# Ergodic and statistical properties of smooth systems

Adam Kanigowski

04.05.2021 Midwest Seminar based on joint work with (1) D. Dolgopyat, F. Rodriguez-Hertz (2) D. Dolgopyat, C. Dong, P. Nandori

# $T: (X, \kappa) \to (X, \kappa)$ – a (probability) measure preserving automorphism.

#### Definitions

- ergodic every measurable A for which T(A) = A satisfies  $\kappa(A) \in \{0, 1\}$ .
- 2 weakly mixing if  $\frac{1}{N} \sum_{n \le N} |\kappa(T^n A \cap B) \kappa(A)\kappa(B)| \to 0$ for every measurable A, B.
- **3** mixing  $-\kappa(T^NA \cap B) \to \kappa(A)\kappa(B)$  for every measurable A, B.
- A has positive entropy— if there is a finite partition with a linear growth of information.
- 5 K-system if every (non-trivial) factor of T has positive entropy.
- **6** Bernoulli if T is isomorphic to a Bernoulli shift.

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$$\Sigma = \{1, \dots, d\}^{\mathbb{Z}};$$
  
•  $\mathbf{p} = (p_1, \dots, p_d), \sum_{i=1}^d p_i = 1 - \text{probability vector}$   
•  $\sigma : (\Sigma, \mathbf{p}^{\mathbb{Z}}) \to (\Sigma, \mathbf{p}^{\mathbb{Z}}) - \text{Bernoulli shift},$ 

$$\sigma((x_i)_{i\in\mathbb{Z}})=(x_{i+1})_{i\in\mathbb{Z}}.$$

#### Bernoulli systems

T is Bernoulli if T is isomorphic to a Bernoulli shift.

#### Relations

ergodic  $\subsetneq$  weak mixing  $\subsetneq$  mixing  $\subsetneq$  K  $\subsetneq$  Bernoulli.

The above inclusion also hold in smooth category.

### K non Bernoulli

First Example – Ornstein  $(T, T^{-1})$  - transformation – Kalikow

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The above properties are qualitative and do not require smooth structure.

Let  $f \in C^k(M, \mu)$ , be  $\mu$  preserving, where  $\mu$  is a smooth measure on M.

Definition: Central Limit Theorem

We say that f satisfies the classical CLT if for every  $\phi \in C^k$  with  $\mu(\phi) = 0,$ 

$$\frac{1}{\sqrt{N}}S_N(\phi) := \frac{1}{\sqrt{N}}\sum_{n \le N} \phi \circ f^n \to \mathcal{N}(0, \sigma_{\phi}^2),$$

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# Trivial CLT

 $R_{\alpha}x = x + \alpha$ , for a.e.  $\alpha$ ,

$$S_N(\phi) = o(N^{\epsilon}), \text{ for every } \epsilon > 0.$$

#### Definition: Exponential mixing

 $f \in C^{k}(M,\mu)$  is exponentially mixing if there exists  $C, \eta > 0$  and  $\ell \in \mathbb{N}$  such that for every  $\phi, \psi \in C^{\ell}$ ,

$$\left|\mu(\phi\circ f^n\cdot\psi)-\mu(\phi)\mu(\psi)\right|< C\|\phi\|_{\ell}\|\psi\|_{\ell}e^{-\eta n}.$$

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# Some results

# Known results: rigidity

**I** *K*-property implies mixing (of all orders) (Kolmogorov).

K- property implies Bernoulli in dimension 2 (Pesin).

 exponential mixing of all orders implies CLT (Chernov, Bjorklund-Gorodnik).

- every manifold of dim ≥ 2 supports a Bernoulli diffeomorphism (Katok, Brin-Katok-Rudolph);
- K not Bernoulli examples (Kalikow, Katok, Rudolph, K-Rodriguez-Hertz-Vinhage), generalized (T, T<sup>-1</sup>)-maps;
- 3 non-weakly mixing systems satisfying CLT (Kifer-Conze), Anosov  $\times R_{\alpha}$ .

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# exponential mixing obviously implies mixing.

CET implies ergodicity but not weak mix

#### Questions:

- Does exponential mixing imply positive entropy, higher order mixing, K,Bernoulli?
- 2 Does exponential mixing imply CLT?
- Boes CLT imply positive entropy? (J.-P. Thouvenot)

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## Theorem 1 (Dolgopyat, K, Rodriguez-Hertz)

Exponential mixing implies Bernoulli.

#### Consequences

- **1** Exponential mixing implies positive entropy (and K).
- Exponential mixing implies mixing of all orders.
- From ergodic point of view exponentially mixing systems are classified by entropy.

#### Example:

The system:

$$\begin{pmatrix} e^t & 0 \\ 0 & e^{-t} \end{pmatrix} \times \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}$$

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Does (non-trivial) CLT +K imply Bernoulli?(J.-P. Thouvenot)

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$$T: (X, \mu) \to (X, \mu) \text{ map (or flow)},$$
  
2  $\alpha: (Y, \nu) \to (Y, \nu) \text{ an } \mathbb{R}^d \text{ (or } \mathbb{Z}^d) \text{ action, } d \ge 1$   
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 $F(x, y) = (Tx, \alpha_{\tau(x)}(y))$   
acting on  $(X \times Y, \mu \times \nu)$ .

#### Classical example: random walk in random scenery

 $(X, \mu) = (Y, \nu) = (\{0, 1\}^{\mathbb{Z}}, (1/2, 1/2)^{\mathbb{Z}}), T = \alpha = \sigma_2$  $\tau(x) = (-1)^{x_0}.$ It is K and NOT (loosely) Bernoulli map (Kalikow). It does NOT satisfy the classical CLT (DDKN).

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## Source of K non-Bernoulli examples:

## **1** Smooth *K* non-Bernoulli examples (Katok).

T = Anosov map,  $\alpha = h_t \times h_t$ ,  $\tau$  smooth, positive and not a coboundary ( $h_t$  is the horocycle flow).

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#### Higher rank abelian actions in the fiber

$$T = \sigma$$
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$$Tx = R_{\theta}x = x + \theta$$
 on  $\mathbb{T}^m$ ,  $\theta \in D(\kappa)$ , where

 $D(\kappa) = \{ v \in \mathbb{R}^m : \langle v, k \rangle \ge C_v \| k \|^{-\kappa} \text{ for } k \in \mathbb{Z}^m \}.$ 

 $Leb(D(\kappa)) = 1$  for  $\kappa > m$ . Let  $\kappa/2 < r < m$  and  $d > 20 \cdot rac{1}{1 - rac{r}{m}}.$ 

- 2 α any smooth R<sup>d</sup>-action which is exponentially mixing of all orders.
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$$F(x,y) := (R_{\theta}x, S_{\tau(x)}(y))$$

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#### Main example of $\alpha$

 $\alpha$  is the Weyl chamber flow on  $SL(d, \mathbb{R})$ , i.e.  $\alpha$  is the group of diagonal matrices (isomorphic to  $\mathbb{R}^{d-1}$ ).

#### Steps for Thm 2

- Show existence of  $\tau$  as in 3.
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## Theorem 1, exponential mixing implies Bernoulli

### MAIN IDEAS:

#### (I) Exponential Mixing implies positive entropy.

- (II) By Pesin entropy formula there exists a non-zero Lyapunov exponent. In particular *f* is non-uniformly partially hyperbolic and there is a full measure set of points with non trivial unstable space.
- (III) Exponential mixing implies equidistribution of unstable leaves at exponential scale (for most points).
- (IV) Equidistribution of unstable leaves implies the K-property.
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## Theorem 1, exponential mixing implies Bernoulli

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## Pesin theory, explanation of (II)

Let  $\dim M = d$ ,  $df_x : T_x M \to T_{f(x)} M$  denote the differential of fat x and let  $df_x^{(n)} := df_{f^{n-1}x} \circ \ldots \circ df_x$ .

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There exists numbers  $(\lambda_i)_{i=1}^d$  called Lyapunov exponents and (for a.e. x) a decomposition  $T_x M = \bigoplus_{i=1}^d E_i(x)$  such that for  $v \in E_i(x)$ , ||v|| = 1, we have

$$\lim_{N\to\infty}\frac{1}{N}\log(\|df_x^{(N)}(v)\|)=\lambda_i.$$

Let  $E^u(x) := \bigoplus_{\lambda_i > 0} E_i(x)$  and  $E^s(x) := \bigoplus_{\lambda_i < 0} E_i(x)$  be respectively the unstable and stable space. Moreover let  $E^c(x) = \bigoplus_{\lambda_i = 0} E_i(x)$  be the center space.

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# Entropy formula

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### Equidistribution of unstable leaves

For every  $\epsilon > 0$  there exists a  $\delta > 0$  and a set  $P_{\epsilon} \subset M$ ,  $\mu(P_{\epsilon}) > 1 - \epsilon$  such that for every  $x \in P_{\epsilon}$  we have the unstable local foliation of size  $\delta$ , i.e.  $W^{u}(x, \delta) \subset M$  such that  $E^{u}_{f^{-n}x}$  is tangent to  $W^{u}(f^{-n}x, \delta)$  and this local foliation is exponentially contracted when iterating backwards. The same with the stable foliation. We call the normalized unstable measure on  $W^{u}(x, \delta)$  by  $m^{u}$ .

#### Theorem

Let f be exponentially mixing. There exists  $\eta'' > 0$  such that for every n and every ball  $B \in M$  of radius  $\geq e^{-\eta'' n}$  we have

 $m^u(W^u(x,\delta)\cap f^{-n}(B))\in (1-\epsilon,1+\epsilon)\mu(B).$
### Reduction

Let f be K.Let  $W^u(x, \delta)$  and  $W^u(y, \delta)$  be nearby unstable leaves of size  $\delta$ . If for every N there exists an almost measure preserving map  $\theta_{x,y,\delta,N} : (W^u(x, \delta), m_x^u) \to (W^u(y, \delta), m_y^u)$  such that

 $f^n z$  and  $f^n \theta z$  are close for most  $0 \le n \le N$ . then f is Bernoulli.

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### THANK YOU!