# Coalescence and total-variation distance of semi-infinite inverse-gamma polymers 

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#### Abstract

We show that two semi-infinite positive temperature polymers coalesce on the scale predicted by KPZ (Kardar-Parisi-Zhang) universality. The two polymer paths have the same asymptotic direction and evolve in the same environment, independently until coalescence. If they start at distance $k$ apart, their coalescence occurs on the scale $k^{3 / 2}$. It follows that the total variation distance of two semi-infinite polymer measures decays on this same scale. Our results are upper and lower bounds on probabilities and expectations that match, up to constant factors and occasional logarithmic corrections. Our proofs are done in the context of the solvable inverse-gamma polymer model, but without appeal to integrable probability. With minor modifications, our proofs give also bounds on transversal fluctuations of the polymer path. As the free energy of a directed polymer is a discretization of a stochastically forced viscous Hamilton-Jacobi equation, our results suggest that the hyperbolicity phenomenon of such equations obeys the KPZ exponent.


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## Contents

1. INTRODUCTION ..... 2
2. MAIN RESULTS ..... 4
2.1. Directed polymer model ..... 4
2.2. Inverse-gamma polymer ..... 5
2.3. Coalescence bounds ..... 6
2.4. Coupling and total variation distance ..... 8
2.5. Transversal fluctuations ..... 9
3. STATIONARY INVERSE-GAMMA POLYMER ..... 9
4. EXIT TIME ESTIMATES. ..... 12
4.1. Proof of Theorem 4.1 ..... 13
4.2. Proof of Theorem 4.2 ..... 28
4.3. Coupled polymer measures. ..... 38
5. COALESCENCE OF SEMI-INFINITE POLYMERS ..... 39
5.1. Busemann functions and semi-infinite polymers ..... 39
5.2. Coupling the forward and backward semi-infinite polymers ..... 40
5.3. Proofs of Theorems 2.1, 2.3, and 2.7, and Corollary 2.5 ..... 42
6. TOTAL VARIATION DISTANCE BOUNDS ..... 44
7. TRANSVERSAL FLUCTUATION LOWER BOUND ..... 45
APPENDIX A: TECHNICAL COMPLEMENTS ..... 48
A.1. Moderate deviation of the bulk free energy ..... 48
A.2. Proof of Propositions 4.7 and 4.8 ..... 48
A.3. Nonrandom properties. ..... 49
A.4. Radon-Nikodym derivative calculation ..... 54
A.5. Sub-exponential random variables ..... 55
A.6. Random walk estimates ..... 56
ACKNOWLEDGMENTS ..... 56
REFERENCES. ..... 57

## 1 | INTRODUCTION

This paper focuses on a probability model for nearest-neighbor up-right random walk paths on the two-dimensional square lattice. The lattice vertices are assigned independent and identically distributed random variables called weights, and the energy of a path is defined as the sum of the weights along the path. The point-to-point quenched polymer measures are probability measures on admissible paths connecting pairs of sites. The probability of a path is proportional to the exponential of its energy.

This model is known as the two-dimensional directed lattice polymer with bulk disorder and was introduced in the statistical physics literature by Huse and Henley [22] in 1985 to represent the domain wall in the ferromagnetic Ising model with random impurities. This model is expected to be a member of the Kardar-Parisi-Zhang (KPZ) universality class and has been extensively studied over the past three decades, becoming a paradigmatic model in the field of nonequilibrium statistical mechanics. See the surveys [11-14, 20, 21, 30, 31, 37].

The directed last-passage percolation (LPP) model on the square lattice is a zero-temperature version of the random polymer model. In LPP, we consider the ground states, which are admissible
paths that maximize the energy, and are referred to as geodesics. This particular LPP model with up-right nearest-neighbor lattice paths is also called the corner growth model.

In LPP, a path that starts from a given lattice vertex and only moves up or right is called a semi-infinite geodesic if each finite piece of the path is a geodesic between its endpoints. The existence, directedness, and uniqueness or nonuniqueness of semi-infinite geodesics have been well-studied and understood (see [16, 17, 38] for details). Notably, it has been demonstrated in [16] that these semi-infinite geodesics can be obtained as limits of finite geodesics, as the endpoint moves off toward infinity in a particular direction. Furthermore, it has been shown in the same paper that semi-infinite geodesics starting at different vertices but having the same asymptotic direction eventually coalesce, that is, they intersect and then move together.

The study of semi-infinite polymer measures in the case of random directed lattice polymers was carried out in [18, 25]. Similar to LPP, [25] established that semi-infinite polymer measures that start from different vertices and share the same asymptotic velocity can be coupled in such a way that their paths coalesce with probability one. As a consequence, the marginals of any two semi-infinite polymer measures that correspond to the same asymptotic velocity are asymptotic to each other. This phenomenon, known as hyperbolicity, has been found to be linked to various phenomena such as stochastic synchronization and the oneforce-one solution principle (see, e.g., [1, 26]). In this work, our focus is on providing precise quantitative bounds on the convergence rates, showcasing how this hyperbolicity obeys the KPZ exponents. Currently, such sharp estimates are only available in the so-called solvable cases, where the weight distribution is chosen in a specific way, allowing for explicit analytic computations.

With nearest-neighbor up-right paths and independent and identically distributed vertex weights, the only known solvable LPP models are the ones with either exponential or geometric weight distribution. In the only known solvable directed polymer model, the weights have a negative log-gamma distribution. This solvable directed polymer model was first introduced by the second author in [32] and has since been referred to as the inverse-gamma or log-gamma polymer.

Our main contributions in this paper are sharp quantitative bounds on the rates of coalescence of the coupled paths and convergence of the marginals in the inverse-gamma polymer model. The corresponding estimates for LPP with exponential weights were obtained in [5] using integrable probability methods, and in [34] using coupling with stationary versions of the model, which relies less on the solvability of the model. In this paper, we adopt the latter approach and further develop it to handle the additional layer of randomness that arises in the case of semi-infinite polymer measures, where the random environment only determines the path measures. Along the way, we provide various new estimates on the exit point of stationary polymers and we improve one existing estimate, namely the last inequality in (4.1).

## Organization of the paper

In Section 2, we present the setting and our main results concerning the coalescence point, total variation distance, and transversal fluctuations. The connection to hyperbolicity in stochastic Hamilton-Jacobi equations is addressed briefly in Remark 2.10. Exit time estimates in the stationary inverse-gamma polymer are a crucial tool in our proofs. We introduce the stationary polymer in Section 3 and provide the exit time estimates in Section 4. The proofs of the coalescence results are presented in Section 5, while the proofs of the total variation distance estimates can be found in Section 6. The proofs of the transversal fluctuations results are provided in Section 7. Various auxiliary results are gathered in the Appendix.

## Notation and conventions

Subscripts indicate restricted subsets of the reals and integers: for example, $\mathbb{Z}_{>0}=\{1,2,3, \ldots\}$ and $\mathbb{Z}_{>0}^{2}=\left(\mathbb{Z}_{>0}\right)^{2}$ is the strictly positive first quadrant of the planar integer lattice.

On $\mathbb{R}^{2}$ we have the following conventions for points $x=\left(x_{1}, x_{2}\right)$ and $y=\left(y_{1}, y_{2}\right)$. Coordinatewise order: $x \leqslant y$ if and only if $x_{1} \leqslant y_{1}$ and $x_{2} \leqslant y_{2}$. The $\ell^{1}$ norm is $|x|_{1}=\left|x_{1}\right|+\left|x_{2}\right|$. The origin of $\mathbb{R}^{2}$ is denoted by both 0 and $(0,0)$. The two standard basis vectors are $e_{1}=(1,0)$ and $e_{2}=(0,1)$.

For integers $m \leqslant n$, the integer interval is denoted by $\llbracket m, n \rrbracket=\{m, m+1, \ldots, n\}$. For planar points $a \leqslant b$ in $\mathbb{Z}^{2}, \llbracket a, b \rrbracket=\left\{x \in \mathbb{Z}^{2}: a \leqslant x \leqslant b\right\}$ is the rectangle in $\mathbb{Z}^{2}$ with corners $a$ and $b$. The northeast boundary of a rectangle $\llbracket a, b \rrbracket$, denoted by $\partial^{\mathrm{NE}} \llbracket a, b \rrbracket$, is the set of vertices $v \in \llbracket a, b \rrbracket$ such that $v \cdot e_{1}=b \cdot e_{1}$ or $v \cdot e_{2}=b \cdot e_{2}$. The notation $\llbracket a, b \rrbracket$ is an integer line segment in $\mathbb{Z}^{2}$ if $a$ and $b$ are on the same horizontal or vertical line. In particular, $\llbracket a-e_{1}, a \rrbracket$ and $\llbracket a-e_{2}, a \rrbracket$ denote unit edges.

The total variation distance between two probability measures $\mu$ and $\nu$ on a measurable space $(\Omega, \mathcal{F})$ is $d_{\mathrm{TV}}(\mu, \nu)=\sup _{A \in \mathcal{F}}|\mu(A)-\nu(A)|$. For a probability measure $\mu, X \sim \mu$ means the random variable $X$ has distribution $\mu$.

## 2 | MAIN RESULTS

## 2.1 | Directed polymer model

Let $\left\{Y_{z}\right\}_{z \in \mathbb{Z}^{2}}$ be a collection of positive weights on the sites of the planar integer square lattice. For vertices $u \leqslant v$ in $\mathbb{Z}^{2}, \mathbb{X}_{u, v}$ denotes the collection of up-right paths $x .=\left\{x_{i}\right\}_{0 \leqslant i \leqslant n}$ where $n=$ $|u-v|_{1}, x_{0}=u, x_{n}=v$ and $x_{i+1}-x_{i} \in\left\{e_{1}, e_{2}\right\}$ for all $i \in \llbracket 0, n-1 \rrbracket$. Define the point-to-point polymer partition function between the two vertices $u \leqslant v$ by

$$
Z_{u, v}=\sum_{x_{.} \in \mathbb{X}_{u, v}} \prod_{i=0}^{|u-v|_{1}} Y_{x_{i}} .
$$

We use the convention $Z_{u, v}=0$ if $u \leqslant v$ fails. The quenched polymer measure is a probability measure on the set $\mathbb{X}_{u, v}$ and is defined by

$$
Q_{u, v}\left\{x_{.}\right\}=\frac{1}{Z_{u, v}} \prod_{i=0}^{|u-v|_{1}} Y_{x_{i}} .
$$

In general, the positive weights $\left\{Y_{z}\right\}_{z \in \mathbb{Z}^{2}}$ can be seen as a random environment if they are chosen as independent and identically distributed positive random variables defined on some probability space $(\Omega, \mathcal{F}, \mathbb{P})$. Under the moment assumption

$$
\mathbb{E}\left[\left|\log Y_{x}\right|^{p}\right]<\infty \quad \text { for some } p>2,
$$

there exists a concave, positively homogeneous, nonrandom continuous function $\Lambda: \mathbb{R}_{\geqslant 0}^{2} \rightarrow \mathbb{R}$ that satisfies the shape theorem (see [25, section 2.3]):

$$
\begin{equation*}
\lim _{n \rightarrow \infty} \sup _{z \in \mathbb{Z}_{\geqslant 0}^{2}:|z|_{1} \geqslant n} \frac{\left|\log Z_{0, z}-\Lambda(z)\right|}{|z|_{1}}=0 \quad \text { P-almost surely. } \tag{2.1}
\end{equation*}
$$

$\Lambda$ is called the (limiting) free energy density or, by analogy with stochastic growth models, the shape function. Regularity properties of $\Lambda$ such as strict convexity or differentiability are not known in general.

Fix a base point $v \in \mathbb{Z}^{2}$ and let $x_{N} \geqslant v$ in $\mathbb{Z}^{2}$ be a sequence of lattice points going to infinity in a deterministic direction $\xi$, that is, $x_{N} /\left|x_{N}\right|_{1} \xrightarrow[N \rightarrow \infty]{ } \xi /|\xi|_{1}$. The $\xi$-directed semi-infinite polymer measure is obtained as the weak limit

$$
\begin{equation*}
Q_{v, x_{N}} \rightharpoonup \Pi_{v}^{\xi} \quad \text { as } N \rightarrow \infty \tag{2.2}
\end{equation*}
$$

provided this weak limit exists $\mathbb{P}$-almost surely. The probability measure $\Pi_{v}^{\xi}$ is the quenched path measure of a random walk in a random environment (RWRE) on $\mathbb{Z}^{2}$ started at $v$. An RWRE is Markov chain whose transition probability depends on the environment in a translation-covariant way. In the polymer case, these transition probabilities are given by limiting ratios of partition functions. If the shape function $\Lambda$ (as a function of directions) has sufficient local regularity around the direction $\xi$, then the limiting measure $\Pi_{v}^{\xi}$ exists [25, Theorem 3.8].

## 2.2 | Inverse-gamma polymer

This paper focuses exclusively on the inverse-gamma polymer. A real random variable $X$ has the inverse-gamma distribution with shape parameter $\mu \in(0, \infty)$, abbreviated as $X \sim \mathrm{Ga}^{-1}(\mu)$, if its reciprocal $X^{-1}$ has the gamma distribution with shape parameter $\mu$. Equivalently, $X$ has probability density function

$$
f_{X}(x)=\frac{1}{\Gamma(\mu)} x^{-1-\mu} e^{-x^{-1}} \mathbb{1}_{(0, \infty)}(x)
$$

where $\Gamma(a)=\int_{0}^{\infty} s^{a-1} e^{-s} d s$ is the gamma function. The inverse-gamma polymer is defined by letting $\left\{Y_{z}\right\}_{z \in \mathbb{Z}^{2}}$ be independent and identically distributed inverse-gamma distributed random variables. We will fix the shape parameter $\mu$ in the rest of the paper. While many of the constants in the proofs depend on $\mu$, we will not explicitly mention this fact.

In the current state of the subject, $\Lambda$ in (2.1) can be written down explicitly only in the inverse-gamma case. Then the regularity of $\Lambda$ required for (2.2) can be verified explicitly. Hence, for each given direction $\xi$ in the open first quadrant and each initial vertex $v \in \mathbb{Z}^{2}$, the measure $\Pi_{v}^{\xi}$ exists almost surely [18, Theorem 7.1]. Its transition probability is given in Equation (5.2).

Let $\Psi_{0}$ and $\Psi_{1}$ be the digamma and trigamma functions, defined by $\Psi_{0}(z)=\frac{d}{d z} \log \Gamma(z)$ and $\Psi_{1}(z)=\Psi_{0}^{\prime}(z)=\frac{d^{2}}{d z^{2}} \log \Gamma(z)$. In the study of the inverse-gamma polymer, it is convenient to index the spatial directions $\xi$ by the parameter $\rho \in(0, \mu)$ through

$$
\begin{equation*}
\xi[\rho]=\left(\frac{\Psi_{1}(\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)}, \frac{\Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)}\right) . \tag{2.3}
\end{equation*}
$$

We call $\xi[\rho]$ the characteristic direction associated to the parameter $\rho$. This notion acquires its full meaning when we discuss the stationary inverse-gamma polymer in Section 3. The formula for


FIGURE 1 These pictures illustrate the likely events which are the complements of the rare events bounded in Theorems 2.1 and 2.3. The open circle marks the coalescence point of two $\xi[\rho]$-directed semi-infinite polymer paths. On the left $r$ is large and the initial points are far apart on the scale $N^{2 / 3}$. Consequently the two paths are unlikely to coalesce before exiting the rectangle. On the right $\delta$ is small and coalescence inside the rectangle is likely.
the shape function $\Lambda$ is cleanest in terms of the characteristic direction: from [32, (2.16)]

$$
\Lambda(\xi[\rho])=-\frac{\Psi_{1}(\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \cdot \Psi_{0}(\mu-\rho)-\frac{\Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \Psi_{0}(\rho)
$$

Throughout the paper, $N$ is a scaling parameter that goes to infinity. We define the particular sequence of lattice points

$$
\begin{equation*}
v_{N}=\left(\left\lfloor N \xi[\rho] \cdot e_{1}\right\rfloor,\left\lfloor N \xi[\rho] \cdot e_{2}\right\rfloor\right) \in \mathbb{Z}_{\geqslant 0}^{2} \tag{2.4}
\end{equation*}
$$

that go to infinity in the characteristic direction $\xi[\rho]$. We simplify the notation for the semi-infinite polymer distribution to $\Pi_{v}^{\rho}=\Pi_{v}^{\xi[\rho]}$.

## 2.3 | Coalescence bounds

For two initial vertices $a, b \in \mathbb{Z}^{2}$, let $\mathcal{H}_{a, b}^{\rho}$ denote the classical coupling measure of the Markov chains $\Pi_{a}^{\rho}$ and $\Pi_{b}^{\rho}$, as defined by Thorisson [35, chapter 2]. Under the distribution $\mathcal{H}_{a, b}^{\rho}$, the two paths evolve jointly as a Markov chain on $\mathbb{Z}^{2} \times \mathbb{Z}^{2}$ with marginal distributions $\Pi_{a}^{\rho}$ and $\Pi_{b}^{\rho}$. The joint transition probability is defined on $\mathbb{Z}^{2} \times \mathbb{Z}^{2}$ so that the two paths move independently until they meet, after which they move together. When this meeting happens we say that the two paths coalesced. By [25, Theorem A.1], for a given $\rho$, coalescence happens $\mathcal{H}_{a, b}^{\rho}$-almost surely, for almost every environment.

We quantify the speed of coalescence by specifying the lattice subset in which the coalescence first happens. For $A \subset \mathbb{Z}^{2}$, let $\Gamma^{A}$ denote the collection of pairs of semi-infinite up-right paths in $\mathbb{Z}^{2}$ that first meet at a vertex inside the set $A$. Then, $\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{\llbracket 0, v_{N} \rrbracket}\right)$ is the quenched probability that the coalescence of the paths from $a$ and $b$ happens inside the set $\llbracket 0, v_{N} \rrbracket$. Similarly, $\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \rrbracket}\right)$ is the quenched probability that the coalescence happens outside $\llbracket 0, v_{N} \rrbracket$. The two theorems below give upper and lower bounds on the expectations of these quenched probabilities in two distinct cases: when the initial points are close together and when they are far apart on the scale $N^{2 / 3}$. These two scenarios are illustrated in Figure 1.

Theorem 2.1. Let $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, N_{0}, \delta_{0}$ depending only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $N^{-2 / 3} \leqslant \delta \leqslant \delta_{0}$, we have

$$
C_{1} \delta \leqslant \mathbb{E}\left[\mathcal{H}_{\left[\delta N^{2 / 3}\right\rfloor e_{1},\left[\delta N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \rrbracket}\right)\right] \leqslant C_{2}|\log \delta|^{10} \delta .
$$

Remark 2.2. The restriction $\delta \geqslant N^{-2 / 3}$ is needed only for the lower bound of the theorem and only for the trivial reason that the expectation vanishes when $\delta<N^{-2 / 3}$ because then the two paths start together at the origin.

Theorem 2.3. Let $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, r_{0}, c_{0}, N_{0}$ that depend only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r_{0} \leqslant r \leqslant c_{0} N^{1 / 3}$, we have

$$
e^{-C_{1} r^{3}} \leqslant \mathbb{E}\left[\mathcal{H}_{\left[r N^{2 / 3}\right\rfloor e_{1},\left\lfloor r N^{2 / 3}\right] e_{2}}\left(\Gamma^{\llbracket 0, v_{N} \rrbracket}\right)\right] \leqslant e^{-C_{2} r^{3}} .
$$

Remark 2.4. Again, the upper bound $r \leqslant c_{0} N^{1 / 3}$ is only needed for the lower bound in the theorem.

The estimates above do not depend on starting the paths on an antidiagonal. The following corollary gives two of the four additional estimates. The other two follow from the theorems. Also, $e_{1}$ and $e_{2}$ are interchangeable by symmetry.

Corollary 2.5. Let $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C, N_{0}, \delta_{0}, r_{0}$ that depend only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}, r \geqslant r_{0}$ and $N^{-2 / 3} \leqslant \delta \leqslant \delta_{0}$, we have

$$
\mathbb{E}\left[\mathcal{H}_{0,\left\lfloor r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\Gamma^{\llbracket 0, v_{N} \rrbracket}\right)\right] \leqslant e^{-C r^{3}} \quad \text { and } \quad \mathbb{E}\left[\mathcal{H}_{0,\left\lfloor\delta N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \mathbb{1}}\right)\right] \geqslant C \delta .
$$

By planar monotonicity and a change of variable, our estimates can also be stated for two semi-infinite polymer paths that start at fixed locations. If the initial points are of order $k$ apart, then their meeting takes place on the scale $k^{3 / 2}$, as captured in the corollary below. We shift the rectangle with the initial points so that the constants do not depend at all on the initial points. The coordinate-wise minimum of two lattice points $a=\left(a_{1}, a_{2}\right)$ and $b=\left(b_{1}, b_{2}\right)$ is denoted by $a \wedge b=\left(a_{1} \wedge b_{1}, a_{2} \wedge b_{2}\right)$.

Corollary 2.6. Let $\varepsilon \in(0, \mu / 2)$ and $a \neq b$ in $\mathbb{Z}^{2}$. Let $k=|a-b|_{1} \geqslant 1$. There exist positive constants $C_{1}, C_{2}, r_{0}, c_{0}$ that depend only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], k \geqslant 1, r \geqslant r_{0}$ and $\delta \geqslant c_{0} k^{-1 / 2}$ we have

$$
\begin{aligned}
& C_{1} r^{-2 / 3} \leqslant \mathbb{E}\left[\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash\left\{a \wedge b+\llbracket 0, v_{r k^{3} / 2} \rrbracket\right\}}\right)\right] \leqslant C_{2}(\log r)^{10} r^{-2 / 3} \quad \text { and } \\
& e^{-C_{2} \delta^{-2}} \leqslant \mathbb{E}\left[\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{a \wedge b+\llbracket 0, v_{\delta k^{3} / 2} \rrbracket}\right)\right] \leqslant e^{-C_{1} \delta^{-2}} .
\end{aligned}
$$

The next result gives tail bounds for the quenched probability of fast coalescence, of optimal exponential order.

Theorem 2.7. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, C_{3}, C_{4}, r_{0}, c_{0}, N_{0}$ that depend only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r_{0} \leqslant r \leqslant c_{0} N^{1 / 3}$, we have

$$
\begin{aligned}
e^{-C_{1} r^{3}} & \leqslant \mathbb{P}\left(\mathcal{H}_{\left\lfloor r N^{2 / 3}\right\rfloor e_{1},\left\lfloor r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\Gamma^{\left[0, v_{N} \mathbb{1}\right.}\right) \geqslant 1-e^{-C_{2} r^{2} N^{1 / 3}}\right) \\
& \leqslant \mathbb{P}\left(\mathcal{H}_{\left\lfloor r N^{2 / 3}\right\rfloor e_{1},\left\lfloor r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\Gamma^{\left[0, v_{N} \rrbracket\right.}\right) \geqslant e^{-C_{3} r^{2} N^{1 / 3}}\right) \leqslant e^{-C_{4} r^{3}} .
\end{aligned}
$$

## 2.4 | Coupling and total variation distance

As the quenched noncoalescence probability $\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash\left(a \wedge b+\llbracket 0, v_{r k^{3} / 2} \mathbb{I}\right)}\right)$ is nonincreasing in $r$, Corollary 2.6 implies the almost sure convergence $\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash\left(a \wedge b+\llbracket 0, v_{r k}{ }^{3 / 2} \rrbracket\right)}\right) \rightarrow 0$ as $r \rightarrow \infty$. This says that the polymer distributions $\Pi_{a}^{\rho}$ and $\Pi_{b}^{\rho}$ couple almost surely. To state this precisely, let $\chi_{N}=\chi_{N}(\gamma)$ denote the vertex where a semi-infinite up-right path $\gamma$ started inside $\llbracket 0, v_{N} \rrbracket$ first meets the northeast boundary $\partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket$. If $\left(\gamma^{a}, \gamma^{b}\right)$ denote the paths under $\mathcal{H}_{a, b}^{\rho}$, then for $a, b \in \mathbb{Z}_{\geqslant 0}^{2}$ we have

$$
\begin{equation*}
\mathcal{H}_{a, b}^{\rho}\left\{\chi_{N}\left(\gamma^{a}\right)=\chi_{N}\left(\gamma^{b}\right) \text { for large enough } N\right\}=1 . \tag{2.5}
\end{equation*}
$$

The standard coupling inequality (stated in (6.1) in Section 6) implies that the total variation distance between the distributions induced on $\partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket$ converges to zero almost surely:

$$
\begin{equation*}
\lim _{N \rightarrow \infty} d_{\mathrm{TV}}\left(\Pi_{a}^{\rho}\left\{\chi_{N} \in \cdot\right\}, \Pi_{b}^{\rho}\left\{\chi_{N} \in \cdot\right\}\right)=0 \quad \mathbb{P} \text {-almost surely. } \tag{2.6}
\end{equation*}
$$

The next two theorems establish bounds on this convergence. In the same spirit as in the earlier results, when the initial points are close on the scale $N^{2 / 3}$, the total variation distance on the northeast boundary of a rectangle of size $N$ is small. In the opposite case the starting points are far apart on the scale $N^{2 / 3}$ and the total variation distance is close to 1 .

Theorem 2.8. Let $\varepsilon \in(0, \mu / 2)$. There exist finite strictly positive constants $\delta_{0}, N_{0}, C$ that depend on $\varepsilon$ such that, whenever $0<\delta \leqslant \delta_{0}, N \geqslant N_{0}$ and $\rho \in[\varepsilon, \mu-\varepsilon]$,

$$
\mathbb{E}\left[d_{\mathrm{TV}}\left(\Pi_{\left[\delta N^{2 / 3}\right] e_{1}}^{\rho}\left(\chi_{N} \in \cdot\right), \Pi_{\left[\delta N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N} \in \cdot\right)\right)\right] \leqslant C|\log \delta|^{10} \delta
$$

Theorem 2.9. Let $\varepsilon \in(0, \mu / 2)$. There exist finite positive constants $r_{0}, N_{0}, C$ depending on $\varepsilon$ such that whenever $N \geqslant N_{0}, r_{0} \leqslant r \leqslant N^{1 / 3}$ and $\rho \in[\varepsilon, \mu-\varepsilon]$, we have

$$
\mathbb{E}\left[d_{\mathrm{TV}}\left(\Pi_{\left[r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\chi_{N} \in \bullet\right), \Pi_{\left[r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N} \in \cdot\right)\right)\right] \geqslant 1-e^{-C r^{3}}
$$

The proofs of the two theorems are given in Section 6.
Remark 2.10 (Hyperbolicity in stochastic equations). The free energy of a directed polymer can be viewed as a discretization of a stochastically forced viscous Hamilton-Jacobi equation. This connection goes back to [23, 24]. In this vein, semi-infinite polymer measures can be used to
construct stationary eternal solutions to such equations. Article [1] treats a semidiscrete case and [26] the KPZ equation. In particular, the limit (2.6) is a version of hyperbolicity that appears in stochastic synchronization (also called the one force-one solution principle) of such equations. This is the positive temperature analogue of the inviscid phenomenon whereby action minimizers are asymptotic to each other in the infinite past. See, for example, [1, Theorem 4.4]. Our results above show that, in the case at hand, this form of hyperbolicity obeys the KPZ wandering exponent. On universality grounds one can predict that this is true in some generality in one space dimension for stochastically forced viscous Hamilton-Jacobi equations with nonlinear Hamiltonians.

## 2.5 | Transversal fluctuations

Finally, we present a result concerning the transversal fluctuation of the finite independent and identically distributed polymer. This result is derived by making a slight modification to the proof of the upper bound for fast coalescence, as stated in Theorem 2.1. It is expected for the midpoint of polymer from $(0,0)$ to $(N, N)$ to fluctuate around the diagonal on the scale $N^{2 / 3}$. The upper bound on the transversal fluctuation was first proved in the work [32], and we provide here the lower bound, that is, we show that it is rare for the midpoint of the polymer to be too close to the diagonal.

To state the result, let us introduce some notation. Let $\{$ mid $\leqslant k\}$ denote the collection of directed paths between $-v_{N}$ and $v_{N}$ that intersect the $\ell^{\infty}$ ball of radius $k$, centered at the origin.

Theorem 2.11. Let $\varepsilon \in(0, \mu / 2)$. There exist finite strictly positive constants $\delta_{0}, N_{0}, C$ that depend on $\varepsilon$ such that, whenever $0<\delta \leqslant \delta_{0}, N \geqslant N_{0}$ and $\rho \in[\varepsilon, \mu-\varepsilon]$,

$$
\mathbb{E}\left[Q_{-v_{N}, v_{N}}\left\{\operatorname{mid} \leqslant \delta N^{2 / 3}\right\}\right] \leqslant C|\log \delta|^{10} \delta .
$$

Remark 2.12. The midpoint transversal fluctuation can be generalized to other positions along the path, as long as they are order $N$ away from $-v_{N}$ and $v_{N}$.

Remark 2.13. Our proof technique also yields the following lower bound on the fluctuation of the endpoint of the point-to-line polymer. Let $Q_{0, N}^{\mathrm{p} 21}$ denote the point-to-line quenched path measure on the collection of directed paths from $(0,0)$ to the anti-diagonal line $x+y=2 N$. And let \{end $\leqslant$ $k\}$ denote the sub-collection of these paths that intersect the $\ell^{\infty}$ ball of radius $k$, centered at $(N, N)$. It holds that

$$
\begin{equation*}
\mathbb{E}\left[Q_{0, N}^{\mathrm{p} 21}\left\{\mathrm{end} \leqslant \delta N^{2 / 3}\right\}\right] \leqslant C|\log \delta|^{10} \sqrt{\delta} . \tag{2.7}
\end{equation*}
$$

We get the weaker $\sqrt{\delta}$ instead of $\delta$ because the antidiagonal version of the independence property of Busemann increments on horizontal or vertical lines for two different directions is not known.

## 3 | STATIONARY INVERSE-GAMMA POLYMER

One of the main tools we use in our proofs is a stationary version of the polymer model, which we now describe.

The stationary inverse-gamma polymer with southwest boundary is defined on a quadrant instead of the entire $\mathbb{Z}^{2}$. It requires a parameter parameter $\rho \in(0, \mu)$ and a base vertex $v \in \mathbb{Z}^{2}$. To each $z \in v+\mathbb{Z}_{>0}^{2}$ we attach a weight $Y_{z} \sim \mathrm{Ga}^{-1}(\mu)$. On the $e_{1}$ - and $e_{2}$-boundary of $v+\mathbb{Z}_{\geqslant 0}^{2}$, we place (edge) weights

$$
\begin{equation*}
I_{v+k e_{1}}^{\rho} \sim \mathrm{Ga}^{-1}(\mu-\rho) \quad \text { and } \quad J_{v+k e_{2}}^{\rho} \sim \mathrm{Ga}^{-1}(\rho), \quad k \geqslant 1 . \tag{3.1}
\end{equation*}
$$

All these weights in the quadrant are independent. We refer to the $Y$ weights as the bulk weights and to the $I^{\rho}$ and $J^{\rho}$ weights as the $\rho$-boundary weights. Subsection 5.1 explains the reason behind thinking of $I^{\rho}$ and $J^{\rho}$ as edge weights instead of vertex weights.

We use the same $\mathbb{P}$ to denote the joint distribution of the weights $\left(Y, I^{\rho}, J^{\rho}\right)$. For $w \in v+\mathbb{Z}_{\geqslant 0}^{2}$, we define the partition function of the stationary polymer by

$$
Z_{v, w}^{\rho}=\sum_{x \in \mathbb{X}_{v, w}} \prod_{i=0}^{|w-v|_{1}} \widetilde{Y}_{x_{i}}, \text { where for } x \in v+\mathbb{Z}_{\geqslant 0}^{2}, \quad \widetilde{Y}_{x}= \begin{cases}1 & \text { if } x=v, \\ I_{x-e_{1}, x}^{\rho} & \text { if } x \in v+\mathbb{Z}_{>0} e_{1} \\ J_{x-e_{2}, x}^{\rho} & \text { if } x \in v+\mathbb{Z}_{>0} e_{2} \\ Y_{x} & \text { for } x \in v+\mathbb{Z}_{>0}^{2}\end{cases}
$$

The corresponding quenched polymer measure is defined as

$$
Q_{v, w}^{\rho}\left(x_{\mathbf{\bullet}}\right)=\frac{1}{Z_{v, w}^{\rho}} \prod_{i=0}^{|w-v|_{1}} \widetilde{Y}_{x_{i}}, \quad x_{\mathbf{0}} \in \mathbb{X}_{v, w} .
$$

Next we state the theorem that explains why the process $Z^{\rho}$ is called ratio-stationary, or simply stationary. For a subset $A \subset \mathbb{Z}^{2}$, let $A^{<}=\cup_{x \in A}\left(x+\mathbb{Z}_{<0}^{2}\right)$.

Theorem 3.1 [32, Theorem 3.3] and [18, eq. (3.6)]. Fix $\rho \in(0, \mu)$. For each $u \in v+\left(\mathbb{Z}_{>0} \times \mathbb{Z}_{\geqslant 0}\right)$, $w \in v+\left(\mathbb{Z}_{\geqslant 0} \times \mathbb{Z}_{>0}\right)$, and $x \in v+\mathbb{Z}_{\geqslant 0}^{2}$ we have

$$
\frac{Z_{v, u}^{\rho}}{Z_{v, u-e_{1}}^{\rho}} \sim \mathrm{Ga}^{-1}(\mu-\rho), \quad \frac{Z_{v, w}^{\rho}}{Z_{v, w-e_{2}}^{\rho}} \sim \mathrm{Ga}^{-1}(\rho), \quad \text { and } \quad \frac{1}{Z_{v, x+e_{1}}^{\rho} / Z_{v, x}^{\rho}+Z_{v, x+e_{2}}^{\rho} / Z_{v, x}^{\rho}} \sim \mathrm{Ga}^{-1}(\mu)
$$

Translation invariance: The distribution of the process

$$
\left\{\frac{Z_{v, z+u}^{\rho}}{Z_{v, z+u-e_{1}}^{\rho}}, \frac{Z_{v, z+w}^{\rho}}{Z_{v, z+w-e_{2}}^{\rho}}: u \in \mathbb{Z}_{>0} \times \mathbb{Z}_{\geqslant 0}, w \in \mathbb{Z}_{\geqslant 0} \times \mathbb{Z}_{>0}\right\}
$$

does not depend on the translation $z \in v+\mathbb{Z}_{\geqslant 0}^{2}$. Furthermore, let $A=\left\{y_{i}\right\}_{i \in \mathcal{I}}$ be any finite or infinite down-right path in $v+\mathbb{Z}_{\geqslant 0}^{2}$, indexed by an interval $\mathcal{I} \subset \mathbb{Z}$. (This means that each increment satisfies $y_{i+1}-y_{i} \in\left\{e_{1},-e_{2}\right\}$.) Then, the nearest-neighbor ratios $\left\{Z_{v, y_{i+1}}^{\rho} / Z_{v, y_{i}}^{\rho}\right\}$ along the path and the weights $\left\{\left(Z_{v, x+e_{1}}^{\rho} / Z_{v, x}^{\rho}+Z_{v, x+e_{2}}^{\rho} / Z_{v, x}^{\rho}\right)^{-1}: x \in A^{<} \cap\left(v+\mathbb{Z}^{2}\right)\right\}$ are mutually independent.

A key quantity in the coupling approach to polymers and LPP models is the exit time. For an up-right path $\gamma$, we define $\tau(\gamma) \in \mathbb{Z} \backslash\{0\}$ as the signed number of steps taken before the first turn,
where the plus sign corresponds to $e_{1}$ steps and the minus sign to $e_{2}$ steps. For example, $\tau(\gamma)=-3$ means that the first four steps of $\gamma$ consist of three consecutive $e_{2}$ steps followed by an $e_{1}$ step. For $v, w \in \mathbb{Z}$, when additional clarity is needed, we use the notation $\tau_{v, w}$ to denote the restriction of the function $\tau$ to the domain $\mathbb{X}_{v, w}$. When the path $\gamma$ starts at the base vertex $v$ of the stationary polymer process, $|\tau|$ equals the number of boundary weights seen by the path before it exits the boundary. This justifies the term exit time for $\tau(\gamma)$.

With the function $\tau$, we define the restricted partition function $Z_{v, w}(a \leqslant \tau \leqslant b)$ similarly to $Z_{v, w}$, except that we sum only over the subset of paths $\left\{x . \in \mathbb{X}_{v, w}: a \leqslant \tau_{v, w}\left(x_{\mathbf{0}}\right) \leqslant b\right\}$.

Because the weights on the boundary are stochastically larger than the bulk weights, the path prefers to stay on the boundary. For each $\rho \in(0, \mu)$ the characteristic direction $\xi[\rho]$ is the unique direction in which the pulls of the $e_{1}$ - and $e_{2}$-boundaries balance out. The sampled path between the origin and $v_{N}$ tends to take order $N^{2 / 3}$ steps on the boundary. Precise exit time estimates are stated in in Section 4.

The stationary inverse-gamma polymer with northeast boundary is analogous to the previously defined model, except that it is defined on a third quadrant and uses boundary edge weights placed on the northeast boundary. Thus, it also requires a parameter $\rho \in(0, \mu)$ and a base vertex $v \in \mathbb{Z}^{2}$, but it is defined on the quadrant $v-\mathbb{Z}_{\geqslant 0}^{2}$. To each $z \in v+\mathbb{Z}_{<0}^{2}$ we attach a bulk (vertex) weight $Y_{z} \sim \mathrm{Ga}^{-1}(\mu)$. On the $e_{1}$ - and $e_{2}$-boundary of $v-\mathbb{Z}_{\geqslant 0}^{2}$, we place edge weights

$$
\begin{align*}
& I_{\llbracket v+(k-1) k e_{1}, v+k e_{1} \rrbracket}^{\rho}=I_{v+(k-1) k e_{1}, v+k e_{1}}^{\rho} \sim \mathrm{Ga}^{-1}(\mu-\rho),  \tag{3.2}\\
& J_{\llbracket v+(k-1) k e_{2}, v+k e_{2} \rrbracket}^{\rho}=J_{v+(k-1) k e_{2}, v+k e_{2}}^{\rho} \sim \mathrm{Ga}^{-1}(\rho), \quad k \leqslant 0 .
\end{align*}
$$

All these weights in the quadrant are independent. Here too, we use $\mathbb{P}$ to denote the joint distribution of $\left(Y, I^{\rho}, J^{\rho}\right)$ and write $Z_{u, v}^{\rho, \mathrm{NE}}$ and $Q_{u, v}^{\rho, \mathrm{NE}}$ for, respectively, the partition function and quenched measure for the polymer with northeast boundary. Precisely, for $u \in v-\mathbb{Z}_{\geqslant 0}^{2}$, define

$$
Z_{u, v}^{\rho, \mathrm{NE}}=\sum_{x \in \in \mathbb{X}_{u, v}} \prod_{i=0}^{|v-u|_{1}} \widetilde{Y}_{x_{i}}, \text { where for } x \in v-\mathbb{Z}_{\geqslant 0}^{2}, \quad \widetilde{Y}_{x}= \begin{cases}1 & \text { if } x=v, \\ I_{x, x+e_{1}}^{\rho} & \text { if } x \in v-\mathbb{Z}_{>0} e_{1}, \\ J_{x, x+e_{2}}^{\rho} & \text { if } x \in v-\mathbb{Z}_{>0} e_{2}, \\ Y_{x} & \text { for } x \in v-\mathbb{Z}_{>0}^{2} .\end{cases}
$$

The quenched polymer measure is defined by

$$
Q_{u, v}^{\rho, \mathrm{NE}}\left(x_{.}\right)=\frac{1}{Z_{u, v}^{\rho, \mathrm{NE}}} \prod_{i=0}^{|v-u|_{1}} \widetilde{Y}_{x_{i}} .
$$

Remark 3.2. We work mostly with the stationary model with southwest boundary and, therefore, we only flesh out the location of the boundary when it is the northeast boundary that is being used.

By symmetry, the analogous version of Theorem 3.1 holds for the stationary polymer with northeast boundary.

## 4 | EXIT TIME ESTIMATES

In this section, we prove exit time estimates for the stationary polymer model with southwest boundary, introduced in Section 3. These results will be used to derive the coalescence estimate in Section 5 and the total variation bounds in Section 6.

The first theorem below concerns the case when the polymer paths have an unusually large exit time. The upper bound for the annealed measure is proved in [15, 28]. We improve this estimate into a bound for the quenched tail. The related upper bound in the zero-temperature model is [8, Theorem 2.4]. The proof in [8] uses a technical result from [7, Theorem 10.5]. We will present a simpler proof in this paper.

Theorem 4.1. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $r_{0}, N_{0}, c_{0}$, and $C_{i}, i \in \llbracket 1,6 \rrbracket$, that depend only on $\varepsilon$ such that for all $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r_{0} \leqslant r \leqslant c_{0} N^{1 / 3}$, we have

$$
\begin{align*}
e^{-C_{1} r^{3}} & \leqslant \mathbb{P}\left(\min _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \geqslant 1-e^{-C_{2} r^{2} N^{1 / 3}}\right) \\
& \leqslant \mathbb{P}\left(Q_{0, v_{N}+(1,1)}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \geqslant e^{-C_{3} r^{2} N^{1 / 3}}\right) \leqslant e^{-C_{4} r^{3}} \tag{4.1}
\end{align*}
$$

and

$$
e^{-C_{5} r^{3}} \leqslant \mathbb{E}\left[\min _{x \notin \mathbb{0}, v_{N} \mathbb{1}} Q_{0, v_{N}}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\}\right] \leqslant \mathbb{E}\left[Q_{0, v_{N}+(1,1)}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\}\right] \leqslant e^{-C_{6} r^{3}} .
$$

The next theorem is about the polymer paths having unusually small exit times. The estimate improves upon the result from [9] where these types of estimates were used to rule out the existence of nontrivial bi-infinite polymer measures. This technique was first developed for the nonexistence of bi-infinite geodesics in the corner growth model [2] and subsequently applied to coalescence estimates for semi-infinite geodesics in [34].

Theorem 4.2. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, N_{0}, \delta_{0}$ that depend only on $\varepsilon$ such that for all $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}, N^{-2 / 3}<\delta \leqslant \delta_{0}$, we have

$$
\begin{equation*}
\mathbb{P}\left(\max _{x \notin \mathbb{0}, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \leqslant C_{1}|\log \delta|^{10} \delta \tag{4.2}
\end{equation*}
$$

and

$$
\begin{equation*}
C_{1} \delta \leqslant \mathbb{E}\left[\max _{x \notin \mathbb{0}, u_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}\right] \leqslant C_{2}|\log \delta|^{10} \delta . \tag{4.3}
\end{equation*}
$$

We close this section by extending the above estimates to any coupling of stationary polymer measures. Let $\widetilde{Q}_{0, A}^{\rho}$ be any coupling of the measures $\left\{Q_{0, x}^{\rho}: x \in A\right\}$. This is then a probability measure on the product space $\prod_{y \in A} \mathbb{X}_{0, y}$. We view the elements of this product space as vectors and then for $x \in A$, the $x$ th coordinate of such a vector would be the path that ends at $x$. For $x \in A$, define the map $\widetilde{\tau}_{0, x}: \prod_{y \in A} \mathbb{X}_{0, y} \rightarrow \mathbb{Z}$ that records the exit time of the path in the $x$ th coordinate of any vector in $\prod_{y \in A} \mathbb{X}_{0, y}$.

Theorem 4.3. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, r_{0}, c_{0}, N_{0}$ that depend only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r_{0} \leqslant r \leqslant c_{0} N^{1 / 3}$, we have

$$
\mathbb{P}\left(\widetilde{Q}_{0, \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}^{\rho}\left(\bigcap_{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{\left|\widetilde{\tau}_{0, x}\right| \geqslant r N^{2 / 3}\right\}\right) \geqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right) \geqslant e^{-C_{2} r^{3}}
$$

and

$$
\mathbb{E}\left[\widetilde{Q}_{0, \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}^{\rho}\left(\bigcap_{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{\left|\widetilde{\tau}_{0, x}\right| \geqslant r N^{2 / 3}\right\}\right\} \geqslant e^{-C r^{3}} .\right.
$$

Theorem 4.4. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C, N_{0}, \delta_{0}$ that depend only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}, K \geqslant 1$ and $0<\delta \leqslant \delta_{0}$, we have

$$
\mathbb{E}\left[\widetilde{Q}_{0, \partial \mathrm{NE} \llbracket 0, v_{N} \rrbracket}^{\rho}\left(\bigcup_{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{\left|\widetilde{\tau}_{0, x}\right| \leqslant \delta N^{2 / 3}\right\}\right)\right] \leqslant C|\log \delta|^{10} \delta .
$$

## 4.1 | Proof of Theorem 4.1

The expectation bounds in Theorem 4.1 follow directly from the tail bounds. We split the proof of the tail bounds into the following two lemmas.

Lemma 4.5. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, r_{0}, N_{0}$ depending only on $\varepsilon$ such that for all $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r \geqslant r_{0}$, we have

$$
\mathbb{P}\left(Q_{0, v_{N}}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \leqslant e^{-C_{2} r^{3}} .
$$

Lemma 4.6. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, r_{0}, N_{0}, c_{0}$ depending only on $\varepsilon$ such that for all $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r_{0} \leqslant r \leqslant c_{0} N^{1 / 3}$, we have

$$
\mathbb{P}\left(\min _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \geqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right) \geqslant e^{-C_{2} r^{3}}
$$

### 4.1.1 | Proof of Lemma 4.5

We start with two calculations for the shape function $\Lambda$. Their proofs use Taylor expansions and are thus postponed to Appendix A.2.

The first proposition below captures the loss of free energy due to curvature.
Proposition 4.7. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, N_{0}, c_{0}$ depending only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}, 1 \leqslant s \leqslant c_{0} N^{1 / 3}$, we have

$$
\Lambda\left(v_{N}-\left\lfloor s N^{2 / 3}\right\rfloor e_{1}+\left\lfloor s N^{2 / 3}\right\rfloor e_{2}\right)-\left\lfloor s N^{2 / 3}\right\rfloor \Psi_{0}(\mu-\rho)+\left\lfloor s N^{2 / 3}\right\rfloor \Psi_{0}(\rho)-\Lambda\left(v_{N}\right) \leqslant-C_{1} s^{2} N^{1 / 3}
$$



FIGURE 2 Sketch of Corollary 4.10. On the left is a path in the event $\tau_{0, v_{N}-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}} \geqslant 1$. On the right, a second base point is placed at $-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}$ and the edge weights on the $e_{2}$-axis based at 0 are determined by the ratio variables of the polymer based at $-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}$. By Lemma A.7, $Q_{0, v_{N}-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}}\{\tau \geqslant 1\}=Q_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, v_{N}-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}}\left\{\tau \geqslant\left\lfloor r N^{2 / 3}\right\rfloor+1\right\}$, and Theorem 4.9 can be applied.

The second proposition is essentially a bound on the nonrandom fluctuation when the endpoint varies around $v_{N}$.

Proposition 4.8. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, N_{0}, c_{0}$ depending only on $\varepsilon$ such that for each $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}, 0 \leqslant s \leqslant 3$, we have

$$
\left|\Lambda\left(v_{N}-\left\lfloor s N^{2 / 3}\right\rfloor e_{1}+\left\lfloor s N^{2 / 3}\right\rfloor e_{2}\right)-\left\lfloor s N^{2 / 3}\right\rfloor \Psi_{0}(\mu-\rho)+\left\lfloor s N^{2 / 3}\right\rfloor \Psi_{0}(\rho)-\Lambda\left(v_{N}\right)\right| \leqslant C_{1} N^{1 / 3}
$$

Next, we recall the established annealed version of the exit time estimate, which, through the Markov inequality, yields the expression (4.4). This equation represents a nonoptimal variant of the upper bound presented in Lemma 4.5, as it has $e^{-C r^{3}}$ instead of $e^{-C r^{2} N^{1 / 3}}$ within the probability measure. It is important to note that this nonoptimal version alone is not enough to prove the coalescence estimate later on. Consequently, Lemma 4.5 assumes a pivotal role in advancing the arguments laid out in the paper.

Theorem 4.9 [15, 28]. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, C_{3}, r_{0}, N_{0}$ that depend only on $\varepsilon$ such that for for all $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r \geqslant r_{0}$, we have

$$
\mathbb{E}\left[Q_{0, v_{N}}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\}\right] \leqslant e^{-C_{1} r^{3}}
$$

And by Markov inequality,

$$
\begin{equation*}
\mathbb{P}\left(Q_{0, v_{N}}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \geqslant e^{-C_{2} r^{3}}\right) \leqslant e^{-C_{3} r^{3}} \tag{4.4}
\end{equation*}
$$

Lemma A. 7 allows us to obtain the following corollary from Theorem 4.9. The proof of Corollary 4.10 is by now standard and is summarized in Figure 2 and its caption.

Corollary 4.10. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C_{1}, C_{2}, C_{3}, r_{0}, N_{0}$ that depend only on $\varepsilon$ such that for for all $\rho \in[\varepsilon, \mu-\varepsilon], N \geqslant N_{0}$ and $r \geqslant r_{0}$, we have

$$
\mathbb{P}\left(Q_{0, v_{N}-r N^{2 / 3} e_{1}}^{\rho}\{\tau \geqslant 1\} \geqslant e^{-C_{1} r^{3}}\right) \leqslant e^{-C_{2} r^{3}}
$$

and

$$
\mathbb{E}\left[Q_{0, v_{N}-r N^{2 / 3} e_{1}}^{\rho}\{\tau \geqslant 1\}\right] \leqslant e^{-C_{3} r^{3}} .
$$

The same result holds when $v_{N}-r N^{2 / 3} e_{1}$ is replaced by $v_{N}+r N^{2 / 3} e_{2}$.
With these results, we obtain the following estimate for the maximum free energy.

Proposition 4.11. For each $\varepsilon \in(0, \mu / 2)$, there exist positive constants $C_{1}, C_{2}, N_{0}, c_{0}$ depending on $\varepsilon$ such that for each $N \geqslant N_{0}$ and $1 \leqslant r \leqslant c_{0} N^{2 / 3}$, we have

$$
\mathbb{P}\left(\max _{k \in \llbracket 0,3\left\lfloor N^{2 / 3}\right\rfloor \rrbracket}\left\{\log Z_{0, v_{N}+(-k, k)}-\Lambda\left(v_{N}+(-k, k)\right)\right\} \geqslant C_{1} r N^{1 / 3}\right) \leqslant e^{-C_{2} r^{3 / 2}} .
$$

Proof. To start, let us separate the probability that we are trying to bound into two parts.

$$
\begin{gather*}
\mathbb{P}\left(\max _{k \in \llbracket 0,3\left\lfloor N^{2 / 3} \rrbracket \rrbracket\right.}\left\{\log Z_{0, v_{N}+(-k, k)}-\Lambda\left(v_{N}+(-k, k)\right)\right\} \geqslant C^{\prime} r N^{1 / 3}\right) \\
\leqslant \mathbb{P}\left(\max _{k \in \llbracket 0,3\left\lfloor N^{2 / 3}\right\rfloor \rrbracket}\left\{\log Z_{0, v_{N}+(-k, k)}-\log Z_{0, v_{N}}-\left[\Lambda\left(v_{N}+(-k, k)\right)-\Lambda\left(v_{N}\right)\right]\right\} \geqslant \frac{C^{\prime}}{2} r N^{1 / 3}\right)  \tag{4.5}\\
+\mathbb{P}\left(\log Z_{0, v_{N}}-\Lambda\left(v_{N}\right) \geqslant \frac{C^{\prime}}{2} r N^{1 / 3}\right) \tag{4.6}
\end{gather*}
$$

Using Proposition A.1, (4.6) $\leqslant e^{-C r^{3 / 2}}$. To bound (4.5), we reformulate the problem into a bound for running maxima of random walks. First, by Proposition 4.8 , if $C^{\prime} \geqslant 4 C_{1}$ and $r \geqslant 1$, then

$$
\begin{equation*}
(4.5) \leqslant \mathbb{P}\left(\max _{k \in \llbracket 0,3\left[N^{2 / 3}\right\rfloor \rrbracket}\left\{\log Z_{0, v_{N}+(-k, k)}-\log Z_{0, v_{N}}-\left[k \Psi_{0}(\mu-\rho)-k \Psi_{0}(\rho)\right]\right\} \geqslant \frac{C^{\prime}}{4} r N^{1 / 3}\right) . \tag{4.7}
\end{equation*}
$$

Next, we will show that the quantity $\log Z_{0, v_{N}+(-k, k)}-\log Z_{0, v_{N}}$ can be compared to a random walk with independent and identically distributed steps. To do this, we will place boundary weights on the south-west boundary of $(-1,-1)+\mathbb{Z}_{\geqslant 0}^{2}$ with parameters $\lambda=\rho-q_{0} \sqrt{r} N^{-1 / 3}$ and $\mu-\lambda$. Here, $q_{0}$ will be fixed sufficiently large so that the situation from Figure 3 happens: if we trace the $-\xi[\lambda]$-directed ray from $v_{N}$, it crosses the vertical line $x=-1$ above the point $\left(-1, r N^{2 / 3}\right)$. Then the $c_{0}$ from the statement of our proposition can be now fixed sufficiently small so that $\lambda$ stays between $(0, \mu)$. These choices depend only on $\varepsilon$.

Because $\left(-1, r N^{2 / 3}\right)$ is far away from $(-1,-1)$ on the scale $N^{2 / 3}$, by a similar argument to Corollary 4.10, we have

$$
\begin{equation*}
\mathbb{P}\left(Q_{(-1,-1), v_{N}}^{\lambda}\{\tau \geqslant 1\} \geqslant 1 / 10\right) \leqslant e^{-C r^{3}} . \tag{4.8}
\end{equation*}
$$

Let us denote the complement of the event above as

$$
A=\left\{Q_{(-1,-1), v_{N}}^{\lambda}\{\tau \leqslant-1\}>9 / 10\right\} .
$$



FIGURE 3 The random walk set up in Proposition 4.11.

In the calculation below, let $Z_{(-1,0), x}^{\lambda, \mathrm{W}}$ denote the partition function for up-right paths from $(-1,0)$ to $x$, which uses the same weights as $Z_{(-1,-1), x}^{\lambda}$ does on the west boundary but uses the original (bulk) weights on $\mathbb{Z}_{\geqslant 0}^{2}$. For each $i=0,1, \ldots, 3\left\lfloor N^{2 / 3}\right\rfloor-1$, we have

$$
\begin{aligned}
& \frac{Z_{0, v_{N}+(-i-1, i+1)}}{Z_{0, v_{N}+(-i, i)}} \\
& \leqslant \frac{Z_{(-1,0), v_{N}+(-i-1, i+1)}^{\lambda, \mathrm{W}}}{Z_{(-1,0), v_{N}+(-i, i)}^{\lambda, \text { est }}}=\frac{Z_{(-1,0), v_{N}+(-i-1, i+1)}^{\lambda, \mathrm{W}}}{Z_{(-1,0), v_{N}+(-i, i)}^{\lambda, \mathrm{W}} \cdot \frac{I_{\mathbb{I}(-1,-1),(-1,0) \rrbracket}^{\lambda}}{I_{\mathbb{I}(-1,-1),(-1,0) \rrbracket}^{\lambda}} \quad \text { by Proposition A.3 }} \\
& =\frac{Z_{(-1,-1), v_{N}+(-i-1, i+1)}^{\lambda}(\tau \leqslant-1)}{Z_{(-1,-1), v_{N}+(-i, i)}^{\lambda}(\tau \leqslant-1)}=\frac{Q_{(-1,-1), v_{N}+(-i-1, i+1)}^{\lambda}(\tau \leqslant-1)}{Q_{(-1,-1), v_{N}+(-i, i)}^{\lambda}(\tau \leqslant-1)} \cdot \frac{Z_{(-1,-1), v_{N}+(-i-1, i+1)}^{\lambda}}{Z_{(-1,-1), v_{N}+(-i, i)}^{\lambda}} \\
& \leqslant \frac{10}{9} \frac{Z_{(-1,-1), v_{N}+(-i-1, i+1)}^{\lambda}}{Z_{(-1,-1), v_{N}+(-i, i)}^{\lambda}} \quad \text { on the event } A .
\end{aligned}
$$

By Theorem 3.1, we can define

$$
S_{k}^{\lambda}=\sum_{i=0}^{k-1} \log \frac{Z_{(-1,-1), v_{N}+(-i-1, i+1)}^{\lambda}}{Z_{(-1,-1), v_{N}+(-i, i)}^{\lambda}}
$$

which is an independent and identically distributed random walk whose step has the same distribution as $\log G_{1}-\log G_{2}$, where $G_{1}$ and $G_{2}$ are independent, respectively, $\mathrm{Ga}(\mu-\lambda)$ and $\mathrm{Ga}(\lambda)$ random variables. Consequently, we have

$$
\begin{equation*}
(4.7) \leqslant \mathbb{P}\left(\max _{k \in \llbracket 0,3\left\lfloor N^{2 / 3}\right\rfloor \rrbracket}\left\{S_{k}^{\lambda}-\left[k \Psi_{0}(\mu-\rho)-k \Psi_{0}(\rho)\right]\right\} \geqslant \frac{C^{\prime}}{8} r N^{1 / 3}\right)+\mathbb{P}\left(A^{c}\right), \tag{4.9}
\end{equation*}
$$

where $\mathbb{P}\left(A^{c}\right) \leqslant e^{-C r^{3}}$. Note $\mathbb{E}\left[S_{k}^{\lambda}\right]=k \Psi_{0}(\mu-\lambda)-k \Psi_{0}(\lambda)$, and using Taylor expansion and the fact that $k \leqslant 3 N^{2 / 3}$, we have

$$
\left|\mathbb{E}\left[S_{k}^{\lambda}\right]-\left[k \Psi_{0}(\mu-\rho)-k \Psi_{0}(\rho)\right]\right| \leqslant C \sqrt{r} N^{1 / 3} .
$$

Finally, taking $C^{\prime} \geqslant 16 C$, the probability in (4.9) is bounded as follows

$$
\begin{aligned}
& \mathbb{P}\left(\max _{k \in \llbracket 0,3\left\lfloor N^{2 / 3}\right\rfloor \rrbracket}\left\{S_{k}^{\lambda}-\left[k \Psi_{0}(\mu-\rho)-k \Psi_{0}(\rho)\right]\right\} \geqslant \frac{C^{\prime}}{8} r N^{1 / 3}\right) \\
& \quad \leqslant \mathbb{P}\left(\max _{k \in \llbracket 0,3\left\lfloor N^{2 / 3}\right\rfloor \rrbracket}\left\{S_{k}^{\lambda}-\mathbb{E}\left[S_{k}^{\lambda}\right]\right\} \geqslant \frac{C^{\prime}}{16} r N^{1 / 3}\right) \leqslant e^{-C^{\prime \prime} r^{3 / 2}}
\end{aligned}
$$

where the last inequality follows from Theorem A.11.

With this result, we are ready to prove Lemma 4.5. The proof uses arguments for a stationary polymer with an antidiagonal boundary instead of a southwest boundary, which we will now define. Let $S_{(0,0)}$ be the bi-infinite staircase paths (with alternating $e_{1}$ and $-e_{2}$ steps) through $(0,0)$

$$
\begin{equation*}
S_{(0,0)}=\{\ldots,(-1,1),(-1,0),(0,0),(0,-1),(1,-1), \ldots\} . \tag{4.10}
\end{equation*}
$$

Next, we attach boundary weights along $S_{(0,0)}$, which are all independent. For each horizontal edge to the left and right of $(0,0)$, we attach $\mathrm{Ga}(\mu-\rho)$ and $\mathrm{Ga}^{-1}(\mu-\rho)$ weights. For each vertical edge to the left and right of $(0,0)$, we attach $\mathrm{Ga}^{-1}(\rho)$ and $\mathrm{Ga}(\rho)$ weights. For $k \in \mathbb{Z}$, let $H_{k}$ denote the product of the edge weights from $S_{(0,0)}$ between $(0,0)$ and $(k,-k)$.

The partition function for this polymer with antidiagonal boundary is defined by

$$
Z_{0, x}^{\rho, \text { dia }}=\sum_{k \in \mathbb{Z}} H_{k} \cdot \widetilde{Z}_{(k,-k), x},
$$

where $\widetilde{Z}$ is the point-to-point partition but without using the weight at its starting point. The corresponding polymer measure $Q_{0, x}^{\rho \text { dia }}$ is a probability measure on paths that start at 0 , move along the antidiagonal, taking either only $e_{1}-e_{2}$ steps or only $e_{2}-e_{1}$ steps, and then enter the bulk by taking an $e_{1}$ or $e_{2}$ step, after which they only take steps in $\left\{e_{i}, i=1,2\right\}$. For such a path $\gamma$, we define $\tau^{\mathrm{dia}}(\gamma) \in \mathbb{Z} \backslash\{0\}$ as the signed number of steps taken before entering the bulk, where the plus sign corresponds to $e_{1}-e_{2}$ steps and the minus sign to $e_{2}-e_{1}$ steps. For $k \in \mathbb{Z}$, let us define the partition function over paths with exit point $k$ as

$$
\begin{equation*}
Z_{0, x}^{\rho, \text { dia }}\left(\tau^{\mathrm{dia}}=k\right)=H_{k} \cdot \widetilde{Z}_{(k,-k), x} . \tag{4.11}
\end{equation*}
$$

Proof of Lemma 4.5. First, by Lemma A.9, it suffices to prove our estimate for the stationary polymer with the antidiagonal boundary defined above. By a slight abuse of notation, let us denote $Z^{\rho}=Z^{\rho, \text { dia }}$, and $Q^{\rho}=Q^{\rho, \text { dia }}$. There is no confusion because we will only be working with the antidiagonal boundary in the remainder of this proof (instead of southwest boundary).

By a union bound, it suffices to prove that there exist positive constants $C_{1}, C_{2}, s_{0}, c_{0}$ such that for each $N \geqslant N_{0}$ and $s_{0} \leqslant s \leqslant c_{0} N^{1 / 3}$, we have

$$
\mathbb{P}\left(\max _{k N^{-2 / 3} \in(s, s+1]} Q_{0, v_{N}}^{\rho}\left\{\tau^{\mathrm{dia}}=k\right\} \geqslant e^{-C_{1} s^{2} N^{1 / 3}}\right) \leqslant e^{-C_{2} s^{3}} .
$$

To show this, we rewrite the quenched probability above in terms of the free energies,

$$
\begin{align*}
& \mathbb{P}\left(\log Z_{0, v_{N}}^{\rho}-\max _{k N^{-2 / 3} \in(s, s+1]} \log Z_{0, v_{N}}^{\rho}\left\{\tau^{\mathrm{dia}}=k\right\} \leqslant C^{\prime} s^{2} N^{1 / 3}\right) \\
& \leqslant \mathbb{P}\left(\left[\log Z_{0, v_{N}}^{\rho}-\Lambda\left(v_{N}\right)\right]\right.  \tag{4.1.1}\\
& \quad-\max _{k N^{-2 / 3} \in(s, s+1]}\left[\log Z_{0, v_{N}}^{\rho}\left\{\tau^{\mathrm{dia}}=k\right\}-\left(\Lambda\left(v_{N}+(-k, k)\right)-k \Psi_{0}(\mu-\rho)+k \Psi_{0}(\rho)\right)\right] \\
& \left.\quad \leqslant C^{\prime} s^{2} N^{1 / 3}+\max _{k N^{-2 / 3} \in(s, s+1]}\left(\Lambda\left(v_{N}+(-k, k)\right)-k \Psi_{0}(\mu-\rho)+k \Psi_{0}(\rho)-\Lambda\left(v_{N}\right)\right)\right) .
\end{align*}
$$

Applying Proposition 4.7, if we fix $C^{\prime}$ in (4.12) sufficiently small, then, we may replace the right side of the inequality in (4.12) by $-c^{\prime} s^{2} N^{1 / 3}$ for some small positive constant $c^{\prime}$.

Let $\left\{Z_{i}\right\}_{i=1}^{\infty}$ denote a sequence of independent and identically distributed random variables with the same distribution given by $-\log G_{1}+\log G_{2}$, where $G_{1}$ and $G_{2}$ are independent, respectively, $\mathrm{Ga}(\mu-\rho)$ and $\mathrm{Ga}(\rho)$ random variables. The $Z_{i}$ 's will play the role of the boundary weight at $(i,-i)$, $i \geqslant 1$. Now continuing with a union bond, we have

$$
\begin{gather*}
(4.12) \leqslant \mathbb{P}\left(\log Z_{0, v_{N}}^{\rho}-\Lambda\left(v_{N}\right) \leqslant-\frac{1}{5} c^{\prime} s^{2} N^{1 / 3}\right)  \tag{4.13}\\
+\mathbb{P}\left(\max _{k N^{-2 / 3} \in(s, s+1]}\left(\sum_{i=1}^{k} Z_{i}+k \Psi_{0}(\mu-\rho)-k \Psi_{0}(\rho)\right) \geqslant \frac{1}{5} c^{\prime} s^{2} N^{1 / 3}\right)  \tag{4.14}\\
+\mathbb{P}\left(\max _{k N^{-2 / 3} \in(s, s+1]}\left(\log \widetilde{Z}_{(-k, k), v_{N}}-\Lambda\left(v_{N}+(-k, k)\right) \geqslant \frac{1}{5} c^{\prime} s^{2} N^{1 / 3}\right),\right. \tag{4.15}
\end{gather*}
$$

and (4.13) $\leqslant e^{-C s^{3}}$ by Proposition A.2, $(4.14) \leqslant e^{-C s^{3}}$ by Proposition A. 12 and Theorem A.11, $(4.15) \leqslant e^{-C s^{3}}$ by Proposition 4.11. Finally, we note that even the $\widetilde{Z}$ free energy does not use the first weight, but Proposition 4.11 (which was originally stated for $Z$ instead of $\widetilde{Z}$ ) still applies because using a union bound we can get

$$
\mathbb{P}\left(\max _{0 \leqslant k \leqslant 3 N^{2 / 3}} \log Y_{(-k, k)} \geqslant \varepsilon s^{2} N^{1 / 3}\right) \leqslant N^{2 / 3} e^{-c s^{2} N^{1 / 3}} \leqslant C e^{-s^{2} N^{1 / 3}} \leqslant C^{\prime} e^{-c^{\prime} s^{3}}
$$

Also, we are applying Proposition 4.11 by first shifting the picture to move the $v_{N}$ in (4.7) to the origin, flipping it about the antidiagonal, and then using, in the proposition, $a v_{N}$ that is not exactly the $v_{N}$ in the lemma, but rather $v_{N}-s N^{2 / 3}(1,-1)$. This is allowed because the proposition is stated uniformly for a whole interval of characteristic directions.

### 4.1.2 | Proof of Lemma 4.6

To prove Lemma 4.6, we tilt the probability measure to make the event likely and pay for this with a bound on the Radon-Nikodym derivative. This argument was introduced in [3] in the context of the asymmetric simple exclusion process and later adapted to lower bound proofs of the longitudinal fluctuation exponent [33] and large exit time probability [34] in the stationary lastpassage percolation process. The key idea here is to perturb the parameter $\rho$ of the stationary
polymer model to $\rho \pm r N^{-1 / 3}$. This allows us to control the exit point on the scale $N^{2 / 3}$. The general idea of utilizing perturbations of order $N^{-1 / 3}$ goes back to the seminal paper [10]. We now give the details.

For $v \in \mathbb{Z}_{\geqslant 0}^{2}$ let $\partial^{\mathrm{NE}} \llbracket 0, v \rrbracket$ denote the north-east boundary of the rectangle $\llbracket 0, v \rrbracket$, that is, the sites $u \in \llbracket 0, v \rrbracket$ with $u \cdot e_{1}=v \cdot e_{1}$ or $u \cdot e_{2}=v \cdot e_{2}$.

Note that it is enough to prove the claimed bound with $\min _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>r N^{1 / 3}\right\}$ replaced by

$$
\min _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\},
$$

as

$$
\begin{aligned}
& \min _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \\
& \quad=\min _{x \notin \llbracket 0, v_{N} \rrbracket} \sum_{z \in \partial^{\mathrm{NE}} \llbracket\left[0, v_{N} \rrbracket\right.} Q_{0, x}^{\rho}\left\{|\tau|>r N^{2 / 3}\right. \text { and the path passes through z\}} \\
& \quad=\min _{x \notin \llbracket 0, v_{N} \rrbracket} \sum_{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, z}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} Q_{0, x}^{\rho}\{\text { path passes through } z\} \\
& \\
& \geqslant \min _{x \notin \llbracket 0, v_{N} \rrbracket} \sum_{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left(\min _{z^{\prime} \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, z^{\prime}}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\}\right) Q_{0, x}^{\rho}\{\text { passes through } z\} \\
& \quad=\min _{z^{\prime} \in \partial^{\mathrm{NE} \llbracket 0, v_{N} \rrbracket}} Q_{0, z^{\prime}}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} .
\end{aligned}
$$

Take $c \in\left(0, \frac{\varepsilon}{4 \mu^{2}} \wedge \frac{1}{2}\right.$ ], with $\varepsilon$ as in the statement of the lemma. Below, we will choose an exact value for $c$, which will still only depend on $\varepsilon$ (and $\mu$ ).

Given positive $r$ and $N$, define the perturbed parameters $\lambda=\rho+r N^{-1 / 3}$ and $\eta=\rho-r N^{-1 / 3}$. The choice of $c$ guarantees that if

$$
\begin{equation*}
r \leqslant c\left((\mu-\rho)^{2} \wedge \rho^{2}\right) N^{1 / 3} \tag{4.16}
\end{equation*}
$$

then $\eta<\rho<\lambda$ are all contained in $[\varepsilon / 2, \mu-\varepsilon / 2]$.
Given positive constants $a<b$, define a new environment $\widetilde{\mathbb{P}}$ by changing the original boundary weights (whose distribution we will denote by $\mathbb{P}^{\rho}$ ) on parts of the axes. Precisely, $\widetilde{\mathbb{P}}$ is the joint distribution, under $\mathbb{P}^{\rho}$ of

$$
\begin{aligned}
\tilde{\omega}_{k e_{1}} & \sim \mathrm{Ga}^{-1}(\mu-\lambda) & & \text { for } k \in\left[\left[\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor+1,\left\lfloor b r N^{2 / 3}\right\rfloor\right]\right] \\
\tilde{\omega}_{k e_{2}} & \sim \mathrm{Ga}^{-1}(\eta) & & \text { for } \in\left[\left[\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor+1,\left\lfloor b r N^{2 / 3}\right\rfloor\right]\right] \\
\tilde{\omega}_{z} & \sim \omega_{z} & & \text { for all other } z \in \mathbb{Z}_{\geqslant 0}^{2} .
\end{aligned}
$$

The $\widetilde{\omega}$ weights in the first two lines are all independent and independent of the $\omega$ weights. The exact values of $a$ and $b$ will be determined further down and will only depend on $\varepsilon>0$ (and c). Essentially, they will be chosen so that, in the picture in the left panel of Figure 4, the two thick dotted lines passing through $v_{N}$ and having slopes $\xi[\lambda]$ and $\xi[\eta]$ rest inside the highlighted regions on the axes. Then, under the new random environment $\widetilde{\mathbb{P}}$, we will show that there exists


FIGURE 4 Left: Two dotted lines have slopes $\xi[\lambda]$ and $\xi[\eta]$. Right: Decomposition of the north and east boundaries of $\llbracket 0, v_{N} \rrbracket$ into regions $\mathcal{L}$ (light gray) and $\mathcal{D}$ (dark gray). A small perturbation of $v_{N}$ to $w_{N}$ keeps the endpoint of the $-\xi[\lambda]$ ray from $w_{N}$ in the interval $\left[\operatorname{arN}^{2 / 3}, b r N^{2 / 3}\right]$.
some constant $C_{1}$ such that for $N$ and $r$ large,

$$
\begin{equation*}
\widetilde{\mathbb{P}}\left(\min _{x \in \partial \mathrm{NE} \llbracket 0, v_{N} \rrbracket} Q_{0, x}\left\{|\tau|>\operatorname{ar} N^{2 / 3}\right\} \geqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right) \geqslant 1 / 2 . \tag{4.17}
\end{equation*}
$$

We finish the proof of the theorem, assuming this inequality. Denote the event inside (4.17) by $S$ and let $f=\frac{d \widetilde{\mathbb{P}}}{d \mathbb{P}^{\rho}}$, where $\mathbb{P}^{\rho}$ is the marginal of $\mathbb{P}$, that is, the probability measure with independent $\rho$-boundary weights and bulk weights. By the Cauchy-Schwarz inequality, we have

$$
1 / 2 \leqslant \widetilde{\mathbb{P}}(S)=\mathbb{E}^{\rho}\left[1_{S} f\right] \leqslant \mathbb{P}^{\rho}(S)^{1 / 2} \mathbb{E}^{\rho}\left[f^{2}\right]^{1 / 2} \leqslant \mathbb{P}^{\rho}(S)^{1 / 2} e^{C r^{3}},
$$

where the last bound for the second moment of $f$ follows from Proposition A.10. This implies

$$
\begin{equation*}
\mathbb{P}\left(\min _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>\operatorname{ar} N^{2 / 3}\right\} \geqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right) \geqslant e^{-C_{2} r^{3}} . \tag{4.18}
\end{equation*}
$$

To recover the statement of our lemma without the constant $a$ in (4.18), just modify $C_{1}$ and $C_{2}$.
Next, we will show (4.17) that will finish the proof of the theorem. To do this, we will show that for $r$ and $N$ large, we have

$$
\begin{equation*}
\widetilde{\mathbb{P}}\left(\max _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\}<e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \geqslant 1-C r^{-3} . \tag{4.19}
\end{equation*}
$$

This and the similar estimate for the event $\left\{-1 \geqslant \tau_{0, x} \geqslant-\operatorname{ar} N^{2 / 3}\right\}$ imply (4.17) when $r$ is taken large. Note that here we will pick the values of $a<b$ and $c \in\left(0, \frac{\varepsilon}{4 \mu^{2}} \wedge \frac{1}{2}\right]$ only for the bound (4.19). When applying the same argument to the other case, we obtain another set of constants $a^{\prime}<b^{\prime}$ and $c^{\prime} \in\left(0, \frac{\varepsilon}{4 \mu^{2}} \wedge \frac{1}{2}\right]$ that are possibly different. Then, we replace $a$ and $a^{\prime}$ by $a \wedge a^{\prime}, b$ and $b^{\prime}$ by $b \vee b^{\prime}$, and $c$ and $c^{\prime}$ by $c \wedge c^{\prime}$.

Recall the perturbed parameter $\lambda=\rho+r N^{-1 / 3}$. If $c \in\left(0, \frac{\varepsilon}{4 \mu^{2}} \wedge \frac{1}{2}\right]$ and $r$ and $N$ satisfy condition (4.16), then $\lambda$ satisfies

$$
\begin{equation*}
\varepsilon / 2<\rho<\lambda \leqslant \rho+c\left((\mu-\rho)^{2} \wedge \rho^{2}\right) \leqslant \mu-\varepsilon / 2 . \tag{4.20}
\end{equation*}
$$



FIGURE 5 Left: The dotted lines have characteristic slope $\xi[\lambda]$. Consequently, with high probability, the sampled $\lambda$ polymer from 0 to $w_{N}$ exits through the interval $\llbracket a r N^{2 / 3} e_{1}, b r N^{2 / 3} e_{1} \rrbracket$. Right: Illustration of estimate (4.25).

We estimate the difference of the reciprocal slopes (i.e., $\frac{\text { change of } x}{\text { change of } y}$ ) of the vectors $\xi[\lambda]$ and $\xi[\rho]$. By definition

$$
\frac{\xi[\lambda] \cdot e_{1}}{\xi[\lambda] \cdot e_{2}}-\frac{\xi[\rho] \cdot e_{1}}{\xi[\rho] \cdot e_{2}}=\frac{\Psi_{1}\left(\rho+r N^{-1 / 3}\right)}{\Psi_{1}\left(\mu-\rho-r N^{-1 / 3}\right)}-\frac{\Psi_{1}(\rho)}{\Psi_{1}(\mu-\rho)}
$$

As $\Psi_{1}$ is smooth and takes positive values on compact intervals strictly contained inside $(0, \mu)$, we can Taylor expand the quotient $g(z)=\frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\mu-\rho-z)}$ around $z=0$. This gives

$$
\begin{equation*}
\left|\left(\frac{\xi[\lambda] \cdot e_{1}}{\xi[\lambda] \cdot e_{2}}-\frac{\xi[\rho] \cdot e_{1}}{\xi[\rho] \cdot e_{2}}\right)-\left(-k_{1} r N^{-1 / 3}\right)\right| \leqslant k_{2} r^{2} N^{-2 / 3} \tag{4.21}
\end{equation*}
$$

for all $\rho$ and $\lambda$ such that $\varepsilon / 2<\rho<\lambda<\mu-\varepsilon / 2$. Here, $k_{1}$ and $k_{2}$ are positives constant depending only on $\rho, \mu$, and $\varepsilon$. Take $c \in\left(0, \frac{\varepsilon}{4 \mu^{2}} \wedge \frac{1}{2}\right.$ ] to satisfy

$$
\begin{equation*}
c \leqslant \frac{1}{100} \frac{k_{1}}{k_{2}} . \tag{4.22}
\end{equation*}
$$

Then, for $r$ and $N$ satisfying (4.16),

$$
\begin{equation*}
k_{2} r^{2} N^{-2 / 3}<\frac{1}{10} k_{1} r N^{-1 / 3} . \tag{4.23}
\end{equation*}
$$

And from (4.21) and (4.23) above, we obtain

$$
\begin{equation*}
-2 k_{1} r N^{-1 / 3} \leqslant \frac{\xi[\lambda] \cdot e_{1}}{\xi[\lambda] \cdot e_{2}}-\frac{\xi[\rho] \cdot e_{1}}{\xi[\rho] \cdot e_{2}} \leqslant-\frac{1}{2} k_{1} r N^{-1 / 3} . \tag{4.24}
\end{equation*}
$$

Now, start two rays at $(0,0)$ in the directions $\xi[\rho]$ and $\xi[\lambda]$ and let $u_{N}$ be the lattice point closest to the $\xi[\lambda]$-directed ray such that $u_{N} \cdot e_{2}=v_{N} \cdot e_{2}$. (See the right panel of Figure 5.) Then (4.24) implies that there exist two fixed positive constants $l_{1}, l_{2}$ depending only on $\rho, \mu$, and $\varepsilon$ such that

$$
\begin{equation*}
l_{1} r N^{2 / 3} \leqslant v_{N} \cdot e_{1}-u_{N} \cdot e_{1} \leqslant l_{2} r N^{2 / 3} . \tag{4.25}
\end{equation*}
$$

For now, we define

$$
a=\frac{1}{10} l_{1} \quad \text { and } \quad b=10 l_{2},
$$

and note that the above value of $a$ will be lowered if necessary, later in the argument.
Fix a positive constant $q \leqslant \frac{1}{10} l_{1}$, let us define

$$
\begin{equation*}
w_{N}=v_{N}-\left\lfloor q r N^{2 / 3}\right\rfloor e_{1} . \tag{4.26}
\end{equation*}
$$

As shown on the right of Figure 4, the point $w_{N}$ splits $\partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket$ into the dark region $\mathcal{D}$ and the light region $\mathcal{L}$. We will first work with the dark region and show

$$
\begin{equation*}
\widetilde{\mathbb{P}}\left(\max _{x \in D} Q_{0, x}\left\{1 \leqslant \tau \leqslant a r N^{2 / 3}\right\} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \geqslant 1-C r^{-3} . \tag{4.27}
\end{equation*}
$$

Let us look at another polymer measure $R_{0, x}$ that is restricted to paths that start with an $e_{1}$ step from the origin, then

$$
R_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\}=\frac{Z_{0, x}\left(1 \leqslant \tau \leqslant\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor\right)}{Z_{0, x}(1 \leqslant \tau)} .
$$

From the following three facts,

- $Q_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \leqslant R_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\}$,
- $R_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\}+R_{0, x}\left\{\tau>\operatorname{ar} N^{2 / 3}\right\}=1$, and
- by Lemma A.4, $R_{0, w_{N}}\left\{\tau>\operatorname{ar} N^{2 / 3}\right\} \leqslant R_{0, x}\left\{\tau>\operatorname{ar} N^{2 / 3}\right\}$ for each $x \in \mathcal{D}$,
we have

$$
Q_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \leqslant R_{0, w_{N}}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \quad \text { for each } x \in \mathcal{D}
$$

Thus, in order to show (4.27), it suffices to show

$$
\begin{equation*}
\widetilde{\mathbb{P}}\left(R_{0, w_{N}}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \geqslant 1-C r^{-3} . \tag{4.28}
\end{equation*}
$$

To show (4.28), we will find a high probability event

$$
A=A_{1} \cap A_{2} \cap A_{3} \cap A_{4}
$$

with $\widetilde{\mathbb{P}}(A) \geqslant 1-C r^{-3}$ such that on $A$,

$$
\begin{equation*}
Z_{0, w_{N}}\left(\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor b r N^{2 / 3}\right\rfloor\right) \geqslant e^{C^{\prime} r^{2} N^{1 / 3}} Z_{0, w_{N}}\left(1 \leqslant \tau \leqslant\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor\right) \tag{4.29}
\end{equation*}
$$

as this implies

$$
\begin{aligned}
R_{0, w_{N}}\left\{\tau>a r N^{2 / 3}\right\} & \geqslant \frac{Z_{0, w_{N}}\left(\left\lfloor a r N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor b r N^{2 / 3}\right\rfloor\right)}{Z_{0, w_{N}}(1 \leqslant \tau)} \\
& \geqslant e^{C^{\prime} r^{2} N^{1 / 3}} \frac{Z_{0, w_{N}}\left(1 \leqslant \tau \leqslant\left\lfloor a r N^{2 / 3}\right\rfloor\right)}{Z_{0, w_{N}}(1 \leqslant \tau)} \\
& =e^{C^{\prime} r^{2} N^{1 / 3}} R_{0, w_{N}}\left\{1 \leqslant \tau \leqslant a r N^{2 / 3}\right\},
\end{aligned}
$$

which together with

$$
R_{0, w_{N}}\left\{1 \leqslant \tau \leqslant \operatorname{arN} N^{2 / 3}\right\}+R_{0, w_{N}}\left\{\tau>\operatorname{arN} N^{2 / 3}\right\}=1
$$

gives

$$
R_{0, w_{N}}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \leqslant \frac{1}{1+e^{C^{\prime} r^{2} N^{1 / 3}}} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}} \quad \text { on } A .
$$

Next, we define $A_{1}, A_{2}, A_{3}$ and $A_{4}$ and their intersection gives $A$. Let $Z^{\lambda}$ and $Z^{\rho}$ denote the partition functions with the $\lambda$ - and $\rho$-boundary weights, and where all boundary weights are independent. Then, the $e_{1}$-boundary weights from $\widetilde{\mathbb{P}}$ can be seen as a mixture of these $\lambda$ - and $\rho$-weights. The desired inequality (4.29) (under $\widetilde{\mathbb{P}}$ ) can be rewritten as

$$
\begin{aligned}
\left(\frac{Z_{0, w_{N}}^{\lambda}}{Z_{0, w_{N}}^{\rho}} \prod_{i=1}^{\left\lfloor\operatorname{arN} N^{2 / 3}\right\rfloor} \frac{I_{(i, 0)}^{\rho}}{I_{(i, 0)}^{\lambda}}\right) \frac{Z_{0, w_{N}}^{\lambda}\left(\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor b r N^{2 / 3}\right\rfloor\right)}{Z_{0, w_{N}}^{\lambda}} & \\
& \geqslant e^{C^{\prime} r^{2} N^{1 / 3}} \frac{Z_{0, w_{N}}^{\rho}\left(1 \leqslant \tau \leqslant\left\lfloor a r N^{2 / 3}\right\rfloor\right)}{Z_{0, w_{N}}^{\rho}},
\end{aligned}
$$

which is implied by the inequality

$$
\left(\frac{Z_{0, w_{N}}^{\lambda}}{Z_{0, w_{N}}^{\rho}} \prod_{i=1}^{\left\lfloor a r N^{2 / 3}\right\rfloor} \frac{I_{(i, 0)}^{\rho}}{I_{(i, 0)}^{\lambda}}\right) \frac{Z_{0, w_{N}}^{\lambda}\left(\left\lfloor a r N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor b r N^{2 / 3}\right\rfloor\right)}{Z_{0, w_{N}}^{\lambda}} \geqslant e^{C^{\prime} r^{2} N^{1 / 3}}
$$

Because $w_{N}$ is a point of order $r N^{2 / 3}$ units away from $u_{N}$ (recall $u_{N}$ is along the $\xi[\lambda]$ characteristic ray defined above (4.25)), there is an event $A_{1}$ with $\mathbb{P}\left(A_{1}\right) \geqslant 1-e^{-C r^{3}}$ such that the $\lambda$ quenched probability appearing above (i.e., the last ratio of partition functions on the left-hand side) satisfies

$$
\frac{Z_{0, w_{N}}^{\lambda}\left(\left\lfloor a r N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor b r N^{2 / 3}\right\rfloor\right)}{Z_{0, w_{N}}^{\lambda}} \geqslant 1 / 2 \quad \text { on the event } A_{1} .
$$

This is proved as Lemma 4.12 at the end of this section, and the idea is illustrated on the right of Figure 5.

Once on the event $A_{1}$, (4.29) would follow from having

$$
\begin{equation*}
\frac{Z_{0, w_{N}}^{\lambda}}{Z_{0, w_{N}}^{\rho}} \prod_{i=1}^{a r N^{2 / 3}} \frac{I_{(i, 0)}^{\rho}}{I_{(i, 0)}^{\lambda}} \geqslant e^{C^{\prime} r^{2} N^{1 / 3}} \tag{4.30}
\end{equation*}
$$

with possibly a different $C^{\prime}$. This inequality should hold with a high probability if $a>0$ is taken sufficiently small. We will work with the logarithmic version of (4.30)

$$
\log Z_{0, w_{N}}^{\lambda}-\log Z_{0, w_{N}}^{\rho}-\left(\sum_{i=1}^{a r N^{2 / 3}} \log \left(I_{(i, 0)}^{\lambda}\right)-\log \left(I_{(i, 0)}^{\rho}\right)\right) .
$$

We start by showing that

$$
\begin{equation*}
\mathbb{E}\left[\log Z_{0, w_{N}}^{\lambda}\right]-\mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right] \geqslant c_{1} r^{2} N^{1 / 3} \tag{4.31}
\end{equation*}
$$

for some $\varepsilon$-dependent constant $c_{1}$, and this constant $c_{1}$ will be used for the rest of the proof. First, note the exact values of the expectations are

$$
\begin{aligned}
& \mathbb{E}\left[\log Z_{0, w_{N}}^{\lambda}\right]=\Psi_{0}(\mu-\lambda)\left(\left\lfloor\Psi_{1}(\rho) N\right\rfloor-\left\lfloor q r N^{2 / 3}\right\rfloor\right)+\Psi_{0}(\lambda)\left\lfloor\Psi_{1}(\mu-\rho) N\right\rfloor \\
& \mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right]=\Psi_{0}(\mu-\rho)\left(\left\lfloor\Psi_{1}(\rho) N\right\rfloor-\left\lfloor q r N^{2 / 3}\right\rfloor\right)+\Psi_{0}(\rho)\left\lfloor\Psi_{1}(\mu-\rho) N\right\rfloor
\end{aligned}
$$

Using a Taylor expansion,

$$
\begin{aligned}
\Psi_{0}(\mu-\lambda) & =\Psi_{0}(\mu-\rho)+\Psi_{1}(\mu-\rho)\left(-r N^{-1 / 3}\right)+\frac{1}{2} \Psi_{1}^{\prime}(\mu-\rho)\left(-r N^{-1 / 3}\right)^{2}+R_{1} \\
\Psi_{0}(\lambda) & =\Psi_{0}(\rho)+\Psi_{1}(\rho)\left(r N^{-1 / 3}\right)+\frac{1}{2} \Psi_{1}^{\prime}(\rho)\left(r N^{-1 / 3}\right)^{2}+R_{2}
\end{aligned}
$$

Due to condition (4.20), we have $\left|R_{i}\right| \leqslant C\left(r N^{-1 / 3}\right)^{3}$ for both $i \in\{1,2\}$ and with an $\varepsilon$-dependent constant $C>0$. Plugging these two formulae back into the right side of (4.31), the linear terms from the expansions cancel out. By further lowering the value of $c$ from (4.16) if necessary, $R_{1}$ and $R_{2}$ can be absorbed into the $\left(r N^{-1 / 3}\right)^{2}$ terms, and there exist two positive constants $D_{1}$ and $D_{2}$ depending only on $\varepsilon, \mu$ and $c$ such that

$$
\mathbb{E}\left[\log Z_{0, w_{N}}^{\lambda}\right]-\mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right] \geqslant D_{1} r^{2} N^{1 / 3}-D_{2} q r^{2} N^{1 / 3},
$$

where the parameter $q$ is from (4.26). By fixing $q$ sufficiently small, we obtain the desired estimate (4.31).

Next, with the constant $c_{1}$ from (4.31), we define the two events

$$
\begin{aligned}
& A_{2}=\left\{\log Z_{0, w_{N}}^{\lambda} \geqslant \mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right]+\frac{c_{1}}{2} r^{2} N^{1 / 3}\right\}, \\
& A_{3}=\left\{\log Z_{0, w_{N}}^{\rho} \leqslant \mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right]+\frac{c_{1}}{10} r^{2} N^{1 / 3}\right\},
\end{aligned}
$$

and we will show $\mathbb{P}\left(A_{2}\right) \wedge \mathbb{P}\left(A_{3}\right) \geqslant 1-\mathrm{Cr}^{-3}$.
First, we work with $\mathbb{P}\left(A_{2}\right)$. For $\theta, x>0$, let us define $L(\theta, x)$ as in [32, (3.17)],

$$
L(\theta, x)=\int_{0}^{x}\left(\Psi_{0}(\theta)-\log y\right) x^{-\theta} y^{\theta-1} e^{x-y} d y .
$$

In the next calculation, the first equality is the statement in [32, Theorem 3.7],

$$
\begin{align*}
\operatorname{Var}\left[\log Z_{0, w_{N}}^{\rho}\right] & =w_{N} \cdot e_{2} \Psi_{1}(\rho)-w_{N} \cdot e_{1} \Psi_{1}(\mu-\rho)+2 \mathbb{E}\left[E^{Q_{0, w_{N}}^{\rho}}\left[\sum_{i=1}^{0 \vee \tau} L\left(\mu-\rho, I_{i e_{1}}^{\rho}\right)\right]\right] \\
& \leqslant C\left(r N^{2 / 3}+\mathbb{E}\left[E^{Q_{0, w_{N}}^{\rho}}\left[\tau \mathbb{1}_{\{\tau \geqslant 1\}}\right]\right]+1\right) \quad \text { (by Lemma 4.2 of [32]) } \\
& \leqslant C\left(r N^{2 / 3}+\mathbb{E}\left[E^{Q_{0, v_{N}}^{\rho}\left[\tau 1_{\{\tau \geqslant 1\}}\right]}\right]+1\right) \quad \text { (by Lemma A.5) } \\
& \leqslant C r N^{2 / 3}+C^{\prime} N^{2 / 3} \quad \quad \text { (by (4.32) of [32]). }
\end{align*}
$$

Now, we upper bound the compliment

$$
\begin{aligned}
\mathbb{P}\left(A_{2}^{c}\right) & =\mathbb{P}\left\{\log Z_{0, w_{N}}^{\lambda}<\mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right]+\frac{c_{1}}{2} r^{2} N^{1 / 3}\right\} \\
& \leqslant \mathbb{P}\left\{\log Z_{0, w_{N}}^{\lambda}<\mathbb{E}\left[\log Z_{0, w_{N}}^{\lambda}\right]-\frac{c_{1}}{2} r^{2} N^{1 / 3}\right\} \quad(\text { by }(4.31)) \\
& \leqslant \frac{4}{c_{1}^{2} r^{4} N^{2 / 3}} \operatorname{Var}\left[\log Z_{0, w_{N}}^{\lambda}\right] \\
& \leqslant \frac{4}{c_{1}^{2} r^{4} N^{2 / 3}}\left(\mathbb{V a r}\left[\log Z_{0, w_{N}}^{\rho}\right]+c_{3} r N^{2 / 3}\right) \quad \text { (by Lemma 4.1 of [32]) } \\
& \leqslant C r^{-3}
\end{aligned}
$$

The fact $\mathbb{P}\left(A_{3}\right) \geqslant 1-\mathrm{Cr}^{-3}$ comes from the Markov inequality

$$
\mathbb{P}\left(A_{3}^{c}\right)=\mathbb{P}\left\{\log Z_{0, w_{N}}^{\rho}>\mathbb{E}\left[\log Z_{0, w_{N}}^{\rho}\right]+\frac{c_{1}}{10} r^{2} N^{1 / 3}\right\} \leqslant \frac{100}{c_{1}^{2} r^{4} N^{2 / 3}} \operatorname{Var}\left[\log Z_{0, w_{N}}^{\rho}\right] \leqslant C r^{-3}
$$

Next, we define another high probability event $A_{4}$ by

$$
A_{4}=\left\{\sum_{i=1}^{a r N^{2 / 3}}\left(\log I_{(i, 0)}^{\lambda}-\log I_{(i, 0)}^{\rho}\right) \leqslant \frac{c_{1}}{10} r^{2} N^{1 / 3}\right\}
$$

If $a$ is chosen sufficiently small compared to $c_{1}$, then by Proposition A. 12 and Theorem A.11,

$$
\mathbb{P}\left(A_{4}\right) \geqslant 1-e^{-C r^{3}} .
$$

Finally, on the event

$$
A_{1} \cap A_{2} \cap A_{3} \cap A_{4}
$$

our desired estimate (4.30) (after taking logarithm) will hold

$$
\log Z_{0, w_{N}}^{\lambda}-\log Z_{0, w_{N}}^{\rho}-\left(\sum_{i=1}^{\operatorname{ar} N^{2 / 3}} \log I_{(i, 0)}^{\lambda}-\log I_{(i, 0)}^{\rho}\right) \geqslant \frac{c_{1}}{10} r^{2} N^{1 / 3} \geqslant C^{\prime} r^{2} N^{1 / 3}
$$

This finishes the argument for the dark region and (4.27).
For the light region,

$$
\begin{aligned}
& \widetilde{\mathbb{P}}\left(\max _{x \in \mathcal{L}} Q_{0, x}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \\
& \geqslant \mathbb{P}\left(\max _{x \in \mathcal{L}} Q_{0, x}^{\rho}\left\{1 \leqslant \tau \leqslant \operatorname{ar} N^{2 / 3}\right\} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \\
& \geqslant \mathbb{P}\left(\max _{x \in \mathcal{L}} Q_{0, x}^{\rho}\{1 \leqslant \tau\} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}}\right)
\end{aligned}
$$

$$
\begin{aligned}
& =\mathbb{P}\left(Q_{0, w_{N}}^{\rho}\{1 \leqslant \tau\} \leqslant e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \quad \text { (by Lemma A.5) } \\
& =1-\mathbb{P}\left(Q_{0, w_{N}}^{\rho}\{1 \leqslant \tau\}>e^{-C^{\prime} r^{2} N^{1 / 3}}\right) \\
& \geqslant 1-e^{-C r^{3}} \quad \text { (by Corollary 4.10). }
\end{aligned}
$$

The proof of Lemma 4.6 is complete.
Lemma 4.12 is an auxiliary estimate for the proof of Lemma 4.6. Recall that $\lambda=\rho+r N^{-1 / 3}$ and satisfies the condition (4.20). As shown on the right of Figure $4, u_{N}$ and $v_{N}$ on the north boundary satisfies (4.25). Using the parameters $l_{1}$ and $l_{2}$ in (4.25), we fix

$$
\begin{equation*}
a \leqslant \frac{1}{10} l_{1}, \quad b \geqslant 10 l_{2}, \quad q \leqslant \frac{1}{10} l_{1} . \tag{4.33}
\end{equation*}
$$

Recall $w_{N}=v_{N}-q r N^{2 / 3} e_{1}$ is a point on the north boundary of $\llbracket 0, v_{N} \rrbracket$. Lemma 4.12 shows that for small enough $a>0$ and large enough $b>0$, the sampled polymer path between the origin and $w_{N}$ exits the $e_{1}$-axis through the interval $\llbracket a r N^{2 / 3} e_{1}, b r N^{2 / 3} e_{1} \rrbracket$ with high probability under $\mathbb{P}^{\lambda}$. This is illustrated on the left of Figure 5.

Lemma 4.12. Let $\varepsilon \in(0, \mu / 2)$, and fix $\rho \in[\varepsilon, \mu-\varepsilon]$ and constants $a, b, q$ as in (4.33). There exist positive constants $C_{1}, C_{2}, C_{3}, r_{0}$, and $N_{0}$ that depend only on $\varepsilon$ such that, for any $r>r_{0}, N \geqslant N_{0}$ with $\lambda=\rho+r N^{-1 / 3}$ satisfying (4.20), we have

$$
\mathbb{P}\left(Q_{0, w_{N}}^{\lambda}\left\{a r N^{2 / 3} \leqslant \tau \leqslant b r N^{2 / 3}\right\} \leqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right) \leqslant e^{-C_{2} r^{3}}
$$

and

$$
\mathbb{E}\left[Q_{0, w_{N}}^{\lambda}\left\{a r N^{2 / 3} \leqslant \tau \leqslant b r N^{2 / 3}\right\}\right] \geqslant 1-e^{-C_{3} r^{3}} .
$$

Proof. First, note we have the following horizontal distance bound between $w_{N}$ and $u_{N}$, where $u_{N}$ is defined previously above (4.25)

$$
\frac{1}{2} l_{1} r N^{2 / 3} \leqslant w_{N} \cdot e_{1}-u_{N} \cdot e_{1} \leqslant l_{2} r N^{2 / 3}
$$

Let $z_{N}$ be the integer point closest to where the $-\xi[\lambda]$-directed ray starting at $w_{N}$ crosses the $e_{1-}$ axis (illustrated as the white dot in Figure 5), then the distance between the origin and $z_{N}$ satisfies the same bound

$$
\begin{equation*}
\frac{1}{2} l_{1} r N^{2 / 3} \leqslant z_{N} \cdot e_{1} \leqslant l_{2} r N^{2 / 3} . \tag{4.34}
\end{equation*}
$$

In the next part, we will show that sampled polymer path between the origin and $w_{N}$ will exist on the $e_{1}$-axis near $z_{N}$. More precisely, we show for $r>r_{0}$ and $N \geqslant N_{0}$ such that (4.20) holds, then

$$
\begin{align*}
& \mathbb{P}\left(Q_{0, w_{N}}^{\lambda}\left\{\tau<\operatorname{ar} N^{2 / 3}\right) \geqslant e^{-C r^{2} N^{1 / 3}}\right) \leqslant e^{-C^{\prime} r^{3}},  \tag{4.35}\\
& \mathbb{P}\left(Q_{0, w_{N}}^{\lambda}\left\{\tau>b r N^{2 / 3}\right) \geqslant e^{-C r^{2} N^{1 / 3}}\right) \leqslant e^{-C^{\prime} r^{3}} . \tag{4.36}
\end{align*}
$$

First, we show (4.36). In the estimate below, the first inequality follows from Lemma A.5; the next equality comes from moving the base from the origin to $z_{N}$ as a nested polymer; the final


FIGURE 6 The vertex $z_{N}$ is shown as the white dot. Applying Lemma A. 8 in the proof of Lemma 4.12 to assert that $Q_{0, w_{N}}^{\lambda}\left\{\tau \leqslant \operatorname{ar} N^{2 / 3}\right\}=Q_{\left(\left\lfloor a r N^{2 / 3}\right\rfloor,-h\right), w_{N}}^{\lambda}\{\tau<-h\}$, which is small.
inequality comes from applying Lemma 4.6 to the nested polymer where the starting and end points are in the $\xi[\lambda]$ direction,

$$
\begin{aligned}
& \mathbb{P}\left(Q_{0, w_{N}}^{\lambda}\left\{\tau>b r N^{2 / 3}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \\
& \leqslant \mathbb{P}\left(Q_{0, v_{N}}^{\lambda}\left\{\tau>b r N^{2 / 3}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \\
& =\mathbb{P}\left(Q_{z_{N}, v_{N}}^{\lambda}\left\{\tau>b r N^{2 / 3}-z_{N} \cdot e_{1}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \\
& =\mathbb{P}\left(Q_{z_{N}, v_{N}}^{\lambda}\left\{\tau>\frac{b}{2} r N^{2 / 3}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \\
& \leqslant e^{-C_{2} r^{3}}
\end{aligned}
$$

This proves (4.36).
To prove (4.35) choose $h$ so that ( $\left\lfloor\operatorname{ar} N^{2 / 3}\right\rfloor,-h$ ) is the closest integer point to the $(-\xi[\lambda])$ directed ray starting at $w_{N}$ (see the right of Figure 6). Lemma A. 8 gives

$$
\begin{aligned}
\mathbb{P}\left(Q_{0, w_{N}}^{\lambda}\{\tau\right. & \left.\left.\leqslant \operatorname{ar} N^{2 / 3}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \\
& =\mathbb{P}\left(Q_{\left.\left\lfloor a r N^{2 / 3}\right\rfloor,-h\right), w_{N}}^{\lambda}\{\tau<-h\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) .
\end{aligned}
$$

Lemma 4.6 states that it is unlikely for the sampled polymer paths from $Q_{\left(\left\lfloor\operatorname{arN} N^{2 / 3}\right\rfloor,-h\right), w_{N}}$ to exit late in the scale $N^{2 / 3}$ from the $y$-axis because the direction is the characteristic one $\xi[\lambda]$. Thus, it suffices to show $h$ is bounded below by some $k(\rho) r N^{2 / 3}$.

Using the lower bound from (4.34), the distance between $z_{N}$ and $\left\lfloor a r N^{2 / 3}\right\rfloor e_{1}$ is bounded below by $4 \operatorname{ar} N^{2 / 3}$. The slope of the line going through $w_{N}$ and $z_{N}$ is roughly $\xi[\lambda]$, because recall $z_{N}$ is defined to be the closes integer point to the crossing point between the $-\xi[\lambda]$-directed ray from $w_{N}$ and the $e_{1}$-axis. Thus, its slope is contained inside a compact interval strictly inside $(0, \mu)$.


FIGURE 7 The north and east boundaries of $\llbracket 0, v_{N} \|$ are decomposed into $\mathcal{L}^{ \pm}$(light gray) and $\mathcal{D}$ (dark gray). The parameter $q$ is less than some small constant that depends only on $\rho$.

Thus, we have

$$
\begin{equation*}
h \geqslant k(\rho) r N^{2 / 3} \tag{4.37}
\end{equation*}
$$

which finishes the proof.

## 4.2 | Proof of Theorem 4.2

First, note that instead of

$$
\max _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\},
$$

it suffices to work with

$$
\max _{x \in \partial^{\mathrm{NE}} \mathbb{0} 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}
$$

as

$$
\begin{align*}
& \max _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\} \\
& \quad=\max _{x \notin \llbracket 0, v_{N} \rrbracket \rrbracket} \sum_{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right. \text { and passes through z\}} \\
&  \tag{4.38}\\
& =\max _{x \notin \llbracket 0, v_{N} \rrbracket} \sum_{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, z}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\} Q_{0, x}^{\rho}\{\text { passes through } z\} \\
& \\
& \leqslant \max _{\left.x \notin \llbracket 0, v_{N} \rrbracket \rrbracket_{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} \sum_{z^{\prime} \in \partial^{\mathrm{NE}} \llbracket\left[0, v_{N} \rrbracket\right.} Q_{0, z^{\prime}}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}\right) Q_{0, x}^{\rho}\{\text { passes through } z\}} \quad=\max _{z^{\prime} \in \partial^{\mathrm{NE} \llbracket 0, v_{N} \rrbracket}} Q_{0, z^{\prime}}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\} .
\end{align*}
$$

Decompose the northeast boundary $\partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket$ into three parts $\mathcal{D}$ and $\mathcal{L}^{ \pm}$as in Figure 7, with

$$
w_{N}^{+}=v_{N}-\left\lfloor q r N^{2 / 3}\right\rfloor e_{1} \quad \text { and } \quad w_{N}^{-}=v_{N}-\left\lfloor q r N^{2 / 3}\right\rfloor e_{2}
$$

where $q \leqslant 1$ is a small positive constant to be chosen later above (4.42), and

$$
r=|\log \delta| .
$$

The dark gray set $\mathcal{D}$ comprises the vertices between $w_{N}^{+}$and $w_{N}^{-}$on the northeast corner of the boundary of the rectangle $\llbracket 0, v_{N} \rrbracket$. Recall that we assume in the theorem that

$$
\begin{equation*}
N>\delta^{-3 / 2} \tag{4.39}
\end{equation*}
$$

This is natural because otherwise the probability in the statement of the theorem would be zero. Introduce the perturbed parameters

$$
\begin{equation*}
\lambda=\rho+r N^{-1 / 3} \quad \text { and } \quad \eta=\rho-r N^{-1 / 3} . \tag{4.40}
\end{equation*}
$$

We require the following bounds to hold for these two parameters

$$
\begin{equation*}
\rho<\lambda \leqslant \rho+\frac{\rho \wedge(\mu-\rho)}{2}<\mu \quad \text { and } \quad 0<\rho-\frac{\rho \wedge(\mu-\rho)}{2} \leqslant \eta<\rho \tag{4.41}
\end{equation*}
$$

The point of the choice $\rho \pm \frac{\rho \wedge(\mu-\rho)}{2}$ is only to bound $\lambda$ and $\eta$ from above and below by two constants strictly inside $(0, \mu)$ and only depending on $\varepsilon$. The above two requirements can be rewritten as

$$
N \geqslant\left(\frac{2 r}{\rho \wedge(\mu-\rho)}\right)^{3}
$$

With (4.39), this bound on $N$ is automatically satisfied as soon as $\delta^{-3 / 2} \geqslant\left(\frac{2 r}{\rho \wedge(\mu-\rho)}\right)^{3}$. As $r=$ $|\log \delta|$, we can ensure this by lowering the value of $\delta_{0}$.

Now we show that if one takes $q$ and $\alpha$ small enough, then the $\xi[\eta]$ - and $\xi[\lambda]$-directed rays started at the points $\pm\left\lfloor\alpha r N^{2 / 3}\right\rfloor e_{1}$ will avoid $\mathcal{D}$ as shown in Figure 8. To this end, recall $\xi[\rho]$ defined in 2.3. Let $u_{N}$ be the point where the $\xi[\lambda]$-ray starting from $\left\lfloor\alpha r N^{2 / 3}\right\rfloor e_{1}$ crosses the north boundary of $\llbracket 0, v_{N} \rrbracket$. Then the $e_{1}$-coordinates of $w_{N}^{+}$and $u_{N}$ can be lower bounded by

$$
\begin{align*}
& \left(\frac{\Psi_{1}(\lambda)}{\Psi_{1}(\mu-\lambda)} \cdot \frac{\Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)}-\frac{\Psi_{1}(\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)}\right) N-\alpha r N^{2 / 3}-q r N^{2 / 3}-5 \\
& \quad=\frac{\Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \cdot\left(\frac{\Psi_{1}(\lambda)}{\Psi_{1}(\mu-\lambda)}-\frac{\Psi_{1}(\rho)}{\Psi_{1}(\mu-\rho)}\right) N-\alpha r N^{2 / 3}-q r N^{2 / 3}-5 \\
& \geqslant C_{1}(\varepsilon) r N^{2 / 3}-\alpha r N^{2 / 3}-q r N^{2 / 3}-5, \tag{4.42}
\end{align*}
$$

where the inequality comes from Taylor's theorem because $\Psi_{1}$ is a smooth function on a compact interval inside $(0, \mu)$ depending on $\varepsilon$. Here, $C_{1}(\varepsilon)$ is a finite positive constant that only depends on $\varepsilon$. The inequality holds provided $r N^{-1 / 3} \leqslant c(\varepsilon)$ for some positive $c(\varepsilon)$ that only depends on $\varepsilon$ and this can be guaranteed to hold by lowering the threshold $\delta_{0}$ because

$$
r N^{-1 / 3} \leqslant|\log \delta| \delta^{1 / 2} \leqslant \delta_{0}^{1 / 3}
$$



FIGURE 8 Illustration of the set $\mathcal{D}$, the nested polymer, and three characteristic directions. The parameters $q=\alpha$ are less than some small constant that depends only on $\rho, \delta$ is a small positive constant in $\left(0, \delta_{0}\right)$, and $r$ is a large constant with $r=|\log \delta|$.

Now choosing

$$
\begin{equation*}
q=\alpha=C_{1}(\varepsilon) / 10 \tag{4.43}
\end{equation*}
$$

we obtain

$$
\begin{equation*}
(4.42) \geqslant C_{2}(\varepsilon) r N^{2 / 3}, \tag{4.44}
\end{equation*}
$$

and this gives us the desired picture for $\xi[\lambda]$ shown in Figure 8. The argument for the $\xi[\eta]$-directed ray is similar. For what follows we also want to guarantee that $\delta<\alpha r=\alpha|\log \delta|$. This can be done by decreasing the value of $\delta_{0}$ after having fixed $\alpha$. This completes the setup described in Figure 8.

Consider the set $\mathcal{D}$ shown in Figure 7 in dark gray and also in Figure 8. Place the stationary polymer model on $0+\mathbb{Z}_{\geqslant 0}^{2}$ as a nested polymer inside a larger stationary polymer model on the quadrant $-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}+\mathbb{Z}_{\geqslant 0}^{2}$. From the relation between two nested polymers given by Lemma A.7, we have

$$
\begin{align*}
& \mathbb{P}\left(\max _{z \in \mathcal{D}} Q_{0, z}^{\rho}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right)  \tag{4.45}\\
= & \mathbb{P}\left(\max _{z \in \mathcal{D}} Q_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left\{\left\lfloor r N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor r N^{2 / 3}\right\rfloor+\delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \\
\leqslant & \mathbb{P}\left(\max _{z \in D} \frac{Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\left\lfloor r N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor r N^{2 / 3}\right\rfloor+\delta N^{2 / 3}\right)}{Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\left\lfloor r N^{2 / 3}\right\rfloor-\alpha r N^{2 / 3}+1 \leqslant \tau \leqslant\left\lfloor r N^{2 / 3}\right\rfloor+\alpha r N^{2 / 3}\right)} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \\
= & \mathbb{P}\left(\operatorname { m i n } _ { z \in \mathcal { D } } \left\{\log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\left\lfloor r N^{2 / 3}\right\rfloor-\alpha N^{2 / 3}+1 \leqslant \tau \leqslant\left\lfloor r N^{2 / 3}\right\rfloor+\alpha N^{2 / 3}\right)\right.\right. \\
& \left.\left.\quad-\log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\left\lfloor r N^{2 / 3}\right\rfloor+1 \leqslant \tau \leqslant\left\lfloor r N^{2 / 3}\right\rfloor+\delta N^{2 / 3}\right)\right\} \leqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right)
\end{align*}
$$

$$
\begin{aligned}
& \leqslant \mathbb{P}\left(\operatorname { m i n } _ { z \in \mathcal { D } } \left\{\max _{i \in \llbracket-\alpha N^{2 / 3}+1, \alpha N^{2 / 3} \rrbracket} \log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor+i\right)\right.\right. \\
& \quad-\log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor\right)+\log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor\right) \\
& \left.\left.\quad-\max _{k \in \llbracket 1, \delta N^{2 / 3} \rrbracket} \log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor+k\right)\right\} \leqslant 2|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right),
\end{aligned}
$$

and by a union bound, the last probability above can be upper bounded as follows

$$
\begin{align*}
& \leqslant \mathbb{P}\left(\operatorname { m i n } _ { z \in \mathcal { D } } \left\{\max _{i \in \llbracket-\alpha N^{2 / 3}+1, \alpha N^{2 / 3} \rrbracket} \log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor+i\right)\right.\right. \\
& \left.\left.-\log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor\right)\right\} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right)  \tag{4.46}\\
& +\mathbb{P}\left(\operatorname { m a x } _ { z \in \mathcal { D } } \left\{\max _{k \in \llbracket 1, \delta N^{2 / 3} \rrbracket} \log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor+k\right)\right.\right. \\
& \left.\left.\quad \quad \log Z_{-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, z}^{\rho}\left(\tau=\left\lfloor r N^{2 / 3}\right\rfloor\right)\right\} \geqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) .
\end{align*}
$$

Before we continue our bound, let us simplify our notation. For $z \in \mathcal{D}$ and $i \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+$ $1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket$, define horizontal increments

$$
\widetilde{I}_{(i, 1)}^{z}=\frac{Z_{(i-1,1), z}}{Z_{(i, 1), z}}
$$

which live on the horizontal line $y=1$. With these increments, define a two-sided multiplicative walk $\left\{M_{n}^{z}\right\}_{n \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket}$ by setting $M_{0}^{z}=1$ and

$$
\begin{equation*}
M_{n}^{z} / M_{n-1}^{z}=I_{(n, 0)}^{\rho} / \widetilde{I}_{(n, 1)}^{z} \tag{4.47}
\end{equation*}
$$

where $I_{(n, 0)}^{\rho}$ are the boundary weights from the stationary polymer in the quadrant $-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}+$ $\mathbb{Z}_{\geqslant 0}^{2}$. Note that $n=0$ corresponds to $\tau=\left\lfloor r N^{2 / 3}\right\rfloor$, which is exit at the origin.

Then, (4.46) can be upper bounded as

$$
\begin{gather*}
\text { (4.46) }=\mathbb{P}\left(\min _{z \in \mathcal{D}} \max _{n \in \llbracket-\alpha r N^{2 / 3}+1, \alpha r N^{2 / 3} \rrbracket} \log M_{n}^{z} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right)  \tag{4.48}\\
+\mathbb{P}\left(\max _{z \in \mathcal{D}} \max _{n \in \llbracket 1, \delta N^{2 / 3} \rrbracket} \log M_{n}^{z} \geqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right)  \tag{4.49}\\
\leqslant \mathbb{P}\left(\left\{\min _{z \in \mathcal{D}} \max _{n \in \llbracket 1,\left\lfloor\frac{1}{2} \alpha r N^{2 / 3} \rrbracket \rrbracket\right.} \log M_{n}^{z} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right\}\right. \\
\left.\bigcap\left\{\min _{z \in \mathcal{D}} \max _{n \in \llbracket-\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor, 0 \rrbracket} \log M_{n}^{z} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right\}\right)  \tag{4.50}\\
+\mathbb{P}\left(\max _{z \in \mathcal{D}} \max _{n \in \llbracket 1,\left\lfloor\delta N^{2 / 3} \rrbracket \rrbracket\right.} \log M_{n}^{z} \geqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) . \tag{4.51}
\end{gather*}
$$



FIGURE 9 Setup for the stationary polymer with ratios of partition functions.

For any $z \in \mathcal{D}$, Lemma A. 3 gives

$$
M_{n}^{z} \geqslant M_{n}^{w_{N}^{+}} \quad \text { for } n \geqslant 1 \quad \text { and } \quad M_{n}^{z} \geqslant M_{n}^{w_{N}^{-}} \quad \text { for } n \leqslant 0
$$

Therefore, we may bound (4.50) and (4.51) by

$$
\begin{gather*}
(4.50)+(4.51) \leqslant \mathbb{P}\left(\left\{\max _{n \in \llbracket 1,\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor \rrbracket} \log M_{n}^{w_{N}^{+}} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right\}\right.  \tag{4.52}\\
\left.\bigcap\left\{\max _{n \in \llbracket-\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor, 0 \rrbracket} \log M_{n}^{w_{N}^{-}} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right\}\right) \\
+\mathbb{P}\left(\max _{n \in \llbracket 1,\left\lfloor\delta N^{2 / 3}\right\rfloor \rrbracket} \log M_{n}^{w_{N}^{-}} \geqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \tag{4.53}
\end{gather*}
$$

Next, to each edge on the north and east sides of the rectangle $\llbracket-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}, v_{N}+e_{1}+e_{2} \rrbracket$, we attach both $\lambda$ - and $\eta$-edge weights, coupled as in [9, Theorem B.4]. We denote these weights by $I_{v_{n}+k e_{1}+e_{2}}^{\lambda, \mathrm{NE}}, J_{v_{n}+e_{1}+k e_{2}}^{\lambda, \mathrm{NE}}, I_{v_{n}+k e_{1}+e_{2}}^{\eta, \mathrm{NE}}$, and $J_{v_{n}+e_{1}+k e_{2}}^{\eta, \mathrm{NE}}, k \leqslant 1$. Together with the bulk weights in $\llbracket-\left\lfloor r N^{2 / 3}\right\rfloor e_{1}+e_{2}, v_{N} \rrbracket$, these define stationary polymers with northeast boundary. Let us denote their partition functions by $Z_{x, v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}$ and $Z_{x, v_{N}+e_{1}+e_{2}}^{\eta, \mathrm{NE}}$ for $x \in \llbracket\left(-\left\lfloor r N^{2 / 3}\right\rfloor, 1\right), v_{N} \rrbracket$. The corresponding polymer measures are denoted by $Q_{x, v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}$ and $Q_{x, v_{N}+e_{1}+e_{2}}^{\eta, \mathrm{NE}}$, respectively. This is depicted in Figure 9.

On the horizontal line $y=1$, let us also define for $i \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket$

$$
\begin{equation*}
I_{(i, 1)}^{\lambda, \mathrm{NE}}=\frac{Z_{(i-1,1), v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}}{Z_{(i, 1), v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}} \quad \text { and } \quad I_{(i, 1)}^{\eta, \mathrm{NE}}=\frac{Z_{(i-1,1), v_{N}+e_{1}+e_{2}}^{\eta, \mathrm{NE}} .}{Z_{(i, 1), v_{N}+e_{1}+e_{2}}^{\eta, \mathrm{NE}} .} \tag{4.54}
\end{equation*}
$$

Lemma 4.13. There exists a positive constant $C$, depending only on $\varepsilon$, such that for $\alpha, r, N$ as chosen above, and for any integers $a, b \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket$, the event

$$
\begin{equation*}
A=\left\{\frac{1}{2} \prod_{i=a}^{b} I_{(i, 1)}^{\eta, \mathrm{NE}} \leqslant \prod_{i=a}^{b} \widetilde{I}_{(i, 1)}^{w_{N}^{-}} \leqslant \prod_{i=a}^{b} \widetilde{I}_{(i, 1)}^{w_{N}^{+}} \leqslant 2 \prod_{i=a}^{b} I_{(i, 1)}^{\lambda, \mathrm{NE}}\right\} \tag{4.55}
\end{equation*}
$$

satisfies $\mathbb{P}\left(A^{c}\right) \leqslant e^{-C r^{3}}$.
Proof. Due to the relative positions of $w_{N}^{ \pm}$, Lemma A. 3 implies the middle inequality in the definition of $A$. We will prove the desired bound for the inequality on the right, that is,

$$
\begin{equation*}
\mathbb{P}\left(\prod_{i=a}^{b} \widetilde{I}_{i}^{w_{N}^{+}} \leqslant 2 \prod_{i=a}^{b} I_{i}^{\lambda, \mathrm{NE}}\right) \geqslant 1-e^{-C r^{3}} . \tag{4.56}
\end{equation*}
$$

The argument for the inequality on the left is similar and will be omitted.
Let $\tau^{\mathrm{NE}}$ be defined similarly to $\tau$, but acting on down-left paths. Namely, it gives the number of steps the path takes before making its first corner. We will again use the convention that $\tau^{\mathrm{NE}}>0$ if the first step of the path is $-e_{1}$ and $\tau^{\mathrm{NE}}<0$ if the first step is $-e_{2}$.

Our estimate essentially follows from the following two facts. The first fact is that the random variable

$$
Q_{\left\lfloor\alpha r N^{2 / 3}\right\rfloor e_{1}+e_{2}, v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}\left\{\tau^{\mathrm{NE}} \geqslant q r N^{2 / 3}\right\}
$$

is, almost surely, less than or equal to

$$
Q_{(a, 1), v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}\left\{\tau^{\mathrm{NE}} \geqslant q r N^{2 / 3}\right\} \quad \forall a \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket .
$$

This follows directly from Lemma A.5, although note that here we exit from the NE boundary instead of the SW boundary. The second fact is that there exist positive constants $C_{1}$ and $C_{2}$ such that

$$
\begin{equation*}
\mathbb{P}\left(Q_{\left\lfloor\alpha r N^{2 / 3}\right\rfloor e_{1}+e_{2}, v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}\left\{\tau^{\mathrm{NE}} \geqslant q r N^{2 / 3}\right\} \geqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right) \geqslant 1-e^{-C_{2} r^{3}} . \tag{4.57}
\end{equation*}
$$

To see this, observe that

$$
\mathbb{P}\left(Q_{\left\lfloor\alpha r N^{2 / 3}\right\rfloor e_{1}+e_{2}, v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}\left\{\tau^{\mathrm{NE}} \leqslant q r N^{2 / 3}\right\} \geqslant e^{-C_{1} r^{2} N^{1 / 3}}\right) \leqslant e^{-C_{2} r^{3}}
$$

is the same as (4.35), except here we rotate the picture by $180^{\circ}$. The key idea is illustrated in Figure 10. Note the similarities between Figures 6 and 10. From Figure 10, the calculation $z_{N} \cdot e_{2}-v_{N} \cdot e_{2}-1 \geqslant C r N^{2 / 3}$ is omitted because it is similar to (4.37).

Let $Z_{(b, 1), w_{N}^{+}+e_{2}}^{\lambda, \mathrm{N}}$ denote the partition function for up-right paths from $(b, 1)$ to $w_{N}^{+}+e_{2}$, which uses the same weights as $Z_{(b, 1), w_{N}^{+}+e_{2}}^{\lambda, \mathrm{NE}}$ does on the north boundary but uses the original (bulk) weights on $w_{N}^{+}+\mathbb{Z}_{\leqslant 0}^{2}$.

On the high probability event

$$
\begin{equation*}
\left\{Q_{\left\lfloor\alpha r N^{2 / 3}\right\rfloor e_{1}+e_{2}, v_{N}+e_{1}+e_{2}}^{\lambda, \mathrm{NE}}\left\{\tau^{\mathrm{NE}} \geqslant q r N^{2 / 3}\right\} \geqslant 1-e^{-C_{1} r^{2} N^{1 / 3}}\right\}, \tag{4.58}
\end{equation*}
$$



FIGURE 10 Illustration of (4.57). By Lemma A.8, $Q_{a_{N}, b_{N}}^{\lambda, \mathrm{NE}}\left(\tau^{\mathrm{NE}} \leqslant q r N^{2 / 3}\right)=Q_{a_{N}, z_{N}}^{\lambda, \mathrm{NE}}\left(\tau^{\mathrm{NE}}<-\left(z_{N} \cdot e_{2}-v_{N} \cdot e_{2}-1\right)\right)$, and this is unlikely because $z_{N} \cdot e_{2}-v_{N} \cdot e_{2}-1 \geqslant C r N^{2 / 3}$.
we have

$$
\begin{aligned}
& \prod_{i=a+1}^{b} \widetilde{I}_{(i, 1)}^{w_{N}^{+}}=\frac{Z_{(a, 1), w_{N}^{+}}^{Z_{(b, 1), w_{N}^{+}}}}{} \\
& \leqslant \frac{Z_{(a, 1), w_{N}^{+}+e_{2}}^{N}}{Z_{(b, 1), w_{N}^{+}+e_{2}}^{N}} \quad \quad \quad \text { By Lemma A.3) } \\
&=\frac{Z_{(a, 1), w_{N}^{+}+e_{2}}^{N} \prod_{i=1}^{\left\lfloor q r N^{2 / 3}\right\rfloor+1} I_{v_{N}+e_{1}+e_{2}-i e_{1}}^{\lambda, \mathrm{NE}}}{Z_{(b, 1), w_{N}^{+}+e_{2}}^{N} \prod_{i=1}^{\left\lfloor q r N^{2 / 3}\right\rfloor+1} I_{v_{N}+e_{1}+e_{2}-i e_{1}}^{\lambda, \mathrm{NE}}} \\
&=\frac{Z_{(a, 1), v_{N}+e_{1}+e_{2}}^{Z_{(b, 1), v_{N}+e_{1}+e_{2}}^{\mathrm{NE}}\left(\tau^{\mathrm{NE}} \geqslant\left\lfloor q r N^{2 / 3}\right\rfloor\right)}}{} \begin{array}{l}
=\frac{Q_{(a, 1), v_{N}+e_{1}+e_{2}}^{\mathrm{NE}}\left(\tau^{\mathrm{NE}} \geqslant\left\lfloor q r N^{2 / 3}\right\rfloor\right)}{Q_{(b, 1), v_{N}+e_{1}+e_{2}}^{\mathrm{NE}}\left(\tau^{\mathrm{NE}} \geqslant\left\lfloor q r N^{2 / 3}\right\rfloor\right)} \prod_{i=a+1}^{b} I_{(i, 1)}^{\lambda, \mathrm{NE}} \\
\end{array} \\
& \leqslant \frac{1}{1-e^{-C_{1} r^{2} N^{1 / 3}} \prod_{i=a+1}^{b} I_{(i, 1)}^{\lambda, \mathrm{NE}} \quad(\text { on the event }(4.58)) \text {. }}
\end{aligned}
$$

With the new horizontal increments $I^{\lambda, \mathrm{NE}}$ and $I^{\eta, \mathrm{NE}}$, define two more two-sided multiplicative random walks $M_{n}^{\lambda}$ and $M_{n}^{\eta}$ with $M_{0}^{\lambda}=M_{0}^{\eta}=1$,

$$
M_{n}^{\lambda} / M_{n-1}^{\lambda}=I_{(n, 0)}^{\rho} / I_{(n, 1)}^{\lambda, \mathrm{NE}}, \quad \text { and } \quad M_{n}^{\eta} / M_{n-1}^{\eta}=I_{(n, 0)}^{\rho} / I_{(n, 1)}^{\eta, \mathrm{NE}}
$$

On the event $A$ from (4.55), we get

$$
\begin{equation*}
\frac{1}{2} M_{n}^{\lambda} \leqslant M_{n}^{w_{N}^{+}} \leqslant 2 M_{n}^{\eta} \text { for } n \geqslant 1 \text { and } \frac{1}{2} M_{n}^{\eta} \leqslant M_{n}^{w_{N}^{-}} \leqslant 2 M_{n}^{\lambda} \text { for } n \leqslant 0 . \tag{4.5}
\end{equation*}
$$

Now we can bound

$$
\begin{gather*}
\mathbb{P}(\text { event in }(4.52) \cap A) \leqslant \mathbb{P}\left(\left\{\max _{n \in \llbracket 1,\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor \rrbracket} \log M_{n}^{\lambda} \leqslant 6|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right\}\right.  \tag{4.60}\\
\left.\bigcap\left\{\max _{n \in \llbracket-\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor, 0 \rrbracket} \log M_{n}^{\eta} \leqslant 6|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right\}\right), \\
\mathbb{P}(\text { event in }(4.53) \cap A) \leqslant \mathbb{P}\left(\max _{n \in \llbracket 1,\left\lfloor\delta N^{2 / 3}\right\rfloor \rrbracket} \log M_{n}^{\eta} \geqslant \frac{1}{2}|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) . \tag{4.61}
\end{gather*}
$$

[9, Theorem B.4] states that the increment variables $\left\{I_{(i, 1)}^{\lambda, \mathrm{NE}}\right\}_{i \geqslant 1} \cup\left\{I_{(i, 1)}^{\eta, \mathrm{NE}}\right\}_{i \leqslant 0}$ are independent, and these are independent of the boundary weights $\left\{I_{(i, 0)}^{\rho}\right\}$ by construction. Thus, we get

$$
\begin{align*}
(4.60) \leqslant \mathbb{P}\left(\max _{n \in \llbracket 1,\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor \rrbracket} \log M_{n}^{\lambda}\right. & \left.\leqslant 6|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \\
& \times \mathbb{P}\left(\max _{n \in \llbracket-\left\lfloor\frac{1}{2} \alpha r N^{2 / 3}\right\rfloor, 0 \rrbracket} \log M_{n}^{\eta} \leqslant 6|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) . \tag{4.62}
\end{align*}
$$

The next step is a random walk estimate because the steps of the walks $\log M_{n}^{\lambda}$ and $\log M_{n}^{\eta}$ are given by the difference of two independent log-gamma random variables, which are sub-exponential random variables. Using Proposition A.13, we see that (4.62) $\leqslant C|\log \delta|^{6} \delta$. Using Theorem A.11, we also have $(4.61) \leqslant C \delta$.

To summarize, we have shown

$$
\begin{align*}
\mathbb{P}(\text { event in }(4.45)) & \leqslant 2 \mathbb{P}\left(A^{c}\right)+\mathbb{P}(\text { event in }(4.52) \cap A)+\mathbb{P}(\text { event in }(4.53) \cap A) \\
& \leqslant 2 e^{-C|\log \delta|^{3}}+C|\log \delta|^{6} \delta  \tag{4.63}\\
& \leqslant|\log \delta|^{10} \delta .
\end{align*}
$$

This completes the proof of the desired bound (4.2) with the maximum taken over the dark region $\mathcal{D} \subset \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket$ in Figure 7 .

For the endpoints in $\mathcal{L}^{+}$, we have the following estimate,

$$
\begin{aligned}
& \mathbb{P}\left(\max _{z \in \mathcal{L}^{+}} Q_{0, z}^{\rho}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \\
& \leqslant \mathbb{P}\left(\max _{z \in \mathcal{L}^{+}} Q_{0, z}^{\rho}\{1 \leqslant \tau\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right)
\end{aligned}
$$



FIGURE 11 We have $Q_{0, w_{N}^{-}}\left\{\tau \leqslant \delta N^{2 / 3}\right\}=Q_{\left(\left[\delta N^{2 / 3}\right],-h\right), w_{N}^{-}}\{\tau<-h\}$ which is rare because $h$ is lower bounded by $C r N^{2 / 3}$. The lower bound on $h$ follows from the fact the vertical distance between $v_{N}$ and $w_{N}^{-}$is of order $r N^{2 / 3}$.

$$
\begin{array}{ll}
\leqslant \mathbb{P}\left(Q_{0, w_{N}^{+}}^{\rho}\{1 \leqslant \tau\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) & \text { (by Lemma A.5) } \\
\leqslant e^{-C|\log \delta|^{3}} & \quad \text { (by Corollary } 4.10)
\end{array}
$$

Similarly, for the $\mathcal{L}^{-}$region, we have

$$
\begin{aligned}
\mathbb{P}\left(\max _{z \in \mathcal{L}^{-}} Q_{0, z}^{\rho}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) & \leqslant \mathbb{P}\left(\max _{z \in \mathcal{L}^{-}} Q_{0, z}^{\rho}\left\{\tau \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \\
& \leqslant \mathbb{P}\left(Q_{0, w_{N}^{-}}^{\rho}\left\{\tau \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \\
& \leqslant e^{-C|\log \delta|^{3}}
\end{aligned}
$$

The idea for the last inequality is illustrated in Figure 11, essentially again following from Lemma A. 5 and Corollary 4.10. This finishes the argument for the $\mathcal{L}^{-}$region. The bound (4.2) is thus proved.

The probability bound implies the upper bound in (4.3):

$$
\mathbb{E}\left[\max _{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, z}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}\right] \leqslant \delta+\mathbb{P}\left(\max _{z \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, z}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\} \geqslant \delta\right) \leqslant C|\log \delta|^{10} \delta
$$

We turn to the lower bound in (4.3). By Lemma 4.6, there exist two constants $r_{0}$ and $N_{0}$ (depending on $\varepsilon$ ) such that, for $N \geqslant N_{0}$,

$$
\begin{equation*}
\mathbb{E}\left[Q_{0, v_{N}+e_{1}+e_{2}}^{\rho}\left\{|\tau| \leqslant r_{0} N^{2 / 3}\right\}\right] \geqslant \frac{1}{2} \tag{4.64}
\end{equation*}
$$



FIG URE 12 Left: Partition for the collection of paths in (4.64). The origin is not necessarily a partition point. Right: An illustration for (4.68). The nested polymer with its quenched measure $Q_{z, v_{N}^{\prime}}^{(0)}$ is shown in black.

Abbreviate $v_{N}^{\prime}=v_{N}+e_{1}+e_{2}$. Given $\delta \geqslant N^{-2 / 3}$, partition $\left[-r_{0}, r_{0}\right]$ as

$$
-r_{0}=p_{0}<p_{1}<\cdots<p_{\left\lfloor\frac{2 r_{0}}{\delta}\right\rfloor}<p_{\left\lfloor\frac{2 r_{0}}{\delta}\right\rfloor+1}=r_{0}
$$

with mesh size $p_{i+1}-p_{i} \leqslant \delta$. See the left side of Figure 12. By (4.64), there exists an integer $i^{\star} \in$ [ $\left.0,\left\lfloor\frac{2 r_{0}}{\delta}\right\rfloor\right]$ such that

$$
\begin{equation*}
\mathbb{E}\left[Q_{0, v_{N}^{\prime}}^{\rho}\left\{p_{i \star} N^{2 / 3} \leqslant \tau \leqslant p_{i^{\star}+1} N^{2 / 3}\right\}\right] \geqslant \frac{\frac{1}{2} \delta}{2 r_{0}}=C(\varepsilon) \delta . \tag{4.65}
\end{equation*}
$$

As we cannot control the exact location of $i^{\star}$, we compensate by varying the endpoint around $v_{N}^{\prime}$. Let

$$
A_{N}=\llbracket v_{N}^{\prime}-r_{0} N^{2 / 3} e_{1}, v_{N}^{\prime} \rrbracket \cup \llbracket v_{N}^{\prime}-r_{0} N^{2 / 3} e_{2}, v_{N}^{\prime} \rrbracket
$$

denote the set of lattice points on the boundary of the rectangle $\llbracket 0, v_{N}^{\prime} \rrbracket$ within distance $r_{0} N^{2 / 3}$ of the upper right corner $v_{N}^{\prime}$. We claim that for any integer $i \in\left[0,\left\lfloor\frac{2 r_{0}}{\delta}\right\rfloor\right]$,

$$
\begin{equation*}
\mathbb{E}\left[\max _{z \in A_{N}} Q_{0, z}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}\right] \geqslant \mathbb{E}\left[Q_{0, v_{N}^{\prime}}^{\rho}\left\{p_{i^{\star}} N^{2 / 3} \leqslant \tau \leqslant p_{i^{\star}+1} N^{2 / 3}\right\}\right] . \tag{4.66}
\end{equation*}
$$

Then bounds (4.65) and (4.66) imply

$$
\begin{equation*}
\mathbb{E}\left[\max _{z \in A_{N}} Q_{0, z}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}\right] \geqslant C(\rho) \delta, \tag{4.67}
\end{equation*}
$$

and the lower bound in (4.3) follows directly from (4.67).
It remains to prove claim (4.66). If $p_{i^{\star}} \leqslant 0 \leqslant p_{i^{\star}+1}$, (4.66) is immediate. We argue the case $p_{i^{\star}+1}>p_{i^{\star}}>0$, the other one being analogous. Set $z=\left(\left\lfloor p_{i^{\star}} N^{2 / 3}\right\rfloor-1\right) e_{1}$ and apply Lemma A. 7
to the polymer with the nested quenched measure $Q_{z, .}^{(0)}$. See the right side of Figure 12. Then

$$
\begin{align*}
& \mathbb{E}\left[Q_{0, v_{N}^{\prime}}^{\rho}\left\{p_{i^{\star}} N^{2 / 3} \leqslant \tau \leqslant p_{i^{\star}+1} N^{2 / 3}\right\}\right] \\
& \leqslant \mathbb{E}\left[Q_{z, v_{N}^{\prime}}^{(0)}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\}\right]  \tag{4.68}\\
& \left.=\mathbb{E}\left[Q_{0, v_{N}^{\prime}-\left(\left\lfloor p_{i \star} N^{2 / 3}\right\rfloor-1\right) e_{1}}^{\rho}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\}\right]\right] \quad \text { (by shift-invariance) } \\
& \leqslant \mathbb{E}\left[\max _{z \in A_{N}} Q_{0, z}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\}\right] .
\end{align*}
$$

Theorem 4.2 is proved.

## 4.3 | Coupled polymer measures

Proof of Theorem 4.3. From Theorem 4.1, there exists an event $A$ with probability at least $e^{-C_{1} r^{3}}$ such that on $A$, we have

$$
\min _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau|>r N^{2 / 3}\right\} \geqslant 1-e^{-C_{2} r^{2} N^{1 / 3}} .
$$

By a union bound, on the event $A$ we have

$$
\widetilde{Q}_{0, \Omega^{\mathrm{NE}} \llbracket 0, v_{N} \mathbb{\rrbracket}}^{\rho}\left(\bigcap_{x \in \partial \mathrm{NE} \llbracket 0, v_{N} \rrbracket}\left\{\left|\widetilde{\tau}_{0, x}\right|>r N^{2 / 3}\right\}\right) \geqslant 1-N e^{-C_{2} r^{2} N^{1 / 3}} \geqslant 1-e^{-C_{3} r^{2} N^{1 / 3}}
$$

provided that $r_{0}, N_{0}$ are sufficiently large. With this, we have finished the proof of this theorem.

Proof of Theorem 4.4. By Theorem 4.2, on the high probability event $B$ with probability at least $1-C_{1} \delta|\log \delta|^{10}$, we have

$$
\max _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{|\tau| \leqslant \delta N^{2 / 3}\right\} \leqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}
$$

With the assumption that $\sqrt{\delta} N^{1 / 3} \geqslant 1$, a union bound implies that on $B$,

$$
\widetilde{Q}_{0, \partial \mathrm{NE} \llbracket 0, v_{N} \rrbracket}\left(\bigcup_{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{\widetilde{\tau}_{0, x} \leqslant \delta N^{2 / 3}\right\}\right) \leqslant N e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}} \leqslant \delta .
$$

The claim of the theorem follows.

## 5 | COALESCENCE OF SEMI-INFINITE POLYMERS

In this section, we will define the semi-infinite polymer measures and prove Theorems 2.1 and 2.3 about their coalescence. The proof will use a duality between forward and backward polymer measures, which we describe in Subsection 5.2.

## 5.1 | Busemann functions and semi-infinite polymers

Following [18, Theorem 4.1], for any fixed $\rho \in(0, \mu), \mathbb{P}$-almost surely, the limits

$$
\begin{equation*}
B^{\rho}(x, y)=\lim _{N \rightarrow \infty}\left(\log Z_{x, v_{N}}-\log Z_{y, v_{N}}\right), \tag{5.1}
\end{equation*}
$$

exist for any $x, y \in \mathbb{Z}^{2}$ and satisfy

$$
Y_{z}^{-1}=e^{-B^{\rho}\left(z, z+e_{1}\right)}+e^{-B^{\rho}\left(z, z+e_{2}\right)}
$$

and

$$
B^{\rho}(x, y)+B^{\rho}(y, z)=B^{\rho}(x, z),
$$

for all $x, y, z \in \mathbb{Z}^{2}$. Furthermore, for any $z \in \mathbb{Z}^{2}, I_{z}^{\rho}=e^{B^{\rho}\left(z-e_{1}, z\right)} \sim \mathrm{Ga}^{-1}(\mu-\rho), J_{z}^{\rho}=e^{B^{\rho}\left(z-e_{2}, z\right)} \sim$ $\mathrm{Ga}^{-1}(\rho)$, and if we fix any vertex $v \in \mathbb{Z}^{2}$, then the weights $Y_{z}, I_{v-k e_{1}}^{\rho}, J_{v-k e_{2}}^{\rho}, z \in v-\mathbb{Z}_{>0}^{2}, k \geqslant 0$, are mutually independent and thus define a stationary polymer with northeast boundary on $v-\mathbb{Z}_{\geqslant 0}^{2}$. The partition function and quenched polymer measure will be denoted by $Z_{\cdot, v}^{\rho, \mathrm{NE}}, Q_{\cdot, v}^{\rho, \mathrm{NE}}$. Similarly, if we define

$$
\breve{Y}_{z}^{\rho}=\frac{1}{e^{-B^{\rho}\left(z-e_{1}, z\right)}+e^{-B^{\rho}\left(z-e_{2}, z\right)}}, \quad z \in \mathbb{Z}^{2}
$$

then $\check{Y}_{z}^{\rho} \sim \mathrm{Ga}^{-1}(\mu)$ for all $z \in \mathbb{Z}^{2}$, and for any vertex $v \in \mathbb{Z}^{2}$ the weights $\check{Y}_{z}^{\rho}, I_{v+k e_{1}}^{\rho}, J_{v+k e_{2}}^{\rho}, z \in$ $v+\mathbb{Z}_{>0}^{2}, k \geqslant 1$, are mutually independent and defined a stationary polymer with southwest boundary on $v+\mathbb{Z}_{\geqslant 0}^{2}$. The partition function and quenched polymer measure will be denoted by $\breve{Z}_{v, \bullet}^{\rho, S W}, \breve{Q}_{v, \bullet}^{\rho, S W}$. Thus, for any $v \in \mathbb{Z}^{2}, \breve{Q}_{v, \bullet}^{\rho, S W}$ has the same distribution as the generic $Q_{v, \bullet}^{\rho}$ we introduced in Section 3 and used in Section 4. (This distributional equality is a special feature of the inverse-gamma polymer.)

The $\xi[\rho]$-directed (forward) semi-infinite polymer measure starting at $z$, denoted by $\Pi_{z}^{\rho}$, is a Markov chain on $\mathbb{Z}^{2}$ with transition probabilities

$$
\begin{align*}
& \pi^{\rho}\left(x, x+e_{1}\right)=\frac{J_{x+e_{2}}^{\rho}}{I_{x+e_{1}}^{\rho}+J_{x+e_{2}}^{\rho}}=Y_{x} e^{-B^{\rho}\left(x, x+e_{1}\right)}, \\
& \pi^{\rho}\left(x, x+e_{2}\right)=\frac{I_{x+e_{1}}^{\rho}}{I_{x+e_{1}}^{\rho}+J_{x+e_{2}}^{\rho}}=Y_{x} e^{-B^{\rho}\left(x, x+e_{2}\right)} . \tag{5.2}
\end{align*}
$$

The $\xi[\rho]$-directed backward semi-infinite polymer measure starting at $z$, denoted by $\check{\Pi}_{z}^{\rho}$, is a Markov chain on $\mathbb{Z}^{2}$ with transition probabilities

$$
\begin{equation*}
\check{\pi}^{\rho}\left(x, x-e_{1}\right)=\frac{J_{x}^{\rho}}{I_{x}^{\rho}+J_{x}^{\rho}}=\check{Y}_{x}^{\rho} e^{-B^{\rho}\left(x-e_{1}, x\right)} \quad \text { and } \quad \check{\pi}^{\rho}\left(x, x-e_{2}\right)=\frac{I_{x}^{\rho}}{I_{x}^{\rho}+J_{x}^{\rho}}=\check{Y}_{x}^{\rho} e^{-B^{\rho}\left(x-e_{2}, x\right)} . \tag{5.3}
\end{equation*}
$$

The next proposition relates the semi-infinite polymers to the stationary ones. For $u \in \mathbb{Z}^{2}$ and $v \in u+\mathbb{Z}_{>0}^{2}$ let $\Pi_{u, v}^{\rho}$ be the distribution of the Markov chain that starts at $u$, has transition probabilities $\pi^{\rho}\left(x, x+e_{i}\right), i \in\{1,2\}$, if $x \in \llbracket u, v-e_{1}-e_{2} \rrbracket$, and when it gets to $v-\mathbb{Z}_{>0} e_{i}, i \in\{1,2\}$, it takes $e_{i}$ steps to get to $v$ and end there. Similarly, let $\check{\Pi}_{v, u}^{\rho}$ be the distribution of the Markov chain that starts at $v$, has transition probabilities $\check{\pi}^{\rho}\left(x, x-e_{i}\right), i \in\{1,2\}$, if $x \in \llbracket u+e_{1}+e_{2}, v \rrbracket$, and when it gets to $u+\mathbb{Z}_{>0} e_{i}, i \in\{1,2\}$, it takes $-e_{i}$ steps to get to $u$ and end there.

Define, similarly to $\mathbb{X}_{u, v}$, the set $\mathbb{X}_{v, u}$ of down-left paths starting at $v$ and ending at $u$. For $x . \in$ $\mathbb{X}_{u, v}$, respectively, $\in \mathbb{X}_{v, u}$, let $\bar{x} . \in \mathbb{X}_{v, u}$, respectively, $\in \mathbb{X}_{u, v}$, be the path that traverses $x$. in the reverse direction.

Proposition 5.1. We have $\mathbb{P}$-almost surely, for any $u \in \mathbb{Z}^{2}$ and $v \in u+\mathbb{Z}_{>0}^{2}$, for any $x . \in \mathbb{X}_{u, v}$,

$$
\Pi_{u, v}^{\rho}\left(x_{\bullet}\right)=Q_{u, v}^{\rho, \mathrm{NE}}\left(x_{\bullet}\right) \quad \text { and } \quad \check{\Pi}_{v, u}^{\rho}\left(\bar{x}_{\bullet}\right)=\breve{Q}_{u, v}^{\rho, \mathrm{SW}}\left(x_{\bullet}\right) .
$$

Proof. We prove the second claim, the first one being similar. Let $\ell=|v-u|_{1}$ and index the path $x_{\text {. so that }} x_{0}=u$ and $x_{\ell}=v$. We will consider the case where $x_{1}=e_{1}$ and the proof in the other case is identical. Let $k \geqslant 1$ be such that $x_{k}=u+k e_{1}$ and $x_{k+1}=u+k e_{1}+e_{2}$. Then

$$
\begin{aligned}
\check{\Pi}_{v, u}^{\rho}(\bar{x} .) & =\prod_{i=k}^{\ell-1} \check{\pi}^{\rho}\left(x_{i+1}, x_{i}\right)=\prod_{i=k}^{\ell-1} \breve{Y}_{x_{i+1}}^{\rho} e^{-B^{\rho}\left(x_{i}, x_{i+1}\right)} \\
& =e^{-B^{\rho}\left(x_{k}, v\right)} \prod_{i=k}^{\ell-1} \breve{Y}_{x_{i+1}}^{\rho}=e^{-B^{\rho}(u, v)} \prod_{i=1}^{k} I_{u+i e_{1}}^{\rho} \prod_{i=k}^{\ell-1} \breve{Y}_{x_{i+1}}^{\rho} .
\end{aligned}
$$

Adding the above over all paths $x_{.} \in \mathbb{X}_{u, v}$ gives

$$
1=e^{-B^{\rho}(u, v)} \breve{Z}_{u, v}^{\rho, S W}
$$

Consequently,

$$
\check{\Pi}_{v, u}^{\rho}\left(\bar{x}_{.}\right)=\frac{\prod_{i=1}^{k} I_{u+i e_{1}}^{\rho} \prod_{i=k}^{\ell-1} \breve{Y}_{x_{i+1}}^{\rho}}{\check{Z}_{u, v}^{\rho, S \mathrm{~S}}}=\check{Q}_{u, v}^{\rho, \mathrm{SW}}\left(x_{\mathrm{s}}\right) .
$$

## 5.2 | Coupling the forward and backward semi-infinite polymers

We now couple the polymer measures $\left\{\Pi_{z}^{\rho}: z \in \mathbb{Z}^{2}\right\}$ following the construction in [25, appendix A]. To this end, introduce a collection of independent and identically distributed Uniform $[0,1]$ random variables $\left\{\theta_{z}\right\}_{z \in \mathbb{Z}^{2}}$ that are also independent of the random environment $\left\{Y_{z}: z \in \mathbb{Z}^{2}\right\}$. Let $\mathbf{P}$ denote the distribution of $\theta$.

Define a directed random graph $g^{\rho}$ on $\mathbb{Z}^{2}$, according to the following rule

$$
g^{\rho}(x)= \begin{cases}e_{1} & \text { if } \theta_{x} \leqslant \frac{J_{x+e_{2}}^{\rho}}{I_{x+e_{1}}^{\rho}+J_{x+e_{2}}^{\rho}} \\ e_{2} & \text { if } \theta_{x}>\frac{I_{x+e_{1}}^{\rho}}{I_{x+e_{1}}^{\rho}+J_{x+e_{2}}^{\rho}}\end{cases}
$$

From $g^{\rho}$, we can construct a semi-infinite path $X^{\rho, z}$ defined by

$$
\begin{equation*}
X_{0}^{\rho, z}=z \quad \text { and } \quad X_{n+1}^{\rho, z}=X_{n}^{\rho, z}+g^{\rho}\left(X_{n}^{\rho, z}\right) . \tag{5.4}
\end{equation*}
$$

It is clear from the construction that for $\mathbb{P}$-almost every $Y_{.}$, the distribution of $X^{\rho, z}$ under $\mathbf{P}$ is exactly $\Pi_{z}^{\rho}$. Namely, we have $\mathbb{P}$-almost surely, for any $z \in \mathbb{Z}^{2}$ and any finite up-right path $x$. starting at $z$,

$$
\begin{equation*}
\mathbf{P}\left\{X_{0}^{\rho, z}=x_{0}\right\}=\Pi_{z}^{\rho}\left\{x_{0}\right\} . \tag{5.5}
\end{equation*}
$$

We next couple the backward semi-infinite polymer measures together with the forward ones. To this end, define another (dual) random graph $\breve{g}^{\rho}$ by

$$
\breve{g}^{\rho}(x)= \begin{cases}-e_{1} & \text { if } g^{\rho}\left(x-e_{1}-e_{2}\right)=e_{1}, \\ -e_{2} & \text { if } g^{\rho}\left(x-e_{1}-e_{2}\right)=e_{2}\end{cases}
$$

Define the down-left semi-infinite paths $\breve{X}^{\rho, z}$ according to

$$
\begin{equation*}
\check{X}_{0}^{\rho, z}=z \quad \text { and } \quad \check{X}_{n+1}^{\rho, z}=\check{X}_{n}^{\rho, z}+\breve{g}^{\rho}\left(\breve{X}_{n}^{\rho, z}\right) . \tag{5.6}
\end{equation*}
$$

By construction, for $\mathbb{P}$-almost every $Y_{\text {. }}$, the distribution of $\breve{X}^{\rho, z}$ under $\mathbf{P}$ is that of a Markov chain on $\mathbb{Z}^{2}$ with steps in $\left\{-e_{1},-e_{2}\right\}$ and transition probabilities

$$
\begin{aligned}
\frac{J_{x-e_{1}}^{\rho}}{I_{x-e_{2}}^{\rho}+J_{x-e_{2}}^{\rho}} & =\frac{e^{B^{\rho}\left(x-e_{1}-e_{2}, x-e_{1}\right)}}{e^{B^{\rho}\left(x-e_{1}-e_{2}, x-e_{2}\right)}+e^{B^{\rho}\left(x-e_{1}-e_{2}, x-e_{1}\right)}}=\frac{e^{-B^{\rho}\left(x-e_{1}, x\right)}}{e^{-B^{\rho}\left(x-e_{2}, x\right)}+e^{-B^{\rho}\left(x-e_{1}, x\right)}} \\
& =\frac{e^{B^{\rho}\left(x-e_{2}, x\right)}}{e^{B^{\rho}\left(x-e_{1}, x\right)}+e^{B^{\rho}\left(x-e_{2}, x\right)}}=\check{\pi}^{\rho}\left(x, x-e_{1}\right)
\end{aligned}
$$

to go from $x$ to $x-e_{1}$ and, similarly,

$$
\frac{I_{x-e_{2}}^{\rho}}{I_{x-e_{2}}^{\rho}+J_{x-e_{2}}^{\rho}}=\check{\pi}^{\rho}\left(x, x-e_{2}\right)
$$

to go from $x$ to $x-e_{2}$.
Remark 5.2. Note that the graph $g^{\rho}$ and its coupled paths $\left\{X^{\rho, z}: z \in \mathbb{Z}^{2}\right\}$ are constructed to form a forest that covers all of $\mathbb{Z}^{2}$. By [25, Theorem A.2], this forest is in fact a spanning tree, with probability 1 under $\mathbb{P}$. The paths $\left\{\check{X}^{\rho, z}-\left(e_{1}+e_{2}\right) / 2: z \in \mathbb{Z}^{2}\right\}$ form the dual forest that spans the dual lattice $\mathbb{Z}^{2}-\left(e_{1}+e_{2}\right) / 2$. Again, by [25, Theorem A.2], this dual forest is also a spanning forest $\mathbb{P}$-almost surely.


FIG URE 13 The sampled polymers starting $\left(\left\lfloor\delta N^{2 / 3}\right\rfloor, 0\right)$ and $\left(0,\left\lfloor\delta N^{2 / 3}\right\rfloor\right)$ (gray dotted lines) coalesce outside $\llbracket 0, v_{N} \rrbracket$. Equivalently, some dual point $x^{*}=x-(1 / 2,1 / 2)$ outside of $\llbracket 0, v_{N} \rrbracket-(1 / 2,1 / 2)$ sends a dual polymer $\breve{X}^{\rho, x}-(1 / 2,1 / 2)$ (black dotted line) into the rectangle $\llbracket(0,0),\left(\left\lfloor\delta N^{2 / 3}\right\rfloor,\left\lfloor\delta N^{2 / 3}\right\rfloor\right) \rrbracket$.

For $z \in \mathbb{Z}_{>0}^{2}$ let $\widetilde{X}^{\rho, z} \in \mathbb{X}_{z, 0}$ be the random path that follows $\breve{X}^{\rho, z}$ from $z$ until the first time it hits the axes $\mathbb{Z}_{>0} e_{i}, i \in\{1,2\}$, and then goes to 0 taking only $-e_{1}$ or only $-e_{2}$ steps. For $A \subset \mathbb{Z}_{>0}^{2}$ let $\widetilde{Q}_{0, A}^{\rho}$ be the distribution under $\mathbf{P}$ of the paths $\left\{\widetilde{X}_{\cdot}^{\rho, z}: z \in A\right\}$. By Proposition 5.1, this is a coupling of the measures $\left\{\widetilde{Q}_{0, v}^{\rho, \mathrm{SW}}: v \in A\right\}$ and by their construction, the paths $\left\{\widetilde{X}^{\rho, z}: z \in A\right\}$ are $\widetilde{Q}_{0, A}^{\rho}$-almost surely ordered.

## 5.3 | Proofs of Theorems 2.1, 2.3, and 2.7, and Corollary 2.5

We note that the probability $\mathcal{H}_{a, b}^{\rho}\left(\Gamma^{A}\right)$ is the same as the probability under $\mathbf{P}$ that the coalescence point of the coupled paths $X^{\rho, a}$ and $X^{\rho, b}$ belongs to $A$.

Proof of Theorem 2.1. As shown in Figure 13, the duality mentioned in Remark 5.2 implies that the sampled polymer paths coalesce outside of the rectangle $\llbracket 0, v_{N} \rrbracket$ if and only if there exists some $x$ on the northeast boundary of $\llbracket 0, v_{N} \rrbracket$ such that the polymer $\widetilde{X}^{\rho, x}$ satisfies $\left|\tau_{0, x}\right| \leqslant \delta N^{2 / 3}$.

By this equivalence, the expectation in Theorem 2.1 is equal to the expectation in Theorem 4.4,

$$
\mathbb{E}\left[\mathcal{H}_{\left[\delta N^{2 / 3}\right\rfloor e_{1},\left\lfloor\delta N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \rrbracket}\right)\right]=\mathbb{E}\left[\widetilde{Q}_{0, \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}^{\rho}\left(\bigcup_{x \in \mathrm{~N}^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{\left|\widetilde{\tau}_{0, x}\right| \leqslant \delta N^{2 / 3}\right\}\right)\right] .
$$

Finally, for the exit time expectation on the right-hand side, the upper bound follows from Theorem 4.4. The lower bound follows from (4.38) and (4.3) in Theorem 4.2 because the probability of a union of events is bounded below by the maximum of the probabilities of the individual events.

Proof of Theorems 2.3 and 2.7. As shown in Figure 14, if the two sampled forward polymers starting at $\left(\left\lfloor r N^{2 / 3}\right\rfloor, 0\right)$ and $\left(0,\left\lfloor r N^{2 / 3}\right\rfloor\right)$ coalesce inside $\llbracket 0, v_{N} \rrbracket$, then by duality, this happens if and only


FIGURE 14 None of the backward polymers (black dotted lines) will enter the gray square because they are shielded away from it by the coalescing forward polymers (gray dotted lines).
for each $x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket$ the polymer $\widetilde{X}^{\rho, x}$ satisfies $\left|\tau_{0, x}\right| \geqslant r N^{2 / 3}$. Then, we have

$$
\begin{equation*}
\mathcal{H}_{\left[r N^{2 / 3}\right\rfloor e_{1},\left[r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\Gamma^{\llbracket 0, v_{N} \rrbracket}\right) \stackrel{d}{=} \widetilde{Q}_{0, \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}^{\rho}\left(\bigcap_{x \in \partial^{\mathrm{NE} \llbracket 0, v_{N} \rrbracket}}\left\{\left|\widetilde{\tau}_{0, x}\right| \geqslant r N^{2 / 3}\right\}\right) . \tag{5.7}
\end{equation*}
$$

The expectation and the tail probabilities of the right-hand side can be lower bounded using Theorem 4.3. And they are upper bounded by Theorem 4.1 because the probability of an intersection of events is bounded above by the minimum of the probabilities of the individual events.

Proof of Corollary 2.5. To prove the first inequality we will lower bound its complement. By duality, it suffices to show that for some small $q$ depending only on $\varepsilon$,

$$
\begin{align*}
\mathbb{E}\left[\mathcal{H}_{0,\left\lfloor r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \rrbracket}\right)\right] & =\mathbb{E}\left[\widetilde{Q}_{0,2 \mathrm{NE} \llbracket 0, v_{N} \mathbb{I}}^{\rho}\left(\bigcup_{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{1 \leqslant \widetilde{\tau}_{0, x} \leqslant r N^{2 / 3}\right\}\right)\right] \\
& \geqslant \mathbb{E}\left[Q_{0, v_{N}-q r N^{2 / 3} e_{2}}^{\rho}\left\{1 \leqslant \tau \leqslant r N^{2 / 3}\right\}\right] \\
& \geqslant 1-e^{-C r^{3}} . \tag{5.8}
\end{align*}
$$

The last inequality (5.8) follows from an argument similar to the proof of Lemma 4.12. Here, instead of perturbing the directional parameter, we simply perturb our end point from $v_{N}$ to $v_{N}-q r N^{2 / 3} e_{2}$. Then, as shown in Figure 15, if we fix $q$ sufficiently small, then the $-\xi[\rho]$ directed ray starting at $v_{N}-q r N^{2 / 3} e_{2}$ will hit the $e_{1}$-axis within $\llbracket a r N^{2 / 3}, b r N^{2 / 3} \rrbracket$, for some $0<a<b<1$. This again just follows from Taylor's theorem and we omit the details. Then the rest of the argument is exactly the same as in Lemma 4.12.

To prove the second inequality in the claim of the corollary we start with the following calculation, where the first equality comes from duality and the same calculation from (4.38) gives us


FIGURE 15 An illustration for the inequality (5.8). Starting from the point $v_{N}-q r N^{2 / 3} e_{2}$, the $-\xi[\rho]$-directed ray will hit the $e_{1}$-axis between $\llbracket \operatorname{ar} N^{2 / 3}, b r N^{2 / 3} \rrbracket$ for some $0<a<b<1$, provided that $q$ is fixed sufficiently small.
the inequality when we switch from $\}\} \max _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} \ldots \prime$ to $\left.\}\right\} \max _{x \notin \llbracket 0, v_{N} \rrbracket} \ldots{ }^{\prime \prime}$

$$
\begin{aligned}
\mathbb{E}\left[\mathcal{H}_{0,\left\lfloor\delta N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \rrbracket}\right)\right] & =\mathbb{E}\left[\widetilde{Q}_{0, \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}^{\rho}\left(\bigcup_{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket}\left\{1 \leqslant \widetilde{\tau}_{0, x} \leqslant \delta N^{2 / 3}\right\}\right)\right] \\
& \geqslant \mathbb{E}\left[\max _{x \in \partial^{\mathrm{NE}} \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\}\right] \\
& \geqslant \mathbb{E}\left[\max _{x \notin \llbracket 0, v_{N} \rrbracket} Q_{0, x}^{\rho}\left\{1 \leqslant \tau \leqslant \delta N^{2 / 3}\right\}\right]
\end{aligned}
$$

The last expectation can be lower bounded by $C \delta$. The proof is very similar to that of the lower bound in (4.3). More precisely, by (5.8), we can fix two constants $r_{0}$ and $N_{0}$ (depending on $\varepsilon$ ) such that, for $N \geqslant N_{0}$,

$$
\begin{equation*}
\mathbb{E}\left[Q_{0, v_{N}-q r_{0} N^{2 / 3} e_{2}+e_{1}}^{\rho}\left\{1 \leqslant \tau \leqslant r_{0} N^{2 / 3}\right\}\right] \geqslant \frac{1}{2} . \tag{5.9}
\end{equation*}
$$

Note that using the endpoint $v_{N}-q r_{0} N^{2 / 3} e_{2}+e_{1}$ instead of $v_{N}-q r_{0} N^{2 / 3} e_{2}$ does not change the proof of this lower bound.

Now, (5.9) replaces the input (4.64), and we form our partition $\left\{p_{i}\right\}$ in the range $\left[1, r_{0}\right]$ instead of $\left[-r_{0}, r_{0}\right]$. Then, the rest of the proof is the same as the lower bound proof in (4.3).

## 6 | TOTAL VARIATION DISTANCE BOUNDS

Proof of Theorem 2.8. The claim follows from the fact that if $U$ and $V$ are two random variables with distributions $\mu$ and $\nu$, respectively, and if $\mathbf{P}$ is any coupling of the two random variables, then

$$
\begin{equation*}
d_{\mathrm{TV}}(\mu, \nu) \leqslant \mathbf{P}(U \neq V) \tag{6.1}
\end{equation*}
$$

Consider the paths $X^{\rho, \delta N^{2 / 3} e_{i}}, i \in\{1,2\}$, defined in Subsection 5.2. Then, $\chi_{N}\left(X^{\rho, \delta N^{2 / 3} e_{1}}\right) \neq$ $\chi_{N}\left(X^{\rho, \delta N^{2 / 3} e_{2}}\right)$ implies the two paths did not coalesce inside $\llbracket 0, v_{N} \rrbracket$. Hence, if $\mathbf{P}$ is the probability measure from Subsection 5.2, then

$$
\mathbf{P}\left\{\chi_{N}\left(X^{\rho, \delta N^{2 / 3} e_{1}}\right) \neq \chi_{N}\left(X^{\rho, \delta N^{2 / 3} e_{2}}\right)\right\} \leqslant \mathcal{H}_{\left\lfloor\delta N^{2 / 3}\right\rfloor e_{1},\left\lfloor\delta N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\Gamma^{\mathbb{Z}^{2} \backslash \llbracket 0, v_{N} \rrbracket}\right) .
$$

Now the upper bound claimed in the theorem follows directly from Theorem 2.1.
Proof of Theorem 2.9. We will first look at $u$ only in the north boundary of $\llbracket 0, v_{N} \rrbracket$, which we denote as $\partial^{\mathrm{N}} \llbracket 0, v_{N} \rrbracket$, and we will show that

$$
\sum_{u \in \partial^{\mathrm{N}} \llbracket 0, v_{N} \rrbracket}\left|\Pi_{\left[r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\chi_{N}=u\right)-\Pi_{\left[r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N}=u\right)\right| \quad \text { is close to } 1 .
$$

A similar argument can be applied to the east boundary to show that sum is also close to 1 . And combining the two calculations for the north and east boundaries would finish the proof.

From Proposition 5.1 and Theorem 4.1,

$$
\begin{aligned}
& \mathbb{P}\left(\Pi_{\left\lfloor r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N} \in \partial^{N} \llbracket 0, v_{N} \rrbracket\right) \geqslant 1-e^{-c r^{2} N^{1 / 3}}\right) \geqslant 1-e^{-C r^{3}}, \\
& \mathbb{P}\left(\Pi_{\left[r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\chi_{N} \in \partial^{N} \llbracket 0, v_{N} \rrbracket\right) \leqslant e^{-c r^{2} N^{1 / 3}}\right) \geqslant 1-e^{-C r^{3}} .
\end{aligned}
$$

To finish the proof, on the intersection of the two events above, we have

$$
\begin{aligned}
\sum_{u \in \partial^{\mathrm{N}} \llbracket 0, v_{N} \rrbracket} & \left|\Pi_{\left\lfloor r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\chi_{N}=u\right)-\Pi_{\left\lfloor r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N}=u\right)\right| \\
& \geqslant \sum_{u \in \partial^{\mathrm{N}} \llbracket 0, v_{N} \rrbracket}\left(\Pi_{\left\lfloor r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\chi_{N}=u\right)-\Pi_{\left[r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N}=u\right)\right) \\
& =\Pi_{\left[r N^{2 / 3}\right\rfloor e_{2}}^{\rho}\left(\chi_{N} \in \partial^{N} \llbracket 0, v_{N} \rrbracket\right)-\Pi_{\left[r N^{2 / 3}\right\rfloor e_{1}}^{\rho}\left(\chi_{N} \in \partial^{N} \llbracket 0, v_{N} \rrbracket\right) \\
& \geqslant 1-2 e^{-c r^{2} N^{1 / 3}} .
\end{aligned}
$$

## 7 | TRANSVERSAL FLUCTUATION LOWER BOUND

In this section, we prove Theorem 2.11, but omit some of the details because the whole proof is similar to the proof of the upper bound in Theorem 4.2.

First, for $i \in\{1,2\}$, let us define $\left\{\operatorname{mid}_{i} \leqslant \delta N^{2 / 3}\right\}$ to be the collection of paths between $-v_{N}$ and $v_{N}$ that crosses the segment between $-\delta N^{2 / 3} e_{i}$ and $\delta N^{2 / 3} e_{i}$. As

$$
\left\{\operatorname{mid} \leqslant \delta N^{2 / 3}\right\} \subset\left\{\operatorname{mid}_{1} \leqslant \delta N^{2 / 3}\right\} \cup\left\{\operatorname{mid}_{2} \leqslant \delta N^{2 / 3}\right\},
$$

by a union bound and the symmetry between $i=1$ and 2 it suffices to prove that

$$
\mathbb{E}\left[Q_{-v_{N}, v_{N}}\left\{\operatorname{mid}_{1} \leqslant \delta N^{2 / 3}\right\}\right] \leqslant C|\log \delta|^{10} \delta .
$$

We prove this by showing that

$$
\begin{equation*}
\mathbb{P}\left(Q_{-v_{N}, v_{N}}\left\{\operatorname{mid}_{1} \leqslant \delta N^{2 / 3}\right\} \geqslant e^{-|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}}\right) \leqslant C|\log \delta|^{10} \delta . \tag{7.1}
\end{equation*}
$$

Let $r=|\log \delta|$ and fix $\alpha$ sufficiently small (now depending only on $\mu$ ) as in the proof of Theorem 4.2. The next calculation follows the same steps as (4.45), except that we now set $\rho=\mu / 2$ and consider the dark region $\mathcal{D}$ as a single point $v_{N}$.
left side of (7.1)

$$
\begin{aligned}
& =\mathbb{P}\left(\log Z_{-v_{N}, v_{N}}-\log Z_{-v_{N}, v_{N}}\left\{\operatorname{mid}_{1} \leqslant \delta N^{2 / 3}\right\} \leqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \\
& \leqslant \mathbb{P}\left(\log Z_{-v_{N}, v_{N}}\left\{\operatorname{mid}_{1} \leqslant r N^{2 / 3}\right\}-\log Z_{-v_{N}, v_{N}}\left\{\operatorname{mid}_{1} \leqslant \delta N^{2 / 3}\right\} \leqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \\
& \leqslant \mathbb{P}\left(\max _{\left.|k| \leqslant \mid r N^{2 / 3}\right\rfloor}\left[\log Z_{-v_{N}, k e_{1}}+\log Z_{(k, 1), v_{N}}\right]\right. \\
& \left.\quad-\max _{|j| \leqslant\left\lfloor\delta N^{2 / 3}\right\rfloor}\left[\log Z_{-v_{N}, k e_{1}}+\log Z_{(k, 1), v_{N}}\right] \leqslant 2|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \\
& =\mathbb{P}\left(\max _{|k| \leqslant\left\lfloor r N^{2 / 3}\right\rfloor}\left[\log \frac{Z_{-v_{N}, k e_{1}}}{Z_{-v_{N},(0,0)}}+\log \frac{Z_{(k, 1), v_{N}}}{Z_{e_{2}, v_{N}}}\right]\right.
\end{aligned}
$$

$$
\left.-\max _{1 \leqslant j \leqslant\left\lfloor\delta N^{2 / 3}\right\rfloor}\left[\log \frac{Z_{-v_{N}, j e_{1}}}{Z_{-v_{N},(0,0)}}+\log \frac{Z_{(j, 1), v_{N}}}{Z_{e_{2}, v_{N}}}\right] \leqslant 2|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right)
$$

$$
\begin{equation*}
\leqslant \mathbb{P}\left(\max _{|k| \leqslant\left\lfloor r N^{2 / 3}\right\rfloor}\left[\log \frac{Z_{-v_{N}, k e_{1}}}{Z_{-v_{N},(0,0)}}+\log \frac{Z_{(k, 1), v_{N}}}{Z_{e_{2}, v_{N}}}\right] \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \tag{7.2}
\end{equation*}
$$

$$
\begin{equation*}
+\mathbb{P}\left(\max _{1 \leqslant j \leqslant\left\lfloor\delta N^{2 / 3}\right\rfloor}\left[\log \frac{Z_{-v_{N}, j e_{1}}}{Z_{-v_{N},(0,0)}}+\log \frac{Z_{(j, 1), v_{N}}}{Z_{e_{2}, v_{N}}}\right] \geqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) . \tag{7.3}
\end{equation*}
$$

Next, let us define

$$
\widetilde{I}_{(i, 1)}^{v_{N}}=\frac{Z_{(i-1,1), v_{N}}}{Z_{(i, 1), v_{N}}}, \quad \widetilde{I}_{(i, 0)}^{-v_{N}}=\frac{Z_{-v_{N},(i, 1)}}{Z_{-v_{N},(i-1,1)}},
$$

and a two-sided multiplicative walk $\left\{M_{n}^{\prime}\right\}_{n \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket}$ by setting $M_{0}^{\prime}=1$ and

$$
M_{n}^{\prime} / M_{n-1}^{\prime}=\widetilde{I}_{(n, 0)}^{v_{N}} / \widetilde{I}_{(n, 1)}^{v_{N}} .
$$

Then, the two probabilities can be rewritten as

$$
\begin{array}{r}
(7.2)+(7.3)=\mathbb{P}\left(\max _{n \in \llbracket-\alpha r N^{2 / 3}+1, \alpha r N^{2 / 3} \rrbracket} \log M_{n}^{\prime} \leqslant 3|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) \\
+\mathbb{P}\left(\max _{n \in \llbracket 1, \delta N^{2 / 3} \rrbracket} \log M_{n}^{\prime} \geqslant|\log \delta|^{2} \sqrt{\delta} N^{1 / 3}\right) . \tag{7.4}
\end{array}
$$

Note how the right-hand side is similar to (4.48) + (4.49), except for having $M_{n}^{\prime}$ instead of $M_{n}$, and the region $\mathcal{D}$ is reduced to the single vertex $v_{N}$. Next, we give a sketch of how to carry over the estimate from the proof of Theorem 4.2 to the random walk in this proof. The essential step is to upper and lower bound the walk $M_{n}^{\prime}$ by two other walks with independent and identically distributed steps. This was done for $M_{n}$ previously in (4.59). After that, the bound on the two probabilities above comes from the same estimates as in the proof of Theorem 4.2.

First, let us summarize how the desired random walk bound was obtained in the proof of Theorem 4.2. Recall $\lambda$ and $\mu$, defined in (4.40). Lemma 4.13 showed that with probability at least $1-e^{-C r^{3}}$, for each $a, b \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket$,

$$
\frac{1}{2} \prod_{i=a}^{b} I_{(i, 1)}^{\eta, \mathrm{NE}} \leqslant \prod_{i=a}^{b} \widetilde{I}_{(i, 1)}^{v_{N}} \leqslant 2 \prod_{i=a}^{b} I_{(i, 1)}^{\lambda, \mathrm{NE}}
$$

where $I_{(i, 1)}^{\bullet, \mathrm{NE}} \sim \mathrm{Ga}^{-1}(\bullet)$. Furthermore, as stated below (4.61), there is a coupling such that the random variables

$$
\begin{equation*}
\left\{I_{(i, 1)}^{\eta, \mathrm{NE}}, I_{(j, 1)}^{\lambda, \mathrm{NE}}: i \leqslant 0, j \geqslant 1\right\} \text { are independent. } \tag{7.5}
\end{equation*}
$$

By symmetry (or rotating the picture $180^{\circ}$ ), the exact same argument can be applied to $\widetilde{I}_{(i, 0)}^{-v_{N}}$, where now these edge weights are calculated to the point $-v_{N}-\left(e_{1}+e_{2}\right)$ instead of to $v_{N}+\left(e_{1}+\right.$ $\left.e_{2}\right)$. We get that with probability at least $1-e^{-C r^{3}}$, for each $a, b \in \llbracket-\left\lfloor\alpha r N^{2 / 3}\right\rfloor+1,\left\lfloor\alpha r N^{2 / 3}\right\rfloor \rrbracket$,

$$
\frac{1}{2} \prod_{i=a}^{b} I_{(i, 0)}^{\eta, \mathrm{SW}} \leqslant \prod_{i=a}^{b} \widetilde{I}_{(i, 0)}^{-v_{N}} \leqslant 2 \prod_{i=a}^{b} I_{(i, 0)}^{\lambda, \mathrm{SW}}
$$

where $I_{(i, 0)}^{\cdot, \mathrm{SW}} \sim \mathrm{Ga}^{-1}(\cdot)$ are edge weights that are calculated to $-v_{N}-\left(e_{1}+e_{2}\right)$ and with a boundary placed on the south-west edges of the quadrant $-v_{N}-\left(e_{1}+e_{2}\right)+\mathbb{Z}_{\geqslant 0}^{2}$. As above, the random variables

$$
\begin{equation*}
\left\{I_{(i, 0)}^{\lambda, \mathrm{SW}}, I_{(j, 0)}^{\eta, \mathrm{SW}}: i \leqslant 0, j \geqslant 0\right\} \text { are independent. } \tag{7.6}
\end{equation*}
$$

Note how the parameters switched sides, as compared to (7.5).
Next, define two two-sided multiplicative random walks $M_{n}^{+}, M_{n}^{-}$with $M_{0}^{ \pm}=1$ and

$$
\begin{aligned}
& M_{n}^{+} / M_{n-1}^{+}=I_{(n, 0)}^{\lambda, \mathrm{SW}} / I_{(n, 1)}^{\eta, \mathrm{NE}} \\
& M_{n}^{-} / M_{n-1}^{-}=I_{(n, 0)}^{\eta, \mathrm{SW}} / I_{(n, 1)}^{\lambda, \mathrm{NE}}
\end{aligned}
$$

We get

$$
\frac{1}{2} M_{n}^{-} \leqslant M_{n}^{\prime} \leqslant 2 M_{n}^{+} \text {for } n \geqslant 1 \text { and } \frac{1}{2} M_{n}^{+} \leqslant M_{n}^{\prime} \leqslant 2 M_{n}^{-} \text {for } n \leqslant 0 .
$$

These bounds play the role of (4.59). With this, go back to (7.4) and follow the same argument as the one we used to bound (4.48) + (4.49), but with $M_{n}, M_{n}^{\lambda}$, and $M_{n}^{\mu}$ replaced by $M_{n}^{\prime}, M_{n}^{-}$, and $M_{n}^{+}$, respectively. We should point out that an essential fact that is used in the step analogous to (4.62) is the independence of the walks $\left\{M_{n}^{-}: n \geqslant 1\right\}$ and $\left\{M_{n}^{+}: n \leqslant 0\right\}$, which follows from (7.5) and (7.6). We omit the rest of the details.

## APPENDIX A: TECHNICAL COMPLEMENTS

## A. 1 | Moderate deviation of the bulk free energy

We present here two estimates that we use in the proof of (4.1). The first tail bound can be derived for the inverse-gamma polymer by combining [4, Theorem 1.7], which utilizes integrable probability methods, with [19, Theorem 2.2]. For the O'Connell-Yor polymer, the bound was established in [27] as Proposition 2.1 without the use of integrable probability. A proof of the bound for the inverse-gamma polymer, without the use of integrable probability, will appear in [15]. This result can be found in Theorem 4.3 .1 of the Ph.D. thesis [36].
Proposition A.1. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C$, $N_{0}$ depending on $\varepsilon$ such that for each $N \geqslant N_{0}, t \geqslant 1$, and each $\rho \in[\varepsilon, \mu-\varepsilon]$, we have

$$
\mathbb{P}\left(\log Z_{0, v_{N}}-\Lambda\left(v_{N}\right) \geqslant t N^{1 / 3}\right) \leqslant e^{-C \min \left\{t^{3 / 2}, t N^{1 / 3}\right\}} .
$$

The next tail bound is [6, Proposition 3.8]. The analogous bound for the O'Connell-Yor polymer appears as [27, Proposition 3.4].

Proposition A.2. Let $\varepsilon \in(0, \mu / 2)$. There exist positive constants $C, N_{0}$ depending on $\varepsilon$ such that for each $N \geqslant N_{0}, t \geqslant 1$ and and each $\rho \in[\varepsilon, \mu-\varepsilon]$, we have

$$
\mathbb{P}\left(\log Z_{0, v_{N}}-\Lambda\left(v_{N}\right) \leqslant-t N^{1 / 3}\right) \leqslant e^{-C \min \left\{t^{3 / 2}, t N^{1 / 3}\right\}}
$$

## A. 2 | Proof of Propositions 4.7 and 4.8

Let $\varepsilon \in(0, \mu / 2)$ and fix $\rho \in[\varepsilon, \mu-\varepsilon]$. We start with a few derivative calculations.

$$
\begin{align*}
& \left.\frac{d}{d z} \frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}\right|_{z=0}=\frac{\Psi_{2}(\rho) \Psi_{1}(\mu-\rho)+\Psi_{1}(\rho) \Psi_{2}(\mu-\rho)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}  \tag{A.1}\\
& \left.\frac{d}{d z} \frac{\Psi_{1}(\mu-\rho-z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}\right|_{z=0}=-\frac{\Psi_{2}(\rho) \Psi_{1}(\mu-\rho)+\Psi_{1}(\rho) \Psi_{2}(\mu-\rho)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}} \\
& \left.\frac{d^{2}}{d z^{2}} \frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}\right|_{z=0}=-\frac{2 \Psi_{2}(\rho)\left(\Psi_{2}(\rho)-\Psi_{2}(\mu-\rho)\right)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}+\frac{\Psi_{3}(\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \\
& +\Psi_{1}(\rho)\left(\frac{2\left(\Psi_{2}(\rho)-\Psi_{2}(\mu-\rho)\right)^{2}}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{3}}-\frac{\Psi_{3}(\mu-\rho)+\Psi_{3}(\rho)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}\right) \\
& \quad+\Psi_{1}(\mu-\rho)\left(\frac{2\left(\Psi_{2}(\rho)-\Psi_{2}(\mu-\rho)\right)^{2}}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{3}}-\frac{\Psi_{3}(\mu-\rho)+\Psi_{3}(\rho)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}\right) \\
& \left.\frac{d^{2}}{d z^{2}} \frac{\Psi_{1}(\mu-\rho-z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}\right|_{z=0}=\frac{2 \Psi_{2}(\mu-\rho)\left(\Psi_{2}(\rho)-\Psi_{2}(\mu-\rho)\right)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}+\frac{\Psi_{3}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \\
& \quad \Psi_{1}(\rho+z) \\
& \frac{d}{d z}\left(\frac{\Psi_{1}(\mu-\rho-z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)} \Psi_{0}(\mu-\rho-z)+\left.\frac{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}{\left.\Psi_{0}(\rho+z)\right)}\right|_{z=0}\right.  \tag{A.2}\\
& =\frac{\left(\Psi_{0}(\mu-\rho)-\Psi_{0}(\rho)\right)\left(\Psi_{2}(\rho) \Psi_{1}(\mu-\rho)+\Psi_{1}(\rho) \Psi_{2}(\mu-\rho)\right)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}
\end{align*}
$$

$$
\begin{aligned}
& \left.\frac{d^{2}}{d z^{2}}\left(\frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)} \Psi_{0}(\mu-\rho-z)+\frac{\Psi_{1}(\mu-\rho-z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)} \Psi_{0}(\rho+z)\right)\right|_{z=0} \\
& =\frac{2\left(\Psi_{0}(\rho) \Psi_{2}(\mu-\rho)-\Psi_{2}(\rho) \Psi_{0}(\mu-\rho)\right)\left(\Psi_{2}(\rho)-\Psi_{2}(\mu-\rho)\right)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}} \\
& \quad+\frac{\Psi_{3}(\rho) \Psi_{0}(\mu-\rho)+\Psi_{0}(\rho) \Psi_{3}(\mu-\rho)-\Psi_{2}(\rho) \Psi_{1}(\mu-\rho)-\Psi_{1}(\rho) \Psi_{2}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \\
& \quad+\left(\Psi_{1}(\rho) \Psi_{0}(\mu-\rho)+\Psi_{0}(\rho) \Psi_{1}(\mu-\rho)\right)\left(\frac{2\left(\Psi_{2}(\rho)-\Psi_{2}(\mu-\rho)\right)^{2}}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{3}}-\frac{\Psi_{3}(\mu-\rho)+\Psi_{3}(\rho)}{\left(\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)\right)^{2}}\right) .
\end{aligned}
$$

Because of the bijection in (2.3), there exists a $z$ such that

$$
\begin{equation*}
N \xi[\rho+z]=v_{N}-\left\lfloor s N^{2 / 3}\right\rfloor e_{1}+\left\lfloor s N^{2 / 3}\right\rfloor e_{2} \tag{A.3}
\end{equation*}
$$

From (A.1), we see that the derivative of $\frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}$ at $z=0$ is strictly negative. By continuity, it is also strictly negative on a neighborhood of 0 . This and the mean value theorem imply that

$$
\begin{equation*}
z \in\left[c_{1} s N^{-1 / 3}, c_{2} s N^{-1 / 3}\right] \tag{A.4}
\end{equation*}
$$

for some positive constant $c_{1}, c_{2}$ depending on $\varepsilon$.
The quantity appearing on the left side of Propositions 4.7 and 4.8 is essentially the following (we ignore the integer floor function),

$$
\begin{aligned}
& -N\left[\frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)} \Psi_{0}(\mu-\rho-z)+\frac{\Psi_{1}(\mu-\rho-z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)} \Psi_{0}(\rho+z)\right] \\
& +N\left[\frac{\Psi_{1}(\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \Psi_{0}(\mu-\rho)+\frac{\Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} \Psi_{0}(\rho)\right] \\
& +N \Psi_{0}(\mu-\rho)\left[\left(\frac{\Psi_{1}(\rho+z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}-\frac{\Psi_{1}(\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)}\right]\right. \\
& +N \Psi_{0}(\rho)\left[\frac{\Psi_{1}(\mu-\rho-z)}{\Psi_{1}(\rho+z)+\Psi_{1}(\mu-\rho-z)}-\frac{\Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)}\right] .
\end{aligned}
$$

In the above, we used (A.3) to write $\left\lfloor s N^{2 / 3}\right\rfloor=\left(v_{N}-N \xi[\rho+z]\right) \cdot e_{1}=\left(N \xi[\rho+z]-v_{N}\right) \cdot e_{2}$.
By performing Taylor expansions in $z$ and using the computations presented earlier in this section, we observe a number of cancellations, ultimately turning the above expression into

$$
\frac{N}{2} \cdot \frac{\Psi_{1}(\rho) \Psi_{2}(\mu-\rho)+\Psi_{2}(\rho) \Psi_{1}(\mu-\rho)}{\Psi_{1}(\rho)+\Psi_{1}(\mu-\rho)} z^{2}+N \cdot \mathcal{O}\left(z^{3}\right)
$$

This and (A.4) imply the claimed bounds in Propositions 4.7 and 4.8 , provided that a sufficiently small value of $c_{0}$ is chosen.

## A. 3 | Nonrandom properties

The following monotonicity property of the ratios of partition functions is in [9, Lemma A.2].

Lemma A.3. Let $x, y, z \in \mathbb{Z}^{2}$ be such that $x \cdot e_{1} \leqslant y \cdot e_{1}, x \cdot e_{2} \geqslant y \cdot e_{2}$, and $x, y \leqslant z$, then

$$
\begin{equation*}
\frac{Z_{x, z}}{Z_{x, z-e_{1}}} \leqslant \frac{Z_{y, z}}{Z_{y, z-e_{1}}} \quad \text { and } \quad \frac{Z_{x, z}}{Z_{x, z-e_{2}}} \geqslant \frac{Z_{y, z}}{Z_{y, z-e_{2}}} . \tag{A.5}
\end{equation*}
$$

The above lemma implies the following results about the monotonicity between the ratio of partition functions and exit times.

Lemma A.4. Let $z \in \mathbb{Z}_{\geqslant 0}^{2}$ and let $k, l \in \mathbb{Z}_{\geqslant 1}$ be such that $l \leqslant k$. Then

$$
\frac{Z_{0, z}(\tau \geqslant l)}{Z_{0, z-e_{1}}(\tau \geqslant l)} \leqslant \frac{Z_{0, z}(\tau \geqslant k)}{Z_{0, z-e_{1}}(\tau \geqslant k)} \quad \text { and } \quad \frac{Z_{0, z}(\tau \geqslant l)}{Z_{0, z-e_{2}}(\tau \geqslant l)} \geqslant \frac{Z_{0, z}(\tau \geqslant k)}{Z_{0, z-e_{2}}(\tau \geqslant k)} .
$$

Proof. Note that $\frac{Z_{0, z}(\tau \geqslant l)}{Z_{0, z-e_{1}}(\tau \geqslant l)}=\frac{Z_{l e_{1}, z}}{Z_{l e_{1}, z-e_{1}}}$ and $\frac{Z_{0, z}(\tau \geqslant k)}{Z_{0, z-e_{1}}(\tau \geqslant k)}=\frac{Z_{k e_{1}, z}}{Z_{k e_{1}, z-e_{1}}}$. Then Lemma A. 3 gives us the inequality

$$
\frac{Z_{l e_{1}, z}}{Z_{l e_{1}, z-e_{1}}} \leqslant \frac{Z_{k e_{1}, z}}{Z_{k e_{1}, z-e_{1}}}
$$

The other inequality with $e_{2}$ follows from a similar argument.
The next lemma is an immediate consequence of Lemma A.4. It suggests that shifting the endpoint to the right or down increases the likelihood of the polymer taking more $e_{1}$ steps at the beginning.

Lemma A.5. For any $k, l, m \in \mathbb{Z}_{\geqslant 0}$ and $x \in \mathbb{Z}_{\geqslant 0}^{2}$ such that $x+l e_{1}-m e_{2} \in \mathbb{Z}_{\geqslant 0}^{2}$,

$$
Q_{0, x}\{\tau \geqslant k\} \leqslant Q_{0, x+l e_{1}-m e_{2}}\{\tau \geqslant k\} .
$$

Proof. Note that the proof of Lemma A. 4 also gives

$$
\frac{Z_{0, x}}{Z_{0, x-e_{1}}} \leqslant \frac{Z_{0, z}(\tau \geqslant k)}{Z_{0, x-e_{1}}(\tau \geqslant k)} \quad \text { and } \quad \frac{Z_{0, x}}{Z_{0, x-e_{2}}} \geqslant \frac{Z_{0, x}(\tau \geqslant k)}{Z_{0, x-e_{2}}\left(\tau_{0, x} \geqslant k\right)} .
$$

Rearrange to get

$$
\begin{equation*}
Q_{0, x}\{\tau \geqslant k\}=\frac{Z_{0, x}(\tau \geqslant k)}{Z_{0, x}} \leqslant \frac{Z_{0, x+e_{1}}(\tau \geqslant k)}{Z_{0, x+e_{1}}}=Q_{0, x+e_{1}}\{\tau \geqslant k\} \tag{A.6}
\end{equation*}
$$

and

$$
\begin{equation*}
Q_{0, x}\{\tau \geqslant k\} \geqslant Q_{0, x+e_{2}}\{\tau \geqslant k\} . \tag{A.7}
\end{equation*}
$$

Applying the two inequalities (A.6) and (A.7) repeatedly gives us the statement of our lemma.
Fix $u \in \mathbb{Z}^{2}$, we will define a polymer with a general down-right boundary with the base at $u$. Let $\mathcal{Y}_{u}=\left\{y_{i}\right\}_{i \in \mathbb{Z}}$ be a bi-infinite downright path going through $u$. We use the convention that $y_{0}=u$ and $y_{i} \cdot e_{1} \leqslant y_{j} \cdot e_{1}$ if $i \leqslant j$.

Next, let us place positive edge weights $\left\{S_{y_{i-1}, y_{i}}\right\}$ along $\mathcal{Y}_{u}$, and we will define the following function $H$. Let $H_{u, u}=1$. For each $x_{0}=y_{m}$ for some $m>0$, define

$$
H_{u, x_{0}}=\prod_{n=1}^{m} \widetilde{Y}_{y_{n-1}, y_{n}} \quad \text { where } \widetilde{Y}_{y_{n-1}, y_{n}}= \begin{cases}S_{y_{n-1}, y_{n}} & \text { if } y_{n}-y_{n-1}=e_{1}, \\ 1 / S_{y_{n-1}, y_{n}} & \text { if } y_{n}-y_{n-1}=-e_{2} .\end{cases}
$$

For each $x_{0}=y_{-m}$ for some $m>0$, define

$$
H_{u, x_{0}}=\prod_{n=0}^{-m+1} \widetilde{Y}_{y_{n}, y_{n-1}} \quad \text { where } \widetilde{Y}_{y_{n}, y_{n-1}}= \begin{cases}1 / S_{y_{n}, y_{n-1}} & \text { if } y_{n}-y_{n-1}=e_{1} \\ S_{y_{n}, y_{n-1}} & \text { if } y_{n}-y_{n-1}=-e_{2}\end{cases}
$$

Recall $\mathcal{Y}_{u}^{\geqslant}=\cup_{n}\left(y_{n}+\mathbb{Z}_{\geqslant 0}^{2}\right)$ and $\mathcal{Y}_{u}^{>}=\cup_{n}\left(y_{n}+\mathbb{Z}_{>0}^{2}\right)$. For each $y \in \mathcal{Y}_{u}$ and $v \in \mathcal{Y}_{u}^{>}$, define the set of paths

$$
\mathbb{X}_{y, v}^{\mathcal{Y}_{u}}=\left\{x . \in \mathbb{X}_{y, v}: x_{1} \in \mathcal{Y}_{u}^{>}\right\} .
$$

This set is empty if both $y+e_{i}, i \in\{1,2\}$, are on $\mathcal{Y}_{u}$. For $v \in \mathcal{Y}_{u}^{>}$, define the partition function

$$
Z_{u, v}^{\mathcal{y}_{u}}=\sum_{y \in \mathcal{Y}_{u}} \sum_{x_{x . \in X_{y, v}^{y_{u}}}} H_{u, y} \prod_{i=1}^{|y-v|_{1}} Y_{x_{i}},
$$

where $\left\{Y_{z}\right\}$ are the bulk weights for $z \in \mathcal{Y}_{u}^{>}$. For $v \in \mathcal{Y}_{u}$ let $Z_{u, v}^{\mathcal{Y}_{u}}=H_{u, v}$. The corresponding quenched path measure will be denoted as $Q_{u, v}^{\mathcal{Y}_{u}}$. Note that these partition functions satisfy the following induction: for $w \in \mathcal{Y}_{u}^{>}$,

$$
\begin{equation*}
Z_{u, w}^{y_{u}}=\left(Z_{u, w-e_{1}}^{y_{u}}+Z_{u, w-e_{2}}^{y_{u}}\right) Y_{w} . \tag{A.8}
\end{equation*}
$$

Given a polymer model defined on $\mathcal{Y}_{u}^{\geqslant}$. We fix another bi-infinite down-right path $\mathcal{Z}_{v} \subset \mathcal{Y}_{u}^{\geqslant}$ and define the following nested polymer model rooted at $v$. It has the same bulk weights, and on the new boundary $\mathcal{Z}_{v}=\left\{z_{n}\right\}$, the weights are given by

$$
S_{z_{n-1}, z_{n}}= \begin{cases}\frac{z_{u, z_{n}}^{y_{u}}}{z_{u, z_{n-1}}^{y_{u}}} & \text { if } z_{n}-z_{n-1}=e_{1}, \\ \frac{z_{u, z_{n-1}}^{y_{u}}}{z_{u, z_{n}}^{y_{u}}} & \text { if } z_{n}-z_{n-1}=-e_{2} .\end{cases}
$$

We will denote this nested polymer measure by $Q_{v, \bullet}^{\mathcal{Z}_{v},\left(\mathcal{Y}_{u}\right)}$.
Lemma A.6. Fix $u, v \in \mathbb{Z}^{2}$ and two down-right bi-infinite paths $\mathcal{Y}_{u}$ and $\mathcal{Z}_{v}$ with $\mathcal{Z}_{v} \subset \mathcal{Y}_{u}^{\geqslant}$. Then for $w \in \mathcal{Z}_{v}^{\geqslant 0}$,

$$
\begin{equation*}
Z_{v, w}^{\mathcal{Z}_{u},\left(\mathcal{Y}_{u}\right)}=\frac{Z_{u, w}^{\mathcal{Y}_{u}}}{Z_{u, v}^{\mathcal{Y}_{u}}} . \tag{A.9}
\end{equation*}
$$

Consequently, for each $w \in \mathcal{Z}_{v}^{\geqslant 0}$ and $i \in\{1,2\}$,

$$
\begin{equation*}
\frac{Z_{u, w+e_{i}}^{\mathcal{Y}_{u}}}{Z_{u, w}^{\mathcal{Y}_{u}}}=\frac{Z_{v, w+e_{i}}^{\mathcal{Z}_{v}\left(\mathcal{Y}_{u}\right)}}{Z_{v, w}^{\mathcal{Z}_{v}\left(\mathcal{Y}_{u}\right)}} . \tag{A.10}
\end{equation*}
$$

Proof. When $w \in \mathcal{Z}_{v}$ the equality (A.9) comes straight from the definitions. Then it follows for $w \in \mathcal{Z}_{v}^{>}$because the two sides satisfy the same induction (A.8).

Lemma A.7. Fix $u, v \in \mathbb{Z}^{2}$ and two down-right bi-infinite paths $\mathcal{Y}_{u}$ and $\mathcal{Z}_{v}$ with $\mathcal{Z}_{v} \subset \mathcal{Y}_{u}^{\geqslant}$. Let $i \in\{1,2\}$ and $z \in \mathcal{Z}_{v}$ be such that $z+e_{i}$ is inside $\mathcal{Z}_{v}^{>0}$. Then, for each $w \in \mathcal{Z}_{v}^{>0}$.

$$
Q_{u, w}^{\mathcal{Y}_{u}}\left\{\text { path goes through } \llbracket z, z+e_{i} \rrbracket\right\}=Q_{v, w}^{\mathcal{Z}_{v}\left(\mathcal{Y}_{u}\right)}\left\{\text { path goes through } \llbracket z, z+e_{i} \rrbracket\right\} \text {. }
$$

Proof. We prove the case with $i=2$, the other case being symmetric. Then

$$
\begin{aligned}
& Q_{u, w}^{y_{u}}\left\{\text { path goes through the edge } \llbracket z, z+e_{2} \rrbracket\right\}=\frac{Z_{u, z}^{y_{u}} \cdot Z_{z+e_{2}, w}}{Z_{u, w}^{\mathcal{Y}_{u}}}=\frac{\frac{z_{u, z}^{y_{u}}}{Z_{u, v}^{y_{u}}} \cdot Z_{z+e_{2}, z}}{\frac{Z_{u, w}^{y_{u}}}{Z_{u, v}^{y_{u}}}} \\
& \quad=\frac{Z_{v, z}^{\mathcal{Z}_{v},\left(\mathcal{Y}_{u}\right)} \cdot Z_{z+e_{2}, z}}{Z_{v, w}^{\mathcal{Z}_{v}\left(\mathcal{Y}_{u}\right)} \quad \text { by Lemma A. } 6} \\
& =Q_{v, w}^{\mathcal{Z}_{v},\left(\mathcal{Y}_{u}\right)}\left\{\text { path goes through the edge } \llbracket z, z+e_{2} \rrbracket\right\} .
\end{aligned}
$$

See the top panel in Figure A1 for an illustration.
Next, we restrict attention to stationary polymers with southwest and antidiagonal boundaries. To simplify the notation, we will denote the respective partition functions by $Z_{u, \bullet}$ and $Z_{u, \cdot}^{\text {dia }}$. The corresponding polymer measures are denoted by $Q_{u,}$, and $Q_{u, .}^{d i a}$. For the antidiagonal boundaries, the bi-infinite paths are given by $S_{u}=u+S_{(0,0)}$, where $S_{(0,0)}$ is given in (4.10). For the nested polymers, we will always assume the outer polymer has an antidiagonal boundary, and the nested partition functions with antidiagonal and southwest boundaries are denoted, respectively, by $Z_{v, \bullet}^{(u) \text { dia }}$ and $Z_{v, \bullet}^{(u)}$. The corresponding polymer measures are denoted by $Q_{v, \bullet}^{(u)}$ and $Q_{v, \bullet}^{(u) \text { dia }}$.

The following two lemmas relate the exit times of two polymer processes with different starting points. They are illustrated on the bottom of Figure A1.

Lemma A.8. Fix two base points $(0,0)$ and $(m,-n)$ with $m, n>0$. Take $u$ with $u \leqslant(0,0)$ and $u \leqslant$ $(m,-n)$. Let $Z_{0, .}^{(u)}$ and $Z_{(m,-n), .}^{(u)}$, be the partition functions of the polymers with southwest boundaries, rooted at $(0,0)$ and $(m,-n)$, respectively, nested inside a polymer rooted at $u$ and having antidiagonal boundary $S_{u}$. Then for $v \in\left((0,0)+\mathbb{Z}_{>0}^{2}\right) \cap\left((m,-n)+\mathbb{Z}_{>0}^{2}\right)$,

$$
Q_{0, v}^{(u)}\{\tau \leqslant m\}=Q_{(m,-n), v}^{(u)}\{\tau<-n\} .
$$



FIGURE A1 Top: Illustration of Lemma A. 7 in the special case when $\mathcal{Y}_{u}$ and $\mathcal{Z}_{v}$ are southwest boundaries. Bottom: Illustration of Lemmas A. 8 and A.9. Note that any directed path between $u$ and $v$ goes through a gray edge/arrow if and only if it goes through a black edge/arrow.

Proof. This lemma follows from Lemma A. 7 as we have the equalities

$$
\begin{aligned}
& Q_{0, v}^{(u)}\{\tau \leqslant m\} \\
& =Q_{u, v}^{\text {dia }\left\{\left\{\text { path goes through edges }\left\{\left[\left[a, a+e_{2}\right]\right]: 0<a \cdot e_{1} \leqslant m \text { and } a \cdot e_{2}=0\right\}\right\} \cup\right.} \\
& \text { \{path goes through edges } \left.\left.\left\{\left[\left[a, a+e_{1}\right]\right]: 0<a \cdot e_{2} \leqslant v \cdot e_{2} \text { and } a \cdot e_{1}=0\right\}\right\}\right\} \\
& =Q_{u, v}^{\text {dia }}\left\{\text { path goes through edges }\left\{\left[\left[b, b+e_{1}\right]\right]: 0<b \cdot e_{2} \leqslant v \cdot e_{2} \text { and } b \cdot e_{1}=m\right\}\right\} \\
& =Q_{(m,-n), v}^{(u)}\{\tau<-n\} .
\end{aligned}
$$

Recall the exit time from the antidiagonal boundary, defined above (4.11).
Lemma A.9. Fix two base points $(0,0)$ and $(r, r)$ with $r \in \mathbb{Z}_{>0}$. Take $u \in-\mathbb{Z}_{>0}^{2}$. Let $Z_{0, .}^{(u)}$ and $Z_{(r, r), \text {. }}^{(u) \text { dia }}$ be the partition functions of the polymers with southwest and antidiagonal boundaries, rooted at $(0,0)$ and $(r, r)$, respectively, nested inside a polymer rooted at $u$ and having antidiagonal boundary $S_{u}$. Then for $v \in(r, r)+\mathbb{Z}_{>0}^{2}$,

$$
Q_{0, v}^{(u)}\{\tau \geqslant 2 r\}=Q_{(r, r), v}^{(u), \text { dia }}\left\{\tau^{\mathrm{dia}} \geqslant r\right\} .
$$

Proof. This lemma again follows from Lemma A. 7 as we have the equalities
$Q_{0, v}^{(u)}\{\tau \geq 2 r\}$
$=Q_{u, z}^{\text {dia }}\left\{\right.$ path goes through edges $\left\{\left[\left[a, a+e_{2}\right]\right]: 2 r \leqslant a \cdot e_{1} \leqslant v \cdot e_{1}\right.$ and $\left.\left.a \cdot e_{2}=0\right\}\right\}$
$=Q_{u, z}^{\text {dia }}\left\{\left\{\right.\right.$ path goes through edges $\left.\left\{\left[\left[(b, 2 r-b),(b, 2 r-b)+e_{1}\right]\right]: 2 r \leqslant b<v \cdot e_{1}\right\}\right\} \cup$
\{path goes through edges $\left.\left.\left\{\left[\left[(b, 2 r-b),(b, 2 r-b)+e_{2}\right]\right]: 2 r \leqslant b \leqslant v \cdot e_{1}\right\}\right\}\right\}$
$=Q_{(r, r), v}^{(u), d i a}\left\{\tau^{d i a} \geqslant r\right\}$.

## A. 4 Radon-Nikodym derivative calculation

Given $a>0, N \in \mathbb{Z}_{>0}$, and $\rho>0$, let $P^{\rho}$ denote the probability distribution on the product space $\Omega=\mathbb{R}^{\left\lfloor a N^{2 / 3}\right\rfloor}$ under which the coordinates $X_{i}(\omega)=\omega_{i}$ are independent and identically distributed $\mathrm{Ga}^{-1}(\rho)$ random variables.
Proposition A.10. Fix $\mu>0$ and $\varepsilon \in(0, \mu / 2)$. There exists a positive constant $C$ that only depends on $\varepsilon$ and $\mu$ and such that the following holds. Take any $a>0, b \in \mathbb{R}$, and $N \in \mathbb{Z}_{>0}$, and any $\rho \in$ $[\varepsilon, \mu-\varepsilon]$. Take $|b| \leqslant \frac{1}{4} \varepsilon N^{1 / 3}$ and let $f$ denote the Radon-Nikodym derivative

$$
f=\frac{d P^{\rho+b N^{-1 / 3}}}{d P^{\rho}} .
$$

Then

$$
E^{P \rho}\left[f^{2}\right] \leqslant e^{C a b^{2}}
$$

Proof. Let us denote $\lambda=\rho+b N^{-1 / 3}$. From a direct computation, we obtain

$$
\begin{align*}
E^{P \rho}\left[f^{2}\right] & =\int\left(\prod_{i=1}^{\left\lfloor a N^{2 / 3}\right\rfloor} \frac{\frac{1}{\Gamma(\lambda)} \frac{1}{\omega_{i}^{\lambda+1}} e^{-\frac{1}{\omega_{i}}}}{\frac{1}{\Gamma(\rho)} \frac{1}{\omega_{i}^{\rho+1}} e^{-\frac{1}{\omega_{i}}}} P(d \omega)\right. \\
& =\left(\frac{\Gamma(\rho)^{2}}{\Gamma(\lambda)^{2}} \frac{1}{\Gamma(\rho)} \int_{0}^{\infty} \frac{1}{x^{2 \lambda-\rho+1}} e^{-\frac{1}{x}} d x\right)^{\left\lfloor a N^{2 / 3}\right\rfloor} \\
& =\left(\frac{\Gamma(\rho) \Gamma(2 \lambda-\rho)}{\Gamma(\lambda)^{2}}\right)^{\left\lfloor a N^{2 / 3}\right\rfloor} \tag{A.11}
\end{align*}
$$

We continue by taking the logarithm of (A.11),

$$
\log (A \cdot 11)=\left\lfloor a N^{2 / 3}\right\rfloor(\log \Gamma(\rho)+\log \Gamma(2 \lambda-\rho)-2 \log \Gamma(\lambda))
$$

Note that $\rho=\lambda-b N^{-1 / 3}$ and $2 \lambda-\rho=\lambda+b N^{-1 / 3}$. We can thus assume that $b>0$, the other case being symmetric. Next, note that if we Taylor expand

$$
\begin{equation*}
\log \Gamma(\rho)+\log \Gamma(2 \lambda-\rho)-2 \log \Gamma(\lambda) \tag{A.12}
\end{equation*}
$$

then both the zeroth and the first derivative terms cancel out.

The assumption $0<b \leqslant \frac{1}{4} \varepsilon N^{1 / 3}$ implies that

$$
0<\varepsilon \leqslant \rho<\lambda<2 \lambda-\rho \leqslant \mu-\frac{\varepsilon}{2}<\mu .
$$

In addition, $\log \Gamma(\cdot)$ is a smooth function on $\mathbb{R}_{>0}$. Thus, the second derivative term and the remainder from the expansion can be upper bounded using a constant $C^{\prime}$ depending only on $\varepsilon$ and $\mu$ and we get

$$
(A .12) \leqslant C^{\prime} b^{2} N^{-2 / 3}+C^{\prime} b^{3} N^{-1} .
$$

Again, by the assumption on $b, C^{\prime} b^{2} N^{-2 / 3}+C^{\prime} b^{3} N^{-1} \leqslant(1+\varepsilon / 4) C^{\prime} b^{2} N^{-2 / 3}$. The claim follows with $C=(1+\varepsilon / 4) C^{\prime}$.

## A. 5 | Sub-exponential random variables

Let $\left\{X_{i}\right\}$ be a sequence of independent and identically distributed sub-exponential random variables with parameters $K_{0}>0$ and $\lambda_{0}>0$. This means

$$
\begin{equation*}
\mathbb{E}\left[e^{\lambda\left(X_{1}-\mathbb{E}\left[X_{1}\right]\right)}\right] \leqslant e^{K_{0} \lambda^{2}} \quad \text { for all } \lambda \in\left[0, \lambda_{0}\right] . \tag{A.13}
\end{equation*}
$$

Define $S_{0}=0$ and $S_{k}=X_{1}+\cdots+X_{k}-k \mathbb{E}\left[X_{1}\right]$ for $k \geqslant 1$. The following theorem captures the right tail behavior of the running maximum.

Theorem A.11. Assume (A.13). Then

$$
\mathbb{P}\left(\max _{0 \leqslant k \leqslant n} S_{k} \geqslant t \sqrt{n}\right) \leqslant \begin{cases}e^{-t^{2} /\left(4 K_{0}\right)} & \text { if } t \leqslant 2 \lambda_{0} K_{0} \sqrt{n}, \\ e^{-\frac{1}{2} \lambda_{0} t \sqrt{n}} & \text { if } t \geqslant 2 \lambda_{0} K_{0} \sqrt{n} .\end{cases}
$$

Proof. As $S_{k}$ is a mean zero random walk, $e^{\lambda S_{k}}$ is a nonnegative sub-martingale for $\lambda \geqslant 0$. By Doob's maximal inequality,

$$
\mathbb{P}\left(\max _{0 \leqslant k \leqslant n} S_{k} \geqslant t \sqrt{n}\right)=\mathbb{P}\left(\max _{0 \leqslant k \leqslant n} e^{\lambda S_{k}} \geqslant e^{\lambda t \sqrt{n}}\right) \leqslant \frac{\mathbb{E}\left[e^{\lambda S_{n}}\right]}{e^{\lambda t \sqrt{n}}}=\frac{\mathbb{E}\left[e^{\lambda\left(X_{1}-\mathbb{E}\left[X_{1}\right]\right)}\right]^{n}}{e^{\lambda t \sqrt{n}}} \leqslant e^{n K_{0} \lambda^{2}-\lambda t \sqrt{n}},
$$

where in the last inequality we applied (A.13), for which we now assume $\lambda \in\left[0, \lambda_{0}\right]$. On this interval, the exponent $h(\lambda)=n K_{0} \lambda^{2}-\lambda t \sqrt{n}$ is minimized at $\lambda_{t}=\min \left\{\lambda_{0}, \frac{t}{2 K_{0} \sqrt{n}}\right\}$ and

$$
h\left(\lambda_{t}\right)= \begin{cases}-\frac{t^{2}}{4 K_{0}} & \text { if } t \leqslant 2 \lambda_{0} K_{0} \sqrt{n} \\ n K_{0} \lambda_{0}^{2}-\lambda_{0} t \sqrt{n} \leqslant-\frac{1}{2} \lambda_{0} t \sqrt{n} & \text { if } t \geqslant 2 \lambda_{0} K_{0} \sqrt{n}\end{cases}
$$

The proof is complete.
Next, we verify that log gamma and log inverse gamma random variables are sub-exponential. Recall that if $X \sim \mathrm{Ga}(\alpha)$, then $\mathbb{E}[\log X]=\Psi_{0}(\alpha)$, where $\Psi_{0}$ is the digamma function, that is, $\Psi_{0}(\alpha)=(\log \Gamma(\alpha))^{\prime}$.

Proposition A.12. Fix $\varepsilon \in(0, \mu / 2)$. There exist positive constants $K_{0}, \lambda_{0}$ depending on $\varepsilon$ such that for each $\alpha \in[\varepsilon, \mu-\varepsilon]$ and $X \sim \mathrm{Ga}(\alpha)$, we have

$$
\mathbb{E}\left[e^{\lambda\left(\log X-\Psi_{0}(\alpha)\right)}\right] \leqslant e^{K_{0} \lambda^{2}} \quad \text { for all } \lambda \in\left[-\lambda_{0}, \lambda_{0}\right] .
$$

Proof. First, note that $\mathbb{E}\left[X^{\lambda}\right]=\frac{\Gamma(\alpha+\lambda)}{\Gamma(\alpha)}$, provided that $\alpha+\lambda>0$. This last condition can be guaranteed for all $\alpha>\varepsilon$ by taking $\lambda_{0}$ small enough (depending on $\varepsilon$ ). Then, by Taylor's theorem,

$$
\begin{aligned}
\log \mathbb{E}\left[e^{\lambda\left(\log X-\Psi_{0}(\alpha)\right)}\right] & =\log \left(\mathbb{E}\left[X^{\lambda}\right] e^{-\lambda \Psi_{0}(\alpha)}\right)=\log \Gamma(\alpha+\lambda)-\log \Gamma(\alpha)-\lambda \Psi_{0}(\alpha) \\
& =\Psi_{1}(\alpha) \frac{\lambda^{2}}{2}+o\left(\lambda^{2}\right) \leqslant K_{0} \lambda^{2},
\end{aligned}
$$

provided $\lambda_{0}$ is taken sufficiently small depending on $\varepsilon$. The constant $K_{0}$ can be chosen to not depend on $\alpha \in[\varepsilon, \mu-\varepsilon]$ because $\Psi_{1}$ is a smooth function on $\mathbb{R}_{>0}$.

## A. 6 | Random walk estimates

Let $\left\{X_{i}\right\}_{i \in \mathbb{Z}_{>0}}$ be an independent and identically distributed sequence of random variables with

$$
\mathbb{E}\left[X_{i}\right]=\mu, \quad \operatorname{Var}\left[X_{i}\right]=1 \quad \text { and } \quad \mathbb{E}\left[\left|X_{i}-\mu\right|^{3}\right]=c_{3}<\infty .
$$

Define $S_{k}=\sum_{i=1}^{k} X_{i}$ for $k \geqslant 1$. We have the following proposition that bounds the probability that the running maximum of a random walk is small.
Proposition A.13. There exists a positive constant $C$ such that for any $l>0$, we have

$$
\begin{equation*}
\mathbb{P}\left(\max _{1 \leqslant k \leqslant N} S_{k}<l\right) \leqslant C\left(c_{3} l+c_{3}^{2}\right)(|\mu|+1 / \sqrt{N}) . \tag{A.14}
\end{equation*}
$$

This result follows directly from the following two results from [29].
Lemma A. 14 [29] Lemma 5. There exists an absolute constant $C$ such that for any $l>0$

$$
\begin{equation*}
\mathbb{P}\left(\max _{1 \leqslant k \leqslant N} S_{k}<l\right)-\mathbb{P}\left(\max _{1 \leqslant k \leqslant N} S_{k}<0\right) \leqslant C\left(c_{3} l+c_{3}^{2}\right)(|\mu|+1 / \sqrt{N}) . \tag{A.15}
\end{equation*}
$$

Lemma A. 15 [29] Lemma 7. There exists an absolute constant C such that

$$
\begin{equation*}
\mathbb{P}\left(\max _{1 \leqslant k \leqslant N} S_{k}<0\right) \leqslant C c_{3}^{2}(|\mu|+1 / \sqrt{N}) \tag{A.16}
\end{equation*}
$$

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