

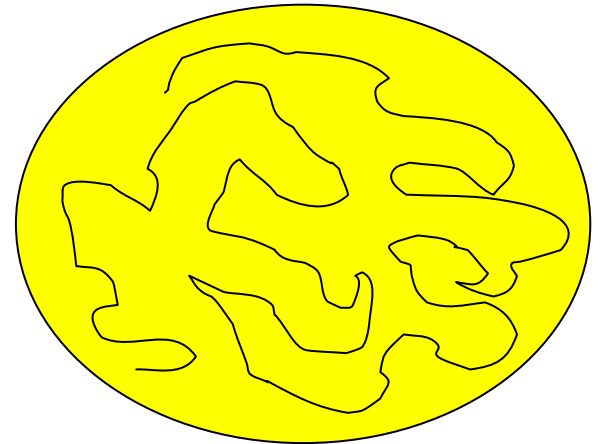
1次元間欠写像における 長時間平均のふるまい

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What is ergodicity of the non-equilibrium state?

1. Introduction (Ergodic problems and Intermittent phenomena)
2. Universal Distributions
(Mittag Leffler, Generalized Arcsine, Stable)
3. Concluding Remarks



Ergodic Theory

Birkhoff's ergodic theorem. (1931) Let T be measure preserving transformation of the probability space (X, \mathcal{B}, m) , then

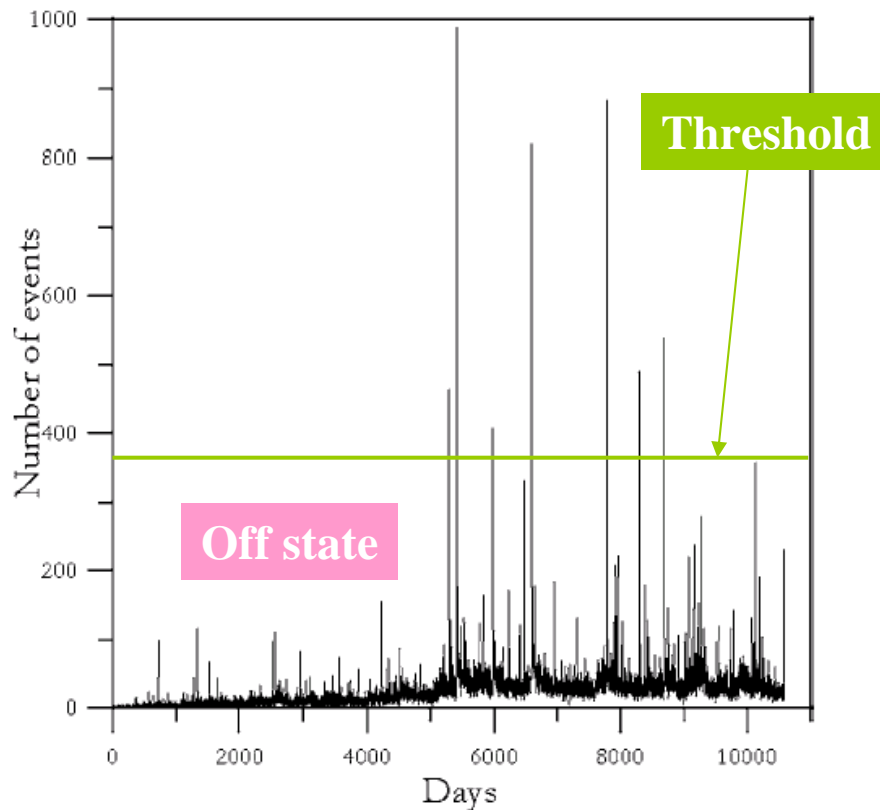
$$\frac{1}{n} \sum_{k=0}^{n-1} f \circ T^k \rightarrow \int_X f dm \quad \text{a.e. } \forall f \in L^1(m).$$

Question

$$\frac{1}{n} \sum_{k=0}^{n-1} f \circ T^k \rightarrow ? \quad \text{a.e. } \forall f \notin L^1(m).$$

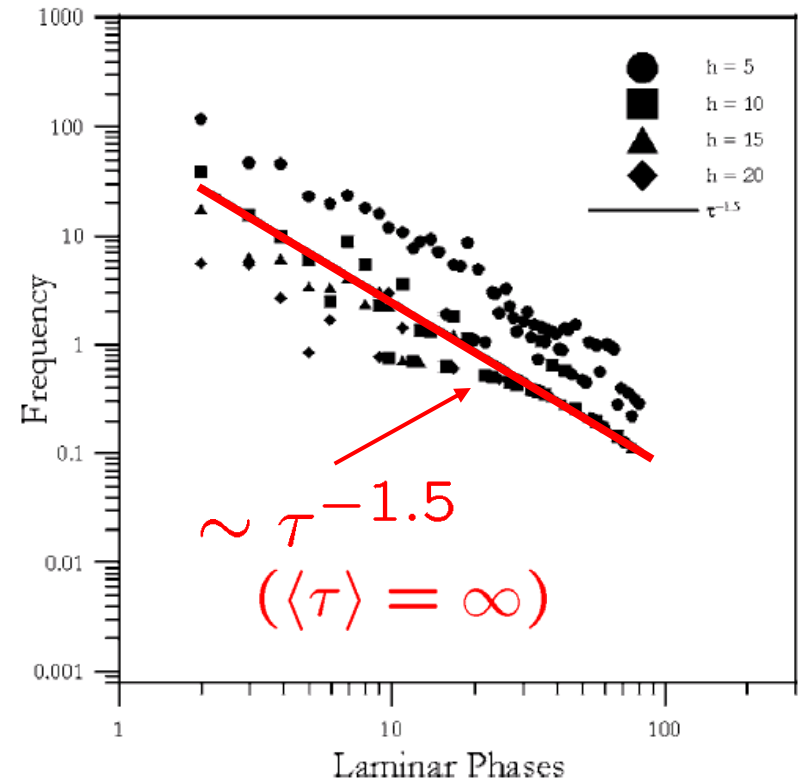
Intermittent Phenomena

M. Bottiglieri and C. Godano, **On-off intermittency in earthquake occurrence**, *Phys.Rev. E* **75**, 026101 (2007).



$\Delta t = 1$ (day)

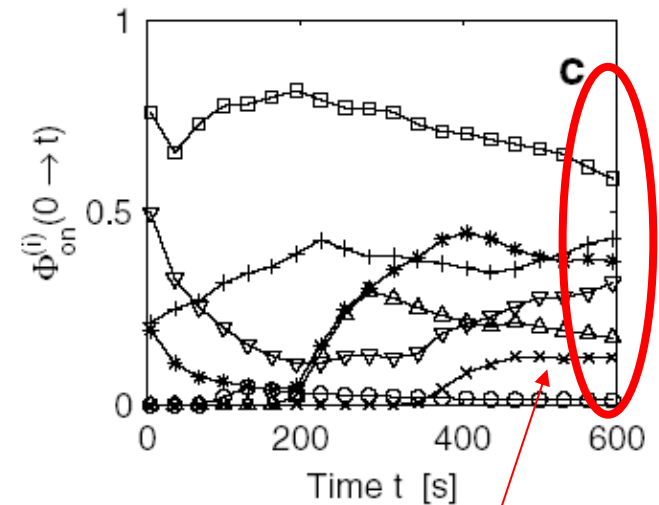
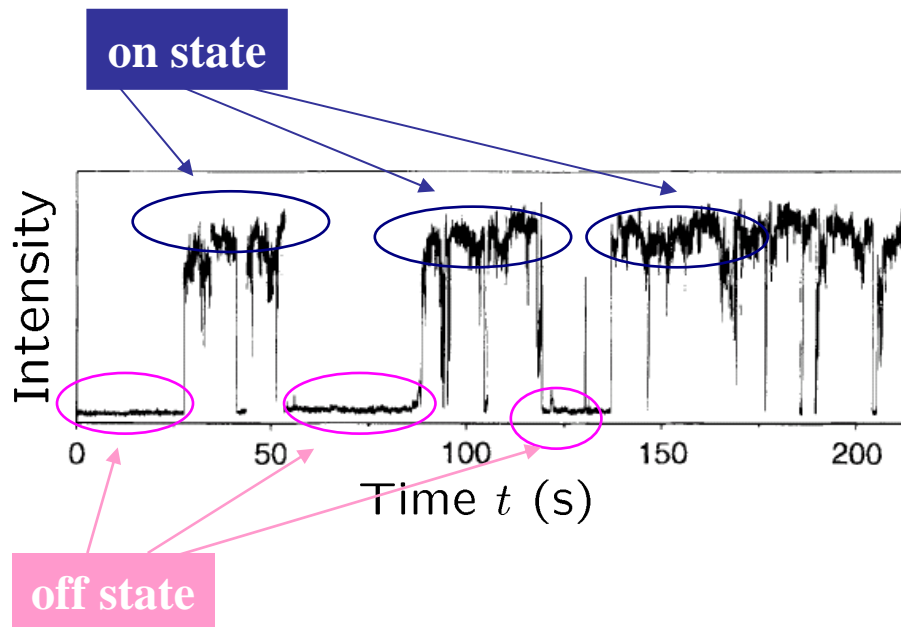
(the Southern California catalog, 1973-2003)



$\Delta t = 600$ (min)

Nonergodicity

X.Brokmann et al., **Statistical aging and nonergodicity in the fluorescence of single nanocrystals**, *Phys. Rev. Lett.* **90**, 120601 (2003).



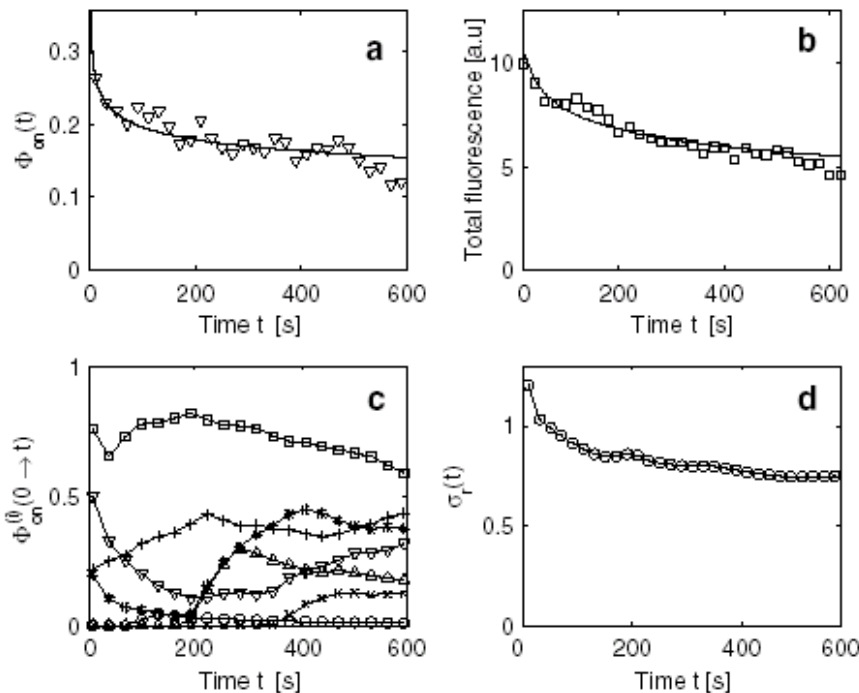
Characteristics

**Power law, $1/f$ spectrum,
Non-stationarity, Non-ergodicity**

The ratio of the on state time $\Phi_{on}(t)$ does not converge.

Nonergodicity and Non-stationarity

Nonstationary and non-ergodic behavior



the fraction of QDs in the on state

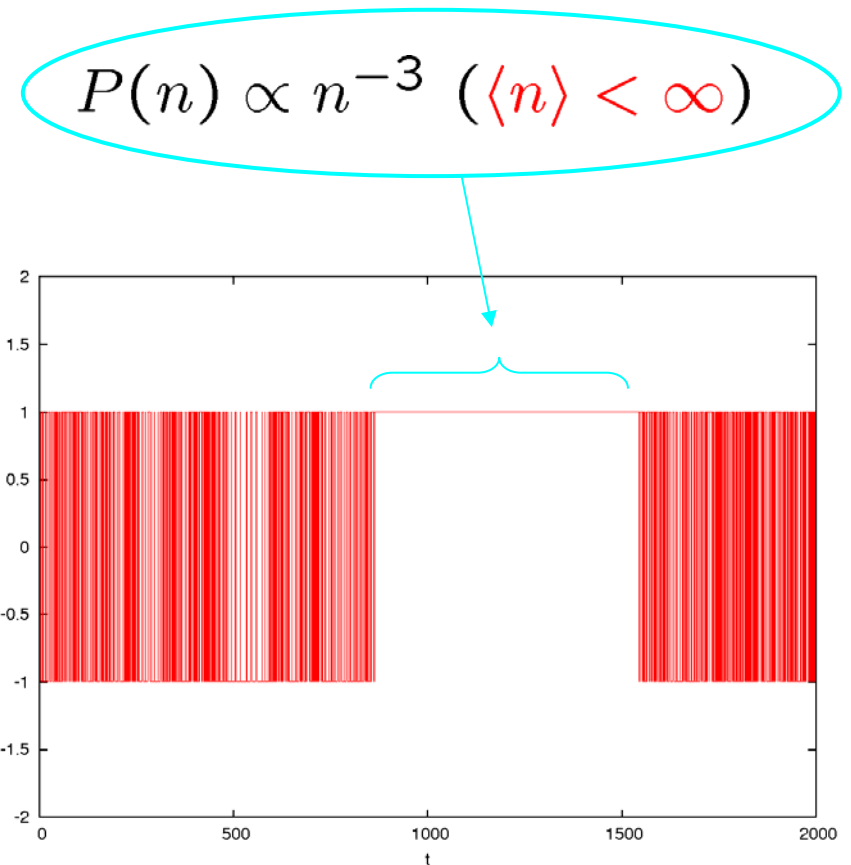
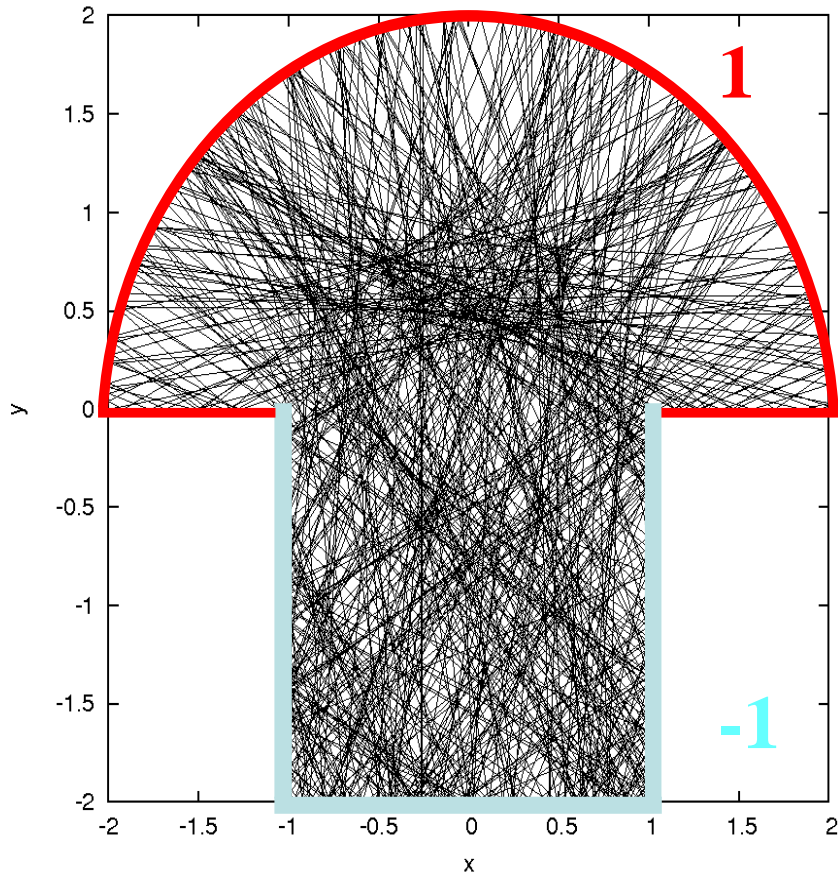
$$\Phi_{\text{on}}(t) = \frac{1}{215} \sum_{k=1}^{215} 1_{[I_0, \infty)}(I_k(t)) \sim t^{-\beta_1}$$

the fraction of time in the on state
(for the 7th QD)

$$\Phi_{\text{on}}^{(i)}(t) = \frac{T_{\text{on}}^{(i)}(t)}{t}$$

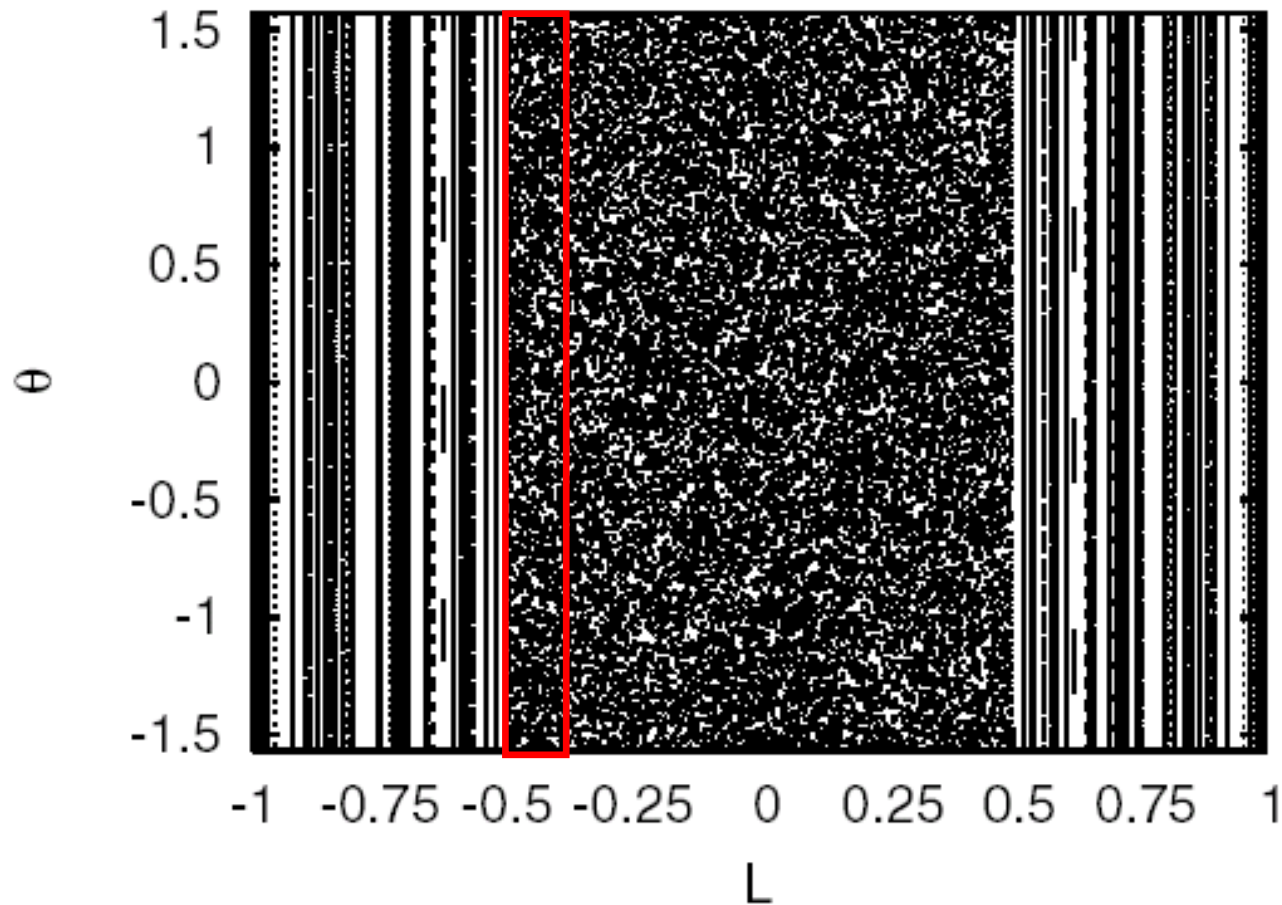
Time average does not converge to constant value.

Mushroom Billiard



T. Miyaguchi, **Escape time statistics for mushroom billiard**, *Phy. Rev. E* **75** 066215 (2007). (Aizawa lab. Waseda univ. Satoru Tsugawa)

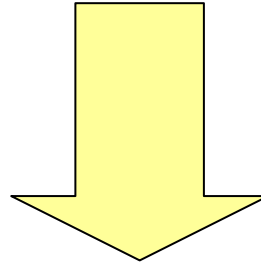
Remark



Survival probability $\Rightarrow P(n) \propto n^{-2}$ ($\langle n \rangle = \infty$)

Purpose

Analyzing **the distribution of the time average** of some observation functions in **infinite measure systems** which are related to intermittent phenomena.



To make clear the non-stationary phenomena and **the foundation of the ergodicity in the non-equilibrium state**

Conditions

Let c denote the inner endpoint of the monotonicity intervals, and let $I_0 = (0, c)$, $I_1 = (c, 1)$. Then T is assumed to satisfy the following conditions:

- (i) $T|_{I_j}$ has a C^1 -extension $\bar{T}|_{\bar{I}_j}$ to \bar{I}_j , $(\bar{T}|_{\bar{I}_j})' > 0$, and $\overline{T(I_j)} = [0, 1]$ ($j = 1, 2$).
- (ii) T is convex in a neighbourhood of 0.
- (iii) T admits an invariant measure μ equivalent to λ , such that the density $d\mu/d\lambda$ has a version of the form

$$h(x) = \tilde{h}(x)/x^p, \quad x \in (0, 1], \text{ where } p \geq 1,$$
 and \tilde{h} is positive, continuous and of bounded variation on $[0, 1]$.
- (iv) The function $\psi = \frac{h \circ T \cdot T'}{h}$ is increasing on I_0 .

Infinite measure systems

♠ Lasota-Yorke map T is defined on $[0, 1]$ as

$$Tx = \begin{cases} \frac{x}{x-1} & x \in [0, 1/2) \\ 2x - 1 & x \in [1/2, 1], \end{cases}$$

which has invariant density $\rho(x) = 1/x$, and $Tx - x \sim x^2$ ($x \rightarrow 0$).

♠ Boole transformation $T : \mathbb{R} \rightarrow \mathbb{R}$ is defined as

$$Tx = x - 1/x$$

which has **uniform** invariant density.

Invariant density can not be normalized.

♠ Log-Weibull map

$$Tx = \begin{cases} x + x^2 e^{-1/x} & x \in [0, a] \\ (x - a)/(1 - a) & x \in (a, 1]. \end{cases}$$

The invariant density has an **essential singularity** at 0.

$$\rho(x) = g(x)e^{1/x}/x, \quad g \text{ continuous and positive on } [0, 1].$$

The residence time distribution obeys the **Log-Weibull distribution**.

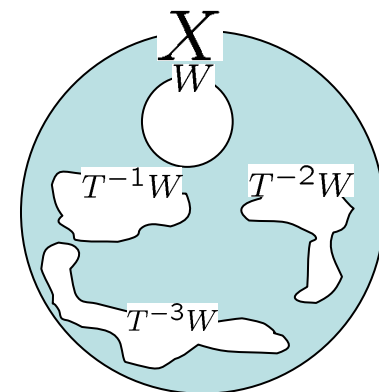
Conservative and Dissipative

Def. Wandering set

A set $W \subset X$ is called a *wandering set* if the sets $\{T^{-n}W\}_{n=0}^{\infty}$ are disjoint.

$\mathcal{D}(T)$ (dissipative part): a countable union of wandering sets

$\mathcal{C}(T)$ (conservative part): $\mathcal{C}(T) = X \setminus \mathcal{D}(T)$



Def. Conservative and Dissipative

The transformation T is called conservative if $\mathcal{C}(T) = X \bmod \mu$, and *dissipative* if $\mathcal{D}(T) = X \bmod \mu$.

Remark

$\mu(X) < \infty \Rightarrow$ conservative

DKA Limit Theorem

Darling – Kac – Aaronson Limit Theorem (1981)

If T is a conservative, ergodic measure preserving transformation of (X, \mathcal{B}, m) , then $\exists a_n$ s.t. $\forall f \in \underline{L^1_+}(m)$

$$\frac{1}{a_n} \sum_{k=0}^{n-1} f(T^k \cdot) \rightarrow M_\alpha \quad \text{as } n \rightarrow \infty.$$

Random variable (Mittag-Leffler distribution)

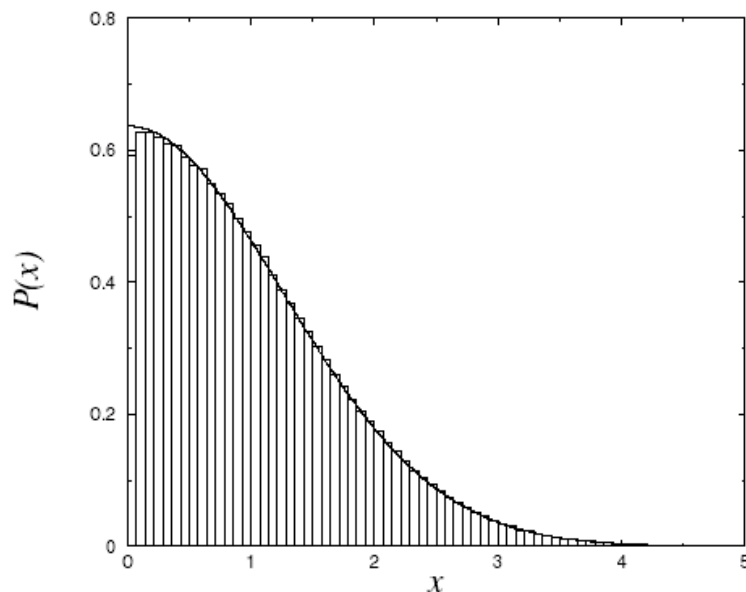
$$E(e^{-zM_\alpha}) = \sum_{k=0}^{\infty} \frac{\Gamma(1+\alpha)^k (-z)^k}{\Gamma(1+k\alpha)}$$

$$f > 0 \text{ and } \int_X |f| dm < \infty$$

Lyapunov exponent

$$f(x) = \log |(Tx)'|$$

Mittag-Leffler Distribution

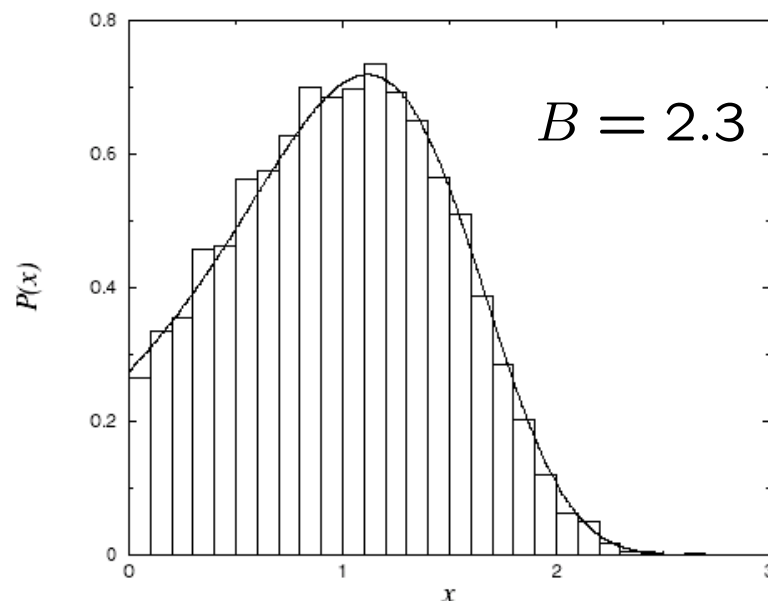


Lyapunov exponent for Boole transformation

$$Tx = x - \frac{1}{x}$$

(return sequence : $a_n = \frac{n^{1/2}}{\sqrt{2\pi}}$)

$$f(x) = \log |(Tx)'|$$



Lyapunov exponent for MB map

$$Tx = \begin{cases} x + 2^{B-1}x^B & x \in [0, 1/2] \\ x - 2^{B-1}(1-x)^B & x \in (1/2, 1] \end{cases}$$

(return sequence : $a_n \propto n^{\beta-1}$)

Skew Modified Bernoulli Map

The skew modified Bernoulli map is closely related to the **intermittent phenomena**.

(Rayleigh-Benard convection, Lorentz model)

Skew Modified Bernoulli map

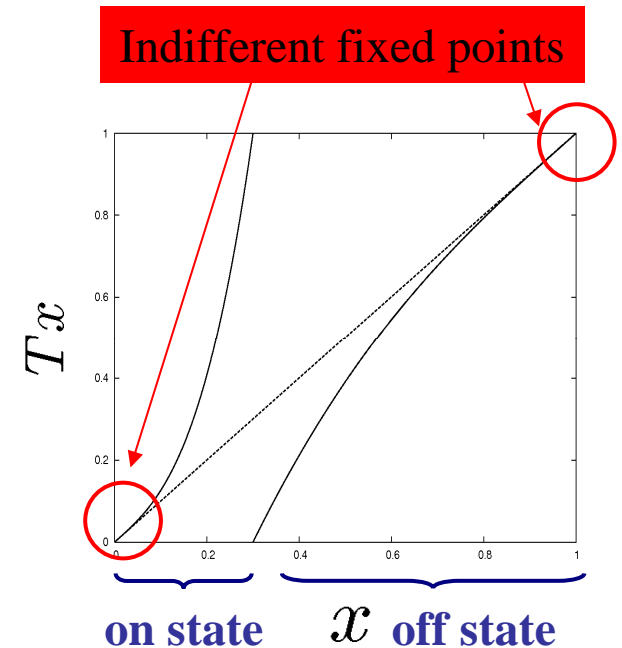
$$Tx = \begin{cases} x + (1 - c) \left(\frac{x}{c}\right)^B & x \in [0, c] \\ x - c \left(\frac{1-x}{1-c}\right)^B & x \in (c, 1] \end{cases}$$

Invariant measure (**Infinite measure**)

$$\rho(x) \propto x^{1-B} + (1-x)^{1-B}$$

$B < 2$: finite measure

$B \geq 2$: infinite measure



MB map ($B=3.0, c=0.3$)

Lamperti-Thaler Generalized Arcsine Law

Let T be the skew modified Bernoulli map, then

$$\lim_{n \rightarrow \infty} \Pr \left(\frac{1}{n} \sum_{k=0}^{n-1} 1_{[0,c]} \circ T^k \leq x \right) = G_{\alpha_1, \alpha_2}(x),$$

where $\alpha_1 = \beta - 1$,

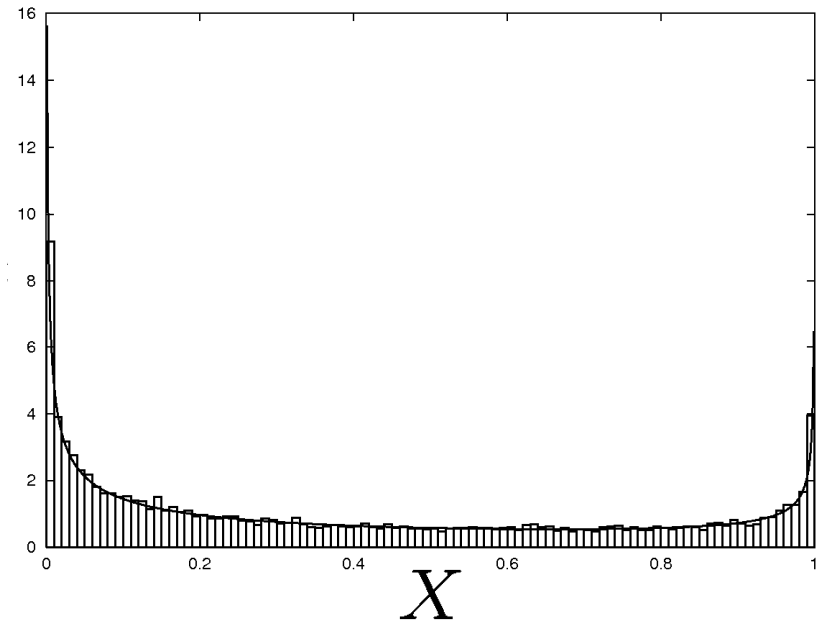
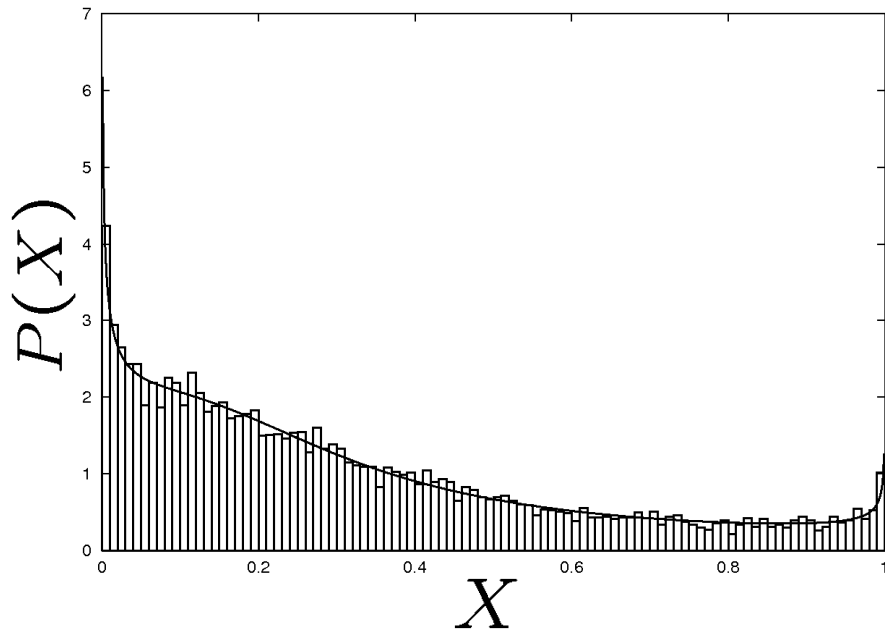
$$\alpha_2 = \frac{1 + (B - 1)c}{1 + (B - 1)(1 - c)} \left(\frac{1 - c}{c} \right)^{\frac{2}{B-1}}$$

and the p.d.f. $G'_{\alpha_1, \alpha_2}(x)$ is given by

$$G'_{\alpha_1, \alpha_2}(x) = \frac{\alpha_2 \sin \pi \alpha_1}{\pi} \frac{x^{\alpha_1-1} (1-x)^{\alpha_1-1}}{\alpha_2^2 x^{2\alpha_1} + 2\alpha_2 x^{\alpha_1} (1-x)^{\alpha_1} \cos \pi \alpha_1 + (1-x)^{2\alpha_1}}$$

M. Thaler, A limit theorem for sojourns near indifferent fixed points of one-dimensional maps, *Ergod. Th. & Dynam. Sys.* 22 1289-1312 (2002).

Generalized Arcsine Law



Probability density function of $X = \frac{1}{n} \sum_{k=0}^{n-1} 1_{[0,c]} \circ T^k$

Remark on the invariant density and mean

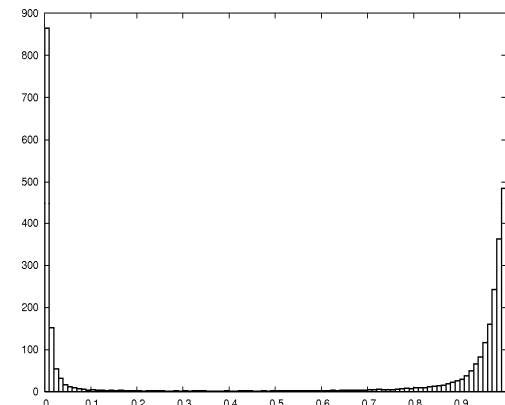
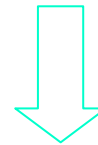
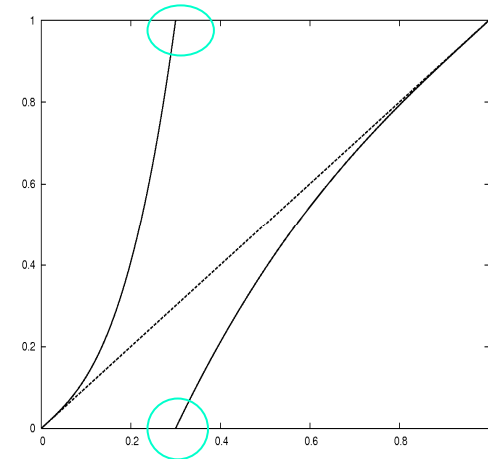
$$P_+(n_+) \propto \left(\frac{n_+}{c_+} \right)^{-\frac{B}{B-1}} \quad (\text{On state})$$

$$P_-(n_-) \propto \left(\frac{n_-}{c_-} \right)^{-\frac{B}{B-1}} \quad (\text{Off state})$$

$$\frac{c_+}{c_-} \propto \frac{\langle n_+ \rangle}{\langle n_- \rangle} = q(c) = \frac{1 - \alpha_2}{\alpha_2}$$

$$q(c) \neq 1 \quad (c \neq 1/2) \text{ and } q'(c) > 0$$

The invariant density is **not symmetric**.



Universal Distributions

Time Average of the observation function

$$\frac{1}{a_n} \sum_{k=0}^{n-1} f \circ T^k = \frac{f + f \circ T + \dots + f \circ T^{n-1}}{a_n} \rightarrow \mathbf{X} \text{ as } n \rightarrow \infty$$

where $a_n = n^\gamma$.

Random variables

Distributional Limit Theorems for the Time Average

$f(x)$	invariant measure	γ	distribution of X
L_m^1	finite	1	<i>delta</i> (Birkhoff, 1931)
L_m^1	infinite	$\frac{1}{B-1}$	<i>Mittag-Leffler</i> (Aaronson, 1981)
non- L_m^1	finite	$\frac{\alpha}{2-B}$	Stable (Akimoto [1])
$L_{loc}^1(m, (0, 1))$	infinite	1	Generalized Arcsine (Akimoto [1])
non- L_m^1	infinite	$\frac{\alpha}{B-1} + 1$	Stable (Akimoto [1])

[1] T. Akimoto, Generalized Arcsine Law and Stable Law in an Infinite Measure System, arXiv:0801.1382v.

$L^1_{\text{loc},m}(\Omega)$ with finite mean

Definition

Locally integrable

Finite mean

A function $f(x)$ is said to be *locally integrable with finite mean* with respect to the measure m on Ω provided that $f \in L^1_m(K) \forall$

compact set $K \subset \Omega$: $\int_K |f(x)| dm < \infty$; and $\lim_{\epsilon \rightarrow 0} \frac{\int_{\epsilon}^{1-\epsilon} f dm}{\int_{\epsilon}^{1-\epsilon} dm} < \infty$.

In this case we call f the $L^1_{\text{loc},m}(\Omega)$ function with finite mean.

Examples in the MB map

$$\spadesuit f(x) = \begin{cases} 1 & x \in [0, 1/2) \\ -1 & x \in [1/2, 1) \end{cases}$$

$$\spadesuit f(x) = x$$

are the $L^1_{\text{loc},m}(0, 1)$ function with finite mean. (**non- L^1_μ functions**).

Generalized Arcsine Law

Let T be the skew MB map and f be the $L^1_{\text{loc},m}(0,1)$ function with finite mean and $f(0) = a, f(1) = b$. Further, there exists δ such that $0 < \delta < 1$ and $f(x)$ is continuous in $[0, \delta] \cup [1 - \delta, 1]$. Then the time average converges in distribution to $Y_{\alpha_1, \alpha_2, a, b}$:

$$\frac{1}{n} \sum_{k=0}^{n-1} f(T^k \cdot) \rightarrow Y_{\alpha_1, \alpha_2, a, b} \text{ as } n \rightarrow \infty$$

where $\gamma = 1/(B - 1)$ and the p.d.f. of $Y_{\alpha_1, \alpha_2, a, b}$ for $a > b$ is given by

Generalized Arcsine distribution

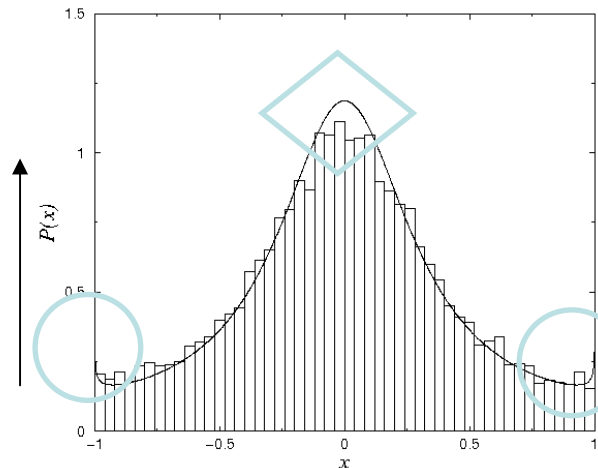
Numerical Simulations ($c = 1/2$)

$$\underline{B = 2.2} \quad (\alpha_1 \cong 0.83)$$

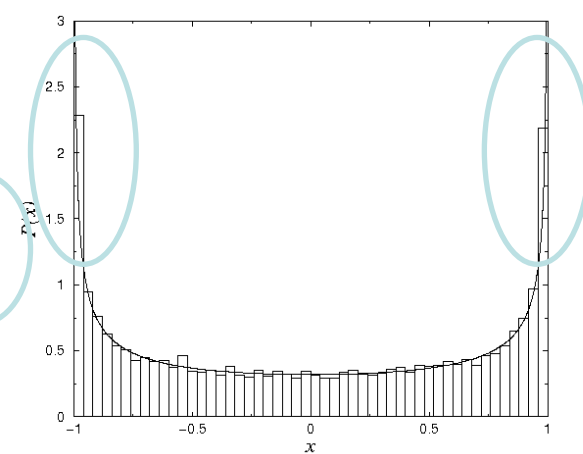
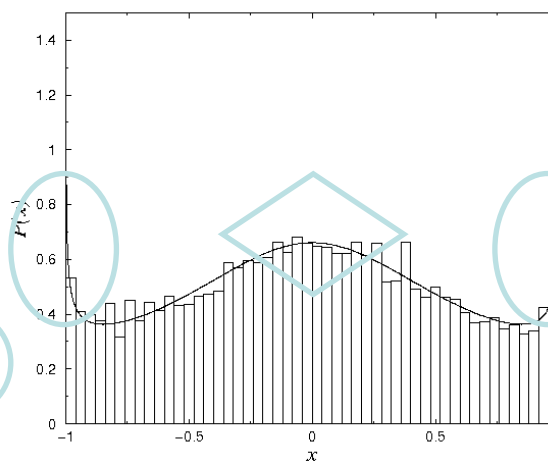
$$\underline{B = 2.4} \quad (\alpha_1 \cong 0.71)$$

$$\underline{B = 3.0} \quad (\alpha_1 = 0.5)$$

p.d.f. of the time average



Time average



Distributions of the time average

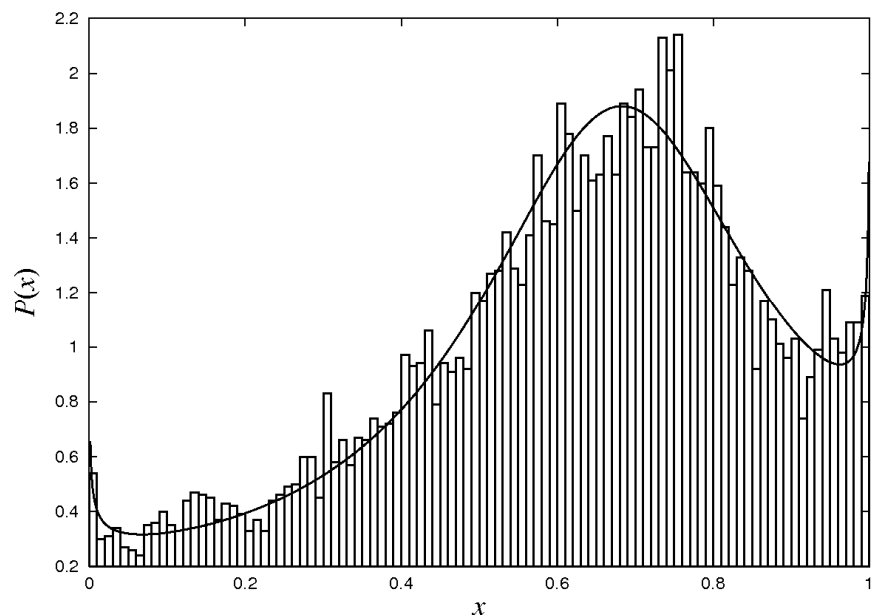
Observation function

$$f(x) = \begin{cases} 1 & (x \leq \frac{1}{2}) \\ -1 & (x > \frac{1}{2}) \end{cases}$$

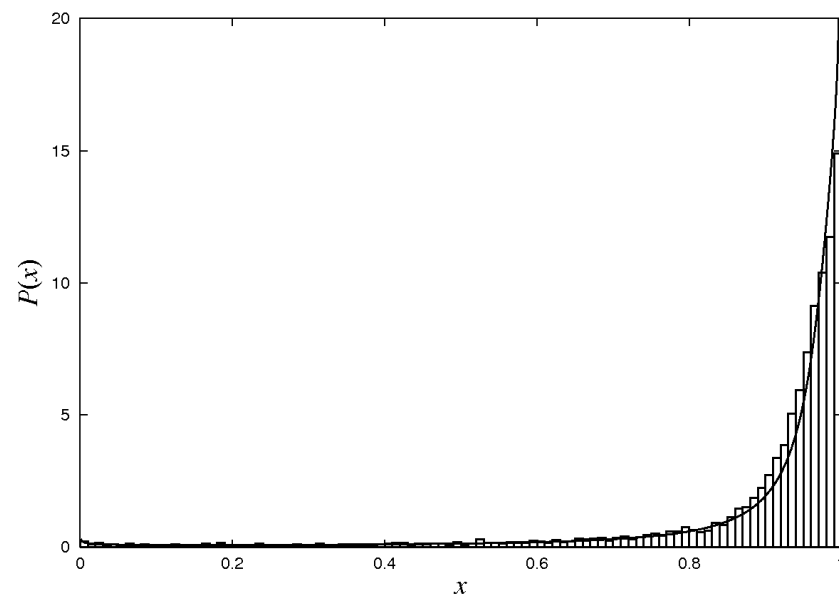
As the value of B becomes large, the middle peak becomes low and edge peaks become high.

Numerical Simulations

$B = 2.3$ and $c = 0.4$



$B = 2.5$ and $c = 0.1$



Probability density function of $x = \frac{1}{n} \sum_{k=0}^{n-1} f \circ T^k$

$(f(y) = y^2)$

Application to Correlation Function

For all $n \geq 0$

$$C(n) = \frac{1}{N} \sum_{k=0}^{N-1} f(x_k)g(x_{k+n}) \rightarrow Y_{\alpha_1, \alpha_2, a, b} \quad N \rightarrow \infty$$

where $f \cdot g \circ T^n \in L^1_{\text{loc}, m}(0, 1)$ with finite mean and $f(0)g(0) = a, f(1)g(1) = b$.

Correlation function is **intrinsically random** (Generalized Arcsine distribution) and **never decays**.

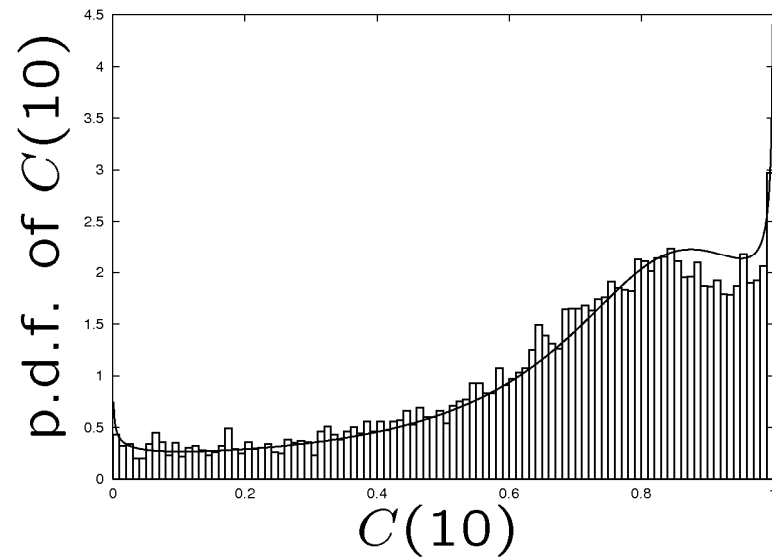
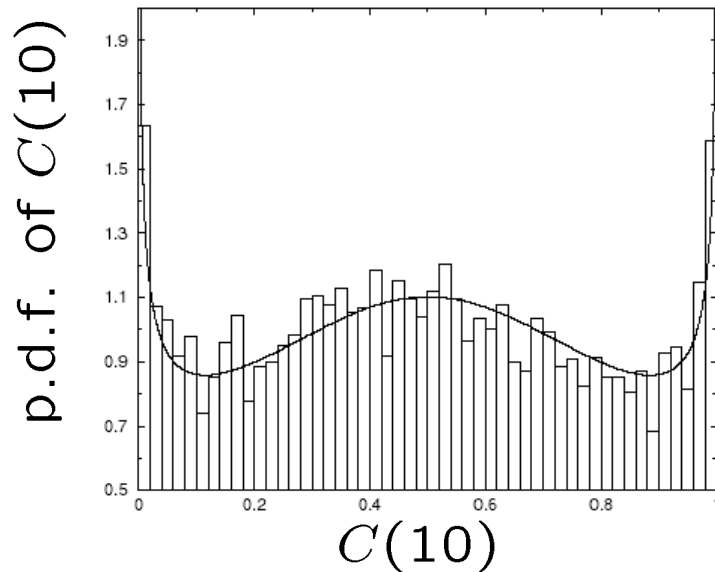
Remark. The convergence becomes slow as n becomes large.

$$\left(C(n) = \frac{1}{N} \sum_{k=0}^{N-1} x_k x_{k+n} \right)$$

Correlation Function

$B = 2.5$ and $C = 0.5$

$B = 2.4$ and $C = 0.4$



Probability density function of $C(10) = \frac{1}{n} \sum_{k=0}^{n-1} x_k x_{k+10}$

Remark on Wiener-Khintchine Theorem

Wiener-Khintchine Theorem

$$S(k, N) \cong \sum_{n=1}^N C(n) \cos(2\pi kn/N)$$

where $C(n)$ is the correlation function. In the case of the power law decay, namely, $C(n) \sim n^{-\alpha}$, **there is a possibility that $S(k, N)$ diverges**. Actually, $S(k, N)$ diverges when the exponent α is smaller than 1.

$$\begin{aligned} S(k, N) &\cong \sum_{n=1}^N n^{-\alpha} \cos(2\pi kn/N) = N^{1-\alpha} \frac{1}{N} \sum_{n=1}^N \left(\frac{n}{N}\right)^{-\alpha} \cos\left(2\pi k \frac{n}{N}\right) \\ &\sim c(k) N^{1-\alpha} \cong k^{\alpha-1} N^{1-\alpha} \quad \text{as } N \rightarrow \infty \end{aligned}$$

where $c(k) = \int_0^1 x^{-\alpha} \cos(2\pi kx) dx = k^{\alpha-1} \frac{1}{(2\pi)^{-\alpha}} \int_0^1 x^{-\alpha} \cos x dx < \infty$.

Power Spectrum

$$S_k = |\hat{x}_k|^2 = \frac{1}{N^2} \sum_{j=1}^N \sum_{l=1}^N x_j x_l \cos \left(\frac{2\pi k}{N} (j - l) \right)$$

$C(n)$: random $\Rightarrow S_k$: random

Question

1. The distribution of S_k
2. Scaling of S_k ($S_k \sim N^{-\nu}$)

Distribution of S_0

$$S_0 = \frac{1}{N^2} \sum_{j=1}^N \sum_{l=1}^N x_j x_l = \underbrace{\left(\frac{1}{N} \sum_{j=1}^N x_j \right)}_{\text{Generalized arcsine distribution}} \underbrace{\left(\frac{1}{N} \sum_{l=1}^N x_l \right)}_{\text{Generalized arcsine distribution}}$$

$\rightarrow (Y_{\alpha_1, \alpha_2, 0, 1})^2$

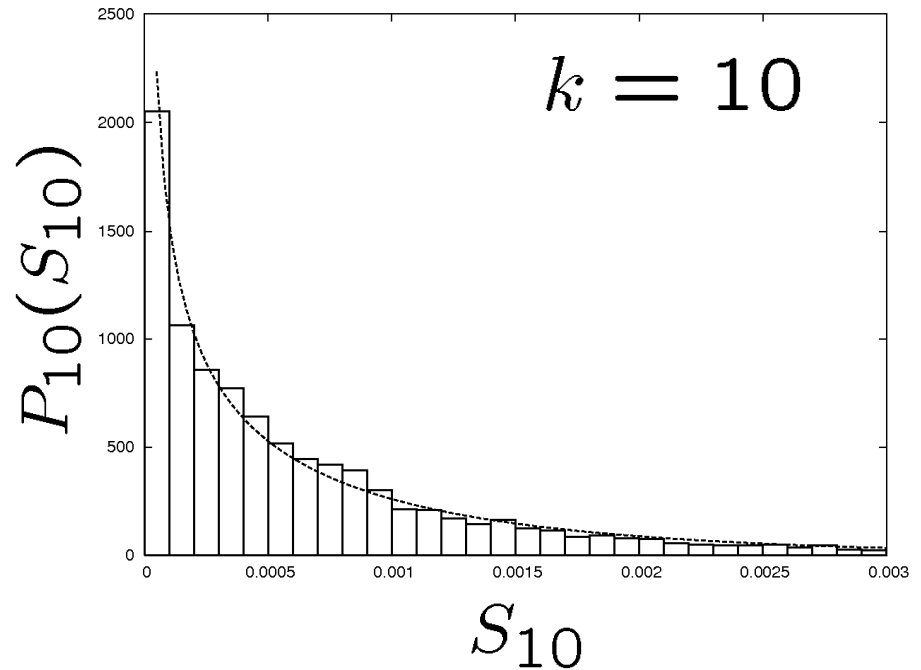
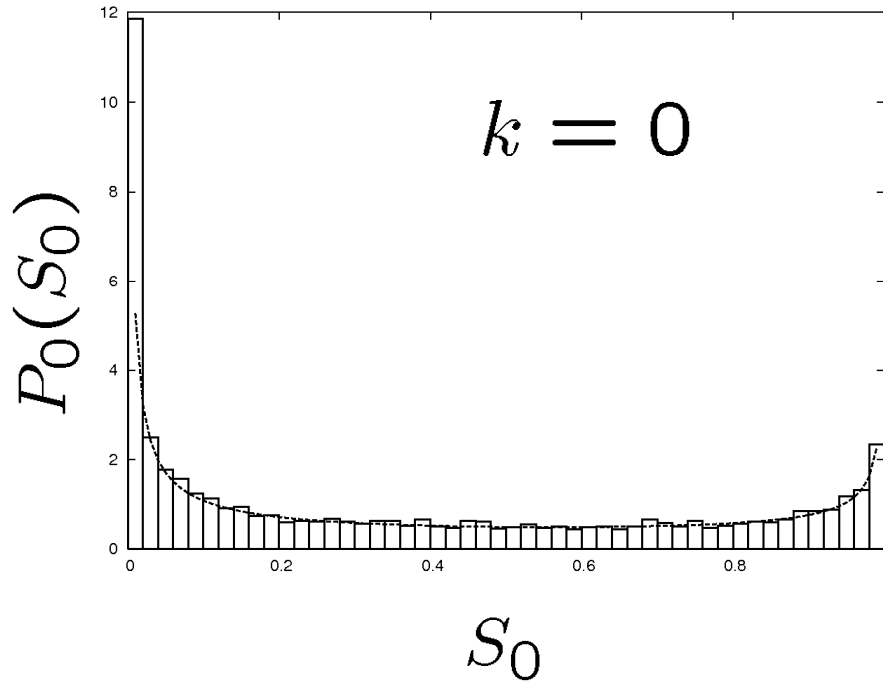
The probability density function of $x = S_0$ is given by

$$P_0(x) = \frac{1}{2\sqrt{x}} G'_{\alpha_1, \alpha_2}(\sqrt{x})$$

$$\left[\Pr(S_0 < x) = \Pr(Y_{\alpha_1, \alpha_2, 0, 1} < \sqrt{x}) \right]$$

Numerical Simulations

$B = 3.0$ and $c = 0.5$



$$P_0(x) = \frac{1}{2\pi\sqrt{x}\sqrt{\sqrt{x}(1-\sqrt{x})}}$$

$$P_{10}(x) = Ax^{-0.48}e^{-750x}$$

(Gamma distribution)

$L^1_{loc,m}(\Omega)$ with infinite mean

Definition

Locally integrable

Infinite mean

A function $f(x)$ is said to be *locally integrable with finite mean* with respect to the measure μ on Ω provided that $f \in L^1_\mu(K) \forall$ compact

set $K \subset \Omega$: $\int_K |f(x)| dm < \infty$; and $\lim_{\epsilon \rightarrow 0} \frac{\int_\epsilon^{1-\epsilon} f dm}{\int_\epsilon^{1-\epsilon} dm} = \infty$.

In this case we call f the $L^1_{loc,m}(\Omega)$ function with infinite mean.

$$\frac{1}{a_n} \sum_{k=0}^{n-1} f \circ T^k \rightarrow ? \quad \text{a.e. } \forall f \in L^1_{loc,m}(0,1) \text{ with infinite mean.}$$

Stable Distributions

Theorem 1. For fixed $0 < \alpha < 1$ the function $\gamma_\alpha(\lambda) = e^{-\lambda^\alpha}$ is the Laplace transform of a distribution G_α with the following properties:

G_α is stable; more precisely, if $\mathbf{X}_1, \dots, \mathbf{X}_n$ are independent variables with the distribution G_α , then $(\mathbf{X}_1 + \dots + \mathbf{X}_n)/n^{1/\alpha}$ has again the distribution G_α .

$$x^\alpha [1 - G_\alpha(x)] \rightarrow \frac{1}{\Gamma(1 - \alpha)} \quad x \rightarrow \infty,$$

$$e^{x^{-\alpha}} G_\alpha(x) \rightarrow 0 \quad x \rightarrow 0.$$

Power law phenomena

Earthquake, fluorescence intermittency of nanocrystals,
motion of bacteria, chaotic dynamics, finance

Theorem and Conjecture

Finite measure case ($B < 2$)

$$\frac{2}{\Gamma(1-\alpha)n^{1/\alpha}} \sum_{k=0}^{n-1} f \circ T^k \rightarrow G_\alpha \quad n \rightarrow \infty,$$

where $f(x) = x^{-\beta}$ ($\beta \geq 2 - B$) and $\alpha = \beta/(2 - B)$.

Infinite measure case ($B \geq 2$)

$$\frac{1}{b_n} \sum_{k=0}^{n-1} f \circ T^k \rightarrow G_{1/\gamma} \quad n \rightarrow \infty,$$

where $b_n \propto n^\gamma$ and $\gamma = \frac{\beta}{B-1} + 1$.

$$x^\beta f(x) = O(1), \quad x \rightarrow 0, \quad (1-x)^\alpha f(x) = O(1) \quad x \rightarrow 1.$$

Finite Measure Case ($B < 2$)

Invariant density

$$\rho(x) = \frac{2-B}{2} \{x^{1-B} + (1-x)^{1-B}\}.$$

$$\{\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_n\}$$

Birkhoff's ergodic theorem tells us that the probability density function of the sequence $\{Tx, T^2x, \dots, T^nx\}$ obeys the invariant density as $n \rightarrow \infty$. The distribution of $\mathbf{Y} = f(\mathbf{X})$ is given by

$$\Pr(\mathbf{Y} < x) = \Pr(\mathbf{X} > x^{-1/\alpha}) = 1 - \frac{1}{2}x^{-\frac{2-B}{\alpha}}(1 - x^{-\frac{B-1}{\alpha}}).$$

Therefore

$$1 - \Pr(\mathbf{Y} < x) \sim \frac{1}{2}x^{-\frac{2-B}{\alpha}} \quad x \rightarrow \infty.$$

Infinite Measure Case ($B \geq 2$)

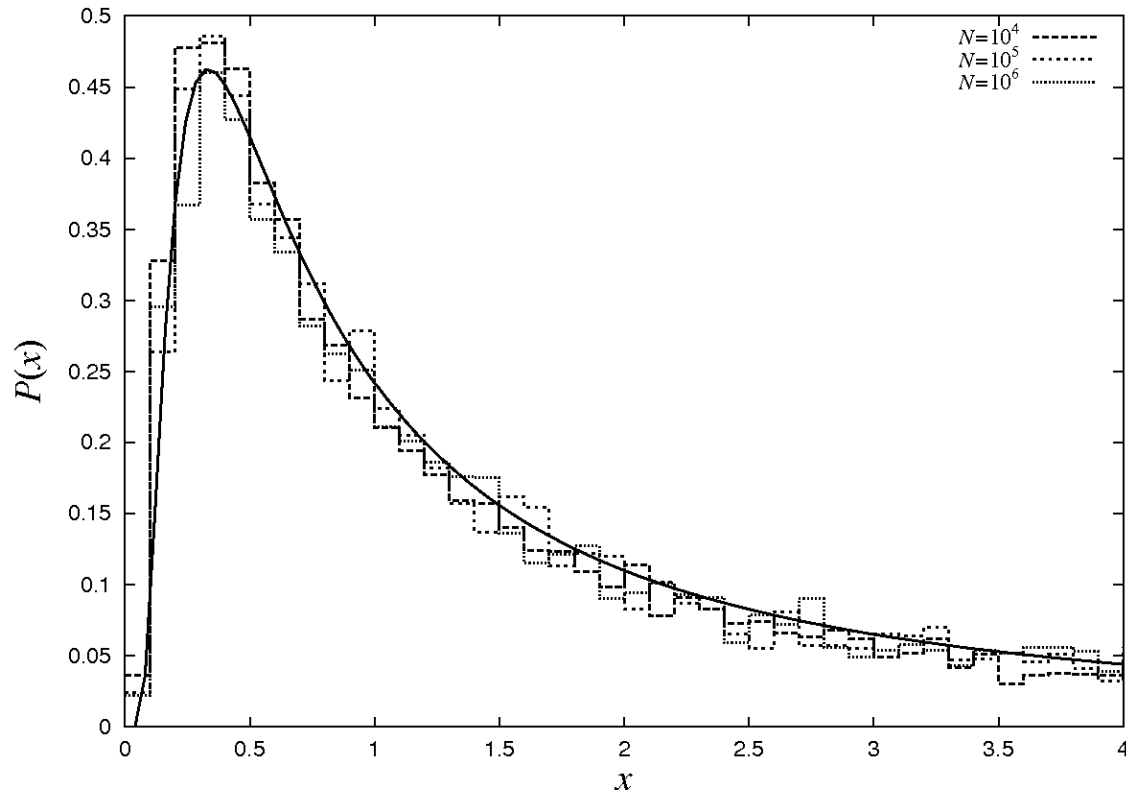


Fig. 1. The probability density function for the scaled time average of $f(x)$ ($B = 3.0, \beta = 2.0$). The fitting curve is a stable distribution with $\gamma = 2.0$.

Convergence to the invariant density

THEOREM Let $T : [0, 1] \rightarrow [0, 1]$ satisfy the conditions (i)-(iv) with return index α . Then, for all Riemann-integrable functions u on $[0, 1]$,

$$w_n(T)P^n u \rightarrow \left(\frac{1}{\Gamma(\alpha)\Gamma(2-\alpha)} \int_0^1 u d\lambda \right) h$$

uniformly on compact subsets of $(0, 1]$.

$$w_n \sim \begin{cases} \log n, & \alpha = 1 \\ n^{1-\alpha}, & \alpha < 1 \end{cases}$$

Lasota-Yorke map

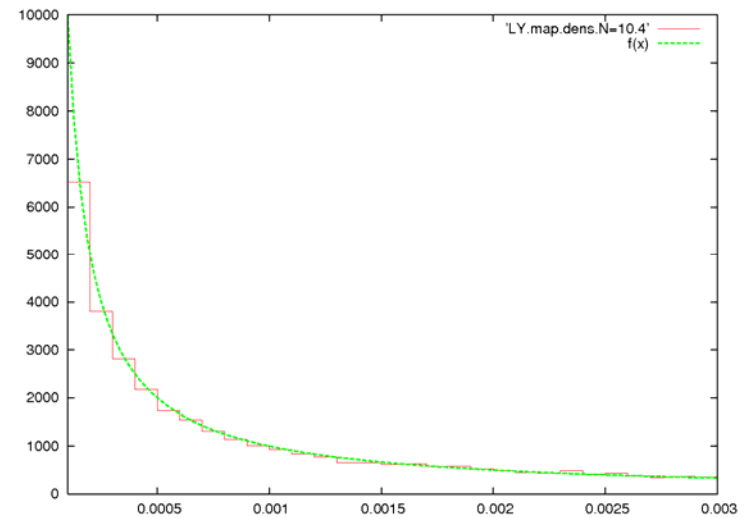
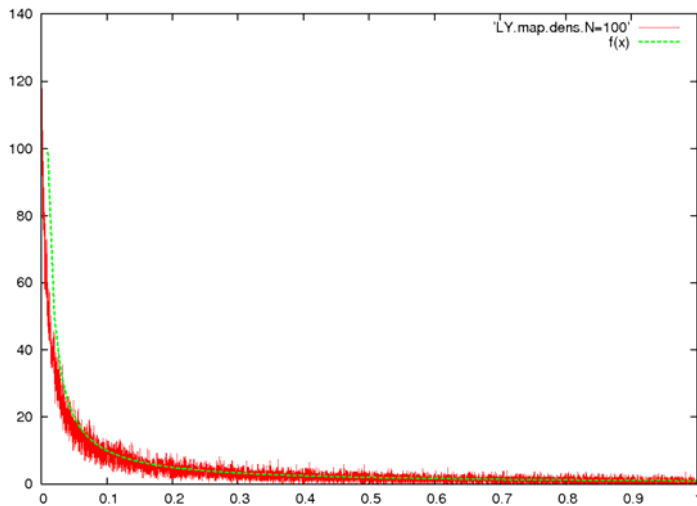
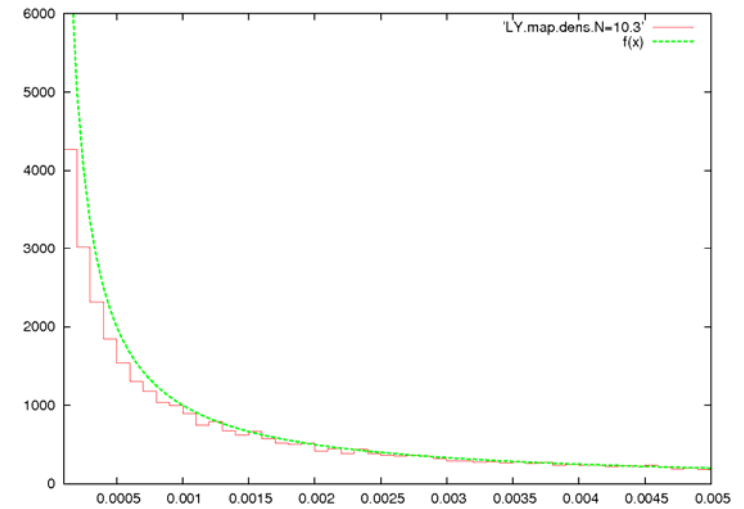
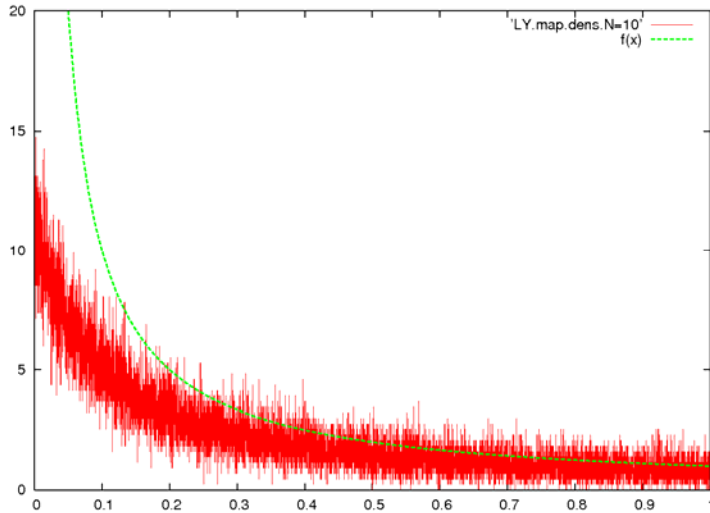
$$T(x) = \begin{cases} x/(1-x), & x \in [0, 1/2), \\ 2x-1, & x \in [1/2, 1], \end{cases}$$

which has invariant density $h(x) = 1/x$.

$$(\log n)P^n u \rightarrow \left(\int_0^1 u d\lambda \right) h \quad n \rightarrow \infty$$

Numerical Simulations

$(\log n)P^n u(x)$



x

x

The Scaling Exponent

Assumption $P^n u \rightarrow \frac{1}{2}\delta(x) + \frac{1}{2}\delta(1-x)$

Let P be the Perron-Frobenius operator. (Evolution of the density) For $u \in L^1(0,1)$ and $\int_0^1 u(x)dx = 1$

$$P^n u = \begin{cases} 1/2\epsilon_n & (x < \epsilon_n) \\ 0 & (\epsilon \leq x \leq 1 - \epsilon_n) \\ 1/2\epsilon_n & (1 - \epsilon_n < x) \end{cases} \quad \text{and } \epsilon \propto n^{-\frac{1}{B-1}}.$$

We have $\langle f \circ T^n \rangle = \int_0^1 f(x) P^n u(x) dx = O(\epsilon_n^{-\beta}) = O(n^{\frac{\beta}{B-1}})$.

Therefore $\sum_{k=0}^{n-1} \langle f \circ T^k \rangle \propto n^{\frac{\beta}{B-1}+1}$

Concluding Remarks

In infinite measure dynamical systems the time average of some observation functions converges in distribution (**Generalized Arcsine Law, Stable Law**).

$f(x)$	invariant measure	γ	distribution of X
L_m^1	finite	1	<i>delta</i> (Birkhoff, 1931)
L_m^1	infinite	$\frac{1}{B-1}$	<i>Mittag-Leffler</i> (Aaronson, 1981)
$\text{non-}L_m^1$	finite	$\frac{\alpha}{2-B}$	<i>Stable</i> (Akimoto [1])
$L_{\text{loc}}^1(m, (0, 1))$	infinite	1	<i>Generalized Arcsine</i> (Akimoto [1])
$\text{non-}L_m^1$	infinite	$\frac{\alpha}{B-1} + 1$	<i>Stable</i> (Akimoto [1])

Non-stationary time series
(Fluorescence of nanocrystals)

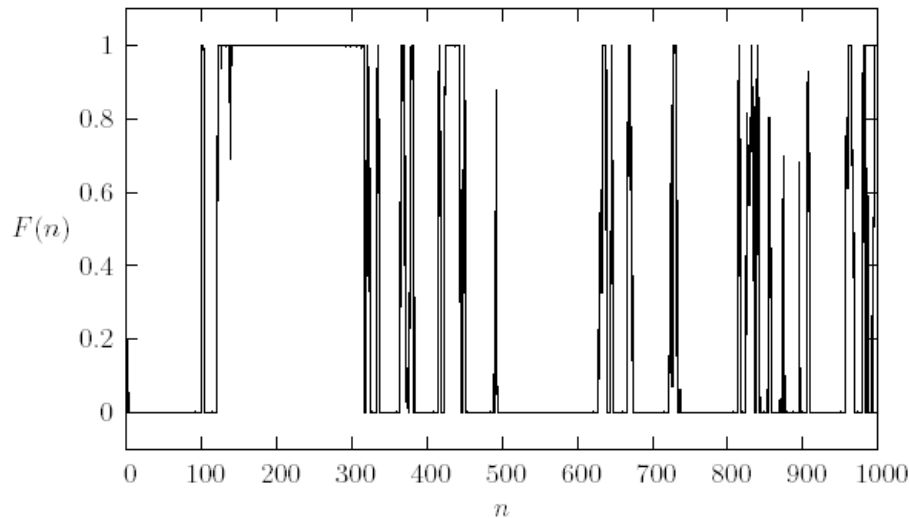


Stationary random variables

Time average

Concluding Remarks

Ergodicity of non-equilibrium state in dynamical system is related to infinite measure systems.



Macroscopic observable $F(n)$ is random.

$$\left(F(n) = \frac{1}{N} \sum_{k=n-N+1}^n f(T^k x) \right)$$