

# Notes on Furstenberg Theorem

Ajorda Jiao

## Contents

1	Introduction	1
2	Cocycles and ergodic theorems	2
3	Main results and some examples	4
4	Stationary measures	5
5	Original proof of $d = 2$	7
6	Invariance Principle	9

## §1 Introduction

Recall strong law of large numbers.

### Theorem 1.1 (Strong Law of Large Numbers)

$X_0, X_1, \dots, X_n, \dots$  a sequence of i.i.d. random values,  $\mathbb{E}|X_0| < \infty$ , then

$$\frac{1}{n}(X_0 + X_1 + \dots + X_{n-1}) \rightarrow \mathbb{E} X_0 \text{ a.s. .}$$

**Remark 1.2** — It can be regarded as a corollary of Birkhoff's ergodic theorem.

### Corollary 1.3

$X_0 > 0, \mathbb{E} \log X_0 > 0$ , then

$$X_0 X_1 \cdots X_{n-1} \rightarrow \infty, \quad \text{exponentially fast a.s. .}$$

We want to generalize this result to some non-commutative case. Let  $\nu$  be a probability measure on  $\text{SL}(d, \mathbb{R})$ , let  $A_0, A_1, \dots, A_n, \dots$  be a sequence of i.i.d. random matrices with common distribution  $\nu$ . Let

$$A^n := A_{n-1} \cdots A_1 A_0,$$

we want to show that  $\|A^n\| \sim e^{\lambda n}$  under some assumptions. A natural integrable condition is

$$\int \log \|A_0\| d\nu < \infty.$$

**Definition 1.4.** We define the **extremal Lyapunov exponents** as

$$\lambda_+ := \lim_{n \rightarrow \infty} \frac{1}{n} \log \|A^n\|, \quad \lambda_- := \lim_{n \rightarrow \infty} \frac{1}{n} \log \|(A^n)^{-1}\|^{-1}.$$

They are called the upper and the lower Lyapunov exponent, respectively.

**Definition 1.5.** Let  $\tilde{A} := \lim_{n \rightarrow \infty} (A^{n*} A^n)^{\frac{1}{2n}}$ , assume the eigenvalues of  $\tilde{A}$  are

$$e^{\lambda_1} \geq e^{\lambda_2} \geq \dots \geq e^{\lambda_d}.$$

The set  $\{\lambda_1, \lambda_2, \dots, \lambda_d\}$  is called **Lyapunov spectrum**.

**Remark 1.6** — We use the Lyapunov exponents to measure the increasing speed. Our aim is to proof  $\lambda_+ > 0$  under some assumptions.

## §2 Cocycles and ergodic theorems

**Definition 2.1.** Let  $(X, \mathcal{F}, \mu)$  be a probability space and  $f : X \rightarrow X$  be a measure-preserving map. Let  $A : X \rightarrow \text{GL}(d, \mathbb{R})$  be a measurable function. The **linear cocycle** defined by  $A$  over  $f$  is the transformation:

$$F : X \times \mathbb{R}^d \rightarrow X \times \mathbb{R}^d, \quad (x, v) \mapsto (f(x), A(x)v).$$

**Definition 2.2.** Let  $F : X \times \mathbb{R}^d \rightarrow X \times \mathbb{R}^d$  be a linear cocycle. The **projective cocycle** associated with it is defined as

$$\mathbb{P}F : X \times \mathbb{RP}^{d-1} \rightarrow X \times \mathbb{RP}^{d-1}, \quad (x, [v]) \mapsto (f(x), [A(x)v]).$$

### Example 2.3

Take  $X = \text{SL}(d, \mathbb{R})^{\mathbb{N}}$  with probability measure  $\mu = \nu^{\mathbb{N}}$ . Let  $f : X \rightarrow X$  be the shift map. The measurable function  $A : X \rightarrow \text{GL}(d, \mathbb{R})$  is defined as  $x = (A_0, A_1, \dots) \mapsto A_0$ . Let  $A^n(x) = A_{n-1} \cdots A_1 A_0$ , consider the linear cocycle defined by  $A$  over  $f$ , denoted by  $F : X \times \mathbb{R}^d \rightarrow X \times \mathbb{R}^d$ . Then  $F^n(x, v) = (f^n x, A^n(x)v)$ .

For general linear cocycle  $F$ , assume  $F^n(x, v) = (f^n x, A^n(x)v)$ , we can defined the Lyapunov exponents of  $F$  by  $A^n(x)$ .

Then the Lyapunov exponents of a random matrices sequence is identified with the Lyapunov exponents of the linear cocycle  $F$  constructed in the example above.

The following two ergodic theorems guarantee the existence of Lyapunov exponents and Lyapunov spectrum. Moreover,  $(X, f)$  is ergodic with respect to  $\mu$ , hence the  $f$ -invariance of the Lyapunov exponents implies that they are constants almost everywhere.

**Theorem 2.4** (Kingman's Sub-additive Ergodic Theorem)

Let  $(X, \mu)$  be a probability space and  $f : X \rightarrow X$  be a measure preserving map. Let  $(g_n)_{n=1}^\infty$  be a sequence of measurable functions such that  $g_1^+ \in L^1(X, \mu)$ , satisfying the subadditivity condition

$$g_{n+m} \leq g_m + g_n \circ f^m \quad \text{for all } m, n \geq 1.$$

Then there exists an  $f$ -invariant function  $g : X \rightarrow \mathbb{R} \cup \{-\infty\}$ , such that

$$\frac{1}{n} g_n \rightarrow g \quad \mu\text{-a.s.}$$

Moreover,

$$\frac{1}{n} \int g_n d\mu \rightarrow \int g d\mu = \inf_{n \geq 1} \frac{1}{n} \int g_n d\mu.$$

**Theorem 2.5** (Oseledets' Multiplicative Ergodic Theorem)

Let  $F$  be a linear cocycle on  $(X, \mathcal{F}, \mu)$  defined by  $A : X \rightarrow \text{GL}(d, \mathbb{R})$  over  $f : X \rightarrow X$  satisfying the integrability condition  $\log^+ \|A(\cdot)\| \in L^1(X, \mu)$ . Then there exists a forward invariant set  $\tilde{X} \in \mathcal{F}$  with full measure such that for each  $x \in \tilde{X}$ , the following statements hold:

- (i)  $\bar{A}(x) := \lim_{n \rightarrow \infty} (A^n(x)^* A^n(x))^{\frac{1}{2n}}$  exists.
- (ii) Let  $e^{\lambda_{p(x)}(x)} < \dots < e^{\lambda_2(x)} < e^{\lambda_1(x)}$  be the different eigenvalues of  $\bar{A}(x)$  and let  $U_{p(x)}(x), \dots, U_2(x), U_1(x)$  be the corresponding eigenspaces with multiplicities  $d_i(x) := \dim U_i(x)$ . Then

$$p(x) = p(fx), \quad \lambda_i(x) = \lambda_i(fx), \quad d_i(x) = d_i(fx).$$

- (iii) Put  $V_{p(x)+1}(x) := \{0\}$ , and for  $i = 1, 2, \dots, p(x)$ ,  $V_i(x) = U_{p(x)} \oplus \dots \oplus U_i(x)$ , so that

$$V_{p(x)}(x) \subset \dots \subset V_i(x) \subset \dots \subset V_1(x) = \mathbb{R}^d$$

defined a filtration of  $\mathbb{R}^d$ . For each  $v \in V_i(x) \setminus V_{i+1}(x)$ , the Lyapunov exponent of  $v$  exists and coincides with  $\lambda_i(x)$ , i.e.,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log \|A^n(x)v\| = \lambda_i(x).$$

- (iv) For all  $i = 1, 2, \dots, p(x)$ ,  $A(x)V_i(x) \subset V_i(fx)$ .

Moreover, the maps  $x \mapsto p(x)$ ,  $x \mapsto \lambda_i(x)$ ,  $x \mapsto d_i(x)$ ,  $x \mapsto U_i(x)$ ,  $x \mapsto V_i(x)$  (the last two convergences are in the sense of  $X \rightarrow \cup_{k=1}^d G_k(d)$ , where  $G_k(d)$  is the Grassmannian manifold of  $k$ -dimensional subspaces of  $\mathbb{R}^d$ ) are measurable.

### §3 Main results and some examples

**Definition 3.1.** We call  $\nu$  is **irreducible**, if there is no proper subspace  $V \subseteq \mathbb{R}^d$ , such that  $A(V) \subseteq V$  for  $\nu$ -a.e.  $A$ .

**Definition 3.2.** We call  $\nu$  is **strongly irreducible**, if there is no proper subspace  $V \subseteq \mathbb{R}^d$ , such that  $A(V) \subseteq V$  for  $\nu$ -a.e.  $A$ .

**Definition 3.3.** We call  $\nu$  is **non-compact**, if the support of  $\nu$  is not contained in a compact subgroup of  $\mathrm{SL}(d, \mathbb{R})$ .

**Remark 3.4** —  $\nu$  is compact if and only if there exists  $P \in \mathrm{SL}(d, \mathbb{R})$  such that  $\nu(P^{-1}\mathrm{SO}(d, \mathbb{R})P) = 1$ .

#### Theorem 3.5 (Furstenberg)

Strongly irreducible + non-compact  $\implies \lambda_+ > 0 > \lambda_-$ .

Let  $T_\nu$  be the semigroup generated by  $\mathrm{supp} \nu$ .

**Definition 3.6.** We call  $\nu$  is **contracting**, if  $\exists \{B_n\} \subseteq T_\nu$  such that  $\|B_n\|^{-1} B_n \rightarrow B$  with  $\mathrm{rank} B = 1$ .

**Remark 3.7** — Contracting is stronger than non-compact, because non-compact just guarantee that  $\mathrm{rank} B \leq d - 1$ .

**Definition 3.8.** We call  $\nu$  is  **$p$ -strongly irreducible** or  **$p$ -contracting**, if the action of  $(\mathrm{SL}(d, \mathbb{R}), \nu)$  on  $\wedge^p(\mathbb{R}^d)$  is strongly irreducible or contracting, respectively.

#### Theorem 3.9 (Furstenberg)

Strongly irreducible + contracting  $\implies \lambda_1 > 0 > \lambda_2$ .

#### Theorem 3.10 (Furstenberg)

$p$ -strongly irreducible +  $p$ -contracting  $\implies$  Lyapunov spectrum is simple.

Another result is proved by Gol'dsheid and Margulis, which conditions are much easier to verify.

#### Theorem 3.11 (Gol'dsheid, Margulis)

Assume  $\nu$  is a probability measure on  $\mathrm{GL}(d, \mathbb{R})$  satisfies the integrability condition and  $T_\nu$  is Zariski dense in  $\mathrm{GL}(d, \mathbb{R})$ . Then the Lyapunov spectrum is simple.

### Some examples

We show some counter examples for  $d = 2$ .

#### Example 3.12

If  $\nu$  supports on  $P^{-1}\text{SO}(2, \mathbb{R})P$  for some  $P \in \text{SL}(2, \mathbb{R})$ , then  $\|A^n\| \leq \|P^{-1}\| \|P\|$  is bounded almost everywhere.

#### Example 3.13

If  $\nu$  admits an invariant direction:

Consider  $\nu$  supports on  $\left\{ \begin{bmatrix} 1 & s \\ 0 & 1 \end{bmatrix} : s \in \mathbb{R} \right\}$ . Assume  $A_0 = \begin{bmatrix} 1 & X \\ 0 & 1 \end{bmatrix}$  where  $X$  is a positive random value with  $0 < \mathbb{E} X < \infty$ . Then  $\|A^n\| \rightarrow \infty$  but just with linear speed.

#### Example 3.14

If two directions preserved by  $\nu$ -a.e.  $A$ :

Let  $M_1 = \begin{bmatrix} e & 0 \\ 0 & e^{-1} \end{bmatrix}$ ,  $M_2 = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ , let  $\nu = \frac{1}{2}\delta_{M_1} + \frac{1}{2}\delta_{M_2}$ . As we can let  $t_k$  be the  $k$ -th  $n$  with  $A_n = M_2$  and  $\xi_k = t_k - t_{k-1} - 1$ . Then

$$\log \|A^n\| \sim \frac{\xi_1 - \xi_2 + \cdots + \xi_{k-1} - \xi_k}{n} \rightarrow 0 \text{ a.e. .}$$

## §4 Stationary measures

For every element  $A \in \text{GL}(d, \mathbb{R})$  and probability measure  $\eta$  on  $\mathbb{RP}^{d-1}$ . The linear map  $A : \mathbb{RP}^{d-1} \rightarrow \mathbb{RP}^{d-1}$ ,  $[v] \mapsto [Av]$  induces a probability measure  $A_*\eta$  on  $\mathbb{RP}^{d-1}$ .

Let  $\nu$  be a probability measure on  $\text{GL}(d, \mathbb{R})$  (or  $\text{SL}(d, \mathbb{R})$ ) and  $\eta$  be a probability measure on  $\mathbb{RP}^{d-1}$ . Then we define the convolution  $\nu * \eta$  be the probability measure on  $\mathbb{RP}^{d-1}$  such that for any continuous function  $f$  on  $\mathbb{RP}^{d-1}$ ,

$$\int f([v]) d(\nu * \eta)([v]) = \iint f([Av]) d\nu(A) d\eta([v]).$$

Or we can write as

$$\nu * \eta = \int A_*\eta d\nu(A).$$

Now, we define the  $\nu$ -stationary measure on  $\mathbb{RP}^{d-1}$ . That is if we consider  $\text{GL}(n, d)$  acting on  $\mathbb{RP}^{d-1}$  with law  $\nu$ , it gives a random walk on the projective space. The stationary measure is the probability measure on  $\mathbb{RP}^{d-1}$  which is invariant under the random walk.

**Definition 4.1.** A probability measure  $\eta$  on  $\mathbb{RP}^{d-1}$  is called a  **$\nu$ -stationary measure** if  $\nu * \eta = \eta$ .

As an analogue to the existence of invariant probability measure of the continuous map on a compact metric space, the following proposition tells us the stationary measure always exist on the projective space.

**Proposition 4.2**

For any probability measure  $\nu$  on  $\mathrm{GL}(n, d)$ , there are some  $\nu$ -stationary measure on  $\mathbb{RP}^{d-1}$ .

*Proof.* Let  $\xi$  be an arbitrary probability measure on  $\mathbb{RP}^{d-1}$ , and let

$$\xi_n = \frac{1}{n} \sum_{k=0}^{n-1} \nu^k * \xi.$$

Then  $\xi_n$  is a sequence of probability measure on  $\mathbb{RP}^{d-1}$ . Because the space of probability measures on a compact metric space is compact with respect to the weak-\* topology. Let  $\eta$  be a limit point of  $(\xi_n)_{n=0}^\infty$ , then  $\eta$  is a  $\nu$ -stationary measure on  $\mathbb{RP}^{d-1}$ .  $\square$

**Properties of stationary measures**

Now we focus on the probability measure  $\nu$  on  $\mathrm{SL}(d, \mathbb{R})$  which satisfies two conditions of Furstenberg's theorem. Let  $F$  be the linear cocycle on  $\mathrm{SL}(d, \mathbb{R})^{\mathbb{N}} \times \mathbb{R}^d$  which we construct in the example before.

**Proposition 4.3**

Let  $\eta$  be a  $\nu$ -stationary measure on  $\mathbb{RP}^{d-1}$ . If the cocycle  $F$  is strongly irreducible then  $\eta(V) = 0$  for any proper projective subspace  $V$ .

*Proof.* Otherwise, let  $d_0$  be the smallest dimension such there are some subspaces  $V$  with positive measure. Let  $c$  be the largest measure of those subspaces. Let

$$\mathcal{M} = \{V \subseteq \mathbb{RP}^{d-1} : \dim V = d_0, \eta(V) = c\}.$$

Then for every  $V_1 \neq V_2 \in \mathcal{M}$ , we have  $\eta(V_1 \cap V_2) = 0$ , hence  $\mathcal{M}$  is finite. Because  $\eta$  is  $\nu$ -invariant, then for every  $V \in \mathcal{M}$ ,  $A^{-1}V \in \mathcal{M}$  for  $\nu$ -a.e.  $A$ . Hence  $\mathcal{M}$  is a finite collection of proper subspaces of  $\mathbb{RP}^{d-1}$  which is invariant under  $\nu$  almost every  $A \in \mathrm{SL}(d, \mathbb{R})$ .  $\square$

**Remark 4.4** — The probability measure on the projective space with this property is said to be **proper**.

For any proper probability measure  $\zeta$  on the projective space, we can define  $B_*\zeta$  for any  $B \neq 0 \in M(d, \mathbb{R})$ . Note that for  $B_n, B \in M(d, \mathbb{R}) \setminus \{0\}$ , if  $B_n \rightarrow B$ , then  $(B_n)_*\zeta \rightarrow B_*\zeta$  in the weak\* topology.

**Proposition 4.5**

Let  $\zeta$  be a probability measure on  $\mathbb{RP}^{d-1}$  such that the measure of each proper subspace is zero. Then the stabilizer  $H(\zeta) := \{A \in \mathrm{SL}(d, \mathbb{R}) : A_*\zeta = \zeta\}$  is a compact subgroup in  $\mathrm{SL}(d, \mathbb{R})$ .

*Proof.* The fact that  $H(\zeta)$  is a subgroup of  $\mathrm{SL}(d, \mathbb{R})$  follows from the definition directly. It is also closed in the weak-\* topology. It suffices to show that the norm  $\|B\|$  is bounded for  $B \in H(\zeta)$ . Otherwise, let  $(B_n)_{n=1}^\infty \subseteq H(\zeta)$  be a sequence of matrices such that  $\|B_n\| \rightarrow \infty$ . Let  $\tilde{B}_n = B_n / \|B_n\|$ , by passing to a subsequence, without loss of generality, suppose that  $\tilde{B}_n \rightarrow \tilde{B}$ . We have  $\tilde{B}_n_*\zeta = B_n_*\zeta = \zeta \rightarrow \zeta$ , but  $\tilde{B}$  does not have full rank which contradicts with the condition of  $\zeta$ .  $\square$

## §5 Original proof of $d = 2$

Now we focus on  $d = 2$  and show the original proof of Furstenberg. Let  $(X, \mu) = (\mathrm{SL}(2, \mathbb{R})^{\mathbb{N}}, \nu^{\mathbb{N}})$ ,  $F : X \times \mathbb{R}^2 \rightarrow X \times \mathbb{R}^2$  be the linear cocycle and  $\mathbb{P}F : X \times \mathbb{RP}^1 \rightarrow X \times \mathbb{RP}^1$  be the associated projective cocycle. Let  $\eta$  be a  $\nu$ -stationary measure on  $\mathbb{RP}^1$  which is proper by the previous discussion. Then  $\mu \times \eta$  is invariant under  $\mathbb{P}F$ .

By Oseledets' multiplicative ergodic theorem, for  $\mu$ -almost every  $x \in X$ , there exists an one dimensional subspace  $E(x) \in \mathbb{R}^2$  such that for all  $v \in \mathbb{R}^2 \setminus E(x)$ , it holds  $\frac{1}{n} \log \|A^n(x)v\| \rightarrow \lambda_+$ . Because  $\eta$  is proper, we conclude that for  $m = \mu \times \eta$  all  $(x, [v]) \in X \times \mathbb{RP}^1$ ,  $\frac{1}{n} \log \|A^n(x)v\| \rightarrow \lambda_+$ .

Now, we define  $\Phi : X \times \mathbb{RP}^1$  given by  $\Phi(x, [v]) = \log \frac{\|A(x)v\|}{\|v\|}$ . Then,

$$\frac{1}{n} \sum_{k=0}^n \Phi \circ F^k(x, [v]) = \frac{1}{n} \log \frac{\|A^n(x)v\|}{\|v\|} \rightarrow \lambda_+ \quad m\text{-a.e.}$$

Note that the left hand side also tends to Birkhoff average of  $\Phi$ . Hence we get an identity

$$\lambda_+ = \int \Phi dm = \iint \frac{\log \|A(x)v\|}{\|v\|} d\nu d\eta.$$

This identity tells us by applying the stationary measure, we can regard the Lyapunov exponent as a Birkhoff average. The following proposition shows that it suffices to prove the Birkhoff sum divergent almost everywhere.

### Proposition 5.1

Let  $T : (Y, m) \rightarrow (Y, m)$  be a measure preserving map of a probability space. If  $\varphi \in L^1(m)$  satisfying

$$\sum_{k=0}^{n-1} \varphi \circ T^k \rightarrow +\infty \quad m\text{-a.s.},$$

then  $\int \varphi dm > 0$ .

Now, we reduce the problem to show that  $\|A^n(x)v\| \rightarrow \infty (n \rightarrow \infty)$  for  $m$ -a.e.  $(x, [v])$ .

### Convergence of measures

For  $A \in \mathrm{SL}(2, \mathbb{R})$ , let  $A^*$  be the transpose of  $A$ . For  $\nu$  be a probability measure on  $\mathrm{SL}(2, \mathbb{R})$ , let  $\nu^*$  be the probability measure on  $\mathrm{SL}(2, \mathbb{R})$  push forward of  $\nu$  under the transpose. Let  $\zeta$  be a  $\nu^*$ -stationary measure on  $\mathbb{RP}^1$ , which is also proper.

### Lemma 5.2

For  $\mu$ -a.e.  $x \in X$ , there exists a probability measure  $\zeta_x$  on  $\mathbb{RP}^1$  such that

$$(A^n(x)^*)_* \zeta \rightarrow \zeta_x.$$

Moreover, for  $\nu^*$ -a.e.  $B \in \mathrm{SL}(2, \mathbb{R})$ ,  $(A^n(x)^* B)_* \zeta \rightarrow \zeta_x$  in the weak\* sense.

*Proof.* Fix  $f \in C(\mathbb{RP}^1)$ , consider the function  $P : \text{SL}(n, \mathbb{R}) \rightarrow \mathbb{R}$  given by

$$P(A) = \int_{\mathbb{RP}^1} f([Av]) d\zeta([v]).$$

Suppose  $\mathcal{F}$  is the  $\sigma$ -algebra associated with the probability space  $(X, \mu)$ . Let  $\mathcal{F}_n$  be the sub  $\sigma$ -algebra of  $\mathcal{F}$  generated by the cylinders of length  $n$ . Then we have

$$\begin{aligned} \mathbb{E}(P(A^{n+1}(x)^*) | \mathcal{F}_n) &= \int P(A^n(x)^* B^*) d\nu(B) \\ &= \iint f([A^n(x)^* Bv]) d\nu^*(B) d\zeta([v]) \\ &= \int f([A^n(x)v]) d(\nu^* * \zeta)([v]) \\ &= \int f([A^n(x)v]) d\nu^*([v]) = P(A^n(x)^*). \end{aligned}$$

Besides,  $\|P(A^n(x)^*)\|_2 \leq \|f\|_\infty < \infty$ , hence  $(P(A^n(\cdot)^*))_{n=1}^\infty$  is a  $L^2$  bounded martingale. By the martingale convergence theorem, there is an  $Lf \in L^2$  such that  $P(A^n(x)^*) \rightarrow Lf(x)$   $\mu$ -a.e. and in  $L^2$ . Then

$$\mathbb{E}(|P(A^{n+1}(x)^*) - P(A^n(x)^*)|^2) \rightarrow 0.$$

Where we have

$$\mathbb{E} \left( \iint |f([A^n(x)^* Bv]) - f([A^n(x)^* v])|^2 d\nu^*(B) d\zeta([v]) \right) \rightarrow 0,$$

this shows that for  $\nu^*$ -a.e.  $B \in \text{SL}(2, \mathbb{R})$ ,  $P(A^n(x)^* B) \rightarrow Lf(x)$   $\mu$ -a.e..

Take a countable dense set of  $f$  in  $C(\mathbb{RP}^1)$ , then there is a  $\mu$ -full measure set of  $x$  such that  $Lf(x)$  exists for all  $f$ . Then the functional  $f \mapsto Lf(x)$  gives a probability measure  $\zeta_x$  on  $\mathbb{RP}^1$ . These  $\zeta_x$  satisfy the condition.  $\square$

### Lemma 5.3

The limit measure  $\zeta_x$  is a Dirac measure.

*Proof.* Fix a generic point  $x$ , we know that  $A^n(x)^* \zeta \rightarrow \zeta_x$  and  $A^n(x)^* B\nu \rightarrow \nu_x$  for  $\nu^*$ -a.e.  $B$ . Choose a limit point of  $\|A^n(x)^*\|^{-1} A^n(x)^*$ , denoted by  $A$ . Then  $A_* \zeta = \zeta_x = (AB)_* \zeta$  for  $\nu^*$ -a.e.  $B \in \text{SL}(2, \mathbb{R})$ . If  $A$  is invertible, by proposition 4.5,  $\nu^*$  must supports on a compact subgroup of  $\text{SL}(2, \mathbb{R})$ , contradiction. Then  $A$  must be non-invertible, which shows that  $\zeta_x$  is a Dirac measure.  $\square$

**Remark 5.4** — Denote this Dirac measure by  $\delta_z = \delta_{z(x)}$ , the proof of lemma shows that the  $z(x)$  is independent of the choice of stationary measure. Moreover, the distribution of  $z(x)$  on  $\mathbb{RP}^1$  is same as  $\zeta$ , hence we can prove the uniqueness of the stationary measure.

*Proof of Furstenberg Theorem of  $d = 2$ .*

Firstly, for  $\mu$ -a.e.  $x \in X$ , we have  $(A^n(x)^*)_* \zeta \rightarrow \delta_z$ . Given generic  $x \in X$ , there must  $\|A^n(x)\| \rightarrow \infty$  otherwise  $A^n(x)$  have a limit point in  $\text{SL}(2, \mathbb{R})$  and the limit measure can't be Dirac.



Then we consider a limit point of  $\|A^n(x)\|^{-1}A^n(x)^*$ , denote by  $A(x)$ . Note that  $\text{rank } A(x) = 1$  and  $\text{Range } A(x) = z(x) \cdot \mathbb{R}$  where  $\|z(x)\| = 1$ . As  $n \rightarrow \infty$ , we have

$$\frac{\|A_n v\|}{\|A_n\|} = \sup_{\|u\|=1} \left\langle \frac{A_n v}{\|A_n\|}, u \right\rangle = \sup_{\|u\|=1} \left\langle v, \frac{(A^n)^* u}{\|A_n\|} \right\rangle \rightarrow \sup_{\|u\|=1} \langle v, Au \rangle = |\langle v, z \rangle|.$$

In particular,  $\|A_n(x)v\| \rightarrow \infty$  otherwise  $v \perp z(x)$ . Let  $\eta$  be the stationary measure of  $\nu$  which is proper, the former discussion shows that  $\|A_n(x)v\| \rightarrow \infty$  for  $m = \mu \times \eta$  almost every  $(x, [v])$ . The theorem follows.  $\square$

## §6 Invariance Principle

**Definition 6.1.** Let  $m$  be a probability measure on the product space  $X \times Y$  that projects to the probability measure  $\mu$  on  $X$ . A **disintegration** of  $m$  along vertical fibers is a measurable family  $\{m_x : x \in X\}$  of probability measures on  $Y$  satisfying

$$m(E) = \int_X m_x\{v : (x, v) \in E\} d\mu \quad \text{for any measurable } E \subseteq X \times Y.$$

The measures  $m_x$  on each fiber are called the **conditional probabilities** of  $m$ .

**Fact 6.2.** A disintegration along a vertical fiber does exist. Moreover, the disintegration is unique up to a full  $\mu$ -measure set.

Assume that  $X$  is a separable complete metric space, let  $\mu$  be an invariant probability measure on  $X$  with respect to  $f : X \rightarrow X$ . Let  $F : X \times \mathbb{R}^d \rightarrow X \times \mathbb{R}^d$  be a linear cocycle defined over  $f$  with extremal Lyapunov exponents  $\lambda_+(x), \lambda_-(x)$ . The following theorem of Ledrappier shows that if the upper Lyapunov exponent and the lower Lyapunov exponent coincide almost everywhere, then any  $\mathbb{P}F$ -invariant measure must have some invariance property on the conditional probabilities on the fiber.

### Theorem 6.3 (Ledrappier)

Assume that  $\lambda_-(x) = \lambda_+(x)$  for  $\mu$ -almost every  $x \in X$ . Then

$$m_{f(x)} = A(x)_* m_x \quad \text{for } \mu\text{-almost every } x \in X,$$

for any disintegration  $\{m_x : x \in X\}$  of any  $\mathbb{P}F$ -invariant probability measure  $m$  on  $X \times \mathbb{R}^{d-1}$  that projects down to  $\mu$ .

### Proof of Ledrappier's theorem

Let  $m$  be a probability measure on  $X \times \mathbb{R}^{d-1}$  invariant under the projective cocycle  $\mathbb{P}F$  which projects to  $\mu$  and let  $\{m_x : x \in X\}$  be a disintegration of  $m$ . For each  $x \in X$ , let

$$A(x)_*^{-1} m_{f(x)} = \zeta_x + \xi_x,$$

where  $\zeta_x \ll m_x$  and  $\xi_x \perp m_x$ . Let  $J(x, \cdot)$  be the Radon-Nikodym derivate of  $\zeta_x$  with respect to  $m_x$ , then we have

$$dA(x)_*^{-1} m_{f(x)} = J(x, \cdot) dm_x + d\xi_x.$$

**Remark 6.4** — The Radon-Nikodym derivate  $J$  reflects the contraction of  $A(x)$  on the projective space with respect to the conditional probability on each fiber.

**Definition 6.5.** The **fibred entropy** of  $m$  is defined by

$$h(m) = - \int \log J dm.$$

**Proposition 6.6**

The fibred entropy  $h(m)$  is always non-negative. If  $h(m) = 0$  then  $A(x)_*m_x = m_{f(x)}$  holds for  $\mu$ -a.e.  $x$ .

*Proof.* By Jensen's inequality, we have

$$h(m) = \int_{\{J>0\}} -\log J dm + \infty m\{J=0\} \geq -\log \int_{\{J>0\}} J dm + \infty m\{J=0\} \geq 0.$$

When the equality holds, there will be:  $m\{J=0\} = 0$ ,  $\log J$  is a constant  $m$ -almost everywhere and  $\int J dm = 1$ . The last equality shows that  $\xi_x = 0$  for  $\mu$ -almost  $x \in X$ . Hence  $J \equiv 1$  holds  $m$ -almost everywhere. Combining those discussions, we proofs the claim.  $\square$

Besides, we have another estimate of the fibred entropy. The difference of the extremal Lyapunov exponents reflects the contraction on  $\mathbb{RP}^{d-1}$  with respect to the projective metric. And the fibred entropy measures the contraction on  $\mathbb{RP}^{d-1}$  with respect to the conditional probability. As an analogues of the Ruelle inequality in the smooth ergodic theory, it is not surprising that the fibred entropy is bounded by the differences. The following proposition shows this relationship between the fibred entropy and the extremal Lyapunov exponents.

Assume, in addition,  $m$  is ergodic with respect to  $(X \times \mathbb{RP}^{d-1}, \mathbb{P}F)$ .

**Proposition 6.7**

$$0 \leq h(m) \leq d(\lambda_+ - \lambda_-).$$

**Remark 6.8** — The constant  $d$  can be replaced by  $d - 1$  but doesn't matter.

Note that Ledrappier's theorem follows from proposition 6.6 and proposition 6.7 immediately. It suffices to show proposition 6.7.

Consider any  $\Delta > \lambda_+ - \lambda_-$ , for any  $\varepsilon > 0$ , let  $J_\varepsilon = J + \varepsilon$  and  $h_\varepsilon(m) = - \int J_\varepsilon$ . Suppose, for contradiction,  $h_\varepsilon(m) > d\Delta + 4\varepsilon$  for some  $\Delta$ , where  $\varepsilon$  is small enough.

**Lemma 6.9**

Each fiber  $\{x\} \times \mathbb{RP}^{d-1}$  admits partitions  $\mathcal{P}_n(x)$  defined for  $n$  large enough, such that

- (i)  $\#\mathcal{P}_n(x) \leq e^{n(d\Delta+2\varepsilon)}$ ,
- (ii)  $\text{diam}\mathcal{P}_n(x) \leq e^{-n(\Delta+2\varepsilon)}$ ,
- (iii)  $m_x(\partial\mathcal{P}_n(x, v)) = 0$  for all  $v \in \mathbb{RP}^{d-1}$ , where  $\mathcal{P}_n(x, v)$  denote the atom of  $\mathcal{P}_n(x)$  that contains the point  $v$ .

For each  $0 \leq k \leq n$ , let  $\mathcal{P}_{n,k}(x)$  be a partition of  $\{x\} \times \mathbb{RP}^{d-1}$  given by the pull back of  $\mathcal{P}_n(f^k(x))$  under  $A^k(x)$ . That is  $\mathcal{P}_{n,k}(x, v) = A^{-k}(x)\mathcal{P}_n(f^k(x, v))$  for each  $(x, v) \in X \times \mathbb{RP}^{d-1}$ . Consider the function

$$J_{n,k,\varepsilon}(x, v) = J_{n,k}(x, v) + \varepsilon = \frac{m_{f(x)}(\mathcal{P}_{n,k}(F(x, v)))}{m_x(\mathcal{P}_{n,k+1}(x, v))} + \varepsilon.$$

**Lemma 6.10**

$$\sup_{0 \leq k \leq n} \|\log J_{n,k,\varepsilon} - \log J_\varepsilon\|_{L^1(m)} \rightarrow 0, \quad n \rightarrow \infty.$$

*Proof of Proposition 6.7.* Let  $J_{n,\varepsilon}(x, v) = \prod_{k=0}^{n-1} J_{n,n-1-k,\varepsilon} \circ F^k(x, v)$ , and let

$$J_n(x, v) = \prod_{k=0}^{n-1} J_{n,n-1-k} \circ F^k(x, v) = \frac{m_{f^n(x)}(\mathcal{P}_n(F^n(x, v)))}{m_x(\mathcal{P}_{n,n}(x, v))} \leq J_{n,\varepsilon}(x, v).$$

We have

$$\frac{1}{n} \log J_{n,\varepsilon} = \frac{1}{n} \sum_{k=0}^{n-1} \log J_{n,n-1-k,\varepsilon} \circ F^k(x, v).$$

Because we assume that  $m$  is ergodic,

$$\frac{1}{n} \sum_{k=0}^{n-1} \log J_\varepsilon \circ F^k(x, v) \rightarrow \int \log J_\varepsilon dm = -h_\varepsilon(m) \quad \text{in } L^1(m).$$

The previous lemma shows when  $n$  tends to infinity,  $\frac{1}{n} \sum_{k=0}^{n-1} \log J_{n,n-1-k,\varepsilon} \circ F^k(x, v)$  is

closed to  $\frac{1}{n} \sum_{k=0}^{n-1} \log J_\varepsilon \circ F^k(x, v)$  in  $L^1(m)$ , hence  $\frac{1}{n} \log J_{n,\varepsilon} \rightarrow -h_\varepsilon(m)$  in  $L^1(m)$ . By passing to a subsequence, we can get a sequence  $n_j \rightarrow \infty$  such that

$$\frac{1}{n_j} \log J_{n_j,\varepsilon}(x, v) \rightarrow -h_\varepsilon(m) \quad \text{for } m\text{-a.e. } (x, v) \in X \times \mathbb{RP}^{d-1}.$$

Then,

$$\limsup_j m_{f^{n_j}(x)}(\mathcal{P}_{n_j}(F^{n_j}(x, v))) \leq -h_\varepsilon(m).$$

For each large  $j$ , there is  $E_j \subseteq X \times \mathbb{RP}^{d-1}$ , such that  $m(E_j) > \frac{1}{2}$  and

$$m_{f^{n_j}(x)}(\mathcal{P}_{n_j}(F^{n_j}(x, v))) \leq e^{-n_j(h_\varepsilon(m)-\varepsilon)} \quad \text{for all } (x, v) \in E_j.$$

Hence  $m_{f^{n_j}(x)}(F^{n_j}(E_j) \cap (f^{n_j}(x) \times \mathbb{RP}^{d-1})) \leq e^{-n_j(h_\varepsilon(m)-\varepsilon)} \cdot e^{n_j(d\Delta+2\varepsilon)} \leq e^{-n_j\varepsilon}$  by the assumption  $h_\varepsilon(m) > d\Delta + 4\varepsilon$ . This follows that  $m(F^{n_j}(E_j)) \leq e^{-n_j\varepsilon} \rightarrow 0$ , as  $j \rightarrow \infty$ . Which contradicts with  $m(E_j) > \frac{1}{2}$  and  $F$  preserves the measure  $m$ .  $\square$