Notes on Furstenberg Theorem

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§1 Introduction

Recall strong law of large numbers.

Theorem 1.1 (Strong Law of Large Numbers)

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

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

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

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

We want to generalized this result to some non-commutative case. Let ν be a probability measure on $\mathrm{SL}(d,\mathbb{R})$, let $A_0,A_1,\cdots,A_n,\cdots$ be a sequence of i.i.d. random matrices with commen distribution ν . Let

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

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

$$\int \log \|A_0\| \,\mathrm{d}\nu < \infty.$$

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

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

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

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

$$e^{\lambda_1} \geqslant e^{\lambda_2} \geqslant \cdots \geqslant e^{\lambda_d}$$
.

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}, μ) be a probability space and $f: X \to X$ be a measure-preserving map. Let $A: X \to \mathrm{GL}(d, \mathbb{R})$ be a measurable function. The **linear cocycle** defined by A over f is the transformation:

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

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

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

Example 2.3

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

$$g_{n+m} \leqslant g_m + g_n \circ f^m$$
 for all $m, n \geqslant 1$.

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

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

Moreover,

$$\frac{1}{n} \int g_n d\mu \to \int g d\mu = \inf_{n \geqslant 1} \frac{1}{n} \int g_n d\mu.$$

Theorem 2.5 (Oseledets' Multiplicative Ergodic Theorem)

Let F be a linear cocycle on (X, \mathcal{F}, μ) defined by $A: X \to \mathrm{GL}(d, \mathbb{R})$ over $f: X \to X$ 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) $\bar{A}(x) := \lim_{n \to \infty} (A^n(x)^* A^n(x))^{\frac{1}{2n}}$ exists.
- (ii) Let $e^{\lambda_{p(x)}(x)} < \cdots < e^{\lambda_2(x)} < e^{\lambda_1(x)}$ be the different eigenvalues of $\bar{A}(x)$ and let $U_{p(x)}(x), \cdots, U_2(x), U_1(x)$ be the corresponding eigenspaces with multiplicities $d_i(x) \coloneqq \dim U_i(x)$. 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 $i = 1, 2, \dots, p(x), V_i(x) = U_{p(x)} \oplus \dots U_i(x)$, so

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

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.,

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

(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 $X \to \bigcup_{k=1}^d G_k(d)$, 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 ν is **irreducible**, if there is no proper subspace $V \subseteq \mathbb{R}^d$, such that $A(V) \subseteq V$ for ν -a.e. A.

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

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

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

Theorem 3.5 (Furstenberg)

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

Let T_{ν} be the semigroup generated by supp ν .

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

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

Definition 3.8. We call ν is p-strongly irreducible or p-contracting, if the action of $(SL(d, \mathbb{R}), \nu)$ 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 ν is a probability measure on $GL(d, \mathbb{R})$ satisfies the integrability condition and T_{ν} is Zariski dense in $GL(d, \mathbb{R})$. Then the Lyapunov spectrum is simple.

Some examples

We show some counter examples for d=2.

Example 3.12

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

Example 3.13

If ν admits an invariant direction:

Consider ν 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 $||A^n|| \to \infty$ but just with linear speed.

Example 3.14

If two directions preserved by ν -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

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

§4 Stationary measures

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

Let ν be a probability measure on $\mathrm{GL}(d,\mathbb{R})$ (or $\mathrm{SL}(d,\mathbb{R})$) and η 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

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

Now, we define the ν -stationary measure on \mathbb{RP}^{d-1} . That is if we consider $\mathrm{GL}(n,d)$ acting on \mathbb{RP}^{d-1} with law ν , 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

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

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 ν on GL(n,d), there are some ν -stationary measure on $\mathbb{R}\mathbb{P}^{d-1}$

Proof. Let ξ 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 ξ_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 η be a limit point of $(\xi_n)_{n=0}^{\infty}$, then η is a ν -stationary measure on \mathbb{RP}^{d-1} .

Properties of stationary measures

Now we focus on the probability measure ν 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 η be a ν -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 η is ν -invariant, then for every $V \in \mathcal{M}$, $A^{-1}V \in \mathcal{M}$ for ν -a.e. A. Hence \mathcal{M} is a finite collection of proper subspaces of \mathbb{RP}^{d-1} which is invariant under ν 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 ζ 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 $B_n \to B$, then $(B_n)_*\zeta \to B_*\zeta$ in the weak* topology.

Proposition 4.5

Let ζ 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 $||B_n|| \to \infty$. Let $\widetilde{B}_n = B_n/||B_n||$, by passing to a subsequence, without lost of generality, suppose that $\widetilde{B}_n \to \widetilde{B}$. We have $\widetilde{B}_{n*}\zeta = B_{n*}\zeta = \zeta \to \zeta$, but \widetilde{B} does not have full rank which contradicts with the condition of ζ .

§5 Original proof of d=2

Now we focus on d=2 and show the original proof of Furstenberg. Let $(X,\mu)=(\mathrm{SL}(2,\mathbb{R})^{\mathbb{N}},\nu^{\mathbb{N}}), F:X\times\mathbb{R}^2\to X\times\mathbb{R}^2$ be the linear cocycle and $\mathbb{P}F:X\times\mathbb{RP}^1\to X\times\mathbb{RP}^1$ be the associated projective cocycle. Let η be a ν -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 μ -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 $\frac{1}{n} \log \|A^n(x)v\| \to \lambda_+$. Because η is proper, we conclude that for $m = \mu \times \eta$ all $(x, [v]) \in X \times \mathbb{RP}^1, \frac{1}{n} \log \|A^n(x)v\| \to \lambda_+$.

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

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

Note that the left hand side also tends to Birkhoff average of Φ . 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 if suffices to prove the Birkhoff sum divergent almost everywhere.

Proposition 5.1

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

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

then $\int \varphi dm > 0$.

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

Convergence of measures

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

Lemma 5.2

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

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

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

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

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

Suppose \mathcal{F} is the σ -algebra associated with the probability space (X,μ) . Let \mathcal{F}_n be the

sub σ -algebra of \mathcal{F} generated by the cylinders of length n. Then we have

$$\mathbb{E}(P(A^{n+1}(x)^*)|\mathcal{F}_n) = \int P(A^n(x)^*B^*) d\nu(B)$$

$$= \iint f([A^n(x)^*Bv]) d\nu^*(B) d\zeta([v])$$

$$= \int f([A^n(x)v]) d(\nu^* * \zeta)([v])$$

$$= \int f([A^n(x)v]) d\nu^*([v]) = P(A^n(x)^*).$$

Besides, $||P(A^n(x)^*)||_2 \le ||f||_{\infty} < \infty$, 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 $P(A^n(x)^*) \to Lf(x)$ μ -a.e. and in L^2 . Then

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

Where we have

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

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

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

Lemma 5.3

The limit measure ζ_x is a Dirac measure.

Proof. Fix a generic point x, we know that $A^n(x)^*\zeta \to \zeta_x$ and $A^n(x)^*B\nu \to \nu_x$ for ν^* -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 ν^* -a.e. $B \in \mathrm{SL}(2,\mathbb{R})$. If A is invertible, by proposition 4.5, ν^* must supports on a compact subgroup of $\mathrm{SL}(2,\mathbb{R})$, contradiction. Then A must be non-invertible, which shows that ζ_x is a Dirac measure.

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 ζ , hence we can prove the uniqueness of the stationary measure.

Proof of Furstenberg Theorem of d = 2.

Firstly, for μ -a.e. $x \in X$, we have $(A^n(x)^*)_*\zeta \to \delta_z$. Given generic $x \in X$, there must $||A^n(x)|| \to \infty$ otherwise $A^n(x)$ have a limit point in $SL(2,\mathbb{R})$ 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 rank A(x) = 1 and Range $A(x) = z(x) \cdot \mathbb{R}$ where ||z(x)|| = 1. As $n \to \infty$, we have

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

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

§6 Invariance Principle

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

$$m(E) = \int_X m_x \{v : (x, v) \in E\} d\mu$$
 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 μ -measure set.

Assume that X is a separable complete metric space, let μ be an invariant probability measure on X with respect to $f: X \to X$. Let $F: X \times \mathbb{R}^d \to X \times \mathbb{R}^d$ 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 μ -almost every $x \in X$. Then

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

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

Proof of Ledrappier's theorem

Let m be a probability measure on $X \times \mathbb{RP}^{d-1}$ invariant under the projective cocycle $\mathbb{P}F$ which projects to μ 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 ζ_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 **fibered entropy** of m is defined by

$$h(m) = -\int \log J \mathrm{d}m.$$

Proposition 6.6

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

Proof. By Jensen's inequality, we have

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

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 μ -almost $x \in X$. Hence $J \equiv 1$ holds m-almost everywhere. Combining those discussions, we proofs the claim

Besides, we have another estimate of the fibered entropy. The difference of the extremal Lyapunov exponents reflects the contraction on \mathbb{RP}^{d-1} with respect to the projective metric. And the fibered 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 fibered entropy is bounded by the differences. The following proposition shows this relationship between the fibered entropy and the extremal Lyapunov exponents.

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

Proposition 6.7

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

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 Δ , where ε is small enough.

Lemma 6.9

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

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

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

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

Lemma 6.10

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

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

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

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,

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

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 $\frac{1}{n} \log J_{n,\varepsilon} \to -h_{\varepsilon}(m)$ in $L^1(m)$. By passing to a subsequence, we can get a sequence $n_j \to \infty$ such that

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

Then,

$$\lim \sup_{i} m_{f^{n_j}(x)}(\mathscr{P}_{n_j}(F^{n_j}(x,v))) \leqslant -h_{\varepsilon}(m).$$

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

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

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