

Quantitative results in Poincaré recurrence for mixing processes

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Based on recent works with M. Abadi (IME-USP), V. Amorim (IFSP), JR. Chazottes (Polytechnique), N. Haydn (USC), S. Vaienti (Aix-Marseille).

- 1 **Setting**
- 2 First return time and number of returns
- 3 Fluctuations of the return time

Dynamical system notation is easier for this talk:

Consider (Ω, T, μ) , $T : \Omega \rightarrow \Omega$, and a T -invariant probability measure $\mu \circ T^{-1} = \mu$.

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To fix idea: think about an i.i.d. sequence of $\text{Ber}(p)$ r.v.'s!!

Présentation

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Hitting time of x in the measurable $B \subset \Omega$

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Theorem (Poincaré Recurrence Theorem (1890))

T -invariance $\Rightarrow \mu_B(\tau_B(x) < \infty) = 1$ for B with $\mu(B) > 0$.

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Ergodicity $\Rightarrow \mathbb{E}_{B \tau_B} = \frac{1}{\mu(B)} \quad (\mu(B) > 0)$.

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Theorem (The Kac Lemma (1947))

Ergodicity $\Rightarrow \mathbb{E}_{B\tau_B} = \frac{1}{\mu(B)}$ ($\mu(B) > 0$).

→ First quantitative result.

What about $\mu_B(\tau_B \mu(B) > t)$?

Theorem (Folklore)

For suitable dynamics and suitable $B_n(x) \rightarrow x \in \Omega$

$$\mu_{B_n}(\tau_{B_n} \mu(B_n) > t) \xrightarrow{n \rightarrow \infty} e^{-t}, \quad \underline{\mu - a.e. x \in \Omega.}$$

The scaling factor $\mu(B_n)$ comes from the Kac Lemma.

“suitable dynamics”: mixing conditions (stronger than ergodicity)

“suitable $B_n(x)$ ”: Think about “cylinders/patterns”.

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The new scaling parameter $\lambda(B_n(x))$ accounts for periodic x 's.

And $\lambda(B_n(x)) \rightarrow 1$ for aperiodic x 's.

Abadi *et al.* (2000's): Several possible choices for $\lambda(B_n(x))$, but a simple one is

$$\mu_{B_n}(\tau_{B_n} > \text{"first possible return to } B_n\text{"})$$

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Theorem (Pitskel (1991), Hirata (1993)...)

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Theorem (Haydn and Vaienti (2009) for cylinder sets)

For suitable dynamics and *cylinder sets or balls* $B_n(x) \rightarrow x \in \Omega$

$$\mathcal{L}(N_n) \Rightarrow \text{Pólya-Aeppli}(t\lambda) \quad , \quad \forall x \in \Omega$$

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Pólya-Aeppli($t\lambda$) is the Compound Poisson $\sum_{i=1}^M X_i$ with

- X_i 's are i.i.d. $\text{Geo}(\lambda)$
- $M \sim \text{Pois}(t\lambda)$.

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- G., Haydn and Vaienti (to appear, 2023)

-> for symbolic dynamical systems.

Motivation: approximate synchronization of stochastic processes.

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For ergodic measure μ

$$\frac{\log r_n}{n} \rightarrow h_\mu \quad \mu - \text{a.e. } x \in \Omega$$

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What about the fluctuations?

Theorem (Collet, Galves and Schmitt (1999))

If μ satisfies

- strongly mixing,
- with $\sigma^2 := \lim \text{Var} \left(\frac{\log r_n}{\sqrt{n}} \right) > 0$,

then:

$$\mathcal{L} \left(\frac{\log r_n/n - h_\mu}{\sigma/\sqrt{n}} \right) \Rightarrow \mathcal{N}(0, 1),$$

and there exists a $u_0 > 0$ s.t. for $u \in [0, u_0[$

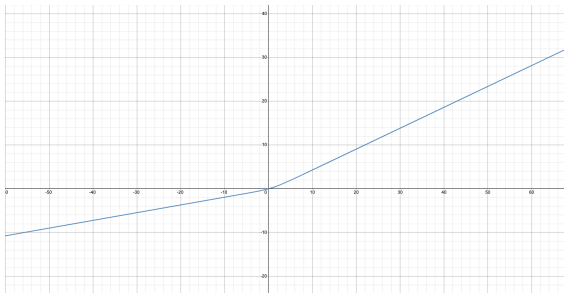
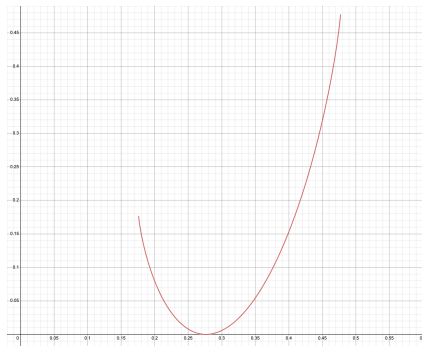
$$- \lim \frac{1}{n} \log \mu \left(\frac{\log r_n}{n} > h + u \right) = m(h + u)$$

$$- \lim \frac{1}{n} \log \mu \left(\frac{\log r_n}{n} < h - u \right) = m(h - u),$$

where m is the Legendre-Fenchel transform of

$$\mathcal{M}(q) := \lim_n \frac{1}{n} \log \mathbb{E} \left(\mu(X_1^n)^{-1} \right)^q, \quad q \in \mathbb{R}.$$

Observation: Example for i.i.d. $\text{Ber}(1/3)$



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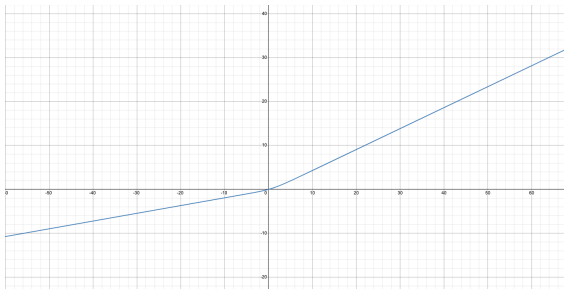
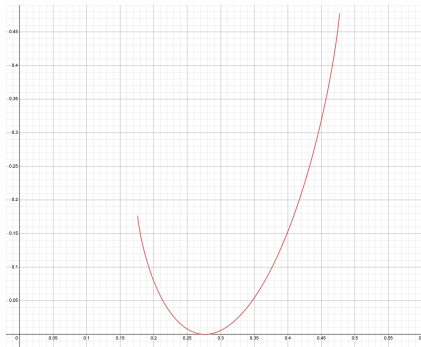
Same conditions as before:

- Λ exists and equals $\lim_n \frac{1}{n} \log \max_{a_0^{n-1}} \mu(a_0^{n-1})$
- and $\mathcal{R}(q)$ exists and

$$\begin{aligned} \mathcal{R}(q) &= \mathcal{M}(q) && \text{for } q > q^* \\ &= \Lambda && \text{for } q \leq q^* \end{aligned}$$

where $q^* \in (-1, 0)$ is the solution of the equation $\mathcal{M}(q) = \Lambda$.

Consequence:



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These **local** considerations yield “ansatz”

$$r_n \longleftrightarrow \mu(X_1^n)^{-1}$$

and

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Grassberger and Procaccia (1983): return time picture to estimate the dimension.

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Haydn *et al.* (2002) shown, by simulations, that this doesn't work.

THANK YOU!!!