One-dimensional polymers in random environments: stretching vs. folding

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Abstract

In this article we study a non-directed polymer model on $\mathbb{Z}$, that is a one-dimensional simple random walk placed in a random environment. More precisely, the law of the random walk is modified by the exponential of the sum of potentials $\beta \omega_x - h$ sitting on the range of the random walk, where $(\omega_x)_{x \in \mathbb{Z}}$ are i.i.d. random variables (the disorder) and $\beta \geq 0$ (disorder strength) and $h \in \mathbb{R}$ (external field) are two parameters. When $\beta = 0, h > 0$, this corresponds to a random walk penalized by its range; when $\beta > 0, h = 0$, this corresponds to the “standard” polymer model in random environment, except that it is non-directed. In this work, we allow the parameters $\beta, h$ to vary according to the length of the random walk and we study in detail the competition between the stretching effect of the disorder, the folding effect of the external field (if $h \geq 0$) and the entropy cost of atypical trajectories. We prove a complete description of the (rich) phase diagram and we identify scaling limits of the model in the different phases. In particular, in the case $\beta > 0, h = 0$ of the non-directed polymer, if $\omega_x$ has a finite second moment we find a range size fluctuation exponent $\xi = 2/3$.

Keywords: Random Polymer, Random walk, Range, Heavy-tail distributions, Weak-coupling limit, Super-diffusivity, Sub-diffusivity

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1 Introduction

We consider here a simple symmetric random walk on $\mathbb{Z}^d$, $d \geq 1$, placed in a time-independent random environment, see [23]. The interaction with the environment occurs on the range of the random walk, i.e. on the sites visited by the walk. This model may be seen as a disordered version of random walks penalized by their range (in the spirit of [11, 19]). One closely related model is the celebrated directed polymer in random environment model (see [15] for a review), which has attracted interest from both the mathematical and physics communities over the last forty years, and can be used to describe a polymer chain placed in a solvent with impurities.

1.1 The model

Let $S := (S_n)_{n \geq 0}$ be a simple symmetric random walk on $\mathbb{Z}^d$, $d \geq 1$, starting from 0, whose trajectory represents a (non-directed) polymer. Let $\mathbf{P}$ denote its law. The random environment, or disorder, is modeled by a field $\omega := (\omega_x)_{x \in \mathbb{Z}^d}$ of i.i.d. random variables. We let $\mathbb{P}$ denote the law of $\omega$, and $\mathbb{E}$ the expectation with respect to $\mathbb{P}$ (assumptions on the law of $\omega$ are detailed in Section 1.2 below).

For $\beta \geq 0$ (the disorder strength, or inverse temperature) and $h \in \mathbb{R}$ (an external field), we define for all $N \in \mathbb{N}$ the following Gibbs transformation of the law $\mathbf{P}$, called the polymer measure:

$$ \frac{d\mathbf{P}^\omega_{N,\beta,h}}{d\mathbf{P}}(S) := \frac{1}{Z^\omega_{N,\beta,h}} \exp \left( \sum_{x \in \mathbb{Z}^d} (\beta \omega_x - h) \mathbb{1}_{\{x \in R_N\}} \right), \quad (1.1) $$

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where $\mathcal{R}_N = \{S_0, S_1, \ldots, S_N\}$ is the range of the random walk up to time $N$, and

$$Z_{N, \beta, h}^\omega := \mathbb{E}\left[ \exp\left( \sum_{x \in \mathbb{Z}^d} (\beta \omega_x - h) \mathbb{1}_{\{x \in \mathcal{R}_N\}} \right) \right] = \mathbb{E}\left[ \exp\left( \beta \sum_{x \in \mathcal{R}_N} \omega_x - h |\mathcal{R}_N| \right) \right]$$

is the partition function of the model, defined so that $\mathbf{P}_{N, \beta, h}^\omega$ is a probability measure.

Let us stress the main differences with the standard directed polymer model: (i) here, the random walk does not have a preferred direction; (ii) there is an additional external field $h \in \mathbb{R}$; (iii) the random walk can only pick up one weight $\beta \omega_x - h$ at a site $x \in \mathbb{Z}^d$, so returning to an already visited site does not bring any reward or penalty (in the directed polymer model, the environment is renewed each time).

We now wish to understand the typical behavior of polymer trajectories $(S_0, \ldots, S_N)$ under the polymer measure $\mathbf{P}_{N, \beta, h}^\omega$. Two important quantities we are interested in are:

- the range size exponent $\xi$, loosely defined as $\mathbb{E}|\mathcal{R}_N| = N^\xi$;
- the fluctuation exponent $\chi$, loosely defined as $\text{Var}(\log Z_{N, \beta, h}^\omega) = N^{2\chi}$.

In view of (1.1), there are three quantities that may influence the behavior of the polymer: the energy collected from the random environment $\omega$; the penalty $h$ (or reward depending on its sign) for having a large range; the entropy cost of the exploration of the random walk $S$. If $\beta = 0$ and $h > 0$, then we recover a random walk penalized by its range. This model is by now quite well understood: the random environment collected from the polymer measure $\mathbf{P}_{N, \beta, h}^\omega$ is the partition function of the model, defined so that $\mathbf{P}_{N, \beta, h}^\omega$ is a probability measure.

### 1.2 Setting of the paper

In this article, we focus on the case of the dimension $d = 1$: the behavior of the model is already very rich and we are able to obtain sharp results. Let us mention that in dimension $d \geq 2$ some aspects of the model are considered in [3], but many questions remain open.

Our main assumption on the environment is that $\omega_x$ is in the domain of attraction of some $\alpha$-stable law, with $\alpha \in (0, 2]$, $\alpha \neq 1$. More precisely, we assume the following

**Assumption 1.** If $\alpha = 2$ we assume that $\mathbb{E}[\omega_0] = 0$ and $\mathbb{E}[\omega_0^2] = 1$. If $\alpha \in (0, 1) \cup (1, 2)$ we assume that $P(\omega_0 > t) \sim pt^{-\alpha}$ and $P(\omega_0 < -t) \sim qt^{-\alpha}$ as $t \to \infty$, with $p + q = 1$ (and $p > 0$); if $q = 0$, we interpret it as $P(\omega_0 < -t) = o(t^{-\alpha})$. Moreover, if $\alpha \in (1, 2)$, we also assume that $\mathbb{E}[\omega_0] = 0$.

Let us stress that Assumption 1 ensures that:

- if $\alpha = 2$, then $\omega_t$ is in the normal domain of attraction, so that $(\frac{1}{\sqrt{n}} \sum_{i=1}^{\lfloor nt \rfloor} \omega_i)_{u \leq t \leq v}$ converges to a two-sided (standard) Brownian Motion;
- if $\alpha \in (0, 1) \cup (1, 2)$, then $\omega_t$ is in the domain of attraction of some non-Gaussian stable law and $(\frac{1}{\sqrt{n}} \sum_{i=1}^{\lfloor nt \rfloor} \omega_i)_{u \leq t \leq v}$ converges to a two-sided $\alpha$-stable Lévy process.

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1We use the standard notation $a_n \asymp b_n$ if $\limsup_{n \to +\infty} \frac{b_n}{a_n} < +\infty$ and $\liminf_{n \to +\infty} \frac{b_n}{a_n} < +\infty$.

2We use the standard notation $a(t) \sim b(t)$ if $\lim_{t \to +\infty} \frac{a(t)}{b(t)} = 1$ and $a(t) = o(b(t))$ if $\lim_{t \to +\infty} \frac{a(t)}{b(t)} = 0$. 

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We leave the case \( \alpha = 1 \) aside mostly for simplicity: indeed, to obtain a process convergence as above, a non-zero centering term is in general needed (even in the symmetric case \( p = q \), see [21 IX.8], or [5]); however most of our analysis applies in that case. We also focus on pure power tails when \( \alpha \in (0,2) \), simply to lighten notation and simplify the statements: our results could easily be adapted to the case of regularly varying tails.

Henceforth we refer to \((X_t)_{t \in \mathbb{R}}\) as the two-sided Brownian motion if \( \alpha = 2 \) and as the two-sided Lévy process defined below if \( \alpha \in (0,1) \cup (1,2) \). We refer to Chapter 1 of [1] for an overview on Lévy processes.

**Notation 1.** Let \( \alpha \in (0,2) \). We let \( X^{(1)} = (X^{(1)}_u)_{u \geq 0} \) and \( X^{(2)} = (X^{(2)}_u)_{u \geq 0} \) be two i.i.d. standard Brownian motions if \( \alpha = 2 \) and two i.i.d. (\( \alpha \)-stable) Lévy processes with no Brownian component, no drift, and Lévy measure \( \nu(dx) = \alpha (p \mathbb{1}_{x>0} + q \mathbb{1}_{x<0})|x|^{-1-\alpha} \, dx \) if \( \alpha \in (0,2) \).

We now define a coupling between the discrete environment \((\omega_x)_{x \in \mathbb{Z}}\) and the processes \( X^{(1)} \) and \( X^{(2)} \), using the construction proposed in [25]. Let us consider the space \( \mathcal{D} = \mathcal{D}(\mathbb{R}_+, \mathbb{R}) \) of all càdlàg real functions equipped with the Skorokhod metric \( \bar{d} \). We let

\[
\Sigma^+_j = \sum_{x=0}^{j} \omega_x, \quad \Sigma^-_j = \sum_{x=-j}^{-1} \omega_x \quad \text{for } j \geq 0,
\]

and for \( u, v \geq 0 \)

\[
X^{(1)}_{N,u,v} = X^{(1)}_{N,v}(\omega) := N^{-\frac{1}{\alpha}} \Sigma^+_j, \quad X^{(2)}_{N,u,v} = X^{(2)}_{N,u}(\omega) := N^{-\frac{1}{\alpha}} \Sigma^-_j.
\]

For \( 0 < \alpha < 2 \) the two (independent) processes \( X^{(1)}_N = (X^{(1)}_{N,u})_{u \geq 0} \) and \( X^{(2)}_N = (X^{(2)}_{N,u})_{u \geq 0} \) are càdlàg (they are \( \mathcal{D} \)-valued random variables) and they converge in distribution to \( X^{(1)} = (X^{(1)}_u)_{u \geq 0} \) and \( X^{(2)} = (X^{(2)}_u)_{u \geq 0} \) as in Notation [1] which are \( \mathcal{D} \)-valued random variables. Then, as done in [25 Section 3], we can build on the same probability space a sequence of random fields \( \omega^{(N)} = (\omega^{(N)}_x)_{x \in \mathbb{Z}} \) parametrized by \( N \), such that \( \omega^{(N)} \) has the same law as the original environment \( \omega \) for every \( N \) and for which the processes \( X^{(1)}_N(\omega^{(N)}) \) and \( X^{(2)}_N(\omega^{(N)}) \) converge a.s. in the Skorokhod metric on \( \mathcal{D} \) to \( X^{(1)} \) and \( X^{(2)} \) respectively—we refer to chapter VI in [21] for a characterisation of the convergence of sequences in \( \mathcal{D} \). A coupling can also be realized in the case \( \alpha = 2 \), see e.g. [10 Ch. 2]. We denote by \( \bar{\mathbb{P}} \) the law of the coupling and in order to lighten notation we will denote \( \bar{\omega} \) instead of \( \omega^{(N)} \) in such a coupling, letting the dependence on \( N \) be implicit.

### 1.3 Presentation of a first result

In the present paper, we allow \( \beta \) and \( h \) to vary with the size of the system, giving rise to a large diversity of possible behaviors. Before we go into these details, let us already state how our results translate in the case of fixed parameters \( \beta, h \).

We define

\[
M^+_N := \max_{0 \leq n \leq N} S_n \geq 0 \quad \text{and} \quad M^-_N := \min_{0 \leq n \leq N} S_n \leq 0
\]

the right-most and left-most points of the random walk after \( N \) steps. In particular, the size of the range is \( M^+_N - M^-_N \).

**Theorem 1.1.** Consider the coupling \( \bar{\mathbb{P}} \) defined above.

1. **Case \( \alpha \in (1,2) \).**

   (a) If \( \beta \geq 0 \) and \( h > 0 \). Then, for any \( \varepsilon > 0 \), we have that

   \[
   \lim_{N \to \infty} \frac{1}{N^{3/2}} \log \bar{Z}^\omega_{N,\beta,h} = -\frac{3}{2} (h \pi)^{2/3} \quad \bar{\mathbb{P}}\text{-a.s.}
   \]

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On the other hand, in the case \( \alpha \) the size of the range of the random walk in relation to the environment. More precisely, with high probability, the coupling is not needed. Additionally, the location of the maximizer of \( \sup_{u,v\geq 0} \{Y_{u,v}^{(2)}\} \) is well-defined, and for any \( \varepsilon > 0 \),
\[
\lim_{N \to \infty} \mathbb{P}_{N,\beta,h}^\omega \left( \left| \frac{1}{N^2} \left( M_N - M^\dagger_N \right) - \pi^\frac{2}{3} h^{-\frac{2}{3}} \right| > \varepsilon \right) = 0, \quad \mathbb{P}\text{-a.s.}
\]

(b) If \( \beta > 0 \) and \( h = 0 \). Then, letting
\[
Y_{u,v}^{(2)} = Y_{u,v}^{\beta,(2)} := \beta(X_v^{(1)} + X_u^{(2)}) - \frac{1}{2}(u \wedge v + u + v)^2
\]
for \( u, v \geq 0 \), we have
\[
\lim_{N \to \infty} \frac{1}{N^2} \log Z_{N,\beta,h}^\omega = \sup_{u,v\geq 0} \left\{ Y_{u,v}^{(2)} \right\} \in (0, +\infty), \quad \text{in} \ \hat{\mathbb{P}}\text{-probability.}
\]

(c) If \( \beta \geq 0 \) and \( h < 0 \). Then for any \( \varepsilon > 0 \), we have that
\[
\lim_{N \to \infty} \mathbb{P}_{N,\beta,h}^\omega \left( \left| \frac{1}{N} |S_N| - \tanh |h| \right| > \varepsilon \right) = 0, \quad \mathbb{P}\text{-a.s.}
\]

2. Case \( \alpha \in (0, 1) \). Let \( \beta > 0 \) and \( h \in \mathbb{R} \). Then letting
\[
Y_{u,v}^{(3)} = Y_{u,v}^{\beta,(3)} := \beta(X_v^{(1)} + X_u^{(2)}) - \infty 1_{\{u \wedge v + u + v > 1\}}
\]
for \( u, v \geq 0 \), we have
\[
\lim_{N \to \infty} \frac{1}{N^2} \log Z_{N,\beta,h}^\omega = \sup_{u,v\geq 0} \left\{ Y_{u,v}^{(3)} \right\} \in (0, +\infty), \quad \hat{\mathbb{P}}\text{-a.s.}
\]

Additionally, \( (\mathcal{U}^{(3)}, \mathcal{V}^{(3)}) := \arg \max_{u,v\geq 0} \{ Y_{u,v}^{(3)} \} \) is well-defined and for any \( \varepsilon > 0 \),
\[
\lim_{N \to \infty} \mathbb{P}_{N,\beta,h}^\omega \left( \left| \frac{1}{N} (M_N - M_N^\dagger) - (-\mathcal{U}^{(3)}, \mathcal{V}^{(3)}) \right| > \varepsilon \right) = 0, \quad \hat{\mathbb{P}}\text{-a.s.}
\]

Let us stress that, when \( \alpha \in (1, 2] \), the range size (see Definition 5.1 below for a proper definition) of the polymer is:

(a) of order \( N^{1/3} \) if \( h > 0 \) — folded phase, this is included in Theorem 3.6

(b) of order \( N^{\alpha/(2\alpha-1)} \) if \( h = 0 \), \( \beta > 0 \) — extended phase, this is included in Theorem 3.3

(c) of order \( N \) if \( h < 0 \) — extended phase, this is included in Theorem 1.2

On the other hand, in the case \( \alpha \in (0, 1) \), the range size is always of order \( N \), for whatever value of \( h \in \mathbb{R} \) — extended phase, this is included in Theorem 3.4 below.

Remark 1.1. The main advantage of using the coupling \( \hat{\mathbb{P}} \) defined above is that we are able to discuss the size of the range of the random walk in relation to the environment. More precisely, with high probability (or \( \mathbb{P}\text{-a.s., depending on the case} \), the end-points of the range under the polymer measure converge to the maximizer of the variational problem. This is what we present in Theorem 1.7 and later in Section 3 (Regions 2, 3 and 4, see Figure 4). Let us mention that there are some convergences in \( \hat{\mathbb{P}}\text{-probability that we are not able to upgrade to} \ \mathbb{P}\text{-a.s. convergences: this is due to a lack of control of the convergence of} \ (\mathcal{X}^\dagger_j)_{j\geq 1} \text{ on different scales in Lemmas 5.2 and 5.3, see Remark 5.1. Note that this problem disappears we have a good control on the coupling (see again Remark 1.7) or when} \ \xi = 1. \ \text{On the other hand, in the case when the limiting behavior is not affected by the environment (for example in Regions 1, 5, 6 discussed in Section 3), the coupling is not needed.} \)
1.4 Varying the parameters $\beta$ and $h$

In order to observe a transition between a folded phase ($h > 0$, $\beta = 0$) and an unfolded phase ($h = 0$, $\beta > 0$, or $h < 0$), a natural idea is to consider parameters $\beta$ and $h$ that depend on the size of the system, i.e. $\beta := \beta_N$ and $h := h_N$. There are then some sophisticated balances between the energy gain, the range penalty and the entropy cost as we tune $\beta_N$ and $h_N$. Our main results identify the different regimes for the behavior of the random walk: we provide a complete (and rich) phase diagram (see Figures 1-2-3 below), and describe each phase precisely (range size and fluctuation exponents, limit of the rescaled log-partition function).

In the rest of the paper, we therefore consider the following setting:

$$\beta_N := \hat{\beta} N^{-\gamma} \quad \text{and} \quad h_N := \hat{h} N^{-\zeta},$$

where $\gamma, \zeta \in \mathbb{R}$ describe the asymptotic behavior of $\beta_N, h_N$, and $\hat{\beta} > 0, \hat{h} \in \mathbb{R}$ are two fixed parameters. We could consider a slightly more general setting, adding some slowly varying function in the asymptotic behavior of $\beta_N$ or $h_N$: we chose to stick to the simpler strictly power-law case, to avoid lengthy notation and more technical calculations.

2 Some heuristics: presentation of the phase diagrams

In this section, we focus on the case $h \geq 0$; the case $h < 0$ is considered in Section 3 (it has a less rich behavior and is somehow simpler, see Remark 2 below). In analogy with the directed polymer model in a heavy-tailed random environment [6, 7], the presence of heavy-tails (Assumption 1) strongly impacts the behavior of the model: the phase diagrams are different according to whether $\alpha \in (1, 2], \alpha \in \left(\frac{1}{2}, 1\right)$ or $\alpha \in (0, \frac{1}{2})$.

Let us denote $\xi$ the typical range size exponent of the random walk under the polymer measure $P_{\hat{\beta} N, \hat{h} N}$ (see Definition 3.1 below for a proper definition), and let us develop some heuristics to determine $\xi \in [0, 1]$.

First of all, thanks to Lemmas 4.1, 4.3 in Appendix, we have for $0 < a < b$,

$$\log P(|R_N| \in (a N^\xi, b N^\xi)) \approx \log P\left(\max_{0 \leq n \leq N} |S_n| \in (a N^\xi, b N^\xi)\right) \approx \begin{cases} -N^{2\xi - 1}, & \text{if } \xi \geq \frac{1}{2}, \\ -N^{1-2\xi}, & \text{if } \xi \leq \frac{1}{2}. \end{cases}$$

(2.1)

If $\xi > 1/2$, this corresponds to a “stretching” of the random walk, whereas when $\xi < 1/2$, this corresponds to a “folding” of the random walk. We refer to (2.1) as the entropic cost of having range size $N^\xi$.

Then, if the range size is of order $N^\xi$, then under Assumption 1 and in view of (1.4), we get that

$$\beta_N \sum_{x \in R_N} \omega_x \approx \hat{\beta} N^{\frac{\xi}{2} - \gamma}, \quad h_N |R_N| \approx \hat{h} N^{\xi - \zeta}.$$ 

(2.2)

We refer to the first term as the “disorder” term, and to the second one as the “range” term (recall we focus for now on the case $\hat{h} > 0$ so the “range” term is always with a minus sign). All together, if the range size is of order $N^\xi$, then log $Z_{\hat{\beta} N, \hat{h} N}$ should get a contribution from three terms:

$$\text{disorder} \approx \hat{\beta} N^{\frac{\xi}{2} - \gamma}, \quad \text{range} \approx -\hat{h} N^{\xi - \zeta}, \quad \text{entropy} \approx - \begin{cases} N^{1-2\xi} & \text{if } \xi \leq 1/2, \\ N^{2\xi - 1} & \text{if } \xi \geq 1/2. \end{cases}$$

(2.3)

In (2.3), there is therefore a competition between the “disorder” (first term), the “range” (second term), and the “entropy” (last term). We now discuss how a balance can be achieved between these terms depending on $\gamma$ and $\zeta$ (and how they determine $\xi$). There are three main possibilities:

(i) there is a “disorder”-“entropy” balance (and the “range” term is negligible);

(ii) there is a “range”-“entropy” balance (and the “disorder” term is negligible);
(iii) there is a “range”-“disorder” balance (and the “entropy” term is negligible).

To summarize, all three regimes can occur (depending on $\gamma, \zeta$) if $\alpha \in (1, 2]$; on the other hand, regime (iii) disappears if $\alpha \in (0, 1)$, and regime (i) disappears if $\alpha \in (0, \frac{1}{2})$. We now determine for which values of $\gamma, \zeta$ one can observe the different regimes above: we consider the three subcases $\alpha \in (1, 2]$, $\alpha \in (\frac{1}{2}, 1)$ and $\alpha \in (0, \frac{1}{2})$ separately.

2.1 Phase diagram for $\alpha \in (1, 2]$

Instead of looking for “disorder”-“entropy”, “range”-“entropy” or “range”-“disorder” balance, we will find conditions to have the “disorder” term much larger, much smaller, or of the order of the “range” term.

Case I (“disorder” $\gg$ “range”). This corresponds to having $\xi/\alpha - \gamma > \xi - \zeta$. In that case, the random walk should not feel the penalty for having a large range, so we should have $\xi \geq 1/2$. The competition occurs only between energy and entropy: one could achieve a balance if $\xi/\alpha - \gamma = 2\xi - 1$, that is if

$$\xi = \frac{\alpha}{2\alpha - 1}(1 - \gamma) \quad \text{when} \quad \gamma < \frac{(2\alpha - 1)\zeta - (\alpha - 1)}{\alpha},$$  \hspace{1cm} (2.4)

where the condition on $\gamma$ derives from the fact that $\xi/\alpha - \gamma > \xi - \zeta$, i.e. $\gamma < \zeta - \xi \frac{\alpha - 1}{\alpha}$, in the regime considered here. However, since $\xi \leq 1$, we should have $\xi = 1$ when $\gamma$ is too small, more precisely when $\gamma \leq -\frac{\alpha - 1}{\alpha}$. Thus, we should have have

$$\xi = 1 \quad \text{when} \quad \gamma \leq -\frac{\alpha - 1}{\alpha} \text{ and } \gamma < \zeta - \frac{\alpha - 1}{\alpha}. \hspace{1cm} (2.5)$$

Also, since $\xi \geq 1/2$, we should have $\xi = 1/2$ if $\gamma$ is too large, more precisely when $\gamma \geq \frac{1}{2\alpha}$. Thus, we should have

$$\xi = \frac{1}{2} \quad \text{when} \quad \gamma \geq \frac{1}{2\alpha} \text{ and } \gamma < \zeta - \frac{\alpha - 1}{2\alpha}. \hspace{1cm} (2.6)$$

Case II (“disorder” $\ll$ “range”). This corresponds to having $\xi/\alpha - \gamma < \xi - \zeta$. In that case, the random walk feels the penalty for having a large range, so we should have $\xi \leq 1/2$. The competition occurs only between range and entropy: one could achieve a balance if $\xi - \zeta = 1 - 2\xi$, that is if

$$\xi = \frac{1 + \zeta}{3} \quad \text{when} \quad \gamma > \frac{(2\alpha + 1)\zeta - (\alpha - 1)}{3\alpha}, \hspace{1cm} (2.7)$$

where the condition on $\gamma$ derives from the fact that $\xi/\alpha - \gamma > \xi - \zeta$, i.e. $\gamma > \zeta - \xi \frac{\alpha - 1}{\alpha}$, in the regime considered here. Since $\xi \in [0, 1/2]$, similarly to (2.6), (2.7) we should have that

$$\xi = 0 \quad \text{when} \quad \zeta \leq -1 \text{ and } \gamma > \zeta, \hspace{1cm} (2.8)$$

and

$$\xi = \frac{1}{2} \quad \text{when} \quad \zeta \geq \frac{1}{2} \text{ and } \gamma > \zeta - \frac{\alpha - 1}{2\alpha}. \hspace{1cm} (2.9)$$

Case III (“disorder” $\approx$ “range” $\gg$ “entropy”). This corresponds to having $\xi/\alpha - \gamma = \xi - \zeta$, that is

$$\xi = \frac{\alpha}{\alpha - 1}(\zeta - \gamma). \hspace{1cm} (2.10)$$

In this regime, the entropy cost should be negligible compared to the disorder gain, and we should therefore have that $\xi/\alpha - \gamma > 1 - 2\xi$ if $\xi \leq 1/2$ and $\xi/\alpha - \gamma > 2\xi - 1$ for $\xi \geq 1/2$: after some calculation (and using (2.10)), we find the following condition on $\gamma$

$$\frac{(2\alpha - 1)\zeta - (\alpha - 1)}{\alpha} < \gamma < \frac{(2\alpha + 1)\zeta - (\alpha - 1)}{3\alpha}. \hspace{1cm} (2.11)$$
Moreover, since \( \xi \in [0, 1] \), we must have
\[
\zeta - \frac{\alpha - 1}{\alpha} \leq \gamma \leq \zeta.
\]

(2.12)

To summarize, for \( \alpha \in (1, 2] \), we have identified six different regimes according to the value of \( \gamma, \zeta \): they are described as follows. A representation of these regions in the \((\zeta, \gamma)\)-diagram is given in Figure 1 below.

**Region 1** “disorder”, “range” \( \prec \) “entropy” (Case I-II degenerate)
\[
R_1 = \{\xi = \frac{1}{2}, \gamma > \frac{1}{2}, \zeta > \frac{1}{2}\};
\]

**Region 2** “range” \( \prec \) “disorder” \( \approx \) “entropy” (Case I)
\[
R_2 = \left\{\xi = \frac{\alpha^2}{\alpha - 1}(1 - \gamma), \frac{1 - \alpha}{\alpha} < \gamma < \left(\frac{2\alpha - 1}{\alpha}\right)(\gamma - \zeta) \wedge \frac{1}{2}\alpha\right\};
\]

**Region 3** “range”, “entropy” \( \prec \) “disorder” (Case I degenerate)
\[
R_3 = \{\xi = 1, \gamma < -\frac{\alpha - 1}{\alpha}, \gamma < \zeta - \frac{\alpha - 1}{\alpha}\};
\]

**Region 4** “entropy” \( \prec \) “range” \( \approx \) “disorder” (Case III)
\[
R_4 = \left\{\xi = \frac{\alpha}{\alpha - 1}(\zeta - \gamma), \left(\frac{2\alpha - 1}{\alpha}\right)(\gamma - \zeta) \wedge \left(\zeta - \frac{\alpha - 1}{\alpha}\right) < \gamma < \left(\frac{2\alpha + 1}{3\alpha}\right)(\gamma - \zeta) \wedge \zeta\right\};
\]

**Region 5** “disorder” \( \prec \) “range” \( \approx \) “entropy” (Case II)
\[
R_5 = \left\{\xi = \frac{1 + \zeta}{3}, \gamma > \left(\frac{2\alpha + 1}{3\alpha}\right)(\gamma - \zeta), -1 < \zeta < \frac{1}{2}\right\};
\]

**Region 6** “disorder”, “entropy” \( \prec \) “range” (Case II degenerate)
\[
R_6 = \{\xi = 0, \gamma > \zeta, \zeta < -1\}.
\]

### 2.2 Phase diagram for \( \alpha \in (0, 1) \)

Let us highlight the main differences with the case \( \alpha \in (1, 2] \): the region \( R_4 \) no longer exists when \( \alpha < 1 \), and the region \( R_2 \) also disappears when \( \alpha < 1/2 \). Indeed, region \( R_4 \) corresponds to the case “disorder” \( \approx \) “range”, in which we have \( \xi = \frac{\alpha^2}{\alpha - 1}(\gamma - \zeta) \): it is easy to check that for \( \alpha \in (0, 1) \) there is no \( \gamma \) that can satisfy (2.12), which suggests that there is no “disorder”-“range” balance possible. In the same manner, when \( \alpha \in (0, \frac{1}{2}) \) there is no \( \gamma \) that satisfy \( \frac{1 - \alpha}{\alpha} < \gamma < \frac{1}{2\alpha} \) (see the definition of \( R_2 \) above), which suggests that there is no “disorder”-“entropy” balance possible: region \( R_2 \) no longer exists. We also refer to Section 3.2 (Comment 2) for further comments on why regions \( R_4 \) and \( R_2 \) disappear for \( \alpha < 1 \) and \( \alpha < 1/2 \) respectively.

All together, for \( \alpha \in (\frac{1}{2}, 1) \) we obtain the \((\zeta, \gamma)\)-diagram presented in Figure 2 below: the different regions are described as follows:
\[
R_1 = \{\xi = \frac{1}{2}, \gamma > \frac{1}{2\alpha}, \zeta > \frac{1}{2}\},
\]
\[
R_2 = \left\{\xi = \frac{\alpha^2}{\alpha - 1}(1 - \gamma), \frac{1 - \alpha}{\alpha} < \gamma < \left(\frac{2\alpha - 1}{\alpha}\right)(\gamma - \zeta) \wedge \frac{1}{2}\alpha\right\},
\]
\[
R_3 = \{\xi = 1, \gamma < -\frac{\alpha}{\alpha}, \gamma < \zeta - \frac{\alpha - 1}{\alpha}\},
\]
Remark 2.1. In the case \( h < 0 \), one can conduct similar computation as in (2.4)–(2.12) and obtain a different phase diagram than those of Figures 1–2–3, see Figures 4 and 5 below (note that regions \( R_1, R_2, R_3 \) are unchanged, since the range term is negligible in these regions). Let us stress that when \( h < 0 \), the “disorder” and “range” terms both play in the same direction and encourage exploration, resulting in a much simpler diagram: only range size exponents \( \xi \geq 1/2 \) are possible, see Section 4 below.

Figure 1: Phase diagram in the case \( \alpha \in (1, 2] \). The region \( R_1 \) and the dashed line \( \gamma = \zeta - \frac{a-1}{2\alpha} \) are the thresholds that split the regions of super-diffusivity and sub-diffusivity. Note that when \( \alpha = 1 \), the four lines \( \gamma = \frac{(2a-1)(\zeta-(\alpha-1))}{a(\alpha-1)} \), \( \gamma = \frac{(2a+1)(\zeta-(\alpha-1))}{3a} \), and \( \gamma = \zeta, \gamma = \zeta - \frac{a-1}{\alpha} \) all merge to the line \( \gamma = \zeta \).

Figure 2: Phase diagram in the case \( \alpha \in (1/2, 1) \). Compared to Figure 1, the region \( R_4 \) no longer exists.

\[
R_5 = \left\{ \xi = \frac{1+\zeta}{3}, \frac{(2a-1)(\zeta-(\alpha-1))}{a(\alpha-1)} \right\} \wedge \left( \zeta - \frac{a-1}{\alpha} < \gamma, \ -1 < \zeta < \frac{1}{2} \right),
\]

\[
R_6 = \left\{ \xi = 0, \ \gamma > \zeta - \frac{a-1}{\alpha}, \ \zeta < \frac{1}{\alpha} - 2 \right\}.
\]

Finally, for \( \alpha \in (0, \frac{1}{2}) \) we obtain the \((\zeta, \gamma)\)-diagram presented in Figure 3 below: the different regions are described as follows:

\[
R_1 = \left\{ \xi = \frac{1}{2}, \ \gamma > \frac{1-a}{\alpha}, \ \zeta > \frac{1}{2} \right\},
\]

\[
R_3 = \left\{ \xi = 1, \ \gamma < \frac{1-a}{\alpha}, \ \zeta < \frac{1}{\alpha} - 2 \right\},
\]

\[
R_5 = \left\{ \xi = \frac{1+\zeta}{3}, \ \frac{1-a}{\alpha} \right\} \wedge \left( \zeta - \frac{a-1}{\alpha} < \gamma, \ -1 < \zeta < \frac{1}{2} \right),
\]

\[
R_6 = \left\{ \xi = 0, \ \gamma > \zeta - \frac{a-1}{\alpha}, \ \zeta < \frac{1}{\alpha} - 2 \right\}.
\]

Remark 2.1. In the case \( h < 0 \), one can conduct similar computation as in (2.4)–(2.12) and obtain a different phase diagram than those of Figures 1–2–3, see Figures 4 and 5 below (note that regions \( R_1, R_2, R_3 \) are unchanged, since the range term is negligible in these regions). Let us stress that when \( h < 0 \), the “disorder” and “range” terms both play in the same direction and encourage exploration, resulting in a much simpler diagram: only range size exponents \( \xi \geq 1/2 \) are possible, see Section 4 below.
3 Main results

Our main results consist in proving the phase diagrams of Figures 1, 2, 3, with a precise description of the behavior of the polymer in each region. In order to state our results, let us introduce some definition.

**Definition 3.1.** If $(p_n)_{n \geq 0}$ is a sequence of positive real numbers, we say that $(S_n)_{0 \leq n \leq N}$ has range size of order $N$ under $P_{N,\beta_N,h_N}$ if

$$
\lim \sup_{n \to \infty} \frac{1}{N} \max_{1 \leq n \leq N} |S_n| \in [\eta, \frac{1}{2}] t_N = 1.
$$

If $(S_n)_{0 \leq n \leq N}$ has range size of order $N^\xi$ under $P_{N,\beta_N,h_N}$, then we say that the range size exponent is $\xi$.

In our results, we encounter doubly indexed processes

$$
Y_{u,v} := X_v^{(1)} + X_u^{(2)} - f(u,v) \quad \text{for } u, v \geq 0,
$$

where $X_v^{(1)}$ and $X_u^{(2)}$ are the Lévy processes of Notation 1 and $f$ is some deterministic function (for instance, a large deviation rate function), such that $(u,v) \to f(u,v)$ is continuous on the set where $f(u,v) < +\infty$, with $f(0,0) = 0$. We then have the following result about the maximizer of the variational problem $\sup_{u,v \geq 0} Y_{u,v}$, that we prove in Appendix A.3: it ensures the well-posedness of $\arg \max_{u,v \geq 0} Y_{u,v}$.

**Proposition 3.1.** Suppose that $P$-a.s. the variational problem $\sup_{u,v \geq 0} Y_{u,v}$ is positive with $Y$ defined in (3.1), and that $Y_{u,v} \to -\infty$ as $\max(u,v) \to +\infty$. Then

$$
P\left( \arg \max_{u,v \geq 0} Y_{u,v} \text{ is a singleton } \{ (\mathcal{U}, \mathcal{V}) \} \text{ with } \mathcal{U} \neq \mathcal{V} \right) = 1.
$$

Let us stress that in the case $\alpha = 2$ of a Brownian motion, [26, Lem. 2.6] (or [29]) proves the uniqueness of the maximizer for one-indexed processes, but not doubly-indexed ones.

### 3.1 Statement of the results

We now prove six different theorems, corresponding to the six possible regions in the phase diagram presented in Figure 1. We mention that we will use $\mathbb{P}$ only when the coupling is needed; otherwise we will keep the notation $P$ (in particular this will be the case when the limit of the rescaled log-partition function is non-random).

In this section, we again focus on the case $h > 0$, but several results hold for a general $h \in \mathbb{R}$: we will highlight when the results are specific to the case $h > 0$. The case $h < 0$ will be discussed separately in Section 4. Note that the case $h = 0$ or $\beta = 0$ can be recovered by taking $\zeta = +\infty$ or $\gamma = +\infty$ respectively, while the case of constant $h$ or $\beta$ can be recovered by taking $\zeta = 0$ or $\gamma = 0$ respectively. One can then recover Theorem 1.1 from Theorems 3.3, 3.4, 3.6 and 4.2 below.
Theorem 3.2 (Region 1). Assume that (1.4) holds with \( \hat{\beta} \geq 0, \hat{h} \in \mathbb{R} \) and

\[
\begin{align*}
\gamma > \frac{1}{\alpha} \quad &\text{and} \quad \zeta > \frac{1}{2}, \quad \text{if} \quad \alpha \in [\frac{1}{2}, 1) \cup (1, 2], \\
\gamma > \frac{1-\alpha}{\alpha} \quad &\text{and} \quad \zeta > \frac{1}{2}, \quad \text{if} \quad \alpha \in (0, \frac{1}{2}).
\end{align*}
\]

Then \( (S_n)_{0 \leq n \leq N} \) has range size of order \( \sqrt{N} \) under \( P^\xi_{N, \beta N, h N} \) (i.e. \( \xi = \frac{1}{2} \)) and we have the following convergence

\[
\lim_{N \to +\infty} Z^\xi_{N, \beta N, h N} = 1 \quad \mathbb{P}\text{-a.s.}.
\]

Moreover, we have \( \lim_{N \to +\infty} \| P^\xi_{N, \beta N, h N} - P \|_{TV} = 0 \) \( \mathbb{P}\text{-a.s.} \), where \( \| \cdot \|_{TV} \) is the total variation distance.

Note that the total variation convergence stated in Theorem 3.2 implies that \( S_N/\sqrt{N} \) converges in distribution to a Brownian motion. This convergence holds \( \mathbb{P}\)-almost surely.

Theorem 3.3 (Region 2). Assume that (1.4) holds with \( \hat{\beta} > 0, \hat{h} \in \mathbb{R} \) and

\[
\frac{1-\alpha}{\alpha} < \gamma < \frac{(2\alpha-1)\zeta-(\alpha-1)}{\alpha} \quad \text{and} \quad \alpha \in (\frac{1}{2}, 1) \cup (1, 2].
\]

Then \( (S_n)_{0 \leq n \leq N} \) has range size of order \( N^\xi \) under \( P^\xi_{N, \beta N, h N} \) with \( \xi = \frac{\alpha}{2\alpha-1}(1-\gamma) \in (\frac{1}{2}, 1) \), and we have the following convergence

\[
\lim_{N \to +\infty} \frac{1}{N^{1-\gamma}} \log Z^\xi_{N, \beta N, h N} = W_2 := \sup_{u,v \geq 0} \left\{ Y_{u,v}^{(2)} \right\} \in (0, +\infty), \quad \text{in} \ \hat{\mathbb{P}}\text{-probability},
\]

where \( Y_{u,v}^{(2)} = Y_{u,v}^{\hat{\beta}, (2)} = \hat{\beta}(X_{u}^{(1)} + X_{v}^{(2)}) - \frac{1}{2}(u \wedge v + u + v) \) is as defined in Theorem 1.1-(1B). Additionally, for any \( \varepsilon > 0 \) we have

\[
\lim_{N \to +\infty} P^\xi_{N, \beta N, h N} \left( \left| \frac{1}{N^\xi}(M_N^+, M_N^-) - (-U^{(2)}, V^{(2)}) \right| > \varepsilon \right) = 0, \quad \text{in} \ \hat{\mathbb{P}}\text{-probability},
\]

where \( (U^{(2)}, V^{(2)}) := \arg \max_{u,v \geq 0} \{ Y_{u,v}^{(2)} \} \) is well-defined thanks to Proposition 3.7.

Let us stress that the case \( \alpha = 2, \hat{\beta} = \hat{\beta}_N = \hat{\beta} > 0 \) and \( h = 0 \) corresponds to the case \( \gamma = 0 \) and \( \zeta = +\infty \): we find in that case that the range size exponent is \( \xi = \frac{2}{3} \).

Theorem 3.4 (Region 3). Assume that (1.4) holds with \( \hat{\beta} > 0, \hat{h} \in \mathbb{R} \) and

\[
\gamma < (\zeta - \frac{a-1}{\alpha}) \wedge \left( \frac{1-\alpha}{\alpha} \right) \quad \text{and} \quad \alpha \in (0, 1) \cup (1, 2].
\]

Then \( (S_n)_{0 \leq n \leq N} \) has range size of order \( N \) under \( P^\xi_{N, \beta N, h N} \) (i.e. \( \xi = 1 \)), and we have the following convergence

\[
\lim_{N \to +\infty} \frac{1}{N^{1-\gamma}} \log Z^\xi_{N, \beta N, h N} = W_3 := \sup_{u,v \geq 0} \left\{ Y_{u,v}^{(3)} \right\} \in (0, +\infty), \quad \hat{\mathbb{P}}\text{-a.s.},
\]

where \( Y_{u,v}^{(3)} = Y_{u,v}^{\hat{\beta}, (3)} = \hat{\beta}(X_{u}^{(1)} + X_{v}^{(2)}) - \infty 1_{\{u \wedge v+u+v>1\}} \) is as defined in Theorem 1.1-(2). Additionally, for any \( \varepsilon > 0 \) we have

\[
\lim_{N \to +\infty} P^\xi_{N, \beta N, h N} \left( \left| \frac{1}{N}(M_N^+, M_N^-) - (-U^{(3)}, V^{(3)}) \right| > \varepsilon \right) = 0, \quad \hat{\mathbb{P}}\text{-a.s.}
\]

where \( (U^{(3)}, V^{(3)}) := \arg \max_{u,v \geq 0} \{ Y_{u,v}^{(3)} \} \) is well-defined thanks to Proposition 3.7.
Theorem 3.5 (Region 4). Assume that \([1.4]\) holds with \(\hat{\beta} > 0, \hat{h} > 0\) and
\[
(\frac{2(\alpha-1)\gamma-\alpha-1}{\alpha}) \lor (\gamma - \frac{\alpha-1}{\alpha}) < \gamma < \left(\frac{2(\alpha+1)\gamma-\alpha-1}{\alpha}\right) \land \gamma \quad \text{and} \quad \alpha \in (1, 2],
\]
Then \(\{S_n\}_{0 \leq n \leq N}\) has range size of order \(N^{\xi}\) under \(P_{N,\hat{\beta},N,hN}\) with \(\xi = \frac{\alpha-1}{\alpha} (\gamma - \gamma) \in (0, 1)\), and we have the following convergence
\[
\lim_{N \to +\infty} \frac{1}{N^{\xi-\gamma}} \log Z_{N,\hat{\beta},N,hN}^{\xi} = W_{\xi} := \sup_{u,v \geq 0} \left\{ Y_{u,v}^{(4)} \right\} \in (0, +\infty), \quad \text{in } \hat{P}\text{-probability}, \tag{3.5}
\]
where \(Y_{u,v}^{(4)} = Y_{u,v}^{\hat{\beta},h,(4)} = \hat{\beta} (X_{u}^{(1)} + X_{v}^{(2)}) - \hat{h} (u+v)\). Additionally, for any \(\varepsilon > 0\) we have
\[
\lim_{N \to +\infty} P_{N,\hat{\beta},N,hN} \left( \left| \frac{1}{N^{\xi}} (M_{N}^{+} - M_{N}^{-}) - \left( -U^{(4)}, V^{(4)} \right) \right| > \varepsilon \right) = 0, \quad \text{in } \hat{P}\text{-probability}
\]
where \((U^{(4)}, V^{(4)}) := \arg \max_{u,v \geq 0} \left\{ Y_{u,v}^{(4)} \right\}\) is well-defined thanks to Proposition 2.7

Theorem 3.6 (Region 5). Assume that \([1.4]\) holds with \(\hat{\beta} > 0, \hat{h} > 0\) and
\[
\begin{cases}
\gamma > \left(\frac{2(\alpha-1)\gamma-\alpha-1}{\alpha}\right) \lor (\gamma - \frac{\alpha-1}{\alpha}) \land -1 < \gamma < \frac{1}{2}, & \text{if } \alpha \in (1, 2], \\
\gamma > \left(\frac{2(\alpha-1)\gamma-\alpha-1}{\alpha}\right) \land (\gamma - \frac{\alpha-1}{\alpha}) \land -1 < \gamma < \frac{1}{2}, & \text{if } \alpha \in \left(\frac{1}{2}, 1\right], \\
\gamma > \frac{1-(\alpha-1)}{\alpha} \land (\gamma - \frac{\alpha-1}{\alpha}) \land -1 < \gamma < \frac{1}{2}, & \text{if } \alpha \in (0, \frac{1}{2}).
\end{cases}
\]
Then \(\{S_n\}_{0 \leq n \leq N}\) has range size of order \(N^{\xi}\) under \(P_{N,\hat{\beta},N,hN}\) with \(\xi = \frac{1+\gamma}{3} \in (0, \frac{1}{2})\), and we have the following convergence
\[
\lim_{N \to +\infty} \frac{1}{N^{\xi-\gamma}} \log Z_{N,\hat{\beta},N,hN}^{\xi} = -\frac{3}{2} (\hat{h} \pi^{2/3} = \sup_{r \geq 0} \left\{ -\hat{h}r - \frac{\pi^{2}}{2r^{2}} \right\}) \quad \mathbb{P}\text{-a.s.} \tag{3.6}
\]
Additionally, for every \(\varepsilon > 0\), we have
\[
\lim_{N \to +\infty} P_{N,\hat{\beta},N,hN} \left( \left| \frac{1}{N^{\xi}} (M_{N}^{+} - M_{N}^{-}) - \pi^{2} \hat{h} \right| > \varepsilon \right) = 0 \quad \mathbb{P}\text{-a.s.}
\]

Theorem 3.7 (Region 6). Assume that \([1.4]\) holds with \(\hat{\beta} > 0, \hat{h} > 0\) and
\[
\begin{cases}
\gamma > \zeta \quad \text{and} \quad \zeta < -1, & \text{if } \alpha \in (1, 2], \\
\gamma > \zeta - \frac{\alpha-1}{\alpha} \quad \text{and} \quad \zeta < -1, & \text{if } \alpha \in (0, 1).
\end{cases}
\]
Then we have the following convergences
\[
\lim_{N \to +\infty} P_{N,\hat{\beta},N,hN} \left( |R_{N}| = 2 \right) = 1, \quad \lim_{N \to +\infty} N^{\xi} \log Z_{N,\hat{\beta},N,hN}^{\xi} = -2\hat{h} \quad \mathbb{P}\text{-a.s.} \tag{3.7}
\]

Let us conclude this section with a result that complements Theorems 3.3, 3.4 and Theorem 3.5 in the case \(\xi > \frac{1}{2}\). It shows that under \(P_{N,\hat{\beta},N,hN}\) trajectories travel ballistically to the closest point between \(-UN^{\xi}\) and \(VN^{\xi}\) and then to the other one. Let us introduce some notation to be able to state the result. For \(u,v \geq 0\) with \(u \neq v\), let \(\sigma_{u,v} = -1\) if \(u < v\) and \(\sigma_{u,v} = +1\) otherwise, let \(c_{u,v} = u \land v + u \lor v\), and define the function
\[
b_{u,v}(t) = \begin{cases}
\sigma_{u,v} c_{u,v} t & \text{for } 0 \leq t \leq \frac{u \lor v}{c_{u,v}}, \\
-\sigma_{u,v} c_{u,v} t + 2\sigma_{u,v} (u \lor v) & \text{for } \frac{u \lor v}{c_{u,v}} \leq t \leq 1,
\end{cases}
\]
that goes with constant speed from 0 to the closest point between \(-u\) and \(v\) and then to the other one. Now, for \(\varepsilon > 0\), let us define the event
\[
\mathcal{B}_{N}^{\varepsilon}(u,v) := \left\{ \sup_{t \in [0,1]} \left| \frac{1}{N^{\xi}} S_{[1N]} - b_{u,v}(t) \right| \leq \varepsilon \right\}. \tag{3.9}
\]
We then have the following result.
Proposition 3.8. Assume that, for some \( \xi \in (\frac{1}{2}, 1] \), for any \( \delta > 0 \) we have

\[
\lim_{N \to \infty} \mathbb{P}_{\hat{N}, \hat{\beta}, \hat{h}, h} \left( \left| \frac{1}{N \xi} (M_N^\xi, M_N^\xi) - \left( -U, V \right) \right| > \delta \right) = 0, \quad \text{in } \hat{\mathbb{P}}\text{-probability (resp. } \hat{\mathbb{P}}\text{-a.s.)},
\]

with \( U, V \geq 0 \) two random variables such that \( U \neq V \) a.s. Then, for any \( \varepsilon > 0 \) we have

\[
\lim_{N \to \infty} \mathbb{P}_{\hat{N}, \hat{\beta}, \hat{h}, h} \left( B_{\hat{N}}(U, V) \right) = 1, \quad \text{in } \hat{\mathbb{P}}\text{-probability (resp. } \hat{\mathbb{P}}\text{-a.s.)}.
\]

Remark 3.1. An analogous result should also hold in the case \( \xi \in (0, \frac{1}{2}) \). Assume that \( (3.10) \) holds with \( \xi \in (0, \frac{1}{2}) \). Then we expect that \( \frac{N^{-1}S_X + U}{U + V} \) converges in distribution (under \( \mathbb{P}_{\hat{N}, \hat{\beta}, \hat{h}, h} \)) towards a random variable \( X \) with density \( \frac{2}{\pi} \sin(\pi x) \mathbb{1}_{[0,1]}(x) \). This result is easy to obtain for a random walk conditioned to remain inside an interval \([ -aN^\xi, bN^\xi ]\), but becomes trickier when the range is conditioned to be exactly \([ -aN^\xi, bN^\xi ]\). We are not aware of any such result for random walks conditioned on their range, but let us mention [12] where a closely related question is considered. We therefore chose not to develop this in the present paper to avoid lengthening it.

3.2 Some comments on the results (case \( \hat{h} > 0 \))

Let us now make some observations on our results.

Comment 1. Our results describe a transition from folded trajectories \( (\xi < 1/2) \) to stretched trajectories \( (\xi > 1/2) \), which is induced by the presence of disorder. Let us illustrate this in the case \( \alpha \in (1,2] \) for simplicity; we refer to the phase diagram of Figure 18. If \( \beta_N = \hat{\beta} > 0 \) and \( h_N = \hat{h} > 0 \), that is if \( \gamma = \zeta = 0 \), we find that trajectories are folded, with range size exponent \( \xi = 1/3 \). Now, if we keep \( h_N = \hat{h} > 0 \) fixed (i.e. \( \zeta = 0 \)) and increase the strength of disorder, that is if we decrease \( \gamma \) (taking \( \gamma < 0 \)), we realize that we have transitions between the following regimes:

(i) if \( \gamma > \frac{1-\alpha}{3\alpha} \), the random walk is folded with range size exponent \( \xi = \frac{1}{3} \) (disorder is not strong enough);

(ii) if \( \frac{1-\alpha}{2\alpha} > \gamma > \frac{1-\alpha}{3\alpha} \), then the random walk is still folded, with range size exponent \( \frac{1}{3} < \xi = \frac{\gamma}{1-\alpha} < \frac{1}{2} \) (disorder makes the random walk less folded);

(iii) if \( \frac{1-\alpha}{2\alpha} > \gamma > \frac{1-\alpha}{\alpha} \), then the random walk is stretched, with range size exponent \( \frac{1}{2} < \xi = \frac{\gamma}{1-\alpha} < 1 \) (disorder is strong enough to stretch the random walk);

(iv) if \( \gamma < \frac{1-\alpha}{\alpha} \), then the random walk is completely unfolded and has range size exponent \( \xi = 1 \).

Analogously, if we keep \( \beta_N = \hat{\beta} > 0 \) fixed (i.e. \( \gamma = 0 \)) and decrease the penalty for the range, that is if we increase \( \zeta \) (taking \( \zeta > 0 \)), we have transitions between the following regimes:

(i) if \( 0 < \zeta < \frac{\alpha-1}{2\alpha+1} \), then the random walk is still folded with range size exponent \( \frac{1}{2} < \xi = \frac{\frac{1}{2} + \zeta}{\frac{1}{2} + \zeta} < \frac{\alpha}{2\alpha+1} < \frac{1}{2} \) (disorder plays no role);

(ii) if \( \frac{\alpha-1}{2\alpha+1} < \zeta < \frac{\alpha-1}{2\alpha} \), then the random walk is still folded with range size exponent \( \frac{\alpha}{2\alpha+1} < \xi = \frac{\zeta}{\alpha-1} < \frac{1}{2} \) (disorder plays a role);

(iii) if \( \frac{\alpha-1}{2\alpha} < \zeta < \frac{\alpha-1}{2\alpha-1} \), then the random walk is stretched, with range size exponent \( 1/2 < \xi = \frac{\gamma}{1-\alpha} < \frac{\alpha}{2\alpha-1} < 1 \) (disorder stretches the random walk);

(iv) if \( \zeta > \frac{\alpha-1}{2\alpha-1} \), then the random walk is stretched and has range size exponent \( \frac{5}{6} \leq \xi = \frac{\alpha}{2\alpha-1} < 1 \) (and the penalty for the range is not felt).
Comment 2. Let us now discuss the limiting distributions for the log-partition function in regions $R_2$, $R_3$, $R_4$. For simplicity, we will restrict ourselves to the case where $u = 0$ in the variational problems (3.3)-(3.4) (which corresponds to considering the case of a random walk constrained to stay non-negative): the variational problems become, respectively

\[
\hat{W}_2 := \sup_{v \geq 0} \left\{ \beta X_v - \frac{1}{2} v^2 \right\}, \quad \hat{W}_3 := \beta \sup_{v \in [0,1]} \{ X_v \}, \quad \hat{W}_4 := \sup_{v \geq 0} \left\{ \beta X_v - \hat{h} v \right\},
\]

with $(X_v)_{v \geq 0} = (X_v^{(1)})_{v \geq 0}$.

a) The variational problem $\hat{W}_3$ is clearly always finite. In the case $\alpha = 2$, $(X_t)_{t \geq 0}$ is a Brownian motion and it is standard to get that $\hat{W}_3$ has the distribution of $\beta |Z|$, with $Z \sim \mathcal{N}(0,1)$. In the case $\alpha \in (0,2)$, $(X_t)_{t \geq 0}$ is a stable Lévy process and we get that $\hat{W}_3$ is a positive $\alpha$-stable random variable (see [9, Ch. VIII], and also [27]).

b) The variational problem $\hat{W}_4$ is finite only when $\alpha > 1$: when $\alpha \in (0,1)$, then $X_v$ grows typically as $v^{1/\alpha} \gg v$ as $v \to \infty$ and we therefore have $\hat{W}_4 = +\infty$. This explains in particular why there is no energy-range balance possible if $\alpha \in (0,1)$ and why region $R_4$ no longer exists in that case. If $\alpha = 2$, $(X_t)_{t \geq 0}$ is a Brownian motion and it is standard to get that $\hat{W}_4$ is an exponential random variable (here with parameter $2h/\beta^2$). If $\alpha \in (1,2)$, $(X_t)_{t \geq 0}$ is a stable Lévy process and $(\beta X_t - \hat{h} t)_{t \geq 0}$ is also a Lévy process: the distribution of its supremum $\hat{W}_4$ has been studied extensively, going back to [2], but the exact distribution does not appear to be known (we refer to the recent papers [14, 28]).

c) The variational problem $\hat{W}_2$ is finite only when $\alpha > \frac{1}{2}$: when $\alpha \in (0,\frac{1}{2})$, then $X_v$ grows typically as $v^{1/\alpha} \gg v^2$ as $v \to \infty$ and we therefore have $\hat{W}_2 = +\infty$. This explains in particular why there is no energy-entropy balance possible if $\alpha \in (0,\frac{1}{2})$, and why region $R_2$ no longer exists in that case. In the case $\alpha = 2$, that is when $(X_t)_{t \geq 0}$ is a standard Brownian motion, then $\hat{W}_4$ has appeared in various contexts and its density is known (its Fourier transform is expressed in terms of Airy function, see for instance [17, 22]). In the case $\alpha \in (\frac{1}{2},2)$, exact asymptotics on the tail of the distribution of $\hat{W}_4$ have been derived in [30]; we are not aware whether the distribution of $\hat{W}_4$ has been studied in more detail.

Comment 3. To keep the paper lighter, we have chosen not to treat the cases of the boundaries between different regions of the phase diagrams. These boundary regions do not really hide anything deep: features of both regions should appear in the limit, and “disorder”, “range” and “entropy” may all compete at the same (exponential) scale. Let us state the limiting variational problems that one should find in some the most interesting boundary cases, in the case $\alpha \in (1,2]$ for simplicity (we refer to the phase diagram of Figure 1):

- Line between regions $R_2$ and $R_4$: $\gamma = \frac{(2 \alpha - 1) \zeta - (\alpha - 1)}{\alpha}$ and $\zeta \in (0,\frac{1}{2})$. Then one should have $\xi = \frac{\alpha(1-\gamma)}{2 \alpha - 2}$ and

\[
\lim_{N \to +\infty} \frac{1}{N^{2\zeta - 1}} \log Z_{N,\beta N,hN}^\omega = \sup_{u,v \geq 0} \left\{ \beta (X_u^{(1)} + X_v^{(2)}) - \hat{h}(u + v) - \frac{1}{2} (u \wedge v + u + v)^2 \right\}
\]

in $\mathbb{P}$-probability.

- Line between regions $R_4$ and $R_5$: $\gamma = \frac{(2 \alpha + 1) \zeta - (\alpha - 1)}{3 \alpha}$ and $\zeta \in (-1,\frac{1}{2})$. Then one should have $\xi = \frac{1+\zeta}{3}$ and

\[
\lim_{N \to +\infty} \frac{1}{N^{1-2\zeta}} \log Z_{N,\beta N,hN}^\omega = \sup_{u,v \geq 0} \left\{ \beta (X_u^{(1)} + X_v^{(2)}) - \hat{h}(u + v) - \frac{\pi}{2(u + v)^2} \right\}
\]

in $\mathbb{P}$-probability, where the last term inside the supremum comes from the entropic cost of “folding” the random walk in the interval $[uN^\zeta, vN^\zeta]$ (see Lemma A.3).

- Line between regions $R_2$ and $R_3$: $\gamma = -\frac{\alpha - 1}{\alpha}$ and $\zeta > 0$. Then one should have $\xi = 1$ and

\[
\lim_{N \to +\infty} \frac{1}{N^{2\zeta - 1}} \log Z_{N,\beta N,hN}^\omega = \sup_{u,v \geq 0} \left\{ \beta (X_u^{(1)} + X_v^{(2)}) - \kappa (u \wedge v + u + v) \right\}
\]
in \( \mathbb{P} \)-probability, where
\[
\kappa(t) := \begin{cases} 
\frac{1}{2}(1 + t) \log(1 + t) + \frac{1}{2}(1 - t) \log(1 - t), & \text{for } t \in [0, 1] \\
+\infty & \text{for } t > 1,
\end{cases}
\] (3.11)
is the rate function for the large deviations of the simple random walk, see Lemma A.2.

- Line between regions \( R_4 \) and \( R_4 \): \( \gamma = \zeta - \frac{2}{\alpha} \) and \( \zeta < 0 \). Then one should have \( \xi = 1 \) and
\[
\lim_{N \to +\infty} \frac{1}{N^{2\zeta-1}} \log Z_{N,\beta N,h N}^\ast = \sup_{u,v \geq 0} \left\{ \beta(X_u^{(1)} + X_v^{(2)}) - h(u + v) - \kappa(u \wedge v + u + v) \right\}
\]
in \( \mathbb{P} \)-probability.

**Comment 4.** In region \( R_5 \), the disorder term does not appear in the variational formula. In the case \( \beta = 0 \) and \( h > 0 \) (i.e. \( \gamma = \infty \), \( \zeta = 0 \)), corresponding to the random walk penalized by its range in a homogeneous way, the behavior of the random walk is well understood: it is confined in a segment of length \( (\pi \frac{2}{\gamma} h^{-\frac{\alpha}{2}})^{N^{1/3}} \) with a random center, see [33] for the continuum limit of the process. In our model, we have shown that trajectories are still confined in a segment of length \( (\pi \frac{2}{\gamma} h^{-\frac{\alpha}{2}})^{N^{1/3}} \). However, disorder should appear in the fluctuations of the log-partition function and in particular we believe that, depending on the strength \( \beta_N \) of the disorder interaction, the center of this segment should be determined by the environment so as to maximize the amount of potentials in that segment; in particular, it should not be random anymore (under \( P_{N,\beta N,h N}^\omega \), for typical realizations of \( \omega \)). This picture should hold in region \( R_5 \) as long as the effect of disorder is sufficiently strong. More precisely, using the terminology of Section 2, the “disorder” term is \( \beta N^\frac{\alpha}{2} - \gamma \), with \( \xi = \frac{1}{\beta}(1 + \zeta) \): its effect does not vanish as long as \( \gamma < \xi/\alpha \), that is as long as \( \gamma < \frac{1}{\alpha \beta}(1 + \zeta) \). In other words, there should be another phase transition inside region \( R_5 \): the random walk is confined in a segment of length \( (\pi \frac{2}{\gamma} h^{-\frac{\alpha}{2}})^{N^\frac{2\xi-1}{\alpha}} \) with \( \xi = \frac{1}{\alpha}(1 + \zeta) \), but under \( P_{N,\beta N,h N}^\omega \) the location of this segment should be non-random (i.e. determined by the realization of \( \omega \)) when \( \gamma < \frac{1}{\alpha \beta}(1 + \zeta) \), and random when \( \gamma > \frac{1}{\alpha \beta}(1 + \zeta) \) (which includes the case \( \beta = 0 \)). We leave this as an open problem.

4 Results in the case \( \hat{h} < 0 \)

4.1 The phase diagram

In the case \( \hat{h} < 0 \), the same type of “energy” vs. “range” vs. “entropy” heuristics as in Section 2 can be carried out. The main difference is that the “range” term is now a reward rather than a penalty and thus plays in the same direction as the “disorder” term and encourages stretching: the range size exponent will always verify \( \xi \geq 1/2 \). Recall that for a polymer with typical range size \( N^\xi \), the “range” term is of order \( N^{\xi - \zeta} \), the “disorder” term is of order \( N^{\xi/\alpha - \gamma} \) and the entropy term is \( N^{2\zeta-1} \) (since \( \xi \geq 1/2 \)), as in (2.2) and (2.3). In a similar fashion than in Section 2 we find that two cases need to be considered.

- **Case I** (“disorder” vs. “range”). As mentioned in Remark 2.7, regions \( R_1, R_2, R_3 \) are unchanged when \( h < 0 \): we refer to (2.4) – (2.5) – (2.6) for the determination of \( \xi \) in these three regions.

- **Case II** (“disorder” vs. “range”). The balance between range and entropy is achieved if \( \xi - \zeta = 2 \xi - 1 \) (with \( \xi \in [\frac{1}{2}, 1] \)), which gives \( \xi = 1 - \zeta \) when \( \gamma > \frac{2(\alpha - 1)\zeta - (\alpha - 1)}{\alpha} \). Also, we have \( \xi = 1 \) when \( \zeta \leq 0 \) and \( \gamma > \frac{\zeta}{\alpha} \), and we have \( \xi = 1/2 \) when \( \zeta \geq 1/2 \) and \( \gamma > \frac{1}{2\alpha} \).

To summarize, we can identify several regimes, according to the values of \( \gamma, \zeta \): there are five regimes when \( \alpha \in (\frac{1}{2}, 2] \), see Figure 4 below; there are four regimes when \( \alpha \in (0, \frac{1}{2}] \), see Figure 5 below.

4.2 Statement of the results

We only state the results in regions \( \hat{R}_4 \) and \( \hat{R}_5 \), since the regions \( R_1, R_2, R_3 \) are treated in Section 3.1 see Theorems 3.2, 3.3 and 3.4 (respectively).
\textbf{Theorem 4.1 (Region $\tilde{R}_4$).} Assume that (1.3) holds with $\beta > 0$, $\hat{h} < 0$ and
\[
\gamma > \frac{(2\alpha - 1)\zeta - (\alpha - 1)}{\alpha} \vee \frac{1 - \alpha}{\alpha}, \quad \zeta \in \left(0, \frac{4}{\alpha}\right).
\]
Then $(S_n)_{0 \leq n \leq N}$ has range size of order $N^\xi$ under $P_{N, \beta, hN}^\omega$ with $\xi = 1 - \zeta \in \left(\frac{4}{\alpha}, 1\right)$, and we have the following convergence
\[
\lim_{N \to +\infty} \frac{1}{N^{\xi - \zeta}} \log Z_{N, \beta, hN}^\omega = \frac{1}{2} \hat{h}^2 = \sup_{u, v \geq 0} \left\{ |\hat{h}|(v - u) - \frac{1}{2}(u \wedge v + u + v)^2 \right\} \quad \mathbb{P}\text{-a.s.} \quad (4.1)
\]
Additionally, let us consider, for $\varepsilon > 0$, the two events
\[
B_{N}^{+, \varepsilon} := \left\{ \sup_{t \in [0, 1]} |N^{-\xi}S_{t[N]} + \hat{h} t| \leq \varepsilon \right\}, \quad B_{N}^{-, \varepsilon} := \left\{ \sup_{t \in [0, 1]} |N^{-\xi}S_{t[N]} - \hat{h} t| \leq \varepsilon \right\},
\]
which corresponds to $(S_n)_{0 \leq n \leq N}$ travelling with roughly constant speed to either $-\hat{h}N^\xi$ or $\hat{h}N^\xi$. Then for any $\varepsilon > 0$, we have
\[
\lim_{N \to +\infty} \left( P_{N, \beta, hN}^\omega (B_{N}^{+, \varepsilon}) + P_{N, \beta, hN}^\omega (B_{N}^{-, \varepsilon}) \right) = 1 \quad \mathbb{P}\text{-a.s.},
\]
and $\hat{P}$-a.s.
\[
\lim_{\varepsilon \downarrow 0} \lim_{N \to +\infty} P_{N, \beta, hN}^\omega (B_{N}^{+, \varepsilon}) = \begin{cases} 
1 \left\{ (X_{h}^{(1)} > X_{h}^{(2)}) \right\} & \text{if } \gamma < \frac{1 - \xi}{\alpha}, \\
\exp \left( \beta X_{h}^{(1)} \right) & \text{if } \gamma = \frac{1 - \xi}{\alpha}, \\
\frac{1}{2} & \text{if } \gamma > \frac{1 - \xi}{\alpha}.
\end{cases} \quad (4.2)
\]
Before we state the result in region $\tilde{R}_5$ (which is somehow degenerate), let us state a result in the case $\zeta = 0$, that is at the boundary of regions $\tilde{R}_4$ and $\tilde{R}_5$. 

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Theorem 4.2 (Boundary $\tilde{R}_4$—$\tilde{R}_5$). Assume that (4.4) holds with $\beta > 0$, $\hat{h} < 0$ and with $\zeta = 0$, $\gamma = \frac{a-1}{a}$. Then we have the following convergence
\[
\lim_{N \to +\infty} \frac{1}{N} \log Z_{N,\beta N,h N}^\omega = \log (\sinh |\hat{h}|) = \sup_{u,v \geq 0} \left\{ |\hat{h}|(u + v) - \kappa(u \wedge v + u + v) \right\} \quad \text{P-a.s.,} \tag{4.3}
\]
with $\kappa(\cdot)$ defined in (3.11). Additionally, for $\varepsilon > 0$, let us consider the two events
\[
B_{N}^{+,\varepsilon} := \left\{ \sup_{t \in [0,1]} |N^{-1}S_{t[N]} - \tanh(|\hat{h}|)t| \leq \varepsilon \right\}, \quad B_{N}^{-,\varepsilon} := \left\{ \sup_{t \in [0,1]} |N^{-1}S_{t[N]} + \tanh(|\hat{h}|)t| \leq \varepsilon \right\},
\]
which corresponds to $(S_n)_{0 \leq n \leq N}$ travelling with roughly constant speed to either $\tanh(|\hat{h}|)N$ or $-\tanh(|\hat{h}|)N$. Then for any $\varepsilon > 0$, we have
\[
\lim_{N \to +\infty} \left( P_{N,\beta N,h N}^\omega (B_{N}^{+,\varepsilon}) + P_{N,\beta N,h N}^\omega (B_{N}^{-,\varepsilon}) \right) = 1 \quad \text{P-a.s.,}
\]
and $\hat{P}$-a.s.
\[
\lim_{\varepsilon \downarrow 0} \lim_{N \to +\infty} P_{N,\beta N,h N}^\omega (B_{N}^{+,\varepsilon}) = \begin{cases} 
1 & \text{if } \gamma < \frac{1}{a}, \\
\frac{1}{2} \left( \frac{\exp (\beta X_{1}^{(2)})}{\exp (\beta X_{1}^{(1)}) + \exp (\beta X_{1}^{(2)})} \right) & \text{if } \gamma = \frac{1}{a}, \\
\frac{1}{2} & \text{if } \gamma > \frac{1}{a}.
\end{cases}
\tag{4.4}
\]
To conclude, we state the result in region $\tilde{R}_5$.

Theorem 4.3 (Region $\tilde{R}_5$). Assume that (4.4) holds with $\beta > 0$, $\hat{h} < 0$ and $\zeta < 0$, $\gamma > \zeta - \frac{a-1}{a}$. Then
\[
\lim_{N \to +\infty} N^{\zeta-1} \log Z_{N,\beta N,h N}^\omega = |\hat{h}| \quad \text{P-a.s.}
\]
Additionally, for $\varepsilon > 0$, let us consider the two events
\[
B_{N}^{+,\varepsilon} := \left\{ \sup_{t \in [0,1]} |N^{-1}S_{t[N]} - t| \leq \varepsilon \right\}, \quad B_{N}^{-,\varepsilon} := \left\{ \sup_{t \in [0,1]} |N^{-1}S_{t[N]} + t| \leq \varepsilon \right\},
\]
which corresponds to $(S_n)_{0 \leq n \leq N}$ travelling with roughly constant speed to either $N$ or $-N$. Then, for any $\varepsilon > 0$, we have
\[
\lim_{N \to +\infty} \left( P_{N,\beta N,h N}^\omega (B_{N}^{+,\varepsilon}) + P_{N,\beta N,h N}^\omega (B_{N}^{-,\varepsilon}) \right) = 1 \quad \text{P-a.s.,}
\]
and $\hat{P}$-a.s.
\[
\lim_{\varepsilon \downarrow 0} \lim_{N \to +\infty} P_{N,\beta N,h N}^\omega (B_{N}^{+,\varepsilon}) = \begin{cases} 
1 & \text{if } \gamma < \frac{1}{a}, \\
\frac{1}{2} \left( \frac{\exp (\beta X_{1}^{(1)})}{\exp (\beta X_{1}^{(1)}) + \exp (\beta X_{1}^{(2)})} \right) & \text{if } \gamma = \frac{1}{a}, \\
\frac{1}{2} & \text{if } \gamma > \frac{1}{a}.
\end{cases}
\tag{4.5}
\]
If $\alpha \in (0,1)$ or if $\alpha \in (1,2]$ and $\gamma > \zeta$, then we can upgrade the result: we have
\[
\lim_{N \to +\infty} \left( P_{N,\beta N,h N}^\omega (S_N = N) + P_{N,\beta N,h N}^\omega (S_N = -N) \right) = 1 \quad \text{P-a.s.}
\tag{4.6}
\]
and (4.5) holds with $\{S_N = N\}$ in place of $B_{N}^{+,\varepsilon}$.
4.3 Further comments on the results in the case $\hat{h} < 0$

Comment 5. Notice that in Theorems 4.1, 4.2, and 4.3 the disorder term disappears in the limiting variational problems and the displacement of $S_N$ under $P_{N,\beta_N, h_N}$ is given by a law of large number, possibly with a random direction. Analogously to Comment 4 above, disorder should also appear in the fluctuations of the log-partition function and in the second order term for the displacement of $S_N$. For simplicity, let us comment further the case of the boundary $R_4 - R_5$ that is Theorem 4.2, i.e. consider the case where $h < 0$ is fixed ($\zeta = 0$) and $\beta_N = \hat{\beta} N^{-\gamma}$ with $\gamma > \frac{1-\alpha}{\alpha}$. In that case, the polymer has a (non-random) velocity $v_h := \tanh |h|$ either to the positive side or to the negative side: assume for simplicity that $X^{(1)}_N > X^{(2)}_N$, so that $\frac{1}{N} S_N$ converges to $v_h$ (and not $-v_h$). Randomness should then have the effect of stretching further (or back) the polymer: let us present some heuristic explanation on what one should expect. If we assume that $M_N = \omega u N^p$ and $M_q = v_h N + v N^p$, for some $\rho \in (\frac{1}{2}, 1)$ and $v \in \mathbb{R}$, $u \geq 0$, then, compared to the case $S_N = v_h N$:

(i) there is an additional range reward $|\hat{h}|(u + v) N^p$;
(ii) there is an additional entropic cost $\kappa'(v_h)(2u + v)N^p + \frac{1}{2} \kappa''(v_h)(2u + v)^2 N^{2p-1}$;

(iii) there is an energy gain of approximately $\hat{\beta}(\bar{X}^{(1)}_v + \bar{X}^{(2)}_v) N^\frac{\alpha}{\gamma} - \gamma$, with $\bar{X}^{(1)}, \bar{X}^{(2)}$ the scaling limits of the fields $(\omega_x)_{x \in \mathbb{Z}}$ on a scale $N^p$.

In particular, there are some cancellations between the range reward and the additional entropic cost: one gets $-\kappa'(v_h)u N^p - \frac{1}{2} \kappa''(v_h)(2u + v)^2 N^{2p-1}$. One should therefore take $u = 0$: if $u > 0$ the term $-\kappa'(v_h)u N^p$ cannot be compensated by $\hat{\beta} \bar{X}^{(1)}_v N^\frac{\alpha}{\gamma} - \gamma$, because $\gamma > \frac{1-\alpha}{\alpha}$. Therefore, for the entropy-energy balance, one has to compare $\frac{1}{2} \kappa''(v_h)u^2 N^{2p-1}$ with $\hat{\beta} \bar{X}^{(2)}_v N^\frac{\alpha}{\gamma} - \gamma$.

All together, this suggests that in the case $\gamma < \frac{1}{2\alpha}$ (which is compatible with $\gamma > \frac{1-\alpha}{\alpha}$ only for $\alpha \in (1, 2]$) it is possible to take $\rho := \frac{1-\alpha}{2\alpha - 1} \in (\frac{1}{2}, 1)$, so that $N^{2p-1} \sim N^\frac{\alpha}{\gamma} - \gamma$. Then, under $P_{N,\beta_N, h_N}$, one should have $M_N = o(N^p)$ and $M_q = v_h N - (1 + o(1))\sqrt{Np}$, where $V$ is the maximizer of the variational problem

$$\sup_{v \in \mathbb{R}} \left\{ \hat{\beta} \bar{X}^{(2)}_v - \frac{1}{2} \kappa''(v_h)v^2 \right\}.$$ 

On the other hand, when $\gamma > \frac{1}{2\alpha}$ then there is no further stretching of the polymer by the disorder: under $P_{N,\beta_N, h_N}$ we should have $S_N = v_h N + O(\sqrt{N})$, as it is the case when $\beta_N = 0$. This goes beyond the scope of this article and we leave this as an open problem.

Comment 6. There is some room for improvement in Theorem 4.3 when $\alpha \in (1, 2]$ and $\zeta \geq \gamma > \zeta - \frac{1-\alpha}{\alpha}$. Indeed, we then have that $\frac{1}{N} S_N$ goes to $\pm 1$, but as in Comment 5 above, disorder should appear in the fluctuations of the log-partition function and in the second order term for the displacement of $S_N$. Let us assume that we are on the event where $\frac{1}{N} S_N$ goes to $+1$ (instead of $-1$) and let us present a heuristic explanation on what one should expect. If we assume that $M_N = -u N^p$ and $M_q = N - v N^p$, for some $\rho \in (0, 1)$ and $u, v \geq 0$ with $2u \leq v$ (because of the constraint $M_q - 2M_N \leq N$), then compared to the case $S_N = N$:

(i) there is a diminution of the range reward by $|\hat{h}|(u - v) N^{p-\zeta}$;
(ii) there is a reduction of the entropic cost by roughly $N^p \log N$, coming from the combinatorial term $\binom{N}{N-\epsilon N^p}$ — this is negligible compared to the range term since $\zeta < 0$;

(iii) there is an energy gain of approximately $\hat{\beta}(\bar{X}^{(2)}_v - \bar{X}^{(1)}_v) N^\frac{\alpha}{\gamma} - \gamma$, where $\bar{X}^{(1)}_v, \bar{X}^{(2)}_v$ are the scaling limits of $(\omega_x)_{x \leq 0}, (\omega_{N-x})_{x \geq 0}$ on a scale $N^p$.

This suggests that for $\alpha \in (1, 2]$ and $\gamma > \gamma > \zeta - \frac{1-\alpha}{\alpha}$, one should take $\rho = \frac{\alpha}{\alpha - 1}(\zeta - \gamma) \in (0, 1)$, so that $N^{p-\zeta} \sim N^\frac{\alpha}{\gamma} - \gamma$. Note that one recovers $\rho = 0$ if $\gamma = \zeta$ (to be compared with (4.10)) and $\rho = 1$ if
\[ \gamma = \zeta - \frac{d - 1}{\alpha} \]  
(i.e. on the boundary of Regions \( R_5 - R_3 \), see Figure 1). Additionally, under \( P_{N, \beta_N, h_N}^x \), one should have \( M_N^x = -(1 + o(1))U N^p \) and \( M_N^x = N - (1 + o(1))V N^p \), where \((U, V)\) are the maximizers of the variational problem  

\[ \sup_{0 \leq a \leq v} \left\{ \beta \left( \hat{X}_u^{(2)} - \hat{X}_v^{(1)} \right) - |\hat{h}|(v - u) \right\}. \]

As for Comment 5, we leave this as an open problem.

### 4.4 Organisation of the proof and useful notation

Let us give an overview of how the rest of the paper is organized:

- In Section 4 we start with the proof of Proposition 3.8 then we prove Theorems 3.2 to 3.7 (in that order), i.e. we prove the phase diagram of Regions \( R_1 \) to \( R_6 \); note that regions \( R_4 \) to \( R_6 \) are specific to the case \( \hat{h} > 0 \). The results in Regions \( R_2, R_4 \) and \( R_5 \) involve competitions between “energy”, “range” or “entropy” (but all have the same scheme of proof), while Regions \( R_1, R_3 \) and \( R_6 \) are extreme cases where only one factor is significant and hence are much simpler. Let us stress here that the statements on range size of trajectories \((S_n)_{0 \leq n \leq N}\) under \( P_{N, \beta_N, h_N}^x \) are direct consequences of the convergence of \((M_N, M_N^x)\) under \( P_{N, \beta_N, h_N}^x \), so we do not write their proof explicitly.

- In Appendix A we prove the remaining Theorems 1.1 to 1.3, i.e. we complete the phase diagram in the case \( \hat{h} < 0 \). Here, the main contribution to the partition function comes from the range term and finding the limit of the rescaled log-partition function is not difficult. The harder part consists in showing that disorder plays a role in deciding whether the random walk moves to the positive or to the negative side: this is done by a careful decomposition of the partition function.

- In Appendix A we regroup several technical estimates: large deviations for the range of the random walk in Section A.1, deviation for sums of \( \omega_x \) (i.e. the proof of Lemma 4.4 below) in Section A.2 the proof of Proposition 3.8 in Section A.3 and some technical estimate on càdlàg path in Section A.4.

### Some further notation and a useful lemma

In the rest of the paper, to lighten notation, we will drop the dependence on \( \beta_N \) and \( h_N \): we write \( P_N^x \) instead of \( P_{N, \beta_N, h_N}^x \) and \( Z_N^x \) instead of \( Z_{N, \beta_N, h_N}^x \). We also use the convenient notation \( Z_N^x(E) \) for the partition function restricted to trajectories \((S_n)_{n \geq 0}\) in \( E \): more precisely,

\[ Z_N^x(E) := E \left[ \exp \left( \sum_{x \geq 0} (\beta_N \omega_x - h_N) \mathbb{1}_{\{x \in R_N^x\}} \right) \mathbb{1}_E \right]. \]  

(4.7)

This way, we have that \( P_N^x(E) = Z_N^x(E)/Z_N^x \).

For any \( j \geq 0 \) let us recall the notation \( \Sigma^+_j := \sum_{x=0}^j \omega_x \) and \( \Sigma^-_j := \sum_{x=-j}^{0} \omega_x \), introduced in (1.2) (with the convention that \( \Sigma^-_0 = 0 \)). We then let

\[ \Sigma^e_\ell := \sup_{0 \leq j < \ell} |\Sigma^-_j| + \sup_{0 \leq j < \ell} |\Sigma^+_j|. \]  

(4.8)

Recall that we have set \( M_N^+ := \max_{0 \leq n \leq N} S_n \geq 0 \) and \( M_N^- := \min_{0 \leq n \leq N} S_n \leq 0 \) the right-most and left-most points of the random walk after \( N \) steps; we also denote

\[ M^e_N := \max_{0 \leq n \leq N} |S_n| = \max(M_N^+, -M_N^-). \]

With these notation, notice that we have \( \sum_{x \in R_N} \omega_x = \Sigma^+_M + \Sigma^-_{-M} \). We now state the following (standard) lemma, that we prove in Appendix A.2 for completeness.

**Lemma 4.4.** Let \( \Sigma^e_\ell \) defined as in (4.8). Then, under Assumption 7 (\( \alpha \in (0, 1) \cup (1, 2] \)), there exists a constant \( c \in (1, +\infty) \) such that for any \( T > 0 \) and any \( \ell \) we have

\[ \mathbb{P}(\Sigma^e_\ell > T) \leq c \ell T^{-\alpha}. \]  

(4.9)

Also, \( \mathbb{P}\text{-a.s.} \) there is a constant \( C = C(\omega) \) such that \( \Sigma^e_\ell \leq C \ell^{1/\alpha} (\log_2 \ell)^{2/\alpha} \) for all \( \ell \geq 1 \).

Finally, while we keep the distinction between \( \mathbb{P} \) and \( \mathbb{P} \), we will only write \( \omega \) (and not \( \hat{\omega} \)) in order to lighten the notation.
5 Proof of the main results

5.1 Ballisticity of trajectories: proof of Proposition 3.8

Let $\varepsilon > 0$ be fixed. For $\delta > 0$, let us define (recall we assume $\xi > \frac{1}{2}$)

$$A_N^\delta := \left\{ \left| \frac{1}{N^\xi} (M_N^-, M_N^+) - (-\mathcal{U}, \mathcal{V}) \right| \leq \delta \right\}.$$  

Then, by assumption, we have that for any $\delta > 0$, $\lim_{N \to +\infty} \frac{Z_N^\varepsilon(A_N^\delta)}{Z_N^\varepsilon} = 1$ in $\tilde{\mathbb{P}}$-probability (resp. $\tilde{\mathbb{P}}$-a.s.), so the proof will be complete if we show that

$$\lim_{N \to +\infty} \frac{Z_N^\varepsilon(A_N^\delta, B_N^\varepsilon(\mathcal{U}, \mathcal{V}^c))}{Z_N^\varepsilon(A_N^\delta)} = 0, \quad \tilde{\mathbb{P}}\text{-a.s.} \quad (5.1)$$

where we refer to (3.9) for the definition of $B_N^\varepsilon(\mathcal{U}, \mathcal{V})$.

To do so, we decompose the partition function as follows (here and in the rest of the paper, we often omit integer parts for simplicity):

$$Z_N^\varepsilon(A_N^\delta, B_N^\varepsilon(\mathcal{U}, \mathcal{V}^c)) = \sum_{|x-y| - N(\mathcal{U}, \mathcal{V})| \leq \delta} Z_N^\varepsilon(M_N^-=x, M_N^+=y, B_N^\varepsilon(\mathcal{U}, \mathcal{V}^c))$$

Now, for $\delta > 0$ small enough, we have thanks to (A.35) (recall $\mathcal{U}, \mathcal{V} \geq 0$ with $\mathcal{U} \neq \mathcal{V}$) that

$$\mathbb{P}\left(M_N^- = -x, M_N^+ = y, B_N^\varepsilon(\mathcal{U}, \mathcal{V}^c)\right) \leq e^{-c N 2^\xi - 1} \mathbb{P}\left(M_N^- = -x, M_N^+ = y\right) \quad (5.2)$$

uniformly for $x, y$ such that $|(|x, y| - N(\mathcal{U}, \mathcal{V})| \leq \delta N^\xi$, for some constant $c = c_{\delta, \varepsilon}(\mathcal{U}, \mathcal{V}) > 0$.

Using (5.2), we obtain that

$$Z_N^\varepsilon(A_N^\delta, B_N^\varepsilon(\mathcal{U}, \mathcal{V}^c)) \leq e^{-c N 2^\xi - 1} \sum_{|x-y| - N(\mathcal{U}, \mathcal{V})| \leq \delta} e^{\beta N (\Sigma_x + \Sigma_y) - h_N(x+y+1)} \mathbb{P}\left(M_N^- = -x, M_N^+ = y\right)$$

$$\leq e^{-c N 2^\xi - 1} \sum_{|x-y| - N(\mathcal{U}, \mathcal{V})| \leq \delta} Z_N^\varepsilon(M_N^- = -x, M_N^+ = y) \leq e^{-c N 2^\xi - 1} Z_N^\varepsilon(A_N^\delta),$$

which shows (5.1). \hfill \Box

5.2 Region R1: Proof of Theorem 3.2

Recall that in Region $R_1$ we have

$$\begin{cases} \gamma > \frac{1}{2\alpha} & \text{and} \quad \zeta > \frac{1}{2}, & \text{if} \quad \alpha \in \left[\frac{1}{2}, 1\right) \cup (1, 2], \vspace{1em} \\
\gamma > \frac{1-\alpha}{\alpha} & \text{and} \quad \zeta > \frac{1}{2}, & \text{if} \quad \alpha \in (0, \frac{1}{2}).\vspace{1em} \\
\end{cases}$$

Let us note that we always have $\gamma > \frac{1}{2\alpha}$, since $\frac{1-\alpha}{\alpha} > \frac{1}{2\alpha}$ when $\alpha < 1/2$.

Convergence of the partition function

Fix $A$ (large) and split the partition function in the following way

$$Z_N^\varepsilon = Z_N^\varepsilon(M_N^x \leq A \sqrt{N}) + Z_N^\varepsilon(M_N^x > A \sqrt{N}). \quad (5.3)$$
Upper bound. Recalling the definition (4.7) of the restricted partition function, one easily sees that
\[
Z_N^\circ(M_N^+ \leq A\sqrt{N}) \leq \exp \left( \beta N^{-\gamma} \Sigma_{A\sqrt{N}}^* + 2A|h|N^{\frac{\gamma}{2}-\zeta} \right) 
\]
\[
\leq \exp \left( \beta N^{-\gamma} \Sigma_{A\sqrt{N}}^* + 2A|h|N^{\frac{\gamma}{2}-\zeta} \right) 
\]
(5.4)

Notice that \(N^{\frac{\gamma}{2}-\zeta}\) goes to 0 as \(N \to \infty\), since \(\zeta > \frac{1}{2}\). Also, by Lemma 4.4 since \(\gamma > \frac{1}{2}\) we get that \(N^{-\gamma} \Sigma_{A\sqrt{N}}^*\) goes to 0 almost surely. We therefore get that \(\limsup_{N \to \infty} Z_N^\circ(M_N^+ \leq A\sqrt{N}) \leq 1\) \(\mathbb{P}\)-a.s.

It remains to show that the second term in (5.3) is also small. We split the partition function as
\[
Z_N^\circ(M_N^+ > A\sqrt{N}) \leq \sum_{k=1}^{\log_2(\frac{1}{\sqrt{N}})} Z_N^\circ(M_N^+ \in (2^{k-1}A\sqrt{N}, 2^kA\sqrt{N}))
\]
\[
\leq \sum_{k=1}^{\log_2(\frac{1}{\sqrt{N}})} \exp \left( \beta N^{-\gamma} \Sigma_{2^kA\sqrt{N}}^* + 2^{k+1}A|h|N^{\frac{\gamma}{2}-\zeta} \right) P(M_N^+ \geq 2^{k-1}A\sqrt{N}).
\]

Then, it is standard to get that \(P(M_N^+ > x) \leq 2 \exp(-\frac{x^2}{2\gamma})\) for any \(x > 0\) and \(N \in \mathbb{N}\) (thanks to Lévy’s inequality and a standard Chernov bound), so that
\[
P(M_N^+ > 2^{k-1}A\sqrt{N}) \leq 2 \exp(-2^{2k-3}A^2).
\]
(5.5)

Hence, choosing \(N\) large enough so that \(|h|N^{\frac{\gamma}{2}-\zeta} \leq 2^{-5}A\) and in particular \(2^{k+1}A|h|N^{\frac{\gamma}{2}-\zeta} \leq 2^{2k-4}A^2\), we get \(\mathbb{P}\)-a.s.
\[
Z_N^\circ(M_N^+ > A\sqrt{N}) \leq \sum_{k=1}^{\log_2(\frac{1}{\sqrt{N}})} 2^{2k} \exp \left( C(\hat{\beta}, A, \omega)2^{k/\alpha}N^{2\gamma/(\alpha\gamma)} - 2^{2k-4}A^2 \right),
\]
(5.6)

where we also used Lemma 4.4 to get an almost sure bound on \(\Sigma_{2^kA\sqrt{N}}^*\). Now, note that uniformly for \(k\) in the sum,
\[
\frac{2^{k/\alpha}N^{2\gamma/(\alpha\gamma)} - 2^{2k-4}A^2}{2^{2k}} \leq (\log_2 N)^{2/\alpha} \times \begin{cases} N^{2\gamma/(\alpha\gamma)} & \text{if } \alpha \in \left[\frac{1}{2}, 2\right], \\ A^{-\frac{1-2\alpha}{\alpha}N^{\frac{1}{2}-\frac{\gamma}{\alpha}}} & \text{if } \alpha \in (0, \frac{1}{2}). \end{cases}
\]

This (uniform) upper bound goes to 0 as \(N \to \infty\), because \(\gamma > \frac{1}{2}\) if \(\alpha \geq \frac{1}{2}\) and \(\gamma > \frac{1-2\alpha}{\alpha}\) if \(\alpha \in (0, \frac{1}{2})\).

All together, we get that \(\mathbb{P}\)-a.s., for \(N\) large enough,
\[
Z_N^\circ(M_N^+ > A\sqrt{N}) \leq \sum_{k=1}^{\log_2(\frac{1}{\sqrt{N}})} 2 \exp \left( -2^{2k-5}A^2 \right) \leq C \exp(-A^2/8).
\]

We have therefore proven that for any \(A > 0\), \(\limsup_{N \to \infty} Z_N^\circ \leq 1 + Ce^{-A^2/8}\). Since \(A\) is arbitrary, this gives that \(\limsup_{N \to \infty} Z_N^\circ \leq 1\).

Lower bound. For the lower bound, we use that
\[
\hat{\beta} N^{-\gamma} \Sigma_{A\sqrt{N}}^* + 2A|h|N^{\frac{\gamma}{2}-\zeta} \to 0 \text{ almost surely.}
\]
Using that \(P(M_N^+ \leq A\sqrt{N}) \geq 1 - 2e^{-A^2/2}\), we therefore get that \(\liminf_{N \to \infty} Z_N^\circ \geq 1 - 2e^{-A^2/2} \mathbb{P}\)-a.s. Since \(A\) is arbitrary, this gives \(\liminf_{N \to \infty} Z_N^\circ \geq 1\) \(\mathbb{P}\)-a.s., which concludes the proof. \(\square\)
Convergence in total variation distance

We show that for any \( \varepsilon \in (0, \frac{1}{8}) \),

\[
\lim \sup_{N \to \infty} \sup_{B} \left| P_N^\infty(B) - P(B) \right| < 5\varepsilon \quad \text{P-a.s.,}
\]  
(5.8)

where \( B \) ranges over all P-measurable sets. This implies the convergence from \( P_N^\infty \) to \( P \) in total variation distance, since \( \varepsilon \) is arbitrary.

Let \( \mathcal{C}_{N,\varepsilon} := \{ \omega : |Z_N^\infty - 1| < \varepsilon \} \): we have proven above that P-a.s. \( \lim_{N \to \infty} \mathds{1}_{\mathcal{C}_{N,\varepsilon}} = 1 \). Note that \( P_N^\infty(B) = Z_N^\infty(B)/Z_N^\infty \). Hence, on the event \( \mathcal{C}_{N,\varepsilon} \), we have

\[
\frac{1}{1 + \varepsilon} (Z_N^\infty(B) - P(B)) - \frac{\varepsilon}{1 + \varepsilon} < P_N^\infty(B) - P(B) < \frac{1}{1 - \varepsilon} (Z_N^\infty(B) - P(B)) + \frac{\varepsilon}{1 - \varepsilon},
\]

where we also used that \( P(B) \leq 1 \). Therefore, to prove \( \text{(5.8)} \), we need to show that \( \lim \sup_{N \to \infty} \sup_{B} |Z_N^\infty(B) - P(B)| < \varepsilon, \text{P-a.s.} \). For \( A > 0 \), we have that

\[
|Z_N^\infty(B) - P(B)| \leq \left| Z_N^\infty(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) - P(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) \right| + Z_N^\infty(M_N^\infty > A\sqrt{N}) + P(M_N^\infty > A\sqrt{N}).
\]

As seen above, we have \( \lim \sup_{N \to \infty} Z_N^\infty(M_N^\infty > A\sqrt{N}) \leq C e^{-A^2/8} \) almost surely, and also \( P(M_N^\infty > A\sqrt{N}) \leq 2e^{-A^2/2} \); these two terms can be made arbitrarily small by taking \( A \) large. Hence it is enough to show that for any \( A > 0 \)

\[
\lim \sup_{N \to \infty} \sup_{B} \left| Z_N^\infty(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) - P(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) \right| = 0 \quad \text{P-a.s.}
\]

Analogously to \( \text{(5.4)} \) and \( \text{(5.7)} \) we have, for any measurable \( B \)

\[
Z_N^\infty(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) \leq \exp \left( \hat{\beta} N^{-\gamma} \Sigma_{A\sqrt{N}}^\infty + 2A|\hat{h}|N^{1/2 - \zeta} \right) P(B \cap \{ M_N^\infty \leq A\sqrt{N} \}),
\]

\[
Z_N^\infty(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) \geq \exp \left( -\hat{\beta} N^{-\gamma} \Sigma_{A\sqrt{N}}^\infty - 2A|\hat{h}|N^{1/2 - \zeta} \right) P(B \cap \{ M_N^\infty \leq A\sqrt{N} \}),
\]

so \( |Z_N^\infty(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) - P(B \cap \{ M_N^\infty \leq A\sqrt{N} \})| \) is bounded by

\[
\left| \exp \left( \hat{\beta} N^{-\gamma} \Sigma_{A\sqrt{N}}^\infty + 2A|\hat{h}|N^{1/2 - \zeta} \right) - 1 \right| + \left| \exp \left( -\hat{\beta} N^{-\gamma} \Sigma_{A\sqrt{N}}^\infty - 2A|\hat{h}|N^{1/2 - \zeta} \right) - 1 \right|,
\]

where we also bounded \( P(B \cap \{ M_N^\infty \leq A\sqrt{N} \}) \) by \( 1 \). Since \( \hat{\beta} N^{-\gamma} \Sigma_{A\sqrt{N}}^\infty + 2A|\hat{h}|N^{1/2 - \zeta} \) goes to 0 almost surely (see Lemma \ref{lemma:bound}), this concludes the proof.

\[ \square \]

5.3 Region R2: proof of Theorem 3.3

Recall that in Region R2 we have

\[ 2\xi - 1 = \frac{\xi}{\alpha} - \gamma > \xi - \zeta \quad \text{with} \quad \alpha \in \left(\frac{1}{2}, 1\right) \cup (1, 2], \]

and that this region does not exist when \( \alpha < 1/2 \). We prove that the range size is of order \( N^\xi \) with \( \xi = \frac{\alpha}{2\alpha - 1}(1 - \gamma) \).

Convergence of the rescaled log-partition function

We fix some \( A \) large and we split the partition function as

\[
Z_N^\infty = Z_N^\infty \left( M_N^\infty \leq AN^\xi \right) + Z_N^\infty \left( M_N^\infty > AN^\xi \right).
\]  
(5.9)

The proof of the convergence is divided into three steps: (1) we show that after taking logarithm and dividing by \( N^{2\xi - 1} \), the first term converges to some random variable \( W_2^A \) as \( N \to \infty \); (2) we show that the second term is small compared to the first one; (3) we let \( A \to \infty \) and we observe that \( W_2^A \) converges to \( W_2 \).
Step 1. We prove the following lemma.

**Lemma 5.1.** In Region $R_2$, we have that $\hat{P}$-a.s., for any $A \in \mathbb{N}$,

$$\lim_{N \to \infty} \frac{1}{N^{2\xi-1}} \log Z^x_N \left(M^x_N \leq AN^\zeta \right) = \mathcal{W}^x_2 := \sup_{u,v \in [0,A]} \left\{ \hat{\beta}(X^{(1)}_u + X^{(2)}_v) - I(u,v) \right\},$$

with $(X^{(1)}_u, X^{(2)}_u)_{u,v \geq 0}$ from Notation \(\square\) and $I(u,v) := \frac{1}{2}(u \wedge v + u + v)^2$.

**Proof.** Let us fix $\delta > 0$ and write

$$Z^{w,\leq}_{\xi} := Z^{w}_{\xi} \left(M^w_N \leq AN^\zeta \right) = \sum_{k_1=0}^{[A/\delta]} \sum_{k_2=0}^{[A/\delta]} Z^{w}_{\xi}(k_1, k_2, \delta),$$  \hspace{1cm} (5.10)

where we define

$$Z^{w}_{\xi}(k_1, k_2, \delta) := Z^{w}_{\xi} \left(M^w_N \in (- (k_1 + 1)\delta N^\zeta, - k_1 \delta N^\zeta], M^w_N \in [k_2 \delta N^\zeta, (k_2 + 1)\delta N^\zeta] \right)$$

(5.11)

(recall $M^w_N := \min_{0 \leq n \leq N} S_n$ and $M^w_N := \max_{0 \leq n \leq N} S_n$). Since there are at most $(A/\delta)^2$ terms in the sum, we easily get that

$$\max_{0 \leq k_1, k_2 \leq A/\delta} \log Z^{w}_{\xi}(k_1, k_2, \delta) \leq \log Z^{w,\leq}_{\xi} \leq 2 \log(A/\delta) + \max_{0 \leq k_1, k_2 \leq A/\delta} \log Z^{w}_{\xi}(k_1, k_2, \delta).$$

Upper bound. An upper bound on $\log Z^{w}_{\xi}(k_1, k_2, \delta)$ is given by

$$\beta_N \left( \Sigma^-_{\{k_1, k_2\} \delta N^\zeta} + \Sigma^+_{\{k_2, k_1\} \delta N^\zeta} \right) + \beta_N R^\delta_N(k_1, k_2, \delta) + |\hat{h}|(k_1 + k_2 + 2)\delta N^\zeta - \zeta + p^\delta_N(k_1, k_2, \delta),$$

where for $u, v \geq 0$ we defined

$$R^\delta_N(u, v) := \max_{u \leq \xi < j \leq (u + \delta)N^\zeta - 1} \left| \Sigma^-_{\{u, \xi\}} \right| + \max_{v \leq \xi < j \leq (v + \delta)N^\zeta - 1} \left| \Sigma^+_{\{v, \xi\}} \right|,$$

and

$$p^\delta_N(u, v) := \log \mathbf{P} \left( M^-_N \in (- (u + \delta)N^\zeta, - uN^\zeta], M^+_N \in [vN^\zeta, (v + \delta)N^\zeta] \right).$$

Let us write $u = k_1 \delta, v = k_2 \delta$ and set $U_0 = U_0(A) = \{0, \delta, 2\delta, \ldots, A\}$: using that $2\xi - 1 = \xi/\alpha - \gamma$, we get that

$$\max_{0 \leq k_1, k_2 \leq A/\delta} \frac{\log Z^{w}_{\xi}(k_1, k_2, \delta)}{N^{2\xi-1}} \leq \max_{u,v \in U_0} \left\{ \beta N^{\xi - \zeta} \left( \Sigma^-_{\{u, N^\zeta\}} + \Sigma^+_{\{v, N^\zeta\}} \right) + \beta N^{\xi - \zeta} R^\delta_N(u, v) \right. \right.$$  \hspace{1cm} (5.15)

$$\left. + |\hat{h}|(u + v + 2\delta)N^{\xi - \zeta} - (2\xi - 1)N^{\zeta - \zeta} - N^{(2\xi - 1) - \zeta} + N^{(2\xi - 1) - \zeta} p^\delta_N(u, v) \right\}.$$

It is easy to see that the third term in the maximum goes to 0 uniformly in $u, v$, since $u + v + 2\delta < 3A$ and since $\xi - \zeta < 2\xi - 1$ in Region $R_2$. Note that thanks to the coupling introduced in Section 1.2 we have that $(N^{-\xi/\alpha} \Sigma^-_{\{u, N^\zeta\}})_{u \in [0,A+\delta]}$ and $(N^{-\xi/\alpha} \Sigma^+_{\{v, N^\zeta\}})_{v \in [0,A+\delta]}$ converge $\hat{P}$-a.s. in the Skorokhod topology to two independent Lévy processes $(X^{(2)}_u)_{u \in [0,A+\delta]}$ and $(X^{(1)}_v)_{v \in [0,A+\delta]}$ (of Notation \(\square\)).

Note also that thanks to Lemma \(\Pi\) (see (A.2)) we have

$$\lim_{N \to \infty} N^{-(2\xi - 1)} p_N(u, v, \delta) = -I(u, v), \quad \text{with} \quad I(u, v) := \frac{1}{2}(u \wedge v + u + v)^2, \quad u, v \geq 0.$$

Since $U_0$ is a finite set, by (1.10) in [32, the limiting Lévy processes $X^{(1)}_u$ and $X^{(2)}_u$ are $\hat{P}$-a.s. continuous at every point in $U_0$. Hence, thanks to Lemma \(\Pi\) $\hat{P}$-a.s., for any $\varepsilon > 0$ there is a random integer $N_0 = N_0(\varepsilon, \delta, \omega)$, such that for all $N \geq N_0$,

$$N^{-\frac{\xi}{2\alpha}} R^\delta_N(u, v) \leq 2\varepsilon + \sup_{u \leq u' \leq u + \delta + \varepsilon} |X^{(2)}_u - X^{(2)}_{u'}| + \sup_{v \leq v' \leq v + \delta + \varepsilon} |X^{(1)}_v - X^{(1)}_{v'}|,$$

$$|N^{-\frac{\xi}{2\alpha}} \Sigma^-_{\{u, N^\zeta\}} - X^{(2)}_u| \leq \varepsilon \quad \text{and} \quad |N^{-\frac{\xi}{2\alpha}} \Sigma^+_{\{v, N^\zeta\}} - X^{(1)}_v| \leq \varepsilon,$$
uniformly for all \( u, v \in U_\delta \). Now letting \( N \to \infty \) and then \( \varepsilon \to 0 \), we readily have that the lim sup as \( N \to +\infty \) of the right-hand side of (5.15) is \( \hat{P} \)-a.s. smaller than
\[
\hat{W}_{2}^{A,\delta} := \max_{u, v \in U_\delta} \left\{ \hat{\beta}(X_u^{(2)} + X_v^{(1)}) + \hat{\beta} \sup_{0 \leq t \leq \delta} |X_{u+t}^{(2)} - X_u^{(2)}| + \hat{\beta} \sup_{0 \leq t \leq \delta} |X_{v+t}^{(1)} - X_v^{(1)}| - I(u, v) \right\}. \tag{5.16}
\]

**Lower bound.** On the other hand, a lower bound on \( \log Z_N^\varepsilon(k_1, k_2, \delta) \) is given by
\[
\beta_N \left( \Sigma_{[k_1, N^\varepsilon]} + \Sigma_{[k_2, N^\varepsilon]} \right) - \beta_N R_N^{\varepsilon}(k_1, k_2, \delta) - |\mathring{h}|(k_2 + k_1 + 2)\delta N^\varepsilon + p_R^\varepsilon(k_1, k_2, \delta).
\]

Thus, setting \( u = k_1, v = k_2, \delta = U_\delta \) as above, we obtain
\[
\max_{0 \leq k_1, k_2 \leq \frac{\delta}{4}} \frac{\log Z_N^\varepsilon(k_1, k_2, \delta)}{N^{2\xi - 1}} \Rightarrow \max_{u, v \in U_\delta} \left\{ \beta N^{-\frac{\delta}{4}} \left( \Sigma_{[u, N^\varepsilon]} + \Sigma_{[v, N^\varepsilon]} \right) - \hat{\beta} N^{-\frac{\delta}{4}} R_N^\varepsilon(u, v) - |\mathring{h}|(u + v + 2)\delta N^{\xi - \varepsilon - (2\xi - 1)} + N^{-(2\xi - 1)} p_R^\varepsilon(u, v) \right\}.
\]

We get as above that the lim inf as \( N \to +\infty \) of the right-hand side is \( \hat{P} \)-a.s. larger than
\[
\tilde{W}_{2}^{A,\delta} := \max_{u, v \in U_\delta} \left\{ \hat{\beta}(X_u^{(2)} + X_v^{(1)}) - \hat{\beta} \sup_{0 \leq t \leq \delta} |X_{u+t}^{(2)} - X_u^{(2)}| - \hat{\beta} \sup_{0 \leq t \leq \delta} |X_{v+t}^{(1)} - X_v^{(1)}| - I(u, v) \right\}. \tag{5.17}
\]

**Conclusion.** Summarizing, we have \( \hat{P} \)-a.s. the upper bound (5.10) and the lower bound (5.17) for \( \limsup_N N^{-(2\xi - 1)} \log Z_N^\varepsilon \) and \( \liminf_N N^{-(2\xi - 1)} \log Z_N^\varepsilon \) respectively. Notice that, since trajectories of Lévy processes are a.s. càd-làg (continuous in the case of Brownian motion), we have that
\[
\lim_{\delta \downarrow 0} \tilde{W}_{2}^{A,\delta} = \lim_{\delta \downarrow 0} \hat{W}_{2}^{A,\delta} = \sup_{u, v \in [0, 1]} \left\{ \hat{\beta}(X_u^{(2)} + X_v^{(1)}) - I(u, v) \right\},
\]
which is exactly \( W_2^A \).

**Step 2.** Next, we prove the following result.

**Lemma 5.2.** In Region \( R_2 \), there is some \( A_0 > 0 \) and some constant \( C = C_\beta \) such that, for all \( A \geq A_0 \), for all \( N \geq 1 \)
\[
\mathbb{P} \left( \frac{1}{N^{2\xi - 1}} \log Z_N^\varepsilon \left( M_N^* > AN^\xi \right) \geq -1 \right) \leq CA^{1 - 2\alpha}.
\]

Since \( \alpha > 1/2 \) in Region \( R_2 \), Lemma 5.2 implies that for any \( \varepsilon > 0 \) we can choose \( A > 0 \) such that \( \frac{1}{N^{2\xi - 1}} \log Z_N^\varepsilon (M_N^* > AN^\xi) < -1 \) with \( \mathbb{P} \)-probability larger than \( 1 - \varepsilon \). Therefore, thanks to Lemma 5.1 and because \( W_2^A \geq 0 \) (by taking \( u = 0 = v \)), one can choose \( A \) such that the second term in (5.9) is negligible compared to the first one in \( \mathbb{P} \)-probability.

**Proof of Lemma 5.2.** Let us write
\[
Z_{N}^{\omega, \varepsilon} := Z_N (M_N^* > AN^\xi) = \sum_{k=1}^{\infty} Z_N^\varepsilon (M_N^* \in (2^{k-1}AN^\xi, 2^k AN^\xi)),
\]
so that
\[
Z_{N}^{\omega, \varepsilon} \leq \sum_{k=1}^{\infty} \exp \left( \hat{\beta} N^{-\gamma} \Sigma_{2kAN^\xi}^* + |\mathring{h}| 2^{k+1} AN^{\xi - \varepsilon} \right) \mathbb{P} (M_N^* \geq 2^{k-1} AN^\xi).
\]

Using that \( \mathbb{P} (M_N^* \geq 2^{k-1} AN^\xi) \leq 2 \exp \left( -2^{2k-3} A^2 N^{2\xi - 1} \right) \), we get by subadditivity
\[
\mathbb{P} \left( Z_{N}^{\omega, \varepsilon} \geq e^{-N^{2\xi - 1}} \right) \leq \sum_{k=1}^{\infty} \mathbb{P} \left( 2 \exp \left( \hat{\beta} N^{-\gamma} \Sigma_{2kAN^\xi}^* - 2^{2k-4} A^2 N^{2\xi - 1} \right) \geq \frac{1}{2^{k+1}} e^{-N^{2\xi - 1}} \right).
\]
where we have used the fact that $2^{k+1}|h|AN^{\xi-\zeta} \leq 2^{2k-4}A^2N^{2\xi-1}$ for large enough $A$, uniformly in $k, N \geq 1$ (using that we have $\xi - \zeta < 2\xi - 1$ in Region $R_2$). Therefore, provided that $A$ is sufficiently large, recalling that $\gamma + 2\xi - 1 = \xi/\alpha$, we end up with

$$P \left( Z_N^w \geq e^{-N^{2\xi-1}} \right) \leq \sum_{k=1}^{\infty} P \left( \beta \Sigma^k_{2^kAN^\xi} \geq 2^{2k-5}A^2N^{\xi/\alpha} \right) \leq \sum_{k=1}^{\infty} c_\beta 2^{(1-2\alpha)k} A^{1-2\alpha},$$

where we have used Lemma 4.3 for the last inequality. Since $\alpha > 1/2$, this concludes the proof of Lemma 5.2.

**Remark 5.1.** If we want to upgrade our convergence in $\hat{P}$-probability to a $\hat{P}$-a.s. convergence, we would need to upgrade Lemma 5.2 to the following:

$$\lim_{A \to \infty} \hat{P} \left( \limsup_{N \to \infty} \frac{1}{N^{2\xi-1}} \log Z_N^w \left( M_N^w > AN^\xi \right) \geq -1 \right) = 0.$$

With the same proof as above, we would need to bound

$$\lim_{N_0 \to \infty} \hat{P} \left( \sup_{N \geq N_0} \left( Z_N^w e^{-N^{2\xi-1}} \right) \geq 1 \right) \leq \lim_{N_0 \to \infty} \sum_{k=1}^{\infty} \hat{P} \left( \sup_{N \geq N_0} (N^{-\xi/\alpha}\Sigma^k_{2^kAN^\xi}) \geq c_\beta 2^{2k}A^2 \right).$$

The proof would be complete if one could exchange the limit and the sum since we have

$$\limsup_{N \to \infty} N^{-\xi/\alpha}\Sigma^k_{2^kAN^\xi} \leq 3X_{2^kA}^\alpha$$

the continuous process analogous to $\Sigma^k$, see e.g. Lemma 2.23 in [24, Ch. VI]. But one needs to apply dominated convergence, that is control the tail of $\sup_{N \geq N_0} (N^{-\xi/\alpha}\Sigma^k_{2^kAN^\xi})$: for that one would need a better control of the convergence in the coupling. Indeed, a sufficient condition to obtain the $\hat{P}$-a.s. convergence is the following: there is some $\varepsilon < 2 - \frac{1}{\alpha}$ such that $\hat{P}$-a.s. there exists a constant $C(\omega)$ such that

$$\sup_{N \geq 1} N^{-\xi/\alpha}\Sigma^k_{2^kAN^\xi} \leq C(\omega)t^\varepsilon X_t^\alpha, \quad \text{uniformly in } t \geq 1.$$

The important part in this condition is that the constant $C(\omega)$ is uniform on all scales $t \in [2^kA, 2^{k+1}A]$.

For instance, this condition is verified if the coupling is exact, that is if the $\omega_i$’s are $\alpha$-stable, in which case we can set $\hat{\omega}^N_i = N^{\xi/\alpha}(X_i^{(1)}(\cdot)_{(i+1)N^\xi} - X_i^{(1)}(\cdot)_{(i+1)N^\xi-\varepsilon})$ for $i \geq 0$ (and analogously for $i < 0$).

**Step 3.** Let us note that, by monotonicity in $A$, we have that $W_2 = \lim_{A \uparrow \infty} W_2^A$ is well-defined (possibly infinite) and non-negative. We prove the following lemma:

**Lemma 5.3.** If $\alpha \in \left( \frac{1}{2}, 2 \right]$, we have that $W_2 := \sup_{u,v \geq 0} \left( \hat{\beta}(X_v^{(1)} + X_u^{(2)}) - I(u,v) \right)$ is $\hat{P}$-a.s. positive and finite.

Combined with Lemmas 5.1-5.2 this readily proves that $N^{-2(\xi-1)} \log Z_N^w$ converges almost surely to $W_2$ as $N \to \infty$.

**Proof.** To show that $W_2 > 0$ almost surely, notice that taking $u = 0$ we have

$$W_2 \geq \sup_{v \geq 0} \left\{ \hat{\beta} X_v^{(1)} - \frac{1}{2} v^2 \right\}.$$

Then, almost surely, we can find some sequence $v_n \downarrow 0$ such that $X_v^{(1)} \geq v_1^{1/\alpha}$ for all $n$ (see e.g. [11 Th. 2.1]): we get that $W_2 \geq \sup_{n \geq 0} \left( \hat{\beta} v_1^{1/\alpha} - \frac{1}{2} v_n^{1/\alpha} \right) > 0$ since $\alpha > 1/2$.

To show that $W_2 < +\infty$ a.s. notice that $I(u,v) = \frac{1}{2}(u \wedge v + u + v)^2 \geq \frac{1}{2}(u^2 + v^2)$: we therefore get that

$$W_2 \leq \sup_{u \geq 0} \left\{ \hat{\beta} X_u^{(2)} - \frac{1}{2} u^2 \right\} + \sup_{v \geq 0} \left\{ \hat{\beta} X_v^{(1)} - \frac{1}{2} v^2 \right\}.$$

Let us consider the second term and show that it is a.s. finite (the first term is identical). For any $\varepsilon > 0$, a.s. $X_v^{(1)} \leq v^{(1+\varepsilon)/\alpha}$ for $v$ large enough, see e.g. [31 Sec. 3]. Hence $\hat{\beta} X_v^{(1)} - \frac{1}{2} v^2 \leq \hat{\beta} v^{(1+\varepsilon)/\alpha} - \frac{1}{2} v^2 \leq 0$ for all $v$ large enough, provided that $(1 + \varepsilon)/\alpha < 2$, which concludes the proof.
We prove that the range size is of order $\log N$.

Let us define, for $r \in (0, 1)$
\[
U_2^{\varepsilon, \varepsilon'} = \{ (u, v) \in (\mathbb{R}_+)^2 : \sup_{(s, t) \in B_\varepsilon(u, v)} \{ \hat{\beta}(X_s^{(1)} + X_u^{(2)}) - I(s, t) \} \geq \mathcal{W}_2 - \varepsilon' \},
\]
where $B_\varepsilon(u, v)$ is the closed ball of center $(u, v)$ and of radius $\varepsilon > 0$. Let us observe that $U_2^{\varepsilon, \varepsilon'}$ is a.s. bounded: we know that a.s. the supremum outside a compact set $[-A(\omega), 0] \times [0, A(\omega)]$ is smaller than $-1 \leq \mathcal{W}_2 - \varepsilon'$, see Lemma 5.3. Moreover, by Lemma 5.3 we can choose $\varepsilon'$ such that $\mathcal{W}_2 - \varepsilon' > 0$.

We now prove that for any $\varepsilon, \varepsilon' \in (0, 1)$, $\lim_{N \to \infty} \mathbb{P}_N(\mathcal{A}_N^{\varepsilon, \varepsilon'}(M_N^\varepsilon) \in U_2^{\varepsilon, \varepsilon'}) = 1$ in $\hat{\mathbb{P}}$-probability. To simplify the notation, we denote the event $\{ \frac{1}{N^\varepsilon}(-M_N^\varepsilon, M_N^\varepsilon) \notin U_2^{\varepsilon, \varepsilon'} \}$ by $A_{N, 2}^{\varepsilon, \varepsilon'}$. We have
\[
\log \mathbb{P}_N(A_{N, 2}^{\varepsilon, \varepsilon'}) = \log Z_N(\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) - \log Z_N^{\varepsilon}. 
\]
From what we showed above, we have that $N^{-(2\varepsilon - 1)} \log Z_N^{\varepsilon}$ converges in $\hat{\mathbb{P}}$-probability to $\mathcal{W}_2$, so the proof will be complete if we show that $N^{-(2\varepsilon - 1)} \log Z_N(\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) < \mathcal{W}_2$ with $\hat{\mathbb{P}}$-probability close to 1. Thanks to Lemma 5.2 we only need to estimate $Z_N(\mathcal{A}_N^{\varepsilon, \varepsilon'}; \mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'})$. For any $\delta > 0$, we perform a similar decomposition as in [5, 10] to get
\[
Z_N^{\varepsilon, \varepsilon'}(\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) := Z_N^{\varepsilon}(M_N^\varepsilon \leq AN^\varepsilon; \mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) = \sum_{k_1 = 0}^{[A/\delta]} \sum_{k_2 = 0}^{[A/\delta]} Z_N(k_1, k_2, \delta; \mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}),
\]
where we defined $Z_N(k_1, k_2, \delta; \mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'})$ as
\[
Z_N \left( M_N^\varepsilon \in (-1 - k_1)\delta N^\varepsilon, -k_1 \delta N^\varepsilon, k_2 \delta N^\varepsilon, (k_2 + 1)\delta N^\varepsilon; \mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'} \right).
\]
By definition of $\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}$, we get that
\[
Z_N^{\varepsilon, \varepsilon'}(\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) \leq \left( \frac{A}{\delta} \right)^2 \max_{(k_1, k_2) \in U_2^{\varepsilon, \varepsilon'}} Z_N(k_1, k_2, \delta),
\]
where $U_2^{\varepsilon, \varepsilon'} := \{ (k_1, k_2) : k_1 \delta, k_2 \delta \in U_\delta, [k_1 \delta, (k_1 + 1)\delta) \times [k_2 \delta, (k_2 + 1)\delta) \notin U_2^{\varepsilon, \varepsilon'} \}$, with $U_\delta = \{ 0, \delta, 2\delta, \ldots, A \}$; by convention the maximum is 0 if $U_2^{\varepsilon, \varepsilon'}$ is empty.

Now, by the same argument as in Step 1, we have that
\[
\lim \sup_{\delta \to 0} \lim_{N \to \infty} \frac{1}{N^\varepsilon} \log Z_N^{\varepsilon, \varepsilon'}(\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) \leq \sup_{(u, v) \notin U_2^{\varepsilon, \varepsilon'}} \{ \hat{\beta}(X_u^{(1)} + X_v^{(2)}) - I(u, v) \} \leq \mathcal{W}_2 - \varepsilon',
\]
by definition of $U_2^{\varepsilon, \varepsilon'}$. This concludes the proof that $\lim_{N \to \infty} \mathbb{P}_N(\mathcal{A}_{N, 2}^{\varepsilon, \varepsilon'}) = 0$ $\hat{\mathbb{P}}$-a.s.

Let us now observe that thanks to Proposition 5.1 the maximiser $(U_2^{(2)}, V_2^{(2)})$ of $\mathcal{W}_2$ is $\hat{\mathbb{P}}$-a.s. unique: hence, $\bigcap_{\varepsilon' > 0} U_2^{\varepsilon'} \subset B_{4_\varepsilon}(U_2^{(2)}, V_2^{(2)})$. Therefore, for any $\varepsilon > 0$ there is a.s. some $\varepsilon' > 0$ such that $U_2^{\varepsilon, \varepsilon'}$ is included in $B_{8\varepsilon}(U_2^{(2)}, V_2^{(2)})$, which concludes the proof.

### 5.4 Region R3: proof of Theorem 3.4

Recall that in Region R3 we have
\[
\gamma < \zeta - \frac{a-1}{\alpha} \quad \text{and} \quad \gamma < \frac{1-\alpha}{\alpha}, \quad \text{with } \alpha \in (0, 1) \cup (1, 2].
\]
We prove that the range size is of order $N$. 

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Convergence of the rescaled log-partition function

First of all, notice that we can reduce to the case \( h_N \equiv 0 \). Indeed, we have the bounds

\[
Z^\omega_N,\beta_N,h_N=0 \times e^{-|h_N|N} \leq Z^\omega_N,\beta_N,h_N \leq Z^\omega_N,\beta_N,h_N=0 \times e^{|h_N|N}.
\]

Since \( h_N = \hat{h}N^{-\zeta} \) with \( \zeta > \gamma + \frac{\alpha - 1}{\alpha} \), we have that \( \lim_{N \to \infty} N^{-\left(\frac{\alpha}{\alpha} - \frac{1}{\alpha}\right)}|h_N|N = 0 \). In the following, we therefore focus on the convergence of \( N^{-\left(\frac{\alpha}{\alpha} - \frac{1}{\alpha}\right)} \log Z^\omega_N,\beta_N,h_N=0 \). We write for simplicity \( Z^\omega_N \) for \( Z^\omega_N,\beta_N,h_N=0 \).

For any \( \delta > 0 \), we can write

\[
Z^\omega_N = \sum_{k_1=0}^{\lfloor 1/\delta \rfloor} \sum_{k_2=0}^{\lfloor 1/\delta \rfloor} Z^\omega_N(k_1, k_2, \delta),
\]

with \( Z^\omega_N(k_1, k_2, \delta) \) as in (5.11) with \( \xi = 1 \). Since there are at most \( N \) steps for the random walk, we can have \( M_N^\delta \leq -k_1\delta N \) and \( M_N^\delta \geq k_2\delta N \) only if \( \delta(k_1 + k_2 + k_1 + k_2) \leq 1 \). Hence, writing \( u = k_1\delta, v = k_2\delta \), and \( U_\delta = \{0, \delta, 2\delta, \ldots, 1\} \), we have

\[
\max_{u \in U_\delta} \log Z^\omega_N(u, v, \delta, \delta) \leq \log Z^\omega_N \leq -2 \log \delta + \max_{u \in U_\delta} \log Z^\omega_N(u, v, \delta, \delta).
\]

Upper bound. We have

\[
\max_{u \in U_\delta} N^{\gamma - \frac{1}{\alpha}} \log Z^\omega_N(u, v, \delta) \leq \max_{u \in U_\delta} \beta \left( N^{-\frac{1}{\alpha}} \Sigma^{-[u,N]} + N^{-\frac{1}{\alpha}} \Sigma^{+[v,N]} + N^{-\frac{1}{\alpha}} R^\delta_N(u, v) \right),
\]

where \( R^\delta_N(u, v) \) is as in (5.13) with \( \xi = 1 \). As in the previous section, we get that \( \mathbb{P}\text{-a.s.} \) the lim sup of the right-hand side is bounded above by

\[
\hat{\mathcal{W}}^\delta_3 := \max_{u \in U_\delta} \left\{ \beta(X_u^{(2)} + X_v^{(1)}) + \beta \sup_{0 \leq t \leq \delta} |X_u^{(2)} - X_u^{(1)}| + \beta \sup_{0 \leq t \leq \delta} |X_v^{(1)} - X_v^{(1)}| \right\}.
\]

Lower bound. We have

\[
\max_{u \in U_\delta} N^{\gamma - \frac{1}{\alpha}} \log Z^\omega_N(u, v, \delta) \geq \max_{u \in U_\delta} \left\{ \beta \left( N^{-\frac{1}{\alpha}} \Sigma^{-[u,N]} + N^{-\frac{1}{\alpha}} \Sigma^{+[v,N]} \right) - N^{-\frac{1}{\alpha}} R^\delta_N(u, v) - N^{1+\gamma - \frac{1}{\alpha}} \log 2 \right\},
\]

where we used that any non-empty event of \( (S_n)_{0 \leq n \leq N} \) has probability at least \( 2^{-N} \). Now, since \( \gamma < \zeta + \frac{1}{\alpha} - 1 \) and \( \gamma < \frac{1}{\alpha} - 1 \), the last two terms in the maximum go to 0; we therefore get that \( \mathbb{P}\text{-a.s.} \) the lim inf of the right-hand side is bounded below by

\[
\tilde{\mathcal{W}}^\delta_3 := \max_{u \in U_\delta} \left\{ \beta(X_u^{(2)} + X_v^{(1)}) - \beta \sup_{0 \leq t \leq \delta} |X_u^{(2)} - X_u^{(1)}| - \beta \sup_{0 \leq t \leq \delta} |X_v^{(1)} - X_v^{(1)}| \right\}.
\]

Conclusion. We can conclude in the same manner as in the proof of Lemma 5.1 letting \( N \to \infty \) and then \( \delta \downarrow 0 \), we get that \( N^{\gamma - \frac{1}{\alpha}} \log Z^\omega_N \) converges \( \mathbb{P}\text{-a.s.} \) to

\[
\lim_{\delta \downarrow 0} \tilde{\mathcal{W}}^\delta_3 = \lim_{\delta \downarrow 0} \mathcal{W}^\delta_3 = \sup_{u \in U_\delta} \{ \beta(X_u^{(2)} + X_v^{(1)}) \},
\]

where the limit holds thanks to the a.s. càdlàg property of trajectories of the Lévy process (or continuity in the Brownian motion case). This is exactly the variational problem \( \mathcal{W}_3 \) defined in Theorem 3.4. Together with the (trivial) fact that \( \mathcal{W}_3 \in (0, +\infty) \) a.s., this concludes the proof of (3.4). \( \square \)
Convergence of \((M_N^*, M_N^-)\)

The proof follows the same strategy as in Region \(R_2\), so we only give a sketch. Let us define the counterpart of \(U_3^{ε,ε'}\) in Region \(R_3\) by

\[
U_3^{ε,ε'} = \{ (u,v) \in (\mathbb{R}_+)^2 : u < v + u + v \leq 1, \sup_{s,t \geq 0, (s,t) \in \mathcal{B}_8(\mathbb{R}_+,v,v)} \{ \hat{\beta}(X_s^{(1)} + X_s^{(2)}) \} \geq W_3 - ε' \}.
\]

Then we denote the event \(\{ 1/\lambda - M_N^-, M_N^+ \} \notin U_3^{ε,ε'} \}) by \(A_{N,3}^{ε,ε'}\). By the same procedure as in Region \(R_2\) (here we can use \(\hat{P}\)-a.s. convergences, since we do not have to restrict trajectories), we can first show that \(\hat{P}\)-a.s.

\[
\limsup_{N \to \infty} \frac{1}{N^{\frac{1}{\alpha} - \gamma}} \log Z_N(A_{N,3}^{ε,ε'}) < W_3 \quad \text{and so} \quad \limsup_{N \to \infty} \frac{1}{N^{\frac{1}{\alpha} - \gamma}} \log P_N(A_{N,3}^{ε,ε'}) < 0.
\]

We then deduce, as done in Region \(R_2\) that \(\hat{P}\)-a.s. \(\lim_{N \to \infty} P_N(\{ 1/\lambda - M_N^-, M_N^+ \} \notin U_3^{ε,ε'} \}) = 1.\) By uniqueness of the maximizer \((U(3), V(3))\) of \(W_3\) (Propposition 3.1), we get that if \(ε'\) is small enough, then \(U_3^{ε,ε'}\) is contained in \(B_\delta(U(3), V(3))\), which completes the proof.

5.5 Region \(R_4\): proof of Theorem 3.5

We prove that in Region \(R_4\) the range size is of order \(N^\xi\) with \(\xi = \alpha - 1/\alpha \). We take \(\hat{h} > 0\), and recall that in this region we have

\[
\left(\frac{(2α-1)\xi - (\alpha-1)}{\alpha} \right) - (\alpha-1) < \gamma < \left(\frac{(2α+1)\xi - (\alpha-1)}{3\alpha} \right),
\]

and that \(\xi - \xi = \xi/\alpha - \gamma > 2\xi - 1\). Recall also that region \(R_4\) does not exist if \(\alpha < 1\).

Convergence of the rescaled log-partition function

For any \(A > 0\), we first write

\[
Z_N^\omega = Z_N^\omega \left( M_N^* \leq AN^\xi \right) + Z_N^\omega \left( M_N^* > AN^\xi \right).
\]

(5.18)

The strategy is similar to that in Region \(R_2\) and we use analogous notation. We proceed in three steps: (1) after taking logarithm and dividing by \(N^{\xi - \xi}\), we show that the first term converges to some limit \(W_4^A\) when \(N \to \infty\); (2) we show that the second term in (5.18) is small compared to the first one; (3) we show that \(W_4^A \to W_4\) as \(A \to \infty\), with \(W_4 \in (0, +\infty)\) almost surely.

Step 1. We prove the following lemma.

Lemma 5.4. In Region \(R_4\), we have that \(\hat{P}\)-a.s., for any \(A \in \mathbb{N}\),

\[
\lim_{N \to \infty} \frac{1}{N^{\xi - \xi}} \log Z_N^\omega \left( M_N^* \leq AN^\xi \right) = W_4^A := \sup_{u,v \in [0,A]} \left\{ \hat{\beta}(X_u^{(1)} + X_u^{(2)}) - \hat{h}(u + v) \right\},
\]

with \((X_u^{(1)}, X_u^{(2)})_{u,v \geq 0}\) from Notation 2.

Proof. For fixed \(\delta > 0\), we write (recall the notation (5.11))

\[
Z_N^{\omega,\xi} := Z_N^\omega \left( M_N^* \leq AN^\xi \right) = \sum_{k_1=0}^{\lfloor A/\delta \rfloor} \sum_{k_2=0}^{\lfloor A/\delta \rfloor} Z_N^\omega(k_1, k_2, \delta).
\]

Since the number of summands above is finite, we can write

\[
\max_{0 \leq k_1, k_2 \leq \frac{A}{\delta}} \log Z_N^\omega(k_1, k_2, \delta) \leq \log Z_N^{\omega,\xi} \leq 2 \log(\frac{A}{\delta}) + \max_{0 \leq k_1, k_2 \leq \frac{A}{\delta}} \log Z_N^\omega(k_1, k_2, \delta).
\]

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Upper bound. We write \( u = k_1 \delta, v = k_2 \delta \) and set \( U_\delta = \{0, \delta, 2\delta, \ldots, A\} \). Recalling that \( \xi - \zeta = \xi / \alpha - \gamma \), we get that

\[
\max_{0 \leq k_1, k_2 \leq \frac{A}{\delta}} \frac{1}{N^{\xi - \zeta}} \log Z_N^w(k_1, k_2, \delta) \\
\leq \max_{u, v \in U_\delta} \left\{ \hat{\beta} N^{-\delta} \left( \Sigma^{- \delta}_{[u \delta N]} + \Sigma^{+ \delta}_{[v \delta N]} \right) + \hat{\beta} N^{-\delta} R_N^d(u, v) - \hat{h}(v + u) + N^{\xi - \zeta} \delta^d_N(u, v) \right\}.
\]

Then, notice that \( \lim_{N \to \infty} N^{\xi - \zeta} \delta^d_N(u, v) = 0 \), thanks to Lemmas \[A.1\] and \[A.3\] since we have \( \xi - \zeta > |2\xi - 1| \). Therefore, similarly to the previous sections, \( \hat{\mathbb{P}} \)-a.s. the lim sup of right-hand side is bounded above by

\[
\hat{W}_4^A, \delta := \max_{u, v \in U_\delta} \left\{ \hat{\beta} \left( X_u^{(2)} + X_v^{(1)} - \sup_{t \in [0, \delta]} |X_u^{(2)}| - \sup_{t \in [0, \delta]} |X_v^{(1)}| \right) - \hat{h}(v + u) \right\}.
\]

Lower bound. On the other hand, we bound \( \log Z_N^w(k_1, k_2, \delta) \) from below by

\[
\beta N \left( \Sigma^{- \delta}_{[k_1 \delta N]} + \Sigma^{+ \delta}_{[k_2 \delta N]} \right) - \beta N R_N^d(k_1 \delta, k_2 \delta) - h_N(k_2 + k_1 + 2) \delta N^\xi - p_N^\delta(k_1 \delta, k_2 \delta).
\]

Thus, setting \( u = k_1 \delta, v = k_2 \delta \) and \( U_\delta = \{0, \delta, \ldots, A\} \) as above, we obtain

\[
\max_{0 \leq k_1, k_2 \leq \frac{A}{\delta}} \frac{1}{N^{\xi - \zeta}} \log Z_N^w(k_1, k_2, \delta) \\
\geq \max_{u, v \in U_\delta} \left\{ \hat{\beta} \left( X_u^{(2)} + X_v^{(1)} - \sup_{t \in [0, \delta]} |X_u^{(2)}| - \sup_{t \in [0, \delta]} |X_v^{(1)}| \right) - \hat{h}(v + u + 2\delta) \right\}.
\]

Hence, similarly to what is done above, \( \hat{\mathbb{P}} \)-a.s. the lim inf of the right-hand side is bounded below by

\[
\hat{W}_4^A, \delta := \max_{u, v \in U_\delta} \left\{ \hat{\beta} \left( X_u^{(2)} + X_v^{(1)} - \sup_{t \in [0, \delta]} |X_u^{(2)}| - \sup_{t \in [0, \delta]} |X_v^{(1)}| \right) - \hat{h}(v + u + 2\delta) \right\}.
\]

Conclusion. The terms \( \hat{W}_4^A, \delta, \hat{W}_4^A \) are almost sure upper and lower bound for the lim sup and lim inf of \( N^{\xi - \zeta} \log Z_N^w \). By the a.s. càd-làg property of trajectories of Lévy processes (or continuity in the Brownian motion case), we have

\[
\lim_{\delta \downarrow 0} \hat{W}_4^A, \delta = \lim_{\delta \downarrow 0} \hat{W}_4^A \delta = \sup_{u, v \in [0, A]} \left\{ \hat{\beta} \left( X_u^{(2)} + X_v^{(1)} \right) - \hat{h}(v + u) \right\},
\]

which is exactly \( W_4^A \). The convergence in Lemma \[5.4\] is therefore achieved by letting \( N \to \infty \) and then \( \delta \to 0 \).

\[ \square \]

Step 2. Next, we prove the following lemma.

**Lemma 5.5.** In region \( R_4 \), there is some \( A_0 > 0 \) and some constant \( C = C_{\beta, \delta} \), such that for all \( A \geq A_0 \) and any \( N \geq 1 \)

\[
\mathbb{P} \left( \frac{1}{N^{\xi - \zeta}} \log Z_N^w(M_N^A > AN^\xi) \geq -1 \right) \leq CA^{1-\alpha}.
\]

Since \( \alpha > 1 \) in region \( R_4 \), this proves that for any \( \epsilon > 0 \) we can choose \( A \) large enough so that \( \frac{1}{N^{\xi - \zeta}} \log Z_N^w(M_N^A > AN^\xi) < -1 \) with \( \mathbb{P} \) probability larger than \( 1 - \epsilon \). Hence, thanks to Lemma \[5.4\] and the fact that \( W_4^A \geq 0 \), the second term in \[5.18\] is negligible compared to the first one in \( \hat{\mathbb{P}} \)-probability. Note that here again, we are not able to upgrade the convergence to a \( \hat{\mathbb{P}} \)-a.s. convergence, see Remark \[5.1\] which also applies here.
Proof. Let us write $Z_N^{\alpha,\gamma} := Z_N^\alpha (M^\alpha_ N > AN^\gamma)$ so

$$Z_N^{\alpha,\gamma} = \sum_{k=1}^{\infty} Z_N^\alpha (M^\alpha_ N \in (2^{k-1}AN^\xi, 2^kAN^\xi]) \leq \sum_{k=1}^{\infty} \exp \left( \beta N^{-\gamma} \Sigma^*_2 AN^\xi - \frac{2^{k-1}}{2} - \frac{A^2}{2^k} \right).$$

By subadditivity, we therefore get that

$$\mathbb{P} \left( Z_N^{\alpha,\gamma} \geq e^{-N^{\xi-\gamma}} \right) \leq \sum_{k=1}^{\infty} \mathbb{P} \left( e^{\beta N^{-\gamma} \Sigma^*_2 AN^\xi - \frac{2^{k-1}}{2} - \frac{A^2}{2^k}} \geq \frac{1}{2^{k+1}} e^{-N^{\xi-\gamma}} \right) \leq \sum_{k=1}^{\infty} \mathbb{P} \left( \beta N^{-\gamma} \Sigma^*_2 AN^\xi \geq \frac{2^{k-2}}{2^k} \right),$$

where the last inequality holds provided that $A$ has been fixed large enough (we also used that $\xi - \zeta = \frac{\xi}{\alpha} - \gamma$). Then Lemma 4.4 gives that each probability in the sum is bounded by a constant times $2^{k(1-\alpha)}A^{1-\alpha}$. Since $\alpha > 1$, summing this over $k$ gives the conclusion of the proof of Lemma 5.5.

Step 3. By monotone convergence, $W_4^\alpha$ converges a.s. to $W_4$: we only need to show that $W_4$ is positive and finite. Combining this with Lemmas 5.4 and 5.5, this completes the proof of (3.5).

Lemma 5.6. If $\alpha \in (1, 2]$, we have that $W_4 := \sup_{u,v \geq 0} \{ \beta (X_v^{(1)} + X_v^{(2)}) - \hat{h}(u + v) \}$ is $\mathbb{P}$-a.s. positive and finite.

Proof. The proof is analogous to the proof of Lemma 5.3. To show that $W_4 > 0$, we use that $W_4 \geq \sup_{v \geq 0} \{ \beta X_v^{(2)} - \hat{h}v \}$. By [10, Th. 2.1], there is a.s. a sequence $v_n \downarrow 0$, such that $X_v^{(1)} \geq v_n^{1/\alpha}$ for all $n$. Hence, for large enough $n$, $W_4 \geq \hat{h}v_n^{1/\alpha} - \frac{h}{\eta} > 0$, since $\alpha > 1$.

To show that $W_4 < \infty$, we use that $W_4 \leq \sup_{u \geq 0} \{ \beta X_v^{(2)} - \hat{h}u \} + \sup_{v \geq 0} \{ \beta X_v^{(1)} - \hat{v} \hat{h} \}$. By [31], we have that for any $\epsilon > 0$, a.s. $X_v^{(1)} \leq \frac{v}{\alpha}(1+\epsilon)/\alpha$ for $v$ large enough. Therefore, if $(1+\epsilon)/\alpha < 1$ (recall $\alpha > 1$), we get that $\beta X_v^{(1)} - \hat{v} \leq \beta v^{(1+\epsilon)/\alpha} - \frac{h}{\eta} \leq 0$ for all $v$ sufficiently large. Similarly we also have that $\beta X_v^{(2)} - \hat{h}v \leq 0$ for all $v$ large enough. This concludes the proof.

Convergence of $(M_N^\alpha, M_N^\gamma)$

As in previous sections, we define

$$U_{4,\epsilon'} = \left\{ (u,v) \in (\mathbb{R}_+)^2 : \sup_{(s,t) \in B^4(u,v)} \{ \hat{\beta}(X_t^{(1)} + X_t^{(2)}) - \hat{h}(s + t) \} \geq W_4 - \epsilon' \right\}$$

and the event $A_{N,A}^{\epsilon,\epsilon'} = \{ \frac{1}{N}\log Z_N^\alpha (A_{N,A}^{\epsilon,\epsilon'}) < W_4 \}$. Then, in an identical manner as in Regions $R_2$, we have that with $\mathbb{P}$-probability close to 1,

$$\frac{1}{N^{\xi-\gamma}} \log Z_N^\alpha (A_{N,A}^{\epsilon,\epsilon'}) < W_4 \quad \text{and so} \quad \frac{1}{N^{\xi-\gamma}} \log \mathbb{P}_N^\alpha (A_{N,A}^{\epsilon,\epsilon'}) < 0,$$

from which one deduces that

$$\lim_{N \to \infty} \mathbb{P}_N^\alpha \left( \frac{1}{N^{\xi-\gamma}} (-M_N^\alpha, M_N^\alpha) \notin U_{4,\epsilon'} \right) = 1,$$

in $\mathbb{P}$-probability. (5.19)

Moreover, if $\epsilon'$ is small enough, then $U_{4,\epsilon'}$ is contained in $B_{8\epsilon}(U^{(4)}, V^{(4)})$, which completes the proof.

5.6 Region $R_5$: proof of Theorem 3.6

In Region $R_5$, we prove that the range size is of order $N^\xi$ with $\xi = \frac{1+\xi}{3} \in (0, \frac{1}{2})$. Note that in this region we take $\hat{h} > 0$ and that we have

$$1 - 2\xi = \xi - \zeta > \frac{\xi}{\alpha} - \gamma \quad \text{and} \quad -1 < \zeta < \frac{1}{2}.$$
Convergence of the rescaled log-partition function

We fix some constant $A = A(\hat{h})$ (large) and we split the partition function as

$$Z_N^* = Z_N^* \left( M_N^* \leq AN^\xi \right) + Z_N^* \left( M_N^* > AN^\xi \right).$$

The strategy of proof is similar to that in Region $R_2$, but with only two steps: (1) we show that after taking logarithm and dividing by $N^{1-2\xi}$, the first term converges to some constant independent of $A$ (if $A$ is large enough); (2) we show that for $A$ large the second term is negligible compared to the first one.

**Step 1.** We prove the following lemma.

**Lemma 5.7.** In Region $R_3$, we have that for any $A > 0$

$$\lim_{N \to +\infty} \frac{1}{N^{1-2\xi}} \log Z_N^* \left( M_N^* \leq AN^\xi \right) = \sup_{u,v \in [0,A]} \left\{ -\hat{h}(u + v) - \hat{I}(u,v) \right\} \quad \mathbb{P}-a.s.,$$

where $\hat{I}(u,v) := \frac{\alpha (u + v)^2}{2}$ for $u, v \geq 0$. By a simple calculation, the supremum is $-\frac{3}{2}(\hat{h}\pi)^{\frac{3}{2}}$ for any $A > \pi^{\frac{3}{2}}\hat{h}^{-\frac{3}{4}}$ and it is achieved at $u + v = \pi^{\frac{3}{2}}\hat{h}^{-\frac{3}{4}}$.

**Proof.** For any fixed $A$, we have the following upper and lower bounds

$$\log \hat{Z}_N^\delta - \beta N^{-\gamma} \Sigma_{AN}^\delta \leq \log Z_N^* \left( M_N^* \leq AN^\xi \right) \leq \log \hat{Z}_N^\delta + \beta N^{-\gamma} \Sigma_{AN}^\delta,$$

where $\hat{Z}_N^\delta := \mathbb{E} \left[ \exp(-h_N|R_N|) I_{\{M_N^* \leq AN^\xi\}} \right]$.

Since in Region $R_5$ we have $1 - 2\xi > \frac{k}{\alpha} - \gamma$, we get that $N^{-(1-2\xi)} \beta N^{-\gamma} \Sigma_{AN}^\delta$ goes to 0 $\mathbb{P}$-a.s. (see e.g. Lemma 5.4). Therefore, we only need to prove that

$$\lim_{N \to +\infty} \frac{1}{N^{1-2\xi}} \log \hat{Z}_N = \sup_{u,v \in [0,A]} \left\{ -\hat{h}(u + v) - \hat{I}(u,v) \right\}$$

there is no disorder anymore). But this convergence is quite standard, since $\hat{I}(u,v)$ is the rate function for the LDP for $(N^{-\xi}M_N^*, N^{-\xi}M_N^*)$: more precisely, by Lemma A.3

$$-\hat{I}(u,v) = \lim_{N \to +\infty} \frac{1}{N^{1-2\xi}} \log \mathbb{P}(M_N^* \geq -uN^\xi; M_N^* \leq vN^\xi).$$

This is enough to conclude thanks to Varadhan’s lemma.

**Step 2.** Next, we prove the following lemma.

**Lemma 5.8.** In Region $R_3$, we have for any fixed $A$

$$\lim_{N \to +\infty} \frac{1}{N^{1-2\xi}} \log Z_N^* \left( M_N^* > AN^\xi \right) \leq -\frac{1}{2} A\hat{h} \quad \mathbb{P}-a.s.$$

Combining this result with Lemma 5.7 readily yields the convergence (5.6), provided that $A > \pi^{\frac{3}{2}}\hat{h}^{-\frac{3}{4}}$ and $\frac{1}{2} A\hat{h} > \frac{3}{2}(\hat{h}\pi)^{\frac{3}{2}}$.

**Proof.** We consider four cases, which correspond to four different conditions on $\gamma, \zeta$ (see Figures 1234):

(i) $\alpha \in (1, 2]$ and $\zeta \in (-1, 1/2)$; (ii) $\alpha \in (0, 1)$ and $\zeta \in (-1, 0)$; (iii) $\alpha \in (\frac{1}{2}, 1)$ and $\zeta \in (0, \frac{1}{2})$; (iv) $\alpha \in (0, \frac{1}{2})$ and $\zeta \in (0, \frac{1}{2})$. We deal with the first two ones at the same time and we treat the third and fourth one afterwards since the strategy of the proof is slightly different.
Cases (i)-(ii). Let us write

\[ Z_N^\alpha(M_N^* > AN^\xi) = \sum_{k=1}^{\log_2(\frac{1}{\alpha}N^{1-\xi})} Z_N^\alpha(M_N^* \in (2^{k-1}AN^\xi, 2^kAN^\xi]) \]

\[ \leq \sum_{k=1}^{\log_2(\frac{1}{\alpha}N^{1-\xi})} \exp \left( C(A, \hat{\beta}, \omega) 2^{k/\alpha} N^{\frac{\xi}{\alpha} - \gamma} (\log_2 N)^{2/\alpha} - \tilde{h}2^{k-1}AN^{\xi-\zeta} \right) \]

where we have used Lemma 4.4 to bound \( \Sigma_k^* \) uniformly for \( k \). Now, uniformly for \( k \) in the sum we have

\[ \frac{2^{k/\alpha} N^{\frac{\xi}{\alpha} - \gamma} (\log_2 N)^{2/\alpha}}{2^k N^{\xi-\zeta}} \leq (\log_2 N)^{2/\alpha} \begin{cases} N^{\frac{\xi}{\alpha} - \gamma - (\xi-\zeta)} & \text{if } \alpha \in (1, 2], \\ A^{-\frac{1}{\alpha}} N^{\frac{\xi}{\alpha} + \gamma - (\xi-\zeta)} & \text{if } \alpha \in (0, 1). \end{cases} \]

Notice that in cases (i)-(ii) the upper bound always goes to 0 as \( N \to \infty \), because \( \frac{\xi}{\alpha} - \gamma > \xi - \zeta \) in the case \( \alpha \in (1, 2] \) and \( \gamma > \zeta - \frac{\alpha-1}{\alpha} \) in the case \( \alpha \in (0, 1) \). Therefore, \( \mathbb{P} \)-a.s., for \( N \) large enough, we have

\[ Z_N^\alpha(M_N^* > AN^\xi) \leq \sum_{k=1}^{\log_2(\frac{1}{\alpha}N^{1-\xi})} \exp \left( -\tilde{h}2^{k-2}AN^{\xi-\zeta} \right) \leq C \exp \left( -\frac{1}{2} \tilde{h}AN^{\xi-\zeta} \right). \]

Since \( \xi - \zeta = 1 - 2\xi \), this concludes the proof.

Cases (iii)-(iv). In that case, we have \( \xi \in (0, 1/2) \) and \( \zeta \in (0, 1/2) \). Hence, we can write

\[ Z_N^\alpha(M_N^* > AN^\xi) = Z_N^\alpha(M_N^* \in (AN^\xi, N^{1-\xi}]) + Z_N^\alpha(M_N^* \in (N^{1-\xi}, N]). \]

(5.22)

For the first term, we have similarly as above that \( Z_N^\alpha(M_N^* \in (AN^\xi, N^{1-\xi}]) \) is bounded by

\[ \sum_{k=1}^{\log_2(\frac{1}{\alpha}N^{1-\xi})} \exp \left( C(A, \hat{\beta}, \omega) 2^{k/\alpha} N^{\frac{\xi}{\alpha} - \gamma} (\log_2 N)^{2/\alpha} - \tilde{h}2^{k-1}AN^{\xi-\zeta} \right). \]

Now, uniformly for \( k \) in the sum we have

\[ \frac{2^{k/\alpha} N^{\frac{\xi}{\alpha} - \gamma} (\log_2 N)^{2/\alpha}}{2^k N^{\xi-\zeta}} \leq (\log_2 N)^{2/\alpha} A^{-\frac{1}{\alpha}} N^{-\gamma} \frac{2\alpha-1}{\alpha} \xi + \frac{\alpha-1}{\alpha} \]

with the upper bound vanishing, because in cases (iii)-(iv) we have \( \gamma > \frac{2\alpha-1}{\alpha} \xi + \frac{\alpha-1}{\alpha} \) for \( \xi \in (0, 1/2) \) (note that if \( \alpha < \frac{3}{2}, \frac{2\alpha-1}{\alpha} < 0 \)). Therefore, \( \mathbb{P} \)-a.s., for \( N \) large enough, we get

\[ Z_N^\alpha(M_N^* \in (AN^\xi, N^{1-\xi}]) \leq \sum_{k=1}^{\log_2(\frac{1}{\alpha}N^{1-\xi})} \exp \left( -\tilde{h}2^{k-2}AN^{\xi-\zeta} \right) \leq C \exp \left( -\frac{1}{2} \tilde{h}AN^{\xi-\zeta} \right). \]

(5.23)

For the second term on the right-hand side of (5.22), we have

\[ Z_N^\alpha(M_N^* \in (N^{1-\xi}, N]) = \sum_{k=1}^{\log_2(N^\xi)} Z_N^\alpha(M_N^* \in (2^{-k}N, 2^{-k+1}N]) \]

\[ \leq \sum_{k=1}^{\log_2(N^\xi)} 2 \exp \left( C(\hat{\beta}, \omega) 2^{-k/\alpha} N^{\frac{1}{\alpha} - \gamma} (\log_2 N)^{2/\alpha} - 2^{-2k-1}N \right), \]

where we have used Lemma 4.4 to bound \( \Sigma_k^* \) by \( C(\omega)2^{-k/\alpha} N^{\frac{1}{\alpha} - \gamma} (\log_2 N)^{2/\alpha}, \mathbb{P} \)-a.s. and also the fact that \( \mathbb{P}(M_N^* > x) \leq 2 \exp(-\frac{x^2}{2N}). \) Now, uniformly for \( k \) in the sum, we have

\[ \frac{2^{-k/\alpha} N^{\frac{1}{\alpha} - \gamma - 1} (\log_2 N)^{2/\alpha}}{2^{-2k}N} \leq (\log_2 N)^{2/\alpha} \begin{cases} N^{-\gamma} & \text{if } \alpha \in (\frac{1}{2}, 1), \\ N^{\frac{\alpha-1}{\alpha}} - \gamma & \text{if } \alpha \in (0, \frac{1}{2}), \end{cases} \]

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which goes to 0 in cases (iii)-(iv), because \( \gamma > \frac{2\alpha - 1}{\alpha} \zeta + \frac{1 - \alpha}{\alpha} \) if \( \alpha \in \left( \frac{1}{2}, 1 \right) \) and \( \gamma > \frac{1 - \alpha}{\alpha} \) if \( \alpha \in (0, \frac{1}{2}) \). Therefore, \( \mathbb{P} \)-a.s., for \( N \) large enough, we have

\[
Z_N^\omega (M_N^+ \in (N^{1-\zeta}, N)) \leq C \log_2 N \exp \left( -\frac{1}{4} N^{1-2\zeta} \right) \leq \exp \left( -\tilde{h} AN^{\zeta-\zeta} \right), \tag{5.24}
\]

where in the last inequality we have used the fact that \( 1 - 2\zeta - (\xi - \zeta) = 1 - \xi - \zeta > 0 \), since \( \xi < \frac{1}{2} \) and \( \zeta < \frac{1}{2} \).

Combining (5.23)-(5.24) with (5.22) and since \( \xi - \zeta = 1 - 2\xi \), this concludes the proof. \( \square \)

**Convergence of** \( M_N^+ - M_N^- \)

Let us define \( c_h := \pi \frac{2}{\hat{h}} \). Let \( \varepsilon > 0 \) and define the event

\[
\mathcal{A}_{N,5}^\varepsilon = \left\{ \left| \frac{1}{N^\zeta} (M_N^+ - M_N^-) - c_h \right| > \varepsilon \right\}.
\]

As in the previous sections, since \( \log \mathbb{P}_N^{\omega} (\mathcal{A}_{N,5}^\varepsilon) = \log Z_N^{\omega} (\mathcal{A}_{N,5}^\varepsilon) - \log Z_N^{\omega} \), using the convergence (5.6) we simply need to show that there is some \( \delta_\varepsilon > 0 \) such that

\[
\lim_{N \to \infty} \mathbb{P} \left( \frac{1}{N^{1-2\zeta}} \log Z_N^{\omega} (\mathcal{A}_{N,5}^\varepsilon) < -\frac{3}{2} (\hat{h})^{2/3} - \delta_\varepsilon \right) = 1
\]

But this is simply due to the fact that analogously to Lemma 5.7, we have \( \mathbb{P} \)-a.s.

\[
\lim_{N \to +\infty} \frac{1}{N^{1-2\zeta}} \log Z_N^{\omega} (\mathcal{A}_{N,5}^\varepsilon) = \sup_{u,v \geq 0, |u+v-c_h| > \varepsilon} \left\{ -\hat{h}(u+v) - I(u,v) \right\} < -\frac{3}{2} (\hat{h})^{2/3},
\]

where the inequality is strict since the supremum in Lemma 5.4 is attained for \( u + v = c_h \). \( \square \)

**5.7 Region R6: proof of Theorem 3.7**

Recall that in Region \( R_6 \), we have

\[
\zeta < (-1) \land \gamma \quad \text{if} \quad \alpha \in (1, 2) \qquad \text{and} \qquad \zeta < (-1) \land \left( \gamma + \frac{\alpha - 1}{\alpha} \right) \quad \text{if} \quad \alpha \in (0, 1).
\]

Let us note that in all cases we have \( \gamma > \zeta \). We split \( Z_N^{\omega} \) in two parts

\[
Z_N^{\omega} = Z_N^{\omega} (|\mathcal{R}_N| = 2) + Z_N^{\omega} (|\mathcal{R}_N| \geq 3). \tag{5.25}
\]

It is clear that

\[
Z_N^{\omega} (|\mathcal{R}_N| = 2) = e^{-2\hat{h} N^{-\zeta}} \left( e^{\beta N^{-\gamma}(\omega_0+\omega_1)2^{-N}} + e^{\beta N^{-\gamma}(\omega_0+\omega_1)2^{-N}} \right),
\]

so that \( N^{\zeta} \log Z_N^{\omega} (|\mathcal{R}_N| = 2) \) converges \( \mathbb{P} \)-a.s. to \( -2\hat{h} \), since \( \zeta < \gamma \) and \( \zeta < -1 \).

We now prove that \( \lim \sup_{N \to \infty} N^{\zeta} \log Z_N^{\omega} (|\mathcal{R}_N| \geq 3) \) is strictly smaller than \( -2\hat{h} \) a.s.: this will imply that the second term in (5.23) is negligible compared to the first one and as a consequence prove that \( \mathbb{P}_N^{\omega} (|\mathcal{R}_N| = 2) \) converges to 1 \( \mathbb{P} \)-a.s.

We write

\[
Z_N^{\omega} (|\mathcal{R}_N| \geq 3) = \sum_{k=3}^{N} Z_k^{\omega} (|\mathcal{R}_N| = k) \leq \sum_{k=3}^{N} \exp \left( -k\hat{h} N^{-\zeta} + C(\omega) \beta N^{-\gamma} k^{1/\alpha}(\log_2 N)^{2/\alpha} \right),
\]

where we have used Lemma 4.4 to bound \( \Sigma^2 \) for the last inequality. Now, uniformly for \( k \) in the sum we have

\[
\frac{k^{1/\alpha} N^{-\gamma}(\log_2 N)^{2/\alpha}}{k N^{-\zeta}} \leq \left( \log_2 N \right)^{2/\alpha} \begin{cases} N^{\zeta-\gamma} & \text{if} \ \alpha \in (1, 2], \\ \frac{1}{\alpha} N^{\frac{1}{\alpha}+\zeta-\gamma} & \text{if} \ \alpha \in (0, 1). \end{cases}
\]
In all this section, we consider only the case \( \hat{h} \). In this region, we prove that the range size is of order \( N^{\xi} \). Since in Region 1, we have

\[
Z_N^\xi(M_N^* > AN^{\xi}) \leq \sum_{k=3}^{N} \exp \left( -\frac{5}{6} k \hat{h} N^{-\xi} \right) \leq C \exp \left( -\frac{5}{2} \hat{h} N^{-\xi} \right).
\]

We therefore get \( \limsup_{N \to \infty} N^{\xi} \log Z_N^\xi(|\mathcal{R}_N| \geq 3) \leq -\frac{5}{2} \hat{h} < -2 \hat{h} \) a.s., which concludes the proof. \( \square \)

## 6 Proof of the remaining results: the case \( \hat{h} < 0 \)

In all this section, we consider only the case \( \hat{h} < 0 \).

### 6.1 Region \( \hat{R}_4 \): proof of Theorem 4.1

In this region, we prove that the range size is of order \( N^{\xi} \) with \( \xi = 1 - \zeta / \alpha \in (\frac{1}{2}, 1) \). Recall that in Region \( \hat{R}_4 \), we have

\[
2\xi - 1 = \xi - \zeta > \frac{\xi}{\alpha} - \gamma \quad \text{and} \quad 0 < \zeta < \frac{1}{2}.
\]

The proofs are almost identical to what is done in regions \( R_5-R_6 \), so we give much less detail. We fix some constant \( A = A_\hat{h} := 32|\hat{h}| \) and we split the partition function as

\[
Z_N^\xi = Z_N^\xi(M_N^* \leq AN^{\xi}) + Z_N^\xi(M_N^* > AN^{\xi}).
\]

**Step 1.** We have the following lemma, analogous to Lemma 5.7.

**Lemma 6.1.** In Region \( \hat{R}_4 \), for any \( A > 0 \) we have the following convergence

\[
\lim_{N \to +\infty} \frac{1}{N^{2\xi-1}} \log Z_N^\xi(M_N^* \leq AN^{\xi}) = \sup_{u,v \in [0,A]} \left\{ |\hat{h}|(u + v) - I(u,v) \right\} \quad \text{\( \mathbb{P}\)-a.s.}
\]

By a simple calculation, the supremum is \( \frac{1}{2} \hat{h}^2 \) for any \( A \geq |\hat{h}| \) and it is attained at \( (u,v) = (0,|\hat{h}|) \) or \((u,v) = (|\hat{h}|,0)\).

**Proof.** Since in Region \( \hat{R}_4 \) we have \( 2\xi - 1 > \frac{\xi}{\alpha} - \gamma \), for any fixed \( A \) we get that \( N^{-(2\xi-1)} \times \beta N^{-\gamma} \Sigma_{AN^{\xi}}^{*} \) almost surely goes to 0. Therefore we only need to prove that

\[
\lim_{N \to +\infty} \frac{1}{N^{2\xi-1}} \log Z_N^0(M_N^* \leq AN^{\xi}) = \sup_{u,v \in [0,A]} \left\{ |\hat{h}|(u + v) - I(u,v) \right\},
\]

where \( Z_N^0 \) denotes the partition function with \( \omega \equiv 0 \) (or equivalently \( \beta_N \equiv 0 \)).

But (6.1) follows from Varadhan’s lemma, since \( I(u,v) \) is the rate function for the LDP for \( (N^{-\xi}M_N^*, N^{-\xi}M_N^+) \), by Lemma A.1. \( \square \)

**Step 2** To conclude the proof of the convergence (4.1), it remains to show the following.

**Lemma 6.2.** In Region \( \hat{R}_4 \), we have for any \( A \geq 32|\hat{h}| \)

\[
\limsup_{N \to +\infty} \frac{1}{N^{2\xi-1}} \log Z_N^\xi(M_N^* > AN^{\xi}) \leq 0 \quad \text{\( \mathbb{P}\)-a.s.}
\]

Together with Lemma 6.1, this readily yields the convergence (4.1).
Proof. We write

\[
Z_N(M_N^* > AN^\xi) = \sum_{k=1}^{\log_2(\frac{1}{N^{1-\xi}})} Z_N(M_N^* \in (2^{k-1}AN^\xi, 2^kAN^\xi)) \leq \sum_{k=1}^{\log_2(\frac{1}{N^{1-\xi}})} 2 \exp \left( C(A, \hat{\beta}, \omega) 2^{k/\alpha} N^{\xi-\gamma} (\log_2 N)^{2/\alpha} + 2^{k+1} |\hat{h}|AN^{\xi-\zeta} - 2^{2k-3} A^2 N^{2\xi-1} \right),
\]

where we have used Lemma 4.4 to get an almost sure bound on \( \Sigma^*_k \), and also the fact that \( \mathbb{P}(M_N^* > x) \leq 2 \exp(-\frac{x^2}{2N}) \). Now, as in the proof of Lemma 5.8, the “disorder” term is seen to be negligible compared to the “range” term (uniformly for \( k \) in the sum): we get that \( \mathbb{P}\text{-a.s., for } N \text{ large enough,} \)

\[
Z_N^\circ(M_N^* > AN^\xi) \leq \sum_{k=1}^{\log_2(\frac{1}{N^{1-\xi}})} 2 \exp \left( 2^{k+2} |\hat{h}|AN^{\xi-\zeta} - 2^{2k-3} A^2 N^{2\xi-1} \right) \leq 2 \log_2(\frac{1}{P}N^{1-\xi}),
\]

where for the last inequality we have used that \( \xi - \zeta = 2\xi - 1 \) and that \( A \geq 2^{5}|\hat{h}| \). This concludes the proof. \( \square \)

Convergence of trajectories. First of all, let us go one step further in the proof of Lemma 6.1. Indeed, in (6.1) the supremum in the variational problem is attained at \( (u, v) = (0, |\hat{h}|) \) or \( (u, v) = (|\hat{h}|, 0) \), so we can deduce that the main contribution to \( Z_N^0 \) (hence to \( Z_N \) in view of the proof of Lemma 6.1) comes from trajectories with \( N^{-\xi}(M_N, M_N^*) \) either close to \( (0, |\hat{h}|) \) or to \( (-|\hat{h}|, 0) \). One can actually show that the main contribution comes from trajectories moving at roughly constant speed to these endpoints (similarly to Proposition 3.8): using (A.4), one easily gets that analogously to (6.1), for any \( \epsilon > 0 \),

\[
\limsup_{N \to \infty} \frac{1}{N^{2\xi-1}} \log Z_N^0 \left( M_N^* \leq AN^\xi, (B_N^{+,\epsilon} \cup B_N^{-,\epsilon})^c \right) < \sup_{u,v \in [0,A]} \left\{ |\hat{h}|(u+v) - I(u,v) \right\}, \quad (6.2)
\]

where we recall the definition of the events \( B_N^{+,\epsilon} \) (recall \( \hat{h} < 0 \)):

\[
B_N^{+,\epsilon} := \left\{ \sup_{t \in [0,1]} |N^{-\xi}S_{[tN]} + \hat{h} t| \leq \epsilon \right\}, \quad B_N^{-,\epsilon} := \left\{ \sup_{t \in [0,1]} |N^{-\xi}S_{[tN]} - \hat{h} t| \leq \epsilon \right\}.
\]

All together, in view of the fact that \( N^{-2(2\xi-1)} \hat{\beta} N^{-\gamma} \Sigma^*_N \) goes to 0 a.s., we get that

\[
\lim_{N \to \infty} \frac{1}{N^{2\xi-1}} \log Z_N^\circ \left( (B_N^{+,\epsilon} \cup B_N^{-,\epsilon})^c \right) < \lim_{N \to \infty} \frac{1}{N^{2\xi-1}} \log Z_N^0,
\]

from which one deduces that

\[
\lim_{N \to \infty} \mathbb{P}_N^\circ(B_N^{+,\epsilon} \cup B_N^{-,\epsilon}) = 1 \quad \mathbb{P}\text{-a.s.}
\]

Given \( \hat{h} \), the events \( B_N^{+,\epsilon} \) are disjoint for \( \epsilon \) small enough: this implies in particular that

\[
\lim_{N \to \infty} \mathbb{P}_N(B_N^{+,\epsilon}) + \mathbb{P}_N(B_N^{-,\epsilon}) = \lim_{N \to \infty} \frac{Z_N^\circ(B_N^{+,\epsilon}) + Z_N^\circ(B_N^{-,\epsilon})}{Z_N^\circ} = 1. \quad (6.3)
\]

Now, denoting again \( Z_N^\circ \) the partition function with \( \omega = 0 \) (or equivalently \( \beta_N = 0 \)) and \( \mathbb{P}_N^0 \) the corresponding measure, we have

\[
e^{\beta_N \Sigma^*(\hat{h}|-\epsilon))N^{\xi-\beta_N R_N^0(0,|\hat{h}|-\epsilon)}} \mathbb{P}_N^0(B_N^{+,\epsilon}) \leq \frac{Z_N^\circ(B_N^{+,\epsilon})}{Z_N^0} \leq e^{\beta_N \Sigma^*(\hat{h}|-\epsilon))N^{\xi}} \mathbb{P}_N^0(B_N^{+,\epsilon}) + \beta_N R_N^2(0,|\hat{h}|-\epsilon) \mathbb{P}_N^0(B_N^{-,\epsilon}),
\]

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where \( R_N^\varepsilon(u,v) \) is defined in \((5.13)\). A similar inequality holds with \( B_N^\varepsilon \) in place of \( B_N^\varepsilon \), simply by replacing \( \Sigma_\varepsilon^{+}((|h| - \varepsilon)N^\varepsilon) \) by \( \Sigma_\varepsilon^{-}((|h| - \varepsilon)N^\varepsilon) \) and \( R_N^{2\varepsilon}(0,|h| - \varepsilon) \) by \( R_N^{2\varepsilon}(|h| - \varepsilon, 0) \).

Therefore, we have

\[
\frac{Z_N^\varepsilon(B_N^{+\varepsilon})}{Z_N^\varepsilon(B_N^{+\varepsilon}) + Z_N^\varepsilon(B_N^{-\varepsilon})} \leq \frac{e^{\beta_N \Sigma_\varepsilon^+(|h| - \varepsilon)N} + e^{\beta_N \Sigma_\varepsilon^-(|h| - \varepsilon)N} + e^{\beta_N R_N^{2\varepsilon}(0,|h| - \varepsilon) + \beta_N R_N^{2\varepsilon}(|h| - \varepsilon, 0)}}{e^{\beta_N \Sigma_\varepsilon^+(|h| - \varepsilon)N} + e^{\beta_N \Sigma_\varepsilon^-(|h| - \varepsilon)N} + e^{\beta_N R_N^{2\varepsilon}(0,|h| - \varepsilon) + \beta_N R_N^{2\varepsilon}(|h| - \varepsilon, 0)}} , \tag{6.4}
\]

and similarly

\[
\frac{Z_N^\varepsilon(B_N^{+\varepsilon})}{Z_N^\varepsilon(B_N^{+\varepsilon}) + Z_N^\varepsilon(B_N^{-\varepsilon})} \geq \frac{e^{\beta_N \Sigma_\varepsilon^+(|h| - \varepsilon)N} + e^{\beta_N \Sigma_\varepsilon^-(|h| - \varepsilon)N} + e^{\beta_N R_N^{2\varepsilon}(0,|h| - \varepsilon) + \beta_N R_N^{2\varepsilon}(|h| - \varepsilon, 0)}}{e^{\beta_N \Sigma_\varepsilon^+(|h| - \varepsilon)N} + e^{\beta_N \Sigma_\varepsilon^-(|h| - \varepsilon)N} + e^{\beta_N R_N^{2\varepsilon}(0,|h| - \varepsilon) + \beta_N R_N^{2\varepsilon}(|h| - \varepsilon, 0)}} . \tag{6.5}
\]

Let us make a few observations. First of all, notice also that by symmetry we get that when \( |\hat{h}| > 0 \)

\[
\lim_{N \to \infty} P_N^\varepsilon(B_N^{+\varepsilon}) = \frac{1}{2}. \tag{6.6}
\]

Recall that \( \lim_{N \to +\infty} N^{-\xi/(\xi + \gamma)} \leq N^{-\xi/(\xi + \gamma)} \leq \lim_{N \to +\infty} N^{-\xi/(\xi + \gamma)} \leq X^{1\varepsilon}_{|h|} - \varepsilon \) and \( X^{1\varepsilon}_{|h|} - \varepsilon \) for fixed \( \hat{h} \), \( \mathbb{P} \)-a.s. the two processes \( X^{(1)}_{|h|} \) and \( X^{(2)}_{|h|} \) are both continuous at \( t = |\hat{h}| \), so \( \lim_{\varepsilon \downarrow 0} X^{(1)}_{|h|} = X^{(2)}_{|h|} \). Note also that, for \( \hat{\mathbb{P}} \)-almost every realization of \( \omega \), for any \( \delta \) one can choose \( \varepsilon > 0 \) small enough so that \( N^{-\xi/(\xi + \gamma)} R_N^{2\varepsilon}(0,|\hat{h}| - \varepsilon) \leq \delta \) for all \( N \) large enough, and similarly for \( N^{-\xi/(\xi + \gamma)} R_N^{2\varepsilon}(|\hat{h}| - \varepsilon, 0) \).

Let us now consider three cases.

(i) If \( \gamma < \xi/\alpha \). On the event \( X^{(1)}_{|h|} < X^{(2)}_{|h|} \), one can choose \( \varepsilon > 0 \) small enough so that

\[
\lim_{N \to +\infty} N^{-\xi/(\xi + \gamma)} \left( \Sigma_{\varepsilon^+}((|h| - \varepsilon)N^\varepsilon) + R_N^{2\varepsilon}(0,|h| - \varepsilon) \right) > 0 .
\]

Since \( \beta_N = \gamma N^{-\gamma} N^{-\frac{\xi}{\xi + \gamma}} \) with \( N^{-\gamma} \to +\infty \), from \((6.4)\) we deduce that a.e. on the event \( X^{(1)}_{|h|} < X^{(2)}_{|h|} \), for \( \varepsilon > 0 \) small enough

\[
\lim sup_{N \to +\infty} P_N^\varepsilon(B_N^{+\varepsilon}) = \lim sup_{N \to +\infty} \frac{Z_N^\varepsilon(B_N^{+\varepsilon})}{Z_N^\varepsilon(B_N^{+\varepsilon}) + Z_N^\varepsilon(B_N^{-\varepsilon})} = 0,
\]

recalling also \((6.6)\). By an identical reasoning, we get that a.e. on the event \( X^{(1)}_{|h|} > X^{(2)}_{|h|} \), for \( \varepsilon > 0 \) small enough \( \lim sup_{N \to +\infty} P_N^\varepsilon(B_N^{-\varepsilon}) = 0 \). Hence, because the event \( X^{(1)}_{|h|} = X^{(2)}_{|h|} \) has probability 0 and recalling \((6.3)\), we can conclude that

\[
\lim_{\varepsilon \downarrow 0} \lim_{N \to +\infty} P_N^\varepsilon(B_N^{+\varepsilon}) = \mathbb{1}_{\{X^{(1)}_{|h|} > X^{(2)}_{|h|}\}} \quad \hat{\mathbb{P}}\text{-a.s.}
\]

where the limit in \( N \) is well-defined provided that \( \varepsilon \) is small enough. A similar statement holds for \( P_N^\varepsilon(B_N^{-\varepsilon}) \), exchanging the role of \( X^{(1)} \) and \( X^{(2)} \).
(ii) If $\gamma = \xi/\alpha$, then similarly as above, from (6.4) and (6.5) we deduce that for any $\delta > 0$, $\hat{P}$-a.s. we can choose $\varepsilon > 0$ small enough so that

$$
\limsup_{N \to +\infty} \mathbb{P}_N^\varepsilon(B_N^{+,\varepsilon}) = \limsup_{N \to +\infty} \frac{Z_N^\varepsilon(B_N^{+,\varepsilon})}{Z_N^\varepsilon(B_N^{+,\varepsilon}) + Z_N^\varepsilon(B_N^{-,\varepsilon})} \leq \frac{e^{\beta X(1)_{-\varepsilon} + \delta}}{e^{\beta X(1)_{-\varepsilon} - \delta}} + e^{\beta X(2)_{-\varepsilon} + \delta},
$$

$$
\liminf_{N \to +\infty} \mathbb{P}_N^\varepsilon(B_N^{+,\varepsilon}) = \liminf_{N \to +\infty} \frac{Z_N^\varepsilon(B_N^{+,\varepsilon})}{Z_N^\varepsilon(B_N^{+,\varepsilon}) + Z_N^\varepsilon(B_N^{-,\varepsilon})} \geq \frac{e^{\beta X(1)_{-\varepsilon} - \delta}}{e^{\beta X(1)_{-\varepsilon} + \delta}} + e^{\beta X(2)_{-\varepsilon} - \delta},
$$

recalling again (6.3) and (6.4). Taking $\delta$ arbitrarily small, we get that

$$
\limsup_{\varepsilon \to 0} \lim\sup_{N \to +\infty} \mathbb{P}_N^\varepsilon(B_N^{+,\varepsilon}) = \lim\limsup_{\varepsilon \to 0} \mathbb{P}_N^\varepsilon(B_N^{+,\varepsilon}) = \frac{e^{\beta X(1)_{-\varepsilon}}}{e^{\beta X(1)_{-\varepsilon}} + e^{\beta X(2)_{-\varepsilon}}} \quad \hat{P}\text{-a.s.}
$$

The statement is analogous for $\mathbb{P}_N^\varepsilon(B_N^{-,\varepsilon})$, exchanging the role of $X(1)$ and $X(2)$.

(iii) If $\gamma > \xi/\alpha$, since $\beta_N = \beta_N^{N^{-\gamma} \gamma N^{-\frac{\xi}{\alpha}}}$ and $N^{-\gamma} \gamma \to 0$, we get from (6.4) and (6.5) that for any $\varepsilon > 0$

$$
\lim_{N \to +\infty} \mathbb{P}_N^\varepsilon(B_N^{+,\varepsilon}) = \lim_{N \to +\infty} \frac{Z_N^\varepsilon(B_N^{+,\varepsilon})}{Z_N^\varepsilon(B_N^{+,\varepsilon}) + Z_N^\varepsilon(B_N^{-,\varepsilon})} = \frac{1}{2} \quad \hat{P}\text{-a.s.},
$$

recalling again (6.3). We also get $\lim_{N \to +\infty} \mathbb{P}_N^\varepsilon(B_N^{-,\varepsilon}) = \frac{1}{2}$, $\hat{P}$-a.s.

This concludes the proof of (4.2). \qed

6.2 Boundary region $\tilde{R}_4$—$\tilde{R}_5$: proof of Theorem 4.2

The proof is similar to that for region $\tilde{R}_4$: one only needs the analogous to Lemma 6.1. The rate function $I(u, v)$ for $(N^{-\xi}M_N^{-}, N^{-\xi}M_N^{+})$ is replaced by the rate function $\kappa(u \wedge v + u + v)$ for $(N^{-1}M_N^{-}, N^{-1}M_N^{+})$, see Lemma 4.2. We end up with:

$$
\lim_{N \to +\infty} \frac{1}{N} \log Z_N^\varepsilon = \sup_{u, v \in [0, 1]} \{ |\hat{h}|(u + v) - \kappa(u \wedge v + u + v) \} \quad \text{P-a.s.}
$$

Then, using that $\kappa(t) = \frac{1}{2}(1 + t) \log(1 + t) + \frac{1}{2}(1 - t) \log(1 - t)$ if $0 \leq t \leq 1$ and $\kappa(t) = +\infty$ if $t > 1$, a straightforward calculation finds that the supremum is attained at $(u, v) = (0, \tanh(|\hat{h}|))$ or $(u, v) = (\tanh(|\hat{h}|), 0)$ and equals $\log(\sinh(|\hat{h}|))$.

Then, by following the same ideas as above, one can show that the events

$$
B_N^{\varepsilon} := \left\{ \sup_{t \in [0, 1]} |N^{-1}S_{[0, 1]} - \tanh(|\hat{h}|)t| \leq \varepsilon \right\}, \quad B_N^{-\varepsilon} := \left\{ \sup_{t \in [0, 1]} |N^{-1}S_{[0, 1]} + \tanh(|\hat{h}|)t| \leq \varepsilon \right\},
$$

verify

$$
\lim_{N \to +\infty} \mathbb{P}_N^\varepsilon(B_N^{+,\varepsilon} \cup B_N^{-,\varepsilon}) = 1 \quad \text{P-a.s.}
$$

From this, one can proceed as above (see in particular (6.3) and (6.5)) to get (4.4). Details are left to the reader.

6.3 Region $\tilde{R}_5$: proof of Theorem 4.3

First of all, notice that

$$
Z_N^\varepsilon \geq Z_N^\varepsilon(\{|S_N| = N\}) \geq e^{-\beta N^{-\gamma} \Sigma_N^{\varepsilon} - hN^{1-\zeta} - N \log 2} \text{ and } Z_N^\varepsilon \leq e^{-hN^{1-\zeta} + \beta N^{-\gamma} \Sigma_N^{\varepsilon}}.
$$

Hence, the P-a.s. convergence $\lim_{N \to +\infty} N\zeta^{-1} \log Z_N^\varepsilon = |\hat{h}|$ is immediate, since $\zeta < 0$ and $\frac{1}{\alpha} - \gamma < 1 - \zeta$. 

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The rest of the proof of Theorem 4.3 is similar to what is done in Sections 6.1 and 6.2 above. In particular, for any \(\varepsilon > 0\) one has that

\[
\lim_{N \to \infty} \sup_{N^{-\varepsilon}} \log Z_N^\varepsilon(|S_N| \leq (1 - \varepsilon)N) \leq -(1 - \frac{1}{2} \varepsilon)\hat{h} \quad \mathbb{P}\text{-a.s.,}
\]

so that \(\lim_{N \to \infty} \mathbb{P}_N^\varepsilon(|S_N| \geq (1 - \varepsilon)N) = 1 \mathbb{P}\text{-a.s.}\). Then, by following the same ideas as above, one can show that the events

\[
B_N^{+\varepsilon} := \left\{ \sup_{t \in [0,1]} |N^{-1}S_{[t]} - t| \leq \varepsilon \right\}, \quad B_N^{-\varepsilon} := \left\{ \sup_{t \in [0,1]} |N^{-1}S_{[t]} + t| \leq \varepsilon \right\},
\]

verify

\[
\lim_{N \to \infty} \mathbb{P}_N^\varepsilon(B_N^{+\varepsilon} \cup B_N^{-\varepsilon}) = 1 \quad \mathbb{P}\text{-a.s.}
\]

From this, one can proceed as above to get \((4.3)\). Details are left to the reader.

**Improvement in the case** \(\alpha \in (0,1)\) or \(\alpha \in (1,2)\) and \(\gamma > \zeta\). We now prove \((4.3)\). As mentioned above, we have \(\lim_{N \to \infty} \mathbb{P}_N^\varepsilon(|S_N| \geq (1 - \varepsilon)N) = 1\) almost surely. Now, we can split the event \(|S_N| > (1 - \varepsilon)N\) according to whether \(M_N^+ > \frac{1}{2} N\) or \(M_N^- < -\frac{1}{2} N\). Hence, we only have to prove that \(\mathbb{P}_N^\varepsilon(\frac{1}{2} N < M_N^- \leq N - 1)\) goes to 0, and similarly for \(M_N^+\).

To this end, we show the following: \(\mathbb{P}\)-a.s., for \(N\) large enough we have

\[
\frac{Z_N^\varepsilon(\frac{1}{2} N < M_N^+ \leq N - 1)}{Z_N^\varepsilon(|S_N| = N)} \leq C \exp \left( \frac{\varepsilon}{2} \hat{h} N^{-\zeta} \right). \tag{6.8}
\]

This will conclude the proof of \((4.3)\) since \(\hat{h} N^{-\zeta} \to -\infty\) (recall \(\zeta < 0\)).

We have \(Z_N^\varepsilon(S_N = N) = 2^{-N}e^{\beta N \Sigma_{2\varepsilon}} N^N\). Hence, using that in the case \(M_N^+ = N - k\) then \(M_N^- \geq -\frac{1}{2} k\) so \(|R_N| \leq N - \frac{1}{2} k\), we get that for \(1 \leq k < \frac{1}{2} N\), after simplifications of the numerator and denominator,

\[
\frac{Z_N^\varepsilon(M_N^+ = N - k)}{Z_N^\varepsilon(S_N = N)} \leq \exp \left( \beta_N \sum_{i=N-k+1}^N \omega_i + \beta_N \Sigma_{2\varepsilon}^+ + \frac{1}{2} \hat{h} N k \right) 2^N \mathbb{P}(M_N^+ = N - k).
\]

Denoting \(\Sigma_{2\varepsilon}^+ := \Sigma_{2\varepsilon}^+ + \sum_{j=1}^N \omega_i\), we then get that for \(\ell \in \{1, \ldots, \log_2 N - 1\}\)

\[
\frac{Z_N^\varepsilon(M_N^+ \in [N - 2^{\ell}, N - 2^{\ell-1}])}{Z_N^\varepsilon(S_N = N)} \leq \exp \left( \beta N^{-\zeta} \Sigma_{2\varepsilon}^+ + \hat{h} N^{-\zeta} 2^{\ell-2} \right) 2^{N+1} \mathbb{P}(S_N \geq N - 2^{\ell}),
\]

where we also used that \(\mathbb{P}(M_N^+ \geq N - 2^{\ell}) = 2\mathbb{P}(S_N \geq N - 2^{\ell}) - 1\) by the reflection principle. Now, since \(N - S_N\) has a Binom\((N, \frac{1}{2})\) distribution, we have

\[
\mathbb{P}(S_N \geq N - 2^{\ell}) = \sum_{i=0}^{2^\ell} 2^{-N} \binom{N}{i} \leq 2^{-N} \frac{N}{2^\ell} \binom{N}{2^\ell},
\]

for \(\ell\) such that \(2^\ell \leq \frac{1}{2} N\). Note that \(2^\ell \binom{N}{2^\ell} \leq N 2^{\ell}\), which is smaller than \(\exp(2^{\ell-3}|\hat{h}| N^{-\zeta})\) for \(N\) large enough (uniformly for the range of \(\ell\) considered). We therefore end up with

\[
\frac{Z_N^\varepsilon(M_N^+ \in [N - 2^{\ell}, N - 2^{\ell-1}])}{Z_N^\varepsilon(S_N = N)} \leq \exp \left( C(\hat{\beta}, \omega)N^{-\gamma}(\log_2 N)^{2/\alpha} 2^{\ell/\alpha} + \hat{h} N^{-\zeta} 2^{\ell-3} \right),
\]

where we have also bounded \(\Sigma_{2\varepsilon}^+\) by a constant \(c = c(\omega)\times \ell^{2/\alpha} 2^{\ell/\alpha}\), analogously to Lemma 4.4. Now, uniformly for \(\ell \in \{1, \ldots, \log_2 N - 1\}\), we have

\[
\frac{N^{-\gamma}(\log_2 N)^{2/\alpha} 2^{\ell/\alpha}}{N^{-\zeta} 2^\ell} \leq (\log_2 N)^{2/\alpha} \left\{ \begin{array}{ll} N^{\zeta-\gamma} & \text{if } \alpha \in (1,2], \\ N^{\frac{1-\alpha}{2} + \gamma} & \text{if } \alpha \in (0,1]. \end{array} \right.
\]
This upper bound goes to 0 as $N \to \infty$ since $\gamma > \zeta$ if $\alpha \in (1, 2]$ and $\gamma > \zeta + \frac{1-\alpha}{\alpha}$ if $\alpha \in (0, 1)$. Hence, $\mathbb{P}$-a.s., for $N$ large enough we have

\[
\frac{Z_N^\infty \left( \max N < M_N^+ \leq N-1 \right)}{Z_N^\infty (S_N = N)} = \log_2 N \sum_{\ell = 1}^{N-1} \frac{Z_N^\infty (M_N^+ \in [N - 2^\ell, N - 2^{\ell-1}])}{Z_N^\infty (S_N = N)} \leq \sum_{\ell = 1}^{\log_2 N-1} \exp \left( \hat{h} N^{-\zeta} 2^{\ell-4} \right) \leq C \exp \left( \frac{1}{8} \hat{h} N^{-\zeta} \right),
\]

which gives (6.8).

For the proof of the last statement (i.e. the analogous of (4.5)), notice that

\[
Z_N^\infty (S_N = N) = 2^{-N} e^{\beta N \Sigma^+ - h_N^{\infty}}, \quad Z_N^\infty (S_N = -N) = 2^{-N} e^{\beta N \Sigma^- - h_N^{\infty}},
\]

so that

\[
\frac{Z_N^\infty (S_N = N)}{Z_N^\infty (S_N = N) + Z_N^\infty (S_N = -N)} = \frac{e^{\beta N \Sigma^+}}{e^{\beta N \Sigma^-} + e^{\beta N \Sigma^+}}. \tag{6.9}
\]

Then, we proceed as in the previous sections to get (4.5) with $\{S_N = N\}$ in place of $B_N^+$. Details are left to the reader.

A. Technical estimates

A.1. Estimates on deviation probabilities

We present here some large deviation estimates for the simple random walk that are needed throughout the paper. Recall $M_N^- := \min_{0 \leq n \leq N} S_n$ and $M_N^+ := \max_{0 \leq n \leq N} S_n$.

Stretched

Our first lemma deals with the super-diffusive case: we estimate the probability that $M_N^+ \geq vN^\xi$ and $M_N^- \leq -uN^\xi$ when $\xi \in \left( \frac{1}{2}, 1 \right)$, for $u, v \geq 0$. The one-sided large deviation results are classical, using e.g. explicit calculations for the simple random walk (see [20, Ch. III.7]): we get that if $\xi \in (\frac{1}{2}, 1)$

\[
\lim_{N \to \infty} -\frac{1}{N^{2\xi-1}} \log \mathbb{P}(M_N^+ \geq v N^\xi) = \lim_{N \to \infty} -\frac{1}{N^{2\xi-1}} \log \mathbb{P}(S_N \geq v N^\xi) = \frac{1}{2} v^2.
\]

The case where both the minimum and maximum are required to have large deviations is an easy extension of the reflection principle that

\[
\mathbb{P}(S_N \geq 2a + b) \vee \mathbb{P}(S_N \geq a + 2b) \leq \mathbb{P}(M_N^+ \leq -a; M_N^- \geq b) \leq \mathbb{P}(S_N \geq 2a + b) + \mathbb{P}(S_N \geq a + 2b),
\]

so that $\log \mathbb{P}(M_N^- \leq -uN^\xi; M_N^+ \geq vN^\xi) \sim \log \mathbb{P}(S_N \geq (u \wedge v + u + v)N^\xi)$ as $N \to +\infty$.

Lemma A.1. If $\xi \in \left( \frac{1}{2}, 1 \right)$ then for any $u, v \geq 0$ we have that

\[
\lim_{N \to \infty} -\frac{1}{N^{2\xi-1}} \log \mathbb{P}\left( M_N^- \leq -uN^\xi; M_N^+ \geq vN^\xi \right) = I(u, v) := \frac{1}{2} (u \wedge v + u + v)^2. \tag{A.1}
\]

As an easy consequence of this lemma, we get that for any $\delta > 0$, for any $u, v \geq 0$,

\[
\lim_{N \to \infty} -\frac{1}{N^{2\xi-1}} \log \mathbb{P}\left( M_N^- \in \left( -u + (u, \delta), -u \right] N^\xi; M_N^+ \in \left[ v, v + \delta \right] N^\xi \right) = I(u, v). \tag{A.2}
\]

Using again the reflection principle, it is also not difficult to show that a local version of (A.2) holds: omitting the integer parts for simplicity, we have

\[
\limsup_{N \to \infty} -\frac{1}{N^{2\xi-1}} \log \sup_{x \in [u, u + \delta]} \sup_{y \in [v, v + \delta]} \mathbb{P}(M_N^- = -x N^\xi; M_N^+ = y N^\xi) = I(u, v). \tag{A.3}
\]
We now state a result that shows that the large deviation is essentially realized by the event that the random walk moves ballistically to one end (whichever is the closest) and then ballistically to the other one. For \( u, v \geq 0 \) with \( u \neq v \), recall the definition (A.3) of the function \( b_{u,v} \) that goes with constant speed from 0 to the closest point between \(-u\) and \(v\) and then to the other one. Recall also the notation (3.9):

\[
\mathcal{B}_N^{\varepsilon}(u,v) := \left\{ \sup_{t \in [0,1]} \left| \frac{1}{N^{\varepsilon}} S_t[N] - b_{u,v}(t) \right| \leq \varepsilon \right\}.
\]

We then have the following result, which is a direct consequence of Lemma A.1: let \( u, v \geq 0 \) with \( u \neq v \), then for any \( \varepsilon > 0 \),

\[
\liminf_{N \to \infty} -\frac{1}{N^{2\varepsilon-1}} \log P \left( M_N^- \leq -uN^{\varepsilon}; M_N^+ \geq vN^{\varepsilon}; \mathcal{B}_N^{\varepsilon}(u,v) \right) \geq I(u,v) + c_\varepsilon(u,v), \quad (A.4)
\]

for some constant \( c_\varepsilon(u,v) > 0 \).

As a consequence, for \( \delta > 0 \) small enough such that \((u-\delta)^+ > 0 \) or \((v-\delta)^+ > 0 \) (where \( x^+ := \max\{x,0\} \) and \([u-\delta, u+\delta] \cap [v-\delta, v+\delta] = \emptyset \), we also have

\[
\liminf_{N \to \infty} -\frac{1}{N^{2\varepsilon-1}} \log \sup_{x \in [(u-\delta)^+, u+\delta]} \log P \left( M_N^- = -xN^{\varepsilon}; M_N^+ = yN^{\varepsilon}; \mathcal{B}_N^{\varepsilon}(u,v) \right) \geq I((u-\delta)^+, (v-\delta)^+) + c_\varepsilon(u-\delta, v-\delta).
\]

Together with (A.3), we end up with

\[
\liminf_{N \to \infty} -\frac{1}{N^{2\varepsilon-1}} \log \sup_{x \in [(u-\delta)^+, u+\delta]} \log P \left( M_N^- = -xN^{\varepsilon}; M_N^+ = yN^{\varepsilon}; \mathcal{B}_N^{\varepsilon}(u,v) \right) \geq I((u-\delta)^+, (v-\delta)^+) + c_\varepsilon(u-\delta, v-\delta) - I(u+\delta, v+\delta) =: c_{\varepsilon,\delta}(u,v) \quad (A.5)
\]

where \( c_{\varepsilon,\delta}(u,v) > 0 \), provided that \( \delta \) is small enough (how small depends on \( \varepsilon, u, v \)).

Let us also state the large deviation result in the case \( \xi = 1 \). As above, it derives from the fact that \( \log P(M_N^- \leq -uN; M_N^+ \geq vN) \sim \log P(S_N \geq (u+v+u+v)N) \) as \( N \to \infty \).

**Lemma A.2.** For any \( u, v \geq 0 \), we have that

\[
\lim_{N \to \infty} -\frac{1}{N} \log P \left( M_N^- \leq -uN; M_N^+ \geq vN \right) = \kappa(u + v + u + v),
\]

where \( \kappa : \mathbb{R}_+ \to \mathbb{R}_+ \) is the LDP rate function for the simple random walk, that is \( \kappa(t) := \frac{1}{2} (1 + t) \log (1 + t) + \frac{1}{2} (1 - t) \log (1 - t) \) if \( 0 < t \leq 1 \) and \( \kappa(t) = +\infty \) if \( t > 1 \).

Note that analogues of the ballisticity statements (A.4) and (A.5) hold in the case \( \xi = 1 \).

**Folding**

Our second lemma deals with the sub-diffusive case: we estimate the probability that \( M_N^+ \leq vN^{\xi} \) and \( M_N^- \geq -uN^{\xi} \) when \( \xi \in (0, \frac{1}{2}) \), for \( u, v \geq 0 \). The result follows from classical random walk calculations, leading to explicit expressions of ruin probabilities (see Eq. (5.8) in [20, Ch. XIV]); one may refer to [13, Lem. 2.1] and its proof for the following statement.

**Lemma A.3.** If \( \xi \in (0, \frac{1}{2}) \), then for any \( u, v \geq 0 \) we have that

\[
\lim_{N \to \infty} -\frac{1}{N^{1-2\xi}} \log P \left( M_N^- \geq -uN^{\xi}; M_N^+ \leq vN^{\xi} \right) = \frac{\pi^2}{2(u+v)^2}. \quad (A.6)
\]

As an easy consequence of this lemma, we get that for any \( \delta > 0 \) and any \( u, v \geq 0 \),

\[
\lim_{N \to \infty} -\frac{1}{N^{1-2\xi}} \log P \left( M_N^- \in [-u,-u+\delta]N^{\xi}; M_N^+ \in (v-\delta,v]N^{\xi} \right) = \frac{\pi^2}{2(u+v)^2}. \quad (A.7)
\]
A.2 Proof of Lemma 4.4

We start with the first part of the statement. First of all, notice that the bound is trivial if $\ell T^{-\alpha} > 1$: we therefore assume that $\ell T^{-\alpha} \leq 1$. Using Etemadi’s inequality (see [10] Thm. 2.2.5)) we get that

$$P(\Sigma^*_\ell > T) \leq 3 \max_{k \in \{1, \ldots, \ell\}} P(|\Sigma_k^*| > \frac{1}{6} T) + 3 \max_{k \in \{1, \ldots, \ell\}} P(|\Sigma_k^*| > \frac{1}{6} T).$$

We only bound $P(|\Sigma_k^*| > \frac{1}{6} T)$, since the same bound will hold for $P(|\Sigma_k^*| > \frac{1}{6} T)$. The case $\alpha = 2$ is a consequence of Kolmogorov’s maximal inequality and the case $\alpha \in (0, 2)$ ($\alpha \neq 1$) follows from the so-called big-jump (or one-jump) behavior. Let us give an easy proof: define $\tilde{\omega}_x := \omega_x 1_{\{\omega_x < T\}}$, so that

$$P(|\Sigma_k^*| > \frac{1}{6} T) \leq P(\exists 0 \leq x \leq k, |\omega_x| > T) + P\left(\sum_{x=0}^{k} \tilde{\omega}_x > \frac{1}{6} T\right) \leq (k + 1)P(|\omega_0| > T) + \frac{36}{T^2} (k + 1)E[(\tilde{\omega}_0)^2] + k(k + 1)E[|\tilde{\omega}_0|^2],$$

where we used a union bound for the first term and Markov’s inequality (applied to $(\sum_{x=0}^{k} \tilde{\omega}_x)^2$) for the second. Now, the first term is clearly bounded by a constant times $k T^{-\alpha}$ thanks to Assumption 1. For the second term, we use again Assumption 1 to get that if $\alpha \in (0, 1) \cup (1, 2)$, $E[(\tilde{\omega}_0)^2] \leq c T^{2-\alpha}$ and $E[|\tilde{\omega}_0|^2] \leq c T^{1-\alpha}$ (when $\alpha \in (1, 2)$ we use this for last inequality that $E[|\tilde{\omega}_0|] = 0$). Therefore, we end up with the bound

$$P(|\Sigma_K^*| > \frac{1}{6} T) \leq cT^{-\alpha} + cT^{2-2\alpha} \leq 2cT^{-\alpha},$$

where we have used that $\ell T^{-\alpha} \leq 1$ for the last inequality.

For the second part of the statement, notice that $P(\Sigma_k^* > k^{2/\alpha} 2^{k/\alpha}) \leq c k^{-2}$; hence, by Borel–Cantelli, $\mathbb{P}$-a.s. there is a constant $C' = C'(\omega)$ such that $\Sigma_k^* \leq C' k^{2/\alpha} 2^{k/\alpha}$ for all $k \geq 0$. Since $\Sigma_k^*$ is monotone in $\ell$, we get that $\mathbb{P}$-a.s. there is a constant $C = C(\omega)$ such that $\Sigma_k^* \leq C(\log_2 \ell)^{2/\alpha} \ell^{1/\alpha}$ for all $\ell \geq 1$.\hfill $\square$

A.3 Uniqueness of the maximizer: proof of Proposition 3.1

We start with the following lemma, at the core of the proof.

Lemma A.4. Let $(X^{(1)}_u)_{u \geq 0}$ and $(X^{(2)}_u)_{u \geq 0}$ be two independent $\alpha$-stable Lévy processes and $f : (\mathbb{R}_+) \to \mathbb{R}$ be any function. Denote $\mathcal{I}_{a,b} := [c, d] \times [a, b]$. Then for any two disjoint rectangles $\mathcal{I}_{a,b}^d$ and $\mathcal{I}_{a',b'}^{d'}$, we have that

$$P\left(\sup_{(u,v) \in \mathcal{I}_{a,b}^d} \{X^{(1)}_u + X^{(2)}_v + f(u,v)\} = \sup_{(u',v') \in \mathcal{I}_{a',b'}^{d'}} \{X^{(1)}_u + X^{(2)}_v + f(u',v')\}\right) = 0.$$

Proof. Let us assume that $a' > b$ (other cases are treated similarly). Then, the difference of the supremums can be written as $Z = Z' - (X^{(1)}_{a'} - X^{(1)}_b)$, with

$$Z = \sup_{(u,v) \in \mathcal{I}_{a,b}^d} \{X^{(1)}_u - X^{(1)}_b + X^{(2)}_v + f(u,v)\}, \quad Z' = \sup_{(u',v') \in \mathcal{I}_{a',b'}^{d'}} \{X^{(1)}_u - X^{(1)}_a + X^{(2)}_v + f(u',v')\}.$$

Note that $Z, Z'$ are independent of $X^{(1)}_{a'} - X^{(1)}_b$: we therefore get that

$$P(X^{(1)}_{a'} - X^{(1)}_b = Z' - Z) = 0,$$

since $X^{(1)}_{a'} - X^{(1)}_b$ has no atom (it is $\alpha$-stable).\hfill $\square$
Proof of Proposition A.7. By Lemma A.4 and subadditivity, we have
\[
P\left(\exists a, b, c, d, a', b', c', d' \in \mathbb{Q}, \text{ s.t. } T_{a,b}^{c,d} \cap T_{a',b'}^{c',d'} = \emptyset, \sup_{(u,v) \in T_{a,b}^{c,d}} Y_{u,v} = \sup_{(u,v') \in T_{a',b'}^{c',d'}} Y_{u',v'} \right) = 0.
\]
Since \(P\)-a.s. \(\sup_{u,v \geq 0} Y_{u,v} > 0\) and \(Y_{u,v} \to -\infty\) as \(u \to +\infty\) or \(v \to +\infty\), then for \(P\)-a.s. realization \((X_{t}^{(1)})_{v \geq 0}, (X_{t}^{(2)})_{u \geq 0}\), there exists some rational constant \(A = A(\omega)\), such that the supremum is achieved on the rectangle \([0, A]^2\). Then a sequential application of dichotomy yields the uniqueness of the maximizer.

Next, we show that
\[
P\left(\exists v \geq 0, \text{ such that } \arg\max_{u,v \geq 0} Y_{u,v} = \{(v, v)\}\right) = 0 \tag{A.8}
\]
Note that the above probability is bounded above by
\[
P\left(\sup_{u,v \geq 0} Y_{u,v} = \sup_{v \geq 0} Y_{v,v}\right) \leq \sup_{0 \leq u \leq v} Y_{u,v} + \sup_{0 \leq u \leq v} Y_{v,u}. \tag{A.9}
\]
It suffices to show that both probabilities on the right-hand side of (A.9) are 0. We only deal with the first term, the second one is identical.

For any \(n, k \geq 0\), let \(B_{n,k} = [\frac{k}{n}, \frac{k+1}{n}] \times [\frac{k}{n}, \frac{k+1}{n}]\). Note that \(\bigcup_{k=1}^{\infty} B_{n,k}\) covers the line \(u = v\) and that, as \(n \to +\infty\), \(\bigcup_{k=1}^{\infty} B_{n,k} \downarrow \{u = v\}\) and \(\{v \leq u\} \setminus \bigcup_{k=1}^{\infty} B_{n,k} \uparrow \{v < u\}\). Hence, by Lemma A.4 and the monotone convergence theorem, we have that
\[
P\left(\sup_{0 \leq v < u} Y_{u,v} = \sup_{0 \leq v} Y_{v,v}\right) = 0.
\]
Furthermore, \(P\)-a.s., for any \(v \geq 0\), we can take \(u_n \downarrow v\), such that \(Y_{u_n,v} \to Y_{v,v}\) by the right continuity of Lévy processes. Hence,
\[
P\left(\sup_{0 \leq u \leq v} Y_{u,v} = \sup_{0 \leq u} Y_{u,v}\right) = 1.
\]
This shows that the upper bound in (A.9) is equal to 0 and concludes the proof of (A.8). \(\square\)

### A.4 Estimates on càd-làg paths at points of continuity

**Lemma A.5.** Let \((\alpha_N(t))_{N \geq 1}\) be a sequence of càd-làg paths on \([0, \infty)\) that converges to a càd-làg path \(\alpha(t)\) for the Skorokhod distance \(d_0\) (cf. [24]). Suppose that \(\alpha\) is continuous at \(u\). Then for any \(\varepsilon, \delta > 0\), there exists \(N_0 = N_0(u, \varepsilon, \delta) > 0\) such that for all \(N \geq N_0\),
\[
|\alpha_N(u) - \alpha(u)| < \varepsilon, \tag{A.10}
\]
\[
\sup_{v \in [u, u+\delta]} |\alpha_N(v) - \alpha_N(u)| < \varepsilon + \sup_{v \in [u, u+\delta+\varepsilon]} |\alpha(v) - \alpha(u)|. \tag{A.11}
\]

**Proof.** We start by proving (A.10). Fix \(\varepsilon\) and let \(\eta = \eta(\varepsilon) > 0\) be some number to be chosen below. Since \(\lim_{N \to \infty} d_0(\alpha_N, \alpha) = 0\), for the above \(\eta > 0\), there exists a sequence of non-decreasing bijections \((\lambda_N(t))_{N \geq 1} : [0, T] \to [0, T]\) with \(T > u\) arbitrary (but fixed) and a large enough integer \(N_0\), such that for all \(N \geq N_0\),
\[
\sup_{t \in [0, T]} |\lambda_N(t) - t| < \eta \quad \text{and} \quad \sup_{t \in [0, T]} |\alpha_N(\lambda_N(t)) - \alpha(t)| < \eta. \tag{A.12}
\]
We have that
\[
|\alpha_N(u) - \alpha(u)| \leq |\alpha_N(u) - \alpha(\lambda_N(u))| + |\alpha(\lambda_N(u)) - \alpha(u)|.
\]
By (A.12) we have \(|\lambda_N(u) - u| < \eta\): if we had fixed \(\eta\) small enough, the second term above is smaller than \(\varepsilon/2\) by continuity while the first term is smaller than \(\varepsilon/2\) by (A.12). Therefore (A.10) is proved.
We now prove (A.11). Using (A.10) (with \( \varepsilon/3 \) instead of \( \varepsilon \)) together with the triangular inequality, we get that 
\[
|\alpha_N(v) - \alpha_N(u)| \leq |\alpha_N(v) - \alpha(u)| + \varepsilon/3 \tag{for \( N \) large enough, so we only need to estimate 
|\alpha_N(v) - \alpha(u)|.
\]
For any fixed \( \delta > 0 \) such that \( \delta + u \leq T \), using the sequence \( \lambda_N \) defined above, we have that
\[
\sup_{u \leq v \leq u + \delta} |\alpha_N(v) - \alpha(u)| = \sup_{\lambda_N^{-1}(u) \leq v \leq \lambda_N^{-1}(u + \delta)} |\alpha_N(\lambda_N(v')) - \alpha(u)|
\leq \sup_{\lambda_N^{-1}(u) \leq v' \leq \lambda_N^{-1}(u + \delta)} |\alpha_N(v') - \alpha(\lambda_N(v'))| + \sup_{\lambda_N^{-1}(u) \leq v' \leq \lambda_N^{-1}(u + \delta)} |\alpha(v') - \alpha(u)|.
\]
The first term above is smaller than \( \varepsilon/3 \) by (A.12). For the second term, always by (A.12), we have that
\[
|\lambda_N^{-1}(u) - u| < \eta, \quad |\lambda_N^{-1}(u + \delta) - u| < \eta + \delta
\]
and hence we need to bound
\[
\sup_{u - \eta \leq v \leq u + \eta + \delta} |\alpha(v) - \alpha(u)| \leq \sup_{u - \eta \leq v \leq u} |\alpha(v) - \alpha(u)| + \sup_{u \leq v \leq u + \delta + \eta} |\alpha(v) - \alpha(u)|.
\]
By continuity, the first term above can be made arbitrarily small by choosing \( \eta \) small, so this proves (A.11). \( \square \)

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