

Persistence of some additive functionals of Sinai's walk

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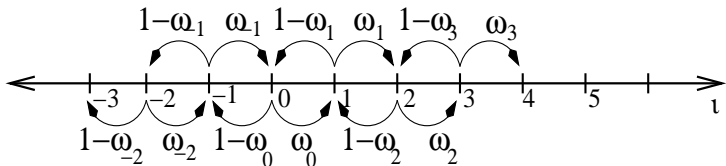
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 - Presentation of the model
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- 3 x -extrema of two-sided Brownian motion
 - Definition
 - Results of Cheliotis
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 - Upper bound
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Presentation of the model

- Random Walks in Random Environment (RWRE) introduced by Chernov (1967, biophysicist, DNA) then also physics, metallurgy
- First Step : Construction of the environment ω

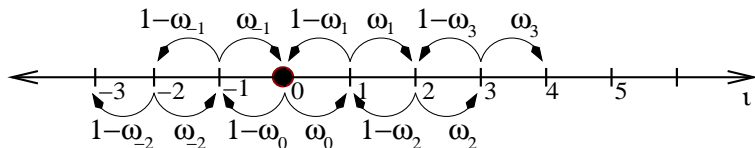
$\omega = (\omega_i)_{i \in \mathbb{Z}}$ i.i.d. $\omega_i \in [\varepsilon_0, 1 - \varepsilon_0]$, law η

$\text{Var}\left(\frac{\omega_0}{1-\omega_0}\right) := \sigma^2 > 0$ (to avoid simple random walks)



Presentation of the model

Second step : Random walk in this environment ω



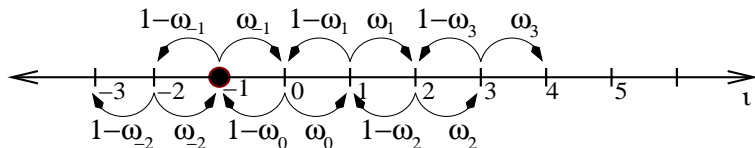
Quenched law :

$$S_0 = 0 \quad \text{and} \quad P_\omega(S_{n+1} = k | S_n = i) = \begin{cases} \omega_i & \text{if } k = i + 1, \\ 1 - \omega_i & \text{if } k = i - 1, \\ 0 & \text{else.} \end{cases}$$

Annealed law : $\mathbb{P}(\cdot) = \int P_\omega(\cdot) \eta(d\omega)$

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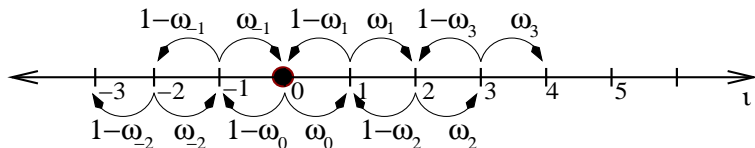
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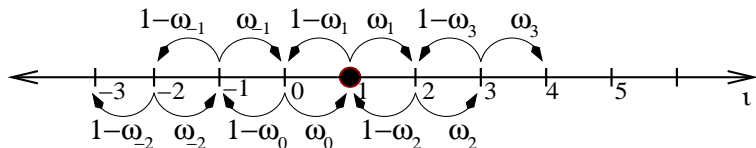
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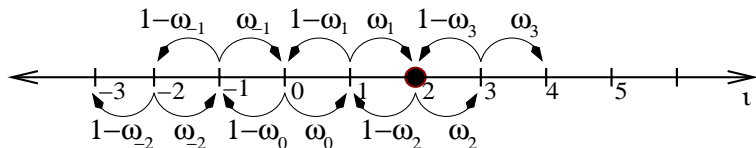
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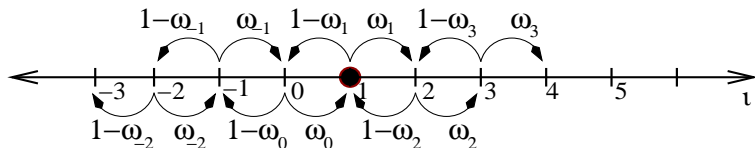
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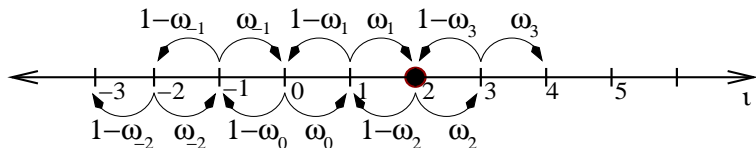
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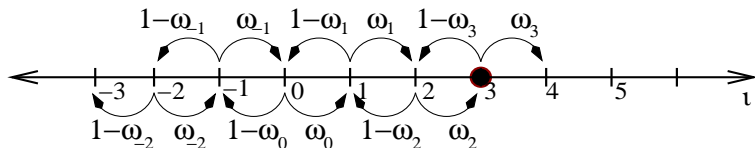
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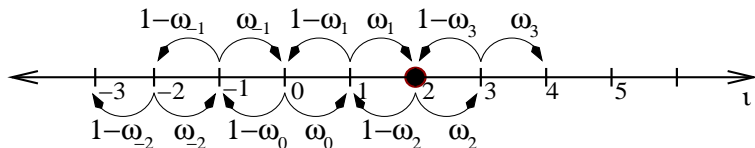
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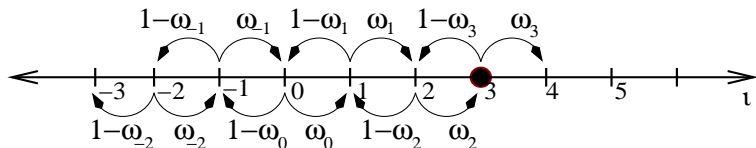
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Classical results

Recurrence criteria (Solomon, 1975)

$$(S_n)_{n \in \mathbb{N}} \text{ recurrent } \eta\text{-a.s.} \iff \mathbb{E} \log \frac{\omega_0}{1-\omega_0} = 0$$

Theorem : Sinai (1982)

In the recurrent case

$$\sigma^2 \frac{S_n}{(\log n)^2} \xrightarrow[n \rightarrow +\infty]{\text{law}} b_\infty,$$

b_∞ symmetric, non degenerate, non gaussian

Sinai walk : ellipticity ($\omega_i \in [\varepsilon_0, 1 - \varepsilon_0]$ a.s.)

$\text{Var}\left(\frac{\omega_0}{1-\omega_0}\right) := \sigma^2 > 0$ (to avoid simple random walks)

recurrent ($\mathbb{E} \log \frac{\omega_0}{1-\omega_0} = 0$)

Question

Theorem (Sinai, 1992)

For a simple random walk $(R_n)_{n \in \mathbb{N}}$, as $n \rightarrow +\infty$,

$$\mathbb{P} \left(\sum_{k=0}^n R_k > 0 \quad \forall 1 \leq n \leq N \right) \asymp 1/N^{1/4}.$$

Question : what if we replace Simple random Walk $(R_n)_n$ by Sinai's Walk $(S_n)_n$?

Main result

Theorem (persistence, AD 2014)

Let f be a function $\mathbb{Z} \rightarrow \mathbb{R}$, such that

- $f(0) = 0$;
- $\forall x > 0, f(x) \geq 1$; $\forall x < 0, f(x) \leq -1$;
- $|f(x)| \leq \exp(|x|^{o(1)})$ as $x \rightarrow \pm\infty$.

We consider $(S_n)_{n \in \mathbb{N}}$ Sinai walk and $u \leq 0$. As $N \rightarrow +\infty$,

$$\mathbb{P} \left(\sum_{k=0}^n f(S_k) > u \quad \forall 1 \leq n \leq N \right) = \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2} + o(1)}}.$$

Persistence exponent $\frac{3-\sqrt{5}}{2}$: first appears in Le Doussal, Monthus and Fisher (1999) then Cheliotis (2005)

Main result - some notation

Local time of $(S_n)_{n \in \mathbb{N}}$ at time $n \in \mathbb{N}$:

$$L(A, n) := \sum_{k=0}^n \mathbb{1}_{\{S_k \in A\}}, \quad A \subset \mathbb{Z},$$

$$L(x, n) := L(\{x\}, n), \quad x \in \mathbb{Z}.$$

Useful since for every function g ,

$$\sum_{k=0}^n g(S_k) = \sum_{x \in \mathbb{Z}} g(x) L(x, n), \quad n \in \mathbb{N},$$

Main result - examples

Example 1

For $f(x) = \mathbb{1}_{\{x>0\}} - \mathbb{1}_{\{x<0\}}$, we get as $N \rightarrow +\infty$

$$\mathbb{P} [L(\mathbb{N}^*, n) > L(\mathbb{Z}_-, n) \quad \forall 1 \leq n \leq N] = \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2} + o(1)}}.$$

Main result - examples

Example 2

Let $\alpha > 0$, $\text{sgn}(x) := \mathbb{1}_{\{x>0\}} - \mathbb{1}_{\{x<0\}}$ for $x \in \mathbb{R}$, and $f(x) = \text{sgn}(x)|x|^\alpha$ for $x \in \mathbb{Z}$. We get for $u \leq 0$, as $N \rightarrow +\infty$,

$$\mathbb{P} \left(\frac{1}{n} \sum_{k=0}^n \text{sgn}(S_k) |S_k|^\alpha > 0 \quad \forall 1 \leq n \leq N \right) = \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2} + o(1)}}.$$

For $\alpha = 1$: persistence of the *temporal average* $\frac{1}{n} \sum_{k=0}^n S_k$

Functions increasing more rapidly, such as for $\alpha > 0$,

$$f(x) = \text{sgn}(x) |x|^{\log(2+|x|)^\alpha}, \quad x \in \mathbb{Z}$$

Potential : definition

Definition : Potential

$$V(n) := \begin{cases} \sum_{i=1}^n \log \frac{1 - \omega_i}{\omega_i} & \text{if } n > 0 \\ 0 & \text{if } n = 0 \\ -\sum_{i=n+1}^0 \log \frac{1 - \omega_i}{\omega_i} & \text{si } n < 0 \end{cases}$$

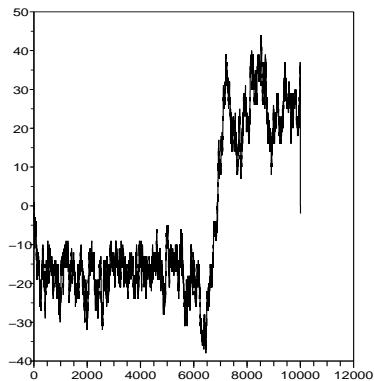
Scale function

$$\tau(p) := \inf \{k \in \mathbb{N}, S_k = p\}, \quad p \in \mathbb{Z}.$$

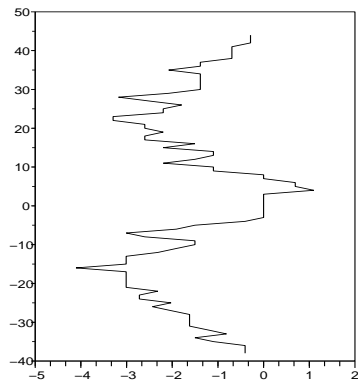
$$P_\omega^q[\tau(r) < \tau(p)] = \left(\sum_{k=p}^{q-1} e^{V(k)} \right) \left(\sum_{k=p}^{r-1} e^{V(k)} \right)^{-1}, \quad p < q < r,$$

Numerical simulation

Marche aleatoire en milieu aleatoire S



Potentiel V



Some useful classical estimates

If $g < h < i$

$$E_{\omega}^h[\tau(g) \wedge \tau(i)] \leq \varepsilon_0^{-1}(i - g)^2 \exp \left[\max_{g \leq \ell \leq k \leq i-1} (V(k) - V(\ell)) \right].$$

If $p < z < q < r$ or $p < q < z < r$,

$$\begin{aligned} & E_{\omega}^q[L(z, \tau(p) \wedge \tau(r))] \\ &= \frac{P_{\omega}^q[\tau(z) < \tau(p) \wedge \tau(r)]}{\omega_z P_{\omega}^{z+1}[\tau(z) > \tau(r)] + (1 - \omega_z) P_{\omega}^{z-1}[\tau(z) > \tau(p)]}. \end{aligned}$$

Coupling with Brownian motion

Komlós–Major–Tusnády strong approximation theorem

$\exists C_1 > 0, C_2 > 0, C_3 > 0$, independent of $K \in \mathbb{N}^*$, such that, possibly in an enlarged probability space, $\exists (W(t), t \in \mathbb{R})$, two-sided standard Brownian motion such that

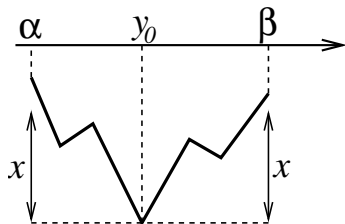
$$\mathbb{P} \left(\sup_{-K \leq i \leq K} |V(i) - \sigma W(i)| > C_1 \log K \right) \leq C_2 K^{-C_3}.$$

Definition of x -extrema

Definition (Neveu and Pitman, 1989)

For w continuous function $\mathbb{R} \rightarrow \mathbb{R}$, $x > 0$, and $y_0 \in \mathbb{R}$:
 y_0 is an x -*minimum* for w if $\exists \alpha < y_0 < \beta$, such that

- $w(y_0) = \inf\{w(y), y \in [\alpha, \beta]\}$,
- $w(\alpha) \geq w(y_0) + x$,
- $w(\beta) \geq w(y_0) + x$.

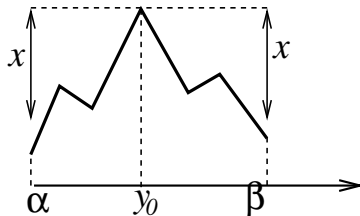
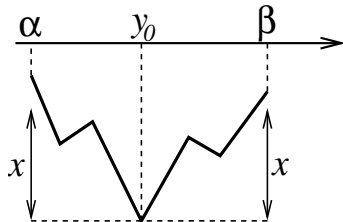


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x-extrema of two sided Brownian motion

Two-sided standard Brownian motion $(W(u), u \in \mathbb{R})$.
With probability 1 (Cheliotis) :



$$\forall x > 0, \quad \{x\text{-extrema of } W\} = \{x_k(W, x), k \in \mathbb{Z}\},$$

- $(x_k(W, x))_{k \in \mathbb{Z}}$ strictly increasing,
- unbounded from above and below,
- $x_0(W, x) \leq 0 < x_1(W, x)$;
- $\forall x > 0, \forall k \in \mathbb{Z}$,
 $x_{k+1}(W, x)$ x-maximum $\Leftrightarrow x_k(W, x)$ x-minimum.

x-extrema of two sided Brownian motion

For each $x > 0$,

$$b_W(x) := \begin{cases} x_0(W, x) & \text{if } x_0(W, x) \text{ is an } x\text{-minimum,} \\ x_1(W, x) & \text{otherwise.} \end{cases}$$

Results of Cheliotis

Cheliotis, 2005

$$\mathbb{P}(b_W \text{ does not change its sign in } [1, x]) \sim_{x \rightarrow +\infty} \frac{C}{x^{\frac{3-\sqrt{5}}{2}}}.$$

Cheliotis, 2005

Let $\mu_t = \text{law of } \frac{1}{t} \#\{\text{changes of sign of } b_W \text{ in } [1, e^t]\}$.

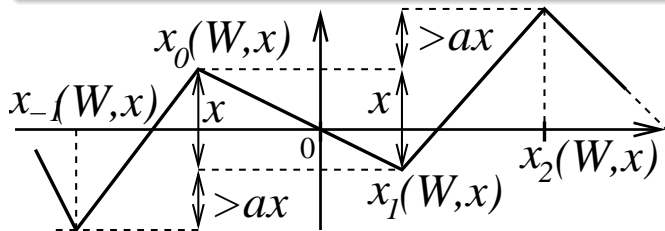
$(\mu_t)_t$ satisfies a Large Deviation Principle, with speed t , and rate function I .

I is continuous on \mathbb{R}_+ , $I(0) = \frac{3-\sqrt{5}}{2}$, $I(x) = +\infty$ if $x < 0$.

Sketch of the proof (for $u = 0$, $f(x) = x$) : Upper bound

Definition : strong change of sign

Consider $x > 0$. For $a > 0$, x is an a -strong change of sign of b_W
 $\Leftrightarrow e(T_0(x)) = 0$, $e(T_{-1}(x)) > ax$, and $e(T_1(x)) > ax$.



Lemma

$$\mathbb{P} \left(\begin{array}{l} \text{no } a\text{-strong change of sign in the } k \text{ first} \\ \text{changes of sign of } b_W \text{ in } [x, +\infty) \end{array} \right) \leq (1 - e^{-2a})^k$$

Upper bound : bad environments

Let $\varepsilon > 0$. ω is a **bad environment** if

- $\exists W$ Brownian motion, such that

$$\sup_{-K \leq x \leq K} \left| V(\lfloor x \rfloor) - \sigma W(x) \right| \leq C_1 \log K, \quad K = (\log N)^{\frac{3-\sqrt{5}}{2C_3} + 4},$$

- (technical conditions)

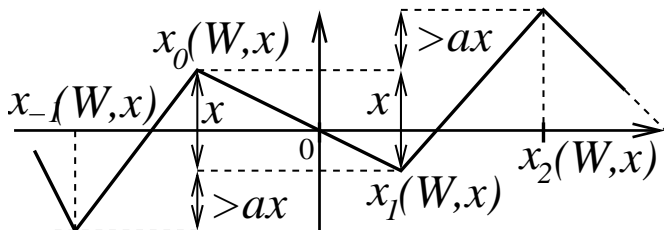
$$d_{\sigma W_+}(5 \log N) \leq (\log N)^4, \quad d_{\sigma W_-}(5 \log N) \leq (\log N)^4,$$

where $d_X(r) := \inf\{t \geq 0, X(t) - \inf_{[0,t]} X \geq r\}$,

- The number of sign changes of b_W in $[(\log N)^\varepsilon, (\log N)^{1-\varepsilon}]$ is $\geq 2\varepsilon \log \log N + 2$
- Among the $k = 2\varepsilon \log \log n$ first changes of sign of b_W in $[(\log N)^\varepsilon, +\infty)$, at least one is a -strong, for $a = \frac{1}{2} \exp\left(\frac{3-\sqrt{5}}{\varepsilon}\right)$.

$$\eta(\text{bad environments}) \geq 1 - \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2}-g(\varepsilon)}}, \quad \lim_{0^+} g = 0^+.$$

On a bad environment, with x is an a -strong change of sign and $(\log N)^\varepsilon \leq x \leq (\log N)^{1-\varepsilon}$,



$$\eta(\text{bad environments}) \geq 1 - \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2}-g(\varepsilon)}}, \quad \lim_{0^+} g = 0^+.$$

On a bad environment ω ,

$$\begin{aligned} P_\omega \left(\exists n \in [1, M], \sum_{k=1}^n S_k \leq 0 \right) &\geq 1 - \frac{4}{(\log N)^2} \\ \Rightarrow \mathbb{P} \left(\exists n \in [1, M], \sum_{k=1}^n S_k \leq 0 \right) \\ &\geq \int_{\{\text{bad environments}\}} P_\omega \left(\exists n \in [1, M], \sum_{k=1}^n S_k \leq 0 \right) \eta(d\omega) \\ &\geq \left(1 - \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2}-g(\varepsilon)}} \right) \left(1 - \frac{4}{(\log N)^2} \right). \end{aligned}$$

Lower bound : good environments

Let $\varepsilon > 0$. ω is a **good environment** if

- $-2\delta k \leq V(k) \leq -\delta k \quad 0 \leq k \leq \varepsilon \log \log N$
- $h := (\log N)^\varepsilon, \theta_0 := \inf\{k \geq \varepsilon \log \log N, V(k) \leq -5h\},$

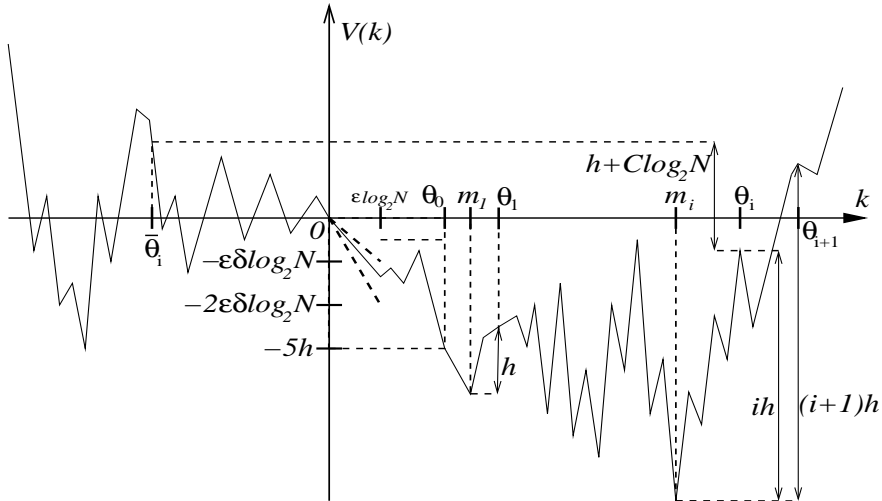
$$\forall \varepsilon \log \log N \leq k \leq \theta_0, \quad V(k) \leq -(\delta/2)\varepsilon \log \log N,$$

- $\theta_0 \leq \varepsilon \log \log N + (\log N)^{\varepsilon\delta/4}$
- $\widehat{V}(k) = V(k), k < 0, \widehat{V}(k) = V(k + \theta_0) - V(\theta_0), k \geq 0.$
 $\exists W$ Brownian motion, such that

$$\sup_{-K \leq x \leq K} |\widehat{V}(\lfloor x \rfloor) - \sigma W(x)| \leq C_1 \log K, \quad K = (\log N)^{\frac{3-\sqrt{5}}{2C_3} + 4},$$

- $d_{\sigma W_+}(5 \log N) \leq (\log N)^4, \quad d_{\sigma W_-}(5 \log N) \leq (\log N)^4,$
- $\forall x \in [1, 5 \log N], \quad b(x) > 0$

Good environment : $\varepsilon > 0$, $h := (\log N)^\varepsilon$



$$\eta(\text{good environemnts}) \geq \frac{1}{(\log N)^{\frac{3-\sqrt{5}}{2} + c\epsilon}}$$

On a good environment ω ,

$$\theta_i := \inf\{k > 0, V(k) - \inf_{[0,k]} V \geq ih\}, \quad i \geq 1.$$

By induction, for all $i \geq 1$ such that $ih \leq 4 \log N$,

$$P_\omega[F_i(N)] \geq c - i(\log N)^{-6},$$

where

$$F_i(N) := \left\{ \sum_{k=1}^n S_k > 0 \quad \forall 1 \leq n \leq \tau(\theta_i) \right\} \cap \left\{ \sum_{k=1}^{\tau(\theta_i)} S_k \geq \frac{\exp(ih)}{2(\log N)^c} \right\}.$$

Lower bound, conclusion

For i_N such that $3 \log N \leq i_N h \leq 4 \log N$,

$$F_{i_N}(N) \subset \{\tau(\theta_{i_N}) \geq N\}$$

$$\Rightarrow P_\omega \left(\sum_{k=0}^N S_k > 0 \quad \forall 1 < n < N \right) \geq P_\omega[F_{i_N}(N)] \geq c - \frac{i_N}{(\log N)^6} \geq \frac{c}{2}$$

$$\Rightarrow \mathbb{P} \left(\sum_{k=0}^N S_k > 0 \quad \forall 1 < n < N \right)$$

$$\geq \int_{\{\text{Good environemnts}\}} P_\omega \left(\sum_{k=0}^N S_k > 0 \quad \forall 1 < n < N \right) \eta(d\omega)$$

$$\geq (c/2) \eta(\text{good environemnts}) \geq \frac{c/2}{(\log N)^{\frac{3-\sqrt{5}}{2} + c\epsilon}}$$

Thank you for your attention.