

Some sufficient condition for the ergodicity of the Lévy transform

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Lévy transformation of the path space

- ▶ β is a Brownian motion

$$\mathbf{T}\beta = \int \text{sign}(\beta_s) d\beta_s = |\beta| - L^0(\beta).$$

- ▶ \mathbf{T} is a transformation of the path space.
- ▶ \mathbf{T} preserves the Wiener measure.
- ▶ Is \mathbf{T} ergodic?
- ▶ A deep result of Marc Malric claims that the Lévy transform is **topologically recurrent**, i.e., on an almost sure event

$$\{\mathbf{T}^n\beta : n \geq 0\} \cap G \neq \emptyset, \quad \text{for all nonempty open } G \subset C[0, \infty).$$

- ▶ We use only a weaker form, also due to Marc Malric, the **density of zeros of iterated paths**, i.e.:

$$\bigcup_{n=0}^{\infty} \{t > 0 : (\mathbf{T}^n\beta)_t = 0\} \text{ is dense in } [0, \infty).$$

Ergodicity and Strong mixing (reminder)

$$T : \Omega \rightarrow \Omega, \quad \mathbf{P} \circ T^{-1} = \mathbf{P}$$

- ▶ T is ergodic, if

- ▶ $\frac{1}{N} \sum_{n=0}^{N-1} \mathbf{P}(A \cap T^{-n}B) \rightarrow \mathbf{P}(A)\mathbf{P}(B),$ for all $A, B.$

- ▶ or, $\frac{1}{N} \sum_{n=0}^{N-1} X \circ T^n \rightarrow \mathbf{E}(X),$ for each r.v. $X \in L^1.$

- ▶ or, the invariant σ -field, is trivial.

- ▶ T is strongly mixing if $\mathbf{P}(A \cap T^{-n}B) \rightarrow \mathbf{P}(A)\mathbf{P}(B),$ for all $A, B.$

Ergodicity and weak convergence

In our case $\Omega = C[0, \infty)$ is a polish space (complete, separable, metric space).

Theorem

Ω polish, T is a measure preserving transform of $(\Omega, \mathcal{B}(\Omega), \mathbf{P})$. Then

- ▶ T is *ergodic* iff $\frac{1}{n} \sum_{k=0}^{n-1} \mathbf{P} \circ (T^0, T^k)^{-1} \xrightarrow{w} \mathbf{P} \otimes \mathbf{P}$ as $n \rightarrow \infty$.
- ▶ T is *strongly mixing* iff $\mathbf{P} \circ (T^0, T^n)^{-1} \xrightarrow{w} \mathbf{P} \otimes \mathbf{P}$ as $n \rightarrow \infty$.

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▶ T is **strongly mixing** iff $\mathbf{P} \circ (T^0, T^n)^{-1} \xrightarrow{w} \mathbf{P} \otimes \mathbf{P}$ as $n \rightarrow \infty$.

Note that both families of measures are **tight**:

$$\left\{ \mathbf{P} \circ (T^0, T^n)^{-1} : n \geq 0 \right\} \quad \text{and} \quad \left\{ \frac{1}{n} \sum_{k=0}^{n-1} \mathbf{P} \circ (T^0, T^k)^{-1} : n \geq 0 \right\}$$

If $C \subset \Omega$ compact, with $\mathbf{P}(\Omega \setminus C) < \varepsilon$ then

$$\mathbf{P} \left((T^0, T^k) \notin C \times C \right) \leq \mathbf{P} \left(T^0 \notin C \right) + \mathbf{P} \left(T^k \notin C \right) < 2\varepsilon.$$

Convergence of finite dim. marginals ($\xrightarrow{f.d.}$) is enough

Some notations:

- ▶ β is the canonical process on $\Omega = C[0, \infty)$,
- ▶ $h : [0, \infty) \times C[0, \infty)$ progressive, $|h| = 1$ $dt \otimes d\mathbf{P}$ a.e.

$$T : \Omega \rightarrow \Omega, \quad T\beta = \int_0^\cdot h(s, \beta) d\beta_s, \quad (\text{e.g. } h(s, \beta) = \text{sign}(\beta_s)).$$

- ▶ $\beta^{(n)} = T^n \beta$ is the n -th iterated path.
- ▶ $h^{(0)} = 1$, $h_s^{(n)} = \prod_{k=0}^{n-1} h(s, \beta^{(k)})$ for $n > 0$, so $\beta_t^{(n)} = \int_0^t h_s^{(n)} d\beta_s$.

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Then

- ▶ The distribution of $(\beta, \beta^{(n)})$ is $\mathbf{P} \circ (T^0, T^n)^{-1}$
- ▶ Let κ_n is uniform on $\{0, 1, \dots, n-1\}$ and independent of β .
The law of $(\beta, \beta^{(\kappa_n)})$ is $\frac{1}{n} \sum_{k=0}^{n-1} \mathbf{P} \circ (T^0, T^k)^{-1}$.
- ▶ T is strongly mixing, iff $(\beta, \beta^{(n)}) \xrightarrow{f.d.} \text{BM-2}$.
- ▶ Similarly, T is ergodic, iff $(\beta, \beta^{(\kappa_n)}) \xrightarrow{f.d.} \text{BM-2}$.
- ▶ Reason: Tightness + $\xrightarrow{f.d.}$ = convergence in law.

Characteristic function

- ▶ Fix $t_1, \dots, t_r \geq 0$ and $\alpha = (a_1, \dots, a_r, b_1, \dots, b_r) \in \mathbb{R}^{2r}$
- ▶ The characteristic function of $(\beta_{t_1}, \dots, \beta_{t_r}, \beta_{t_1}^{(n)}, \dots, \beta_{t_r}^{(n)})$ at α is

$$\varphi_n = \mathbf{E} \left(e^{i \left(\int f(s) d\beta_s + \int g(s) d\beta_s^{(n)} \right)} \right) = \mathbf{E} \left(e^{i \int (f(s) + g(s) h_s^{(n)}) d\beta_s} \right),$$

where $f = \sum_{j=1}^r a_j \mathbb{1}_{[0, t_j]}$, $g = \sum_{j=1}^r b_j \mathbb{1}_{[0, t_j]}$.

- ▶ Finite dim. marginals has the right limit, if for all choices $r \geq 1$, $\alpha \in \mathbb{R}^{2r}$, $t_1, \dots, t_r \geq 0$

$$\varphi_n \rightarrow \exp \left\{ -\frac{1}{2} \int f^2 + g^2 \right\} \quad \text{for strong mixing,}$$

$$\frac{1}{n} \sum_{k=0}^{n-1} \varphi_k \rightarrow \exp \left\{ -\frac{1}{2} \int f^2 + g^2 \right\} \quad \text{for ergodicity.}$$

Estimate for $|\varphi_n - \varphi|$, where $\varphi = e^{-\frac{1}{2} \int f^2 + g^2}$

- ▶ $M_t = \int_0^t (f(s) + h_s^{(n)} g(s)) d\beta_s$.
- ▶ M is a closed martingale and so is $Z = \exp \left\{ iM + \frac{1}{2} \langle M \rangle \right\}$.
- ▶ $Z_0 = 1 \implies \mathbf{E}(Z_\infty) = 1$.
- ▶ $\langle M \rangle_\infty = \int_0^\infty f^2(s) + g^2(s) ds + 2 \int_0^\infty h_s^{(n)} f(s) g(s) ds$
- ▶

$$\varphi = \varphi \mathbf{E}(Z_\infty) = \mathbf{E} \exp \left\{ i \int_0^\infty (f d\beta + g d\beta^{(n)}) + \int_0^\infty f g h^{(n)} \right\}$$

- ▶ Recall that $fg = \sum_j a_j b_j \mathbb{1}_{[0, t_j]}$. Then with $X_n(t) = \int_0^t h_s^{(n)} ds$

$$|\varphi_n - \varphi| \leq \mathbf{E} \left| 1 - e^{\int_0^\infty f g h^{(n)}} \right| \leq e^{\int |fg|} \mathbf{E} \left| \int_0^\infty f g h^{(n)} \right| \leq C \sum_{j=1}^r \mathbf{E} |X_n(t_j)|,$$

where $C = C(f, g) = C(\alpha, t_1, \dots, t_r)$ does not depend on n .

$X_n(t) = \int_0^t h_s^{(n)} ds \xrightarrow{P} 0$ for all $t \geq 0$ would be enough

Theorem

1. If $X_n(t) \xrightarrow{P} 0$ for all $t \geq 0$, then T is strongly mixing.
2. T is ergodic **if and only if** $\frac{1}{n} \sum_{k=0}^{n-1} X_k^2(t) \xrightarrow{P} 0$ for all $t \geq 0$.

Strong mixing:

- ▶ The only missing part is the convergence of finite dimensional marginals.
- ▶ If $X_n(t) \xrightarrow{P} 0$ then $\mathbf{E} |X_n(t)| \rightarrow 0$ since $|X_n(t)| \leq t$.
- ▶ Then $|\varphi_n - \varphi| \leq C \sum_j \mathbf{E} |X_n(t_j)| \rightarrow 0 \implies (\beta, \beta^{(n)}) \xrightarrow{f.d.} \text{BM-2}$.

Remember, that:

- ▶ $(\beta, \beta^{(n)}) \xrightarrow{f.d.} \text{BM-2}$ + tightness gives: $(\beta, \beta^{(n)}) \xrightarrow{D} \text{BM-2}$.
- ▶ $(\beta, \beta^{(n)}) \xrightarrow{D} \text{BM-2} \Leftrightarrow T$ strong mixing.

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Ergodicity. \Leftarrow .

- ▶ By Cauchy-Schwarz and $|X_k(t)| \leq t$

$$\mathbf{E} \left(\frac{1}{n} \sum_{k=0}^{n-1} |X_k(t)| \right) \leq \mathbf{E}^{1/2} \left(\frac{1}{n} \sum_{k=0}^{n-1} X_k^2(t) \right) \rightarrow 0.$$

- ▶ Then $\left| \frac{1}{n} \sum_{k=0}^{n-1} \varphi_k - \varphi \right| \leq C \sum_j \mathbf{E} \frac{1}{n} \sum_{k=0}^{n-1} |X_k(t_j)| \rightarrow 0 \implies (\beta, \beta^{(\kappa_n)}) \xrightarrow{f.d.} \text{BM-2}$.

Remember that

- ▶ κ_n is uniform on $\{0, \dots, n-1\}$ and independent of β
- ▶ $(\beta, \beta^{(\kappa_n)}) \xrightarrow{f.d.} \text{BM-2}$ + tightness gives: $(\beta, \beta^{(\kappa_n)}) \xrightarrow{D} \text{BM-2}$.
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Ergodicity. \Rightarrow (outline of the proof)

- Fix $0 < s < t$. Then the following limits exist a.s and in L^2 :

$$Z_u = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} h_s^{(k)} h_u^{(k)} \quad \text{for } s \leq u \leq t, \quad Z = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} h_s^{(k)} (\beta_t^{(k)} - \beta_s^{(k)})$$

moreover $|Z|$ and $|Z_u|$ are invariant for T , hence they are non-random.

- Then $h_s^{(k)} (\beta_t^{(k)} - \beta_s^{(k)}) = \int_s^t h_s^{(k)} h_u^{(k)} d\beta_u$ and

$$Z = \int_s^t Z_u d\beta_u = \int_s^t |Z_u| d\tilde{\beta}_u, \quad \text{where } \tilde{\beta} = \int_s^t \text{sign}(Z_u) d\beta_u.$$

- $Z \sim N(0, \sigma^2)$ since $|Z_u|$ is non-random. But $|Z|$ is also non-random.
 $\implies Z = 0. \implies Z_u = 0. \implies \frac{1}{n} \sum_{k=0}^{n-1} X_k^2(t) \rightarrow 0.$

A variant of the mean ergodic theorem

- ▶ T is a measure preserving transformation of Ω ,
- ▶ ε_0 is r.v. taking values in $\{-1, +1\}$, $\varepsilon_k = \varepsilon_0 \circ T^k$.
- ▶ For $\xi \in L^2(\Omega)$, $U\xi = \xi \circ T\varepsilon_0$ is an isometry.
- ▶ von Neumann's mean ergodic theorem says, that

$$\frac{1}{n} \sum_{k=0}^{n-1} U^k \xi \rightarrow P\xi, \in L^2$$

where P is the projection onto

$$\{X \in L^2 : X \circ T\varepsilon_0 = UX = X\}$$

- ▶ $|P\xi|$ is invariant under T .
- ▶ what is $U^k \xi$?

$$U\xi = \xi \circ T\varepsilon_0, \quad U^2\xi = \xi \circ T^2\varepsilon_1\varepsilon_0, \quad \dots \quad U^k\xi = \xi \circ T^k \prod_{j=0}^{k-1} \varepsilon_j,$$

- ▶ Almost sure convergence also holds by the subadditive ergodic theorem.

Lévy transformation

- ▶ The Lévy transformation \mathbf{T} is scaling invariant, that is, if for $x > 0$ $\Theta_x : C[0, \infty) \rightarrow C[0, \infty)$ denotes $\Theta_x(w)(t) = xw(t/x^2)$ then

$$\Theta_x \mathbf{T} = \mathbf{T} \Theta_x$$

- ▶ As before $\beta^{(n)} = \beta \circ \mathbf{T}^n$, $h_t^{(n)} = \prod_{k=0}^{n-1} \text{sign}(\beta_t^{(k)})$, $X_n(t) = \int_0^t h_s^{(n)} ds$.

By scaling we get:

Theorem

1. If $X_n(1) \xrightarrow{P} 0$ as $n \rightarrow \infty$, then \mathbf{T} is strongly mixing.
2. \mathbf{T} is ergodic, if and only if $\frac{1}{n} \sum_{k=0}^{n-1} X_k^2(1) \xrightarrow{P} 0$ as $n \rightarrow \infty$.

Behaviour of $X_n(1)$

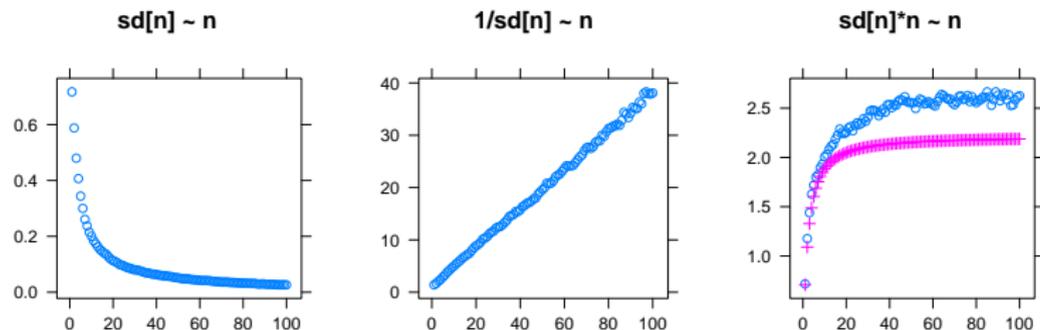


Figure: $sd^2[n] = \mathbf{E}(X_n^2(1))$, sample size: 2000, number of steps of SRW: 10^6 . Pink crosses denotes $n \times \mathbf{E}^{1/2}(\tilde{X}_n^2)$, where $\tilde{X}_n = \int_0^1 \prod_{k=0}^{n-1} \text{sign}(\tilde{\beta}_s^{(k)}) ds$ with independent BM-s $(\tilde{\beta}^{(k)})_{k \geq 0}$. We have $\mathbf{E}^{1/2}(\tilde{X}_n^2) \approx \frac{\pi}{n\sqrt{2}} \approx \frac{2.22}{n}$.

Conjecture

$\mathbf{E}(X_n^2(1)) = O(1/n^2)$. This would give: $X_n(1) = \int_0^1 h_s^{(n)} ds \rightarrow 0$ almost surely.

Simplification

▶ Goal: $X_n(1) = \int_0^1 h_s^{(n)} ds \xrightarrow{P} 0$.

▶ Enough: $X_n(1) \rightarrow 0$ in L^2 .

▶

$$\mathbf{E} \left(X_n^2(1) \right) = 2 \int_{0 < u < v < 1} \mathbf{E} \left(h_u^{(n)} h_v^{(n)} \right) dudv = 2 \int_{0 < u < v < 1} \text{cov} \left(h_u^{(n)}, h_v^{(n)} \right) dudv.$$

▶ Enough:

$$\mathbf{E} \left(h_s^{(n)} h_1^{(n)} \right) \rightarrow 0, \quad \text{for } 0 < s < 1,$$

by boundedness and scaling.

▶ New goal: fixing $s \in (0, 1)$,

$$\mathbf{P} \left(h_s^{(n)} h_1^{(n)} = 1 \right) \approx \mathbf{P} \left(h_s^{(n)} h_1^{(n)} = -1 \right), \quad \text{for } n \text{ large.}$$

Idea: **coupling**.

Coupling I.

- ▶ Assume that $S : C[0, \infty) \rightarrow C[0, \infty)$ preserves \mathbf{P} .
- ▶ Denote by $\tilde{\beta}^{(n)}$ the shadow path $\beta^{(n)} \circ S$.
- ▶ Assume also that there is an event A such that on A the sequences

$$\text{sign}(\beta_s^{(n)}) \text{sign}(\beta_1^{(n)}) \quad \text{and} \quad \text{sign}(\tilde{\beta}_s^{(n)}) \text{sign}(\tilde{\beta}_1^{(n)})$$

differ at exactly one index denoted by ν .

Then

$$\limsup_{n \rightarrow \infty} \left| \mathbf{E} \left(h_s^{(n)} h_1^{(n)} \right) \right| \leq \mathbf{P}(A^c).$$

Reason:

$$\left| \mathbf{E} h_s^{(n)} h_1^{(n)} \right| = \left| \mathbf{E} \left(\frac{h_s^{(n)} h_1^{(n)} + \tilde{h}_s^{(n)} \tilde{h}_1^{(n)}}{2} \right) \right| \leq \mathbf{P}(A^c) + \mathbf{P}(n < \nu).$$

Coupling.

Proposition

If there is a stopping time τ , s.t.

- ▶ $s < \tau < 1$,
- ▶ exists $\nu < \infty$, s.t. $\beta_\tau^{(\nu)} = 0$,
- ▶ $\min_{0 \leq k < \nu} |\beta_\tau^{(k)}| > C\sqrt{1 - \tau}$,

$$\implies \limsup_n \left| \mathbf{E} h_s^{(n)} h_1^{(n)} \right| \leq \mathbf{P} \left(\max_{s \in [0,1]} |\beta_s| > C \right).$$

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- ▶ S reflects β after τ :

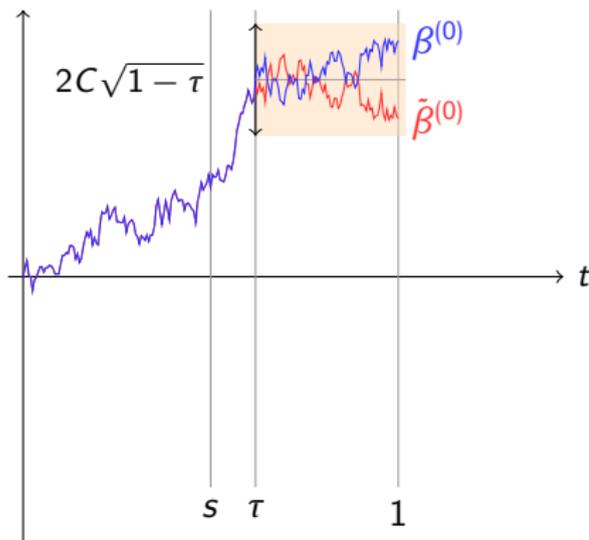
$$(S\beta)_t = \tilde{\beta}_t = \beta_{t \wedge \tau} - (\beta_t - \beta_{t \wedge \tau}).$$

- ▶ $A = \left\{ \max_{t \in [\tau, 1]} |\beta_t^{(0)} - \beta_\tau^{(0)}| \leq C\sqrt{1-\tau} \right\}$.

- ▶ Then

$$\mathbf{P}(A^c) = \mathbf{P} \left(\max_{s \in [0,1]} |\beta_s| > C \right)$$

by strong Markov property and scaling.



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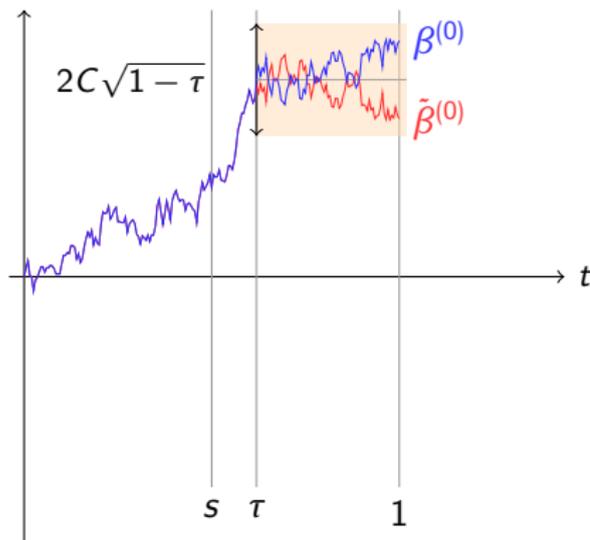
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- ▶ We need that

$$\text{sign}(\beta_s^{(n)} \beta_1^{(n)}) \text{ differs from } \text{sign}(\tilde{\beta}_s^{(n)} \tilde{\beta}_1^{(n)})$$

at exactly one place, when $n = \nu$.

- ▶ Recall that $\mathbf{T}\beta = |\beta| - L$.



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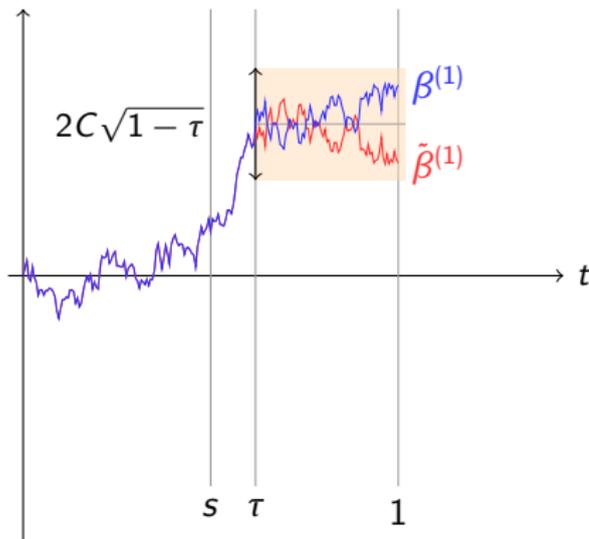
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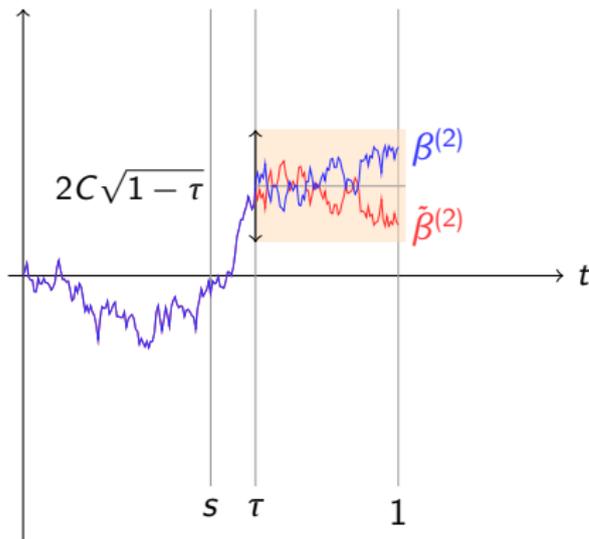
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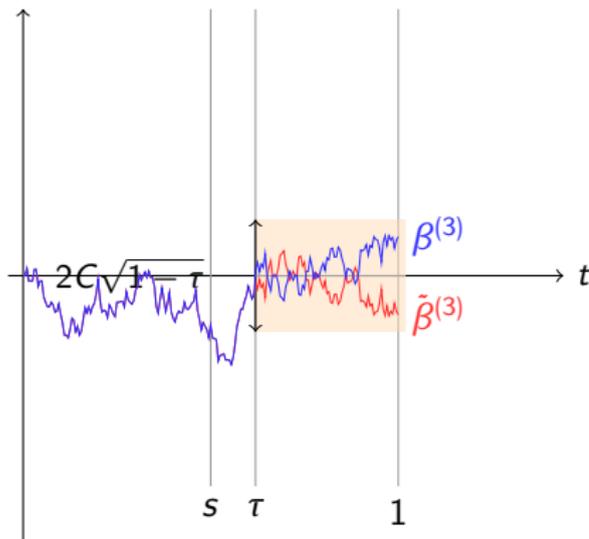
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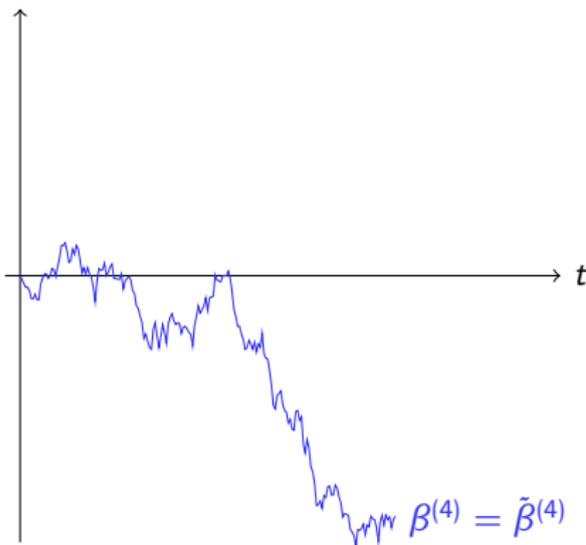
- ▶ $A = \left\{ \max_{t \in [\tau, 1]} |\beta_t^{(0)} - \beta_\tau^{(0)}| \leq C\sqrt{1-\tau} \right\}$.

- ▶ We need that

$$\text{sign}(\beta_s^{(n)} \beta_1^{(n)}) \text{ differs from } \text{sign}(\tilde{\beta}_s^{(n)} \tilde{\beta}_1^{(n)})$$

at exactly one place, when $n = \nu$.

- ▶ Recall that $\mathbf{T}\beta = |\beta| - L$.



Simplification

Proposition

If there is a *random time* τ , s.t.

1. $s < \tau < 1$,
2. exists $\nu < \infty$, s.t. $\beta_\tau^{(\nu)} = 0$,
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$$\tau_n = \inf \left\{ t \geq s : \beta_t^{(n)} = 0, \min_{0 \leq k < n} |\beta_t^{(k)}| \geq C\sqrt{(1-t) \vee 0} \right\}, \quad \tilde{\tau} = \inf_n \tau_n.$$

- ▶ $\tau_n, \tilde{\tau}$ are stopping times.
- ▶ By the condition $s \leq \tilde{\tau} < 1$.
- ▶ If for some $\omega \in \Omega$, $\tilde{\tau}(\omega) < \tau_n(\omega)$ for all n then by continuity

$$\inf_{n \geq 0} |\beta_{\tilde{\tau}}^{(n)}| \geq C\sqrt{1 - \tilde{\tau}} > 0 \quad \text{at } \omega.$$

- ▶ This can only happen with probability zero due to **Malric's density theorem!!**

Good time points

Definition

For $s \in (0, 1)$, $C > 0$

$$A(C, s) = \left\{ t \geq 0 : \exists \gamma, n, s \cdot t < \gamma < t, \beta_\gamma^{(n)} = 0, \right. \\ \left. \min_{0 \leq k < n} |\beta_\gamma^{(k)}| > C\sqrt{t - \gamma} \right\}$$

is the set of **good** time points.

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Goal:

$$\mathbf{P}(1 \in A(C, s)) = 1, \quad \text{for all } C > 0, s \in (0, 1).$$

Set of good points II.

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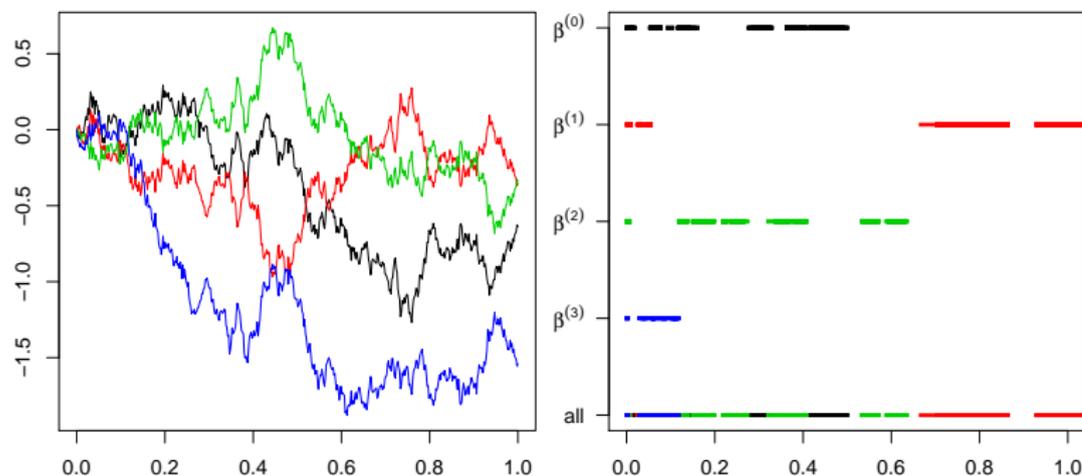
- ▶ $\mathbf{P}(t \in A(C, s))$ does not depend on t .
 - ▶ $\mathbf{P}(1 \in A(C, s)) = 1 \Leftrightarrow A(C, s)$ has full Lebesgue measure almost surely
- Proof: Let Z exponential independent of $\beta^{(0)}$. Then

$$1 = \mathbf{P}(Z \in A(C, s)) = \int_0^\infty \mathbf{P}(t \in A(C, s)) e^{-t} dt.$$

New goal:

The random set of good time points $A(C, s)$ is of full Lebesgue measure almost surely.

Good time points, a picture $s = .9$ and $C = 2$



If γ is a zero of $\beta^{(n)}$ and $\min_{0 \leq k < n} |\beta_{\gamma}^{(k)}| = \xi > 0$ then

$$I = (\gamma, \gamma + L) \subset A(C, s), \quad \text{where} \quad L = \frac{\xi^2}{C^2} \wedge \frac{(1-s)\gamma}{s}.$$

$A(C, s)$ is a dense open set! **May have small Lebesgue measure.**

Porous sets

Definition

The set $H \subset \mathbb{R}$ is **porous** at x if

$$\limsup_{r \rightarrow 0} \frac{\text{length of the largest subinterval of } (x - r, x + r) \setminus H}{2r} > 0.$$

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- ▶ For $H = [0, \infty) \setminus A(C, s)$ the set of bad time points

$$\mathbf{P}(H \text{ is porous at } 1) = 1 \implies \forall t > 0, \mathbf{P}(H \text{ is porous at } t) = 1$$

$$\implies \mathbf{P}(H \text{ is porous at a.e. } t > 0) = 1 \implies \mathbf{P}(\lambda(H) = 0) = 1$$

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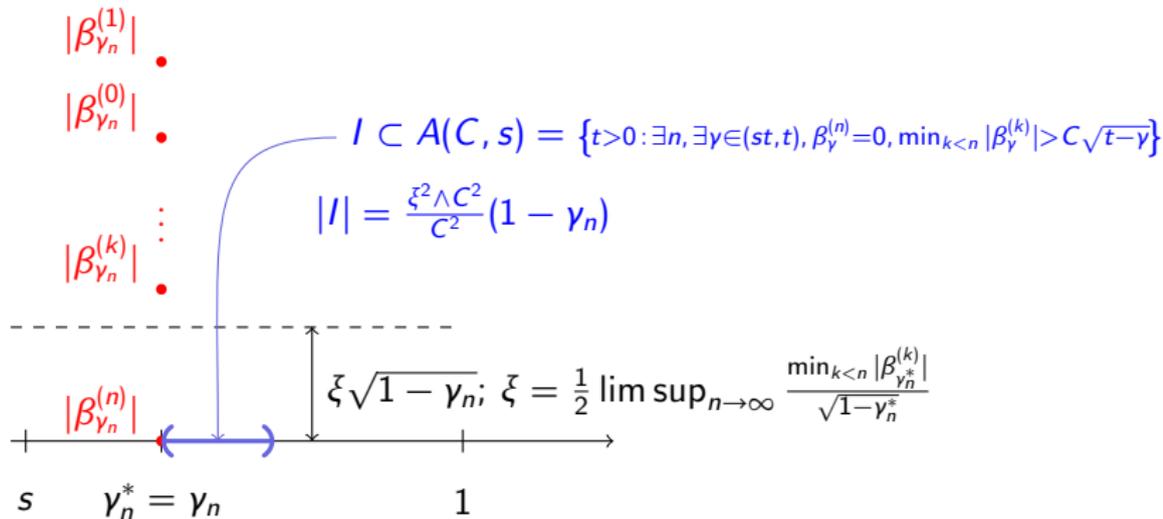
The set of good time points contains sufficiently large intervals near 1.

$\limsup_{n \rightarrow \infty} \frac{\min_{0 \leq k < n} |\beta_{\gamma_n^*}^{(k)}|}{\sqrt{1 - \gamma_n^*}} > 0$ a.s. is enough for strong mixing

► Here $\gamma_n = \sup \{ t \leq 1 : \beta_t^{(n)} = 0 \}$, $\gamma_n^* = \max_{0 \leq k \leq n} \gamma_k$.

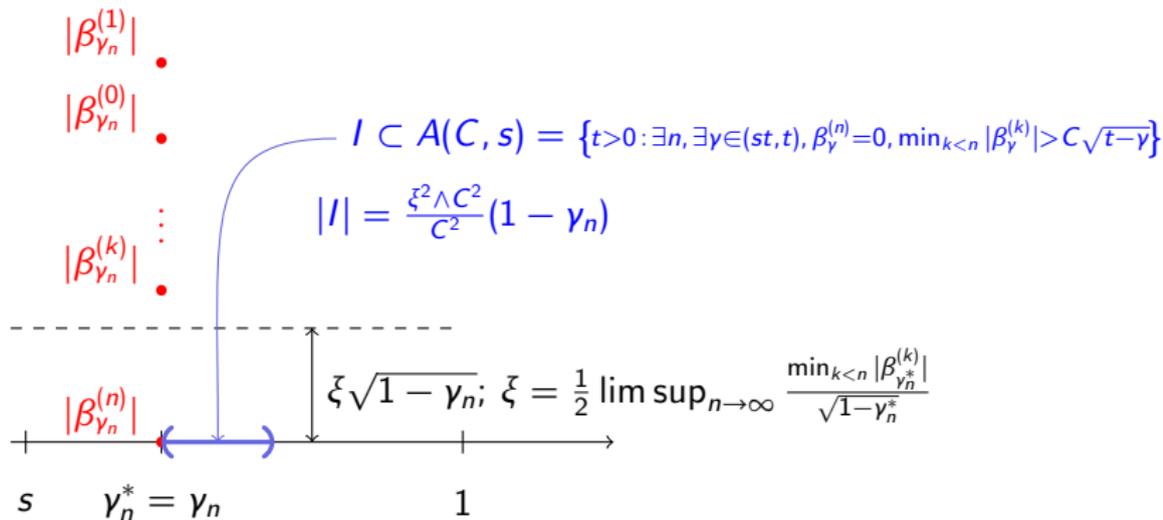
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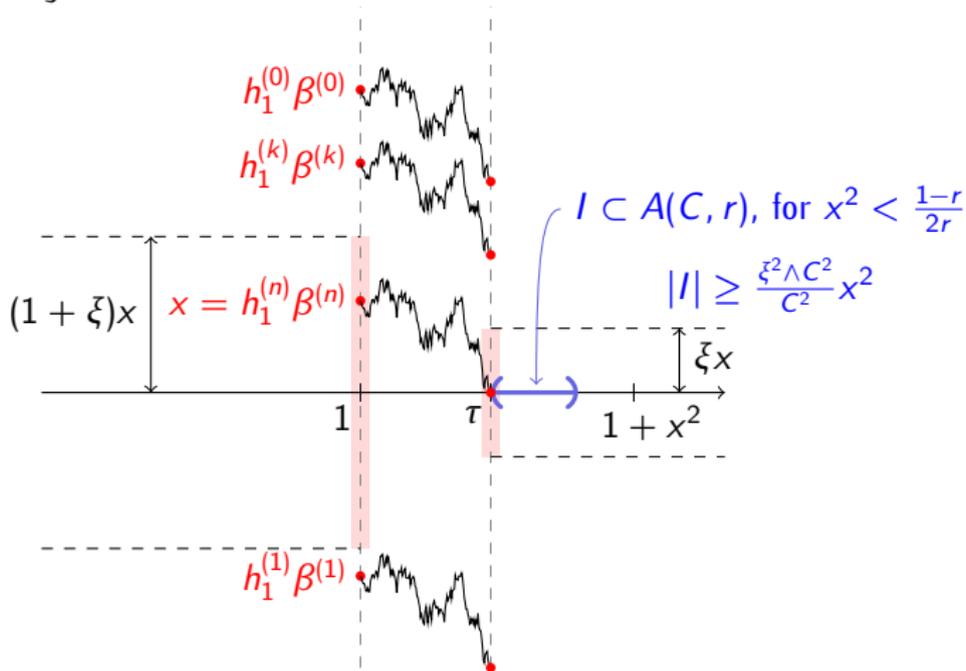
- ▶ $I \subset A(C, s)$, the length of I is proportional to $(1 - \gamma_n) = \delta'$.
- ▶ $A(C, s)$ is of full Lebesgue measure for all C, s , etc..

$\liminf_{n \rightarrow \infty} \frac{Z_{n+1}}{Z_n} < 1$ a.s. also guarantees strong mixing

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- ▶ This condition is obtained similarly, by considering the right neighborhood of 1.



Remark on $X = \liminf \frac{Z_{n+1}}{Z_n}$ and $Y = \limsup \frac{\min_{0 \leq k < n} |\beta_{\gamma_n^*}^{(k)}|}{\sqrt{1 - \gamma_n^*}}$

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Working a bit harder, one can obtain that both X and Y are invariant, and

- ▶ Either $Y = 0$ a.s.,
- ▶ or $0 < \mathbf{P}(Y = 0) < 1$ and \mathbf{T} is not ergodic,
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Remark: There is a hope that $\mathbf{P}(X = 1) = 1$ is impossible. Then

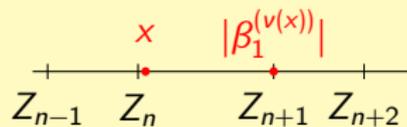
- ▶ $\mathbf{P}(X = 1) > 0 \implies X$ is not constant, hence X is a nontrivial invariant variable.
- ▶ Both X , and Y characterize ergodicity: $X < 1 \Leftrightarrow Y > 0 \Leftrightarrow \mathbf{T}$ is strongly mixing.

$$\liminf_{x \searrow 0} \frac{|\beta_1^{(v(x))}|}{x} < 1 \Leftrightarrow X = \liminf_{n \rightarrow \infty} \frac{Z_{n+1}}{Z_n}$$

▶ Here $v(x) = \inf \{ n \geq 0 : |\beta_1^{(n)}| < x \}$ and $Z_n = \min_{k \leq 0 < n} |\beta_1^{(k)}|$.

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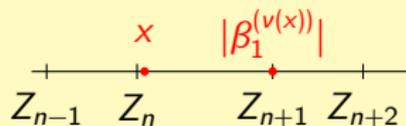


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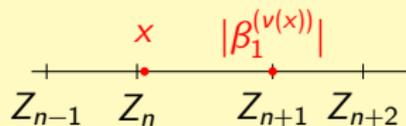
▶ **Proof:** $\mathbb{1}_{(X > 1 - \delta)} \leq \liminf \mathbb{1}_{(|\beta_1^{(v(x))}|/x > 1 - \delta)}$.

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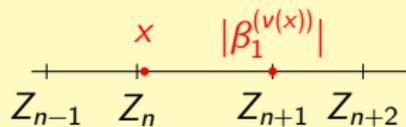
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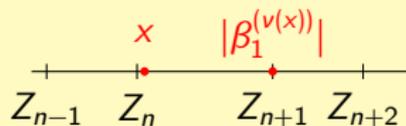
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$$\mathbf{P}(X = 1) \leq \inf_K \inf_{\delta} \mathbf{P}(xv(x) > K) + (1 + K)\delta.$$

Is $\{xv(x) : x > 0\}$ tight?

Recall that

$\sup_{x \in (0,1)} \mathbf{E}(xv(x)) < \infty \implies \{xv(x) : x \in (0,1)\}$ is tight (by Markov inequality) $\implies \mathbf{T}$ is strongly mixing.

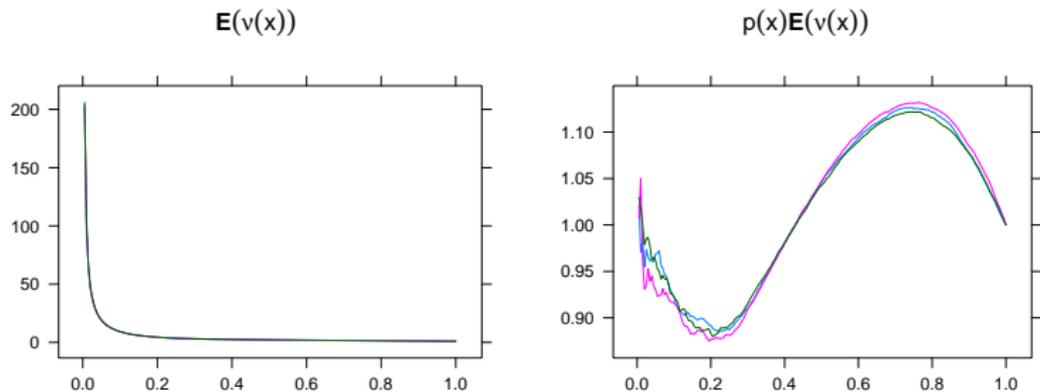


Figure: $\mathbf{E}(v^*(x))$ estimated from long runs of a SRW (number of iteration: 10^5 , number of steps of SRW: 10^9).

On the x -axis the probability $\rho(x) = \mathbf{P}(|\beta_1^{(0)}| < x)$ is given.

Density of $\frac{1}{x}|\beta_1^{(v(x))}|$

- ▶ Consider the natural extension of $(\Omega, \mathcal{B}, \mathbf{P}, T)$. Then T is an invertible measure preserving transformation on the extension.

That is

- ▶ $\Omega = \mathbb{C}[0, \infty)^{\mathbb{Z}}$,
- ▶ for $\omega = (\omega_n)_{n \in \mathbb{Z}}$ $(T\omega)_n = \omega_{n+1}$ and $\beta^{(n)}(\omega) = \omega_n$,
- ▶ \mathbf{P} is such that $\beta^{(k)}, \beta^{(k+1)}, \dots$ has the same joint law as $(\beta, \mathbf{T}^1\beta, \dots)$ for all $k \in \mathbb{Z}$.

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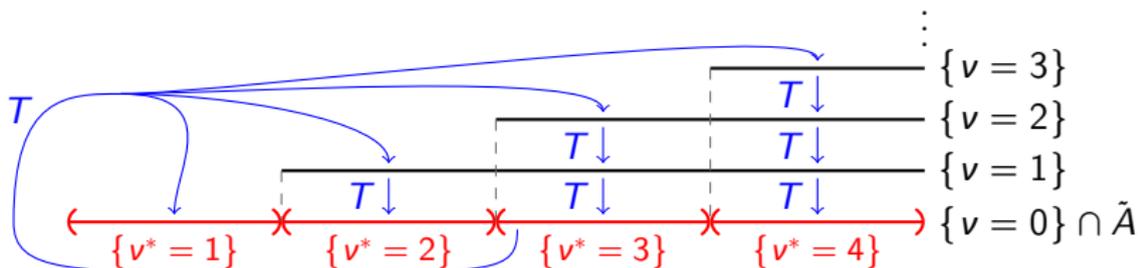
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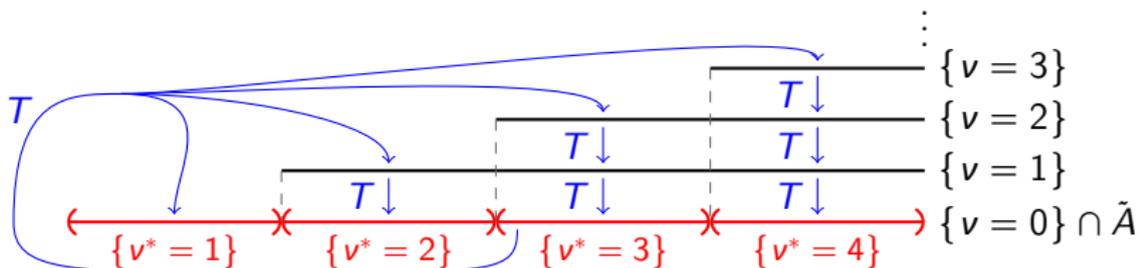
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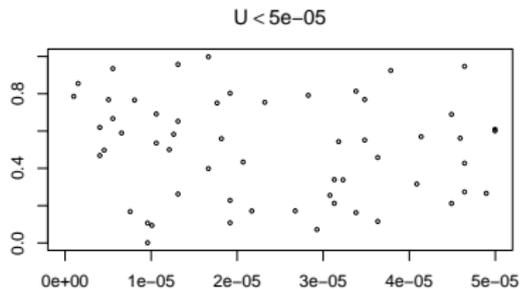
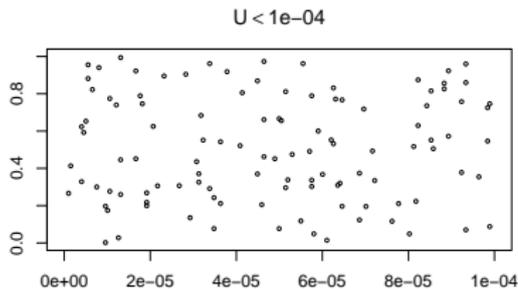
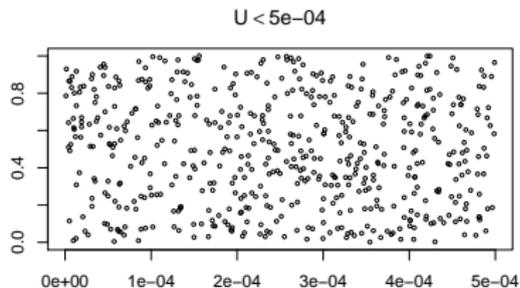
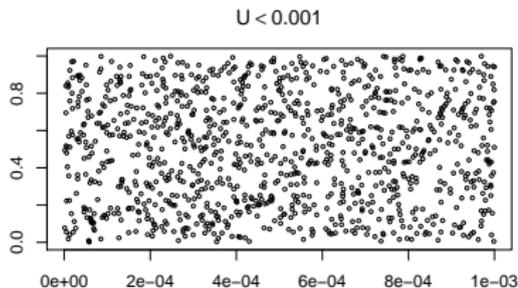
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- ▶ The density f_x of $\frac{1}{x}|\beta_1^{(v(x))}|$ is obtained by conditioning

$$f_x(y) = 2\varphi(yx) \mathbf{E} \left(xv^*(x) \mid |\beta_1^{(0)}| = yx \right), \quad \text{for } y \in (0, 1)$$

The density $\mathbf{E} \left(xv^*(x) \mid |\beta_1^{(0)}| = yx \right)$?



Joint behaviour of $|\beta_1^{(0)}|$ and $v^*(x)$ given $|\beta_1^{(0)}| < x$. Both are rescaled to uniform variables.

$\frac{1}{x}|\beta_1^{(0)}|$ seems to be conditionally independent of $xv^*(x)$,

(From one long random walk: number of steps 10^{13} , number of iterated paths 10^6 .)

$$\lim_{x \rightarrow 0^+} \mathbf{E} \left(x v^*(x) \mid |\beta_1^{(0)}| = yx \right) = ?$$

- ▶ **Conjecture:** $\frac{1}{x} |\beta_1^{(v(x))}|$ converges in distribution to a uniform variable. Actually the density seems to go to 1 as $x \rightarrow 0^+$.
- ▶ Playing with two types of expected return times one can show that

$$\liminf_{x \rightarrow 0^+} \mathbf{P} \left(|\beta_1^{(v(x))}| < x/2 \right) > 0.$$

- ▶ This is enough

$$\liminf_{x \rightarrow 0^+} \frac{|\beta_1^{(v(x))}|}{x} < 1 \quad \text{with positive probability.}$$

- ▶ Recall that then both

$$X = \liminf_{x \rightarrow 0^+} \frac{\min_{0 \leq k \leq n} |\beta_1^{(k)}|}{\min_{0 \leq k < n} |\beta_1^{(k)}|}, \quad Y = \limsup_{x \rightarrow 0^+} \frac{\min_{0 \leq k < n} |\beta_{\gamma_n^*}^{(k)}|}{\sqrt{1 - \gamma_n^*}}$$

characterize ergodicity: $X < 1 \Leftrightarrow Y > 0 \Leftrightarrow \mathbf{T}$ is strongly mixing $\Leftrightarrow \mathbf{T}$ is ergodic.

Conclusion

- ▶ Marc Malric has proved that the orbit of a typical sample path meets every open set.
- ▶ To prove strong mixing only certain open sets has to be considered.
- ▶ For these open sets
 - **Tightness** of the family rescaled hitting times would be enough.
 - or a quantitative result is needed: **the expected hitting times do not growth faster than the inverse of the size of these open sets.**

Thank you for your attention!

Happy birthday!