# Non-stationary version of Furstenberg Theorem on random matrix products

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Thermodynamic Formalism: Non-additive Aspects and Related Topics

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The results are joint with Victor Kleptsyn (CNRS, Rennes University)

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### Powers of a matrix and their rate of growth

Suppose  $A \in SL(d, \mathbb{R})$ . What can be said about the rate of growth of  $||A^n||$  as  $n \to \infty$ ?

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$$\lim_{n\to\infty}\frac{1}{n}\log\|A^n\|=\lambda,$$

where  $\lambda$  is a logarithm of a maximal eigenvalue of A (i.e. logarithm of the spectral radius).

### Furstenberg-Kesten Theorem

Let  $\mu$  be a probability distribution on  $SL(d, \mathbb{R})$ . Choose matrices  $A_1, A_2, \ldots$  randomly and independently with respect to the distribution  $\mu$ . Denote  $A^n = A_n \ldots A_2 A_1$ . What can one say about the rate of growth of  $||A^n||$  as  $n \to \infty$ ?

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### Furstenberg-Kesten Theorem

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Theorem (Furstenberg-Kesten, 1960) Suppose that

$$\int \log \|A\| d\mu(A) < \infty.$$

Then almost surely there exists a limit

$$\lim_{n\to\infty}\frac{1}{n}\log\|A^n\|=\lambda,$$

where  $\lambda = \lambda(\mu)$  is a non-random constant (Lyapunov exponent).

# Example 1

Set 
$$A = \begin{pmatrix} \cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha \end{pmatrix}$$
,  $B = \begin{pmatrix} \cos \beta & \sin \beta \\ -\sin \beta & \cos \beta \end{pmatrix}$ . As  $A_i$  let us choose  $A$  or  $B$  (with probablity 1/2),  $A^n = A_n \dots A_2 A_1$ .

Then

$$\lim_{n\to\infty}\frac{1}{n}\log\|A^n\|=0.$$

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# Example 2

Set 
$$A = \begin{pmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$$
,  $B = \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & 2 \end{pmatrix}$  (i.e.  $B = A^{-1}$ ). As  $A_i$  let us choose  $A$  or  $B$  (with probability 1/2),  $A^n = A_n \dots A_2 A_1$ .

Then almost surely

$$\lim_{n\to\infty}\frac{1}{n}\log\|A^n\|=0.$$

# Example 3

Set 
$$A = \begin{pmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$$
,  $B = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ . As  $A_i$  let us choose  $A$  or  $B$  (with probability 1/2),  $A^n = A_n \dots A_2 A_1$ .

Then almost surely

$$\lim_{n\to\infty}\frac{1}{n}\log\|A^n\|=0.$$

Let  $\mu$  be a probability distribution on  $SL(d, \mathbb{R})$ . Choose matrices  $A_1, A_2, \ldots$  randomly and independently with respect to the distribution  $\mu$ . Denote  $A^n = A_n \ldots A_2 A_1$ . What can one say about the rate of growth of  $||A^n||$  as  $n \to \infty$ ?

Let  $\mu$  be a probability distribution on  $SL(d, \mathbb{R})$ . Choose matrices  $A_1, A_2, \ldots$  randomly and independently with respect to the distribution  $\mu$ . Denote  $A^n = A_n \ldots A_2 A_1$ . What can one say about the rate of growth of  $||A^n||$  as  $n \to \infty$ ?

Theorem (Furstenberg Theorem, 1963)

Suppose that

•  $\int \log \|A\| d\mu(A) < \infty;$ 

 $\bullet$  the smallest closed subgroup of  $SL(d,\mathbb{R})$  that contains  $\mathsf{supp}\,\mu$  is not compact, and

• there is no finite collection of proper subspaces on  $\mathbb{R}^d$  invariant under action of each of the linear maps from supp  $\mu$ .

Then almost surely the Lyapunov exponent

$$\lim_{n\to\infty}\frac{1}{n}\log\|A^n\|=\lambda$$

is strictly positive,  $\lambda > 0$ .

#### Remark 1:

- If d = 2, then the conditions
- $\bullet$  the smallest closed subgroup of  $SL(d,\mathbb{R})$  that contains  $\mathsf{supp}\,\mu$  is not compact, and
- there is no finite collection of proper subspaces on  $\mathbb{R}^d$  invariant under action of each of the linear maps from supp  $\mu$

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can be replaced by

• There exist no measure  $\nu$  on  $\mathbb{RP}^1$  invariant under all  $f_A$ ,  $A \in supp \mu$ .

Remark 2:

To see why the condition

$$\int \log \|A\| d\mu(A) < \infty$$

is relevant, consider the case d = 1. If  $\{a_n\}$ ,  $a_n \in \mathbb{R}$ ,  $a_n > 0$ , are iid random variable, then to study  $a_1 \cdot a_2 \cdot \ldots \cdot a_n$ , consider  $\log a_1 + \log a_2 + \ldots + \log a_n$ . We have

$$\frac{\log a_1 + \log a_2 + \ldots + \log a_n}{n} \to \mathbb{E} \log a_1,$$

if  $\mathbb{E} \log a_1 < \infty$ .

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Cosider discrete Schrödinger operator on  $l^2(\mathbb{Z})$ :

$$(Hu)(n) = u(n+1) + u(n-1) + V(n)u(n),$$

where  $\{V(n)\}$  are iid random variables.

Cosider discrete Schrödinger operator on  $l^2(\mathbb{Z})$ :

$$(Hu)(n) = u(n+1) + u(n-1) + V(n)u(n),$$

where  $\{V(n)\}$  are iid random variables.

If  $E \in \mathbb{R}$  is an eigenvalue, then Hu = Eu, that is,

$$u(n+1) + u(n-1) + V(n)u(n) = Eu(n)$$

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$$u(n+1)+u(n-1)+V(n)u(n)=Eu(n),$$

then

$$\begin{pmatrix} u(n+1)\\ u(n) \end{pmatrix} = \begin{pmatrix} E - V(n) & -1\\ 1 & 0 \end{pmatrix} \begin{pmatrix} u(n)\\ u(n-1) \end{pmatrix}.$$
  
Denote  $\Pi_{E,n} = \begin{pmatrix} E - V(n) & -1\\ 1 & 0 \end{pmatrix}.$ 

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Denote  $\Pi_{E,n} = \begin{pmatrix} E - V(n) & -1 \\ 1 & 0 \end{pmatrix}.$   
So if  $\{V(n)\}$  are iid random variables, by Furstenberg Theorem, with probability 1 for a fixed value of energy  $E$ 

$$\lim_{n\to\infty}\frac{1}{n}\log\|\Pi_{E,n}\Pi_{E,n-1}\dots\Pi_{E,1}\|=\lambda>0$$

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What if

$$V(n) = V_{background}(n) + V_{random}(n)$$
 ?

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**Question:** Suppose  $A_i \in SL(d, \mathbb{R})$  is distributed w.r.t.  $\mu_i$  (independent, but not identically distributed)? What one can say about the growth rate of  $||T_n||$ , where  $T_n = A_n \cdot \ldots \cdot A_1$ ?

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Theorem (G, Kleptsyn, 2022)

Assume K is a compact set in the space of probability distributions on  $SL(2,\mathbb{R})$  such that

• For some  $\gamma, C > 0$  and any  $\mu \in K$  we have  $\int \|A\|^{\gamma} < C$ ;

• ("measures condition") For any  $\mu \in K$ , there are no probability measures  $\nu_1, \nu_2$  on  $\mathbb{RP}^1$  such that  $(f_A)_*\nu_1 = \nu_2$  for all  $A \in \text{supp } \mu$ . Choose any sequence  $\{\mu_i\} \subset K^{\mathbb{N}}$ , and let  $A_i$  be chosen w.r.t.  $\mu_i$ , independently. Then

1) For some  $\lambda > 0$  we have  $\mathbb{E} \log ||T_n|| = L_n > \lambda n$  for all  $n \in \mathbb{N}$ ; 2) Almost surely,  $\lim_{n\to\infty} \frac{1}{n} (\log ||T_n|| - L_n) = 0$ .

Choose any sequence  $\{\mu_i\} \subset K^{\mathbb{N}}$ , and let  $A_i$  be chosen w.r.t.  $\mu_i$ , independently. Then

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- 1) For some  $\lambda > 0$  we have  $\mathbb{E} \log ||T_n|| = L_n > \lambda n$  for all  $n \in \mathbb{N}$ ;
- 2) Almost surely,  $\lim_{n\to\infty} \frac{1}{n} (\log ||T_n|| L_n) = 0.$

#### Addendum:

Almost surely, there exists a unit vector  $\bar{v} \in \mathbb{R}^2$  such that  $T_n \bar{v} \to 0$ , and  $\lim_{n \to \infty} \frac{1}{n} (\log |T_n \bar{v}| + L_n) = 0.$ 

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#### Non-stationary Anderson Model

Consider a discrete Schrödiger operator H on  $\ell^2(\mathbb{Z})$ 

$$[Hu](n) = u(n+1) + u(n-1) + V(n)u(n).$$

Suppose that  $\{V(n)\}$  are independent (but not necessarily identically distributed) random variables, distributed with respect non-degenerate probability measures  $\{\mu_n\}$  with compact support. In particular, non-stationary Anderson-Bernoulli Model (measure  $\mu_n$  is supported on two points, and depends on n) satisfies our conditions. Denote

$$\mathbb{P}=\prod_{n=-\infty}^{\infty}\mu_n$$

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#### Non-stationary Anderson Model

$$[Hu](n) = u(n+1) + u(n-1) + V(n)u(n).$$

**Theorem (Spectral Anderson Localization)** Suppose that a random potential  $\{V(n)\}$  of operator H is defined by independent probability distributions  $\{\nu_n\}$ , such that

1) supp 
$$\mu_n \subseteq [-K, K];$$

2)  $Var(\mu_n) > \varepsilon$ ,

where  $\varepsilon > 0, K < \infty$  are some uniform constant. Then  $\mathbb{P}$ -almost surely operator H has p.p. spectrum, with exponentially decaying eigenfunctions.

Theorem (the case  $d \ge 2$ ) Assume additionally that

• for any  $\mu \in K$ , there are no union of proper subspaces U, U' in  $\mathbb{R}^d$  such that A(U) = U' for all  $A \in \text{supp } \mu$ .

Then the same conclusion holds. That is, if we choose any sequence  $\{\mu_i\} \subset K^{\mathbb{N}}$ , and set  $A_i$  to be chosen w.r.t.  $\mu_i$ , independently, then

1) For some  $\lambda > 0$  we have  $\mathbb{E} \log ||T_n|| = L_n > \lambda n$  for all  $n \in \mathbb{N}$ ;

2) Almost surely,  $\lim_{n\to\infty} \frac{1}{n} (\log ||T_n|| - L_n) = 0.$ 

#### Remark:

The statement that  $\liminf_{n\to\infty} \frac{1}{n}L_n > 0$ , was proven by I.Goldsheid (independently and by different methods).

Theorem (Large Deviation Estimates) Under the assumptions above, for any  $\varepsilon$  there exists  $\delta > 0$  such that for all large enough  $n \in \mathbb{N}$  we have

$$\mathbb{P}\left\{\left|\log \|T_n\| - L_n\right| > \varepsilon n\right\} < e^{-\delta n}$$

Consider any closed Riemannian manifold M and the set of its  $C^1$ -diffeomorphisms  $\text{Diff}^1(M)$ . Set

$$\operatorname{Jac}(f)|_{x} = |\det df|_{x}| = \left. \frac{d\operatorname{Leb}_{M}}{df_{*}\operatorname{Leb}_{M}} \right|_{x}$$

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We will measure the maximum volume contraction rate of a diffeomorphism by the following quantity:

$$\mathcal{N}(f) := \max_{x \in M} \operatorname{Jac}(f)|_x^{-1} = \max_{x \in M} \left. \frac{df_* \operatorname{Leb}_M}{d \operatorname{Leb}_M} \right|_x$$

Let  $K_M$  be a compact subset of the space of probability measures on  $\text{Diff}^1(M)$  (equipped with the weak-\* convergence topology). Let M,  $K_M$  satisfy the following assumptions:

• (log-moment condition) For any  $\mu \in K_M$  one has

$$\int_{{\rm Diff}^1({\mathcal M})} \log {\mathcal N}(f) d\mu(f) < \infty$$

(measures condition) For any μ ∈ K<sub>M</sub> there are no Borel probability measures ν<sub>1</sub>, ν<sub>2</sub> on M such that f<sub>\*</sub>ν<sub>1</sub> = ν<sub>2</sub> for μ-almost every f ∈ Diff<sup>1</sup>(M).

Theorem

If M,  $K_M$  satisfy the assumptions above, then there exists h > 0 such that for any n and any  $\mu_1, \ldots, \mu_n \in K_M$  we have

 $\mathbb{E} \log \mathcal{N}(F_n) \geq nh$ ,

where  $F_n = f_n \circ \cdots \circ f_1$ , and every  $f_i$  is chosen independently with respect to the corresponding measure  $\mu_i$ , so that the expectation is taken over the distribution  $\mu_1 \times \mu_2 \times \ldots \times \mu_n$ .

#### Theorem

If M,  $K_M$  satisfy the assumptions above, then for any given sequence  $(\mu_i)_{i \in \mathbb{N}}$ ,  $\mu_i \in K_M$  of measures on  $\operatorname{Diff}^1(M)$  we have

$$\liminf_{n\to\infty}\frac{1}{n}\mathbb{E}\,\log\mathcal{N}(F_n)\geq h>0,$$

where  $F_n = f_n \circ \cdots \circ f_1$ , and the expectation is taken with respect to the infinite product measure  $\prod_i \mu_i$ .

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#### Remark:

If 
$$A \in SL(d, \mathbb{R})$$
, then  $\mathcal{N}(f_A) = \|A\|^d$ .

### Probability of cancellations

For a given (large but fixed) k we decompose the product of matrices of length n = km,

$$T_n = A_n \dots A_1,$$

into m groups of products of length k:

$$T_n = (A_n \ldots A_{k(m-1)+1}) \ldots (A_k \ldots A_1) = B_m \ldots B_1,$$

where

$$B_j:=(A_{kj}\ldots A_{k(j-1)+1}).$$

One has

$$\log ||T_n|| = \log ||B_m \dots B_1|| \le \sum_{j=1}^m \log ||B_j||.$$

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### **Probability of cancellations**

$$T_n = (A_n \dots A_{k(m-1)+1}) \dots (A_k \dots A_1) = B_m \dots B_1,$$
  
 $B_j = (A_{kj} \dots A_{k(j-1)+1}).$ 



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### Stationary case and stationary measures

In the stationary case:

#### Definition:

Let  $\mu$  be a probability distribution on  $SL(d, \mathbb{R}^d)$ . A probability measure  $\nu$  on  $\mathbb{RP}^{d-1}$  is stationary if

$$\mu * \nu = \mathbb{E}_{\mu}((f_A)_*\nu) = \int (f_A)_*\nu d\mu(A) = \nu$$

**Claim:** Under natural conditions on  $\mu$  (that hold generically), a stationary measure is unique and has no atoms.

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Let X be a metric compact. For a measure  $\mu$  on the space of homeomorphisms  $\operatorname{Homeo}(X)$ , we say that there is

- no finite set with a deterministic image, if there are no two finite sets F, F' ⊂ X such that f(F) = F' for μ-a.e. f ∈ Homeo(X);
- no measure with a deterministic image, if there are no two probability measures ν, ν' on X such that f<sub>\*</sub>ν = ν' for μ-a.e. f ∈ Homeo(X).

Let X be a metric compact. For a measure  $\mu$  on the space of homeomorphisms  $\operatorname{Homeo}(X)$ , we say that there is

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- no measure with a deterministic image, if there are no two probability measures ν, ν' on X such that f<sub>\*</sub>ν = ν' for μ-a.e. f ∈ Homeo(X).

Denote by  $\mathfrak{Max}(\nu)$  the weight of a maximal atom of a probability measure  $\nu$ . In particular, if  $\nu$  has no atoms, then  $\mathfrak{Max}(\nu) = 0$ .

Theorem (Atoms Dissolving)

Let  $K_X$  be a compact set of probability measures on Homeo(X).

Assume that for any μ ∈ K<sub>X</sub> there is no finite set with a deterministic image. Then for any ε > 0 there exists n such that for any probability measure ν on X and any sequence μ<sub>1</sub>,..., μ<sub>n</sub> ∈ K<sub>X</sub> we have

$$\mathfrak{Max}\left(\mu_{n}*\cdots*\mu_{1}*\nu\right)<\varepsilon.$$

In particular, for any probability measure  $\nu$  on X and any sequence  $\mu_1, \mu_2, \ldots \in K_X$  we have

$$\lim_{n\to\infty}\mathfrak{Max}(\mu_n*\cdots*\mu_1*\nu)=0.$$

Theorem (Atoms Dissolving)

Let  $K_X$  be a compact set of probability measures on Homeo(X).

If for any μ ∈ K<sub>X</sub> there is no measure with a deterministic image, then the convergence is exponential and uniform over all sequences μ<sub>1</sub>, μ<sub>2</sub>,... from K<sup>N</sup> and all probability measures ν. That is, there exists λ < 1 such that for any n, any ν and any μ<sub>1</sub>, μ<sub>2</sub>,... ∈ K<sub>X</sub>

$$\mathfrak{Max}(\mu_n*\cdots*\mu_1*\nu)<\lambda^n.$$

Let X be a compact metric space, and  $\mu$  be a probability measure on the space of continuous maps C(X, X). The iterations of the corresponding random dynamical system are the sequences of compositions

$$f_1, f_2 \circ f_1, \ldots, f_n \circ \cdots \circ f_1, \ldots,$$

where  $f_i : X \to X$  are chosen randomly and independently w.r.t. the measure  $\mu$ .

The following theorem corresponds to the Birkhoff Ergodic Theorem for the stationary random dynamics:

Theorem (Random Birkhoff Ergodic Theorem) For any ergodic stationary measure  $\nu$  on X, for any  $\varphi \in C(X, \mathbb{R})$ , for  $\nu$ -a.e.  $x \in X$ ,  $\mu^{\mathbb{N}}$ -almost surely one has

$$\frac{1}{n}\sum_{k=0}^{n-1}\varphi(f_k\circ\ldots\circ f_1(x))\to \int_X\varphi(x)d\nu(x),$$

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where  $f_1, f_2, \ldots$  are chosen randomly and independently, with respect to the distribution  $\mu$ .

Assume that a "test function"  $\varphi \in C(X)$  is given. Also, for every n we assume that a measure  $\mu_n$  on C(X, X) is given (if the dynamics is assumed to be invertible, one can ask instead for the measure on the set of homeomorphisms of X). Denote

$$\mathbb{P} := \prod_{n=1}^{\infty} \mu_n, \tag{1}$$

and our goal is to describe time averages along  $\mathbb P\text{-almost}$  every sequence of iterations.

Let us denote by  $\mathcal{M} = \mathcal{M}(X)$  the space of Borel probability measures on the compact metric space X. We consider it to be equipped with the Wasserstein distance (that is one of the ways to metrize the weak-\* topology in the space of probability measures).

**Standing Assumption:** We will say that a sequence of distributions  $\mu_1, \mu_2, \mu_3, \ldots$  on C(X, X) satisfies the Standing Assumption if for any  $\delta > 0$  there exists  $m \in \mathbb{N}$  such that the images of any two initial measures after averaging over m random steps after any initial moment n become  $\delta$ -close to each other:

$$\forall \nu, \nu' \in \mathcal{M}, \quad \forall n \in \mathbb{N}, \\ \operatorname{dist}_{\mathcal{M}}(\mu_{n+m} * \dots * \mu_{n+1} * \nu, \mu_{n+m} * \dots * \mu_{n+1} * \nu') < \delta.$$

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Theorem (G, Kleptsyn, 2023)

Suppose the sequence of distributions  $\mu_1, \mu_2, \mu_3, \ldots$  satisfies the Standing Assumption above. Given any Borel probability measure  $\nu_0$  on X, define

$$\nu_n := \mu_n * \nu_{n-1}, \quad n = 1, 2, \dots$$

Then for any  $\varphi \in C(X, \mathbb{R})$  and any  $x \in X$ , almost surely

$$\frac{1}{n}\left|\sum_{k=1}^{n}\varphi(f_k\circ\ldots\circ f_1(x))-\sum_{k=1}^{n}\int_X\varphi\,d\nu_k\right|\to 0\quad \text{as}\quad n\to\infty,$$

where  $\nu_n := \mu_n * \nu_{n-1}$ , n = 1, 2, ..., and  $\nu_0$  is arbitrary.

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Moreover, an analogue of the Large Deviations Theorem holds. Namely, for any  $\varepsilon > 0$  there exist  $C, \delta > 0$  such that for any  $x \in X$ 

$$\forall n \in \mathbb{N},$$

$$\mathsf{P}\left(\frac{1}{n}\left|\sum_{k=1}^{n}\varphi(f_k \circ \ldots \circ f_1(x)) - \sum_{k=1}^{n}\int_{X}\varphi \,d\nu_k\right| > \varepsilon\right) < < C \exp(-\delta n).$$

# **Random Iterated Function Systems**

#### Proposition:

Let X be a compact metric space, and  $\lambda \in (0, 1)$  be a constant. Suppose  $\{\mu_i\}_{i \in \mathbb{N}}$  be a sequence of probability distributions in the space of contractions  $X \to X$  with Lipschitz constant at most  $\lambda$ . Then Standing Assumption holds, i.e. for any Borel probability measures  $\nu, \nu'$  on X we have

$$\operatorname{dist}_{\mathcal{M}}(\mu_{n+m}*\ldots*\mu_{n+1}*\nu,\mu_{n+m}*\ldots*\mu_{n+1}*\nu')\to 0 \ \text{ as } \ m\to\infty,$$

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uniformly in  $(\nu, \nu')$  and  $n \in \mathbb{N}$ .

# **Random Matrix Products**

Suppose K is a compact set in the space of probability distributions on  $SL(2, \mathbb{R})$  such that for any  $\mu \in K$  the measures condition is satisfied, i.e. there are no probability measures  $\nu_1, \nu_2$  on  $\mathbb{RP}^1$  such that  $(f_A)_*(\nu_1) = \nu_2$  for all  $A \in \operatorname{supp} \mu$ . Slightly abusing the notation, we will treat  $\mu$  also as a measure on the space of projective maps  $f_A : \mathbb{RP}^1 \to \mathbb{RP}^1$ .

#### **Proposition:**

Suppose the measure condition holds for each  $\mu \in K$ . Then for any sequence  $\{\mu_i\} \in K^{\mathbb{N}}$  and any probability measures  $\nu, \nu' \in \mathcal{M}$  we have:

$$\operatorname{dist}_{\mathcal{M}}(\mu_{n+m}*\ldots*\mu_{n+1}*\nu,\mu_{n+m}*\ldots*\mu_{n+1}*\nu')\to 0 \text{ as } m\to\infty,$$

uniformly in  $(\nu, \nu')$  and  $n \in \mathbb{N}$ .

### **Central Limit Theorem**

Theorem (CLT, La Page, 1982) Suppose  $\{A_i\}$  are random iid  $SL(d, \mathbb{R})$  matrices, distributed w.r.t.  $\mu$ , which is strongly irreducible and contracting, and

$$\int \|A\|^{\gamma} d\mu < \infty$$

for some  $\gamma > 0$ . Then there exists a > 0 such that

$$\frac{\log \|T_n\| - n\lambda_F}{\sqrt{n}} \xrightarrow{dist} \mathcal{N}(0, a^2).$$

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# **Central Limit Theorem**

#### Conjecture:

Under suitable (non-stationary) assumptions, if  $\{A_i\}$  are random independent (but not neccessarily identically distributed)  $SL(d, \mathbb{R})$  matrices, distributed w.r.t.  $\mu_i$ ,  $Var(\log ||T_n||)$  growth linearly, and

$$\frac{\log \|T_n\| - L_n}{\sqrt{Var(\log \|T_n\|)}} \xrightarrow{dist} \mathcal{N}(0, 1).$$

Work in progress, joint with Victor Kleptsyn and Grigorii Monakov.

# Thank you!

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