

A hierarchical renormalization model on trees

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Overview

Model (Derrida and Retaux) :

Let X_0 be random variable taking values in $\{0, 1, 2, \dots\}$. Define

$$X_{n+1} \stackrel{\text{law}}{=} (X_n^{(1)} + X_n^{(2)} - 1)^+, \quad \forall n \geq 0,$$

with two independent copies $X_n^{(1)}, X_n^{(2)}$ of X_n .

Question :

What can we say about the asymptotic behaviors of X_n as $n \rightarrow \infty$?

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Outline

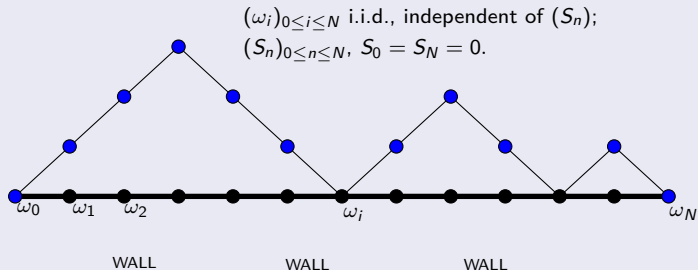
1 Motivation

- Pinning model on a hierarchical structure
- Derrida and Retaux' model

2 Results

3 Proofs

- Proof of the lower bound in Theorem 2
- Proof of the upper bounds in Theorems 1 and 2

The pinning model on \mathbb{Z} 

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- The measure of polymer of length N :

$$P_{N,\omega}(dS) := \frac{1}{Z_N} \exp\left(\sum_{i=1}^N \omega_i \mathbf{1}_{\{S_i=0\}}\right),$$

- Z_N is called the partition function [$\omega_i > 0$ (i attractive); $\omega_i < 0$ (i repulsive)].
- See Giacomin's book (Random Polymer Models, 2007).

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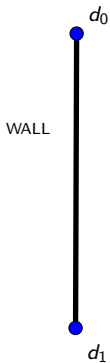
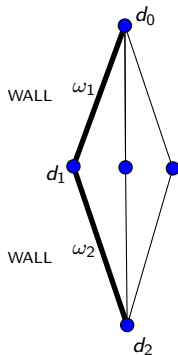
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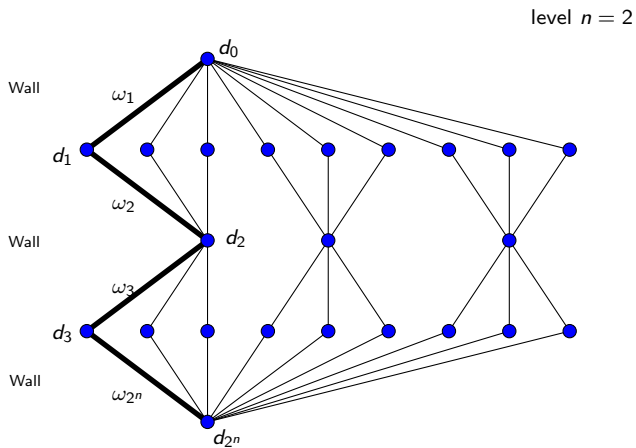
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Pinning model on a hierarchical lattice

Derrida, Hakim and Vannimenus (1992)

- 1 At level 0, there is a unique bond.
- 2 Fix an integer $B \geq 2$ (for e.g. $B = 3$)
- 3 Rule : Each bond gives B branches consisting of 2 bonds each.

Case $B = 3$ Level $n = 0$ Level $n = 1$ 

Case $B = 3$ 

Pinning model on a hierarchical lattice

- At level n , each (direct) trajectory $(S_i)_{1 \leq i \leq 2^n}$ (from d_0 to d_{2^n}) contains 2^n bonds.
- Choose the uniform measure $P_{B,n}$ on all possible trajectories [simple random walk $(S_i)_{1 \leq i \leq 2^n}$ on the hierarchical lattice].
- Let $(\omega_i)_{1 \leq i \leq 2^n}$ be i.i.d. and independent of (S_n) .
- The partition function

$$Z_n := E_{B,n} \exp \left(\sum_{i=1}^{2^n} \omega_i 1_{\{S_{i-1}=d_{i-1}, S_i=d_i\}} \right),$$

where the expectation is only taken with respect to (S_n) .

Pinning model on a hierarchical lattice

- Let $N_n :=$ number of trajectories γ from d_0 to d_{2^n} and

$$R_n := \sum_{\gamma: \gamma_0=d_0, \gamma_{2^n}=d_{2^n}} \exp \left(\sum_{i=1}^{2^n} \omega_i \mathbf{1}_{\{\gamma_{i-1}=d_{i-1}, \gamma_i=d_i\}} \right).$$

- Then $Z_n = \frac{R_n}{N_n}$.
- Easy to see that

$$\begin{aligned} N_{n+1} &= B N_n^2, \\ R_{n+1} &= R_n^{(1)} R_n^{(2)} + (B-1) N_n^2, \end{aligned}$$

with two independent copies $R_n^{(1)}, R_n^{(2)}$ of R_n .

Pinning model on a hierarchical lattice

- Then

$$Z_{n+1} = \frac{R_{n+1}}{N_{n+1}} = \frac{Z_n^{(1)} Z_n^{(2)} + B - 1}{B},$$

with two independent copies $Z_n^{(1)}, Z_n^{(2)}$ of Z_n .

- Stability : $(B, Z) \mapsto (B', Z')$ with $B' := \frac{B}{B-1}, Z' := \frac{Z}{B-1}$.
- See Monthus and Garet (2008), Derrida, Giacomin, Lacoïn and Toninelli (2009), Lacoïn and Toninelli (2009), Giacomin, Lacoïn and Toninelli (2010, 2011), Berger and Toninelli (2013) for the studies of this model [disorder relevance, critical line...]

Pinning model on a hierarchical lattice

- Let $X_n := \log Z_n$. Then

$$\begin{aligned} X_{n+1} &= \log Z_{n+1} \\ &= \log \frac{e^{(X_n^{(1)} + X_n^{(2)})} + B - 1}{B} \\ &= X_n^{(1)} + X_n^{(2)} + \log \frac{1 + (B - 1)e^{-(X_n^{(1)} + X_n^{(2)})}}{B}. \end{aligned}$$

- Let $1 < B < 2$ and define $a := -\log(B - 1) > 0$.
- If $X_n \geq -a$, a.s., then $X_{n+1} \geq -a$ a.s.

Derrida and Retaux (2014)'s model

- Fix $a > 0$. For any $n \geq 0$,

$$X_{n+1} \stackrel{\text{law}}{=} \max(X_n^{(1)} + X_n^{(2)}, -a),$$

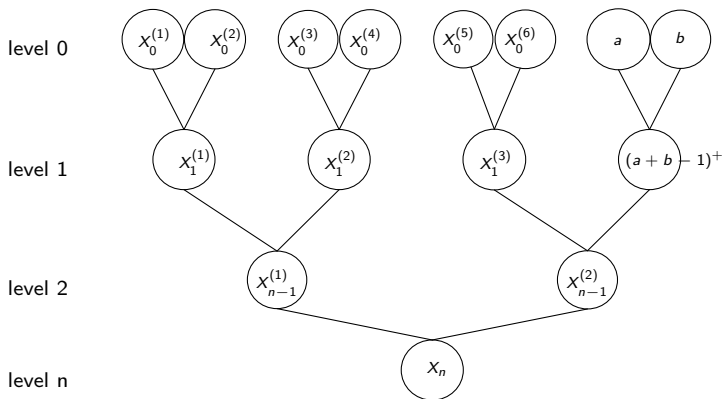
with two independent copies $X_n^{(1)}, X_n^{(2)}$ of X_n .

- Replacing X_n by $X_n + a$ and taking $a = 1$, the recursive equation becomes

$$X_{n+1} \stackrel{\text{law}}{=} (X_n^{(1)} + X_n^{(2)} - 1)^+, \quad \forall n \geq 0,$$

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- 1 Free energy : $F_\infty := \lim_{n \rightarrow \infty} \frac{\mathbb{E}(X_n)}{2^n} \in [0, \infty)$ exists.

Proof : As $X_n \stackrel{\text{law}}{=} (X_{n-1}^{(1)} + X_{n-1}^{(2)} - 1)^+$,

$2\mathbb{E}(X_{n-1}) \geq \mathbb{E}(X_n) \geq 2\mathbb{E}(X_{n-1}) - 1$, implying that

$$F_\infty := \lim_{n \rightarrow \infty} \downarrow \frac{\mathbb{E}(X_n)}{2^n} = \lim_{n \rightarrow \infty} \uparrow \frac{\mathbb{E}(X_n) - 1}{2^n}.$$

- 2 "Percolation on tree" : Let

$$X_0 \stackrel{\text{law}}{=} (1-p)\delta_{\{0\}} + p\delta_{\{Y\}},$$

with $0 \leq p \leq 1$ and $Y > 0$ a positive random variable. Define

$$p_c := \sup\{0 \leq p \leq 1 : F_\infty(p) = 0\}.$$

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Question 1 : Value of p_c .

- Example : $Y \equiv 2$; $p_c = 1/2$?

Question 2 : Behavior of F_∞ .

- If $p_c < 1$, when $p \downarrow p_c$, what is the behavior of $F_\infty(p)$?

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Value of p_c

Theorem (Collet, Eckman, Glaser and Martin 1984)

Suppose that $Y \in \{1, 2, \dots\}$. Then

$$p_c = \frac{1}{1 + \mathbb{E}((Y - 1)2^Y)}.$$

As example, if $Y \equiv 2$, then $p_c = \frac{1}{5}$.

Open question

Find p_c for a general r.v. $Y \in \mathbb{R}_+$; or even for $Y \in \frac{1}{2}\mathbb{N} \dots$

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Corollary

For any general r.v. $Y \in \mathbb{R}_+$,

$$p_c > 0 \iff \mathbb{E}(Y 2^Y) < \infty.$$

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Find a probabilistic proof on the above $L \log L$ -condition.

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Derrida's conjecture on the free energy

Derrida's conjecture :

If $p_c > 0$, then

$$F_\infty(p) = \exp\left(-\frac{K + o(1)}{(p - p_c)^{1/2}}\right), \quad p \downarrow p_c,$$

for some explicit constant $K > 0$.

Why do we need $p_c > 0$?

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Model

Model (Galton-Watson case)

Let ν be an integer-valued random variable such that $m := \mathbb{E}(\nu) \in (1, \infty)$. Consider the recursive equation

$$X_{n+1} \stackrel{\text{law}}{=} \left(\sum_{i=1}^{\nu} X_n^{(i)} - 1 \right)^+,$$

where $X_n^{(1)}, X_n^{(2)}, \dots$, are i.i.d. copies of X_n , and independent of ν .

Suppose $X_0 \stackrel{\text{law}}{=} (1-p)\delta_0 + p\delta_Y$, with $Y > 0$ a.s. (not necessarily integer-valued). Let

$$F_\infty(p) := \lim_{n \rightarrow \infty} \frac{1}{m^n} \mathbb{E}(X_n) \in [0, \infty)$$

and define

$$p_c := \sup\{0 \leq p \leq 1 : F_\infty(p) = 0\}.$$

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What is the value of p_c in the Galton-Watson case?

Only known when ν equals some integer a.s. [Collet et al. (1984)].

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Proposition

In the Galton-Watson case, $p_c = 0$ if and only if $\mathbb{E}(Ym^Y) = \infty$.

Proof :

“If part” is a consequence of Collet et al.'s theorem ;

“Only if” part : an argument by induction to control the generating function of X_n .

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Assumptions

We work under the assumption $\mathbb{E}(Ym^Y) = \infty$ (i.e. $p_c = 0$).

Assume that either

$$\mathbb{P}(Y > x) \approx x^\alpha m^{-x}, \quad x \rightarrow \infty,$$

for some $\alpha \geq -2$,

or

$$\mathbb{P}(Y > x) \approx \theta^{-x}, \quad x \rightarrow \infty,$$

for some $1 < \theta < m$. We study the rate of convergence of $F_\infty(p)$ as $p \rightarrow 0$.

Exponential decay

Theorem 1 (H., Shi (2017+))

Assume that

$$\mathbb{P}(Y > x) \approx x^\alpha m^{-x}, \quad x \rightarrow \infty,$$

for some $\alpha > -2$. Then

$$F_\infty(p) = \exp\left(-p^{-(1+o(1))/(2+\alpha)}\right), \quad p \rightarrow 0.$$

Polynomial decay

Theorem 2 (H., Shi (2017+))

Assume that

$$\mathbb{P}(Y > x) \approx \theta^{-x}, \quad x \rightarrow \infty,$$

for some $1 < \theta < m$. Then

$$F_\infty(p) \approx p^{\frac{\log m}{\log(m/\theta)}}, \quad p \rightarrow 0.$$

- From $X_{n+1} \stackrel{\text{law}}{=} (\sum_{i=1}^{\nu} X_n^{(i)} - 1)^+$, we get that

$$m\mathbb{E}(X_n) - 1 \leq \mathbb{E}(X_{n+1}) \leq m\mathbb{E}(X_n).$$

- Then

$$F_{\infty}(p) = \lim \uparrow m^{-n}(\mathbb{E}(X_n) - \frac{1}{m-1}) = \lim \downarrow m^{-n}\mathbb{E}(X_n).$$

- Hence $F_{\infty}(p) > 0$ if and only if $\exists n \geq 1$ such that $\mathbb{E}(X_n) > \frac{1}{m-1}$.
- Let $n_0 := \inf\{n \geq 1 : \mathbb{E}(X_n) > \frac{1}{m-1}\}$, then

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- Proof of the polynomial decay case : When

$$\mathbb{P}(X_0 > x) \sim p\theta^{-x}, \quad x \rightarrow \infty,$$

with $1 < \theta < m$. Prove that

$$n_0 \approx \frac{\log 1/p}{\log m/\theta}, \quad p \rightarrow 0,$$

where $n_0 := \inf\{n \geq 1 : \mathbb{E}(X_n) > \frac{1}{m-1}\}$.

Proof of Theorem 2 : Lower bound of $F_\infty(p)$, i.e. upper bound of n_0

- Let $M_n := \max_{1 \leq i \leq m^n} X_0^{(i)}$. Then $X_n \geq (M_n - n)^+$.
- Extreme value theory implies that

$$M_n \approx \frac{1}{\log \theta} \left(n \log m - \log \frac{1}{p} \right).$$

- Then

$$\mathbb{E}(X_n) \geq \mathbb{E}(M_n) - n \approx \left(\frac{\log m}{\log \theta} - 1 \right) n - \frac{\log \frac{1}{p}}{\log \theta} > \frac{1}{m-1},$$

for all $n > \left(\frac{1}{m-1} + \frac{\log \frac{1}{p}}{\log \theta} \right) / \left(\frac{\log m}{\log \theta} - 1 \right) \approx \frac{\log(1/p)}{\log m/\theta}$.

Consider the generating functions

$$G_n(s) := \mathbb{E}(s^{X_n}), \quad h(s) := \mathbb{E}(s^{\nu}).$$

Then

$$G_{n+1}(s) = \frac{h(G_n(s))}{s} + \frac{s-1}{s} h(G_n(0)) \leq \frac{h(G_n(s))}{s} + \frac{s-1}{s}.$$

This gives, by iteration, an upper bound on $G_n(s) - 1$. Choose $s = s(p) \sim m$ gives a lower bound for n_0 , i.e. an upper bound for $F_\infty(p)$.

The proof of the upper bound in Theorem 1 is similar, except for the case $\alpha \in (-2, -1)$ which is a slightly more delicate.

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