

Limiting distributions for a one-dimensional random walk in a random environment

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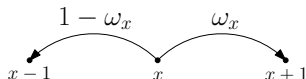
Joint work with Gennady Samorodnitsky

February 17, 2011

RWRE in \mathbb{Z} with i.i.d. environment

An *environment* $\omega = \{\omega_x\}_{x \in \mathbb{Z}} \in \Omega = [0, 1]^{\mathbb{Z}}$.

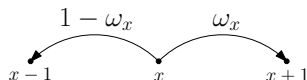
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Quenched law P_ω : fix an environment.

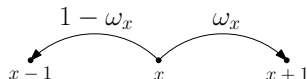
X_n a random walk: $X_0 = 0$, and

$$P_\omega(X_{n+1} = x + 1 | X_n = x) := \omega_x$$

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Averaged law \mathbb{P} : average over environments.

$$\mathbb{P}(G) := \int_{\Omega} P_\omega(G) dP(\omega)$$

Recurrence/Transience and LLN

$$\rho_X = \frac{1 - \omega_X}{\omega_X}$$

Theorem (Solomon '75)

- $E[\log \rho_0] < 0 \Rightarrow X_n \rightarrow +\infty$
- $E[\log \rho_0] > 0 \Rightarrow X_n \rightarrow -\infty$
- $E[\log \rho_0] = 0 \Rightarrow X_n$ is recurrent.

Theorem (Solomon '75)

If $E[\log \rho_0] < 0$ then

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = v_P = \begin{cases} \frac{1 - E[\rho_0]}{1 + E[\rho_0]} & E[\rho_0] < 1 \\ 0 & E[\rho_0] \geq 1. \end{cases}$$

Averaged Limiting Distributions

$\kappa > 0$ solves $E\rho^\kappa = 1$.

Theorem (Kesten, Kozlov, Spitzer '75)

There exists a constant b such that

$$\begin{aligned} (a) \quad \kappa \in (0, 1) &\Rightarrow \lim_{n \rightarrow \infty} \mathbb{P} \left(\frac{X_n}{n^\kappa} \leq x \right) = 1 - L_{\kappa, b}(x^{-1/\kappa}) \\ (b) \quad \kappa \in (1, 2) &\Rightarrow \lim_{n \rightarrow \infty} \mathbb{P} \left(\frac{X_n - nv_P}{n^{1/\kappa}} \leq x \right) = 1 - L_{\kappa, b}(-x) \\ (c) \quad \kappa > 2 &\Rightarrow \lim_{n \rightarrow \infty} \mathbb{P} \left(\frac{X_n - nv_P}{b\sqrt{n}} \leq x \right) = \Phi(x) \end{aligned}$$

where $L_{\kappa, b}$ is a κ -stable distribution function.

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Limits for X_n : $\mathbb{P}(T_n > t) \approx \mathbb{P}(X_t < n)$.

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$$T_n := \inf\{k \geq 0 : X_k = n\}.$$

(Hitting Times)

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$$T_n = \sum_{i=1}^n (T_i - T_{i-1})$$

Quenched Limiting Distribution: Gaussian Regime

Theorem (Goldsheid '06, P. '06)

If $\kappa > 2$ then

$$\lim_{n \rightarrow \infty} P_\omega \left(\frac{T_n - E_\omega T_n}{\sigma \sqrt{n}} \leq x \right) = \Phi(x), \quad P - a.s.$$

where $\sigma^2 = E(\text{Var}_\omega T_1)$, and

$$\lim_{n \rightarrow \infty} P_\omega \left(\frac{X_n - nv_P + Z_n(\omega)}{v_P^{3/2} \sigma \sqrt{n}} \leq x \right) = \Phi(x), \quad P - a.s.$$

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where $Z_n(\omega)$ depends only on the environment.

- Requires a *random* centering.
- Scaling constant is different from averaged CLT.

Quenched Limiting Distribution: Gaussian Regime

Quenched CLT: $T_n = \sum_{i=1}^n (T_i - T_{i-1})$

Use Lindberg-Feller Condition

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Quenched CLT \Rightarrow Averaged CLT:

$$\frac{T_n - n/v_P}{\sqrt{n}} = \frac{T_n - E_\omega T_n}{\sqrt{n}} + \frac{E_\omega T_n - n/v_P}{\sqrt{n}}$$

- Terms on right are asymptotically independent.
- $(E_\omega T_n - n/v_P)/\sqrt{n}$ is approximately mean zero Gaussian.

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Question

What happens when $\kappa < 2$?

Do we get quenched stable laws?

Quenched Limits: $\kappa < 2$

Theorem (P. '07)

If $\kappa < 2$ then $P - a.s.$ there exist random subsequences $n_k = n_k(\omega)$, and $m_k = m_k(\omega)$ such that

$$(a) \quad \lim_{k \rightarrow \infty} P_\omega \left(\frac{T_{n_k} - E_\omega T_{n_k}}{\sqrt{\text{Var}_\omega T_{n_k}}} \leq x \right) = \Phi(x)$$

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FAQ

- Which limit is more likely?

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- What other subsequential limits are possible?

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FAQ

- Which limit is more likely?
- What other subsequential limits are possible?
- Where do the κ -stable distributions come from.

Weak Quenched Limits: $\kappa < 2$

For any $x \in \mathbb{R}$,

$$P_\omega \left(\frac{T_n - E_\omega T_n}{n^{1/\kappa}} \leq x \right)$$

is a random variable.

- Doesn't converge almost surely to a deterministic constant.
- Maybe converges in some weaker sense (in distribution).

Random distributions

\mathcal{M}_1 = Probability distributions on \mathbb{R} (with Prohorov metric ρ).

$$\mu_{n,\omega}(\cdot) = \begin{cases} P_\omega \left(\frac{T_n - E_\omega T_n}{n^{1/\kappa}} \in \cdot \right) & \kappa < 2 \\ P_\omega \left(\frac{T_n - E_\omega T_n}{\sqrt{n}} \in \cdot \right) & \kappa > 2. \end{cases}$$

$\mu_{n,\omega}$ is an \mathcal{M}_1 -valued random variable.

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Weak Quenched Limits ($\kappa < 2$):

$\mu_{n,\omega}$ converges weakly (in distribution) on \mathcal{M}_1 .

Weak Quenched Limiting Distributions: $\kappa < 2$

Theorem (P. & Samorodnitsky '10)

If $\kappa < 2$, then there exists a $\lambda > 0$ such that

$$\mu_{n,\omega} \xrightarrow[n \rightarrow \infty]{} \mu_{\lambda,\kappa},$$

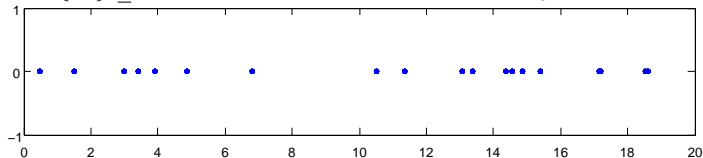
where $\mu_{\lambda,\kappa}$ is a random probability distribution defined by

$$\mu_{\lambda,\kappa}(A) = \mathbf{P} \left(\sum_{i=1}^{\infty} a_i(\tau_i - 1) \in A \mid a_i, i \geq 1 \right)$$

where $\{a_i\}_{i \geq 1}$ are the points of a non-homogeneous Poisson point process with intensity $\lambda \kappa x^{-\kappa-1}$.

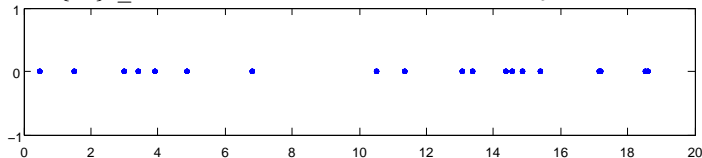
Non-homogeneous Poisson point processes

Let $\{\Gamma_i\}_{i \geq 1}$ be a PPP with constant intensity 1.

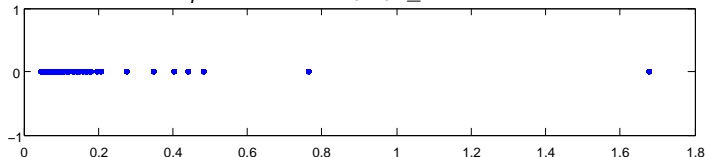


Non-homogeneous Poisson point processes

Let $\{\Gamma_i\}_{i \geq 1}$ be a PPP with constant intensity 1.



Let $a_i = \lambda^{1/\kappa} \Gamma_i^{-1/\kappa}$. Then, $\{a_i\}_{i \geq 1}$ is a PPP with intensity $\lambda \kappa x^{-\kappa-1}$.



Weak Quenched Limits: Averaged Centering ($\kappa < 1$)

Let $\tilde{\mu}_{n,\omega}(\cdot) = P_\omega \left(\frac{T_n}{n^{1/\kappa}} \in \cdot \right)$

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Note: Implies the averaged stable limit laws.

$$\mathbb{P} \left(\frac{T_n}{n^{1/\kappa}} \leq x \right) = E [\tilde{\mu}_{n,\omega}((-\infty, x])] \xrightarrow{n \rightarrow \infty} \mathbf{E} [\tilde{\mu}_{\lambda,\kappa}((-\infty, x])]$$

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- $a_i \tau_i$ are atoms of PPP with intensity $\lambda \kappa \Gamma(\kappa + 1) x^{-\kappa-1} dx$.

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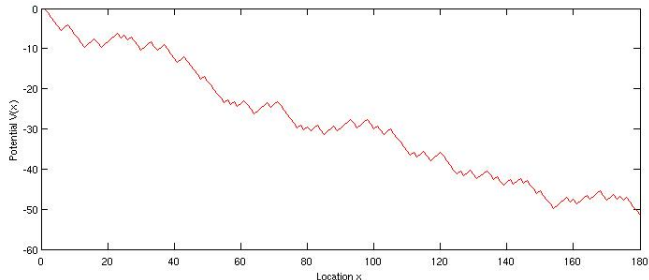
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- $\sum_{i \geq 1} a_i \tau_i$ is a κ -stable random variable.

Proofs

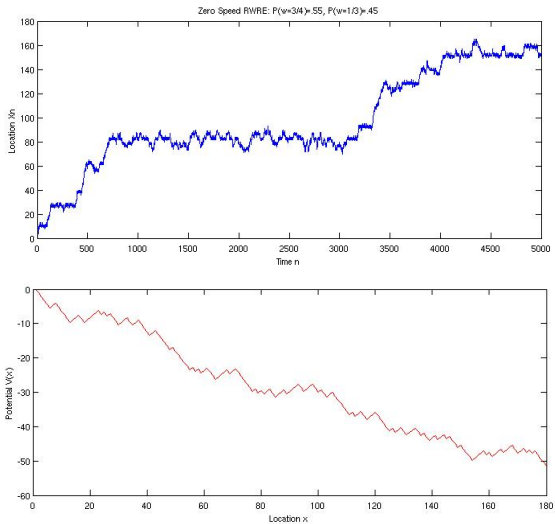
Potential:

$$V(i) := \begin{cases} \sum_{k=0}^{i-1} \log \rho_k, & i > 0 \\ 0, & i = 0 \\ \sum_{k=i}^{-1} -\log \rho_k, & i < 0 \end{cases}$$

Trap: Atypical section where the potential is increasing.



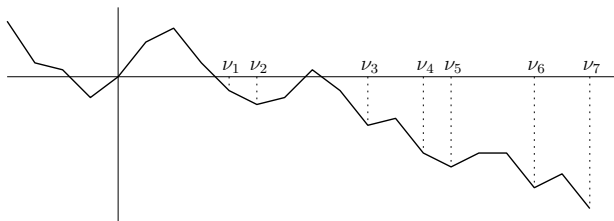
Trapping Effects



Blocks of the environment

Ladder locations $\{\nu_n\}$ defined by $\nu_0 = 0$,

$$\nu_n := \inf\{i > \nu_{n-1} : V(i) < V(\nu_{n-1})\}$$



Reduction to T_{ν_n}

$$\nu_n = \sum_{i=1}^n (\nu_i - \nu_{i-1}).$$

LLN implies

$$\frac{\nu_n}{n} \rightarrow \bar{\nu} = E\nu_1$$

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Enough to study

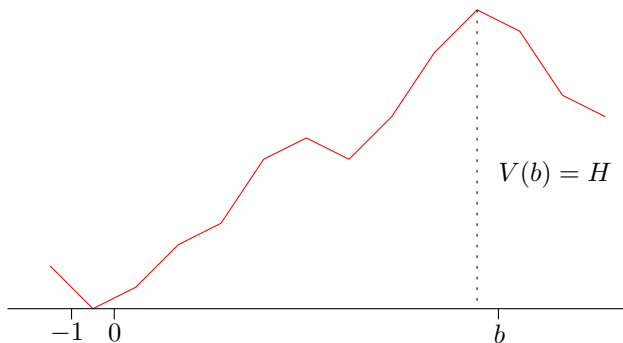
$$\phi_{n,\omega}(\cdot) = P_\omega \left(\frac{T_{\nu_n} - E_\omega T_{\nu_n}}{n^{1/\kappa}} \in \cdot \right)$$

Want to show $\phi_{n,\omega} \implies \mu_{\lambda,\kappa}$ for some λ .

Crossing Hills

Probability of escaping a trap of Height H .

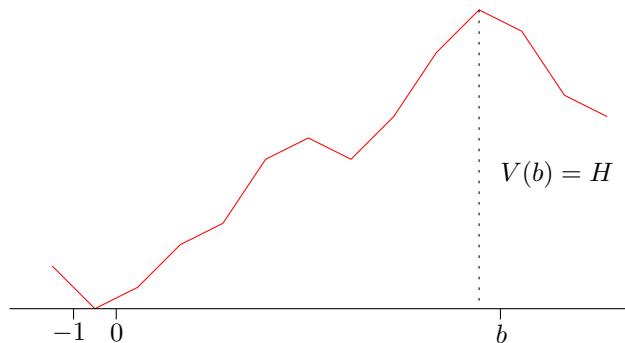
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$$P_{\omega}(T_b < T_{-1}) = \frac{1}{1 + \sum_{j=1}^b e^{V(j)}} \approx e^{-V(b)} = e^{-H}$$

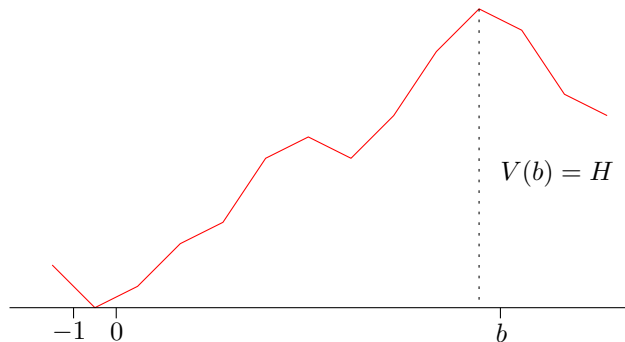


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$$P_{\omega}(T_b < T_{-1}) = \frac{1}{1 + \sum_{j=1}^b e^{V(j)}} \approx e^{-V(b)} = e^{-H}$$

Time to escape trap of height $H \approx \text{Exp}(e^{-H})$.



Comparison with exponentials

Want: $T_\nu \stackrel{Law}{\approx} (E_\omega T_\nu)_\tau$.

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Decompose T_ν into trials

$$T_\nu = S + \sum_{i=1}^G F_i$$

Where

$$S \sim (T_\nu \mid T_\nu < T_0^+)$$

$$F_i \sim (T_0^+ \mid T_0^+ < T_\nu)$$

$$G \sim \text{Geo}(p_\omega), \quad \text{where } p_\omega = P_\omega(T_\nu < T_0^+)$$

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Couple G with an exponential τ

$$G = \left\lfloor \frac{\tau}{\log(1/(1 - p_\omega))} \right\rfloor.$$

Comparison with mixture of exponentials

Let

$$\beta_i = \beta_i(\omega) = E_\omega(T_{\nu_i} - T_{\nu_{i-1}}).$$

The coupling gives

$$\frac{T_{\nu_n} - E_\omega T_{\nu_n}}{n^{1/\kappa}} \approx \frac{1}{n^{1/\kappa}} \sum_{i=1}^n \beta_i(\tau_i - 1)$$

where τ_i are i.i.d. $\text{Exp}(1)$.

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where τ_i are i.i.d. $\text{Exp}(1)$.

$$\lim_{n \rightarrow \infty} n^{-2/\kappa} \text{Var}_\omega \left(T_{\nu_n} - \sum_{i=1}^n \beta_i \tau_i \right) = 0, \quad \text{in } P\text{-probability.}$$

Heuristics of Quenched Limit Laws

Recall,

$$\phi_{n,\omega}(\cdot) = P_\omega \left(\frac{T_{\nu_n} - E_\omega T_{\nu_n}}{n^{1/\kappa}} \in \cdot \right)$$

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Heuristics of Quenched Limit Laws

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Need to understand the distribution of $\{\beta_i\}_{i=1}^n$

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\mathcal{M}_p = Point processes $\sum_{i \geq 1} \delta_{x_i}$ on $(0, \infty)$.

$$N_n = N_n(\omega) = \sum_{i=1}^n \delta_{\beta_i/n^{1/\kappa}}$$

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- 1 $\{\beta_i\}$ is (almost) stationary.
- 2 The β_i have heavy tails: $\beta_1 = E_\omega T_\nu \approx e^H$

$$P(e^H > x) \sim Cx^{-\kappa} \quad \text{as } x \rightarrow \infty.$$

- 3 $\{\beta_i\}$ has good mixing properties.

Define $H : \mathcal{M}_p \longrightarrow \mathcal{M}_1$ by

$$H(\zeta)(\cdot) = \mathbf{P} \left(\sum_{i \geq 1} x_i (\tau_i - 1) \in \cdot \right), \quad \text{for } \zeta = \sum_{i \geq 1} \delta_{x_i}.$$

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H is NOT continuous.