under such a processes. Although our main interest is in the application of these results to quantum processes, the results themselves do not require the maps to be trace non-increasing and require only *positivity* (not complete positivity).

This paper is organized as follows:

1. In §2 we state our main results after formulating certain background notions.
2. In §3, review some definitions and arguments from [22] that are fundamental to the proofs of our main results.
3. In §4, we prove Theorem 1 - [Law of Large Numbers](#).
4. In §5, we prove Theorem 2 - [Central Limit Theorem](#).
5. In §6, we prove Theorem 3, which gives sufficient conditions for the main hypothesis of Theorem 2 - [Central Limit Theorem](#) to hold.

## 2 Formal statement of the main results

### 2.1 Positive Linear Maps

Let  $\mathbb{M}_D = \mathbb{C}^{D \times D}$  denote the space of  $D \times D$  matrices. We consider the space  $\mathbb{M}_D$  with its standard topology as a finite-dimensional vector space. For definiteness, we take this to be the norm topology generated by the *trace norm*,  $\|A\| := \text{Tr} \sqrt{(A^* A)}$  for any  $A \in \mathbb{M}_D$ , but of course the topology is independent of the norm (since  $\mathbb{M}_D$  is finite dimensional). For any matrix  $A \in \mathbb{M}_D$  we denote by  $A^*$  the adjoint matrix (conjugate transpose).

The space of linear operators on  $\mathbb{M}^D$  will be denoted by  $\mathcal{L}(\mathbb{M}_D)$ . We equip the space  $\mathcal{L}(\mathbb{M}_D)$  with the operator norm induced by the trace norm on  $\mathbb{M}_D$ . That is, for  $\phi \in \mathcal{L}(\mathbb{M}_D)$ :

$$\|\phi\| = \sup\{\|\phi(A)\| : A \in \mathbb{M}_D, \|A\| = 1\}. \quad (2.1)$$

For any  $\phi \in \mathcal{L}(\mathbb{M}_D)$  the adjoint of  $\phi$  is the unique map  $\phi^* \in \mathcal{L}(\mathbb{M}_D)$  determined by the identity:

$$\langle A, \phi(B) \rangle = \langle \phi^*(A), B \rangle \text{ for all } A, B \in \mathbb{M}_D, \quad (2.2)$$

where  $\langle A, B \rangle$  denotes the Hilbert-Schmidt inner product,

$$\langle A, B \rangle = \text{tr} A^* B . \quad (2.3)$$

We recall that a map  $\phi \in \mathcal{L}(\mathbb{M}_D)$  is *positive*, if it maps the set of positive semi-definite matrices to itself. It is convenient to introduce notation for certain subsets of positive semi-definite matrices as follows:

1.  $\text{POS}_D$  is the set of all positive semi-definite  $D \times D$  matrices,
2.  $\text{POS}_D^0$  is the set of all positive definite  $D \times D$  matrices,
3.  $\mathbb{S}_D$  is the set of positive semi-definite  $D \times D$  matrices with trace one, and
4.  $\mathbb{S}_D^0$  is the set of positive definite  $D \times D$  matrices with trace one.

The subset  $\mathbb{S}_D$ , being bounded and closed, is compact by the Heine-Borel theorem. Note that  $\phi$  is positive if and only if  $\phi(\mathbb{S}_D) \subset \text{POS}_D$ . We call  $\phi$  *strictly positive* if  $\phi(\mathbb{S}_D) \subset \text{POS}_D^0$ .

Positive maps satisfy a generalization of the Perron-Frobenius Theorem (see [18, 10]): every such map  $\phi$  has an eigenmatrix  $R \in \mathbb{S}_D$  with eigenvalue equal to the spectral-radius  $r(\phi)$ . The map  $\phi$  is called *irreducible* if  $(\mathbf{1} + \phi)^n$  is strictly positive for some  $n$ .<sup>1</sup> By [10, Theorems 2.3 & 2.4] we have the following

**Proposition 2.1.** *If  $\phi$  is an irreducible positive map, then there is a unique  $R \in \mathbb{S}_D$  such that  $\phi(R) = \Lambda R$  for some  $\Lambda \in \mathbb{C}$ . Furthermore, the eigen-matrix  $R$  is non-singular ( $R \in \mathbb{S}_D^0$ ) and the eigenvalue  $\Lambda = r(\phi) > 0$  is the spectral radius of  $\phi$ .*

We call the unique eigenmatrix  $R \in \mathbb{S}_D$  of an irreducible map  $\phi$  the *right Perron-Frobenius eigenmatrix of  $\phi$* . The map  $\phi$  also has a *left Perron-Frobenius eigenmatrix*, which is the Perron-Frobenius eigenmatrix of  $\phi^*$ . (Note that  $\phi$  is irreducible if and only if  $\phi^*$  is.)

The Perron-Frobenius eigenmatrix  $R$  of an irreducible map  $\phi$  may be interpreted as a fixed point of the *projective action of  $\phi$* :

$$\phi \cdot X = \frac{\phi(X)}{\text{tr} \phi(X)} . \quad (2.4)$$

For a general map, the projective action is defined for  $X \in \mathbb{S}_D \setminus \ker \phi$ . However, if  $\ker \phi \cap \mathbb{S}_D = \emptyset$  then the projective action is defined on all of  $\mathbb{S}_D$ . As this condition will play a key role in our analysis, we make the following

**Definition 1.** A positive linear map  $\phi \in \mathcal{L}(\mathbb{M}_D)$  is *non-destructive* if  $\ker \phi \cap \mathbb{S}_D = \emptyset$ . If  $\phi^*$  is non-destructive, we say that  $\phi$  is *non-transient*.

The terminology *non-transient* stems from the fact that if  $\rho \in \ker \phi^* \cap \mathbb{S}_D$  and  $P$  is the projection onto  $\text{ran } \rho$ , then  $\phi^*(P) = 0$  and  $\text{ran } \phi$  is contained in the hereditary sub-algebra  $P^\perp \mathbb{M}_D P^\perp$  where  $P^\perp = I - P$ . Thus the subspace corresponding to  $\text{ran } P$  is a “transient subspace” for  $\phi$ .

A sufficient condition for  $\phi$  to be non-destructive and non-transient is that  $\phi^n$  be strictly positive for some  $n > 0$ . This condition is, in turn, equivalent to  $\phi$  being *irreducible and aperiodic*, i.e., irreducible and having no eigenvalues on the circle  $\{|z| = r(\phi)\}$  except for the Perron-Frobenius eigenvalue.

## 2.2 Limiting results for eigenmatrices of ergodic quantum processes

As in [22], we are interested in sequences  $\Phi^{(n)}$  such that

$$\Phi^{(n)} = \phi_n \circ \dots \circ \phi_1 \quad \text{with} \quad \phi_n = \phi_{0; \theta^n \omega} , \quad (2.5)$$

where  $\omega \mapsto \phi_{0; \omega}$  is a positive map valued random variable defined on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and  $\theta : \Omega \rightarrow \Omega$  is an ergodic map. We recall that a measurable map  $\theta : \Omega \rightarrow \Omega$  is

<sup>1</sup>Equivalently, no *hereditary sub-algebra*,  $P \mathbb{M}_D P$  with  $P$  an orthogonal projection, is invariant under  $\phi$ . See [10].

1. *measure preserving* if  $\mathbb{P}(\theta^{-1}(A)) = \mathbb{P}(A)$  for all  $A \in \mathcal{F}$ , and
2. *ergodic* if it is measure preserving and  $\mathbb{P}(A) = 0$  or 1 whenever  $\theta^{-1}(A) = A$ .

We further recall that either of the following two conditions is sufficient for a measure preserving map  $\theta$  to be ergodic:

1. *essentially  $\theta$ -invariant sets have measure 0 or 1*, i.e.,  $\mathbb{P}(A) = 0$  or 1 whenever  $A \in \mathcal{F}$  with  $\mathbb{P}(A \Delta \theta^{-1}(A)) = 0$ .
2. *essentially  $\theta$ -invariant functions are almost surely constant*, i.e., if  $f \circ \theta = f$  almost surely, then there is  $c \in \mathbb{R}$  such that  $f = c$  almost surely.

See [24] for proofs of these facts and further discussion of ergodic maps.

Now fix a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and an ergodic map  $\theta : \Omega \rightarrow \Omega$ . For a random variable  $X : \Omega \rightarrow \mathcal{S}$ , with  $\mathcal{S}$  some measurable space, we denote the value of  $X$  at  $\omega \in \Omega$  by  $X_\omega$ , and will often omit  $\omega$  from the notation for simplicity. This subscript notation is convenient as we consider map valued random variables which take a matrix as an argument. Let  $\varphi_0 : \Omega \rightarrow \mathcal{L}(\mathbb{M}_D)$  be a positive map valued random variable, where we take the Borel  $\sigma$ -algebra on  $\mathcal{L}(\mathbb{M}_D)$ . For each  $n \in \mathbb{N}$ , define  $\varphi_{n;\omega} = \varphi_{0;\theta^n(\omega)}$ . Let

$$\Phi_\omega^{(n)} = \varphi_{n;\omega} \circ \varphi_{n-1;\omega} \circ \cdots \circ \varphi_{1;\omega}. \quad (2.6)$$

For  $k \geq 0$ , we have

$$\Phi_{\theta^k(\omega)}^{(n)} = \varphi_{n;\theta^k(\omega)} \circ \cdots \circ \varphi_{1;\theta^k(\omega)} = \varphi_{n+k;\omega} \circ \cdots \circ \varphi_{1+k;\omega}; \quad (2.7)$$

as above we may omit  $\omega$  from the notation and simply write this as  $\Phi_{\theta^k}^{(n)} = \varphi_{n+k} \circ \cdots \circ \varphi_{1+k}$ .

In the present work, we study sequences  $\Phi_\omega^{(n)}$  with the property that  $\Phi_\omega^{(n)}$  is eventually strictly positive. We denote by  $\tau_\omega$  the time at which  $\Phi_\omega^{(n)}$  becomes strictly positive and stays strictly positive thereafter:

$$\tau_\omega = \inf\{n \geq 1 : \Phi_\omega^{(n+k)} \text{ is strictly positive } \forall k \geq 0\}. \quad (2.8)$$

Our first assumption is that  $\tau < \infty$  almost surely:

**Assumption 1.** *We have  $\mathbb{P}\{\tau < \infty\} = 1$ , i.e., the sequence  $\Phi_\omega^{(n)}$  is almost surely eventually strictly positive.*

Assumption 1 was also the main assumption of [22], where it was shown to be equivalent to the following two conditions provided that  $\theta$  is invertible (see [22, Lemma 2.1]):

1. there exists  $N_0 \in \mathbb{N}$  such that  $\mathbb{P}(\Phi^{(N_0)} \text{ is strictly positive}) > 0$ , and
2.  $\mathbb{P}\{\varphi_0 \text{ is non-destructive and non-transient}\} = 1$ .

One consequence of this equivalence is that, if  $\theta$  is invertible and Assumption 1 holds, then  $\tau$  can be expressed as

$$\tau = \inf\{n \geq 1 : \Phi_\omega^{(n)} \text{ is strictly positive}\}. \quad (2.9)$$

In particular,  $\tau$  is then a *stopping time* with respect to the filtration  $(\mathcal{F}_n)_{n=0}^\infty$  where  $\mathcal{F}_n$  denotes the  $\sigma$ -algebra generated by  $\phi_0, \dots, \phi_n$ .

Since any strictly positive map is irreducible, Assumption 1 guarantees that the left and right Perron-Frobenius eigenmatrices,  $R_n$  and  $L_n$ , exist for for sufficiently large  $n$ :

$$\Phi_\omega^{(n)}(R_n) = \Lambda_n R_n \quad \text{and} \quad \Phi_\omega^{(n)*}(L_n) = \Lambda_n L_n. \quad (2.10)$$

Here  $\Lambda_n = \Lambda_{n;\omega}$  denotes the spectral radius of  $\Phi_\omega^{(n)}$  and  $L_n, R_n$  are  $\mathbb{S}_D^+$  valued random variables, i.e., they are  $D \times D$  positive definite matrix valued random variables with  $\text{tr}R_n = \text{tr}L_n = 1$ . We have the following

**Lemma 2.2** ([22, Theorem 1]). *Let  $(\varphi_n)_{n \geq 1}$  and  $\Phi^{(n)}$  be as in eq. (2.6) and let  $L_n$  be as in eq. (2.10). If Assumption 1 holds, then there is an  $\mathbb{S}_D^\circ$  valued random variable  $Z'_1$  such that*

$$Z_1 \xrightarrow{a.s.} \lim_{n \rightarrow \infty} L_n \quad (2.11)$$

and, with  $Z_k := Z_1 \circ \theta^{k-1}$ , we have for every  $k \in \mathbb{N}$ ,  $\varphi_k^* \cdot Z_{k+1} = Z_k$  a.s..

*Remark 2.3.* This is half of [22, Theorem 1]. The other half involves the convergence of the right eigenvectors and requires invertibility of the ergodic map  $\theta$ . A close reading of the proof (see [22, Lemma 3.12]) shows that invertibility of  $\theta$  is not necessary for the portion stated here.

### 2.3 Law of Large Numbers

Our first main result is concerned with expectations of the form  $\langle Y, \Phi^{(n)}(X) \rangle$  with  $X, Y \in \mathbb{S}_D$ . The main idea here is that for large  $n$ , the Perron-Frobenius eigenvalue  $\Lambda_n$  of  $\Phi^{(n)}$  typically exhibits exponential growth or decay and dominates the expression, so that we expect

$$\langle Y, \Phi_n(X) \rangle \approx \Lambda_n \frac{\langle Y, R_n \rangle \langle L_n, X \rangle}{\langle L_n, R_n \rangle} + \text{lower order terms,} \quad (2.12)$$

where  $L_n$  and  $R_n$  are the left and right Perron-Frobenius eigenmatrices, respectively, normalized so that  $\text{tr} L_n = \text{tr} R_n = 1$ . Under Assumption 1,  $L_n$  and  $R_n$  are positive definite, so  $\langle Y, R_n \rangle \langle L_n, X \rangle \neq 0$  and eq. (2.12) suggests that

$$\ln \langle Y, \Phi^{(n)}(X) \rangle \approx \ln \Lambda_n + O(1).$$

Thus we expect a Law of Large Numbers,  $\frac{1}{n} \ln \langle Y, \Phi^{(n)}(X) \rangle \rightarrow l$ , where  $l = \lim_n \frac{1}{n} \ln \Lambda_n$ .

To obtain this Law of Large Numbers, we require an integrability assumption for  $\ln \|\varphi_0^*\|$  and for  $\ln v(\varphi_0^*)$ , where for  $\phi \in \mathcal{L}(\mathbb{M}_D)$  we define

$$v(\phi) := \inf \{ \|\phi(X)\| : X \in \mathbb{S}_D \}. \quad (2.13)$$

**Assumption 2.** *We have  $\mathbb{E}[\ln \|\varphi_0^*\|] < \infty$  and  $\mathbb{E}[\ln v(\varphi_0^*)] < \infty$ .*

*Remark 2.4.* We note that any non-destructive map  $\phi$  (in particular, any strictly positive map) must have  $v(\phi) > 0$  because  $\mathbb{S}_D$  is a compact set and the map  $A \mapsto \|\phi(A)\|$  is continuous.

With Assumptions 1 and 2 we have the following

**Theorem 1 - Law of Large Numbers.** *Let  $\Phi^{(n)}$  be a random sequence of positive maps as in eq. (2.6). If Assumptions 1 and 2 hold then*

$$\lim_{n \rightarrow \infty} \sup_{X, Y \in \mathbb{S}_D} \left| \frac{1}{n} \ln \langle Y, \Phi^{(n)}(X) \rangle - l \right| = 0 \quad a.s., \quad (2.14)$$

where  $l = \mathbb{E}[\ln \|\varphi_0^*(Z_1)\|]$  with  $Z_1 = \lim_n L_n$ . Furthermore

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln \|\Phi^{(n)}\| = \lim_{n \rightarrow \infty} \frac{1}{n} \ln \Lambda_n = l \quad a.s., \quad (2.15)$$

with  $\Lambda_n$  the Perron-Frobenius eigenvalue of  $\Phi^{(n)}$ .

*Remark 2.5.* We take  $\ln \langle Y, \Phi^{(n)}(X) \rangle = -\infty$  if  $\langle Y, \Phi^{(n)}(X) \rangle = 0$ ; by Assumption 1 this happens for at most finitely many  $n$ . By Assumption 2,  $l = \mathbb{E}[\ln \|\varphi_0^*(Z_1)\|]$  is finite.

Theorem 1 - Law of Large Numbers is closely related in spirit to the Furstenberg-Kesten theorem [11] and Oseledec's Theorem [23] (see also [12]). By the Furstenberg-Kesten Theorem, the following limit exists

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln \|\Phi^{(n)}\| \quad a.s. = \lambda \quad a.s.,$$

where  $\lambda$  is a deterministic quantity called the *top Lyapunov exponent* of the cocycle  $(X, n) \mapsto \Phi^{(n)}(X)$ . By Oseledet's Theorem, there is a (random) proper subspace  $L \subset \mathbb{M}_D$  such that for  $X \in \mathbb{M}_D \setminus L_{j+1}$  we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln \|\Phi^{(n)}(X)\| = \lambda.$$

The identity eq. (2.14) is the key result in Theorem 1 - **Law of Large Numbers**. Indeed, since  $\Lambda_n = \langle L_n, \Phi^{(n)}(\mathbb{I}) \rangle$  it follows directly from eq. (2.14) that  $l = \lim_n \frac{1}{n} \ln \Lambda_n$  almost surely. Furthermore, as the proof of eq. (2.14) will make clear (see Lemma 4.1), we also have  $\lim_n \frac{1}{n} \ln \|\Phi^{(n)*}(Y)\| = l$  a.s. for any  $Y \in \mathbb{S}_D$ . Since  $\text{span } \mathbb{S}_D = \mathbb{M}_D$ , it follows from Oseledet's Theorem that  $l = \lambda$ , the top Lyapunov exponent, and thus that  $l = \lim_n \frac{1}{n} \ln \|\Phi^{(n)}\|$ . Therefore eq. (2.15) is a consequence of eq. (2.14). Thus to prove Theorem 1 - **Law of Large Numbers** it suffices to prove eq. (2.14). This is accomplished in §4 below.

## 2.4 Central Limit Theorem

Our second main result is a central limit theorem for the fluctuations of  $\ln \langle Y, \Phi^{(n)}(X) \rangle$  around its asymptotic value  $nl$ . For this result we require additional integrability for  $\ln \|\varphi_{0;\omega}^*\|$  and  $\ln v(\varphi_{0;\omega}^*)$ :

**Assumption 2<sub>p</sub>.** For  $p > 1$ , the random variables  $\ln \|\varphi_{0;\omega}^*\|$  and  $\ln v(\varphi_{0;\omega}^*)$  are in  $L^p$ .

To obtain a central limit theorem, we require the ergodic map  $\theta$  to be invertible, and extend the definition of  $\varphi_k$  to  $k < 0$  by  $\varphi_{k;\omega} = \varphi_{0;\theta^k \omega}$ , just as for  $k \geq 0$ . Similarly we define  $Z_{k;\omega} = Z_{1;\theta^{k-1}\omega}$  for  $k \leq 0$ . The key quantities that describe the fluctuations are the deviations of  $\ln \|\varphi_k^*(Z_{k+1})\|$  from its mean:

$$\xi_k := \ln \|\varphi_k^*(Z_{k+1})\| - l, \quad (2.16)$$

where  $l$  is as in Theorem 1 - **Law of Large Numbers**. We also introduce the following reverse filtration  $(\mathcal{F}^n)_{n \in \mathbb{Z}}$  on the probability space:

$$\mathcal{F}^n := \text{sigma algebra generated by } (\varphi_k)_{k \geq n}. \quad (2.17)$$

With these preliminaries, we have the following

**Theorem 2 - Central Limit Theorem.** Let  $\Phi^{(n)}$  be a random sequence of positive maps as in eq. (2.6). Suppose that the ergodic map  $\theta$  is invertible, that Assumption 1 holds, and that Assumption 2<sub>p</sub> holds for some  $p \geq 2$ . If

$$\sum_{n=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q < \infty \quad (2.18)$$

with  $1/p + 1/q = 1$ , then for any sequences  $(X_n)_{n \geq 1}$  and  $(Y_n)_{n \geq 1}$  in  $\mathbb{S}_n$ , the random sequence

$$\left( \frac{1}{\sqrt{n}} \left( \ln \langle Y_n, \Phi^{(n)}(X_n) \rangle - nl \right) \right)_{n \geq 1} \quad (2.19)$$

converges in distribution to a centered normal random variable with variance

$$\sigma^2 := \mathbb{E} \left[ \left( \sum_{k \geq 0} (\mathbb{E}[\xi_{-k} | \mathcal{F}^0] - \mathbb{E}[\xi_{-k} | \mathcal{F}^1]) \right)^2 \right] \geq 0. \quad (2.20)$$

**Remark 2.6.** The proof will show that  $\sigma < \infty$ , but we have allowed the possibility that  $\sigma = 0$ . If  $\sigma = 0$ , the sequence in 2.19 converges to 0 in distribution (and hence in probability). Else, the sequence in 2.19 converges to a centered normal law with variance  $\sigma^2 > 0$ .

We prove Theorem 2 - Central Limit Theorem in §5 below.

The hypothesis eq. (2.18) of Theorem 2 - Central Limit Theorem may not be easy to verify directly. We close this section by introducing *mixing conditions* that are sufficient for eq. (2.18) to hold. Let

$$\mathcal{F}_n := \text{sigma algebra generated by } (\varphi_k)_{k \leq n}. \quad (2.21)$$

Note that  $(\mathcal{F}_n)_{n \in \mathbb{Z}}$  is a filtration, i.e.,  $\mathcal{F}_n \subset \mathcal{F}_{n+1}$ , while  $(\mathcal{F}^n)_{n \in \mathbb{Z}}$  (defined above in eq. (2.17)) is a reverse filtration, i.e.,  $\mathcal{F}^n \supset \mathcal{F}^{n+1}$ . We introduce the following *mixing coefficients*:

$$\alpha_n := \sup_{k \geq 0} \sup \left\{ |\mathbb{P}(A \cap B) - \mathbb{P}(A)\mathbb{P}(B)| : A \in \mathcal{F}_k, B \in \mathcal{F}^{n+k} \right\} \quad (2.22)$$

$$\rho_n := \sup_{k \geq 0} \sup \left\{ \left| \frac{\mathbb{E}[(Y - \mathbb{E}[Y])(X - \mathbb{E}[X])]}{\sigma(Y)\sigma(X)} \right| : Y \in L^2(\mathcal{F}_k), X \in L^2(\mathcal{F}^{n+k}), X, Y \neq 0 \right\} \quad (2.23)$$

We have the following:

**Theorem 3.** *If Assumption 2<sub>p</sub> holds with  $p > 2$  and  $\sum_{n \geq 1} \alpha_n^{(p-2)/p} < \infty$ , then*

$$\sum_{n=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q < \infty,$$

*with  $q$  the conjugate exponent to  $p$ . If Assumption 2<sub>p</sub> holds with  $p = 2$  and  $\sum_{n \geq 1} \rho_n < \infty$ , then*

$$\sum_{n=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_2 < \infty$$

Theorem 3 is proved in §6 below.

### 3 Background results: geometry of $\mathbb{S}_D$ , contraction for positive maps, and ergodic arguments

In this section we review some definitions and arguments from [22] that are fundamental to the proofs below.

#### 3.1 A metric on $\mathbb{S}_D$

Following [22], we define the following metric on  $\mathbb{S}_D$ :

$$d(A, B) := \frac{1 - m(A, B)m(B, A)}{1 + m(A, B)m(B, A)}, \quad (3.1)$$

where

$$m(A, B) = \sup\{\lambda : \lambda B \leq A\} \quad (3.2)$$

for  $A, B \in \mathbb{S}_D$ . The following lemma lists key properties of this metric (see [22, Lemma 3.3, 3.8, 3.9] for further details and proofs):

**Lemma 3.1.** *The function  $d$  defined in eq. (3.1) is a metric on  $\mathbb{S}_D$  satisfying:*

1.  $\frac{1}{2}\|A - B\| \leq d(A, B) \leq 1$  for  $A, B \in \mathbb{S}_D$ .
2.  $d(A, B) < 1$  for  $A, B \in \mathbb{S}_D^\circ$ .
3. If  $A \in \mathbb{S}_d^\circ$ , then  $d(A, B) = 1$  if and only if  $B \in \mathbb{S}_D \setminus \mathbb{S}_D^\circ$ .

4. The set  $\mathbb{S}_D^\circ$  is open in the metric topology generated by  $d$  and  $(\mathbb{S}_D^\circ, d)$  is homeomorphic to  $\mathbb{S}_D^\circ$  in the standard topology (generated by  $d_1(A, B) = \|A - B\|$ ).

In the proofs below, the following simple consequence of the lower bound  $\frac{1}{2}\|A - B\| \leq d(A, B)$  will be useful.

**Lemma 3.2.** *Let  $\phi \in \mathcal{L}(\mathbb{M}_D)$  be a positive map with the property that  $\ker \phi \cap \mathbb{S}_D = \emptyset$ . Then for all  $X, Y \in \mathbb{S}_D$ ;*

$$|\ln \|\phi(X)\| - \ln \|\phi(Y)\|| \leq 2 \frac{\|\phi\|}{v(\phi)} d(X, Y) , \quad (3.3)$$

with  $v(\phi)$  as in eq. (2.13).

*Remark 3.3.* For  $\phi = \varphi_n^*$ , we have  $\ker \phi \cap \mathbb{S}_D = \emptyset$  with probability one under the Assumption 1, see [22, Lemma 2.1]. Under Assumption 2,  $v(\phi)$  is non-zero with probability 1 and the right-hand-side of eq. (3.3) is finite almost surely.

*Proof.* Let  $g : (\mathbb{S}_D, \|\cdot\|) \rightarrow \mathbb{R}$  be defined as  $g(X) = \|\phi(X)\|$ . Since  $\phi$  is positive with no matrix in  $\mathbb{S}_D$  in its kernel we must have that  $g(X) > 0$  for all  $X \in \mathbb{S}_D$ . Since  $\mathbb{S}_D$  is compact in the standard topology, we have that

$$v(\phi) = \min\{\|\phi(Z)\| : Z \in \mathbb{S}_D\} > 0 . \quad (3.4)$$

It follows from the mean value inequality, applied to  $\ln$ , that

$$|\ln \|\phi(X)\| - \ln \|\phi(Y)\|| \leq \frac{\|\phi(X)\| - \|\phi(Y)\|}{v(\phi)} \leq \frac{\|\phi\| \|X - Y\|}{v(\phi)} \quad (3.5)$$

The results follows from lemma 3.1 as  $\|X - Y\| \leq 2d(X, Y)$ . ■

### 3.2 Contraction Coefficient for $\phi$

For any non-destructive positive map  $\phi \in \mathcal{L}(\mathbb{M}_D)$  we define the *contraction coefficient* of  $\phi$ , denoted  $c(\phi)$ , as follows:

$$c(\phi) = \sup\{d(\phi \cdot A, \phi \cdot B) : A, B \in \mathbb{S}_D\} . \quad (3.6)$$

We have the following properties of the contraction coefficient:

**Lemma 3.4** ([22, Lemma 3.14]). *If  $\phi \in \mathcal{L}(\mathbb{M}_D)$  be a non-destructive positive map, then*

1.  $d(\phi \cdot X, \phi \cdot Y) \leq c(\phi)d(X, Y)$  for all  $X, Y \in \mathbb{S}_D$  .
2.  $c(\phi) \leq 1$  and if  $\phi$  is strictly positive then  $c(\phi) < 1$ .
3. If there exist  $X, Y$  such that  $\phi \cdot X \in \mathbb{S}_D^\circ$  and  $\phi \cdot Y \in \mathbb{S}_D \setminus \mathbb{S}_D^\circ$ , then  $c(\phi) = 1$ .
4. For any non-destructive positive map  $\psi$ , we have  $c(\phi \circ \psi) \leq c(\phi)c(\psi)$ .
5. If  $\phi$  is also non-transient, then  $c(\phi) = c(\phi^*)$ .

*Remark 3.5.* We note that the lemma above is stated slightly differently than [22, Lemma 3.14]. However a close reading of the proof in [22] shows that the above version holds.

Under Assumption 1, the maps  $\Phi^n$  defined as in eq. (2.6) become strictly positive in finite time. As a consequence the following result was proved in [22] using Kingman's sub additive ergodic theorem [16, 17, 19]:

**Lemma 3.6** ([22, Lemma 3.11]). *Let  $(\varphi_n)_{n \geq 1}$  and  $\Phi^n$  be as in eq. (2.6). If Assumption 1 holds, then there exists a deterministic constant  $\kappa \in [0, 1]$  such that almost surely*

$$\ln \kappa = \lim_{n \rightarrow \infty} \frac{1}{n} \ln c(\Phi^{(n)})$$

and

$$\ln \kappa = \lim_{n \rightarrow \infty} \frac{1}{n} \mathbb{E} \ln c(\Phi^{(n)}) = \inf_{n \in \mathbb{N}} \frac{1}{n} \mathbb{E} \ln c(\Phi^{(n)}) .$$

**Remark 3.7.** In [22] the ergodic map  $\theta$  is assumed to be invertible. However, a close reading of the proof of [22, Lemma 3.11] shows that invertibility of  $\theta$  is not required.

Lemma 3.6 directly yields the following corollary:

**Corollary 3.8.**  $\lim_{n \rightarrow \infty} c(\Phi^{(n)}) = 0$  almost surely.

The contraction provided by Lemma 3.6 is the driving force behind the convergence  $L_n \rightarrow Z_1$  state in Lemma 2.2. In fact this convergence can be made more quantitative:

**Lemma 3.9** ([22, Lemma 3.12]). *Let  $(\varphi_n)_{n \geq 1}$  and  $\Phi^n$  be as in eq. (2.6) and suppose that Assumption 1 holds. Let  $L_n$  be as in eq. (2.10) and let  $Z_1 = \lim_n L_n$  and  $Z_k = Z_1 \circ \theta^{k-1}$  be as in Lemma 2.2. Then, for each  $Y \in \mathbb{S}_D$  and  $k \in \mathbb{N}$ ,*

$$d((\varphi_k^* \circ \dots \circ \varphi_n^*) \cdot Y, Z_k) \leq c(\varphi_k^* \circ \dots \circ \varphi_n^*)$$

for all sufficiently large  $n$ . In particular, we have  $\lim_n (\varphi_k^* \circ \dots \circ \varphi_n^*) \cdot Y = Z_k$  with probability one.

Below it will be useful to consider the contraction obtained from only a fraction of the process. This is described in the following

**Lemma 3.10.** *Let  $(\varphi_n)_{n \geq 1}$  and  $\Phi^n$  be as in eq. (2.6). Let  $\alpha \in (0, 1)$  and let  $n_\alpha = \lfloor (1 - \alpha)n \rfloor$ , the integer part of  $(1 - \alpha)n$ . If Assumption 1 holds, then*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) = \alpha \ln \kappa \text{ almost surely,} \quad (3.7)$$

where  $\kappa$  is the deterministic constant in Lemma 3.6.

*Proof.* First note that, by Part 4 of Lemma 3.4, we have

$$\ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) \geq \ln c(\varphi_n \circ \dots \circ \varphi_1) - \ln c(\varphi_{n_\alpha} \circ \dots \circ \varphi_1) . \quad (3.8)$$

Thus, by Lemma 3.6,

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) \geq \alpha \ln \kappa \text{ almost surely.} \quad (3.9)$$

To prove the complementary upper bound, i.e., that

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) \leq \alpha \ln \kappa , \quad (3.10)$$

we will show that for each  $m \in \mathbb{N}$

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) \leq \alpha \frac{1}{m} \mathbb{E}[\ln c(\Phi^{(m)})] \text{ almost surely.} \quad (3.11)$$

Eq. (3.10) will then follow by Lemma 3.4.

Let  $m \in \mathbb{N}$  be fixed and consider  $n \in \mathbb{N}$  large enough that  $n - n_\alpha > 2m$ . Let  $p(n) = \lfloor \frac{n_\alpha + m}{m} \rfloor$  and let  $q = q(n) \in \mathbb{N}$  and  $r = r(n) \in \{0, 1, \dots, m-1\}$  be defined by  $n = qm + r$ . Then,

$$n_\alpha + 1 \leq p(n)m + 1 \leq n_\alpha + m < n - m + 1 \leq q(n)m. \quad (3.12)$$

Since  $\ln c(\varphi) \leq 0$  for any  $\varphi \in \mathcal{L}(\mathbb{M}_D)$ , we have, using lemma 3.4, that

$$\ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) \leq \ln c(\varphi_{q(n)m+j} \circ \dots \circ \varphi_{p(n)m+j+1}) \quad (3.13)$$

for any  $0 \leq j \leq m-1$ , where eq. (3.12) guarantees that  $p(n)m + j + 1 \geq 1$  and the composition on the right hand side has non-zero number of factors. Using, 3.4 again we find that

$$\ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) \leq \sum_{k=p(n)}^{q(n)-1} \ln c(\varphi_{km+j+m} \circ \dots \circ \varphi_{km+j+1}) = \sum_{k=p(n)}^{q(n)-1} \ln c(\varphi_m \circ \dots \circ \varphi_1) \circ \theta^{km+j}.$$

Since this holds for any  $j \in \{0, 1, \dots, m-1\}$ , we have

$$\begin{aligned} \ln c(\varphi_n \circ \dots \circ \varphi_{n_\alpha+1}) &\leq \frac{1}{m} \sum_{j=0}^{m-1} \sum_{k=p(n)}^{q(n)-1} \ln c(\varphi_m \circ \dots \circ \varphi_1) \circ \theta^{km+j} \\ &= \frac{1}{m} \sum_{i=p(n)m}^{q(n)m-1} \ln c(\varphi_m \circ \dots \circ \varphi_1) \circ \theta^i \\ &= \sum_{i=0}^{q(n)m-1} \frac{1}{m} \ln c(\Phi^{(m)}) \circ \theta^i - \sum_{i=0}^{p(n)m-1} \frac{1}{m} \ln c(\Phi^{(m)}) \circ \theta^i. \end{aligned}$$

Since  $(\frac{1}{m} \ln c(\varphi_m \circ \dots \circ \varphi_1))^+ \in L^1(\Omega)$  (where  $(\cdot)^+$  denotes the positive part), eq. (3.10) follows from the Birkhoff ergodic theorem.  $\blacksquare$

### 3.3 Invertible ergodic dynamics

In this section, we assume that  $\theta$  is an invertible ergodic map. It is often possible to replace the original dynamical system by a natural extension on which  $\theta$  is invertible; for instance this is possible if  $\theta$  is *essentially surjective*, i.e. if  $\Omega \setminus \theta(\Omega)$  is a sub-null set —see [7]. We will denote this extension also by  $(\Omega, \mathcal{F}, \mathbb{P}, \theta)$  and note that the previously stated results still hold.

Since  $\theta$  is invertible and measure preserving, the inverse map  $\theta^{-1}$  is also a measure preserving ergodic transformation. We extend the definition of  $(\Phi^{(n)})$  to include negative indices as follows

$$\Phi^{(n)}(\omega) = \begin{cases} \varphi_n(\omega) \circ \dots \circ \varphi_1(\omega) & \text{for } n \geq 1, \\ \varphi_0 & \text{for } n = 0, \\ \varphi_{-1}(\omega) \circ \dots \circ \varphi_n(\omega) & \text{for } n \leq -1, \end{cases} \quad (3.14)$$

where  $\varphi_n := \varphi_{\theta^n}$  for all  $n$ . When  $\theta$  is invertible, Assumption 1 guarantees that with probability one  $(\Phi^{(-n)})_{n \geq 1}$  is almost surely eventually strictly positive — see [22, Lemma 3.13].

With this extended dynamical system, we introduce some new notation. Let  $n \in \mathbb{N}$  and define

$$\psi_n = \varphi_{-n}^* \quad \text{and} \quad \Psi^{(n)} = \psi_n \circ \dots \circ \psi_1. \quad (3.15)$$

Note that  $\Psi^{(n)*} = \Phi^{(-n)}$ . We see that  $(\Psi^{(n)})_{n \in \mathbb{N}}$  is almost surely eventually strictly positive. This allows us to define a new stopping time  $\tau'$  as:

$$\tau' = \inf\{n \geq 1 : \Phi^{(n+k)} \text{ and } \Psi^{(n+k)} \text{ are strictly positive } \forall k \geq 0\}, \quad (3.16)$$

satisfying  $\mathbb{P}[\tau' < \infty] = 1$  if  $\theta$  is invertible and Assumption 1 holds.

We have the following result analogous to Lemma 3.6 for the sequence  $(\Psi^{(n)})_{n \geq 1}$ :

**Lemma 3.11.** *If  $\theta$  is invertible and  $(\Phi_{(n)})_{n \geq 1}$  satisfies Assumption 1, then*

$$\ln \kappa \stackrel{a.s.}{=} \lim_{n \rightarrow \infty} \frac{1}{n} \ln c(\Psi^{(n)}), \quad (3.17)$$

where  $\Psi^{(n)}$  is as in eq. (3.15) and  $\kappa$  is the deterministic constant appearing in lemma 3.6. In particular,  $\lim_n c(\Psi^{(n)}) = 0$  almost surely.

**Remark 3.12.** The existence of the deterministic limit on the right hand side of eq. (3.17) follows directly from Lemma 3.6 applied with the sequence  $\Psi^{(n)}$  in place of  $\Phi^{(n)}$ . That the limit equals  $\kappa$  follows from the identity

$$\mathbb{E} \ln c(\Psi^{(n)}) = \mathbb{E} \ln c(\Phi_{\theta^{-n-1}}^{(n)*}) = \mathbb{E} \ln c(\Phi^{(n)}),$$

where we have used the facts that  $\theta$  is measure preserving and that  $c(\phi^*) = c(\phi)$  for any  $\phi$ .

If  $\theta$  is invertible and Assumption 1 holds, then the left and right Perron-Frobenius eigenmatrices  $R_n$  and  $L_n$  for  $\Phi^{(n)}$  exist also for large negative  $n$ . As a result we have the following lemma for the convergence of the right eigenvectors:

**Lemma 3.13** ([22, Lemma 3.14]). *Let  $(\varphi_n)_{n \geq 1}$  and  $\Phi^n$  be as in eq. (2.6) and let  $R_n$  be the right Perron-Frobenius eigenmatrix for  $\Phi^{(n)}$ , see eq. (2.10). If  $\theta$  is invertible and Assumption 1 holds, then there is an  $\mathbb{S}_D^\circ$  valued random variable  $Z'_1$  such that*

$$\lim_{n \rightarrow -\infty} R_n \stackrel{a.s.}{=} Z'_1 \quad (3.18)$$

and, with  $Z'_k := Z'_1 \circ \theta^{-k+1}$ , we have:

1. for every  $k \in \mathbb{N}$ ,  $\psi_k^* \cdot Z'_{k+1} = Z'_k$  a.s., and
2. for each  $Y \in \mathbb{S}_D$  and  $k \in \mathbb{N}$ ,

$$d((\psi_k^* \circ \dots \circ \psi_n^*) \cdot Y, Z'_k) \leq c(\psi_k^* \circ \dots \circ \psi_n^*)$$

for all sufficiently large  $n$ . In particular, we have  $\lim_n (\psi_k^* \circ \dots \circ \psi_n^*) \cdot Y = Z'_k$  a.s..

If instead we take  $n \rightarrow \infty$ , we do not have almost sure convergence of  $R_n$ . However, we do have convergence in distribution:

**Corollary 3.14.** *We have that*

$$R_n \xrightarrow[n \rightarrow \infty]{d} Z'_1 \quad \text{and} \quad L_n \xrightarrow[n \rightarrow -\infty]{d} Z'_1, \quad (3.19)$$

where  $\xrightarrow{d}$  denotes convergence in distribution.

*Proof.* Note that  $R_n = R_{-n; \theta^{n+1}}$ , so that  $R_n \stackrel{d}{=} R_{-n}$ . Since  $\lim_{n \rightarrow \infty} R_{-n} = Z'_1$  a.s., the first limit holds. The proof for the second limit is similar.  $\blacksquare$

## 4 Proof of the Law of Large Numbers

We now describe the proof of Theorem 1 - Law of Large Numbers. Recall from the discussion following the statement of the theorem above, that it suffices to prove eq. (2.14), which states that

$$\lim_{n \rightarrow \infty} \sup_{X, Y \in \mathbb{S}_D} \left| \frac{1}{n} \ln \langle Y, \Phi^{(n)}(X) \rangle - l \right| = 0 \quad \text{a.s.}$$

To this end, note that by Assumption 2 we have  $\mathbb{E}[\ln \|\varphi_k^*(Z_{k+1})\|] < \infty$  for each  $k \in \mathbb{N}$ . Thus by Birkhoff's ergodic theorem we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n \ln \|\varphi_k^*(Z_{k+1})\| \stackrel{a.s.}{=} \mathbb{E} \ln \|\varphi_0^*(Z_1)\| := l.$$

Thus eq. (2.14), and therefore Theorem 1 - Law of Large Numbers, follows from the following

**Lemma 4.1.** *Suppose that Assumption 1 holds and let*

$$D_n = \sup_{X, Y \in \mathbb{S}_D} \{ |\ln \langle Y, \Phi^{(n)}(X) \rangle - \ln \|\Phi^{(n)*}(Y)\| | \} , \quad (4.1)$$

and

$$E_n := \sup_{Y \in \mathbb{S}_D} \left\{ \left| \ln \|\Phi^{(n)*}(Y)\| - \sum_{k=1}^n \ln \|\varphi_{k,\omega}^*(Z_{k+1})\| \right| \right\} . \quad (4.2)$$

for  $n \geq 1$ . Then, with probability one,

1.  $D_n$  is eventually bounded, i.e.,  $\limsup_{n \rightarrow \infty} D_n < \infty$ , and
2.  $\lim_{n \rightarrow \infty} \frac{1}{n} E_n = 0$ .

**Remark 4.2.** Note that from  $\lim_n \frac{1}{n} E_n$  we conclude directly that  $\lim_n \frac{1}{n} \ln \|\Phi^{(n)*}(Y)\| = l$  for every  $Y \in \mathbb{S}_D$ . In particular  $l = \lambda$ , the top Lyapunov exponent of  $\Phi^{(n)}$ , as claimed in the discussion following Theorem 1 - Law of Large Numbers above.

*Proof.* First note that for any  $X, Y \in \mathbb{S}_D$ ,

$$\langle Y, \Phi^{(n)}(X) \rangle = \langle \Phi^{(n)*}(Y), X \rangle \leq \langle \Phi^{(n)*}(Y), \mathbb{I} \rangle = \|\Phi^{(n)*}(Y)\| , \quad (4.3)$$

Here we have used that  $X \leq \mathbb{I}$  and  $\text{tr}M = \|M\|$  for any positive semi-definite matrix.

For the rest of the proof, we restrict to a configuration  $\omega$  such that  $\tau = \tau_\omega < \infty$ . Such configurations form a full measure set by Assumption 1.

Because  $\Phi^{(\tau)}$  is strictly positive, we have  $\min \sigma(\Phi^{(\tau)}(P)) > 0$  for any projection  $P$ , where  $\sigma(\Phi^{(\tau)}(P))$  denotes the spectrum of  $\Phi^{(\tau)}(P)$ . Thus the map  $P \mapsto \min \sigma(\Phi^{(\tau)}(P))$  is a continuous function from the set of rank-1 projections into  $(0, \infty)$ . Since the set of rank-1 projections is compact, we have

$$a := \min \{ \min(\sigma(\Phi^{(\tau)}(P))) : P \text{ is a rank-1 projection} \} > 0.$$

Given  $X, Y \in \mathbb{S}_D$  and  $n > \tau$ , let  $W = \varphi_{\tau+1}^* \circ \dots \circ \varphi_n^*(Y)$ . Because  $X$  has at least one eigenvalue greater than or equal to  $\frac{1}{D}$ , we have  $X \geq \frac{1}{D}P$  for some rank-1 projection  $P$ , and thus

$$\langle Y, \Phi^{(n)}(X) \rangle = \langle W, \Phi^{(\tau)}(X) \rangle \geq \frac{1}{D} \langle W, \Phi^{(\tau)}(P) \rangle \geq \frac{a}{D} \langle W, \mathbb{I} \rangle = \frac{a}{D} \|W\| .$$

Since  $\|\Phi^{(n)*}(Y)\| = \|\Phi^{(\tau)*}(W)\| \leq \|\Phi^{(\tau)*}\| \|W\|$ , we have

$$\langle Y, \Phi^{(n)}(X) \rangle \geq \frac{a}{D \|\Phi^{(\tau)*}\|} \|\Phi^{(\tau)*}(Y)\| . \quad (4.4)$$

Putting eqs. (4.3) and (4.4) together, we see that

$$\ln a - \ln D - \ln \|\Phi^{(\tau)*}\| \leq \ln \langle Y, \Phi^{(n)}(X) \rangle - \ln \|\Phi^{(\tau)*}(Y)\| \leq 0$$

for  $X, Y \in \mathbb{S}_D$  and  $n > \tau$ . It follows that  $\limsup_n D_n \leq \ln D + \ln \|\Phi^{(\tau)*}\| - \ln a < \infty$  whenever  $\tau < \infty$ .

Turning now to the proof that  $\lim_n \frac{1}{n} E_n = 0$ , consider  $n > \tau$ . Note that

$$\|\Phi^{(n)*}(Y)\| = \|\phi_1^*(\phi_2^* \circ \dots \circ \phi_n^*(Y))\| = \|\phi_1^*((\phi_2^* \circ \dots \circ \phi_n^*) \cdot Y)\| \|\phi_2^* \circ \dots \circ \phi_n^*(Y)\| ,$$

where in the final expression we have introduced the projective action by multiplying and dividing by  $\|\phi_2^* \circ \dots \circ \phi_n^*(Y)\| = \text{tr} \phi_2^* \circ \dots \circ \phi_n^*(Y)$ . Taking logarithms and iterating, we find that

$$\ln \|\Phi^{(n)*}(Y)\| = \sum_{k=1}^n \ln \|\varphi_k^*((\varphi_{k+1}^* \circ \dots \circ \varphi_n^*) \cdot Y)\| ,$$

where the empty composition  $\varphi_{n+1}^* \circ \dots \circ \varphi_n^*$  is understood as the identity map. Thus

$$E_n(Y) := \left| \ln \|\Phi^{(n)*}(Y)\| - \sum_{k=1}^n \ln \|\varphi_k^*(Z_{k+1})\| \right| \leq \sum_{k=1}^n E_n^k(Y) ,$$

with  $E_n^k(Y) := \left| \ln \|\varphi_k^*((\varphi_{k+1}^* \circ \dots \circ \varphi_n^*) \cdot Y)\| - \ln \|\varphi_k^*(Z_{k+1})\| \right| . \quad (4.5)$

Using Lemma 3.2, Remark 3.3, and Lemma 3.9 we may bound  $E_n^k(Y)$  as follows

$$E_n^k(Y) \leq 2 \frac{\|\varphi_k^*\|}{v(\varphi_k^*)} c(\varphi_{k+1}^* \circ \dots \circ \varphi_n^*) . \quad (4.6)$$

Now let  $\alpha \in (0, 1)$  and let  $n_\alpha$  be the integer part of  $(1 - \alpha)n$ . We will bound the terms on the right hand side of (4.5) differently according to if  $k < n_\alpha$  or  $k \geq n_\alpha$ . For  $k < n_\alpha$ , we have

$$E_n^k(Y) \leq 2 \frac{\|\varphi_k^*\|}{v(\varphi_k^*)} c(\varphi_{n_\alpha}^* \circ \dots \circ \varphi_n^*) ,$$

where we have used eq. (4.6) and applied Lemma 3.4 to bound  $c(\varphi_{k+1}^* \circ \dots \circ \varphi_n^*) \leq c(\varphi_{n_\alpha}^* \circ \dots \circ \varphi_n^*)$ . For  $k \geq n_\alpha$ , on the other hand, we have

$$E_n^k(Y) \leq \left| \ln \|\varphi_k^*((\varphi_{k+1}^* \circ \dots \circ \varphi_n^*) \cdot Y)\| \right| + \left| \ln \|\varphi_k^*(Z_{k+1})\| \right| \leq 2 \left( |\ln v(\varphi_k^*)| + |\ln \|\varphi_k^*\|| \right) .$$

Thus

$$E_n = \sup_{Y \in \mathbb{S}_D} E_n(Y) \leq S_n^< + S_n^> \quad (4.7)$$

with

$$S_n^< = 2 \sum_{k=1}^{n_\alpha-1} \frac{\|\varphi_k^*\|}{v(\varphi_k^*)} c(\varphi_{n_\alpha}^* \circ \dots \circ \varphi_n^*) ,$$

and

$$S_n^> = 2 \sum_{k=n_\alpha}^n \left( |\ln v(\varphi_k^*)| + |\ln \|\varphi_k^*\|| \right) .$$

We will prove that  $\lim_n S_n^< = 0$  and  $\lim_n \frac{1}{n} S_n^> = O(\alpha)$ .

Note that by Assumption 2 we have  $\mathbb{E}[|\ln(\frac{\|\varphi_0^*\|}{v(\varphi_0^*)})|] < \infty$ . Thus, for any  $\epsilon > 0$ ,

$$\begin{aligned} \infty &> \frac{1}{\epsilon} \mathbb{E}[|\ln(\frac{\|\varphi_0^*\|}{v(\varphi_0^*)})|] \geq \sum_{k=1}^{\infty} \mathbb{P}\left(\ln(\frac{\|\varphi_0^*\|}{v(\varphi_0^*)}) > k\epsilon\right) \\ &= \sum_{k=1}^{\infty} \mathbb{P}\left(\ln(\frac{\|\varphi_0^*\|}{v(\varphi_0^*)}) > k\epsilon\right) = \sum_{k=1}^{\infty} \mathbb{P}\left(\frac{\|\varphi_0^*\|}{v(\varphi_0^*)} > e^{k\epsilon}\right) . \end{aligned}$$

Hence, by the Borel-Cantelli Lemma, we find that  $\limsup_k e^{k\epsilon} \frac{\|\varphi_k^*\|}{v(\varphi_k^*)} \leq 1$  with probability one. Taking  $\epsilon < \alpha |\ln \kappa|$ , we conclude from Lemma 3.10 that

$$\limsup_{n \rightarrow \infty} S_n^< \leq \limsup_{n \rightarrow \infty} n_\alpha e^{\epsilon n_\alpha} c((\varphi_{n_\alpha}^* \circ \dots \circ \varphi_n^*)) = 0.$$

In particular, we also have  $\lim_n \frac{1}{n} S_n^< = 0$ .

Now consider  $S_n^>$ . Since  $\ln v(\varphi_0^*)$  and  $\ln \|\varphi_0^*\|$  are  $L^1$  random variables by Assumption 1, we conclude from the Birkhoff ergodic theorem [3] that

$$\lim_n \frac{1}{n} S_n^> = 2\alpha [\mathbb{E} |\ln \|\varphi_0^*\|| + \mathbb{E} |\ln v(\varphi_0^*)|].$$

We conclude that  $\limsup_n \frac{1}{n} E_n = O(\alpha)$ . Since  $\alpha \in (0, 1)$  was arbitrary, we have  $\lim_n \frac{1}{n} E_n = 0$ .  $\blacksquare$

## 5 Proof the Central Limit Theorem

In this section we prove Theorem 2 - Central Limit Theorem. Let  $(X_n)_{n \geq 1}$  and  $(Y_n)_{n \geq 1}$  be sequences in  $\mathbb{S}_n$ . Then

$$\begin{aligned} \frac{1}{\sqrt{n}} (\ln \langle Y_n, \Phi^{(n)}(X_n) \rangle - nl) &= \frac{1}{\sqrt{n}} (\ln \langle Y_n, \Phi^{(n)}(X_n) \rangle - \ln \|\Phi^{(n)*}(Y_n)\|) \\ &\quad + \frac{1}{\sqrt{n}} (\ln \|\Phi^{(n)*}(Y_n)\| - \sum_{k=1}^n \ln \|\varphi_k^*(\omega)(Z_{k+1}(\omega))\|) + \frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k, \end{aligned}$$

where  $\xi_k = \ln \|\varphi_k^*(Z_{k+1})\| - l$ . Thus

$$\left| \frac{1}{\sqrt{n}} (\ln \langle Y_n, \Phi^{(n)}(X_n) \rangle - nl) - \frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k \right| \leq \frac{1}{\sqrt{n}} (D_n + E_n)$$

with  $D_n$  and  $E_n$  as in eqs. (4.1) and (4.2), respectively. By Lemma 4.1,  $D_n$  is almost surely eventually bounded. Thus to prove that  $(\frac{1}{\sqrt{n}} \ln \langle Y_n, \Phi^{(n)}(X_n) \rangle)_{n \geq 1}$  converges in distribution to a centered normal variable, it suffices to prove the following two results:

1.  $\frac{1}{\sqrt{n}} E_n$  converges to 0 in probability, and
2.  $Q_n := \frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k$  converges in distribution to a centered normal variable with variance given by eq. (2.20) above.

These results are proved in Lemma 5.1 and Lemma 5.2 below, respectively.

**Lemma 5.1.** *Suppose that  $\theta$  is invertible and that Assumption 1 holds. Let  $(E_n)_{n=1}^\infty$  be the variables defined in eq. (4.2). Then  $(E_n)_{n=1}^\infty$  is tight. In particular,  $(\frac{1}{\sqrt{n}} E_n)_{n=1}^\infty$  converges to 0 in probability.*

*Proof.* Following the proof of eq. (4.7) above, but applying in the proof of Lemma 4.1, we have

$$E_n \leq S_n := 2 \sum_{k=1}^n \frac{\|\varphi_k^*\|}{v(\varphi_k^*)} c(\varphi_{k+1}^* \circ \dots \circ \varphi_n^*).$$

We prove that  $E_n$  are tight by showing that  $S_n \stackrel{d}{=} S'_n$  where the random variables  $S'_n$  satisfy  $\sup_n S'_n < \infty$  almost surely.

Consider the variables  $S'_n = S_{n;\theta^{-n}}$ . Since  $c(\phi^*) = c(\phi)$ , we have

$$S'_n = 2 \sum_{k=0}^{n-1} \frac{\|\varphi_{-k}^*\|}{v(\varphi_{-k}^*)} c(\varphi_0 \circ \dots \circ \varphi_{1-k}) .$$

As above the empty composition appearing at  $k = 0$  is understood as the identity map. By the Borel-Cantelli similar to that used to bound  $S_n^<$  in the proof of Lemma 4.1 we see that

$$\limsup_{k \rightarrow \infty} \frac{1}{k} \ln \left( \frac{\|\varphi_{-k}^*\|}{v(\varphi_{-k}^*)} \right) = 0 \quad \text{a.s.}$$

On the other hand by Lemma 3.11 we have

$$\limsup_{k \rightarrow \infty} \frac{1}{k} \ln c(\varphi_0 \circ \dots \circ \varphi_{1-k}) = \ln \kappa < 0 \quad \text{a.s.}$$

It follows that

$$\lim_{n \rightarrow \infty} S'_n = 2 \sum_{k=0}^n \frac{\|\varphi_{-k}^*\|}{v(\varphi_{-k}^*)} c(\varphi_0 \circ \dots \circ \varphi_{1-k}) =: S'_\infty$$

is finite almost surely. Clearly  $S'_\infty = \sup_n S'_n$ .

Since  $S_n \stackrel{d}{=} S'_n$  we have

$$\mathbb{P}[E_n > \epsilon] \leq \mathbb{P}[S_n > \epsilon] = \mathbb{P}[S'_n > \epsilon] \leq \mathbb{P}[S'_\infty > \epsilon] ,$$

so  $(E_n)_{n=1}^\infty$  is tight as claimed. It follows that

$$\mathbb{P}\left[\frac{1}{\sqrt{n}} E_n > \epsilon\right] \leq \mathbb{P}[S'_\infty > \sqrt{n}\epsilon] \rightarrow 0 ,$$

so  $\frac{1}{\sqrt{n}} E_n$  converges to zero in probability. ■

To prove the convergence of  $Q_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k$  to a centered normal law in distribution, we use the martingale approximation method of Gordin [13]. The following proof is adapted from the proof of [14, Lemma 9.2] and is similar to the proof of [20, Theorem 1.1]. The key idea is to find a *reverse martingale difference* with respect to the filtration  $(\mathcal{F}^n)_{n \geq 1}$  and use the Central Limit Theorem for (reverse) martingale differences [2, 4, 6] which was proved independently by Billingsly [1] and [15] for the ergodic case:

**Martingale Difference Central Limit Theorem.** *Let  $(X_n)_{n \geq 1}$  be a stationary ergodic direct or reversed martingale difference with respect to a filtration  $\{\mathcal{A}_n\}_{n \geq 1}$ . If  $X_1 \in L^2$ , then  $(\frac{1}{\sqrt{n}} \sum_{k=1}^n X_k)_{n \geq 1}$  converges in distribution to a centered normal random variable with variance  $\sigma^2 = \mathbb{E}(X_1^2)$ .*

**Lemma 5.2.** *Suppose that  $\theta$  is invertible, that Assumption 1 holds, and Assumption 2<sub>p</sub> holds for some  $p \geq 2$ . Let  $\xi_k = \ln \|\varphi_k^*(Z_{k+1})\| - l$  for  $k \in \mathbb{Z}$ . If*

$$\sum_{n=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q < \infty , \tag{5.1}$$

*with  $\frac{1}{p} + \frac{1}{q} = 1$ , then the sequence  $(Q_n)_{n=1}^\infty$  given by*

$$Q_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k \tag{5.2}$$

*converges in distribution to a centered normal law with variance  $\sigma^2 < \infty$ . Furthermore  $\sigma = 0$  if and only if there exists stationary sequence  $(g_n)_{n \geq 1}$  such that*

$$g_n \in L^q(\mathcal{F}^n) \quad \text{and} \quad \xi_n = g_{n+1} - g_n \tag{5.3}$$

*Proof.* Let  $M := \sum_{k=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^k]\|_q < \infty$  by eq. (5.1). We define

$$g_0 := \sum_{k=1}^{\infty} \mathbb{E}[\xi_{-k} | \mathcal{F}^0], \quad (5.4)$$

and note that

$$\|g_0\|_q \leq \sum_{k=1}^{\infty} \|\mathbb{E}[\xi_{-k} | \mathcal{F}^0]\|_q = M,$$

since  $\theta$  is measure preserving. Since  $\|\cdot\|_1 \leq \|\cdot\|_q$ , the series defining  $g_0$  converges in  $L^1$  and hence absolutely, almost everywhere.

We define

$$\zeta_0 = \sum_{k=0}^{\infty} (\mathbb{E}[\xi_{-k} | \mathcal{F}^0] - \mathbb{E}[\xi_{-k} | \mathcal{F}^1]), \quad (5.5)$$

and note that  $\zeta_0 = \xi_0 + g_0 - g_0 \circ \theta$ . For  $n \in \mathbb{Z}$ , we now define  $\zeta_n = \zeta_0 \circ \theta^n$  and  $g_n = g_0 \circ \theta^n$ , so that

$$\xi_n = \zeta_n + g_{n+1} - g_n. \quad (5.6)$$

Since

$$|\xi_n| \leq \ln \|\phi_n^*\| + \ln v(\phi_n^*) + |l|, \quad (5.7)$$

we have  $\xi_n \in L_p \subset L_q$  by Assumption 2<sub>p</sub>, so  $\zeta_n = \xi_n - g_{n+1} + g_n \in L_q \subset L^1$ . Taking conditional expectation with respect to  $\mathcal{F}^{n+1}$  in eq. (5.5), we see that

$$\mathbb{E}[\zeta_n | \mathcal{F}^{n+1}] = 0, \quad (5.8)$$

i.e.,  $(\zeta_n)_{n \geq 1}$  is a reverse martingale difference (*reverse* because  $(\mathcal{F}^n)_{n \geq 1}$  is a reverse filtration). Now eq. (5.6) shows that

$$\frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k = \frac{1}{\sqrt{n}} \sum_{k=1}^n \zeta_n + \frac{1}{\sqrt{n}} (g_{n+1} - g_1). \quad (5.9)$$

Since  $g_{n+1} = g_1 \circ \theta^n$ , we see that  $g_{n+1} - g_1$  is tight and thus  $\frac{1}{\sqrt{n}} (g_{n+1} - g_1)$  converges to 0 in probability. Therefore, by the Martingale Difference Central Limit Theorem, we will have the required convergence in distribution if we establish that  $\zeta_0 \in L^2$ .

Since  $\zeta_0 = \xi_0 - (g_1 - g_0)$  and  $\xi_0 \in L_p \subset L_2$  by eq. (5.7), it suffices to show that  $g_1 - g_0 \in L^2$ . We have  $g_n \in L^q(\mathcal{F}^n)$ , but this does not suffice as  $q < 2$ . To show that  $g_1 - g_0 \in L^2$  we need to exploit cancellation between the two terms. To this end, let  $\lambda \in (0, 1)$  and define

$$g_0^\lambda = \sum_{k=1}^{\infty} \lambda^{k-1} \mathbb{E}[\xi_{-k} | \mathcal{F}^0], \quad (5.10)$$

and define  $g_n^\lambda = g_0^\lambda \circ \theta^n$  for  $n \in \mathbb{Z}$ . Since  $\|\mathbb{E}[\xi_{-k} | \mathcal{F}^0]\|_p \leq \|\xi_{-k}\|_p = \|\xi_0\|_p$ , the convergence factor  $\lambda^{k-1}$  in eq. (5.10) guarantees that  $g_0^\lambda \in L^p \subset L^2$ . Furthermore, we have

$$\|g_0^\lambda\|_q \leq \sum_{k=1}^{\infty} \lambda^{k-1} \|\mathbb{E}[\xi_{-k} | \mathcal{F}^0]\|_q \leq M, \quad (5.11)$$

since  $\lambda \leq 1$ .

We will now show that  $\|g_1^\lambda - \lambda g_0^\lambda\|_2^2$  is bounded uniformly in  $\lambda$ . We start with the estimate

$$\begin{aligned} \|g_1^\lambda - \lambda g_0^\lambda\|_2^2 &= (1 + \lambda^2) \|g_1^\lambda\|_2^2 - 2\lambda \mathbb{E}[g_0^\lambda g_1^\lambda] \\ &\leq 2[\|g_1^\lambda\|_2^2 - \lambda \mathbb{E}[g_0^\lambda g_1^\lambda]] = 2\mathbb{E}[g_1^\lambda (g_1^\lambda - \lambda \mathbb{E}[g_0^\lambda | \mathcal{F}^1])], \end{aligned}$$

where we have noted that  $\|g_1^\lambda\|_2 = \|g_0^\lambda\|_2$  (since  $(g_n^\lambda)_{n=1}^\infty$  is stationary) and that  $g_1^\lambda$  is  $\mathcal{F}^1$  measurable. Note that

$$g_1^\lambda - \lambda \mathbb{E}[g_0^\lambda | \mathcal{F}^1] = \sum_{k=1}^{\infty} \lambda^{k-1} \mathbb{E}[\xi_{-k+1} | \mathcal{F}^1] - \lambda \sum_{k=1}^{\infty} \lambda^{k-1} \mathbb{E}[\xi_{-k} | \mathcal{F}^1] = \mathbb{E}[\xi_0 | \mathcal{F}^1].$$

Thus

$$\|g_1^\lambda - \lambda g_0^\lambda\|_2^2 \leq 2 \int_{\Omega} \mathbb{E}[\xi_0 | \mathcal{F}^1] g_1^\lambda d\mathbb{P} \leq 2 \|\mathbb{E}[\xi_0 | \mathcal{F}^1]\|_p \|g_1^\lambda\|_q \leq 2 \|\xi_0\|_p M,$$

where we have used Hölder's inequality and eq. (5.11).

Since  $g_1 - g_0 = \lim_{\lambda \uparrow 1} g_1^\lambda - \lambda g_0^\lambda$ , we have

$$\mathbb{E}[(g_1 - g_0)^2] = \mathbb{E}\left[\lim_{\lambda \uparrow 1} (g_1^\lambda - \lambda g_0^\lambda)^2\right] \leq \liminf_{\lambda \uparrow 1} \mathbb{E}[(g_1^\lambda - \lambda g_0^\lambda)^2] \leq 2 \|\xi_0\|_p M,$$

by Fatou's Lemma. Therefore  $g_1 - g_0 \in L^2$ . Thus  $\zeta_n \in L^2$  for each  $n$  and the martingale difference central limit theorem implies that  $(\frac{1}{\sqrt{n}} \sum_{k=1}^n \zeta_k)_{n \geq 1}$  (and thus  $(\frac{1}{\sqrt{n}} \sum_{k=1}^n \xi_k)_{n \geq 1}$ ) converges in distribution to a centered normal random variable with variance  $\sigma^2 = \mathbb{E}[\zeta_0^2]$ .

If  $\sigma = 0$  then we have that  $\zeta_n = 0$  a.s. for each  $n \in \mathbb{Z}$ . In this case, we have  $\xi_n = g_{n+1} - g_n$  for the stationary processes  $(g_n)_{n \in \mathbb{Z}}$  defined above. This concludes the proof of lemma 5.2 ■

This completes the proof of Theorem 2 - Central Limit Theorem. In the next section we discuss the mixing conditions sufficient to prove the hypothesis eq. (2.18).

## 6 Mixing Conditions

In this section we prove Theorem 3, which provides sufficient conditions for the main hypothesis eq. (2.18) of Theorem 2 - Central Limit Theorem. The arguments in this section are based on similar results in [9] and [14]. We rely on the following estimate on averages of sub-multiplicative random variables that combines [14, Lemma 6.2 & Lemma 6.3] — see also [5, Lemma 3 & Lemma 4].

**Lemma 6.1** ([14]). *Consider a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  with an ergodic measure preserving map  $\theta : \Omega \rightarrow \Omega$ , a filtration  $(\mathcal{F}_n)_{n \geq 0}$ , and a reverse filtration  $(\mathcal{F}^n)_{n \geq 0}$ , such that  $\theta^{-1}(\mathcal{F}_{n+1}) = \mathcal{F}_n$  and  $\theta^{-1}(\mathcal{F}^{n+1}) = \mathcal{F}^n$  for each  $n \geq 0$ . Let  $\alpha_n$  and  $\rho_n$  be mixing coefficients defined as defined in eqs. (2.22) and (2.23), respectively. Let  $(M_n)_{n \geq 1}$  be a sequence of  $[0, 1]$ -valued random variables with the following sub-multiplicative property*

$$M_{m+n} \leq M_m M_n \circ \theta^n. \quad (6.1)$$

*If for each  $0 \leq m < n$  it holds that  $M_{n-m} \circ \theta^m$  is both  $\mathcal{F}_n$  and  $\mathcal{F}^m$  measurable, then we have:*

1. *If  $\alpha_n \leq cn^{-\lambda}$  with  $c, \lambda > 0$ , then  $\mathbb{E}[M_n]$  almost vanishes to order  $n^{-\lambda}$ ,*

$$\mathbb{E}[M_n] = O\left(\frac{a_n}{n}\right)^\lambda \quad (6.2)$$

*for any sequence  $(a_n)_{n \geq 1}$  of real numbers such that*

$$\lim_{n \rightarrow \infty} \frac{\ln n}{a_n} = \lim_{n \rightarrow \infty} \frac{a_n}{n} = 0. \quad (6.3)$$

2. *If  $\lim_{n \rightarrow \infty} \rho_n = 0$ , then  $\mathbb{E}[M_n]$  vanishes faster than any polynomial, i.e.,*

$$\mathbb{E}[M_n] = O\left(\frac{1}{n^k}\right). \quad (6.4)$$

*for each  $k \in \mathbb{N}$ ,*

Lemma 6.1 directly implies bounds on  $\mathbb{E}[c(\Phi^{(n)})]$ , stated in the following

**Lemma 6.2.** *Suppose that Assumptions 1 and 2 hold, and let  $\alpha_n$  and  $\rho_n$  be mixing coefficients defined as defined in eqs. (2.22) and (2.23), respectively. For  $r \in (0, 1)$  define*

$$\tau_r = \inf\{n \geq 1 : c(\Psi^{(n)}) \leq r \text{ \& } c(\Phi^{(n)}) \leq r\}.$$

*Then we have that  $\tau_r < \infty$  almost surely. Moreover*

1. *If  $\sum_{k=1}^{\infty} \alpha_k^{1/\lambda} < \infty$ , for some  $\lambda > 0$ , then*

$$\max\left\{\mathbb{P}[\tau_r > n], \mathbb{E}[c(\Phi^{(n)})]\right\} = O\left(\frac{\alpha_n}{n}\right)^{\lambda} \quad (6.5)$$

*for any sequence  $(a_n)_{n \geq 1}$  satisfying eq. (6.3).*

2. *If  $\lim_{n \rightarrow \infty} \rho_n = 0$ , then*

$$\max\left\{\mathbb{P}[\tau_r > n], \mathbb{E}[c(\Phi^{(n)})]\right\} = O\left(\frac{1}{n^k}\right) \quad (6.6)$$

*for any  $k \geq 1$ .*

*Proof.* From Corollary 3.8 and Lemma 3.11 we see that  $\mathbb{P}[\tau_r < \infty] = 1$ . We also have, by Assumption 2, that  $\phi_n$  is non-destructive and non-transient for all  $n \geq 1$ , with probability one. Therefore, we have that almost surely for all  $n \in \mathbb{Z}$ ,  $c(\Psi^{\tau_r+n}), c(\Phi^{\tau_r+n}) < r$ .

Suppose that  $\sum_{k=1}^{\infty} \alpha_k^{1/\lambda} < \infty$ . We start with the observation that  $\alpha_n$  is non-increasing in  $n$ ; this can be seen directly from the definition (2.22) of  $\alpha_n$  using the fact that  $(\mathcal{F}^n)_{n \geq 1}$  is decreasing in  $n$ . Since  $\alpha_n^{1/\lambda}$  is a non-increasing sequence of positive numbers with  $\sum_n \alpha_n^{1/\lambda} < \infty$ , we have  $\lim_{n \rightarrow \infty} n\alpha_n^{\lambda} = 0$ . Therefore we have  $\alpha_n \leq cn^{-\lambda}$ . Now notice that the sequence  $M_n = c(\Phi^{(n)})$ , for  $n \geq 1$ , satisfies the sub-multiplicative condition in Lemma 6.1. Therefore we obtain

$$\mathbb{E}[c(\Phi^{(n)})] \leq c_1 \left(\frac{\alpha_n}{n}\right)^{\lambda} \quad (6.7)$$

for any sequence  $(a_n)_{n \geq 1}$  satisfying eq. (6.3). A similar analysis can be applied to  $(c(\Psi^n))_{n \in \mathbb{N}}$ , resulting in

$$\mathbb{E}[c(\Phi^{(n)})] \leq c_2 \left(\frac{\alpha_n}{n}\right)^{\lambda}. \quad (6.8)$$

Since

$$\mathbb{P}[\tau_r > n] \leq \mathbb{P}[c(\Psi^{(n)}) > r] + \mathbb{P}[c(\Phi^{(n)}) > r] \leq \frac{1}{r} \mathbb{E}[c(\Phi^{(n)})] + \frac{1}{r} \mathbb{E}[c(\Psi^{(n)})], \quad (6.9)$$

we see that eq. (6.5) holds.

If  $\lim_{n \rightarrow \infty} \rho_n = 0$ , then the second part of Lemma 6.1 applies and eq. (6.9) still holds. These two combined give us eq. (6.6).  $\blacksquare$

We are now ready to state the main technical estimate of this section:

**Lemma 6.3.** *Suppose that Assumptions 1 and 2 hold, and let  $\alpha_n$  and  $\rho_n$  be mixing coefficients defined in eqs. (2.22) and (2.23), respectively. Let  $r \in (0, 1)$  and let  $\tau_r$  be as defined in Lemma 6.2. Let  $n_{\alpha}$  denote the integer part of  $(1 - \alpha)n$ , for  $\alpha \in (0, 1)$ .*

1. *If Assumption 2<sub>p</sub> holds with  $p > 2$  then there is  $K < \infty$  such that*

$$\|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q \leq K \left[ \alpha_{n-n_{\alpha}}^{(p-2)/p} + \mathbb{E}[c(\Phi^{(n_{\alpha})})] + (\mathbb{P}[\tau_r > n_{\alpha}])^{1/q} \right], \quad (6.10)$$

*with  $q$  the conjugate exponent to  $p$ .*

2. If Assumption  $2_p$  holds with  $p = 2$  then there is  $K < \infty$  such that

$$\|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_2 \leq K \left[ \rho_{n-n_\alpha} + \mathbb{E}[c(\Phi^{(n_\alpha)*})] + (\mathbb{P}(\tau_r > n_\alpha))^{1/2} \right]. \quad (6.11)$$

Before proving Lemma 6.3, let us show how it implies Theorem 3. First note that if  $(b_n)_{n \geq 1}$  is a sequence of non-negative numbers then

$$\sum_{n=1}^{\infty} b_{n_\alpha} \leq \frac{1}{1-\alpha} \sum_{n=1}^{\infty} b_n, \quad (6.12)$$

$$\sum_{n=1}^{\infty} b_{n-n_\alpha} \leq \frac{1}{\alpha} \sum_{n=1}^{\infty} b_n. \quad (6.13)$$

To see that eq. (6.12) holds, note that given  $m \in \mathbb{N}$ , the number of integers  $n$  such that  $n_\alpha = m$  is bounded by  $\frac{1}{1-\alpha}$ . The proof of eq. (6.13) is similar. Now suppose that Assumption  $2_p$  holds with  $p > 2$  and  $\sum_n \alpha_n^{(p-2)/p} < \infty$ . Then by Lemma 6.2, Lemma 6.3, and eqs. (6.12, 6.13), we have

$$\sum_{n=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q \leq K' \sum_{n \geq 1} \left[ \alpha_n^{\frac{p-2}{p}} + \left( \frac{a_n}{n} \right)^{\frac{p}{p-2}} + \left( \frac{a_n}{n} \right)^{\frac{p-2}{p} \frac{1}{q}} \right]$$

for a suitable  $K' < \infty$  and a slowly increasing sequence  $(a_n)_{n \geq 1}$  satisfying eq. (6.3). Since  $\frac{p}{p-2} > 1$  and  $\frac{p}{p-2} \frac{1}{q} = \frac{p-1}{p-2} > 1$  the right hand side is finite. Similarly, if  $\sum_n \rho_n < \infty$ , then we have

$$\sum_{n=1}^{\infty} \|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q \leq K' \sum_{n \geq 1} \left[ \rho_n + \frac{1}{n^k} \right]$$

for any  $k$ , which is clearly finite. This completes the proof of Theorem 3.

We now turn to the proof of Lemma 6.3:

*Proof of Lemma 6.3.* By Lemma 2.2, we have  $\Phi^{(n_\alpha)} \cdot Z_{n_\alpha+1} = Z_1$ . Therefore

$$\xi_0 = A_n + B_n - \mathbb{E}[A_n],$$

where

$$A_n = \ln \|\varphi_0^*(\Phi^{(n_\alpha)} \cdot Z_{n_\alpha+1})\| - \ln \|\varphi_0^*(\Phi^{(n_\alpha)} \cdot \frac{1}{D} \mathbb{I})\|$$

and

$$B_n = \ln \|\varphi_0^*(\Phi^{(n_\alpha)} \cdot \frac{1}{D} \mathbb{I})\| - \mathbb{E} \left[ \ln \|\varphi_0^*(\Phi^{(n_\alpha)} \cdot \frac{1}{D} \mathbb{I})\| \right].$$

Now consider the event  $\{\tau_r \leq n_\alpha\}$ . On this event,  $c(\Phi^{(n_\alpha)*}) \leq r$ . To bound  $A_n$  on this event we will use the following proposition which we prove below after completing the present proof.

**Proposition 6.4.** *Let  $\psi, \phi \in \mathcal{L}(\mathbb{M}_D)$ . Suppose that  $\psi$  is a positive map and  $\phi$  is a strictly positive map with  $c(\phi) \leq r < 1$ . If  $\psi$  is non-transient, then for any  $A, B \in \mathbb{S}_D$  we have*

$$|\ln \|\psi(\phi \cdot A)\| - \ln \|\psi(\phi \cdot B)\|| \leq c(\phi) \frac{2}{r} \ln \frac{1}{1-r}$$

Using Proposition 6.4, we see that

$$|A_n| \leq \frac{2}{r} \ln \frac{1}{1-r} c(\Phi^{(n_\alpha)*}) \mathbf{1}_{\tau_r \leq n_\alpha} + 2 (|\ln \|\varphi_0^*\|| + |\ln v(\varphi_0^*)|) \mathbf{1}_{\tau_r > n_\alpha} =: A'_n.$$

Therefore, using Hölder's inequality and Assumption 2<sub>p</sub>, we have

$$|\mathbb{E}A_n| \leq \mathbb{E}A'_n \leq C \left( \mathbb{E}[c(\Phi^{(n_\alpha)*})] + (\mathbb{P}(\tau_r > n_\alpha))^{(p-1)/p} \right) \quad (6.14)$$

with  $C < \infty$ . Furthermore we also have that

$$\sup_n \|A_n\|_p \leq \sup_n \|A'_n\|_p < \infty \text{ and } \sup_n \|B_n\|_p < \infty. \quad (6.15)$$

Now, for  $\frac{1}{p} + \frac{1}{q} = 1$  we have that

$$\|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q = \sup_{\{f \in L^p(\mathcal{F}^n) : \|f\|_p \leq 1\}} |\mathbb{E}[f \xi_0]|. \quad (6.16)$$

Hence to bound  $\|\mathbb{E}[\xi_0 | \mathcal{F}^n]\|_q < \infty$  it suffices to find a uniform upper bound for  $\mathbb{E}[\xi f]$  as  $f$  ranges over the unit ball in  $L^p(\mathcal{F}^n)$ . Since  $\xi_0 = A_n + B_n - \mathbb{E}[A_n]$  and  $\mathbb{E}[B_n] = 0$ , we have

$$\begin{aligned} |\mathbb{E}[\xi_0 f]| &\leq |\mathbb{E}[A_n f]| + |\mathbb{E}[B_n f]| + |\mathbb{E}[A_n] \mathbb{E}[f]| \\ &\leq \mathbb{E}[A'_n | f |] + |\mathbb{E}[B_n f] - \mathbb{E}[B_n] \mathbb{E}[f]| + \mathbb{E}[A'_n] \mathbb{E}[|f|] \\ &\leq |\mathbb{E}[A'_n | f |] - \mathbb{E}[A'_n] \mathbb{E}[|f|]| + |\mathbb{E}[B_n f] - \mathbb{E}[B_n] \mathbb{E}[f]| + 2\mathbb{E}[A'_n] \mathbb{E}[|f|] \end{aligned} \quad (6.17)$$

To estimating the right hand side we use the following covariance inequalities involving the mixing coefficients  $\alpha_n$  and  $\rho_n$ .

**Lemma 6.5** ([8, §1.2 Theorem 3] —see also [14, §6.2]). *For each  $n \in \mathbb{N}$ , Let  $\alpha_n$  and  $\rho_n$  be as defined in eqs. (2.22) and (2.23), respectively. For each  $n, k \in \mathbb{N}$ , we have*

$$|\mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]| \leq \rho_n \|X\|_2 \|Y\|_2$$

whenever  $X \in L^2(\mathcal{F}_k)$  and  $Y \in L^2(\mathcal{F}^{n+k})$ , and

$$|\mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]| \leq 8\alpha_n^{1/r} \|X\|_p \|Y\|_q$$

whenever  $X \in L^p(\mathcal{F}_k)$  and  $Y \in L^q(\mathcal{F}^{k+n})$  with  $p, q, r \in [1, \infty]$  such that  $\frac{1}{p} + \frac{1}{q} + \frac{1}{r} = 1$ .

We note that  $A_n, B_n \in L^p(\mathcal{F}_{n_\alpha})$ . If  $p > 2$ , then eq. (6.17) and Lemma 6.5 (with  $q = p$  and  $r = \frac{p}{p-2}$ ) together imply that

$$|\mathbb{E}[\xi_0 f]| \leq 8\alpha_{n-n_\alpha}^{\frac{p}{p-2}} (\|A'_n\|_p + \|B_n\|_p) \|f\|_p + 2\mathbb{E}[A'_n] \|f\|_p,$$

where we have used the estimate  $\mathbb{E}[|f|] \leq \|f\|_p$  in the last term. Eq. (6.10) follows this inequality together with eqs. (6.14, 6.15, 6.16). If  $p = 2$ , then eq. (6.17) and Lemma 6.5 together imply that

$$|\mathbb{E}[\xi_0 f]| \leq \rho_{n-n_\alpha} (\|A'_n\|_2 + \|B_n\|_2) \|f\|_2 + 2\mathbb{E}[A'_n] \|f\|_2.$$

Eq. (6.11) follows from this inequality, again using eqs. (6.14, 6.15, 6.16). This completes the proof of Lemma 6.3.  $\blacksquare$

It remains to prove Proposition 6.4:

*Proof of Proposition 6.4.* From [22, Lemma 3.3], the quantity  $m(A, B)$  appearing in the definition (3.1) of the metric  $d(A, B)$  can be expressed as

$$m(A, B) = \min \left\{ \frac{\text{tr}[XA]}{\text{tr}[XB]} : X \in \mathbb{S}_D \text{ and } \text{tr}[XA] \neq 0 \right\}.$$

Since

$$\frac{\|\psi(\phi \cdot A)\|}{\|\psi(\phi \cdot B)\|} = \frac{\text{tr}\psi^*(\mathbb{I})\phi \cdot A}{\text{tr}\psi^*(\mathbb{I})\phi \cdot B} = \frac{\text{tr}\psi^*(\frac{1}{D}\mathbb{I})\phi \cdot A}{\text{tr}\psi^*(\frac{1}{D}\mathbb{I})\phi \cdot B},$$

we see that

$$m(\phi \cdot A, \phi \cdot B) \leq \frac{\|\psi(\phi \cdot A)\|}{\|\psi(\phi \cdot B)\|} \leq \frac{1}{m(\phi \cdot B, \phi \cdot A)}.$$

Since  $\phi \cdot A, \phi \cdot B$  are positive definite (because  $\phi$  is strictly positive), the various terms appearing in this inequality are all finite and non-zero. Taking logarithms yields

$$\begin{aligned} |\ln \|\psi(\phi \cdot A)\| - \ln \|\psi(\phi \cdot B)\|| &\leq -\ln m(\phi \cdot A, \phi \cdot B) - \ln m(\phi \cdot B, \phi \cdot A) \\ &\leq \ln \frac{1 + d(\phi \cdot A, \phi \cdot B)}{1 - d(\phi \cdot A, \phi \cdot B)} \leq \ln \frac{1 + c(\phi)}{1 - c(\phi)}, \end{aligned}$$

where we have used the definition eq. (3.1) of  $d(\cdot, \cdot)$  and Lemma 3.1 to obtain  $d(\phi \cdot A, \phi \cdot B) \leq c(\phi)$ . Now for  $x \in [0, 1)$  we have

$$\frac{1+x}{1-x} \leq \frac{1}{(1-x)^2}$$

As  $x = c(\phi) \in (0, 1)$  (since  $\phi$  is strictly positive) we have that

$$|\ln \|\psi(\phi \cdot A)\| - \ln \|\psi(\phi \cdot B)\|| \leq 2 \ln \frac{1}{1 - c(\phi)}.$$

Now consider the convex function  $f(x) = \ln 1/(1-x)$  for  $x \in [0, 1)$ . Since  $f$  is convex and  $f(0) = 0$ , we have  $f(tr) \leq tf(r)$  for any  $t, r \in [0, 1)$ . Hence,  $f(\lambda) \leq f(r)\lambda/r$  for any  $\lambda \in [0, r]$ . Thus

$$|\ln \|\psi(\phi \cdot A)\| - \ln \|\psi(\phi \cdot B)\|| \leq c(\phi) \frac{2}{r} \ln \frac{1}{1-r}$$

if  $c(\phi) \leq r$ . ■

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