

A PATCHWORK QUILT SEWN FROM BROWNIAN FABRIC: REGULARITY OF POLYMER WEIGHT PROFILES IN BROWNIAN LAST PASSAGE PERCOLATION

ALAN HAMMOND

ABSTRACT. In last passage percolation models lying in the KPZ universality class, the energy of long energy-maximizing paths may be studied as a function of the paths' pair of endpoint locations. Scaled coordinates may be introduced, so that these maximizing paths, or polymers, now cross unit distances with unit-order fluctuations, and have scaled energy, or weight, of unit order. In this article, we consider Brownian last passage percolation in these scaled coordinates. In the narrow wedge case, when one endpoint of such polymers is fixed, say at $(0, 0) \in \mathbb{R}^2$, and the other is varied horizontally, over $(z, 1)$, $z \in \mathbb{R}$, the polymer weight profile as a function of $z \in \mathbb{R}$ is locally Brownian; indeed, by [Ham17a, Theorem 2.11 and Proposition 2.5], the law of the profile is known to enjoy a very strong comparison to Brownian bridge on a given compact interval, with a Radon-Nikodym derivative in every L^p space for $p \in (1, \infty)$, uniformly in the scaling parameter, provided that an affine adjustment is made to the weight profile before the comparison is made. In this article, we generalize this narrow wedge case and study polymer weight profiles begun from a very general initial condition. We prove that the profiles on a compact interval resemble Brownian bridge in a uniform sense: splitting the compact interval into a random but controlled number of patches, the profile in each patch after affine adjustment has a Radon-Nikodym derivative that lies in every L^p space for $p \in (1, 3)$. This result is proved by harnessing an understanding of the uniform coalescence structure in the field of polymers developed in [Ham17c] using techniques from [Ham17a] and [Ham17b].

1. Introduction	2
1.1. KPZ universality	2
1.2. Probabilistic problems and proof techniques for KPZ	3
1.3. Brownian last passage percolation	5
1.4. Polymer weight profiles from general initial data	6
1.5. Patchwork quilts sewn from Brownian pieces of fabric	6
1.6. The main result: a generic weight profile is a Brownian patchwork quilt	9
2. A geometric view: staircases, zigzags and polymers	11
3. Polymer forests: a rough guide to the proof of the main result	13
4. Important tools	16
4.1. A comment about explicit constants	16
4.2. The rarity of many disjoint polymers	17
4.3. Narrow-wedge polymer weight profiles bear strong comparison to Brownian bridge	18
4.4. Basic results on polymers and weights	19
4.5. Polymer fluctuation	20
4.6. Control on the fluctuation of line-to-point polymers	22
4.7. Conventions governing the presentation of upcoming proofs	22
5. The rough guide elaborated: how rerooting will be carried out	23
6. Rarity of late coalescence	27
7. Polymer coalescence, canopies and intra-canopy weight profiles	30
8. Well-behaved canopy structures are typical	34
9. Discovering the Brownian patchwork quilt: the derivation of the main result	39

Appendix A. Polymer uniqueness	45
Appendix B. Computational derivations	50
B.1. A short piece of working in the proof of Proposition 6.1	51
B.2. Proposition 6.1: derivation	51
B.3. Lemma 8.1: derivation	56
References	69

Contents

1. INTRODUCTION

1.1. KPZ universality. Consider a discrete model of random growth in which the initially healthy integer lattice sites in the upper half-plane become infected. At time zero, a certain subset of such sites on the x -axis are infected. At any given positive integer time, one uninfected site in the upper half-plane that is a nearest neighbour of a presently infected site is selected, uniformly at random. The site then becomes infected, with a ‘transmission’ edge being added from the newly infected site into the already infected set, this edge selected uniformly from the available possibilities.

In this way, the infected region grows, one site at a time. As Figure 1 illustrates, this region is at any given moment the collection of vertices abutting the present set of transmission edges. The transmission edge-set is partitioned into a collection of trees, each rooted at one of the initially infected sites.

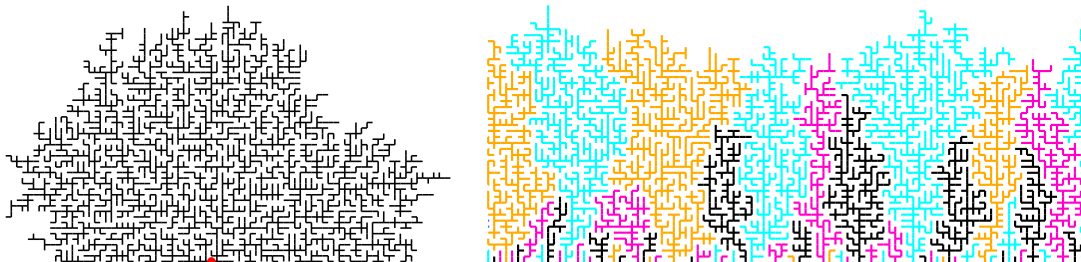


FIGURE 1. The Eden-like infection model growing from two initial conditions. On the left, only the origin is initially infected. On the right, every integer lattice site on the x -axis is so infected.

The $1 + 1$ dimensional Kardar-Parisi-Zhang (KPZ) universality class includes a wide range of interface models suspended over a one-dimensional domain, in which growth in a direction normal to the surface competes with a smoothing surface tension in the presence of a local randomizing force that roughens the surface. The infection model, which is a variant of the Eden model introduced in [Ede61], is expected to lie in the KPZ class. The boundary of the infected region is an interface. When the infected region has diameter of order n , the interface height above a given point has, according to KPZ prediction, typical deviation from the mean of order $n^{1/3}$, while non-trivial correlations in this height as the horizontal coordinate is varied are encountered on scale $n^{2/3}$. Moreover, an exponent of one-half dictates the interface’s regularity, with the interface height being expected to vary between a pair of locations at distance of order at most $n^{2/3}$ on the order of the square root of the distance between these locations.

The broad range of interface models that are rigorously known or expected to lie in the KPZ universality class includes many last passage percolation models, in which the interface models the maximum obtainable value of paths assigned value by integrating over a product measure random environment. The scaling assertions associated to these values of $(1/3, 1/2, 2/3)$ have been rigorously demonstrated for only a few random growth models, each of which enjoys an integrable structure: for example, the seminal work of Baik, Deift and Johansson [BDJ99] rigorously established the one-third exponent, and moreover obtained the GUE Tracy-Widom distributional limit, for the case of Poissonian last passage percolation, while the two-thirds power law for transversal fluctuation was derived for this model by Johansson [Joh00].

1.2. Probabilistic problems and proof techniques for KPZ. The theory of KPZ universality has advanced through physical insights, numerical analysis, and several techniques of integrable or algebraic origin. We will not hazard a summary of literature to support this one-sentence history, but refer to the reader to [Cor12] for a KPZ survey from 2012; in fact, integrable and analytic approaches to KPZ have attracted great interest around and since that time. Now, it is hardly deep or controversial to say that many problems and models in KPZ are intrinsically random: the random interface of the infected region boundary in the two simulations in Figure 1, and the geometry of the randomly evolving forest composed of transmission edges, are two important characters. It would thus seem valuable to approach the problems of KPZ universality from a predominately probabilistic perspective.

The present article is the culmination of a four-paper study of KPZ from such a viewpoint. The companion papers are [Ham17a], [Ham17b] and [Ham17c]. Our main result, Theorem 1.2, draws on important technical ingredients from the other works in order to make a significant probabilistic inference about KPZ universality.

In order to explain the problem in question, we first develop our discussion of KPZ growth. A growth model in the KPZ universality class may be started at time zero from many different initial profiles. One very special case, in which growth is initiated from a unique point, goes by the name ‘narrow wedge’. This corresponds to the growth from a single infected site, the origin, seen in the left sketch of Figure 1. In this case, the limiting description of the late time interface, suitably scaled in light of the one-third and two-thirds powers and up to the subtraction of a parabola, is offered by the Airy_2 process, which is a random function $\mathcal{A} : \mathbb{R} \rightarrow \mathbb{R}$, whose finite dimensional distributions are specified by Fredholm determinants, that was introduced by [PS02]. The one-half power law for interface regularity is expressed by the Hölder-1/2--continuity of \mathcal{A} . It has been anticipated that \mathcal{A} has a locally Brownian structure. Such a statement as this may be interpreted either by taking a *local* limit, in which, for given $x \in \mathbb{R}$, the Gaussianity of $\epsilon^{-1/2}(\mathcal{A}(x+\epsilon) - \mathcal{A}(x))$ is investigated after the low ϵ limit is taken. An in essence stronger, and much more useful, assertion of Brownian structure would concern a *unit-order* scale: examples would be the absolute continuity of $[x, x+K] : y \rightarrow \mathcal{A}(y) - \mathcal{A}(x)$ with respect to Brownian motion (whose rate is two in view of the convention of definition for \mathcal{A}) on any compact interval $[x, x+K]$; or, stronger still, that the resulting Radon-Nikodym derivative lies in L^p -spaces for values of p exceeding one.

Both types of Brownian comparison have been made for the Airy_2 process. Finite dimensional distributional convergence in the local limit has been proved in [Häg08]; this conclusion is strengthened to a functional convergence in [CP15, Theorem 3] by means of stochastically sandwiching the process between variants of the process at equilibrium. Regarding unit-order comparison, the process, after the subtraction of a suitable parabola, may be embedded as the top curve of an \mathbb{N} -indexed ordered ensemble of curves that are in effect mutually avoiding Brownian bridges. Such ensembles

of random curves satisfy a simple ‘Brownian Gibbs’ resampling property which is a very valuable tool for their analysis. Indeed, the tool of Brownian Gibbs resampling was used in [CH14] to prove the absolute continuity statement mentioned in the preceding paragraph. These resampling ideas have been refined in the first article [Ham17a] in our four-paper study to prove a result, [Ham17a, Theorem 1.9], which asserts that the Radon-Nikodym derivative (of the Airy_2 process with respect to Brownian motion on a compact interval) lies in all L^p spaces, for $p \in (1, \infty)$. Actually, there is one caveat: the comparison of the two processes is made only after an affine shift, so that comparison is made with Brownian bridge.

Growth may be initiated from a much more general initial condition than in this narrow wedge case. For example, in Figure 1(right), each site on the x -axis is initially infected. This corresponds to growth from a flat (or zero) initial condition.

For initial conditions that grow at most linearly, it has been anticipated that a limiting description of the suitably scaled late-time interface should exist in these cases also. Indeed, in a recent preprint [MQR17], Matetski, Quastel and Remenik have utilized a biorthogonal ensemble representation found by [Sas05, BFPS07] associated to the totally asymmetric exclusion process in order to find Fredholm determinant formulas for the multi-point distribution of the height function of this growth process begun from an arbitrary initial condition. Using these formulas to take the KPZ scaling limit, the authors construct a scale invariant Markov process that lies at the heart of the KPZ universality class. The time-one evolution of this Markov process may be applied to such general initial data as we have mentioned, and the result is the scaled profile begun from such data, which generalizes the Airy_2 process seen in the narrow wedge case. It is natural to ask what form of local regularity this profile enjoys: is it Hölder-1/2– continuous, and what comparison to Brownian motion can be made? Theorem 4.4 in [MQR17] asserts such Hölder continuity, and also makes a local limit Brownian comparison, proving that there is convergence to Brownian motion (with rate two) in finite dimensional distributions in this limit. Such local limit results for general initial condition profiles have also been derived by [Pim17], for geometric last passage percolation models.

The distinction between asserting Brownian structure in a local limit, and doing so on a unit-order scale, is an important one. The aim of proving [Joh03, Conjecture 1.5] illustrates the difference. Johansson’s conjecture states that the Airy_2 process after subtraction of the parabola x^2 has a unique maximizer; it is important because the maximizer describes the scaled location of the endpoint of a point-to-line maximizing path in last passage percolation models. Local limit Gaussianity does not rule out the presence of two or more maximizers, but the result is a direct consequence [CH14, Theorem 4.3] of unit-order Brownian comparison for the Airy_2 process. (The conjecture in fact has several proofs: Moreno Flores, Quastel and Remenik [MFQR13] via an explicit formula for the maximizer, and an argument of Pimentel [Pim14] showing that any stationary process minus a parabola has a unique maximizer.) Moreover, stronger assertions of unit-order Brownian structure, involving finiteness for higher L^p -norms of the Radon-Nikodym derivative, imply that Brownian motion (or bridge) characteristics obtain on the unit scale in a very strong sense (see [Ham17a, Corollary 1.10]).

The problem of unit-order scale, rather than local limit, Brownian comparison for scaled height profiles begun from general initial data, is an important example of an intrinsically probabilistic question in the theory of KPZ universality. In this article, we study it, for almost arbitrary initial data.

An idea at the heart of our approach is represented by the system of trees seen in the right sketch of Figure 1. (This is only a conceptual connection: the infection model illustrates ideas, and is not being investigated here.) As the trees grow, they compete, with the interface at any given time being partitioned into canopies of presently surviving trees. One important aspect of forest geometry is that it may be expected to respect KPZ scaling, so that a tree surviving when growth has advanced on order n has a canopy of length of order $n^{2/3}$. In the companion papers of our four paper study, such a scaling for forest geometry has been proved. Here we exploit this understanding to study scaled interfaces.

Although our approach is predominately probabilistic, it does rely crucially on certain limited integrable inputs (which are lacking for example in the infection model). The model that we choose for study is Brownian last passage percolation, a model in the KPZ universality class that enjoys attractive probabilistic (and integrable) features. Our principal conclusion, Theorem 1.2, makes a strong Brownian comparison for scaled interface profiles uniformly over all high choices of the length scale parameter for these microscopic models.

We now prepare to state this principal conclusion, first specifying the model under study.

1.3. Brownian last passage percolation. This model was introduced in [OY02]; we will call it Brownian LPP. Let $B : \mathbb{Z} \times \mathbb{R} \rightarrow \mathbb{R}$ denote an ensemble of independent two-sided standard Brownian motions $B(k, \cdot) : \mathbb{R} \rightarrow \mathbb{R}$, $k \in \mathbb{Z}$.

Let $i, j \in \mathbb{Z}$ with $i \leq j$. We denote the integer interval $\{i, \dots, j\}$ by $\llbracket i, j \rrbracket$. Further let $x, y \in \mathbb{R}$ with $x \leq y$. With these parameters given, we consider the collection of non-decreasing lists $\{z_k : k \in \llbracket i+1, j \rrbracket\}$ of values $z_k \in [x, y]$. With the convention that $z_i = x$ and $z_{j+1} = y$, we associate an energy to any such list, namely $\sum_{k=i}^j (B(k, z_{k+1}) - B(k, z_k))$. We may then define the maximum energy, $M_{(x,i) \rightarrow (y,j)}^1$, to be the supremum of the energies of all such lists.

The one-third and two-thirds KPZ scaling considerations that we have outlined are manifest in Brownian LPP. When the ending height j exceeds the starting height i by a large positive integer n , and the location y exceeds x also by n , then the maximum energy grows linearly, at rate $2n$, and has a fluctuation about this mean of order $n^{1/3}$. Moreover, if y is permitted to vary from this location, then it is changes of $n^{2/3}$ in its value that result in a non-trivial correlation of the maximum energy from its original value.

These facts prompt us to introduce scaled coordinates to describe the two endpoint locations, and a notion of scaled maximum energy, which we will refer to as weight. Let n be an element in the positive integers \mathbb{N} , and suppose that $x, y \in \mathbb{R}$ satisfy $y \geq x - 2^{-1}n^{1/3}$. Define

$$\text{Wgt}_{n;(x,0)}^{(y,1)} = 2^{-1/2}n^{-1/3} \left(M_{(2n^{2/3}x,0) \rightarrow (n+2n^{2/3}y,n)}^1 - 2n - 2n^{2/3}(y-x) \right). \quad (1)$$

Consistently with the facts just mentioned, the quantity $\text{Wgt}_{n;(x,0)}^{(y,1)}$ may be expected to be, for given real choices of x and y , a unit-order random quantity, whose law is tight in the scaling parameter $n \in \mathbb{N}$. The quantity describes, in units chosen to achieve this tightness, the maximum possible energy associated to journeys which in the original coordinates occur between $(2n^{2/3}x, 0)$ and $(n + 2n^{2/3}y, n)$. In scaled coordinates, this is a journey between $(x, 0)$ and $(y, 1)$. We view the first coordinate as space and the second as time, so this journey is between x and y over the unit time interval $[0, 1]$.

Underlying this definition is a geometric picture of scaled maximizing paths, or polymers, that achieve these weight values. This picture will be central to our study, and we will explain it shortly, in a way that may serve to further explain the above definition.

1.4. Polymer weight profiles from general initial data. For now, we continue on a rather direct route to stating our principal conclusion. The random function $y \rightarrow \text{Wgt}_{n;(0,0)}^{(y,1)}$ may be viewed as the weight profile obtained by scaled maximizing paths that travel from the origin at time zero to the variable location y at time one. This insistence that the paths must begin at the origin (it is this case that is called the narrow wedge by physicists) is of course rather special. We now make a more general definition, of the f -rewarded line-to-point polymer weight $\text{Wgt}_{n;(*:f,0)}^{(y,1)}$. Here, f is an initial condition, defined on the real line. Paths may begin anywhere on the real line at time zero; they travel to $y \in \mathbb{R}$ at time one. (Because they are free at the beginning and fixed at the end, we refer to these paths as ‘line-to-point’.) They begin with a reward given by evaluating f at the starting location, and then gain the weight associated to the journey they make. The value $\text{Wgt}_{n;(*:f,0)}^{(y,1)}$, which we will define momentarily, denotes the maximum f -rewarded weight of all such paths. In the notation $\text{Wgt}_{n;(*:f,0)}^{(y,1)}$, we again use subscript and superscript expressions to refer to space-time pairs of starting and ending locations. The starting spatial location is being denoted $* : f$. The star is intended to refer to the free time-zero endpoint, which may be varied, and the $: f$ to the reward offered according to where this endpoint is placed.

The next definition specifies essentially the broadest class of f suitable for a study of the weight profiles $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$ for all sufficiently high $n \in \mathbb{N}$.

Definition 1.1. Writing $\bar{\Psi} = (\Psi_1, \Psi_2, \Psi_3) \in (0, \infty)^3$ for a triple of positive reals, we let $\mathcal{I}_{\bar{\Psi}}$ denote the set of measurable functions $f : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ such that $f(x) \leq \Psi_1(1 + |x|)$ and $\sup_{x \in [-\Psi_2, \Psi_2]} f(x) > -\Psi_3$.

For f lying in one of the function spaces $\mathcal{I}_{\bar{\Psi}}$, we now formally define the f -rewarded line-to-point polymer weight $\text{Wgt}_{n;(*:f,0)}^{(y,1)}$ to be

$$\sup_{x \in (-\infty, 2^{-1}n^{1/3} + y]} (\text{Wgt}_{n;(x,0)}^{(y,1)} + f(x)).$$

Our principal conclusion, Theorem 1.2, roughly asserts that the weight profiles $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$, viewed as functions of y in a given compact real interval, enjoy a uniformly strong similarity with Brownian motion, even as the parameters n and f are permitted to vary over all sufficiently high integer values and over the function space $\mathcal{I}_{\bar{\Psi}}$. The phrase ‘uniformly strong similarity’ is a simplification, however, and, in the next section, we explain what form of comparison we will make.

1.5. Patchwork quilts sewn from Brownian pieces of fabric. In essence, our theorem will assert that each weight profile is a Brownian patchwork quilt, with a uniform efficiency in manufacture. Now we define what we mean by this new concept.

1.5.1. Continuous paths, bridges, and the projection between them. For $a, b \in \mathbb{R}$ with $a \leq b$, let $\mathcal{C}_{*,*}([a, b], \mathbb{R})$ denote the space of continuous real-valued functions of the interval $[a, b]$. The use of the pair of subscript stars is intended to indicate that there is no restriction placed on the endpoint values of the member functions. An element f of $\mathcal{C}_{*,*}([a, b], \mathbb{R})$ that vanishes at the endpoints of

$[a, b]$ is here called a bridge. Denote by $\mathcal{C}_{0,0}([a, b], \mathbb{R})$ the collection of bridges. A natural projection maps the first space onto the second, sending a continuous function $f : [a, b] \rightarrow \mathbb{R}$ to the bridge $f(x) - (b-a)^{-1}((b-x)f(a) + (x-a)f(b))$. The latter bridge will be denoted by $f^{[a,b]} : [a, b] \rightarrow \mathbb{R}$. The notation extends to stochastic processes. If X is a $\mathcal{C}_{*,*}([a, b], \mathbb{R})$ -valued random process defined under the law \mathbb{P} , then its bridge projection $X^{[a,b]}$ is a $\mathcal{C}_{0,0}([a, b], \mathbb{R})$ -valued process under the same measure. Any law ν on $\mathcal{C}_{*,*}([a, b], \mathbb{R})$ naturally induces a push-forward law on $\mathcal{C}_{0,0}([a, b], \mathbb{R})$. When ν is the law of standard Brownian motion $X : [a, b] \rightarrow \mathbb{R}$ with $X(a) = 0$, the bridge-valued push forward measure, which is the law of standard Brownian bridge, will be denoted by $\mathcal{B}_{0,0}^{[a,b]}$.

1.5.2. *Patchwork quilts.* How to sew a patchwork quilt? By cutting several pieces of fabric and stitching them together.

Let $a, b \in \mathbb{R}$ with $a < b$ and let $k \in \mathbb{N}$. We aim to form a patchwork quilt, which will be a continuous function $q : [a, b] \rightarrow \mathbb{R}$. Our raw material consists of k pieces of fabric, each a continuous real-valued function on $[a, b]$. This data is listed as the *fabric* sequence: $f_i : [a, b] \rightarrow \mathbb{R}$ for $1 \leq i \leq k$. It is our intention to sew together the consecutive pieces of fabric at a given set of $k-1$ locations in $[a, b]$. We call these locations the *stitch* points, and list them in increasing order $a \leq s_1 < s_2 < \dots < s_{k-1} \leq b$. In this way, the interval $[a, b]$ is divided into k *patches*. With the convention that $a = s_0$ and $b = s_k$, the i^{th} patch is $[s_{i-1}, s_i]$. The i^{th} piece of fabric is cut at the two ends of the i^{th} patch, and the resulting pieces are displaced vertically, in order that they meet at their endpoints, and stitched together. The resulting *patchwork quilt* is the continuous function $q : [a, b] \rightarrow \mathbb{R}$ given by $q(x) = f_i(x) + v_i$ for $x \in [s_{i-1}, s_i]$ and $1 \leq i \leq k$. Here, v_i , $1 \leq i \leq k$, are the vertical shifts: we demand that $v_0 = 0$ (so that $q(a) = f_1(a)$), and determine the later v_i values by insisting on the continuity of q . See Figure 2.

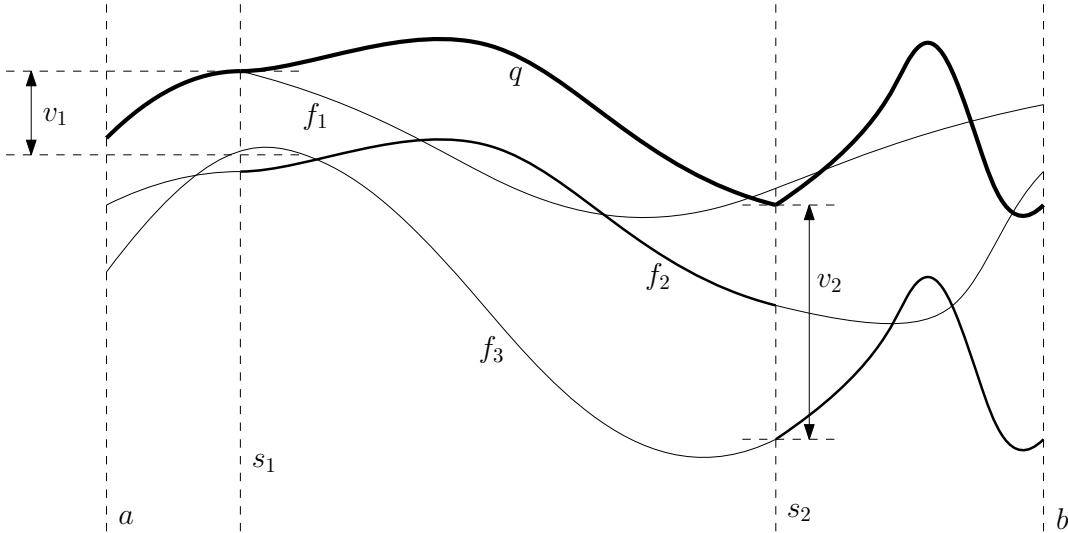


FIGURE 2. A quilt $q : [a, b] \rightarrow \mathbb{R}$ is formed from three pieces of fabric.

1.5.3. *Random continuous curves that uniformly withstand comparison to Brownian bridge.* Let $X : [a, b] \rightarrow \mathbb{R}$ denote a random continuous function defined on a probability space carrying the law \mathbb{P} ; thus, X is a $\mathcal{C}_{*,*}([a, b], \mathbb{R})$ -valued random variable. The process $X^{[a,b]}$ is thus a random bridge on $[a, b]$ whose law may be compared to that of Brownian bridge. For $\beta \in (0, \infty)$, we say that X

withstands L^{β^-} -comparison to Brownian bridge if the Radon-Nikodym derivative of the law of $X^{[a,b]}$ with respect to $\mathcal{B}_{0,0}^{[a,b]}$ has a finite L^η -norm, for every $\eta \in (0, \beta)$.

Suppose now that we instead consider a collection of random continuous functions, each defined under the law \mathbb{P} . The collection is said to *uniformly withstand L^{β^-} -comparison to Brownian bridge* if the above L^η -norm, for any given $\eta \in (0, \beta)$, is bounded above by a finite quantity that is independent of the choice of X in the collection.

1.5.4. *Sewing a random patchwork quilt.* On a probability space with measure \mathbb{P} , suppose given a *fabric* sequence $\{F_i : i \in \mathbb{N}\}$ of random continuous functions defined on $[a, b]$. Suppose further that an almost surely finite random *stitch points* set S satisfying $S \subset [a, b]$ is defined on this probability space.

The patchwork quilt formed by this fabric sequence and stitch point set is itself a random continuous function, defined under the law \mathbb{P} , that we will denote by $\text{Quilt}[\overline{F}, S]$, (where the bar notation indicates a vector, in this case, indexed by \mathbb{N}). At almost every sample point in the probability space, S is finite; at such a point, the quilt is formed from the first $|S| + 1$ fabric sequence elements, and the later elements play no role.

1.5.5. *When random functions may be formed as Brownian patchwork quilts with uniform efficiency.* Let \mathcal{A} be an arbitrary index set. Suppose given a collection of random continuous functions $X_{n,\alpha} : [a, b] \rightarrow \mathbb{R}$, indexed by $(n, \alpha) \in \mathbb{N} \times \mathcal{A}$, defined under the law \mathbb{P} .

Let $\beta_1 > 0$, $\beta_2 \geq 1$ and $\beta_3 > 0$. This collection is said to be uniformly Brownian patchwork $(\beta_1, \beta_2, \beta_3)$ -quiltable if

- if there exist sequences $p, q : \mathbb{N} \rightarrow [0, 1]$ verifying $p_j \leq j^{-\beta_1 + \epsilon}$ and $q_j \leq j^{-\beta_3 + \epsilon}$ for each $\epsilon > 0$ and all j sufficiently high

such that, for each $(n, \alpha) \in \mathbb{N} \times \mathcal{A}$, we may construct under the law \mathbb{P} ,

- (1) an error event $E_{n,\alpha}$ that satisfies $\mathbb{P}(E_{n,\alpha}) \leq q_n$;
- (2) a *fabric* sequence $\overline{F}_{n,\alpha} = \{F_{n,\alpha;i} : i \in \mathbb{N}\}$ (consisting of continuous random functions on $[a, b]$), where the collection $\{F_{n,\alpha;i} : (n, \alpha, i) \in \mathbb{N} \times \mathcal{A} \times \mathbb{N}\}$ *uniformly* withstands $L^{\beta_2^-}$ -comparison to Brownian bridge;
- (3) a *stitch points* set $S_{n,\alpha} \subset [a, b]$ whose cardinality verifies $\mathbb{P}(|S_{n,\alpha}| \geq \ell) \leq p_\ell$ for each $\ell \in \mathbb{N}$;
- (4) and all this in such a way that, for every $(n, \alpha) \in \mathbb{N} \times \mathcal{A}$, the random function $X_{n,\alpha}$ is equal to the patchwork quilt $\text{Quilt}[\overline{F}_{n,\alpha}, S_{n,\alpha}]$ throughout the interval $[a, b]$, whenever the error event $E_{n,\alpha}$ does *not* occur.

Uniformly sewn Brownian patchwork quilts manifest in a reasonably strong sense the notion of unit-order scale Brownian comparison that we discussed in the article's opening paragraphs. A simple example shows, however, that more than a merely abstract tool must be used to remove a given stitch from a quilt. Indeed, consider standard Brownian motion $B : [0, 1] \rightarrow \mathbb{R}$, and let $t : [0, 1] \rightarrow [0, 1]$ denote the tent map that affinely interpolates the points $t(0) = 0$, $t(1/4) = 1$, $t(1/2) = 0$ and $t(1) = 0$. Let H denote the random value such that the modified process $B + tH : [0, 1] \rightarrow \mathbb{R}$ achieves the same maximum value on $[0, 1/2]$ as it does on $[1/2, 1]$. The modified process is absolutely continuous with respect to a suitable vertical shift of Brownian motion on each of the intervals $[0, 1/2]$ and $[1/2, 1]$. However, because B has an almost surely unique maximizer on $[0, 1]$, whereas

$B + tU$ almost surely has two maximizers, the modified process is singular with respect to Brownian motion on $[0, 1]$. That is, there is a quilt description of the modified process with a stitch sewn at one-half, and there is no means of undoing that stitch.

1.6. The main result: a generic weight profile is a Brownian patchwork quilt.

Theorem 1.2. *Let $\bar{\Psi} \in (0, \infty)^3$ satisfy $\Psi_2 \geq 1$. The collection of random continuous functions*

$$[-1, 1] \rightarrow \mathbb{R} : y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$$

indexed by $(n, f) \in 2\mathbb{N} \times \mathcal{I}_{\bar{\Psi}}$ is uniformly Brownian patchwork $(2, 3, 1/252)$ -quiltable.

In order to summarise what has been achieved in proving this theorem, and in what sense further progress may be possible, it is useful to formulate a conjecture about an even stronger Brownian regularity of the weight profiles. For $a \leq b$, we may write $\mathcal{C}_{0,*}([a, b], \mathbb{R})$ for the space of continuous $f : [a, b] \rightarrow \mathbb{R}$ with $f(a) = 0$. We write $\mathcal{B}_{0,*}^{[a,b]}$ for the law of standard Brownian motion $B : [a, b] \rightarrow \mathbb{R}$, with $B(a) = 0$. Let $X : [a, b] \rightarrow \mathbb{R}$ be a $\mathcal{C}_{0,*}([a, b], \mathbb{R})$ -valued random variable. For $\beta \in (0, \infty)$, we say that X *withstands $L^{\beta-}$ -comparison to Brownian motion* if the Radon-Nikodym derivative of the law of X with respect to $\mathcal{B}_{0,*}^{[a,b]}$ has a finite L^η -norm, for every $\eta \in (0, \beta)$. When a collection of such processes X is instead considered, the collection is said to *uniformly withstand $L^{\beta-}$ -comparison to Brownian motion* if the above L^η -norm, for any given $\eta \in (0, \beta)$, is bounded above by a finite quantity that is independent of the choice of X in the collection.

Conjecture 1.3. *Let $\bar{\Psi} \in (0, \infty)^3$ satisfy $\Psi_2 \geq 1$. There exists $n_0 \in \mathbb{N}$ such that the collection*

$$[-1, 1] \rightarrow \mathbb{R} : y \rightarrow \text{Wgt}_{n_0+n;(*:f,0)}^{(y,1)} - \text{Wgt}_{n_0+n;(*:f,0)}^{(-1,1)},$$

indexed by $(n, f) \in \mathbb{N} \times \mathcal{I}_{\bar{\Psi}}$, uniformly withstands $L^{\infty-}$ -comparison to Brownian motion.

If the conjecture is to be believed, then the use of patches, and affine-shifting to obtain bridges, could ultimately be dispensed with. (The conjecture also makes no use of the restriction that n be even that is seen in Theorem 1.2. In contrast to the use of patches and affine shifting, dispensing with n being even in the theorem, perhaps at the expense of slightly stronger hypotheses on the parameters, would be a straightforward matter. In essence this is because the shift $n \rightarrow n + 1$ in Brownian LPP leaves invariant locally Brownian structure.) Theorem 1.2 is a significant advance in understanding because

- it captures Brownian regularity across a very wide range of initial conditions;
- it does so without taking a local limit, under which Gaussianity would arise from a random function W in a limit of low ϵ for the scaled process $\epsilon^{-1/2}(W(x + \epsilon) - W(x))$;
- and it provides a stronger comparison to Brownian behaviour than absolute continuity statements, in which only the finiteness of the L^1 -norm is proved.

The concepts that drive the proof of Theorem 1.2, including the rigorous tools that have been developed to derive the result, are a critical aspect of the theorem's significance. A guiding theme of our approach is to rely on algebraic inputs only in a very limited way – for example, in order to gain control of narrow wedge profiles at given points – and to harness probabilistic techniques in order to reach far stronger conclusions about profiles. (This theme has something in common with the approach recently used in [BSS16] and [BSS17] to resolve the slow bond conjecture for the totally asymmetric exclusion process.) In [Ham17a], it has been understood, using probabilistic

resampling techniques, that narrow wedge profiles closely resemble Brownian motion. Our task here is to relate much more general profiles to these special ones. Theorem 1.2 will be derived by studying, and proving natural properties of, an important polymer forest structure that is associated to the problem of f -rewarded line-to-point polymers. (The trees in the forest roughly correspond to narrow wedge profiles, similarly to the way that the surviving trees in the right sketch of Figure 1 are rooted at certain points on the x -axis, so that each one may be compared to the single, narrow wedge case, tree depicted on the left.) As such, we view Theorem 1.2 as an important practical and conceptual step towards Conjecture 1.3. How close is the conjecture, given the theorem? All of the stitch points should be removed from the quilts, and the affine adjustment that is made, for the purpose of Brownian comparison, to the (putatively unique) fabric sequence element should also be eliminated. These are genuine technical challenges, in whose resolution the structure of the polymer forest may again have a role to play. Whatever the degree of difficulty these challenges may pose, the techniques leading to Theorem 1.2 have built a probabilistic road from an algebraic point of departure into a far wider realm and as such they achieve for an interesting example the aim of broadening KPZ horizons by probabilistic means.

1.6.1. *Brownian comparison and the KPZ point.* We review our principal conclusion in light of the recent construction of the KPZ fixed point in [MQR17] that we mentioned at the outset of the article. The authors of [MQR17] construct a limiting Markov evolution on profiles begun from a class of initial data f that is very similar to the function space $\mathcal{I}_{\bar{\Psi}}$ for given $\bar{\Psi} \in (0, \infty)^3$. It is tempting to say that, in our language, this object is the limiting profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$ with $n = \infty$. However, the construction undertaken in [MQR17] at present works for totally asymmetric simple exclusion, not for Brownian LPP, so that the uniqueness of the $n = \infty$ limiting process remains conjectural for now. It may perhaps be hoped that the techniques of [MQR17] can be brought to bear on Brownian LPP, so that we can speak unambiguously of $y \rightarrow \text{Wgt}_{\infty;(*:f,0)}^{(y,1)}$. If this were the case, then the limit uniqueness understanding in [MQR17] might combine with the unit-order Brownian comparison made in the present article, so that results such as Theorem 1.2 could be asserted in the limiting, $n = \infty$, case; a technical task concerning weak convergence of measures would have to be carried out, however, in order to pass the quilt description to the limit. Were the existence of $y \rightarrow \text{Wgt}_{\infty;(*:f,0)}^{(y,1)}$ to be proved, it would be natural to extend Johansson's conjecture, and thus to suggest that this random function has an almost surely unique maximizer for any f lying in one of the function spaces $\mathcal{I}_{\bar{\Psi}}$. Note that the example at the end of Section 1.5 shows that the passage of the Brownian patchwork quilt description to $n = \infty$ would not in itself be adequate for proving this extended conjecture.

1.6.2. *The broader study of Brownian LPP in scaled coordinates.* This paper has written so that it may be read on its own. However, the paper forms part of a four-paper study of scaled Brownian LPP, alongside [Ham17a], [Ham17b] and [Ham17c], and here we comment on this article's relation to the other papers. The first of the companion papers develops a theory of Brownian Gibbs line ensembles that applies to scaled Brownian LPP. This theory is hidden for the reader of the present article since it is not needed directly; it is used during the proofs of the central results in [Ham17b] and [Ham17c], which will be applied here and recalled presently. The basic idea of the proof in this article, which we will explain in Section 3, after presenting some notation and tools, is crucial to the four-paper study, since the task of implementing it rigorously has required the development of tools throughout the study. The reader may consult [Ham17a, Section 1.2] for an overview of the investigation of scaled Brownian LPP at large. It is worth bearing in mind, however, that the

material in [Ham17a, Subsection 1.2.3], which presents a conceptual overview, follows closely the upcoming Section 3. This choice of presentation in [Ham17a] reflects the conceptual significance for the overall study of the basic idea of the proof of Theorem 1.2.

1.6.3. *Acknowledgments.* The author thanks Riddhipratim Basu, Shirshendu Ganguly and Jeremy Quastel for valuable conversations at many stages of this project. I thank Judit Zádor for the simulations shown in Figure 1.

2. A GEOMETRIC VIEW: STAIRCASES, ZIGZAGS AND POLYMERS

We intend to explain more of the geometric meaning of Theorem 1.2, and give a rough idea of how this result will be proved, before we embark on the proof itself. Indeed, Section 3 will offer such a rough guide to the principal concerned concepts. Before we can offer this outline, we need to develop some preliminaries. In the present section, we revisit the basic setup of Brownian LPP, and discuss our use of scaled coordinates in a more geometric light.

2.0.1. *Staircases.* Recall from Section 1.3 that energy is ascribed to non-decreasing lists and is then maximized. In order to make a study of those lists that attain this maximum energy, we begin by noting that the lists are in bijection with certain subsets of $[x, y] \times [i, j] \subset \mathbb{R}^2$ that we call *staircases*. Staircases offer a geometric perspective on Brownian LPP and perhaps help in visualizing the problems in question.

The staircase associated to the non-decreasing list $\{z_k : k \in \llbracket i + 1, j \rrbracket\}$ is specified as the union of certain horizontal planar line segments, and certain vertical ones. The horizontal segments take the form $[z_k, z_{k+1}] \times \{k\}$ for $k \in \llbracket i, j \rrbracket$. Here, the convention that $z_i = x$ and $z_{j+1} = y$ is again adopted. The right and left endpoints of each consecutive pair of horizontal segments are interpolated by a vertical planar line segment of unit length. It is this collection of vertical line segments that form the vertical segments of the staircase.

The resulting staircase may be depicted as the range of an alternately rightward and upward moving path from starting point (x, i) to ending point (y, j) . The set of staircases with these starting and ending points will be denoted by $SC_{(x,i) \rightarrow (y,j)}$. Such staircases are in bijection with the collection of non-decreasing lists considered earlier. Thus, any staircase $\phi \in SC_{(x,i) \rightarrow (y,j)}$ is assigned an energy $E(\phi) = \sum_{k=i}^j (B(k, z_{k+1}) - B(k, z_k))$ via the associated z -list.

2.0.2. *Energy maximizing staircases are called geodesics.* A staircase $\phi \in SC_{(x,i) \rightarrow (y,j)}$ whose energy attains the maximum value $M_{(x,i) \rightarrow (y,j)}^1$ is called a geodesic from (x, i) to (y, j) . It is a simple consequence of the continuity of the constituent Brownian paths $B(k, \cdot)$ that such a geodesic exists for all choices of $(x, y) \in \mathbb{R}^2$ with $x \leq y$. The geodesic with given endpoints is known to be almost surely unique: take $\ell = 1$ in Lemma A.1, which appears in Appendix A.

2.0.3. *The scaling map.* For $n \in \mathbb{N}$, consider the n -indexed *scaling map* $R_n : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ given by

$$R_n(v_1, v_2) = \left(2^{-1} n^{-2/3} (v_1 - v_2), v_2/n \right).$$

The scaling map acts on subsets C of \mathbb{R}^2 by setting $R_n(C) = \{R_n(x) : x \in C\}$.

2.0.4. *Scaling transforms staircases to zigzags.* The image of any staircase under R_n will be called an n -zigzag. The starting and ending points of an n -zigzag Z are defined to be the image under R_n of such points for the staircase S for which $Z = R_n(S)$.

Note that the set of horizontal lines is invariant under R_n , while vertical lines are mapped to lines of gradient $-2n^{-1/3}$. As such, an n -zigzag is the range of a piecewise affine path from the starting point to the ending point which alternately moves rightwards along horizontal line segments and northwesterly along sloping line segments, where each sloping line segment has gradient $-2n^{-1/3}$; the first and last segment in this journey may be either horizontal or sloping.

2.0.5. *Scaled geodesics are called polymers.* For $n \in \mathbb{N}$, the image of any geodesic under the scaling map R_n will be called an n -polymer, or often simply a polymer.

2.0.6. *Zigzags have weight.* Any n -zigzag Z from $(x, i/n)$ to $(y, j/n)$ is ascribed a scaled weight $\text{Wgt}(Z) = \text{Wgt}_n(Z)$ given by

$$\text{Wgt}(Z) = 2^{-1/2}n^{-1/3} \left(E(S) - 2(j - i) - 2n^{2/3}(y - x) \right)$$

where Z is the image under R_n of the staircase S . Thus, a polymer maximizes weight among the zigzags that share its endpoints, just as a geodesic maximizes energy over staircases.

2.0.7. *Some basic notation.* For $k \geq 1$, we write \mathbb{R}_{\leq}^k for the subset of \mathbb{R}^k whose elements (z_1, \dots, z_k) are non-decreasing sequences. When the sequences are increasing, we instead write $\mathbb{R}_{<}^k$. We also use the notation A_{\leq}^k and $A_{<}^k$. Here, $A \subset \mathbb{R}$ and the sequence elements are supposed to belong to A . We will typically use this notation when $k = 2$.

2.0.8. *Compatible triples.* Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$, which is to say that $n \in \mathbb{N}$ and $t_1, t_2 \in \mathbb{R}$ with $t_1 < t_2$. Taking $x, y \in \mathbb{R}$, does there exist an n -zigzag from (x, t_1) to (y, t_2) ? As far as the data (n, t_1, t_2) is concerned, such an n -zigzag may exist only if

$$t_1 \text{ and } t_2 \text{ are integer multiples of } n^{-1}. \quad (2)$$

We say that data $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$ is a *compatible triple* if it verifies the last condition.

An important piece of notation associated to a compatible triple is $t_{1,2}$, which we will use to denote the difference $t_2 - t_1$. The law of the underlying Brownian ensemble $B : \mathbb{Z} \times \mathbb{R} \rightarrow \mathbb{R}$ is invariant under integer shifts in the first, curve indexing, coordinate. This translates to a distributional invariance of scaled objects under vertical shifts by multiples of n^{-1} , something that makes the parameter $t_{1,2}$ of far greater significance than t_1 or t_2 .

Supposing now that (n, t_1, t_2) is indeed a compatible triple, the condition $y \geq x - 2^{-1}n^{1/3}t_{1,2}$ ensures that the preimage of (y, t_2) under the scaling map R_n lies northeasterly of the preimage of (x, t_1) . Thus, an n -zigzag from (x, t_1) to (y, t_2) exists in this circumstance. We have mentioned that geodesics exist uniquely for given endpoints almost surely. Taking the image under scaling, this translates to the almost sure existence and uniqueness of the n -polymer from (x, t_1) to (y, t_2) . This polymer will be denoted $\rho_{n;(x,t_1)}^{(y,t_2)}$: see Figure 3. In notation that encompasses the special case of $t_1 = 0$ and $t_2 = 1$ seen in (1), the weight of this polymer will be recorded as $\text{Wgt}_{n;(x,t_1)}^{(y,t_2)}$; (in fact, the latter weight can be ascribed meaning as the weight maximum for the given endpoints even in the probability zero event that the polymer in question is not well defined). These two pieces of notation share something in common with several later examples, in which objects are described

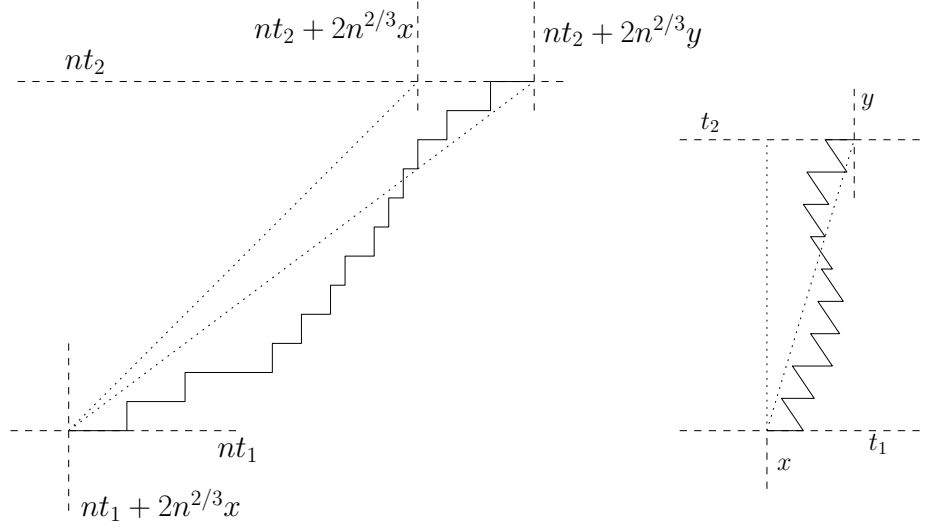


FIGURE 3. Let (n, t_1, t_2) be a compatible triple and let $x, y \in \mathbb{R}$. The endpoints of the geodesic in the left sketch have been selected so that, when the scaling map R_n is applied to produce the right sketch, the n -polymer $\rho_{n;(x,t_1)}^{(y,t_2)}$ results.

in scaled coordinates. Round bracketed expressions in the subscript or superscript will refer to a space-time pair, with the more advanced time in the superscript. Typically some aspect of the n -polymer from (x, t_1) to (y, t_2) is being described when this $\text{TBA}_{n;(x,t_1)}^{(y,t_2)}$ notation is used.

3. POLYMER FORESTS: A ROUGH GUIDE TO THE PROOF OF THE MAIN RESULT

Some of the central players in our study have been introduced: the use of scaled coordinates, polymers with given endpoints such as $\rho_{n;(x,t_1)}^{(y,t_2)}$, and polymer weights such as $\text{Wgt}_{n;(x,t_1)}^{(y,t_2)}$.

Our principal conclusion, Theorem 1.2, asserts that the polymer weight profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$ is a Brownian patchwork quilt, with the quality of the sewing (and the resulting comparison to Brownian motion) not deteriorating even as the initial condition f is permitted to vary over the very broad function class $\mathcal{I}_{\overline{\Psi}}$, and n over all sufficiently high even integers.

A valuable point of departure for beginning to understand how we will obtain this theorem is to consider the special case where $f : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ equals zero at the origin and is otherwise minus infinity. This is the narrow wedge initial condition, in which all competing polymers must begin at zero. A crucial first observation for our proof is that, in this special case, the resulting weight profile is known to enjoy a very strong comparison to Brownian motion. Indeed, a result in [Ham17a] that we will later recall as Theorem 4.3 implies that, for this f , the random function $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$, defined on any given compact interval, withstands L^∞ -comparison to Brownian bridge. Weight profiles arising from such narrow wedge initial conditions, with fixed starting point, would thus seem to represent excellent candidates for the role of fabric sequence elements in the quilt description that we seek to offer of a much more general polymer weight profile.

In the left sketch of Figure 4 is illustrated a geometric view of the narrow wedge weight profile. Note that $\text{Wgt}_{n;(0,0)}^{(y,1)}$ is the weight of the polymer $\rho_{n;(0,0)}^{(y,1)}$. These polymers, indexed by $y \in \mathbb{R}$, all

stream out of the origin at time zero, to arrive at their various ending locations y at time one. The almost sure uniqueness of polymers with given endpoints suggests that, once separated, polymers will not meet again. Thus, the system of polymers should be viewed as a tree, with a root at $(0, 0)$, and a canopy $\mathbb{R} \times \{1\}$ of ending locations.

Return now to a general initial condition $f : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$. The f -rewarded line-to-point polymers may be traced backwards in time from locations $(y, 1)$ with $y \in \mathbb{R}$. They arrive at time zero at a variety of locations, in contrast to the narrow wedge case. They share something with that case, however: as time decreases from one to zero, two polymers that meet will stay together; this may again be expected on the basis of the uniqueness of polymers with given endpoints.

This fact has the implication that we may view the collection of polymers $\rho_{n;(*:f,0)}^{(y,1)}$ indexed by $y \in \mathbb{R}$ as a forest, which we may call the f -rewarded polymer forest. Each constituent tree has a root lying on the x -axis, and a canopy that consists of an interval lying in the line at height one. Indeed, we may partition this copy of the real line into this set of canopies. The polymer weight profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$, when restricted to any given canopy, would seem to have much in common with the narrow wedge weight profile. After all, all the concerned polymers, ending at locations $(y, 1)$, for points y in the given canopy, share their starting location, namely $(r, 0)$, where $r \in \mathbb{R}$ is the root of the canopy in question: see the middle sketch of Figure 4.

This then is the basis of our Brownian patchwork quilt description of the general weight profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$. The patches of the quilt should be the canopies in the f -rewarded polymer forest, so that the stitch point set consists of boundary points between consecutive canopies. The fabric sequence elements take the form $y \rightarrow \text{Wgt}_{n;(r,0)}^{(y,1)}$ where r is the root associated to a tree in the polymer forest. Since r is fixed, the fabric functions should withstand a strong comparison to Brownian bridge, since the narrow wedge weight profiles do.

This plan is a useful guide to the approach by which we will prove Theorem 1.2. There are however significant challenges to be overcome in implementing it. The structure of the problems is elucidated by voicing two objections to the plan:

- Under the plan, the stitch point set is said to be the collection of canopy boundary points. In order to have a useful description, we want this set to be not too big. Certainly, we want to argue that, for all high n , the stitch point set on a unit interval has cardinality that is tight. This raises a geometric question: can we show that the forest of f -rewarded polymers has a tight number of trees per unit length?
- The plan proposes that the f -rewarded weight profile on a given canopy be viewed as $y \rightarrow \text{Wgt}_{n;(r,0)}^{(y,1)}$ where r is the root in question. We may hope that because r is fixed, this object is statistically similar to the narrow wedge case, where say r is zero. There is a significant objection, however. The location r is random, and it may be exceptional. As such, properties of the narrow profile $y \rightarrow \text{Wgt}_{n;(x,0)}^{(y,1)}$, such as locally Brownian structure, that obtain for typical values of $x \in \mathbb{R}$, may disappear for the possibly exceptional choices of root r that we are considering.

We end this rough guide by giving the briefest indication of how we will resolve these problems.

Regarding the first, consider again the f -rewarded polymer forest. Pick a location in the interior of each canopy, so that a point process on $\mathbb{R} \times \{1\}$ results. From each point, draw the f -rewarded line-to-point polymer back to time zero. This system of polymers is pairwise disjoint, because the

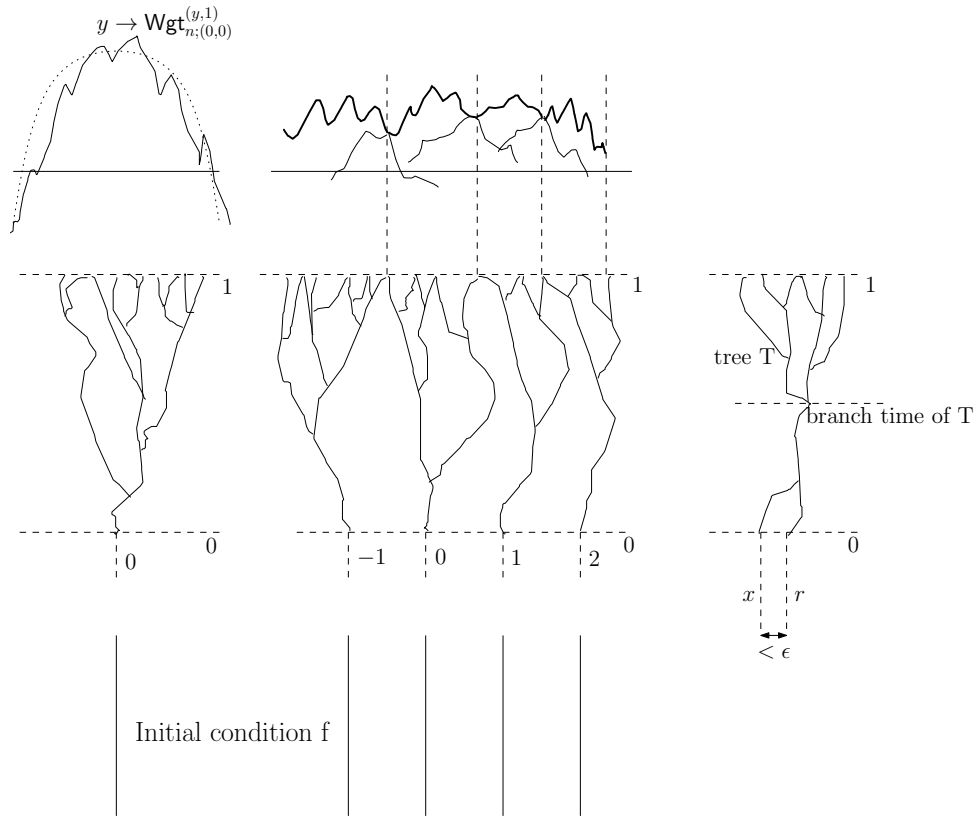


FIGURE 4. In the left and middle sketches, the weight profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$, the f -rewarded polymer forest, and the function f are depicted. On the left, f is zero at zero and otherwise minus infinity, so that the profile is the narrow wedge case $y \rightarrow \text{Wgt}_{n;(0,0)}^{(y,1)}$. In the middle, f is instead zero at integer points. The profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$ is depicted in bold. The dashed vertical lines that contact the line at height one indicate canopy boundary points. In a given canopy with root r , the emboldened profile takes the form $y \rightarrow \text{Wgt}_{n;(r,0)}^{(y,1)}$. The right sketch depicts one tree, rooted at r , in an f -rewarded polymer forest. The proposed rerooting of this tree to a nearby discrete mesh element $x \in \epsilon\mathbb{Z}$ is illustrated.

time-one endpoints lie in different canopies. If the tightness alluded to in the first problem were to fail, we would be drawing, for high n , many disjoint polymers, all ending on a given unit-order interval. This behaviour has been proved to be a rarity for Brownian LPP. As we will recall in Theorem 4.1, it has been proved in [Ham17c, Theorem 1.1] that the probability that there exist k disjoint polymers, beginning and ending in a unit interval, has a superpolynomial tail in k , uniformly in high choices of the scaling parameter $n \in \mathbb{N}$.

Regarding the second difficulty, that root locations r may be exceptional, we may try to solve the problem by a rerooting procedure: see the right sketch in Figure 4. Close to the location r

is an element x in the discrete ϵ -mesh $\epsilon\mathbb{Z}$, where $\epsilon > 0$ is small but fixed. The polymer weight profile that we are trying to describe, $y \rightarrow \text{Wgt}_{n;(r,0)}^{(y,1)}$, may be compared to the narrow wedge profile $y \rightarrow \text{Wgt}_{n;(x,0)}^{(y,1)}$. Because x lies in a discrete mesh, it may be viewed as a typical location, so that the objection raised in the second bullet point does not arise. But why should the nearby profile $y \rightarrow \text{Wgt}_{n;(x,0)}^{(y,1)}$ be expected to offer an accurate description of $y \rightarrow \text{Wgt}_{n;(r,0)}^{(y,1)}$? We will now present an heuristic argument that, since r and x are close, these two functions should, with a high probability determined by this closeness, differ by a random constant as y varies over the canopy of the tree of which r is the root. The tree in question has a *branch* time, with the constituent polymers following a shared course (the tree trunk, if you like) until that time, after which they may go their separate ways to the various locations in the canopy. Now, when the tree is rerooted the short distance from $(r, 0)$ to $(x, 0)$, we may expect that its structure changes by a modification in the form of the polymers only close to time zero: see again the right sketch of Figure 4. Provided this modification has finished by the branch time, it will affect only the form of the tree trunk, and will be shared by all concerned polymers, no matter at which canopy point $(y, 1)$ they end. For this reason, the difference $y \rightarrow \text{Wgt}_{n;(x,0)}^{(y,1)} - \text{Wgt}_{n;(r,0)}^{(y,1)}$ may be expected to equal the discrepancy in weight between the new and the old tree trunks with high probability, and thus typically be independent of the canopy location y .

The actual solution we undertake will modify these suggestions. In particular, the resolution of the second difficulty will proceed by a means that does involve rerooting, but not quite in the way we have just suggested. In any case, this rough guide may help to set the reader's bearings, as we now turn to set out precisely some of the tools that will be needed when it is implemented rigorously.

At roughly this moment in a paper, it is conventional to offer an overview of the structure of the remainder. The story of the proof will unfold in several steps: for example, after the next section of tools, we will return to the rough guide and explain more carefully how rerooting will resolve the second objection. It is probably not helpful to attempt to indicate the global structure of the article for now, beyond pointing that the actual proof of Theorem 1.2 will appear in the final Section 9, and that the road to that section will alternate between heuristic elaboration of the rough guide and rigorous development of the machinery necessary for the rough guide's implementation. We also mention that the paper has two appendices, whose roles will be described in Section 4.4 and Subsection 4.7.2.

4. IMPORTANT TOOLS

If the rough guide is to be implemented, several significant inputs will be needed. It is the task of this section to quote the necessary results.

4.1. A comment about explicit constants. The results that we are quoting state hypotheses on parameters in fairly explicit terms, and this means that we must define some explicit constants before beginning to state the results. Certain results have been proved using a framework of Brownian Gibbs line ensembles, which are in essence systems of mutually avoiding Brownian bridges that arise in Brownian LPP via the Robinson-Schensted-Knuth correspondence. Several of the results we will be quoting are proved using a regularity property of such ensembles, with this property specified in terms of two positive parameters (c, C) . We mention these things now merely in order to explain that these two parameters (c, C) are fixed throughout the paper. It is [Ham17c, Proposition 4.2]

that determines their values. The reader who wishes to understand more about the Brownian Gibbs framework may consult [Ham17a], reading from say the start of Section 1.3 into Chapter 2.

The results to be quoted will make reference to two fixed sequences of constants that are expressed in terms of the given positive values C and c . These two sequences are $\{C_k : k \geq 2\}$ and $\{c_k : k \in \mathbb{N}\}$. They are specified by setting, for each $k \geq 2$,

$$C_k = \max \left\{ 10 \cdot 20^{k-1} 5^{k/2} \left(\frac{10}{3-2^{3/2}} \right)^{k(k-1)/2} C, e^{c/2} \right\} \quad (3)$$

as well as $C_1 = 140C$; and

$$c_k = ((3 - 2^{3/2})^{3/2} 10^{-3/2})^{k-1} c_1,$$

with $c_1 = 2^{-5/2} c \wedge 1/8$.

4.2. The rarity of many disjoint polymers. In the rough guide, we saw that this rarity will be needed to resolve the first objection. In fact, we will use two related results, Theorems 4.1 and 4.2.

Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$ be a compatible triple, and let $I, J \subset \mathbb{R}$ be intervals. Set $\text{MaxDisjtPoly}_{n; (I, t_1)}^{(J, t_2)}$ equal to the maximum cardinality of a disjoint set of n -polymers each of whose start and end points have the respective forms (x, t_1) and (y, t_2) where x is some element of I and y is some element of J .

(This MaxDisjtPoly notation fits within the framework discussed in the last paragraph of Section 2. In this case, the first element in the space-time pairs (I, t_1) and (J, t_2) is an interval, rather than a point.)

The next result is [Ham17c, Theorem 6.2]: the maximum disjoint polymer cardinality has a tail that decays super-polynomially.

Theorem 4.1. *There exist constants $K_0 \geq 1$, $a_0 \in (0, 1)$, $k_0 > 0$ and a sequence of positive constants $\{H_i : i \in \mathbb{N}\}$ for which $\sup_{i \in \mathbb{N}} H_i \exp \{ -2(\log i)^{11/12} \}$ is finite, such that the following holds. Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$ be a compatible triple with $nt_{1,2}$ even. Let $x, y \in \mathbb{R}$, $a, b \in \mathbb{N}$ and $k \in \mathbb{N}$. Write $h = a \vee b$. Suppose that*

$$k \geq k_0 \vee (|x - y| t_{1,2}^{-2/3} + 2h)^3$$

and

$$nt_{1,2} \geq \max \left\{ 2(K_0)^{(12)^{-2}(\log \log k)^2} (\log k)^{K_0}, a_0^{-9} (|y - x| t_{1,2}^{-2/3} + 2h)^9, \right. \\ \left. 10^{325} c^{-36} k^{465} \max \{ 1, (|y - x| t_{1,2}^{-2/3} + 2h)^{36} \} \right\}. \quad (4)$$

Then

$$\mathbb{P} \left(\text{MaxDisjtPoly}_{n; ([x, x+at_{1,2}^{2/3}], t_1)}^{([y, y+bt_{1,2}^{2/3}], t_2)} \geq k \right) \leq k^{-(145)^{-1}(\log \beta)^{-2}(0 \vee \log \log k)^2} \cdot h^{(\log \beta)^{-2}(\log \log k)^2/288+3/2} H_k.$$

Here, β is specified to be $e \vee \limsup_{i \in \mathbb{N}} \beta_i^{1/i}$, where the sequence of constants $\{\beta_i : i \in \mathbb{N}\}$, which verifies $\limsup \beta_i^{1/i} < \infty$, is supplied by [Ham17c, Corollary 5.2].

We make an expository comment about result statements which is illustrated by the last theorem: we have chosen to be fairly explicit in stating hypothesised parameter bounds. One effect of this is to make some statements look complicated, and we now suggest to the reader some ways of reading the results that may serve to focus attention on their basic meaning. First, a basic scaling principle, explained in [Ham17b, Subsection 1.3.10] means that there is no loss of generality in taking $t_1 = 0$ and $t_2 = 1$, and thus $t_{1,2} = 1$, in results stated using this parameter pair. Choosing x, y bounded above in absolute value, by one say, may also be useful. Also taking $a = b = 1$ in the present case, we see that the requirements in the theorem that k is bounded below by a universal constant, and that n is at least a polynomial in k , are quite weak, since our interest is in statements that hold uniformly in high n . In this light, the basic conclusion of the last theorem is seen to be that the probability that k disjoint polymers cross a unit interval in unit time is at most $k^{-\Theta(1)(\log \log k)^2}$, uniformly in high n .

The next result, [Ham17c, Theorem 6.1], asserts a more local form of disjoint polymer rarity: it is unlikely that several disjoint polymers begin and end in a shared pair of short intervals. It shares the parameters K_0 and a_0 , and the sequence $\{\beta_i : i \in \mathbb{N}\}$, with the preceding result, and also uses a further positive constant η_0 .

Theorem 4.2. *Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{>}^2$ be a compatible triple with $nt_{1,2}$ even. Let $m \in \mathbb{N}$, $\epsilon > 0$ and $x, y \in \mathbb{R}$ satisfy the conditions that $m \geq 2$,*

$$\epsilon \leq \min \left\{ (\eta_0)^{4m^2}, 10^{-616} c_m^{22} m^{-115}, \exp \left\{ -C^{3/8} \right\} \right\}, \quad (5)$$

$$nt_{1,2} \geq \max \left\{ 2(K_0)^{m^2} (\log \epsilon^{-1})^{K_0}, 10^{584} c_m^{-48} m^{240} c^{-36} \epsilon^{-222} \max \left\{ 1, |x-y|^{36} t_{1,2}^{-24} \right\}, a_0^{-9} |y-x|^9 t_{1,2}^{-6} \right\}, \quad (6)$$

as well as $|y-x|t_{1,2}^{-2/3} \leq \epsilon^{-1/2}$. Then

$$\begin{aligned} & \mathbb{P} \left(\text{MaxDisjtPoly}_{n; ([x-t_{1,2}^{2/3}\epsilon, x+t_{1,2}^{2/3}\epsilon], t_1)}^{([y-t_{1,2}^{2/3}\epsilon, y+t_{1,2}^{2/3}\epsilon], t_2)} \geq m \right) \\ & \leq \epsilon^{(m^2-1)/2} \cdot 10^{32m^2} m^{15m^2} c_m^{-3m^2} C_m (\log \epsilon^{-1})^{4m^2} \exp \left\{ \beta_m (\log \epsilon^{-1})^{5/6} \right\}. \end{aligned}$$

For this result, it is helpful to bear in mind that m is fixed independently of n , so that the probability that m polymers disjointly cross between two similarly placed intervals of a short width ϵ in unit time is found to be at most $\epsilon^{m^2-1+o(1)}$, uniformly in high n .

4.3. Narrow-wedge polymer weight profiles bear strong comparison to Brownian bridge.

This is a fundamental idea in the rough guide and we now recall the result in question, which is in essence [Ham17a, Corollary 2.12]. For the statement, recall from Subsection 1.5.1 our notation for the space of continuous functions, and that of bridges, defined on a compact real interval, and for the bridge projection between them, as well as the standard Brownian bridge law.

Theorem 4.3. *Let $K \in \mathbb{R}$ and $d \geq 1$.*

Let $x \in \mathbb{R}$. For $n \in \mathbb{N}$ satisfying $K \geq x - 2^{-1}n^{1/3}$, we define a $\mathcal{C}_{,*}([K, K+d], \mathbb{R})$ -valued stochastic process $\mathcal{L} = \mathcal{L}_{n,x} : [K, K+d] \rightarrow \mathbb{R}$ by setting $\mathcal{L}(y) = \text{Wgt}_{n;(x,0)}^{(y,1)}$ for each $y \in [K, K+d]$.*

Endowing the bridge space $\mathcal{C}_{0,0}([K, K+d], \mathbb{R})$ with the topology of uniform convergence, let A be any measurable subset of this space. Then there exists a constant $G > 0$ such that

$$\mathbb{P}(\mathcal{L}^{[K, K+d]}(k, \cdot) \in A) \leq a \cdot G \exp \{G(\log a^{-1})^{5/6}\}.$$

where here a denotes the probability $\mathcal{B}_{0,0}^{[K, K+d]}(A)$. This bound holds for any choice of the parameters (K, d, x) , provided that n is at least a level n_0 that is determined by $K - x$ and d . The constant $G > 0$ may be selected independently of the choice of (K, d, x) subject to $|K - x| + d \leq 2^{-1}c(n+1)^{1/9}$.

Proof. First we mention that the key result here is [Ham17a, Corollary 2.12], which depends on the apparatus of the lengthy paper [Ham17a] in a substantial way. The reader who wants to understand the proof should consult that work. Here, we merely explain formally how to derive the theorem from this result. We mentioned in Section 4.1 that certain regular Brownian Gibbs line ensembles play an important role, and this is certainly true of this theorem. Indeed, [Ham17c, Proposition 4.2] implies that the stochastic process \mathcal{L} is the lowest indexed curve in a line ensemble with $n+1$ curves that, in the language of [Ham17a], is a $(\bar{\varphi}, c, C)$ -regular Brownian Gibbs ensemble, where $\bar{\varphi}$ is the vector $(1/3, 1/9, 1/3)$ and the positive constants c and C are furnished by [Ham17c, Proposition 4.2]. For this reason, [Ham17a, Corollary 2.12] implies the result; note that the index n there assumes the value $n+1$ given by the present context. The final sentence of the theorem, concerning selection of $G > 0$, is a consequence of the final sentence of [Ham17a, Theorem 2.11], which is the result that underlies [Ham17a, Corollary 2.12]. \square

4.4. Basic results on polymers and weights. Some such understanding will be needed, for example, to make sense of polymer forests. Here we state the relevant tools.

Let (n, t_1, t_2) be a compatible triple. We introduce an ordering relation \preceq on the space of n -polymers with lifetime $[t_1, t_2]$. To define it, let $(x_1, x_2), (y_1, y_2) \in \mathbb{R}^2$ and consider a polymer ρ_1 from (x_1, t_1) to (y_1, t_2) and another ρ_2 from (x_2, t_1) to (y_2, t_2) . We declare that $\rho_1 \preceq \rho_2$ if ‘ ρ_2 lies on or to the right of ρ_1 ’: formally, if ρ_2 is contained in the union of the closed horizontal planar line segments whose left endpoints lie in ρ_1 .

First is a simple sandwiching result, [Ham17c, Lemma 5.6].

Lemma 4.4. *Let (n, t_1, t_2) be a compatible triple, and let $(x_1, x_2), (y_1, y_2) \in \mathbb{R}_{\leq}^2$. Suppose that there is a unique n -polymer from (x_i, t_1) to (y_i, t_2) , both when $i = 1$ and $i = 2$. (This circumstance occurs almost surely, and the resulting polymers have been labelled $\rho_{n;(x_1, t_1)}^{(y_1, t_2)}$ and $\rho_{n;(x_2, t_1)}^{(y_2, t_2)}$.) Now let ρ denote any n -polymer that begins in $[x_1, x_2] \times \{t_1\}$ and ends in $[y_1, y_2] \times \{t_2\}$. Then $\rho_{n;(x_1, t_1)}^{(y_1, t_2)} \preceq \rho \preceq \rho_{n;(x_2, t_1)}^{(y_2, t_2)}$.*

Next we discuss f -rewarded polymers.

Definition 4.5. (1) Let $\text{PolyUnique}_{n;0}^1$ denote the set of pairs $(x, y) \in \mathbb{R}^2$ for which there exists an n -zigzag from $(x, 0)$ to $(y, 1)$ whose weight uniquely attains the value $W_{n;(x,0)}^{(y,1)}$. In other words, $(x, y) \in \text{PolyUnique}_{n;0}^1$ precisely when the polymer $\rho_{n;(x,0)}^{(y,1)}$ is well defined.

(2) Let $f : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ be measurable. Let $\text{PolyUnique}_{n;(*:f,0)}^1$ denote the set of $y \in \mathbb{R}$ such that there exists a unique choice of (x, ϕ) , with $x \in \mathbb{R}$ and ϕ an n -zigzag from $(x, 0)$ to $(y, 1)$, such that the weight $\text{Wgt}(\phi)$ is equal to $W_{n;(*:f,0)}^{(y,1)}$. That is, $y \in \mathbb{R}$ is an element of $\text{PolyUnique}_{n;(*:f,0)}^1$ precisely when the f -rewarded line-to-point polymer that ends at $(y, 1)$ is well defined. We call this polymer $\rho_{n;(*:f,0)}^{(y,1)}$.

For $\bar{\Psi} \in (0, \infty)^3$, recall the function space $\mathcal{I}_{\bar{\Psi}}$ from Definition 1.1. The next result will be proved in Appendix A.

Lemma 4.6. (1) *Let $x, y \in \mathbb{R}$. There exists an n -zigzag from $(x, 0)$ to $(y, 1)$ if and only if $y \geq x - n^{1/3}/2$. When the last condition is satisfied, there is almost surely a unique n -polymer from $(x, 0)$ to $(y, 1)$, which is to say, $(x, y) \in \text{PolyUnique}_{n,0}^1$ almost surely.*

(2) *Suppose that $n \in \mathbb{N}$ satisfies $n > 2^{-3/2}\Psi_1^3 \vee 8(\Psi_2 + 1)^3$, and that $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$. Then $[-1, 1] \setminus \text{PolyUnique}_{n;(*:f,0)}^1$ is almost surely of Lebesgue measure zero (and contains given value $y \in [-1, 1]$ with zero probability).*

We mention that the point-to-point polymer uniqueness Lemma 4.6(1) has been expressed in scaled coordinates. Indeed, we have been eager to move promptly to the use of scaled coordinates throughout this article, since our whole perspective makes them essential. However, Brownian LPP, in its unscaled coordinates, underlies everything we do. Lemma 4.6(1) is an example: the basic result on which it depends is Lemma A.1, which is its unscaled counterpart; in fact, the result is more general than this, treating uniqueness of certain multi-geodesics. Although, in keeping with our emphasis on scaled coordinates, this lemma appears only at the end of the paper, in Appendix A, it is an interesting basic result concerning Brownian LPP.

We also need [Ham17b, Lemma 1.6(1)].

Lemma 4.7. *Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$ be a compatible triple. The random function $(x, y) \rightarrow \text{Wgt}_{n;(x,t_1)}^{(y,t_2)}$, which is defined on the set of $(x, y) \in \mathbb{R}^2$ satisfying $y \geq x - 2^{-1}n^{1/3}t_{1,2}$, is continuous almost surely.*

4.5. Polymer fluctuation. The rerooting procedure to resolve the second difficulty in the rough guide will make use of an important aspect of polymer geometry: polymers have Hölder-2/3-regularity. We now quote two results from [Ham17c] to this effect.

Some notational preparation is needed. Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$ be a compatible triple, and let $x, y \in \mathbb{R}$. The polymer $\rho_{n;(x,t_1)}^{(y,t_2)}$ has been defined to be a subset of $\mathbb{R} \times [t_1, t_2]$ containing (x, t_1) and (y, t_2) , but really as n rises towards infinity, it becomes more natural to seek to view it as a random function that maps its lifetime $[t_1, t_2]$ to the real line. In choosing to adopt this perspective, we will abuse notation: taking $t \in [t_1, t_2]$, we will speak of the value $\rho_{n;(x,t_1)}^{(y,t_2)}(t) \in \mathbb{R}$, as if the polymer were in fact a function of $[t_1, t_2]$. Some convention must be adopted to resolve certain microscopic ambiguities as we make use of this new notation, however. First, we will refer to $\rho_{n;(x,t_1)}^{(y,t_2)}(t)$ only when $t \in [t_1, t_2]$ satisfies $nt \in \mathbb{Z}$, a condition that ensures that the intersection of the set $\rho_{n;(x,t_1)}^{(y,t_2)}$ with the line at height t takes place along a horizontal planar interval.

Second, we have to explain which among the points in this interval $\rho_{n;(x,t_1)}^{(y,t_2)} \cap \{(\cdot, t) : \cdot \in \mathbb{R}\}$ we wish to denote by $\rho_{n;(x,t_1)}^{(y,t_2)}(t)$. To present and explain our convention in this regard, we let $\ell_{(x,t_1)}^{(y,t_2)}$ denote the planar line segment whose endpoints are (x, t_1) and (y, t_2) . Adopting the same perspective as for the polymer, we abuse notation to view $\ell_{(x,t_1)}^{(y,t_2)}$ as a function from $[t_1, t_2]$ to \mathbb{R} , so that $\ell_{(x,t_1)}^{(y,t_2)}(t) = t_{1,2}^{-1}((t_2 - t)x + (t - t_1)y)$.

Our convention will be to set $\rho_{n;(x,t_1)}^{(y,t_2)}(t)$ equal to z where (z, t) is that point in the horizontal segment $\rho_{n;(x,t_1)}^{(y,t_2)} \cap \{(\cdot, t) : \cdot \in \mathbb{R}\}$ whose distance from $\ell_{(x,t_1)}^{(y,t_2)}(t)$ is maximal. (An arbitrary tie-breaking rule, say $\rho_{n;(x,t_1)}^{(y,t_2)}(t) \geq \ell_{(x,t_1)}^{(y,t_2)}(t)$, resolves the dispute if there are two such points.) The reason for this very particular convention is that our purpose in using it is to explore, in the soon-to-be-stated Theorem 4.8, upper bounds on the probability of large fluctuations between the polymer $\rho_{n;(x,t_1)}^{(y,t_2)}$ and the line segment $\ell_{(x,t_1)}^{(y,t_2)}$ that interpolates the polymer's endpoints. Our convention ensures that the form of the theorem would remain valid were any other convention instead adopted.

In order to study the intermediate time $(1-a)t_1 + at_2$ (in the role of t in the preceding), we now let $a \in (0, 1)$ and impose that $at_{1,2} \in \mathbb{Z}$: doing so ensures that, as desired, $t \in n^{-1}\mathbb{Z}$, where $t = (1-a)t_1 + at_2$.

Consider also $r > 0$. Define the *polymer deviation regularity* event

$$\text{PolyDevReg}_{n;(x,t_1)}^{(y,t_2)}(a, r) = \left\{ \left| \rho_{n;(x,t_1)}^{(y,t_2)}((1-a)t_1 + at_2) - \ell_{(x,t_1)}^{(y,t_2)}((1-a)t_1 + at_2) \right| \leq rt_{1,2}^{2/3} (a \wedge (1-a))^{2/3} \right\},$$

where \wedge denotes minimum. For example, if $a \in (0, 1/2)$, the polymer's deviation from the interpolating line segment, at height $(1-a)t_1 + at_2$ (when the polymer's journey has run for time $at_{1,2}$), is measured in the natural time-to-the-two-thirds scaled units obtained by division by $(at_{1,2})^{2/3}$, and compared to the given value $r > 0$.

For intervals $I, J \subset \mathbb{R}$, we extend this definition by setting

$$\text{PolyDevReg}_{n;(I,t_1)}^{(J,t_2)}(a, r) = \bigcap_{x \in I, y \in J} \text{PolyDevReg}_{n;(x,t_1)}^{(y,t_2)}(a, r).$$

The perceptive reader may notice a problem with the last definition. The polymer $\rho_{n;(x,t_1)}^{(y,t_2)}$ is well defined almost surely for given endpoints, but this property is no longer assured as the parameters vary over $x \in I$ and $y \in J$. In the case of exceptional (x, y) where several n -polymers move from (x, t_1) to (y, t_2) , we interpret $\rho_{n;(x,t_1)}^{(y,t_2)}$ as the union of all these polymers, for the purpose of defining $\rho_{n;(x,t_1)}^{(y,t_2)}(t)$. This convention permits us to identify worst case behaviour, so that the complementary event $\neg \text{PolyDevReg}_{n;(I,t_1)}^{(J,t_2)}(a, r)$ is triggered by a suitably large fluctuation on the part of any concerned polymer.

The next two results are [Ham17c, Theorem 1.3] and [Ham17c, Proposition 1.4]. In essence, they both assert that a unit lifetime polymer fluctuates, in a short initial or final period of duration a , by more than $a^{2/3}r$, (with r capable of assuming a very broad range of positive values), only with probability at most $\exp\{-\Theta(1)r^{3/4}\}$, uniformly in high choices of $n \in \mathbb{N}$. In addition, Theorem 4.8 asserts this uniformly over polymers that begin and end in a unit-order interval.

Theorem 4.8. *Let $(n, t_1, t_2) \in \mathbb{N} \times \mathbb{R}_{<}^2$ be a compatible triple, with $nt_{1,2}$ even, and let $x, y \in \mathbb{R}$.*

(1) *Let $a \in [1 - 10^{-11}c_1^2, 1)$ satisfy $at_{1,2} \in n^{-1}\mathbb{Z}$. Suppose that $n \in \mathbb{N}$ satisfies*

$$nt_{1,2} \geq \max \left\{ 10^{29}(1-a)^{-25}c^{-18}, 10^{24}c^{-18}(1-a)^{-25}|x-y|^{36}t_{1,2}^{-24} \right\}.$$

Let $r > 0$ be a parameter that satisfies

$$r \geq \max \left\{ 10^9 c_1^{-4/5}, 15C^{1/2}, 11(1-a)^{1/3}t_{1,2}^{-2/3}|x-y| \right\}$$

and $r \leq 3(1-a)^{25/9} n^{1/36} t_{1,2}^{1/36}$.

Writing $I = [x, x + t_{1,2}^{2/3}(1-a)^{2/3}r]$ and $J = [y, y + t_{1,2}^{2/3}(1-a)^{2/3}r]$, we have that

$$\mathbb{P}\left(\neg \text{PolyDevReg}_{n;(I,t_1)}^{(J,t_2)}(a, 2r)\right) \leq 44Cr \exp\left\{-10^{-11}c_1 r^{3/4}\right\}.$$

(2) The same statement holds verbatim when appearances of a are replaced by $1-a$.

Proposition 4.9. Let $(n, t_1, t_2) \in \mathbb{R}_{<}^2$ be a compatible triple, with $nt_{1,2}$ even. Let $x, y \in \mathbb{R}$ and let $a \in [1 - 10^{-11}c_1^2, 1)$ satisfy $at_{1,2} \in n^{-1}\mathbb{Z}$. Suppose that

$$nt_{1,2} \geq \max\left\{10^{29}(1-a)^{-25}c^{-18}, 10^{24}c^{-18}(1-a)^{-25}|x-y|^{36}t_{1,2}^{-24}\right\}.$$

Let $r > 0$ be a parameter that satisfies

$$r \geq \max\left\{10^9c_1^{-4/5}, 15C^{1/2}, 11(1-a)^{1/3}t_{1,2}^{-2/3}|x-y|\right\}$$

and $r \leq 3(1-a)^{25/9} n^{1/36} t_{1,2}^{1/36}$. Then

$$\mathbb{P}\left(\neg \text{PolyDevReg}_{n;(x,t_1)}^{(y,t_2)}(a, r)\right) \leq 22Cr \exp\left\{-10^{-11}c_1 r^{3/4}\right\}.$$

4.6. Control on the fluctuation of line-to-point polymers. To implement the rough guide, we will want to know that the starting endpoints of f -rewarded line-to-point polymers are localized at unit-order distances from their endpoints. We now quote a result, [Ham17b, Lemma 4.1], that makes such an assertion.

Recalling Definition 1.1, let $\bar{\Psi} \in (0, \infty)^3$ and $f \in \mathcal{I}_{\bar{\Psi}}$. For $R \geq 0$, define the event

$$\text{RegFluc}_{n;(*:f,0)}^{\{-1,1\},1}(R) = \left\{\rho_{n;(*:f,0)}^{(-1,1)}(0) \geq -(R+1), \rho_{n;(*:f,0)}^{(1,1)}(0) \leq R+1\right\}.$$

Lemma 4.10. Let $n \in \mathbb{N}$, $R > 0$ and $\bar{\Psi} \in (0, \infty)^3$ satisfy

$$n \geq c^{-18} \max\left\{(\Psi_2 + 1)^9, 10^{23}\Psi_1^9, 3^9\right\},$$

$$R \geq \max\left\{39\Psi_1, 5, 3c^{-3}, 2((\Psi_2 + 1)^2 + \Psi_3)^{1/2}\right\},$$

and $R \leq 6^{-1}cn^{1/9}$. Then, for any $f \in \mathcal{I}_{\bar{\Psi}}$,

$$\mathbb{P}\left(\neg \text{RegFluc}_{n;(*:f,0)}^{\{-1,1\},1}(R)\right) \leq 38RC \exp\left\{-2^{-6}cR^3(2^{-1/2} - 2^{-1})^{3/2}\right\}.$$

4.7. Conventions governing the presentation of upcoming proofs. We close this section of tools by explaining two such conventions, even though they will become operative only when we begin giving proofs in Section 6.

4.7.1. *Boldface notation for quoted results.* During the upcoming proofs, we will naturally be making use of the various tools that we have recalled in this section. The statements of such results involve several parameters, in several cases including (n, t_1, t_2) , spatial locations x and y , and positive real parameters such as r . We will employ a notational device that will permit us to disregard notational conflict between the use of such parameters in the context of the ongoing proof in question, and their role in the statements of quoted results. When specifying the parameter settings of a particular application, we will allude to the parameters of the result being applied in boldface, and thus permit the reuse of the concerned symbols.

4.7.2. *The role of hypotheses invoked during proofs.* When we quote results in order to apply them, we will take care, in addition to specifying the parameters according to the just described convention, to indicate explicitly what the conditions on these parameters are that will permit the quoted result in question to be applied. Of course, it is necessary that the hypotheses of the result being proved imply all such conditions. The task of verifying that the hypotheses of a given result are adequate for the purpose of obtaining all conditions needed to invoke the various results used during its proof may be called the calculational derivation of the result in question. This derivation is necessary, but also in some cases lengthy and unenlightening: a succession of trivial steps. In the case of two results proved in this article, Proposition 6.1 and Lemma 8.1, we have chosen to separate the calculational derivation of most from the rest of the proof. These derivations may be found in Appendix B at the end of the paper.

5. THE ROUGH GUIDE ELABORATED: HOW REROOTING WILL BE CARRIED OUT

We have assembled the key tools for the proof of Theorem 1.2. In order to present a clear heuristic for how the proof will proceed, we now revisit the rough guide from Section 3. Recall that we proposed that the second objection, that root locations may be exceptional, may be resolved by rerooting the trees in the f -rewarded polymer forest to nearby elements in an ϵ -mesh lying in the x -axis, where ϵ is small.

There are some problems in implementing this idea. The weight profile from a given root and from an adjacent ϵ -mesh point will differ by a constant only if the concerned polymers that emanate from these two locations merge quickly enough, before the branch time at which the polymers running from the root to different locations in the canopy in question split apart. This will be become easier to prove if ϵ is small. However, when we come to say that the polymer weight profile $y \rightarrow \mathbf{Wgt}_{n;(x,0)}^{(y,1)}$ for x belonging to the ϵ -mesh withstands a strong comparison to Brownian bridge by invoking Theorem 4.3, our comparison will deteriorate as ϵ decreases, because it will be necessary to sum estimates as a union bound over the order ϵ^{-1} points in a bounded portion of the ϵ -mesh. This will put an opposing, upward, pressure on our choice of the parameter ϵ .

We resolve this difficulty by choosing a means for rerooting that can be proved to have a rather high probability of working. Recall the right sketch in Figure 4, and that we have considered the root r of a tree T in the f -rewarded polymer forest, as well as the left-adjacent ϵ -mesh point x . In proposing to reroot the tree from a beginning at $(r, 0)$ to one at $(x, 0)$, we want to ensure, in order to achieve the desired constant difference of the associated polymer weight profiles, that the coalescence time of the polymer pair running to a given canopy point from $(r, 0)$ and from $(x, 0)$ occurs *before* the branch time of the tree T .

Since x and r differ by at most the small quantity $\epsilon > 0$, it may be expected that this merging occurs quickly; in fact, the order of time at which it occurs is $\epsilon^{3/2}$. In order to work with such a

merging event that has a provably high probability, we will modify the trees we work with, in order that their branch times are necessarily close to one.

To specify the new trees, picture the f -rewarded polymer forest sitting inside the strip $\mathbb{R} \times [0, 1]$, and draw a horizontal line of height $1 - \epsilon^{3/2}$ through the strip. The line cuts the trees into smaller tree tops sitting in the strip $\mathbb{R} \times [1 - \epsilon^{3/2}, 1]$, each with its own canopy. These new canopies, which we will be introduced formally as $(n, * : f, 1 - \epsilon^{3/2})$ -canopies later on, are intervals that subdivide the original collection of canopies into finer pieces. Each new canopy has a portion of an f -rewarded polymer tree associated to it, with a root at time zero, as before. These new trees have the merit of having a branch time that necessarily exceeds $1 - \epsilon^{3/2}$. Consistently with the two-thirds power law that governs polymer fluctuation, the lengths of the new canopies are of order $(\epsilon^{3/2})^{2/3} = \epsilon$, something illustrated in the left sketch of Figure 5.

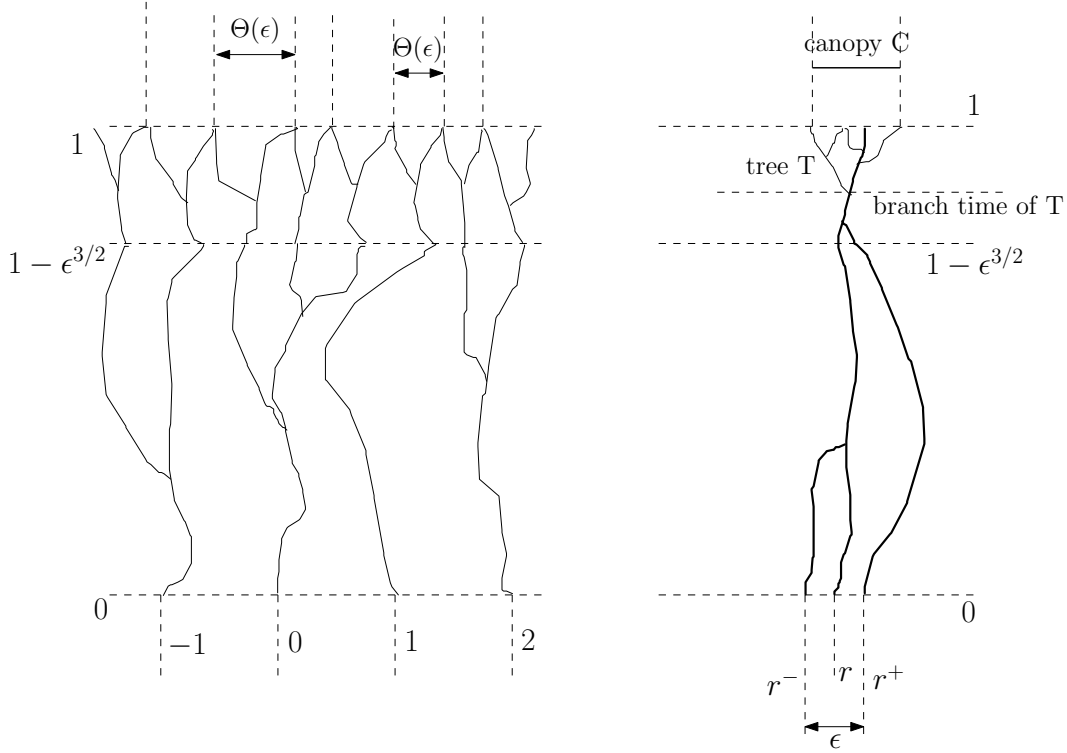


FIGURE 5. *Left:* The f -rewarded polymer forest corresponding to the f that is zero on \mathbb{Z} , and otherwise minus infinity, is depicted. The tree tops are cut along the line at height $1 - \epsilon^{3/2}$, with the boundary points of the resulting new canopies indicated by the vertical dashed lines that touch the height one line. *Right:* The f -rewarded polymer tree T associated to one of these new canopies C . If the canopy does not experience late coalescence then, for any $y \in C$, one of the consecutive pairs in the emboldened triple of polymers must merge before time $1 - \epsilon^{3/2}$. In this instance, the polymer pair from r^- and r so merge, raising the prospect of rerooting the tree T from its root r to the nearby mesh element $r^- \in \epsilon\mathbb{Z}$.

Consider now one of the new canopies C , to which is associated a time-zero root r . Working again with the ϵ -mesh $\epsilon\mathbb{Z}$ embedded in the x -axis, we will consider not only the mesh point to the left of r , but also the one to the right: call these points r^- and r^+ . For any canopy location $y \in C$, we

may consider the triple of polymers $\rho_{n;(r^-,0)}^{(y,1)}$, $\rho_{n;(r,0)}^{(y,1)}$ and $\rho_{n;(r^+,0)}^{(y,1)}$. As we have mentioned, we may expect the merging time between any pair of these polymers to be rather small. We now have the advantage that, with the aim of proving such merging times are less than the branch time of the tree, we need merely prove that they are typically less than $1 - \epsilon^{3/2}$: that is, not necessarily early, but not very late. Moreover, in order to work with a circumstance of suitably high probability, we will consider an event in which it is demanded that only one pair of the concerned polymers coalesces by this time.

Thus, we want to investigate the probability of the event that at least one of the coalescence times between the first and second of the polymers, and between the second and the third, is at most $1 - \epsilon^{3/2}$.

If there exists $y \in C$ for which this presumably typical event fails to occur, we say that the $(n, * : f, 1 - \epsilon^{3/2})$ -canopy C experiences *late* coalescence. How may we bound above the probability of late coalescence? When this event takes place, there exists an element $y \in C$ such that the above triple of polymers remains disjoint throughout the strip $\mathbb{R} \times [0, 1 - \epsilon^{3/2}]$. These three polymers are heading to a shared endpoint at time one, namely y . The two-thirds power law for polymer fluctuation, given rigorous expression by Proposition 4.9, implies that, typically, the three polymers at time $1 - \epsilon^{3/2}$ will lie within a distance of order ϵ of y . Since the polymers begin at r^- , r and r^+ , which are locations lying in a given interval of length ϵ , we see that the polymers restricted to the strip in question begin and end within distance ϵ of each other, and remain pairwise disjoint.

When the length- ϵ intervals at which such a triple of polymers begin and end are given, we may use Theorem 4.1 to find an upper bound on the probability of the existence of such disjoint polymers. Taking $m = 3$ in that result, we find an upper bound of $\epsilon^{(3^2-1)/2+o(1)} = \epsilon^{4+o(1)}$.

Now if we use this approach to determine an upper bound on the probability that there exists an $(n, * : f, 1 - \epsilon^{3/2})$ -canopy that intersects a given unit interval and for which the late coalescence event occurs, we must sum the $\epsilon^{4+o(1)}$ bound over an order of ϵ^{-1} such canopies (each, after all, has length of order ϵ) as well as an order of ϵ^{-1} time-zero length- ϵ intervals that capture the root location associated to a given canopy. Thus an entropy term of ϵ^{-2} appears, leading to an upper bound of the form $\epsilon^{2+o(1)}$ on this event of a uniform absence of late coalescence.

The battle between 4 and -2 has ended with a positive outcome, showing, as desired, the rarity of any instance of late coalescence. Note that the use of the advanced time $1 - \epsilon^{3/2}$ and of a triple, rather than merely a pair, of polymers is needed for this to happen.

The rerooting procedure will then be applied in this typical scenario of absence of late coalescence. For any location y belonging to one of the new canopies C , either the polymers $\rho_{n;(r^-,0)}^{(y,1)}$ and $\rho_{n;(r,0)}^{(y,1)}$ meet before time $1 - \epsilon^{3/2}$, or the polymers $\rho_{n;(r,0)}^{(y,1)}$ and $\rho_{n;(r^+,0)}^{(y,1)}$ do. Suppose for definiteness that the former case applies. If we vary y within the order- ϵ length interval C , the various polymers emanating from $(r, 0)$ will share a common course until their branch time, which exceeds $1 - \epsilon^{3/2}$. Thus, the weight profiles $y \rightarrow \text{Wgt}_{n;(r^-,0)}^{(y,1)}$ and $y \rightarrow \text{Wgt}_{n;(r,0)}^{(y,1)}$ would appear to differ merely by a random constant as y varies over C : see the right sketch in Figure 5. If this can be rigorously demonstrated, it will allow us to treat the $(n, * : f, 1 - \epsilon^{3/2})$ -canopies as the patches in our Brownian patchwork quilt, with the associated fabric sequence element being either $y \rightarrow \text{Wgt}_{n;(r^-,0)}^{(y,1)}$ or $y \rightarrow \text{Wgt}_{n;(r^+,0)}^{(y,1)}$. In fact, there is a perhaps surprising counterexample to the assertion that this difference of weight profiles is necessarily constant. An example of a difficulty is given in Figure 6. When we

implement this approach rigorously, we will employ a trick whereby any given $(n, * : f, 1 - \epsilon^{3/2})$ -canopy will be broken into two, in a way that ensures the desired constancy of the two weight profiles on the resulting pieces.

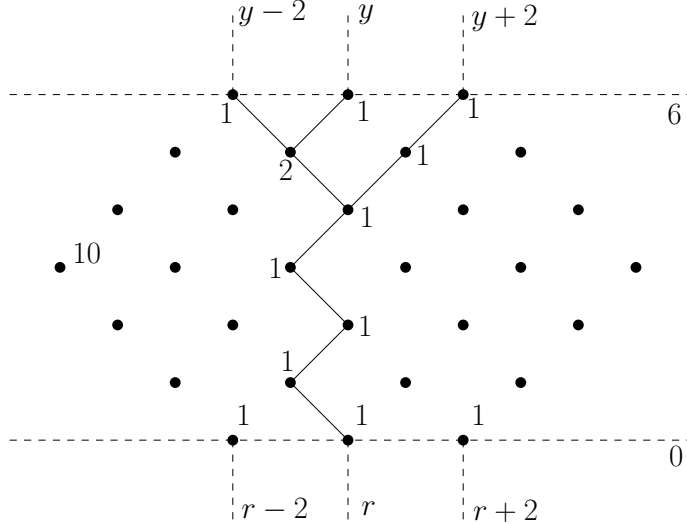


FIGURE 6. In this discrete example of an LPP model, directed paths move from height zero to height six by means of diagonal adjacency. A path’s value is naturally the sum of the lattice values that it finds in its way. All lattice values are zero except those indicated. Depicted is a geodesic tree T , rooted at $(r, 0)$, with a three element canopy $\{y - 2, y, y + 2\}$ at height six. (As it happens, $y = r$.) The tree has branch time equal to four. The canopy does not experience late coalescence, in the sense that, for each $z \in \{y - 2, y, y + 2\}$, either the pair of geodesics from $(r - 2, 0)$ and $(r, 0)$ to $(z, 6)$ merges by time four, or the pair from $(r, 0)$ and $(r + 2, 0)$ does. Despite this, the difference between the geodesic value profiles that map $z \in \{y - 2, y, y + 2\}$ to the value of the geodesic from $(u, 0)$ to $(z, 6)$, for $u = r$ and $u = r - 2$, is not constant. Indeed, the geodesic from $(r - 2, 0)$ to $(y - 2, 6)$ collects the value of ten located at $(r - 5, 3)$, while such an opportunity is not available to the geodesic from $(r, 0)$ to $(y - 2, 6)$. The canopy can however be split into two subintervals so that this problem of non-constancy does not occur.

It is these rerooting ideas that we will develop to resolve the second objection in the rough guide. The quantity $\epsilon > 0$ will be chosen to be random in fact: we will search for a choice on successively smaller dyadic scales until the uniform absence of late coalescence is achieved. Sometimes ϵ will end up being very small, and then the ϵ -mesh locations, such as r^- in the preceding discussion, will number a large quantity of order ϵ^{-1} . This will cause some deterioration in comparison to Brownian bridge when weight profiles such as $y \rightarrow \text{Wgt}_{n;(r^-,0)}^{(y,1)}$ are so compared by means of Theorem 4.3. The result is that while this theorem implies that the weight profile $y \rightarrow \text{Wgt}_{n;(x,0)}^{(y,1)}$ for a given choice of x withstands $L^{\infty-}$ comparison to Brownian bridge, the fabric sequence elements in our Brownian quilt representation of the f -rewarded weight profiles will merely be proved to withstand L^{3-} comparison to Brownian bridge.

6. RARITY OF LATE COALESCENCE

In this section, we specify the late coalescence event. We then state and prove Proposition 6.1; the proof will use Lemma 6.2. The proposition asserts the discussed $\epsilon^{2+o(1)}$ upper bound on the probability of uniform absence of late coalescence. The lemma asserts that it is typical in the case of this absence that there exists a triple of disjoint polymers crossing the strip $\mathbb{R} \times [0, 1 - \epsilon^{3/2}]$ with ϵ -distant endpoints. Thus, the role of the lemma is in essence to reduce the proposition to an application of Theorem 4.2.

For $n \in \mathbb{N}$, $\epsilon > 0$ and $K > 0$, define the *late coalescence* event $\text{LateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon)$ to be the event that there exist $x \in [-K, K]$ and $y \in [-1, 1]$ such that,

- setting $i = \lfloor x\epsilon^{-1} \rfloor$, each of the pairs $(i\epsilon, y)$, (x, y) and $((i+1)\epsilon, y)$ lies in $\text{PolyUnique}_{n,0}^1$, which is to say, each of the polymers $\rho_{n;(i\epsilon,0)}^{(y,1)}$, $\rho_{n;(x,0)}^{(y,1)}$ and $\rho_{n;((i+1)\epsilon,0)}^{(y,1)}$ is well defined;
- and, for any pair among this triple of polymers, the intersection of the pair contains no point whose y -coordinate is less than or equal to $1 - \epsilon^{3/2}$.

We write $\text{NoLateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon)$ for the complementary event $\neg \text{LateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon)$. This last event is the presumably typical event under which there is no late coalescence in the concerned unit-order region that we discussed in the preceding heuristic.

Proposition 6.1. *Let $K \in [2, \infty)$, $\epsilon > 0$ and $n \in 2\mathbb{N}$ satisfy $n\epsilon^{3/2} \in \mathbb{N}$,*

$$\epsilon \leq \min \left\{ 4^{-2}(K+2)^{-4}, (\eta_0)^{72}, 10^{-1342}c_3^{44}, \exp \left\{ -2C^{3/8} \right\} \right\},$$

and

$$n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606}c_3^{-84} \epsilon^{-222}(K+2)^{36}, a_0^{-9}(K+2)^9 2^6 \right\}.$$

Then

$$\mathbb{P} \left(\text{LateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon) \right) \leq \epsilon^2 \cdot 10^{420} K c_3^{-33} C_3 (\log \epsilon^{-1})^{42} \exp \left\{ \beta_3 (\log \epsilon^{-1})^{5/6} \right\}.$$

(The positive constants η_0 , K_0 , a_0 and β_3 are each specified either in Theorem 4.1 or Theorem 4.2.)

Proof. The next result is an important component.

Lemma 6.2. *Let $r > 0$. For $\epsilon \in (0, 2^{-2/3}]$, the occurrence of*

$$\text{LateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon) \cap \text{PolyDevReg}_{n;([-K-1,K+1],0)}^{([-1,1],1)}(1 - \epsilon^{3/2}, r) \quad (7)$$

entails the existence of a pair (i, j) of integers satisfying $|i| \leq K\epsilon^{-1} + 1$ and $|j| \leq \epsilon^{-1} + 1$ such that

$$\text{MaxDisjtPoly}_{n;([i\epsilon, (i+1)\epsilon], 0)} \left([j\epsilon - (r+1)\epsilon - (K+2)\epsilon^{3/2}, j\epsilon + (r+1)\epsilon + (K+2)\epsilon^{3/2}], 1 - \epsilon^{3/2} \right) \geq 3. \quad (8)$$

This use of the PolyDevReg event requires that $1 - \epsilon^{3/2} \in n^{-1}\mathbb{Z}$. This holds due to the Proposition 6.1 hypothesis that $n\epsilon^{3/2} \in \mathbb{N}$.

Proof of Lemma 6.2. Suppose that the event (7) occurs.

The occurrence of $\text{LateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon)$ furnishes $x \in [-K, K]$ and $y \in [-1, 1]$. By setting $i = \lfloor x\epsilon^{-1} \rfloor$ and $j = \lfloor y\epsilon^{-1} \rfloor$, note that the pair (i, j) satisfies the bounds stated in the proposition. Moreover, the intersections of the polymers $\rho_{n;(i\epsilon,0)}^{(y,1)}$, $\rho_{n;(x,0)}^{(y,1)}$ and $\rho_{n;((i+1)\epsilon,0)}^{(y,1)}$ with the strip $\mathbb{R} \times [0, 1 - \epsilon^{3/2}]$ form a pairwise disjoint triple of polymers.

To confirm that this polymer triple verifies (8), the event $\text{PolyDevReg}_{n;([-K-1,K+1],0)}^{([-1,1],1)}(1 - \epsilon^{3/2}, r)$ will be invoked to gain understanding of where the endpoints of these three polymers lie. Specifically, we will allow u to take any value in $\{i\epsilon, x, (i+1)\epsilon\}$, and argue that the upper endpoints satisfy

$$|\rho_{n;(u,0)}^{(y,1)}(1 - \epsilon^{3/2}) - j\epsilon| \leq (r+1)\epsilon + (K+2)\epsilon^{3/2}. \quad (9)$$

Since the lower endpoints clearly satisfy $u \in [i\epsilon, (i+1)\epsilon]$, (9) will be sufficient to verify (8).

To begin arguing for (9), note that, since $\epsilon \leq 1$, the event $\text{PolyDevReg}_{n;(u,0)}^{(y,1)}(1 - \epsilon^{3/2}, r)$ occurs whenever $u \in \{i\epsilon, x, (i+1)\epsilon\}$. Since $1 - \epsilon^{3/2} \geq 1/2$, this entails that

$$\left| \rho_{n;(u,0)}^{(y,1)}(1 - \epsilon^{3/2}) - \ell_{(u,0)}^{(y,1)}(1 - \epsilon^{3/2}) \right| \leq r\epsilon$$

for such u . Since our choice of u lies in $[-K-1, K+1]$, and $y \in [-1, 1]$, the gradient of the planar line segment $\ell_{(u,0)}^{(y,1)}$ is a constant that is at least $(K+2)^{-1}$ in absolute value, so that $|\ell_{(u,0)}^{(y,1)}(1 - \epsilon^{3/2}) - y| \leq (K+2)\epsilon^{3/2}$. We see then that $|\rho_{n;(u,0)}^{(y,1)}(1 - \epsilon^{3/2}) - y| \leq r\epsilon + (K+2)\epsilon^{3/2}$, which alongside $|y - j\epsilon| \leq \epsilon$ completes the derivation of (9) and thus also of (8). \square

We let $r > 0$ be a parameter whose value will be specified later in terms of ϵ . For any given pair (i, j) of integers such that $|i| \leq K\epsilon^{-1} + 1$ and $|j| \leq \epsilon^{-1} + 1$, the event in (8) is a subset of

$$\text{MaxDisjtPoly}_{n;([i\epsilon - (r+1)\epsilon - (K+2)\epsilon^{3/2}, j\epsilon + (r+1)\epsilon + (K+2)\epsilon^{3/2}], 1 - \epsilon^{3/2})} \geq \mathbf{3},$$

and the probability of the latter event may be bounded above by applying Theorem 4.2 with parameter settings $\mathbf{t}_1 = 0$, $\mathbf{t}_2 = 1 - \epsilon^{3/2}$, $\mathbf{x} = i\epsilon$, $\mathbf{y} = j\epsilon$, $\mathbf{m} = \mathbf{3}$, and with ϵ chosen so that $2(\mathbf{t}_{1,2})^{2/3}\epsilon = (r+1)\epsilon + (K+2)\epsilon^{3/2}$. Since $\epsilon \leq (1 - 2^{-3/2})^{2/3}$, $(\mathbf{t}_{1,2})^{2/3} \geq 1/2$; also using $\mathbf{t}_{1,2} \leq 1$, we find that

$$\frac{1}{2} \left((r+1)\epsilon + (K+2)\epsilon^{3/2} \right) \leq \epsilon \leq (r+1)\epsilon + (K+2)\epsilon^{3/2}.$$

(This is the first use of boldface notation, explained in Subsection 4.7.1.) Since $r \geq 1$ and $K+2 \leq \epsilon^{-1/2}$, these bounds imply

$$\epsilon \leq \epsilon \leq (r+2)\epsilon. \quad (10)$$

Using these bounds, we learn from this application of Theorem 4.2 that the event in (8) has probability at most

$$(r+2)^4 \epsilon^4 \cdot 10^{288} 3^{135} c_3^{-27} C_3 (\log \epsilon^{-1})^{36} \exp \{ \beta_3 (\log \epsilon^{-1})^{5/6} \}.$$

Since $x - \mathbf{x}$ and $y - \mathbf{y}$ belong to $[0, \epsilon]$, we see that $|\mathbf{x} - \mathbf{y}| \leq |x - y| + \epsilon$.

Regarding the hypotheses that are needed for this application of Theorem 4.2, note that in view of the latter bound in (10), the condition (5) is met provided that

$$(r+2)\epsilon \leq \min \left\{ (\eta_0)^{36}, 10^{-616} c_3^{22} 3^{-115}, \exp \{ -C^{1/4} \} \right\},$$

while since $\mathbf{t}_{1,2} \geq 2^{-1}$ and $|\mathbf{x} - \mathbf{y}| \leq |x - y| + \epsilon \leq K + 2$, (6) is met when

$$n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{584} c_3^{-48} 3^{240} c^{-36} \epsilon^{-222} \max \{1, (K+2)^{36} 2^{24}\}, a_0^{-9} (K+2)^9 2^6 \right\}.$$

Finally, the hypothesis that $(\mathbf{t}_{1,2})^{-2/3} |\mathbf{y} - \mathbf{x}| \leq \epsilon^{-1/2}$ is ensured by $K + 2 \leq 2^{-1}(r + 2)^{-1/2} \epsilon^{-1/2}$, by (10), $(\mathbf{t}_{1,2})^{2/3} \geq 1/2$ and $|x - y| + \epsilon \leq K + 2$.

We now apply Lemma 6.2, taking a union bound over the integer pair (i, j) that the lemma provides, and using the just obtained upper bound on the probability of (8). We learn that

$$\begin{aligned} & \mathbb{P} \left(\text{LateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon) \cap \text{PolyDevReg}_{n;([-K-1,K+1],0)}^{([-1,1],1)}(1 - \epsilon^{3/2}, r) \right) \\ & \leq (2K\epsilon^{-1} + 3)(2\epsilon^{-1} + 3) \cdot (r + 2)^4 \epsilon^4 \cdot 10^{288} 3^{135} c_3^{-27} C_3 (\log \epsilon^{-1})^{36} \exp \{ \beta_3 (\log \epsilon^{-1})^{5/6} \} \\ & \leq \epsilon^2 \cdot 3K \cdot 3 \cdot 3^4 r^4 \cdot 10^{288} 3^{135} c_3^{-27} C_3 (\log \epsilon^{-1})^{36} \exp \{ \beta_3 (\log \epsilon^{-1})^{5/6} \} \\ & \leq \epsilon^2 \cdot 10^{356} K r^4 c_3^{-27} C_3 (\log \epsilon^{-1})^{36} \exp \{ \beta_3 (\log \epsilon^{-1})^{5/6} \}. \end{aligned} \quad (11)$$

In the penultimate inequality, the bounds $\epsilon \leq 1/3$, $K \geq 1$ and $r \geq 1$ were used in the guise $2K\epsilon^{-1} + 3 \leq 3K\epsilon^{-1}$, $2\epsilon^{-1} + 3 \leq 3\epsilon^{-1}$ and $(r + 2)^4 \leq 3^4 r^4$.

We now find an upper bound on the probability of $\neg \text{PolyDevReg}_{n;([-K-1,K+1],0)}^{([-1,1],1)}(1 - \epsilon^{3/2}, r)$ by applying Theorem 4.8. In this application, we will take $\mathbf{t}_1 = 0$, $\mathbf{t}_2 = 1$, $\mathbf{a} = 1 - \epsilon^{3/2}$ and $\mathbf{r} = r/2$; note that $\mathbf{a} \geq 1/2$ and that $(\mathbf{t}_{1,2})^{2/3} (1 - \mathbf{a})^{2/3} = \epsilon$. The product set $[-K - 1, K + 1] \times [-1, 1]$ may be covered by $\lfloor (4(K + 1)\epsilon^{-1} r^{-1} + 1)(4\epsilon^{-1} r^{-1} + 1) \rfloor$ products of the form $[x, x + \epsilon r/2] \times [y, y + \epsilon r/2]$; Theorem 4.8 will be applied with (\mathbf{x}, \mathbf{y}) ranging over such choices of (x, y) . We find then that

$$\mathbb{P} \left(\neg \text{PolyDevReg}_{n;([-K-1,K+1],0)}^{([-1,1],1)}(1 - \epsilon^{3/2}, r) \right) \leq 64(K + 1)\epsilon^{-2} \cdot 22Cr \exp \{ -10^{-11} 2^{-3/4} c_1 r^{3/4} \} \quad (12)$$

where we also used $r \geq 1$, $K \geq 0$ and $\epsilon \leq 1/4$.

In each of these applications of Theorem 4.8, $|\mathbf{x} - \mathbf{y}| \leq K + 2 \leq 2K$ (since $K \geq 2$). Thus, in order that the applications be made, it is sufficient that each of the following conditions is met: $\epsilon^{3/2} \leq 10^{-11} c_1^2$,

$$n \geq \max \left\{ 10^{29} \epsilon^{-75/2} c^{-18}, 10^{24} 2^{36} c^{-18} \epsilon^{-75/2} K^{36} \right\},$$

$$r \geq 2 \max \left\{ 10^9 c_1^{-4/5}, 15C^{1/2}, 22\epsilon^{1/2} K \right\},$$

and $r \leq 6\epsilon^{25/6} n^{1/36}$; n must also be supposed to be even.

We now make the choice $r = 10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3}$ in order that the exponential term in the right-hand side in (12) equals ϵ^4 . We find then that this right-hand side is at most

$$1408(K + 1)C \cdot 10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3} \epsilon^2 \leq 10^{20} K C c_1^{-4/3} (\log \epsilon^{-1})^{4/3} \epsilon^2$$

since $K \geq 1$. Combining this bound with the upper bound in line (11), substituting the value of r , and using $\epsilon \leq e^{-1}$, $c_3 \leq c_1 \leq 1$ and $C_3 \geq C$, completes the proof of Proposition 6.1. \square

7. POLYMER COALESCENCE, CANOPIES AND INTRA-CANOPIE WEIGHT PROFILES

In order to implement the rerooting plan elaborated in Section 5, we need to describe in more precise terms what we mean by the canopy associated to an f -rewarded polymer tree, or rather the revised notion of an $(n, * : f, 1 - \epsilon^{3/2})$ -canopy, and set down notation for coalescence times of polymers. Importantly, we need to make rigorous sense of the notion that the polymer weight profile for polymers emerging from a given root differs by a random constant from this profile where polymers emerge from a nearby mesh point. This task must overcome the challenge presented by the counterexample in Figure 6, which shows that two profiles can fail to differ by a constant even when all the concerned polymers end in a common canopy that does not experience late coalescence.

This section is devoted to resolving these difficulties. After setting polymer and coalescence notation, we present in Lemma 7.4 and its aftermath the concept of an $(n, * : f, s)$ -canopy, where here s denotes a time between zero and one. This new notion of canopy refers to an interval of y -values such that all the f -rewarded line-to-point polymers that end at time one at points y , when viewed in decreasing time, have coalesced by time s . This is a stronger notion than that of canopy, which corresponds to $s = 0$. The new canopies are subsets of the old ones, and each is associated to a root. Setting $s = 1 - \epsilon^{3/2}$, we obtain the revised notion of canopy introduced informally in Section 5. It is our aim to prove that the f -rewarded polymer weight profile on any given $(n, * : f, 1 - \epsilon^{3/2})$ -canopy (with root r) differs by a random constant from the profile where we reroot to either the left or right adjacent mesh point r^- or r^+ . It is at this point that the problem posed by the Figure 6 counterexample rears its head. Lemma 7.5 is the device permitting a solution of this difficulty. It allows us in Definition 7.6 to introduce a certain *special point* into any given $(n, * : f, 1 - \epsilon^{3/2})$ -canopy. The special point splits the canopy into two pieces. In Lemma 7.7, we establish that, provided that the canopy does not experience late coalescence, the f -rewarded weight profile on any such piece differs by the desired random constant from at least one of the two weight profiles of polymers emanating from the neighbouring mesh points r^- and r^+ .

Definition 7.1. Let $(x_1, x_2) \in \mathbb{R}_{\leq}^2$ and $y \in \mathbb{R}$ be such that (x_1, y) and (x_2, y) are elements in $\text{PolyUnique}_{n; t_1}^{t_2}$. Define the *forward coalescence time*

$$\tau_{n; (\{x_1, x_2\}, t_1)}^{\uparrow; (y, t_2)} \in [t_1, t_2]$$

from the pair $\{x_1, x_2\}$ at time t_1 to the element y at time t_2 to be the infimum of the y -coordinates of points in the intersection $\rho_{n; (x_1, t_1)}^{(y, t_2)} \cap \rho_{n; (x_2, t_1)}^{(y, t_2)}$.

Suppose that $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$. Let $y \in [-1, 1]$. First suppose that $y \in \text{PolyUnique}_{n; (*: f, 0)}^1$, so that the polymer $\rho_{n; (*: f, 0)}^{(y, 1)}$ is uniquely defined. For the purpose of stating the next result (the notation is not reserved after that), set $x = \rho_{n; (*: f, 0)}^{(y, 1)}(0)$ and $i = \lfloor \epsilon^{-1}x \rfloor$. Second, we may further suppose that $(i\epsilon, y), ((i+1)\epsilon, y) \in \text{PolyUnique}_{n; 0}^1$. The polymers $\rho_{n; (i\epsilon, 0)}^{(y, 1)}$ and $\rho_{n; ((i+1)\epsilon, 0)}^{(y, 1)}$ are also then well defined. When $y \in [-1, 1]$ is such that both of these suppositions is valid, so that all three polymers are well defined, we call y an (n, ϵ, f) -triple uniqueness point.

Lemma 7.2. Let $\bar{\Psi} \in (0, \infty)^3$ and suppose that $f \in \mathcal{I}_{\bar{\Psi}}$. Let $n \in \mathbb{N}$ satisfy $n > 2^{-3/2}\Psi_1^3 \vee 8(\Psi_2 + 1)^3$, and let $\epsilon \in (0, 1)$.

- (1) The set of (n, ϵ, f) -triple uniqueness points is a subset of $[-1, 1]$ that has full Lebesgue measure.

- (2) Suppose that $\text{NoLateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon) \cap \text{RegFluc}_{n;(*,0)}^{\{(-1,1],1\}}(K-1)$ occurs. Let $y \in [-1, 1]$ be an (n, ϵ, f) -triple uniqueness point, and associate (x, i) to y as we just did. Then at least one of the quantities $\tau_{n;(\{i\epsilon, x\}, 0)}^{\uparrow; (y, 1)}$ and $\tau_{n;(\{x, (i+1)\epsilon\}, 0)}^{\uparrow; (y, 1)}$ is at most $1 - \epsilon^{3/2}$.

Proof. (2): The f -rewarded line-to-point unit-lifetime polymers that end at $(-1, 1)$ and $(1, 1)$ are almost surely unique by Lemma 4.6(2). Thus the polymer sandwich Lemma 4.4 shows that the occurrence of $\text{RegFluc}_{n;(*,0)}^{\{(-1,1],1\}}(K-1)$ entails that $x \in [-K, K]$. Each of the polymers $\rho_{n;(i\epsilon, 0)}^{(y, 1)}$, $\rho_{n;(*, 0)}^{(y, 1)}$ and $\rho_{n;((i+1)\epsilon, 0)}^{(y, 1)}$ is well defined. When $\text{NoLateCoal}_{n;([-K,K],0)}^{([-1,1],1)}(\epsilon)$ occurs, either the first and second, or the second and third, of these polymers have a point of intersection at y -coordinate at most $1 - \epsilon^{3/2}$. Thus, either $\tau_{n;(\{i\epsilon, x\}, 0)}^{\uparrow; (y, 1)}$ is at most $1 - \epsilon^{3/2}$, or $\tau_{n;(\{x, (i+1)\epsilon\}, 0)}^{\uparrow; (y, 1)}$ satisfies this bound. \square

Definition 7.3. Let $x \in \mathbb{R}$ and $(y_1, y_2) \in \mathbb{R}_{\leq}^2$ be such that (x, y_1) and (x, y_2) are elements in $\text{PolyUnique}_{n; t_1}^{t_2}$. Define the *backward coalescence time*

$$\tau_{n;(x, t_1)}^{\downarrow; (\{y_1, y_2\}, t_2)} \in [t_1, t_2]$$

from the pair $\{y_1, y_2\}$ at time t_2 to the element x at time t_1 to be the supremum of the y -coordinates of points in the intersection $\rho_{n;(x, t_1)}^{(y_1, t_2)} \cap \rho_{n;(x, t_1)}^{(y_2, t_2)}$.

Let $n \in \mathbb{N}$, $t > 0$ and $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$. Suppose that $(y_1, y_2) \in \mathbb{R}_{\leq}^2$ satisfies $y_1, y_2 \in \text{PolyUnique}_{n;(*, f, 0)}^t$. Define the f -rewarded line-to-point backward coalescence time

$$\tau_{n;(*, f, 0)}^{\downarrow; (\{y_1, y_2\}, t)} \in [0, t]$$

from the pair $\{y_1, y_2\}$ at time t (to time zero) to be the supremum of the y -coordinates of points in the intersection $\rho_{n;(*, f, 0)}^{(y_1, t)} \cap \rho_{n;(*, f, 0)}^{(y_2, t)}$.

Suppose that $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$. Recall from Lemma 4.6(2) that when $n \in \mathbb{N}$ satisfies $n > 2^{-3/2}\bar{\Psi}_1^3 \vee 8(\bar{\Psi}_2 + 1)^3$, the set $\text{PolyUnique}_{n;(*, f, 0)}^1$ is a full Lebesgue measure subset of $[-1, 1]$.

Let $s \in (0, 1)$. We identify a subset $R_{n;(*, f, 0)}^1(s)$ of $\text{PolyUnique}_{n;(*, f, 0)}^1 \times \text{PolyUnique}_{n;(*, f, 0)}^1$ by declaring that an element (y_1, y_2) in this product set belongs to $R_{n;(*, f, 0)}^1(s)$ if and only if $\tau_{n;(*, f, 0)}^{\downarrow; (\{y_1, y_2\}, 1)} \geq s$.

Lemma 7.4. Suppose that $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$. For $n \in \mathbb{N}$, write $y_1 R y_2$ for $(y_1, y_2) \in R_{n;(*, f, 0)}^1(s)$. For $n > 2^{-3/2}\bar{\Psi}_1^3 \vee 8(\bar{\Psi}_2 + 1)^3$,

- (1) R is an equivalence relation on $\text{PolyUnique}_{n;(*, f, 0)}^1$.
- (2) Let Q denote any equivalence class of the relation R . There exists an interval $I \subseteq [-1, 1]$ such that $Q = I \cap \text{PolyUnique}_{n;(*, f, 0)}^1$.

Remark. According to the lemma, the interval I could be a singleton, or it might contain at least one of its endpoints. It is in fact not difficult to exclude these circumstances. For example, if $I = \{a\}$, then we might consider the limit as $\epsilon \searrow 0$ of the left-neighbouring line-to-point polymers $\rho_{n;(*, f, 0)}^{(a-\epsilon, 1)}$. It can be argued that this limit, which we might denote by $\rho_{n;(*, f, 0)}^{(a-, 1)}$, is a polymer. Ordinarily, it would equal $\rho_{n;(*, f, 0)}^{(a, 1)}$. Because $I = \{a\}$, however, the new polymer has no point of intersection with

$\rho_{n;(*:f,0)}^{(a,1)}$ at y -coordinate exceeding s except at $(a, 1)$. A right limiting polymer $\rho_{n;(*:f,0)}^{(a+,1)}$ satisfies the same property. Thus, a is not an element of $\text{PolyUnique}_{n;(*:f,0)}^1$. Since Q is not empty, I cannot then be a singleton. Similarly, one may find two (but not three) distinct polymers emanating from $(a, 1)$ if $I = [a, b)$ with $a < b$. Although these ideas serve to elucidate the structure of the canopies, we do not need them in this article, so the details of this argument will be omitted.

Proof of Lemma 7.4. (1): Transitivity of the relation follows because $\tau_{n;(*:f,t_1)}^{\downarrow;(\{y_1,y_3\},t_2)} \geq \tau_{n;(*:f,t_1)}^{\downarrow;(\{y_1,y_2\},t_2)} \wedge \tau_{n;(*:f,t_1)}^{\downarrow;(\{y_2,y_2\},t_2)}$; the other axioms are trivially verified.

(2): Note first that $\text{PolyUnique}_{n;(*:f,0)}^1$ is dense in $[-1, 1]$ by Lemma 4.6(2). Now let (y_1, y_2, y_3) be an increasing sequence of elements of $\text{PolyUnique}_{n;(*:f,0)}^1$ such that $y_1 R y_3$. It suffices to argue that $y_1 R y_2$. Lemma 4.4 implies that $\rho_{n;(*:f,0)}^{(y_1,1)} \preceq \rho_{n;(*:f,0)}^{(y_2,1)} \preceq \rho_{n;(*:f,0)}^{(y_3,1)}$ while the first and third of these polymers are seen to coincide in the strip $\mathbb{R} \times [0, s]$ due to $y_1 R y_3$. Thus, the second polymer also coincides with the other two in this way, so that $y_1 R y_2$. \square

In the circumstances of Lemma 7.4, the interior of the closure of any equivalence class of the relation R is an open real interval, because $\text{PolyUnique}_{n;(*:f,0)}^1$ is dense in $[-1, 1]$. Any such open interval will be called an $(n, * : f, s)$ -canopy. Whenever C is an $(n, * : f, s)$ -canopy, the polymer endpoint $\rho_{n;(*:f,0)}^{(x,1)}(0)$ is independent of $x \in C \cap \text{PolyUnique}_{n;(*:f,0)}^1$. The shared endpoint will be called the root of the canopy C .

Lemma 7.5. *Suppose that $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$. Let $n \in \mathbb{N}$, $s \in (0, 1)$, and let C denote an arbitrary $(n, * : f, s)$ -canopy. Write $C = (y_1, y_2)$ with $y_1, y_2 \in \mathbb{R}$ with $y_1 < y_2$, and denote by r the root of C . Let $y, z \in (y_1, y_2)$, $y < z$, be two elements of $\text{PolyUnique}_{n;(*:f,0)}^1$.*

- (1) *Let $r_- \in \mathbb{R}$ satisfy $r_- < r$, $(r_-, y), (r_-, z) \in \text{PolyUnique}_{n;0}^1$ and $\tau_{n;(\{r_-, r\}, 0)}^{\uparrow;(y,1)} \leq s$. Then the polymer $\rho_{n;(r_-, 0)}^{(z,1)}$ is equal to the concatenation of $\rho_{n;(r_-, 0)}^{(y,1)} \cap (\mathbb{R} \times [0, s])$ and $\rho_{n;(r, 0)}^{(z,1)} \cap (\mathbb{R} \times [s, 1])$.*
- (2) *Let $r_+ \in \mathbb{R}$ satisfy $r < r_+$, $(r_+, y), (r_+, z) \in \text{PolyUnique}_{n;0}^1$ and $\tau_{n;(\{r, r_+\}, 0)}^{\uparrow;(z,1)} \leq s$. Then the polymer $\rho_{n;(r_+, 0)}^{(y,1)}$ is equal to the concatenation of $\rho_{n;(r_+, 0)}^{(z,1)} \cap (\mathbb{R} \times [0, s])$ and $\rho_{n;(r, 0)}^{(y,1)} \cap (\mathbb{R} \times [s, 1])$.*

Proof. The two parts of the lemma are proved similarly and we prove only the first part. To do so, note that, since $r_- \leq r$ and $y \leq z$, Lemma 4.4 implies that

$$\rho_{n;(r_-, 0)}^{(y,1)} \preceq \rho_{n;(r_-, 0)}^{(z,1)} \preceq \rho_{n;(r, 0)}^{(z,1)}. \quad (13)$$

Note that

$$\tau_{n;(\{r_-, r\}, 0)}^{\uparrow;(y,1)} \leq s \leq \tau_{n;(*:f,0)}^{\downarrow;(\{y,z\}, 1)} = \tau_{n;(r, 0)}^{\downarrow;(\{y,z\}, 1)};$$

abbreviating $\tau_- = \tau_{n;(\{r_-, r\}, 0)}^{\uparrow;(y,1)}$ and $\tau_+ = \tau_{n;(r, 0)}^{\downarrow;(\{y,z\}, 1)}$, we see that $s \in [\tau_-, \tau_+]$.

Note that the polymers $\rho_{n;(r, 0)}^{(y,1)}$ and $\rho_{n;(r, 0)}^{(z,1)}$ coincide when intersected with the strip $\mathbb{R} \times [0, \tau_+]$, and that the polymers $\rho_{n;(r_-, 0)}^{(y,1)}$ and $\rho_{n;(r_-, 0)}^{(z,1)}$ coincide when intersected with $\mathbb{R} \times [\tau_-, 1]$. Thus, $\rho_{n;(r_-, 0)}^{(y,1)}$

coincides with $\rho_{n;(r,0)}^{(z,1)}$ on $\mathbb{R} \times [\tau_-, \tau_+]$, which in view of (13) implies that

$$\rho_{n;(r_-,0)}^{(y,1)} \cap (\mathbb{R} \times [\tau_-, \tau_+]) = \rho_{n;(r_-,0)}^{(z,1)} \cap (\mathbb{R} \times [\tau_-, \tau_+]) = \rho_{n;(r,0)}^{(z,1)} \cap (\mathbb{R} \times [\tau_-, \tau_+]). \quad (14)$$

The intersection of $\rho_{n;(r_-,0)}^{(z,1)}$ with $\mathbb{R} \times [\tau_+, 1]$ is a polymer whose endpoints coincide with the intersection of $\rho_{n;(r,0)}^{(z,1)}$ with this strip. Because there is only one polymer, namely $\rho_{n;(r,0)}^{(z,1)}$, that begins at $(r, 0)$ and ends at $(z, 1)$, nor can there be more than one polymer whose endpoints are a given pair of points in $\rho_{n;(r,0)}^{(z,1)}$. Thus, we learn that $\rho_{n;(r_-,0)}^{(z,1)} \cap (\mathbb{R} \times [\tau_+, 1]) = \rho_{n;(r,0)}^{(z,1)} \cap (\mathbb{R} \times [\tau_+, 1])$. By similar reasoning, $\rho_{n;(r_-,0)}^{(y,1)} \cap (\mathbb{R} \times [0, \tau_-]) = \rho_{n;(r_-,0)}^{(z,1)} \cap (\mathbb{R} \times [0, \tau_-])$.

We have learnt that $\rho_{n;(r_-,0)}^{(z,1)}$ is the concatenation of the polymers $\rho_{n;(r_-,0)}^{(y,1)} \cap (\mathbb{R} \times [0, \tau_-])$ and $\rho_{n;(r,0)}^{(z,1)} \cap (\mathbb{R} \times [\tau_-, 1])$. By (14), this assertion also holds when τ_- is replaced by any element of $[\tau_-, \tau_+]$. Choosing this element to be s completes the proof of Lemma 7.5(1). \square

Definition 7.6. Let $\epsilon \in (0, 1)$, and let C denote an arbitrary $(n, * : f, 1 - \epsilon^{3/2})$ -canopy. Write $C = (x_1, x_2)$ with $x_1, x_2 \in \mathbb{R}$ with $x_1 < x_2$; denote by r the root of C ; and set $r_- = \epsilon \lfloor r\epsilon^{-1} \rfloor$. Define the *special point* spec of C to be the infimum of the values of $x \in (x_1, x_2) \cap \text{PolyUnique}_{n;(*:f,0)}^1$ such that $(r_-, x) \in \text{PolyUnique}_{n;0}^1$ and

$$\tau_{n;(\{r_-, r\}, 0)}^{\uparrow; (x, 1)} \leq 1 - \epsilon^{3/2}.$$

We take $\text{spec} = x_2$ when no such value of x exists. Thus, spec is a random point that almost surely lies in $[x_1, x_2]$.

Lemma 7.7. Suppose that $f \in \mathcal{I}_{\bar{\Psi}}$ for some $\bar{\Psi} \in (0, \infty)^3$, and that $n \in \mathbb{N}$ satisfies $n > 2^{-3/2} \bar{\Psi}_1^3 \vee 8(\bar{\Psi}_2 + 1)^3 \vee 8(K + 2)^3$. Let $\epsilon \in (0, 1)$ and $K \geq 1$. Suppose that the event

$$\text{NoLateCoal}_{n;([-K, K], 0)}^{([-1, 1], 1)}(\epsilon) \cap \text{RegFluc}_{n;(*:f, 0)}^{(\{-1, 1\}, 1)}(K - 1)$$

occurs for a given value of $K \geq 1$.

Let C denote an arbitrary $(n, * : f, 1 - \epsilon^{3/2})$ -canopy that intersects $[-1, 1]$. Write $C = (x_1, x_2)$ with $x_1, x_2 \in \mathbb{R}$ with $x_1 < x_2$; denote by r the root of C , and set $r_- = \epsilon \lfloor r\epsilon^{-1} \rfloor$ as well as $r_+ = \epsilon \lceil r\epsilon^{-1} \rceil$. Write $\text{spec} = \text{spec}_C \in [x_1, x_2]$ for the special point of C .

The value of $\text{Wgt}_{n;(r,0)}^{(x,1)} - \text{Wgt}_{n;(r_-,0)}^{(x,1)}$ is independent of $x \in (\text{spec}, x_2) \cap [-1, 1]$, and the value of $\text{Wgt}_{n;(r,0)}^{(x,1)} - \text{Wgt}_{n;(r_+,0)}^{(x,1)}$ is independent of $x \in (x_1, \text{spec}) \cap [-1, 1]$.

Proof. We begin by arguing that the value of $\text{Wgt}_{n;(r,0)}^{(z,1)} - \text{Wgt}_{n;(r_-,0)}^{(z,1)}$ is independent of choices of $z \in (\text{spec}, x_2) \cap [-1, 1]$ for which $(r^-, z), (r, z) \in \text{PolyUnique}_{n;0}^1$. (Such choices will be called z values in the argument that follows. The set of z values is dense by Lemma 4.6(1) and (2), where the first part of the lemma is applicable because $n \geq 8(K + 2)^3$ and $|r^- - z| \leq K + 2$.) To establish this constancy, note that, for any such z , there exists, by the definition of spec , an element $y \in (\text{spec}, z)$ such that $(r, y), (r^-, y) \in \text{PolyUnique}_{n;0}^1$ and $\tau_{n;(\{r_-, r\}, 0)}^{\uparrow; (y, 1)} \leq 1 - \epsilon^{3/2}$. Applying

Lemma 7.5(1) with $\mathbf{s} = 1 - \epsilon^{3/2}$, we find that the polymer $\rho_{n;(r_-,0)}^{(z,1)}$ is equal to the concatenation of $\rho_{n;(r_-,0)}^{(y,1)} \cap (\mathbb{R} \times [0, 1 - \epsilon^{3/2}])$ and $\rho_{n;(r,0)}^{(z,1)} \cap (\mathbb{R} \times [1 - \epsilon^{3/2}, 1])$. The polymer $\rho_{n;(r,0)}^{(z,1)}$ may of course also be viewed as a concatenation of its subpaths in these two strips. The difference in weight

$\text{Wgt}_{n;(r,0)}^{(z,1)} - \text{Wgt}_{n;(r_-,0)}^{(z,1)}$ between these two polymers is thus seen to equal the difference in weight between $\rho_{n;(r_-,0)}^{(y,1)} \cap (\mathbb{R} \times [0, 1 - \epsilon^{3/2}])$ and $\rho_{n;(r,0)}^{(y,1)} \cap (\mathbb{R} \times [0, 1 - \epsilon^{3/2}])$. The latter quantity appears to be independent of z , although it does depend on y , which was determined by z . However, we may introduce $\epsilon > 0$, and insist that $y \in (\text{spec}, \text{spec} + \epsilon)$. So doing, we see that the quantity is indeed independent of the z -value z , provided that $z \geq \text{spec} + \epsilon$. Since $\epsilon > 0$ is arbitrary, the quantity is in fact independent of the z -value without restriction.

That the quantity $\text{Wgt}_{n;(r,0)}^{(x,1)} - \text{Wgt}_{n;(r_-,0)}^{(x,1)}$ is independent of $x \in (\text{spec}, x_2)$ now follows, because the set of z we have been considering is dense in (spec, x_2) , and this quantity is a continuous function of x in this interval by Lemma 4.7.

It remains to argue that $\text{Wgt}_{n;(r,0)}^{(x,1)} - \text{Wgt}_{n;(r_+,0)}^{(x,1)}$ is independent of $x \in (x_1, \text{spec}) \cap [-1, 1]$. To establish this, we first prove that

$$\text{Wgt}_{n;(r,0)}^{(z,1)} - \text{Wgt}_{n;(r_+,0)}^{(z,1)} \text{ is independent of choices of } z \in (x_1, \text{spec}) \quad (15)$$

where $z \in [-1, 1]$ is such that (r^-, z) , (r, z) and (r^+, z) are elements of $\text{PolyUnique}_{n,0}^1$. (In a reuse of terminology, any such z will be called a z -value in the ensuing argument. Again, it is Lemma 4.6(1) and (2) that assure that the set of z -values is dense.) To verify this assertion, note that, since each such z is less than spec , $\tau_{n;(\{r_-,r\},0)}^{\uparrow;(z,1)} > 1 - \epsilon^{3/2}$. However, Lemma 7.2 implies that one or other of $\tau_{n;(\{r_-,r\},0)}^{\uparrow;(z,1)}$ and $\tau_{n;(\{r,r_+\},0)}^{\uparrow;(z,1)}$ is at most $1 - \epsilon^{3/2}$. We learn then that $\tau_{n;(\{r,r_+\},0)}^{\uparrow;(z,1)} \leq 1 - \epsilon^{3/2}$. This inference permits us consider any pair (z_1, z_2) , $z_1 < z_2$, of z -values, and to turn to Lemma 7.5(2) with $\mathbf{y} = z_1$, $\mathbf{z} = z_2$ and $\mathbf{s} = 1 - \epsilon^{3/2}$. From this lemma, we find that the polymer weight $\text{Wgt}_{n;(r_+,0)}^{(z_1,1)}$ is the sum of the weights of the polymers $\rho_{n;(r_+,0)}^{(z_2,1)} \cap (\mathbb{R} \times [0, 1 - \epsilon^{3/2}])$ and $\rho_{n;(r,0)}^{(z_1,1)} \cap (\mathbb{R} \times [1 - \epsilon^{3/2}, 1])$. The weight difference $\text{Wgt}_{n;(r_+,0)}^{(z_1,1)} - \text{Wgt}_{n;(r,0)}^{(z_1,1)}$ is thus seen to equal the difference of the weights of the polymers $\rho_{n;(r_+,0)}^{(z_2,1)} \cap (\mathbb{R} \times [0, 1 - \epsilon^{3/2}])$ and $\rho_{n;(r,0)}^{(z_2,1)} \cap (\mathbb{R} \times [0, 1 - \epsilon^{3/2}])$. This quantity is independent of z_1 , though it does depend on z_2 . However, z_2 may be chosen to be a z -value arbitrarily close to spec , by the density of such values, rendering the quantity independent of z_1 to the left of any neighbourhood of spec . Since the neighbourhood is arbitrary, we confirm (15).

That $\text{Wgt}_{n;(r,0)}^{(x,1)} - \text{Wgt}_{n;(r_+,0)}^{(x,1)}$ is independent of $x \in (x_1, \text{spec})$ follows from the density of z -values in (x_1, spec) and the polymer weight continuity Lemma 4.7. This completes the proof of Lemma 7.7. \square

8. WELL-BEHAVED CANOPY STRUCTURES ARE TYPICAL

We are almost ready to rigorously construct the f -rewarded polymer weight profiles as Brownian patchwork quilts (that is, to prove Theorem 1.2). This section is devoted to some final preparations for this task. Recall that it is settled that the patches that will make up the Brownian quilts will be the intervals obtained by splitting $(n, * : f, 1 - \epsilon^{3/2})$ -canopies at the special points inside these canopies; (the split pieces will be called split canopies when they are formally introduced in the context of the upcoming proof, in Section 9). Recall also that we will actually select the value of $\epsilon > 0$ to be random, by searching through consecutively smaller dyadic scales until a certain favourable event is realized. This event will include the absence of late coalescence, but that is not

the only desired feature. For example, we will also insist that the number of $(n, * : f, 1 - \epsilon^{3/2})$ -canopies not be too high, because this cardinality is in essence the size of the random stitch point set in the constructed quilt, which we wish to argue is not typically too big.

In this section, then, we offer a definition of a *normal coalescence* event, which will play the role of this favourable event. Lemma 8.1 is the only result in the section. It establishes that the normal coalescence event fails with the same order of probability, $\epsilon^{2+o(1)}$, that the event of absence of late coalescence does. This is because the other favourable features that we build into the normal coalescence event, such as the $(n, * : f, 1 - \epsilon^{3/2})$ -canopy cardinality not being abnormally high, are highly typical scenarios.

Let $\bar{\Psi} \in (0, \infty)^3$ and $f \in \mathcal{I}_{\bar{\Psi}}$. For $n \in \mathbb{N}$ and $s \in (0, 1)$, let $\text{Canopy}\#_{n, (*:f,0)}^{([-1,1],1)}(s)$ denote the cardinality of the set of $(n, * : f, s)$ -canopies that intersect $[-1, 1]$.

For $n \in \mathbb{N}$, $\epsilon \in (0, 1)$, $\chi > 0$ and $D > 0$, define the *normal coalescence* event $\text{NormalCoal}_{n, (*:f,0)}^{([-1,1],1)}(D, \epsilon, \chi)$ to be

$$\begin{aligned} \text{NoLateCoal}_{n, ([-D(\log \epsilon^{-1})^{1/3}, D(\log \epsilon^{-1})^{1/3}], 0)}^{([-1,1],1)}(\epsilon) \cap \text{RegFluc}_{n, (*:f,0)}^{\{[-1,1],1\}}(D(\log \epsilon^{-1})^{1/3} - 1) \\ \cap \left\{ \text{Canopy}\#_{n, (*:f,0)}^{([-1,1],1)}(1 - \epsilon^{3/2}) \leq \epsilon^{-1-\chi} \right\}. \end{aligned}$$

The unappetizing aspect of explicit hypothesis bounds is perhaps most conspicuous in the next result (and correspondingly, the calculational derivation of the result in Appendix B is rather long): the basic meaning is that, with D and χ fixed to be large and small, NormalCoal fails with probability at most $\epsilon^{2+o(1)}$, uniformly in high n .

Lemma 8.1. *Suppose that $n \in 2\mathbb{N}$, $\epsilon > 0$, $D > 0$, $\chi > 0$ and $\bar{\Psi} \in (0, \infty)^3$ satisfy the bounds*

$$\begin{aligned} \epsilon \leq \max \left\{ (\eta_0)^{72}, (10^{-41} \chi^{36} D^{-14}) \chi^{-2}, \exp \left\{ -\beta \chi^{-1/2} (3480)^{1/2} \right\}, \right. \\ \left. (2\hat{H})^{-3480} \chi^{-1} (\log \beta)^2, (4k_0)^{-\chi^{-1}}, \exp \left\{ -2K_0^2 \right\}, 10^{-1779} c_3^{44}, \right. \\ \left. \exp \left\{ -D^{-3} \left(78\Psi_1 \vee 4((\Psi_2 + 1)^2 + \Psi_3)^{1/2} \right)^3 \right\}, \exp \left\{ -10^7 c_1^{-9} C^{3/8} \right\} \right\}, \end{aligned}$$

$$n \geq \max \left\{ c^{-18} \max \left\{ (\Psi_2 + 1)^9, 10^{23} \Psi_1^9 \right\}, 4(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{618} D^{36} c_3^{-84} a_0^{-9} \epsilon^{-504} \right\},$$

$n\epsilon^{3/2} \in \mathbb{N}$, $D \geq 10^{16} c_1^{-4/3}$ and $\chi \leq 2^{-1} (1 + 500(\log \beta)^2)^{-1}$. Let $f \in \mathcal{I}_{\bar{\Psi}}$. Then

$$\mathbb{P} \left(\neg \text{NormalCoal}_{n, (*:f,0)}^{([-1,1],1)}(D, \epsilon, \chi) \right) \leq \epsilon^2 \cdot \Omega(\epsilon),$$

where

$$\Omega(\epsilon) = 10^{421} c_3^{-33} C_3 D (\log \epsilon^{-1})^{43} \exp \left\{ \beta_3 (\log \epsilon^{-1})^{5/6} \right\}.$$

The positive constants k_0 , η_0 , K_0 , a_0 , β_3 and β are each specified either in Theorem 4.1 or Theorem 4.2. By \hat{H} , we denote the finite supremum $\sup_{i \in \mathbb{N}} H_i \exp \left\{ -2(\log i)^{11/12} \right\}$ associated to the sequence $\{H_i : i \in \mathbb{N}\}$ of constants provided by Theorem 4.1.

Proof. In this analysis, we will consider the event

$$\begin{aligned} & \bigcap_{y \in \epsilon \mathbb{Z} \cap [-1, 1]} \text{PolyDevReg}_{n; (-D(\log \epsilon^{-1})^{1/3}, 0)}^{(y, 1)} \left(1 - \epsilon^{3/2}, D(\log \epsilon^{-1})^{4/3} \right) \\ & \cap \bigcap_{y \in \epsilon \mathbb{Z} \cap [-1, 1]} \text{PolyDevReg}_{n; (D(\log \epsilon^{-1})^{1/3}, 0)}^{(y, 1)} \left(1 - \epsilon^{3/2}, D(\log \epsilon^{-1})^{4/3} \right). \end{aligned}$$

The event $\text{NormalCoal}_{n; (*:f, 0)}^{([-1, 1], 1)}(D, \epsilon, \chi)$ is an intersection of a NoLateCoal event, a RegFluc event and a Canopy\# event. Let ManyCanopy denote the intersection of the several PolyDevReg events just now recorded, this RegFluc event and the complement of the Canopy\# event. Although the new event entails several circumstances, each of these is typical, except for the complement of the Canopy\# event: something reflected in the name ManyCanopy .

The lemma will be proved by noting that

$$\mathbb{P}(\neg \text{NormalCoal}) \leq \mathbb{P}(\text{LateCoal}) + \mathbb{P}(\text{ManyCanopy}) + \mathbb{P}(\neg \text{RegFluc}) + \mathbb{P}(\neg \cap \text{PolyDevReg}), \quad (16)$$

where $\cap \text{PolyDevReg}$ refers to the event in the preceding display and where we omit all adornments of the events specifying NormalCoal . It is the first term on the right-hand side that dictates the bound in Lemma 8.1: it is of order $\epsilon^{2+o(1)}$. The further three terms will be shown to be at least as small.

We now find upper bounds on the four terms on the right-hand side of (16). The order in which we do so is 2134, the analysis of the second term, $\mathbb{P}(\text{ManyCanopy})$, being slightly more involved.

When ManyCanopy occurs, the bound $\text{Canopy\#}_{n; (*:f, 0)}^{([-1, 1], 1)}(1 - \epsilon^{3/2}) > \epsilon^{-1-\chi}$ holds, and we may select from each $(n; (*:f), 1 - \epsilon^{3/2})$ -canopy that intersects $[-1, 1]$ an element of $y \in \text{PolyUnique}_{n; (*:f, 0)}^1 \cap [-1, 1]$. Discarding the last few values if need be, we may list the resulting points y in the form $\{y_i : i \in \llbracket 1, \lceil \epsilon^{-1-\chi} \rceil \rrbracket\}$. The polymers $\rho_{n; (*:f, 0)}^{(y_i, 1)}$ for $i \in \llbracket 1, \lceil \epsilon^{-1-\chi} \rceil \rrbracket$ are then pairwise disjoint when intersected with the strip $\mathbb{R} \times [1 - \epsilon^{3/2}, 1]$. By the occurrence of the RegFluc event and the polymer sandwiching Lemma 4.4, $\rho_{n; (*:f, 0)}^{(y_i, 1)}(0) \in [-D(\log \epsilon^{-1})^{1/3}, D(\log \epsilon^{-1})^{1/3}]$ for all such indices i . For this reason, the intersection of $\rho_{n; (*:f, 0)}^{(y_i, 1)}$ with the strip $\mathbb{R} \times [0, 1 - \epsilon^{3/2}]$ lies on, or to the right of, $\rho_{n; (-D(\log \epsilon^{-1})^{1/3}, 0)}^{(\epsilon \lceil \epsilon^{-1} y_i \rceil, 1)}$. (In specifying the latter polymer, note that $\epsilon \lceil \epsilon^{-1} y_i \rceil$ is the left-displacement of y_i onto an ϵ -mesh. This polymer is almost surely well defined, by Lemma 4.6(1) with $\mathbf{x} = -D(\log \epsilon^{-1})^{1/3}$ and $\mathbf{y} = \epsilon \lceil \epsilon^{-1} y_i \rceil$.) We may thus bound

$$\begin{aligned} \rho_{n; (*:f, 0)}^{(y_i, 1)}(1 - \epsilon^{3/2}) & \geq \rho_{n; (-D(\log \epsilon^{-1})^{1/3}, 0)}^{(\epsilon \lceil \epsilon^{-1} y_i \rceil, 1)}(1 - \epsilon^{3/2}) \\ & \geq (1 - \epsilon^{3/2}) \epsilon \lceil \epsilon^{-1} y_i \rceil - \epsilon^{3/2} D(\log \epsilon^{-1})^{1/3} - \epsilon D(\log \epsilon^{-1})^{4/3} \\ & \geq y_i - \epsilon^{3/2} y_i - \epsilon - 2\epsilon D(\log \epsilon^{-1})^{4/3} \geq y_i - 4\epsilon D(\log \epsilon^{-1})^{4/3}, \end{aligned}$$

where the second bound is due to the occurrence of $\text{PolyDevReg}_{n; (-D(\log \epsilon^{-1})^{1/3}, 0)}^{(y, 1)}(1 - \epsilon^{3/2}, D(\log \epsilon^{-1})^{4/3})$ and $\epsilon^{3/2} \leq 2^{-1}$, the third to $\epsilon \lceil \epsilon^{-1} y_i \rceil \geq y_i - \epsilon$ and $\epsilon \leq e^{-1}$, and the fourth to $y_i \in [-1, 1]$, $D \geq 1$ and $\epsilon \leq e^{-1}$. A similar upper bound on $\rho_{n; (*:f, 0)}^{(y_i, 1)}(1 - \epsilon^{3/2})$ holds, and we find that

$$\left| \rho_{n; (*:f, 0)}^{(y_i, 1)}(1 - \epsilon^{3/2}) - y_i \right| \leq \epsilon \cdot 4D(\log \epsilon^{-1})^{4/3}. \quad (17)$$

The polymers $\rho_{n;(*:f,0)}^{(y_i,1)}$ number at least $\epsilon^{-1-\chi}$. They may be divided according to which of the intervals $[j\epsilon, (j+1)\epsilon)$, $j \in \llbracket 0, \lfloor \epsilon^{-1} \rrbracket$, contains y_i . Thus, there exists $J \in \llbracket 0, \lfloor \epsilon^{-1} \rrbracket$ for which $[J\epsilon, (J+1)\epsilon)$ contains at least $\epsilon^{-1-\chi}(\epsilon^{-1}+1)^{-1} \geq \epsilon^{-\chi}/2$ (where we used $\epsilon \leq 1$). Set

$$I_1 = [J\epsilon - 4D\epsilon(\log \epsilon^{-1})^{4/3}, (J+1)\epsilon + 4D\epsilon(\log \epsilon^{-1})^{4/3}] \quad \text{and} \quad I_2 = [J\epsilon, (J+1)\epsilon].$$

We see from (17) that, for indices i such that $y_i \in [J\epsilon, (J+1)\epsilon)$, the polymers $\rho_{n;(*:f,0)}^{(y_i,1)}$ restricted to the strip $\mathbb{R} \times [1 - \epsilon^{3/2}, 1]$ are disjoint polymers that begin in $I_1 \times \{1 - \epsilon^{3/2}\}$ and end in $I_2 \times \{1\}$. Thus, we find that $\text{MaxDisjtPoly}_{n; (I_1, 1 - \epsilon^{3/2})}^{(I_2, 1)} \geq \epsilon^{-\chi}/2$ occurs. That is, **ManyCanopy** is a subset of

$$\bigcup_{j \in \llbracket 0, \lfloor \epsilon^{-1} \rrbracket} \left\{ \text{MaxDisjtPoly}_{n; ([j\epsilon - 4D\epsilon(\log \epsilon^{-1})^{4/3}, (j+1)\epsilon + 4D\epsilon(\log \epsilon^{-1})^{4/3}], 1 - \epsilon^{3/2})}^{([j\epsilon, (j+1)\epsilon], 1)} \geq 2^{-1}\epsilon^{-\chi} \right\}.$$

For given $j \in \llbracket 0, \lfloor \epsilon^{-1} \rrbracket$, the probability of the **MaxDisjtPoly** event on display may be bounded above by an application of Theorem 4.1. The theorem's parameters are set: $\mathbf{t}_1 = 1 - \epsilon^{3/2}$, $\mathbf{t}_2 = 1$, $\mathbf{x} = j\epsilon - 4D\epsilon(\log \epsilon^{-1})^{4/3}$, $\mathbf{y} = j\epsilon$, $\mathbf{a} = \lceil 8D(\log \epsilon^{-1})^{4/3} \rceil$, $\mathbf{b} = 1$ and $\mathbf{k} = \lfloor 2^{-1}\epsilon^{-\chi} \rfloor$. With $h = \lceil 8D(\log \epsilon^{-1})^{4/3} \rceil$, we find that the probability in question is at most

$$\begin{aligned} & (4^{-1}\epsilon^{-\chi})^{-(145)^{-1}(\log \beta)^{-2}(0 \vee \log \log \epsilon^{-\chi})^2} \cdot h^{(\log \beta)^{-2}(\log \log \epsilon^{-\chi})^2/288+3/2} H_{\lfloor 2^{-1}\epsilon^{-\chi} \rfloor} \\ & \leq (4^{-1}\epsilon^{-\chi})^{-4^{-1}(145)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \cdot h^{(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288+3/2} \hat{H} \exp \{2(\log \epsilon^{-1})^{11/12}\} \\ & \leq \epsilon^{\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \cdot \hat{H}(2h)^{(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288+3/2} \exp \{2(\log \epsilon^{-1})^{11/12}\}, \end{aligned}$$

where recall that we denote by \hat{H} the finite supremum $\sup_{i \in \mathbb{N}} H_i \exp \{-2(\log i)^{11/12}\}$ associated to the sequence $\{H_i : i \in \mathbb{N}\}$ of constants provided by Theorem 4.1. The form of the first expression is obtained by using $\epsilon \leq 4^{-1/\chi}$ in the form $\mathbf{k} \geq 4^{-1}\epsilon^{-\chi}$. The first displayed inequality made use of $0 \vee \log \log \epsilon^{-\chi} \geq 2^{-1} \log \log \epsilon^{-1}$, which is implied by $\log \chi^{-1} \leq 2^{-1} \log \log \epsilon^{-1}$ or equivalently $\epsilon \leq \exp\{-\chi^{-2}\}$; this inequality is also due to $\chi \leq 1$.

For this application of Theorem 4.1 to be made, since $4^{-1}\epsilon^{-\chi} \leq \mathbf{k} \leq 2^{-1}\epsilon^{-\chi}$, it suffices that, writing $\tau = 4D(\log \epsilon^{-1})^{4/3} + 2\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil$, we have that

$$4^{-1}\epsilon^{-\chi} \geq k_0 \vee (4D(\log \epsilon^{-1})^{4/3} + 2\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil)^3,$$

and

$$\begin{aligned} n\epsilon^{3/2} & \geq \max \left\{ 2(K_0)^{(12)^{-2}(\log \log(2^{-1}\epsilon^{-\chi}))^2} (\log(2^{-1}\epsilon^{-\chi}))^{K_0}, a_0^{-9}\tau^9, \right. \\ & \left. 10^{325}c^{-36}(2^{-1}\epsilon^{-\chi})^{465} \max \{1, \tau^{36}\} \right\}. \end{aligned}$$

Taking a union bound over the $\lfloor \epsilon^{-1} \rfloor + 1 \leq 2\epsilon^{-1}$ choices of j , we find that $\mathbb{P}(\text{ManyCanopy})$ is at most

$$\epsilon^{-1+\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \cdot 2\hat{H}(2h)^{(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288+3/2} \exp \{2(\log \epsilon^{-1})^{11/12}\}. \quad (18)$$

Applying Proposition 6.1 with $\mathbf{K} = D(\log \epsilon^{-1})^{1/3}$ and $\epsilon = \epsilon$, we find that

$$\begin{aligned} & \mathbb{P}\left(\text{LateCoal}_{n;([-D(\log \epsilon^{-1})^{1/3}, D(\log \epsilon^{-1})^{1/3}], 0)}^{([-1, 1], 1)}(\epsilon)\right) \\ & \leq \epsilon^2 \cdot 10^{420} c_3^{-33} C_3 D(\log \epsilon^{-1})^{127/3} \exp\{\beta_3(\log \epsilon^{-1})^{5/6}\}; \end{aligned} \quad (19)$$

since $\mathbf{K} \geq 2$ if and only if $\epsilon \leq \exp\{-8D^{-3}\}$, it is sufficient for the application to be made that $n\epsilon^{3/2} \in \mathbb{N}$,

$$\epsilon \leq \min\left\{\exp\{-8D^{-3}\}, 4^{-2}(D(\log \epsilon^{-1})^{1/3} + 2)^{-4}, (\eta_0)^{72}, 10^{-1342} c_3^{44}, \exp\{-2C^{3/8}\}\right\}$$

and the even integer n verifies

$$n \geq 2 \max\left\{2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606} c_3^{-84} \epsilon^{-222} (D(\log \epsilon^{-1})^{1/3} + 2)^{36}, a_0^{-9} (D(\log \epsilon^{-1})^{1/3} + 2)^9 2^6\right\}.$$

Applying Lemma 4.10 with $\mathbf{R} = D(\log \epsilon^{-1})^{1/3} - 1$, and using $(2^{-1/2} - 1/2)^{3/2} \geq 2^{-4}$, we learn that

$$\mathbb{P}\left(\neg \text{RegFluc}_{n;(*:f, 0)}^{\{-1, 1\}, 1}\left(D(\log \epsilon^{-1})^{1/3} - 1\right)\right) \leq 38D(\log \epsilon^{-1})^{1/3} C \epsilon^{2^{-13} c D^3} \quad (20)$$

since $\mathbf{R} \geq 2^{-1}D(\log \epsilon^{-1})^{1/3}$ due to $\epsilon \leq \exp\{-8D^{-3}\}$. This application is valid when

$$n \geq c^{-18} \max\left\{(\Psi_2 + 1)^9, 10^{23} \Psi_1^9, 3^9\right\}$$

and

$$D(\log \epsilon^{-1})^{1/3} - 1 \in \left[39\Psi_1 \vee 5 \vee 3c^{-3} \vee 2((\Psi_2 + 1)^2 + \Psi_3)^{1/2}, 6^{-1} c n^{1/9}\right].$$

Let $y \in [-1, 1]$ be given. Applying Proposition 4.9 with $\mathbf{t}_1 = 0$, $\mathbf{t}_2 = 1$, $\mathbf{x} = -D(\log \epsilon^{-1})^{1/3}$, $\mathbf{y} = y$, $\mathbf{a} = 1 - \epsilon^{3/2}$ and $\mathbf{r} = D(\log \epsilon^{-1})^{4/3}$, we find that

$$\mathbb{P}\left(\neg \text{PolyDevReg}_{n;(-D(\log \epsilon^{-1})^{1/3}, 0)}^{(y, 1)}\left(1 - \epsilon^{3/2}, D(\log \epsilon^{-1})^{4/3}\right)\right) \leq 22CD(\log \epsilon^{-1})^{4/3} \epsilon^{10^{-11} c_1 D^{3/4}}.$$

Since $y \in [-1, 1]$, the application may be made provided that $\epsilon^{3/2} \leq 10^{-11} c_1^2$; the even integer n satisfies

$$n \geq \max\left\{10^{29} \epsilon^{-75/2} c^{-18}, 10^{24} c^{-18} \epsilon^{-75/2} (D(\log \epsilon^{-1})^{1/3} + 1)^{36}\right\};$$

and

$$D(\log \epsilon^{-1})^{4/3} \in \left[10^9 c_1^{-4/5} \vee 15C^{1/2} \vee 11\epsilon^{1/2} (D(\log \epsilon^{-1})^{1/3} + 1), 3\epsilon^{25/6} n^{1/36}\right].$$

The same upper bound may be found on the quantity

$$\mathbb{P}\left(\neg \text{PolyDevReg}_{n;(D(\log \epsilon^{-1})^{1/3}, 0)}^{(y, 1)}\left(1 - \epsilon^{3/2}, D(\log \epsilon^{-1})^{4/3}\right)\right)$$

by this application, with \mathbf{x} instead set equal to $D(\log \epsilon^{-1})^{1/3}$; the conditions on parameters that permit the application are unchanged. Taking a union bound over the $[2\epsilon^{-1}] + 1 \leq 4\epsilon^{-1}$ values of $y \in \epsilon\mathbb{Z} \cap [-1, 1]$, we find that the failure probability for the intersection of PolyDevReg events in the definition of the NormalCoal event is at most

$$8 \cdot 22CD(\log \epsilon^{-1})^{4/3} \epsilon^{10^{-11} c_1 D^{3/4} - 1}. \quad (21)$$

We now assemble the estimates. The four terms on the right-hand side of (16) are bounded above by (19), (18), (20) and (21). That is,

$$\begin{aligned} & \mathbb{P}\left(\neg \text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, \epsilon, \chi)\right) \\ & \leq \epsilon^2 \cdot 10^{420} c_3^{-33} C_3 D (\log \epsilon^{-1})^{127/3} \exp\{\beta_3 (\log \epsilon^{-1})^{5/6}\} \\ & \quad + \epsilon^{-1+\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \\ & \quad \cdot 2\hat{H}(2h)^{(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288+3/2} \exp\{2(\log \epsilon^{-1})^{11/12}\} \\ & \quad + 38D(\log \epsilon^{-1})^{1/3} C \epsilon^{2-13cD^3} + 8 \cdot 22 CD (\log \epsilon^{-1})^{4/3} \epsilon^{10^{-11}c_1 D^{3/4-1}}. \end{aligned}$$

The second, third and fourth summands are all at most ϵ^2 provided that

$$\begin{aligned} & \epsilon^{-3+\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \\ & \leq 2^{-1} \hat{H}^{-1}(2\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil)^{-(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288-3/2} \exp\{-2(\log \epsilon^{-1})^{11/12}\}, \\ & \quad \epsilon^{2-13cD^3-2} \leq (38)^{-1} D^{-1} (\log \epsilon^{-1})^{-1/3} C^{-1} \end{aligned}$$

and

$$\epsilon^{10^{-11}c_1 D^{3/4-3}} \leq 8^{-1} 22^{-1} c_1 C^{-1} D^{-1} (\log \epsilon^{-1})^{-4/3}.$$

Confirming these three bounds is a calculational matter which is undertaken alongside the calculational derivation of Lemma 8.1 in Appendix B. Briefly, however, the first of these conditions is ensured by our insistence that $\epsilon > 0$ be small enough; and the second and third hold because we insist that D exceed a certain positive lower bound, and then that $\epsilon > 0$ be smaller than a positive constant determined by that lower bound.

Since $c_3 \leq 1$, $C_3 \wedge D \geq 1$, $\beta_3 \geq 0$ and $\epsilon \leq e^{-1}$, we conclude that

$$\mathbb{P}\left(\neg \text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, \epsilon, \chi)\right) \leq \epsilon^2 \cdot 10^{421} c_3^{-33} C_3 D (\log \epsilon^{-1})^{43} \exp\{\beta_3 (\log \epsilon^{-1})^{5/6}\}.$$

This completes the proof of Lemma 8.1. \square

9. DISCOVERING THE BROWNIAN PATCHWORK QUILT: THE DERIVATION OF THE MAIN RESULT

We have built the tools needed to derive Theorem 1.2, and this section is devoted to proving this result. Recall that it is our job to demonstrate that, for some large enough $n_0 \in 2\mathbb{N}$, the weight profiles $[-1, 1] \rightarrow \mathbb{R} : y \rightarrow \text{Wgt}_{n_0+n;(*:f,0)}^{(y,1)}$, indexed by $(n, f) \in 2\mathbb{N} \times \mathcal{I}_{\overline{\Psi}}$, are uniformly Brownian patchwork $(2, 3, 1/252)$ -quilttable. The definition of quilttable entails that a certain cast of characters must be introduced, and proved to enjoy certain properties and relationships. It may be helpful for the reader's bearings that we recall these things now for our particular setting.

Our index set \mathcal{A} is $\mathcal{I}_{\overline{\Psi}}$; its generic element will be called f . The random function $X_{n,f}$, indexed by $(n, f) \in 2\mathbb{N} \times \mathcal{I}_{\overline{\Psi}}$, which we have endeavoured to depict as a Brownian patchwork quilt, is, naturally, $[-1, 1] \rightarrow \mathbb{R} : y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$. The characters whose presence is called for are

- the two sequences p and q , which should verify $p_j \leq j^{-2+o(1)}$ and $q_j \leq j^{-1/252+o(1)}$;
- the error event $E_{n,f}$, whose \mathbb{P} -probability must be at most q_n ;
- the fabric sequence elements $F_{n,f,i}$, indexed by $(n, f, i) \in \mathbb{N} \times \mathcal{I}_{\overline{\Psi}} \times \mathbb{N}$, which will be expected to uniformly withstand L^3 -comparison to Brownian bridge;

- and the stitch points set $S_{n,f}$, whose cardinality will have tail bounded above by the p -sequence, and whose elements we will record in the form $s_{n,f;i}$, $i \in \llbracket 1, \#S_{n_0+n,f} \rrbracket$.

And, of course, we must have the fundamental relationship that ensures that our weight profiles are indeed being exhibited as patchwork quilts: when the error event $E_{n,f}$ fails to occur, $X_{n,f}$ must equal $\text{Quilt}[\bar{F}_{n,f}, S_{n,f}]$.

Moreover, for given $\bar{\Psi} \in (0, \infty)^3$, it suffices to find $n_0 \in 2\mathbb{N}$ such that the above conditions are verified merely if $n \in 2\mathbb{N}$ satisfies $n \geq n_0$. This weakening of the conditions is permitted because the q -sequence may be taken equal to one on an initial finite interval, so that $E_{n,f}$ may be chosen to be the entire sample space for $n < n_0$; thus, no demand is made for such n .

We structure our derivation of Theorem 1.2 by first setting up and discussing some structures that are entailed by the occurrence of the **NormalCoal** event (with suitable parameter settings) that was analysed in the preceding section. After we have been a little progress in this way, we will be able to specify the characters, and prove the properties, whose need has just been recalled, and thus reach the desired conclusion.

We begin this discussion by recalling the quantities D and χ that enter as parameters in the definition of the **NormalCoal** event. Henceforth, we set D equal to $10^{16}c_1^{-4/3}$, a choice that is made in order to permit the application of Lemma 8.1.

Let $\chi > 0$ given. We define an \mathbb{N} -valued random variable Γ_n . Its reciprocal will play the role of the parameter ϵ in the elaborated rough guide from Section 5. We have suggested that we will make a decreasing search for ϵ through dyadic scales. In light of this, it would be natural to define Γ_n to equal 2^j where $j \in \mathbb{N}$ is chosen to be minimal such that the event $\text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, 2^{-j}, \chi)$ occurs. This definition would seem to entail that the event $\text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, \Gamma_n^{-1}, \chi)$ always occurs. However, for this to make sense, it is necessary that the mesh condition $\Gamma_n^{-3/2} \in n^{-1}\mathbb{Z}$ be verified, and so we revisit the definition in order to respect this condition. For each $j \in \mathbb{N}$, we will write $\lceil 2^j \rceil$ for the smallest real value $u \geq 2^j$ such that $u^{-3/2}n \in \mathbb{Z}$. We actually define Γ_n to equal $\lceil 2^j \rceil$ where $j \in \mathbb{N}$ is chosen to be minimal such that the event $\text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, \lceil 2^j \rceil, \chi)$ occurs. This certainly entails that $\text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, \Gamma_n^{-1}, \chi)$ always occurs. It is due to this occurrence that

$$\text{Canopy}\#_{n;(*:f,0)}^{([-1,1],1)}(1 - \Gamma_n^{-3/2}) \leq \Gamma_n^{1+\chi}. \quad (22)$$

Before continuing, we mention that this variation in definition of Γ_n to accomodate the mesh condition is a fairly minor detail. For reference shortly, we note that, for given $j \in \mathbb{N}$, the quantity $\lceil 2^j \rceil$ is less than 2^{j+1} provided that $n \geq (1 - 2^{-3/2})^{-1}2^{3j/2}$. We will see that the j -value associated to Γ_n will very typically verify this condition, so that the concerned dyadic scale does not typically differ between the proposed and actual definitions of Γ_n .

Since $\text{NormalCoal}_{n;(*:f,0)}^{([-1,1],1)}(D, \Gamma_n^{-1}, \chi)$ occurs, we may find a bound on the tail of Γ_n by applying Lemma 8.1. We consider $i \in \mathbb{N}$ and apply this lemma with $\epsilon = \lceil 2^i \rceil^{-1}$, and with \mathbf{D} equal to D , namely $10^{16}c_1^{-4/3}$. Provided that $\chi > 0$ is small enough that $\chi \leq 2^{-1}(1 + 500(\log \beta)^2)^{-1}$, we conclude that, when $n \in 2\mathbb{N}$ and $i \in \mathbb{N}$ satisfy $i \geq i_0(\chi, \bar{\Psi})$ and $n \geq n_1(\bar{\Psi}) + \tilde{C}2^{504i}$,

$$\mathbb{P}(\Gamma_n \geq \lceil 2^i \rceil) \leq \lceil 2^i \rceil^{-2} \cdot \Omega(\lceil 2^i \rceil^{-1}). \quad (23)$$

Here, 2^{-i_0} is the upper bound on ϵ in Lemma 8.1. The quantity $n_1(\bar{\Psi})$ equals the first of the three terms of which the triple maximum lower bound on n is composed. The constant \tilde{C} is specified

by increasing the value $10^{618} c_3^{-84} a_0^{-9} (10^{16} c_1^{4/3})^{36}$ by a K_0 -determined constant so that the assumed lower bound $n \geq n_1(\chi, \bar{\Psi}) + \tilde{C} 2^{504i}$ is enough to ensure that n is at least the second, as well as the third, of the three quantities in the triple maximum lower bound in Lemma 8.1. Naturally, the function Ω is specified by the value in the lemma where \mathbf{D} equals the above value.

For the values of (n, i) that we are considering, it follows from a comment made a moment ago that $\lceil 2^{i-1} \rceil$ is less than 2^i . In view of this, and since $\epsilon \rightarrow \Omega(\epsilon)$ is decreasing, we see from (23) that

$$\mathbb{P}(\Gamma_n \geq 2^i) \leq 2^{-2i} \cdot \Omega(2^{-i-1}). \quad (24)$$

The form of this bound permits us to leave behind the $\lceil \cdot \rceil$ notation, because the discrepancy between the ‘dyadic scale’ and actual definitions of Γ_n has in practice been taken care of.

For $n \in 2\mathbb{N}$ that satisfies $n \geq n_1(\bar{\Psi})$, let $i_{\max}(n) \in \mathbb{N}$ be the maximal value of $i \in \mathbb{N}$ such that $n \geq n_1(\bar{\Psi}) + \tilde{C} 2^{504i}$. Define the *error event* $E_n = \{\Gamma_n \geq 2^{i_{\max}(n)}\}$. Since $2^{i_{\max}(n)}$ is up to a bounded factor equal to $n^{1/504}$ provided that n exceeds a certain value $n_2(\bar{\Psi})$, and $\epsilon \rightarrow \Omega(\epsilon)$ is decreasing, we find that (24) implies that

$$\mathbb{P}(E_n) \leq n^{-2/504} \cdot \Omega(n^{-1/503}). \quad (25)$$

Suppose now that $n \geq n_2(\bar{\Psi})$ and that E_n^c occurs. Consider any $(n; * : f, 1 - \Gamma_n^{-3/2})$ -canopy C that intersects $[-1, 1]$. We now discuss the behaviour of the f -rewarded line-to-point weight profile $y \rightarrow \text{Wgt}_{n;(*:f,0)}^{(y,1)}$ on the interval $y \in C \cap [-1, 1]$; note that there are one or two such $(n; * : f, 1 - \Gamma_n^{-3/2})$ -canopies C (those that contain -1 and 1) for which $C \cap [-1, 1]$ is a smaller interval than is C itself, but that each of the remaining choices of C is equal to $C \cap [-1, 1]$. For any such C (that intersects $[-1, 1]$), the root $\rho_{n;(*:f,0)}^{(y,1)}(0)$ is independent of $y \in C$, and we denote by $\kappa = \kappa(C) \in \mathbb{N}$ the quantity $\kappa = \lfloor \Gamma_n \rho_{n;(*:f,0)}^{(y,1)}(0) \rfloor$ for any $y \in C$. Moreover, recalling Definition 7.6, the closure of the canopy C contains a special point $\text{spec} = \text{spec}_C$. In the case that the special point lies in the interior of C , we consider the two random functions of $y \in (\inf C, \text{spec}) \cap [-1, 1]$ and $y \in (\text{spec}, \sup C) \cap [-1, 1]$ respectively given by

$$\text{Wgt}_{n;(*:f,0)}^{(y,1)} - \text{Wgt}_{n;((\kappa+1)\Gamma_n^{-1},0)}^{(y,1)} \quad \text{and} \quad \text{Wgt}_{n;(*:f,0)}^{(y,1)} - \text{Wgt}_{n;(\kappa\Gamma_n^{-1},0)}^{(y,1)}.$$

When $\text{spec} = \sup C$, we consider only the first of these functions; and when $\text{spec} = \inf C$, only the second. Each function under consideration is constant, as we learn by applying Lemma 7.7 with parameter settings $\epsilon = \Gamma_n^{-1}$ and $\mathbf{K} = D(\log \Gamma_n)^{1/3}$. The lemma’s conclusion is valid only when a **NoLateCoal** and a **RegFluc** event occur. It is the occurrence of **NormalCoal** $_{n;(*:f,0)}^{([-1,1],1)}(D, \Gamma_n^{-1}, \chi)$ which ensures that these events come to pass. Meanwhile, the lemma’s hypothesis that $n > 2^{-3/2} \Psi_1^3 \vee 8(\Psi_2 + 1)^3 \vee 8(\mathbf{K} + 2)^3$ is satisfied, after a possible increase in the value of $n_2 = n_2(\chi, \bar{\Psi})$, because the occurrence of E_n^c entails that $\log \Gamma_n \leq i_{\max}(n) \log 2$, an inequality whose right-hand side is up to a bounded factor at most $\log n$.

In this way, on the event E_n^c , we split each $(n; * : f, 1 - \Gamma_n^{-3/2})$ -canopy I into two pieces (when the special point lies in the interior of C), or leave I untouched (in the other case). We call the intervals so formed $(n; * : f, 1 - \Gamma_n^{-3/2})$ -*split canopies*. Each such split canopy C has the property that the function above that is defined on C is independent of the value of y in C ’s intersection with $[-1, 1]$. Indeed, we may associate to C the *root neighbour index* $\text{RNI} = \text{RNI}(1 - \Gamma_n^{-3/2}, C) \in \{\kappa, \kappa + 1\}$ to be the index of this function which is constant for $y \in C \cap [-1, 1]$.

We write $\#\text{SC}$, the split canopy cardinality, for the number of $(n; * : f, 1 - \Gamma_n^{-3/2})$ -split canopies that are subsets of $(n; * : f, 1 - \Gamma_n^{-3/2})$ -canopies that intersect $[-1, 1]$. We may record the collection of such split canopies that intersect $[-1, 1]$ in the form (ζ_i, ζ_{i+1}) , $1 \leq i \leq \#\text{SC}$, where the increasing real-valued sequence $\{\zeta_i : i \in \llbracket 1, \#\text{SC} + 1 \rrbracket\}$ satisfies $-1 \in (\zeta_1, \zeta_2)$ and $1 \in (\zeta_{\#\text{SC}}, \zeta_{\#\text{SC}+1})$. This sequence's terms are of two types, namely boundary points of $(n; * : f, 1 - \Gamma_n^{-3/2})$ -canopies, and special points interior to such canopies. No two consecutive terms of the latter type are possible, so that $\#\text{SC} \leq 2 \text{Canopy} \#_{n, (*:f,0)}^{([-1,1],1)} (1 - \Gamma_n^{-3/2})$.

We use a shorthand under which, for each $i \in \llbracket 1, \#\text{SC} \rrbracket$, we write RNI_i for the value of the root neighbour index $\text{RNI}(1 - \Gamma_n^{-3/2}, (\zeta_i, \zeta_{i+1}))$ associated to i^{th} of the above $(n; * : f, 1 - \Gamma_n^{-3/2})$ -split canopies. On the event E_n^c , and for each $1 \leq i \leq \#\text{SC}$, we define $Y_{n,i} : [-1, 1] \rightarrow \mathbb{R}$ by setting $Y_{n,i}(y) = \text{Wgt}_{n; (\text{RNI}_i, \Gamma_n^{-1}, 0)}^{(y,1)}$ for $y \in [-1, 1]$. We also use a formal device that permits $Y_{n,i}$ to be defined on the whole probability space. For given (n, i) , it remains to specify $Y_{n,i}$ on the event $E_n \cup \{i \geq \#\text{SC} + 1\}$. We specify that the conditional distribution of $Y_{n,i}$, given this event, equals the standard Brownian bridge law $\mathcal{B}_{1,0,0}^{[-1,1]}$ (which we will denote by μ later in the proof).

Proof of Theorem 1.2. We may now begin the formal derivation of this result, because we are ready to specify the sequences and events (whose need has been recalled at the beginning of this section) that will show that our weight profiles are suitably quiltable. Recall also that it is enough to verify the concerned conditions only when $n \in 2\mathbb{N}$ satisfies $n \geq n_0$, for n_0 determined by $\bar{\Psi}$.

We will next specify the definitions and then explain why they enjoy the necessary properties and relationships. It is the verification that fabric sequence elements uniformly withstand L^3 -comparison to Brownian bridge that still requires a little effort. We present this verification after the others.

First the definitions.

The sequences p and q . We may set $\chi > 0$ so that $2(1 + \chi)^{-1} = 2 - \epsilon/2$ in order that we may take $p_j = j^{-2+\epsilon/2} \cdot 16\Omega((j/4)^{-(1+\chi)^{-1}})$. We set $q_n = n^{-1/252}\Omega(n^{-1/503})$.

The error event $E_{n,f}$. We set $E_{n,f} = E_n$.

The fabric sequence elements. We set $F_{(n,f),i} = Y_{n,i}$ for each $(n, f, i) \in \mathbb{N} \times \mathcal{I}_{\bar{\Psi}} \times \mathbb{N}$.

The stitch points set $S_{n,f}$. This set's cardinality will be chosen to be $\#\text{SC} - 1$. Its elements will be $s_{(n,f),i} = \zeta_i$ for $2 \leq i \leq \#\text{SC}$. We also adopt the convention that $s_{(n,f),0} = -1$ and $s_{(n,f),\#S_{n,f}+1} = 1$.

And now the verification of the desired properties.

The tail of the two sequences. Since $\Omega(\epsilon) = \epsilon^{o(1)}$ as $\epsilon \searrow 0$, there exists $j_0 \in \mathbb{N}$ such that $16\Omega(2^{-1}(j/4)^{-(1+\chi)^{-1}}) \leq j^{\epsilon/2}$ when $j \geq j_0$. Thus, $p_j \leq j^{-2+\epsilon}$ whenever $j \geq j_0$.

Since $\Omega(\epsilon) = \epsilon^{o(1)}$ as $\epsilon \searrow 0$, the sequence q verifies $q_n \leq n^{-1/252+\epsilon}$ for n sufficiently high.

The error event's tail. In view of the definition of q_n and (25), we have that $\mathbb{P}(E_{n,f}) \leq q_n$.

The p -sequence dominates the tail of the stitch points' cardinality. Recall that $\#S_{n,f} \leq \#\text{SC} \leq 2 \text{Canopy} \#_{n, (*:f,0)}^{([-1,1],1)} (1 - \Gamma_n^{-3/2})$. By (22) and (24),

$$\mathbb{P}(\#S_{n,f} \geq 2^{i(1+\chi)+1}) \leq \mathbb{P}(\Gamma_n \geq 2^i) \leq 2^{-2i} \cdot \Omega(2^{-i-1})$$

for $i \in \mathbb{N}$ satisfying $i \geq i_0(\chi, \bar{\Psi})$. Since $\epsilon \rightarrow \Omega(\epsilon)$ is decreasing,

$$\mathbb{P}(\#S_{n,f} \geq j) \leq j^{-2(1+\chi)^{-1}} \cdot 16 \Omega(2^{-1}(j/4)^{-(1+\chi)^{-1}}) = p_j$$

for j sufficiently large.

The weight profiles are indeed patchwork quilts. We must check that, when the error event $E_{n,f}$ fails to occur, $X_{n,f}$ equals $\text{Quilt}[\bar{F}_{n,f}, S_{n,f}]$. This amounts to confirming that, on $E_{n,f}^c$, in each patch $[s_{(n,f),i-1}, s_{(n,f),i}]$, $i \in \llbracket 1, \#S_{n,f} + 1 \rrbracket$, the weight profile $y \rightarrow \text{Wgt}_{n;(*:f),0}^{(y,1)}$ differs from the fabric sequence element $F_{n,f;i} = Y_{n,i}$ by a constant (albeit a random one). Each patch is a split canopy, except the extreme ones, and these are subintervals of split canopies. Thus, this property follows by our construction.

Fabric sequence elements uniformly withstand L^{3-} -comparison to Brownian bridge. All that remains to complete the proof of our main theorem is to confirm this assertion. To wit, we have to show that the Radon-Nikodym derivative of the law of $F_{(n,f),j}^{[-1,1]}$ with respect to the law $\mu = \mathcal{B}_{1;0,0}^{[-1,1]}$ of standard Brownian bridge on $[-1, 1]$ lies in $L^{3-\delta}(d\mu)$ for any $\delta > 0$, with an $L^{3-\delta}$ -norm that, while possibly dependent on $\delta > 0$, is independent of $(n, f, j) \in \mathbb{N} \times \mathcal{I}_{\bar{\Psi}} \times \mathbb{N}$ for $n \geq n_0$.

We now present over several paragraphs an argument leading to this conclusion. Let $\{a_j : j \in \mathbb{N}\}$ denote the enumeration of rationals in $[-1, 1]$ of the form $p2^{-i}$ with $p, i \in \mathbb{N}$ coprime in which these values are recorded in increasing order of the dyadic scale i , and increasingly for values of given scale: the sequence begins $a_1 = -1$, $a_2 = 0$, $a_3 = 1$, and continues by enumerating $-1/2$ and $1/2$, and then $-3/4$, $-1/4$, $1/4$ and $3/4$.

For all $n, m \in \mathbb{N}$, we define the random function $Z_{n,m} : [-1, 1] \rightarrow \mathbb{R}$, setting

$$Z_{n,m}(y) = \text{Wgt}_{n;(a_m,0)}^{(y,1)} \quad \text{for } y \in [-1, 1].$$

For $n \in 2\mathbb{N}$ and $i \in \llbracket 1, \#\text{SC} \rrbracket$, we introduce the \mathbb{N} -valued random variable $M_{n,i}$ by declaring that $a_{M_{n,i}} = \text{RNI}_i \cdot \Gamma_n^{-1}$. Thus, the random function $Y_{n,i} : [-1, 1] \rightarrow \mathbb{R}$ may be written $Y_{n,i}(y) = Z_{n,M_{n,i}}(y)$ for $y \in [-1, 1]$. The root neighbour index RNI_i is defined only when $E_n^c \cap \{\#\text{SC} \leq i\}$ occurs. To specify $M_{n,i}$ on the whole probability space, we further set $M_{n,i} = 0$ on $E_n \cup \{i \geq \#\text{SC} + 1\}$.

For any $n \in 2\mathbb{N}$ and $i \in \mathbb{N}$, $a_{M_{n,i}}$ is a dyadic rational in $[-1, 1]$ whose dyadic scale is at most the base-two logarithm of Γ_n . The number of such rationals in $[-1, 1]$ is $2\Gamma_n + 1$. Thus, for each $k \in \mathbb{N}$, $k \geq 1$,

$$\mathbb{P}\left(\sup_{i \in \mathbb{N}} M_{n,i} \geq 2^k + 2, E_n^c\right) \leq \mathbb{P}(\Gamma_n > 2^{k-1}, E_n^c) = \mathbb{P}(2^k \leq \Gamma_n \leq 2^{i_{\max}(n)-1}),$$

where, in verifying the equality, it may be useful to recall that E_n and $i_{\max}(n)$ have been defined in the paragraph ending at (25). When $k \leq i_{\max}(n) - 1$, we may apply (24) with $\mathbf{i} = k$. The hypothesis that $n \geq n_1(\bar{\Psi}) + \tilde{C} 2^{504k}$ is validated by the definition of $i_{\max}(n)$. We thus find that

$$\mathbb{P}\left(\sup_{i \in \mathbb{N}} M_{n,i} \geq 2^k + 2, E_n^c\right) \leq 2^{-2k} \cdot \Omega(2^{-k-1}) \tag{26}$$

for $i_0 \leq k \leq i_{\max}(n) - 1$. Since the event $2^k \leq \Gamma_n \leq 2^{i_{\max}(n)-1}$ cannot occur if $k \geq i_{\max}(n)$, the bound (26) is seen to be valid provided merely that $k \geq i_0$.

Since $\epsilon \rightarrow \Omega(\epsilon)$ is decreasing, we find that

$$\mathbb{P}\left(\sup_{i \in \mathbb{N}} M_{n,i} \geq j, E_n^c\right) \leq 2^2 j^{-2} \cdot \Omega(j^{-1}/4)$$

whenever $n \in 2\mathbb{N}$ and $j \geq j_0$, where $j_0 = 2^{i_0+1}$.

Let $p_{n,i,j} = \mathbb{P}(M_{n,i} = j, E_n^c)$. Let $A \subseteq \mathcal{C}_{0,0}([-1, 1], \mathbb{R})$ be measurable. For $n \in 2\mathbb{N}$ and $i \in \mathbb{N}$, note that

$$\mathbb{P}(Y_{n,i}^{[-1,1]} \in A, E_n^c) = \sum_{j \in \mathbb{N}} \mathbb{P}(Y_{n,i}^{[-1,1]} \in A, M_{n,i} = j, E_n^c).$$

Let $\ell \in \mathbb{N}$. Note that, when $\ell \geq j_0$,

$$\sum_{j=\ell}^{\infty} \mathbb{P}(Y_{n,i}^{[-1,1]} \in A, M_{n,i} = j, E_n^c) \leq \mathbb{P}(M_{n,i} \geq \ell, E_n^c) \leq \ell^{-2} \cdot 4\Omega(\ell^{-1}/4),$$

and that

$$\sum_{j=1}^{\ell-1} \mathbb{P}(Y_{n,i}^{[-1,1]} \in A, M_{n,i} = j, E_n^c) \leq \sum_{j=1}^{\ell-1} \mathbb{P}(Z_{n,j}^{[-1,1]} \in A, E_n^c).$$

Recall that $Z_{n,j}(y) = \text{Wgt}_{n;(a_j,0)}^{(y,1)}$. The process $Z_{n,j} : [-1, 1] \rightarrow \mathbb{R}$ is an instance of the random function \mathcal{L} seen in Theorem 4.3. Indeed, the process \mathcal{L} in this theorem is specified by three parameters, and with the settings $\mathbf{K} = -1$, $\mathbf{d} = 2$ and $\mathbf{x} = a_j$, it coincides with $Z_{n,j}$ on $[-1, 1]$. We find that, provided that n , even, is at least a certain constant n_2 ,

$$\mathbb{P}(Z_{n,j}^{[-1,1]} \in A) \leq \hat{C}\mu(A) \exp \left\{ (\log(\mu(A)^{-1}\hat{C}))^{5/6} \right\},$$

where the constant \hat{C} is finite; note that the final sentence of Theorem 4.3 implies that \hat{C} may be chosen independently of $j \in \mathbb{N}$ provided that $4 \leq 2^{-1}c(n+1)^{1/9}$ (and this bound is assured by $n \geq n_2$ in view of the value of n_2 that we are about to state). At this moment, the reader may be forgiven for a lack of curiosity concerning an explicit formula for n_2 . Nonetheless, as an aside, we provide one: the result that underlies Theorem 4.3, namely [Ham17a, Theorem 2.11], supplies it. We may take n_2 equal to the maximum of $8^9 c^{-18}$ and

$$(c/2 \wedge 2^{1/2})^{-9} D_1^9 \left(\log \left(2(18)^{3/2} C_1^{3/2} D_1^{3/2} \vee 2 \exp \{ 2 \cdot 10^7 \cdot 2^6 \} \right) \right)^3,$$

where $D_1 = c_1^{-1/3} (2^{-9/2} - 2^{-5})^{-1/3}$.

We find that, when n even is at least n_2 ,

$$\sum_{j=1}^{\ell-1} \mathbb{P}(Y_{n,i}^{[-1,1]} \in A, M_{n,i} = j, E_n^c) \leq (\ell-1) \hat{C}\mu(A) \exp \left\{ (\log(\mu(A)^{-1}\hat{C}))^{5/6} \right\}.$$

When $M_{n,i} = 0$, $Y_{n,i} = Y_{n,i}^{[-1,1]}$ was artificially declared to have the law μ . Thus,

$$\mathbb{P}(Y_{n,i}^{[-1,1]} \in A, M_{n,i} = 0) \leq \mu(A).$$

Set $\ell = \lceil \mu(A)^{-1/3} \rceil$. We find that, when n even is at least n_2 ,

$$\begin{aligned} \mathbb{P}(Y_{n,i}^{[-1,1]} \in A) &\leq \mu(A)^{2/3} \Omega[A] + \mu(A)^{2/3} \hat{C} \exp \left\{ (\log \mu(A)^{-1} \hat{C})^{5/6} \right\} + \mu(A) \\ &\leq \mu(A)^{2/3} (\Omega[A] + \hat{C} + 1) \exp \left\{ (\log \mu(A)^{-1} \hat{C})^{5/6} \right\} \end{aligned}$$

provided that $\mu(A) \leq e^{-1}$ (since $\hat{C} \geq 1$). Here, $\Omega[A]$ denotes $4\Omega(4^{-1} \lceil \mu(A)^{-1/3} \rceil^{-1})$.

Writing $\mathcal{C} = \mathcal{C}_{0,0}([-1, 1], \mathbb{R})$, let $f_{n,i} : \mathcal{C} \rightarrow [0, \infty)$ denote the Radon-Nikodym derivative of the law of $Y_{n,i}^{[-1,1]}$ under \mathbb{P} with respect to μ . For $\ell \in \mathbb{N}$, set $A_{n,i,\ell} = \{g \in \mathcal{C} : 2^\ell \leq f_{n,i}(g) \leq 2^{\ell+1}\}$.

When $n \geq n_2$, we know that

$$2^\ell \mu(A_{n,i,\ell}) \leq \mu(A_{n,i,\ell})^{2/3} (\Omega[A_{n,i,\ell}] + \hat{C} + 1) \exp \left\{ (\log \mu(A_{n,i,\ell})^{-1} \hat{C})^{5/6} \right\},$$

and this implies that

$$\mu(A_{n,i,\ell}) \leq 2^{-3\ell} (\Omega[A_{n,i,\ell}] + \hat{C} + 1)^3 \exp \left\{ 3 (\log \mu(A_{n,i,\ell})^{-1} \hat{C})^{5/6} \right\}.$$

If $\mu(A_{n,i,\ell}) \geq 2^{-3\ell}$, then we find that

$$\mu(A_{n,i,\ell}) \leq 2^{-3\ell} (\Omega(2^{-\ell-2}) + \hat{C} + 1)^3 \exp \left\{ 3(3\hat{C}\ell \log 2)^{5/6} \right\};$$

recall from Lemma 8.1 that $\Omega(2^{-\ell-2}) = O(1) \exp \{O(1)\ell^{5/6}\}$. Thus, we find that, even without imposing the condition $\mu(A_{n,i,\ell}) \geq 2^{-3\ell}$, we may derive the bound

$$\mu(A_{n,i,\ell}) \leq 2^{-3\ell} O(1) \exp \{O(1)\ell^{5/6}\}$$

whenever $n \in 2\mathbb{N}$, $n \geq n_2$ and $\ell, i \in \mathbb{N}$.

Let $\delta > 0$. Note that

$$\int f_{n,i}^{3-\delta} d\mu \leq \sum_{\ell=1}^{\infty} 2^{(3-\delta)(\ell+1)} \mu(A_{n,i,\ell}) \leq O(1) \sum_{\ell=1}^{\infty} 2^{-\delta\ell} \exp \{O(1)\ell^{5/6}\}$$

is bounded above independently of (n, i) satisfying $n \in 2\mathbb{N}$, $n \geq n_2$ and $i \in \mathbb{N}$. This confirms that fabric sequence elements $F_{(n,f);j}^{[-1,1]}$ uniformly withstand L^{3-} -comparison to Brownian bridge. The proof of Theorem 1.2 is complete. \square

APPENDIX A. POLYMER UNIQUENESS

Here we prove the polymer uniqueness Lemma 4.6. We first retreat to unscaled coordinates, and prove the counterpart result in a more general form, addressing uniqueness of the energy maximizer among systems of pairwise disjoint staircases with given endpoints.

To state this general result, Lemma A.1, we introduce some notation. The concerned staircases are not in fact entirely disjoint, not least because their endpoints are shared. Adopting a weaker notion than disjointness, we say that two staircases are horizontally separate if there is no planar horizontal interval of positive length that is a subset of a horizontal interval in both staircases.

Now our notation for collections of pairwise horizontally separate staircases. For $\ell \in \mathbb{N}$, let (x_i, s_i) and (y_i, f_i) , $i \in \llbracket 1, \ell \rrbracket$, be a collection of pairs of elements of $\mathbb{R} \times \mathbb{N}$. (The symbols s and f are used in reference to the staircases' heights at the *start* and *finish*.)

Let $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ denote the set of ℓ -tuples $(\phi_1, \dots, \phi_\ell)$, where ϕ_i is a staircase from (x_i, s_i) to (y_i, f_i) and each pair (ϕ_i, ϕ_j) , $i \neq j$, is horizontally separate. (This set may be empty; in order that it be non-empty, it is necessary that $y_i \geq x_i$ and $f_i \geq s_i$ for $i \in \llbracket 1, \ell \rrbracket$.) Note also that $SC_{(x_1, s_1) \rightarrow (y_1, f_1)}^1$ equals $SC_{(x_1, s_1) \rightarrow (y_1, f_1)}$.

We also associate an energy to each member of $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$. Each of the ℓ elements of any ℓ -tuple in $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ has an energy, as we described in Subsection 2.0.1. Define the energy $E(\phi)$ of any $\phi = (\phi_1, \dots, \phi_\ell) \in SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ to be $\sum_{j=1}^\ell E(\phi_j)$.

When $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell \neq \emptyset$, we further define the maximum ℓ -tuple energy

$$M_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell = \sup \left\{ E(\phi) : \phi \in SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell \right\}.$$

Lemma A.1. *Let $\ell \in \mathbb{N}$ and let (x_i, s_i) and (y_i, f_i) , $i \in \llbracket 1, \ell \rrbracket$, be a collection of pairs of points in $\mathbb{R} \times \mathbb{N}$ such that $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ is non-empty. Then, except on a \mathbb{P} -null set, there is a unique element of $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ whose energy attains $M_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$.*

Proof of Lemma A.1. We model the proof on a resampling argument that shows a simple

Claim. Standard Brownian motion $B : [0, 1] \rightarrow \mathbb{R}$, $B(0) = 0$, has a unique maximizer.

We begin by deriving the claim. In this toy case, we may note that the event that the maximizer is non-unique is a subset of the union over pairs of disjoint intervals $I, J \subset [0, 1]$ with rational endpoints such that $\sup I < \inf J$ of the event $A_{I, J; B}$ that the maximum value $B[\max, I] := \sup_{x \in I} B(x)$ coincides with its counterpart $B[\max, J]$. Writing \mathbb{P} for the probability measure that carries the process B , we may augment the probability space associated to this measure by equipping it with a further *resampled* process B_J^{re} that will also be distributed as standard Brownian motion on $[0, 1]$ under \mathbb{P} . There are some variations on the rules for specifying this new process that prove the desired result in this toy case; we select one that is directly adaptable to the Brownian LPP polymer uniqueness problem that we have in mind. Let $B' : [0, \infty) \rightarrow \mathbb{R}$ denote a standard Brownian motion that is independent of B . Label $J = [j_1, j_2]$ for $j_1, j_2 \in [0, 1]$. For $x \in [0, 1]$, we define

$$B_J^{\text{re}}(x) = \begin{cases} B(x) & x \in [0, j_1], \\ B(j_1) + B'(x - j_1) & x \in J, \\ B(j_1) + B'(j_2 - j_1) + B(x) - B(j_2) & x \in [j_2, 1]. \end{cases}$$

The process B_J^{re} is readily seen to have independent Gaussian increments of the necessary variance and thus to have the law of standard Brownian motion.

As such, we may note that $\mathbb{P}(A_{I, J; B}) = \mathbb{P}(A_{I, J; B_J^{\text{re}}})$, where naturally the latter notation refers to the event of coincidence of interval maxima for the resampled process. Let $\sigma[B]$ denote the σ -algebra generated by $B : [0, 1] \rightarrow \mathbb{R}$. Write $\mathbb{P}_{\sigma[B]}(\cdot) = \mathbb{E}[\mathbf{1}_{\cdot} | \sigma[B]]$ for conditional probability given knowledge of B . Note then that

$$\mathbb{P}(A_{I, J; B_J^{\text{re}}}) = \mathbb{E} \left[\mathbb{P}_{\sigma[B]}(A_{I, J; B_J^{\text{re}}}) \right].$$

Note that $A_{I, J; B_J^{\text{re}}}$ occurs precisely when $B(j_1) + \sup_{x \in [0, j_2 - j_1]} B'(x)$ equals the resampled process maximum on I , which we may denote by $B_J^{\text{re}}[\max, I]$. Because I is assumed to lie to the left of J , $B_J^{\text{re}}[\max, I]$ equals $B[\max, I]$. Conditionally on B , this equality characterizing the occurrence of $A_{I, J; B_J^{\text{re}}}$ may be expressed as asserting that the conditionally random quantity $\sup_{x \in [0, j_2 - j_1]} B'(x)$ equals the known value $B[\max, I] - B(j_1)$. The conditionally random quantity has the conditional law of the maximum of a standard Brownian motion on the given length interval $[0, j_2 - j_1]$; by the reflection principle, this law has a density and thus is non-atomic. Hence, $\mathbb{P}_{\sigma[B]}(A_{I, J; B_J^{\text{re}}})$ is seen to be zero, \mathbb{P} -almost surely.

In this way, we prove the claim: we find that $\mathbb{P}(A_{I,J;B_J^{\text{re}}})$, and thus also $\mathbb{P}(A_{I,J;B})$, equals zero, as we sought to show.

We now adapt this argument to prove Lemma A.1. A horizontal planar line segment with integer height is called a *horizontal rational interval* if its two endpoints have rational x -coordinate. Recall that our data $\ell \in \mathbb{N}$ and $(x_i, s_i), (y_i, f_i), i \in \llbracket 1, \ell \rrbracket$, is supposed such that $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ is non-empty.

Say that an element ϕ of the set $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ *inhabits* a given horizontal rational interval I if the interval I is contained in one of the horizontal segments of one of the ℓ staircase components of ϕ . Say that such an element *avoids* I if each of these components is disjoint from I . Let A_I denote the event that there exist an element of $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ of maximal weight that inhabits I and another such element that avoids I .

Note then that the event that there are two elements of $SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ each of whose energies attains $M_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ is, up to a \mathbb{P} -null set error, contained in $\bigcup A_I$, where the union is taken over all horizontal rational intervals I .

We now prove by resampling that $\mathbb{P}(A_I) = 0$ for any given horizontal rational interval I .

We augment the probability space carrying the law \mathbb{P} with an auxiliary ensemble $B_I^{\text{re}} : \mathbb{Z} \times \mathbb{R} \rightarrow \mathbb{R}$ whose law will coincide with that of B . To do so, let $B' : \mathbb{R} \rightarrow \mathbb{R}$ denote standard two-sided Brownian motion, independent of the ensemble B under \mathbb{P} . Write $I = [z_1, z_2] \times \{k\}$ for $z_1, z_2 \in \mathbb{R}$, $z_1 \leq z_2$, and $k \in \mathbb{Z}$. For $x \in \mathbb{R}$, we define

$$B_I^{\text{re}}(k, x) = \begin{cases} B(k, x) & x \leq z_1, \\ B(k, z_1) + B'(x - z_1) & x \in [z_1, z_2], \\ B(k, z_1) + B'(z_2 - z_1) + B(k, x) - B(k, z_2) & x \geq z_2. \end{cases}$$

Note that the increments of $B_I^{\text{re}}(k, \cdot)$ coincide with those of $B(k, \cdot)$ away from I , while these increments are determined by independent Brownian randomness on this interval. The process $B_I^{\text{re}}(k, \cdot)$ thus shares the law of $B(k, \cdot)$.

We also set $B_I^{\text{re}}(j, x) = B(j, x)$ for all $(j, x) \in \mathbb{Z} \times \mathbb{R}$, $j \neq k$, so that the new ensemble shares the law of B under \mathbb{P} .

Any ℓ -tuple ϕ of staircases has an energy $E(\phi)$ specified by increments of the ensemble B ; it also has a counterpart energy specified in terms of the new ensemble B_I^{re} . We will write $E_I^{\text{re}}(\phi)$ for this new, resampled, energy. Similarly, the event A_I is specified by the randomness of B , and has a counterpart, which we denote by A_I^{re} , when the role of this randomness is played by B_I^{re} .

The equality in law between B and B_I^{re} implies that $\mathbb{P}(A_I) = \mathbb{P}(A_I^{\text{re}})$. To prove Lemma A.1, we have seen that it is enough to show that $\mathbb{P}(A_I) = 0$ for a given horizontal rational interval I . We now complete the proof by fixing such an interval $I = [z_1, z_2] \times \{k\}$ and arguing that $\mathbb{P}(A_I^{\text{re}}) = 0$.

Any element $\phi \in SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ that avoids I satisfies $E_I^{\text{re}}(\phi) = E(\phi)$. Any element $\phi \in SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ that inhabits I undergoes a random but ϕ -independent change of energy under the resampling experiment, satisfying $E_I^{\text{re}}(\phi) = E(\phi) + \Theta$, where $\Theta = B'(z_2 - z_1) - (B(k, z_2) - B(k, z_1))$.

Let the avoidance sum energy ASE denote the supremum of $E(\phi)$ over those $\phi \in SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ that avoid I . Let the inhabitation sum energy ISE denote the supremum of $E(\phi)$ over those $\phi \in$

$SC_{(\bar{x}, \bar{s}) \rightarrow (\bar{y}, \bar{f})}^\ell$ that inhabit I . When these quantities are considered after resampling, with $E_I^{\text{re}}(\phi)$ in place of $E(\phi)$, we prefix the term *resampled* to their names.

Since A_I^{re} is the event that the resampled avoidance and inhabitation sum energies are equal, this event occurs precisely when Θ equals the difference between the avoidance and inhabitation sum energies. Let $\sigma[B]$ denote the σ -algebra generated by the ensemble $B : \mathbb{Z} \times \mathbb{R} \rightarrow \mathbb{R}$ and write $\mathbb{P}_{\sigma[B]}$ for the associated conditional probability. We have then that

$$\mathbb{P}(A_I^{\text{re}}) = \mathbb{E} \left[\mathbb{P}_{\sigma[B]}(\Theta = \text{ASE} - \text{ISE}) \right].$$

To the observer of the ensemble B , $\text{ASE} - \text{ISE}$ is a known quantity, while the conditional distribution of Θ is a normal random variable of mean $B(k, z_1) - B(k, z_2)$ and variance $z_2 - z_1$. Thus, $\mathbb{P}_{\sigma[B]}(\Theta = \text{ASE} - \text{ISE})$ equals zero, \mathbb{P} -almost surely, and so $\mathbb{P}(A_I^{\text{re}})$ equals zero, as we sought to show. This completes the proof of Lemma A.1. \square

Let $h : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ be measurable. For $x, y \in \mathbb{R}$, $x \leq y$ and $s, f \in \mathbb{Z}$, $s \leq f$, we associate to any element $\phi \in SC_{(x,s) \rightarrow (y,f)}^1$ the h -rewarded energy $E(\phi) + h(x)$. The maximum line-to-point h -rewarded energy is then defined to be

$$M_{(*;h,s) \rightarrow (y,f)}^1 := \sup \left\{ E(\phi) + h(x) : \phi \in SC_{(x,s) \rightarrow (y,f)}^1, x \leq y \right\}.$$

Lemma A.2. *Let $y \in \mathbb{R}$. Let $h : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ be measurable, with $h(x) > -\infty$ for some $x < y$, and $\limsup_{x \rightarrow -\infty} h(x)/|x| < 0$. Let $s, f \in \mathbb{Z}$ satisfy $s \leq f$. Then, except on a \mathbb{P} -null set, $M_{(*;h,s) \rightarrow (y,f)}^1$ is finite, and there is a unique choice of (x, ϕ) such that $x \in (-\infty, y]$, $\phi \in SC_{(x,s) \rightarrow (y,f)}^1$ and the h -rewarded energy of ϕ attains $M_{(*;h,s) \rightarrow (y,f)}^1$.*

Proof. In this result, we consider the h -rewarded energy $E(\phi) + h(x)$ of staircases ϕ that end at a given location $(y, f) \in \mathbb{R} \times \mathbb{N}$ and that begin at a location $(x, s) \in \mathbb{R} \times \mathbb{N}$ whose height $s \leq f$ is fixed but whose first coordinate x may vary on $(-\infty, x]$.

We must first argue that the maximum h -rewarded energy $M_{(*;h,s) \rightarrow (y,f)}^1$ adopted among such staircases ϕ is a finite quantity. Due to our assumption that $\limsup_{x \rightarrow -\infty} |x|^{-1}h(x)$ is negative, it suffices to argue that $M_{(x,s) \rightarrow (y,f)}^1$, the energy maximum over such staircases that begin at (x, s) , grows sublinearly in $|x|$ the limit $x \rightarrow -\infty$. (In fact, we also need to know that the supremum of $M_{(x,s) \rightarrow (y,f)}^1$ over x in any bounded interval bordered on the right by y is almost surely finite.) A crude estimate on $M_{(x,s) \rightarrow (y,f)}^1$ is

$$M_{(x,s) \rightarrow (y,f)}^1 \leq \sum_{k=s}^f \left(\sup_{z \in [x,y]} B(k, z) - \inf_{z \in [x,y]} B(k, z) \right)$$

The distribution of the summand is unchanged if we suppose that $B(k, x)$ is zero. Supposing that this is so, the summand $A - B$ may be viewed in the form $A + (-B)$. It is then a sum of two terms that share their distribution. By the reflection principle, this distribution assigns mass $2 \cdot (2\pi)^{-1/2}(y-x)^{-1} \exp\{-2^{-1}(y-x)^{-1}r^2\}$ to any interval (r, ∞) for $r \geq 0$. We see then that

$$\mathbb{P} \left(M_{(x,s) \rightarrow (y,f)}^1 \geq 2kr \right) \leq 2k \cdot 2 \cdot (2\pi)^{-1/2}(y-x)^{-1} \exp\{-2^{-1}(y-x)^{-1}r^2\} \quad (27)$$

for any $r \geq 0$. Fixing any $\eta > 0$, we may choose $r_x = (2^{1/2} + \eta)(y-x)^{1/2}(\log(y-x))^{1/2}$, we may sum this bound over $x \in y - \mathbb{N}$ to find that the event $M_{(x,s) \rightarrow (y,f)}^1 \geq 2kr_x$ may occur for only finitely many such x .

Some understanding of the local regularity of $M_{(x,s)\rightarrow(y,f)}^1$ as x varies is now needed to treat the general case of $x \in (-\infty, y]$. Let $x_1, x_2 \leq y$. Note that

$$M_{(x_1,s)\rightarrow(y,f)}^1 - M_{(x_2,s)\rightarrow(y,f)}^1 \geq B(s, x_2) - B(s, x_1).$$

Writing this in the form $M_{(x_2,s)\rightarrow(y,f)}^1 - M_{(x_1,s)\rightarrow(y,f)}^1 \leq B(s, x_1) - B(s, x_2)$, we may fix $x_1 \leq y$, and find that

$$\sup_{x_2 \in [x_1-1, x_1]} \left(M_{(x_2,s)\rightarrow(y,f)}^1 - M_{(x_1,s)\rightarrow(y,f)}^1 \right) \leq B(s, x_1) - \inf_{x_2 \in [x_1-1, x_1]} B(s, x_2).$$

For any $r > 0$, the right-hand side exceeds r with probability $2 \cdot (2\pi)^{-1/2} \exp\{-2^{-1}r^2\}$ by a second use of the reflection principle. Combining with (27), we see that, for any $x \leq y - 1$,

$$\mathbb{P} \left(\sup_{z \in [x-1, x]} M_{(z,s)\rightarrow(y,f)}^1 \geq 2(k+1)r \right) \leq 2(2k+1)(2\pi)^{-1/2}(y-x)^{-1} \exp\{-2^{-1}(y-x)^{-1}r^2\}. \quad (28)$$

Maintaining our choice $r_x = (2^{1/2} + \eta)(y-x)^{1/2}(\log(y-x))^{1/2}$, we see that, for any given $\eta > 0$, the set of $x \leq y$ for which $M_{(x,s)\rightarrow(y,f)}^1 \geq 2(k+1)r_{x-1}$ is almost surely bounded below. We thus confirm that $M_{(x,s)\rightarrow(y,f)}^1$ grows sublinearly in $|x|$ as $x \rightarrow -\infty$. To prove the finiteness of $M_{(*:h,s)\rightarrow(y,f)}^1$, it remains only to prove the almost sure finiteness of the supremum of $M_{(x,s)\rightarrow(y,f)}^1$ as x varies over any bounded interval bordered by y on the right. However, this is again a consequence of (27) and (28).

To complete the proof of Lemma A.2, it remains to argue that there is a unique choice of (x, ϕ) with $x \leq y$ and $\phi \in D_{(x,s)\rightarrow(y,f)}^1$ for which $E(\phi) + h(x)$ attains the maximum value $M_{(*:h,s)\rightarrow(y,f)}^1$. The resampling proof of Lemma A.1 works here. Note the simplification that we consider staircases rather than ℓ -tuples of staircases: or formally, we take $\ell = 1$.

We revisit the resampling proof, now considering any given horizontal rational interval I contained in $(-\infty, y] \times \{s, f\}$, and defining any staircase in $D_{(x,s)\rightarrow(y,f)}^1$ for any $x \leq y$ to inhabit or avoid I as we did before. The event A_I is now specified by the existence of one staircase in the union of $D_{(x,s)\rightarrow(y,f)}^1$ over $x \leq y$ that inhabits I and of another in the same union that avoids I . As before, it is enough to show that $\mathbb{P}(A_I) = 0$ for any given horizontal rational interval I , this time contained in $(-\infty, y] \times \{s, f\}$. The proof that this probability is zero is unchanged.

This completes the proof of Lemma A.2. □

Proof of Lemma 4.6: (1). The staircase set $SC_{(2n^{2/3}x,0)\rightarrow(n+2n^{2/3}y,n)}^1$, which is in correspondence under the scaling map $R_n : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ with the collection of n -zigzags in question, is non-empty if and only if $y \geq x - n^{1/3}/2$. When this last condition is satisfied, the polymer weight $W_{n;(x,0)}^{(y,1)}$ equals

$$2^{-1/2}n^{-1/3} \left(M_{(2n^{2/3}x,0)\rightarrow(n+2n^{2/3}y,n)}^1 - n - 2n^{2/3}(y-x) \right).$$

The uniqueness of the n -polymer in question then follows from Lemma A.1 with $\ell = 1$, $\mathbf{x}_1 = 2n^{2/3}x$, $\mathbf{y}_1 = n + 2n^{2/3}y$, $\mathbf{s}_1 = 0$ and $\mathbf{f}_1 = n$.

(2). Let $y \in [-1, 1]$. It is enough to check that $y \in \text{PolyUnique}_{n;(*:f,0)}^1$ almost surely. Recall that the f -rewarded line-to-point polymer weight $\mathbb{W}_{n;(*:f,0)}^{(y,1)}$ equals

$$\begin{aligned} & \sup_{x \in (-\infty, n^{1/3}/2+y]} \left(2^{-1/2} n^{-1/3} \left(M_{(2n^{2/3}x,0) \rightarrow (n+2n^{2/3}y,n)}^1 - n - 2n^{2/3}(y-x) \right) + f(x) \right) \\ &= 2^{-1/2} n^{-1/3} \sup_{u \in (-\infty, n+v]} \left(M_{(u,0) \rightarrow (n+v,n)}^1 - n - v + u + h(u) \right), \end{aligned}$$

where $h : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ is given by $h(x) = 2^{1/2} n^{1/3} f(n^{-2/3}x/2)$. We took $v = 2n^{2/3}y$ and $u = 2n^{2/3}x$ to obtain the displayed equality. We seek to apply Lemma A.2 with $\mathbf{h} : \mathbb{R} \rightarrow \mathbb{R} \cup \{-\infty\}$ given by $\mathbf{h}(u) = -n - v + u + h(u)$, $\mathbf{s} = 0$, $\mathbf{f} = n$ and $\mathbf{y} = n + v$. Indeed, $\mathbb{W}_{n;(*:f,0)}^{(y,1)}$ has been shown to equal $2^{-1/2} n^{-1/3} M_{(*:\mathbf{h},\mathbf{s}) \rightarrow (\mathbf{y},\mathbf{f})}^1$, so that the application of the lemma will imply that $y \in \text{PolyUnique}_{n;(*:f,0)}^1$ almost surely. It remains to check that Lemma A.2's hypotheses are satisfied. In this regard, note that

$$\begin{aligned} \mathbf{h}(u) &\leq -n - v + u + 2^{1/2} n^{1/3} \Psi_1 (1 + n^{-2/3}|u|/2) \\ &\leq -n - v + u + 2^{1/2} n^{1/3} \Psi_1 + 2^{-1/2} \Psi_1 n^{-1/3} |u|, \end{aligned}$$

so that the hypothesis that $\limsup_{u \rightarrow -\infty} \mathbf{h}(u)/|u| < 0$ is satisfied when $n > 2^{-3/2} \Psi_1^3$. The other hypothesis, that $\mathbf{h}(u) > -\infty$ for some $u \leq n + v$, is validated due to $\sup_{x \in [-\Psi_2, \Psi_2]} f(x) \geq -\Psi_3$ and $2n^{2/3} \Psi_2 \leq n + v$ or equivalently $n \geq 8(\Psi_2 - y)^3$. \square

APPENDIX B. CALCULATIONAL DERIVATIONS

In Subsection 4.7.2, we explained that one aspect of the proofs of two of our results, Proposition 6.1 and Lemma 8.1, has been separated from the rest. The missing piece is the ‘calculational derivations’ of these results. Here, we present these derivations. In the case of each concerned result, it is now our task to verify that the result’s hypotheses are sufficient for the purpose of verifying every condition that is invoked during the course of the result’s proof.

A guideline of the format of the derivations’ presentation is that first all of the conditions invoked in the proof of the result in question are recorded. These conditions may then be split into lists, according to which parameters are being bounded. Each list is then separately simplified to produce a shorter list that implies the original. Sometimes in carrying out this simplification, new conditions, called *further conditions*, will be introduced. The simplified lists may then be compared, with some further simplifications being noted. The derivation ends when a final list is drawn up whose conditions coincide with the hypotheses of the result under review.

Some notational conventions govern the presentation of the derivations. Conditions are given square bracketed names such as [1] or [n4], shown on the left. These names may be recycled from one calculational derivation to the next. The notation [1, n4] means ‘conditions [1] and [n4]’. Implication is denoted by a right arrow, so that [1, n4] \rightarrow [r3] means ‘conditions [1] and [n4] imply condition [r3]’.

The notation [n2, 4, 5] is used as a shorthand for [n2, n4, n5]. This meaning would be ambiguous, were there named conditions [4] or [5], but we employ this condition only when the meaning is unambiguous.

There are three parts to the section. In the first, a small explicit computation from Proposition 6.1's proof is provided. The next two provide the calculational derivations of this proposition and of Lemma 8.1.

B.1. A short piece of working in the proof of Proposition 6.1. We begin by recording a short piece of working needed at the very end of the proof of this result.

Deriving the upper bound in the conclusion of Proposition 6.1. At the very end of the proof of Proposition 6.1, it is claimed that the sum of $10^{20}KCc_1^{-4/3}(\log \epsilon^{-1})^{4/3}\epsilon^2$ and the expression in line (11) is at most $\epsilon^2 \cdot 10^{420}Kc_3^{-33}C_3(\log \epsilon^{-1})^{42} \exp\{\beta_3(\log \epsilon^{-1})^{5/6}\}$. Here is the working:

$$\begin{aligned} & \epsilon^2 \cdot 10^{356}Kr^4c_3^{-27}C_3(\log \epsilon^{-1})^{36} \exp\{\beta_3(\log \epsilon^{-1})^{5/6}\} + 10^{20}KCc_1^{-4/3}(\log \epsilon^{-1})^{4/3}\epsilon^2 \\ = & \epsilon^2 \cdot 10^{356}K \cdot 10^{176/3}2^{44/3}c_1^{-16/3}(\log \epsilon^{-1})^{16/3} \cdot c_3^{-27}C_3(\log \epsilon^{-1})^{36} \exp\{\beta_3(\log \epsilon^{-1})^{5/6}\} \\ & + 10^{20}KCc_1^{-4/3}(\log \epsilon^{-1})^{4/3}\epsilon^2 \\ \leq & \epsilon^2 \cdot 2^{-1}10^{420}Kc_3^{-33}C_3(\log \epsilon^{-1})^{42} \exp\{\beta_3(\log \epsilon^{-1})^{5/6}\} + 10^{20}KCc_3^{-4/3}(\log \epsilon^{-1})^{4/3}\epsilon^2 \\ \leq & \epsilon^2 \cdot 10^{420}Kc_3^{-33}C_3(\log \epsilon^{-1})^{42} \exp\{\beta_3(\log \epsilon^{-1})^{5/6}\}. \end{aligned}$$

In the penultimate inequality, we used $10^{356+176/3}2^{44/3} = 10^{414+2/3}26007\dots = 10^{414+2/3+4.4\dots} \in [10^{419}, 1/2 \cdot 10^{420}]$, $\epsilon \leq e^{-1}$ and $c_3 \leq c_1 \leq 1$. In the final inequality, we used $C_3 \geq C$, $c_3 \leq 1$ and $\epsilon \leq e^{-1}$.

B.2. Proposition 6.1: derivation. Here we present the calculational derivation of Proposition 6.1

We begin by collecting together all the conditions that are invoked during the proof of this proposition. The three parameters in the proposition's statement are n , ϵ and K . It is in terms of these that the various conditions are expressed, sometimes by means of a quantity $r > 0$. The value of this quantity is during the proof of the proposition set to be equal to

$$r = 10^{44/3}2^{11/3}c_1^{-4/3}(\log \epsilon^{-1})^{4/3}.$$

Record of all the conditions invoked during the proof of Proposition 6.1.

In the application of Theorem 4.2, the bounds $\epsilon \leq (1 - 2^{-3/2})^{2/3}$, $r \geq 1$ and $K + 2 \leq \epsilon^{-1/2}$ are used. The conditions $\mathbf{t}_{1,2} \geq 1/2$ and $(\mathbf{t}_{1,2})^{2/3} \geq 1/2$ are also used. Since $\mathbf{t}_{1,2} = 1 - \epsilon^{3/2}$, these are implied by $\epsilon \leq 2^{-2/3}$. In summary, these bounds are implied by

$$\epsilon \leq 2^{-2/3}$$

$$r \geq 1$$

and

$$K + 2 \leq \epsilon^{-1/2}.$$

The conditions needed to satisfy the hypotheses of Theorem 4.2 are noted to be

$$(r + 2)\epsilon \leq \min \left\{ (\eta_0)^{36}, 10^{-616}c_3^{22}3^{-115}, \exp\{-C^{1/4}\} \right\},$$

$$n \geq 2 \max \left\{ 2(K_0)^9(\log \epsilon^{-1})^{K_0}, 10^{584}c_3^{-48}3^{240}c^{-36}\epsilon^{-222} \max\{1, (K + 2)^{36}2^{24}\}, a_0^{-9}(K + 2)^92^6 \right\}.$$

and

$$K + 2 \leq 2^{-1}(r + 2)^{-1/2}\epsilon^{-1/2}.$$

The application of Lemma 6.2 makes use of $\epsilon \leq 2^{-2/3}$. After this, the bounds $\epsilon \leq 1/3$, $K \geq 1$ and $r \geq 1$ are used.

When Theorem 4.8 is applied, the bounds $r \geq 1$,

$$K \geq 2$$

and

$$\epsilon \leq 1/4$$

are used. In order that this theorem can be applied, it is noted that it suffices that

$$\epsilon^{3/2} \leq 10^{-11}c_1^2,$$

$$n \geq \max \left\{ 10^{29}\epsilon^{-75/2}c^{-18}, 10^{24}2^{36}c^{-18}\epsilon^{-75/2}K^{36} \right\},$$

$$r \geq 2 \max \left\{ 10^9c_1^{-4/5}, 15C^{1/2}, 22\epsilon^{1/2}K \right\},$$

and

$$r \leq 6\epsilon^{25/6}n^{1/36}.$$

Dividing the conditions into several lists.

The listed conditions are now partitioned into lists. A further condition, [ε4], is added at this stage.

The r list. In this list, we gather all conditions that concern the parameter r . Lower bounds on r appear first.

$$[r1] \ r \geq 1$$

$$[r2] \ r \geq 2 \max \left\{ 10^9c_1^{-4/5}, 15C^{1/2}, 22\epsilon^{1/2}K \right\},$$

$$[r3] \ (r + 2)\epsilon \leq \min \left\{ (\eta_0)^{36}, 10^{-616}c_3^{22}3^{-115}, \exp \{ -C^{1/4} \} \right\},$$

$$[r4] \ K + 2 \leq 2^{-1}(r + 2)^{-1/2}\epsilon^{-1/2}.$$

$$[r5] \ r \leq 6\epsilon^{25/6}n^{1/36}.$$

We also have the formula for the value of r :

$$[rV] \ r = 10^{44/3}2^{11/3}c_1^{-4/3}(\log \epsilon^{-1})^{4/3}.$$

The n list. This list consists of the remaining conditions that concern n .

$$[n1] \ n \geq 2 \max \left\{ 2(K_0)^9(\log \epsilon^{-1})^{K_0}, 10^{584}c_3^{-48}3^{240}c^{-36}\epsilon^{-222} \max \{ 1, (K+2)^{36}2^{24} \}, a_0^{-9}(K+2)^92^6 \right\}.$$

$$[n2] \ n \geq \max \left\{ 10^{29}\epsilon^{-75/2}c^{-18}, 10^{24}2^{36}c^{-18}\epsilon^{-75/2}K^{36} \right\},$$

The ϵ list. This list consists of the remaining conditions that concern ϵ .

$$\begin{aligned} [\epsilon 1] \quad & \epsilon \leq 1/4 \\ [\epsilon 2] \quad & \epsilon^{3/2} \leq 10^{-11} c_1^2, \\ [\epsilon 3] \quad & K + 2 \leq \epsilon^{-1/2}. \\ [\epsilon 4] \quad & \epsilon \leq 10^{-65} c_1^6 \end{aligned}$$

The K list. This list consists of the remaining conditions that concern K .

$$[K 1] \quad K \geq 2.$$

Beginning the analysis of the conditions. First we make a claim.

Claim. When all the listed conditions are in force, the bound

$$\epsilon \leq 10^{-65} c_1^6$$

implies that

$$(r + 2)\epsilon \leq \epsilon^{1/2}.$$

Proof. By [rV], the condition $(r + 2)\epsilon \leq \epsilon^{1/2}$ is given by

$$10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3} + 2 \leq \epsilon^{-1/2}$$

which is implied by

$$2 \cdot 10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3} \leq \epsilon^{-1/2}$$

and

$$4 \leq \epsilon^{-1/2};$$

the second of these, namely $\epsilon \leq 4^{-2}$, follows from [\epsilon 2] and $c_1 \leq 1$. Note that

$$2 \cdot 10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3} \leq \epsilon^{-1/2}$$

holds provided that

$$(\log \epsilon^{-1})^{4/3} \leq \epsilon^{-1/4}$$

and

$$2 \cdot 10^{44/3} 2^{11/3} c_1^{-4/3} \leq \epsilon^{-1/4}.$$

The second of these bounds is equivalent to

$$\epsilon \leq 2^{-4} 10^{-176/3} 2^{-44/3} c_1^{16/3}.$$

Since $2^{4+44/3} 10^{176/3} = 2^{56/3} 10^{176/3} = 1.93 \dots \times 10^{64}$, and $c_1 \leq 1$, we see that the last displayed bound is implied by $\epsilon \leq 10^{-65} c_1^6$, which is the hypothesis in the claim.

The first of the two bounds, namely $(\log \epsilon^{-1})^{4/3} \leq \epsilon^{-1/4}$, is equivalent to

$$\epsilon^{-1} \leq \exp \{ \epsilon^{-3/16} \}$$

which is implied by

$$\epsilon^{-1} \leq (6!)^{-1} (\epsilon^{-3/16})^6 = (6!)^{-1} \epsilon^{-9/8}$$

which is implied by $\epsilon \leq (6!)^{-8}$, a condition which follows from the claim's hypothesis $\epsilon \leq 10^{-65} c_1^6$ alongside $c_1 \leq 1$. This proves the claim. \square

Simplifying the r list. We use the claim to simplify the r list.

Note that $[\epsilon 4]$ is the hypothesis of the claim.

We see from the claim that, when $[\epsilon 4]$ holds, $[r 3]$ is implied by

$$[r 6] \quad \epsilon \leq \min \left\{ (\eta_0)^{72}, 10^{-1232} c_3^{44} 3^{-230}, \exp \left\{ -2C^{1/4} \right\} \right\},$$

Note that $[r 2] \rightarrow [r 1]$ due to $c_1 \leq 1$.

The condition $[r 4]$ is equivalent to

$$(r + 2)\epsilon \leq 4^{-1}(K + 2)^{-2}$$

which due to $[\epsilon 4]$ and the claim is implied by

$$[r 7] \quad \epsilon \leq 4^{-2}(K + 2)^{-4}.$$

The condition $[r 5]$ is, in view of $[r V]$ and $\log \epsilon^{-1} \leq \epsilon^{-1}$, implied by

$$10^{44/3} 2^{11/3} c_1^{-4/3} \leq 6\epsilon^{25/6+4/3} n^{1/36}$$

or equivalently

$$10^{528} 2^{132} 6^{-36} c_1^{-48} \epsilon^{-198} \leq n$$

Since $2^{132} 6^{-36} = 5.27 \dots \times 10^{11}$, this last condition is implied by

$$n \geq 10^{540} c_1^{-48} \epsilon^{-198}.$$

Since $c_3 \leq c_1$, $c \leq 1$ and $\epsilon \leq 1$, the last is implied by the second condition in $[n 1]$. That is, $[n 1] \rightarrow [r 5]$.

Thus $[r 2, 6, 7, n 1, \epsilon 4] \rightarrow [r 1, 2, 3, 4, 5]$. We see then that we may simplify the $[r]$ list and present it in the form $[r 2, 6, 7]$.

Simplifying the n list. We now make some straightforward comments about the n conditions.

In the second condition of $[n 1]$, we see the term $\max \{1, (K + 2)^{36} 2^{24}\}$. Obviously this expression is attained by the latter quantity. Thus, the lower bound on n in this second condition equals

$$10^{584} 2^{24} 3^{240} c_3^{-48} c^{-36} \epsilon^{-222} (K + 2)^{36}$$

Since $2^{24} 3^{240} = 5.41 \dots \times 10^{121}$, the displayed value is bounded above by

$$10^{606} c_3^{-48} c^{-36} \epsilon^{-222} (K + 2)^{36}$$

Thus, writing

$$[n 3] \quad n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606} c_3^{-48} c^{-36} \epsilon^{-222} (K + 2)^{36}, a_0^{-9} (K + 2)^9 2^6 \right\},$$

we see that $[n 3] \rightarrow [n 1]$.

The condition $[n 2]$ is comprised of two lower bounds on n . The first of these is implied by the second bound in $[n 3]$ due to $\epsilon \leq 1$, $c_3 \leq 1$ and $c \leq 1$. The second of these is also implied by the same bound for the same reasons. Thus, $[n 3] \rightarrow [n 1]$.

We see then that a simplified form of the n list may take the form of $[n 3]$.

Simplifying the ϵ list.

$[\epsilon 2]$ may be rewritten $\epsilon \leq 10^{-22/3} c_1^{4/3}$ and thus is implied by

$$[\epsilon 5] \quad 10^{-8} c_1^{4/3}.$$

Note that $[\epsilon 5] \rightarrow [\epsilon 1]$ is due to $c_1 \leq 1$.

$[\epsilon 3]$ is equivalent to

$$[\epsilon 6] \epsilon \leq (K + 2)^{-2}$$

Note that $[\epsilon 4] \rightarrow [\epsilon 5]$ is due to $c_1 \leq 1$.

Thus, $[\epsilon 4, 6] \rightarrow [\epsilon 1, 2, 3, 4]$, and thus $[\epsilon 4, 6]$ is a simplified form for the ϵ -list.

The K list. This one-condition list will not be simplified.

Gathering the simplified lists.

We may summarise our progress by recalling that the union of our simplified lists is: $[r2, 6, 7]$, $[n3]$, $[\epsilon 4, 6]$ and $[K1]$. These conditions collectively imply all the bounds needed in the proof of Proposition 6.1. Our next step is to record this collection of conditions again, taking the opportunity to reallocate some into other lists where they more suitably belong. We rename the conditions that change lists, indicating the old and the new names as we record the conditions now.

The updated r list.

$$[r2 = R1] r \geq 2 \max \left\{ 10^9 c_1^{-4/5}, 15C^{1/2}, 22\epsilon^{1/2}K \right\},$$

The updated n list.

$$[n3 = N1] n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606} c_3^{-48} c^{-36} \epsilon^{-222} (K + 2)^{36}, a_0^{-9} (K + 2)^9 2^6 \right\},$$

The updated ϵ list.

$$\begin{aligned} [\epsilon 4 = E1] \epsilon &\leq 10^{-65} c_1^6 \\ [\epsilon 6 = E2] \epsilon &\leq (K + 2)^{-2} \\ [r6 = E3] \epsilon &\leq \min \left\{ (\eta_0)^{72}, 10^{-1232} c_3^{44} 3^{-230}, \exp \left\{ -2C^{1/4} \right\} \right\}, \\ [r7 = E4] \epsilon &\leq 4^{-2} (K + 2)^{-4}. \end{aligned}$$

The K list.

$$[K1] K \geq 2.$$

Further analysis. A few further simplifications will be made. The most important of these address the condition $[R1]$, which is the remaining instance of a condition that involves the parameter r . This condition should be eliminated and replaced by an upper bound on ϵ . To do this, recall the value of r :

$$[rV] r = 10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3}.$$

Since $\epsilon \leq 1$, due to $[E2]$, and $c_1 \leq 1$, $[rV]$ implies the first of the three conditions in $[R1]$. The third of these conditions is in light of $[E4]$ implied by $r \geq 11$; since $C \geq 1$, this bound is implied by the second condition in $[R1]$.

Using $[rV]$, the second condition in $[R1]$ is seen to be equivalent to

$$10^{44/3} 2^{11/3} c_1^{-4/3} (\log \epsilon^{-1})^{4/3} \geq 30C^{1/2}$$

or equivalently

$$\epsilon \leq \exp \left\{ - \left(10^{-44/3} 2^{-11/3} c_1^{4/3} \cdot 30C^{1/2} \right)^{3/4} \right\}$$

Since $10^{-44/3}2^{-11/3} \cdot 30$ and c_1 are both at most one, this last is implied by

$$[R2] \ \epsilon \leq \exp \{ -2C^{3/8} \}$$

Note that $[R2]$ implies the third condition in $[E3]$ due to $C \geq 1$.

We may summarise these inferences by introducing

$$[E5] \ \epsilon \leq \min \left\{ (\eta_0)^{72}, 10^{-1232}c_3^{44}3^{-230}, \exp \{ -2C^{3/8} \} \right\},$$

and noting that $[E1, 2, 4, 5]$ implies the above E list, $[E1, 2, 3, 4]$, as well as the R list $[R1]$.

Having eliminated the parameter r from our conditions, we make now further simplify the E list, which is presently $[E1, 2, 4, 5]$. Note that $[E4] \rightarrow [E2]$. Writing

$$[E6] \ \epsilon \leq \min \left\{ (\eta_0)^{72}, 10^{-1342}c_3^{44}, \exp \{ -2C^{3/8} \} \right\},$$

we see that $[E6] \rightarrow [E5]$ is due to $3^{230} = 5.46 \dots \times 10^{109}$. Moreover, $[E6] \rightarrow [E1]$ because the second condition in $[E6]$ implies $[E1]$ due to $c_3 \leq c_1$. Thus we may update the E list to be $[E4, 6]$.

Regarding the condition $[N1]$, we merely note that, since $c_3 \leq c$, we have $[N2] \rightarrow [N1]$, where

$$[N2] \ n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606}c_3^{-84}\epsilon^{-222}(K+2)^{36}, a_0^{-9}(K+2)^9 2^6 \right\}.$$

The status report is that all conditions have been shown to be implied by the conditions: $[E1, 4, 6]$, $[N2]$ and $[K1]$.

The upper bounds on ϵ , $[E4, 6]$, may be written:

$$\epsilon \leq \min \left\{ 4^{-2}(K+2)^{-4}, (\eta_0)^{72}, 10^{-1342}c_3^{44}, \exp \{ -2C^{3/8} \} \right\},$$

The lower bound on n , $[N2]$, is

$$n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606}c_3^{-84}\epsilon^{-222}(K+2)^{36}, a_0^{-9}(K+2)^9 2^6 \right\}.$$

The lower bound on K , $[K1]$, is

$$K \geq 2.$$

The last three displayed bounds collectively form the hypotheses of Proposition 6.1. This completes the calculational derivation of this proposition.

B.3. Lemma 8.1: derivation. We now turn to the calculational derivation of this lemma. The result has parameters $n \in \mathbb{N}$, $\epsilon > 0$, $D > 0$, $\chi > 0$ and $\bar{\Psi} \in (0, \infty)^3$. It demands certain hypotheses on these parameters. Naturally, our job is to verify that the hypotheses made on the parameters are sufficient to imply all the conditions used during the proof of the result.

Before we begin, we note that one condition, which is not an estimate, is needed in the derivation, when Proposition 6.1 is applied. This is $n\epsilon^{3/2} \in \mathbb{N}$. This condition is indeed hypothesised in Lemma 8.1 and we will not refer to it again.

Record of all the conditions used during the proof of Lemma 8.1

The proof consists of derivations of upper bounds on the four terms on the right-hand side of (16), and then the assembly of these four estimates. We record the used conditions accordingly.

Bounding $\mathbb{P}(\text{ManyCanopy})$.

The first hypothesis is invoked implicitly, when Lemma 4.6(1) is used. The condition needed is $\epsilon[\epsilon^{-1}y_i] \geq -D(\log \epsilon^{-1})^{1/3} - n^{1/3}/2$, where recall that the various points y_i each belong to $[-1, 1]$. Thus, it is sufficient that $1 + \epsilon \leq D(\log \epsilon^{-1})^{1/3} + n^{1/3}/2$. Imposing $\epsilon \leq 1$, we see that it suffices that

$$1 \leq D(\log \epsilon^{-1})^{1/3}$$

and

$$2 \leq n^{1/3}.$$

Next we use $\epsilon^{3/2} \leq 2^{-1}$ and $\epsilon \leq e^{-1}$. After this, $\epsilon \leq 4^{-1/\chi}$. Then we use $\epsilon \leq \exp\{-\chi^{-2}\}$ and $\chi \leq 1$.

In the application of Theorem 4.1, it suffices that

$$4^{-1}\epsilon^{-\chi} \geq k_0 \vee (4D(\log \epsilon^{-1})^{4/3} + 2\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil)^3,$$

and

$$n\epsilon^{3/2} \geq \max \left\{ 2(K_0)^{m^2} (\log(2^{-1}\epsilon^{-\chi}))^{K_0}, a_0^{-9}\tau^9, 10^{325}c^{-36}(2^{-1}\epsilon^{-\chi})^{465} \max\{1, \tau^{36}\} \right\},$$

where we write $\tau = 4D(\log \epsilon^{-1})^{4/3} + 2\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil$.

Bounding $\mathbb{P}(\text{LateCoal})$.

In applying Proposition 6.1, we make use of

$$\epsilon \leq \min \left\{ \exp\{-8D^{-3}\}, 4^{-2}(D(\log \epsilon^{-1})^{1/3} + 2)^{-4}, (\eta_0)^{72}, 10^{-1342}c_3^{44}, \exp\{-2C^{3/8}\} \right\}$$

and

$$n \geq 2 \max \left\{ 2(K_0)^9 (\log \epsilon^{-1})^{K_0}, 10^{606}c_3^{-84}\epsilon^{-222}(D(\log \epsilon^{-1})^{1/3} + 2)^{36}, a_0^{-9}(D(\log \epsilon^{-1})^{1/3} + 2)^9 2^6 \right\}.$$

Bounding $\mathbb{P}(\neg \text{RegFluc})$.

In applying Lemma 4.10, we use the conditions

$$n \geq c^{-18} \max \left\{ (\Psi_2 + 1)^9, 10^{23}\Psi_1^9, 3^9 \right\},$$

and

$$D(\log \epsilon^{-1})^{1/3} - 1 \in \left[39\Psi_1 \vee 5 \vee 3c^{-3} \vee 2((\Psi_2 + 1)^2 + \Psi_3)^{1/2}, 6^{-1}cn^{1/9} \right].$$

Bounding $\mathbb{P}(\neg \cap \text{PolyDevReg})$.

The application of Proposition 4.9 is permitted in view of:

$$\epsilon^{3/2} \leq 10^{-11}c_1^2,$$

$$n \geq \max \left\{ 10^{29}\epsilon^{-75/2}c^{-18}, 10^{24}c^{-18}\epsilon^{-75/2}(D(\log \epsilon^{-1})^{1/3} + 1)^{36} \right\}.$$

and

$$D(\log \epsilon^{-1})^{4/3} \in \left[10^9c_1^{-4/5} \vee 15C^{1/2} \vee 11\epsilon^{1/2}(D(\log \epsilon^{-1})^{1/3} + 1), 3\epsilon^{25/6}n^{1/36} \right].$$

Assembling the estimates. Upper bounds are used at this moment of the proof. These are:

$$\begin{aligned} & \epsilon^{-3+\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \\ & \leq 2^{-1} \hat{H}^{-1} (2 \lceil 8D(\log \epsilon^{-1})^{4/3} \rceil)^{-(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288-3/2} \exp \{ -2(\log \epsilon^{-1})^{11/12} \}, \\ & \epsilon^{2^{-13}cD^3-2} \leq (38)^{-1} D^{-1} (\log \epsilon^{-1})^{-1/3} C^{-1} \end{aligned}$$

and

$$\epsilon^{10^{-11}c_1D^{3/4}-3} \leq 8^{-1} 22^{-1} c_1 C^{-1} D^{-1} (\log \epsilon^{-1})^{-4/3}.$$

Expressing the gathered conditions in lists. We now partition the various conditions into several lists. A few conditions will be expressed in terms of a parameter τ , whose value is now set: Set

$$[\tau] \tau = 4D(\log \epsilon^{-1})^{4/3} + 2 \lceil 8D(\log \epsilon^{-1})^{4/3} \rceil.$$

The $[\epsilon]$ list. Here we record all conditions on ϵ that do not involve n .

$$\begin{aligned} & [\epsilon 1] \quad 1 \leq D(\log \epsilon^{-1})^{1/3} \\ & \quad [\epsilon 2] \quad \epsilon^{3/2} \leq 2^{-1} \\ & \quad [\epsilon 3] \quad \epsilon \leq e^{-1} \\ & \quad [\epsilon 4] \quad \epsilon \leq 4^{-1/\chi} \\ & \quad [\epsilon 5] \quad \epsilon \leq \exp \{ -\chi^{-2} \} \\ & \quad [\epsilon 6] \quad 4^{-1}\epsilon^{-\chi} \geq k_0 \\ & \quad [\epsilon 7] \quad 4^{-1}\epsilon^{-\chi} \geq \tau^3, \\ & \quad [\epsilon 8] \quad \epsilon \leq \exp \{ -8D^{-3} \} \\ & [\epsilon 9] \quad \epsilon \leq 4^{-2} (D(\log \epsilon^{-1})^{1/3} + 2)^{-4} \\ & \quad [\epsilon 10] \quad \epsilon \leq (\eta_0)^{72} \\ & \quad [\epsilon 11] \quad \epsilon \leq 10^{-1342} c_3^{44} \\ & \quad [\epsilon 12] \quad \epsilon \leq \exp \{ -2C^{3/8} \} \\ & \quad [\epsilon 13] \quad D(\log \epsilon^{-1})^{1/3} - 1 \geq 5 \\ & \quad [\epsilon 14] \quad D(\log \epsilon^{-1})^{1/3} - 1 \geq 3c^{-3} \\ & [\epsilon 15] \quad D(\log \epsilon^{-1})^{1/3} - 1 \geq 39\Psi_1 \vee 2((\Psi_2 + 1)^2 + \Psi_3)^{1/2} \\ & \quad [\epsilon 16] \quad \epsilon^{3/2} \leq 10^{-11} c_1^2 \\ & \quad [\epsilon 17] \quad D(\log \epsilon^{-1})^{4/3} \geq 10^9 c_1^{-4/5} \\ & \quad [\epsilon 18] \quad D(\log \epsilon^{-1})^{4/3} \geq 15C^{1/2} \\ & [\epsilon 19] \quad D(\log \epsilon^{-1})^{4/3} \geq 11\epsilon^{1/2} (D(\log \epsilon^{-1})^{1/3} + 1) \\ & [\epsilon 20] \quad \epsilon^{-3+\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \\ & \leq 2^{-1} \hat{H}^{-1} (2 \lceil 8D(\log \epsilon^{-1})^{4/3} \rceil)^{-(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288-3/2} \exp \{ -2(\log \epsilon^{-1})^{11/12} \} \\ & \quad [\epsilon 21] \quad \epsilon^{2^{-13}cD^3-2} \leq (38)^{-1} D^{-1} (\log \epsilon^{-1})^{-1/3} C^{-1} \\ & [\epsilon 22] \quad \epsilon^{10^{-11}c_1D^{3/4}-3} \leq 8^{-1} 22^{-1} c_1 C^{-1} D^{-1} (\log \epsilon^{-1})^{-4/3} \end{aligned}$$

The $[n]$ list. Here are recorded all conditions concerning n .

$$\begin{aligned}
[n1] \quad & 2 \leq n^{1/3} \\
[n2] \quad & n\epsilon^{3/2} \geq 2(K_0)^{(12)^{-2}} (\log \log(2^{-1}\epsilon^{-\chi}))^2 (\log(2^{-1}\epsilon^{-\chi}))^{K_0} \\
[n3] \quad & n\epsilon^{3/2} \geq a_0^{-9} \tau^9 \\
[n4] \quad & n\epsilon^{3/2} \geq 10^{325} c^{-36} (2^{-1}\epsilon^{-\chi})^{465} \max\{1, \tau^{36}\} \\
[n5] \quad & n \geq 4(K_0)^9 (\log \epsilon^{-1})^{K_0} \\
[n6] \quad & n \geq 2 \cdot 10^{606} c_3^{-84} \epsilon^{-222} (D(\log \epsilon^{-1})^{1/3} + 2)^{36} \\
[n7] \quad & n \geq 2a_0^{-9} (D(\log \epsilon^{-1})^{1/3} + 2)^9 2^6 \\
[n8] \quad & n \geq c^{-18} \max\{(\Psi_2 + 1)^9, 10^{23}\Psi_1^9\} \\
[n9] \quad & n \geq 3^9 c^{-18} \\
[n10] \quad & n \geq 10^{29} \epsilon^{-75/2} c^{-18} \\
[n11] \quad & n \geq 10^{24} c^{-18} \epsilon^{-75/2} (D(\log \epsilon^{-1})^{1/3} + 1)^{36} \\
[n12] \quad & D(\log \epsilon^{-1})^{1/3} - 1 \leq 6^{-1} cn^{1/9} \\
[n13] \quad & D(\log \epsilon^{-1})^{4/3} \leq 3\epsilon^{25/6} n^{1/36}
\end{aligned}$$

The $[\chi]$ list. Here appear remaining conditions that concern χ .

$$[\chi 1] \quad \chi \leq 1$$

The $[D]$ list. We also introduce a further condition:

$$[D1] \quad D \geq 1$$

Before trying to simplify these conditions, we state and prove a lemma.

Lemma B.1. *The conditions*

$$[\epsilon 23] \quad \epsilon \leq (10^{-41} \chi^{36} D^{-14})^{\chi^{-2}},$$

$D \geq 1$ and $\chi \leq 2^{1/2}$ imply that

$$22D(\log \epsilon^{-1})^{4/3} \leq \epsilon^{-\chi^2/6}.$$

Proof. The conclusion of the lemma holds precisely when

$$\epsilon^{-1} \leq \exp\{(22)^{-3/4} D^{-3/4} \epsilon^{-\chi^2/8}\}$$

and thus is implied by

$$\epsilon^{-1} \leq (\lceil 16\chi^{-2} \rceil!)^{-1} ((22)^{-3/4} D^{-3/4} \epsilon^{-\chi^2/8})^{\lceil 16\chi^{-2} \rceil}.$$

Our hypothesis that $\chi^2 \leq 2$ implies that $\lceil 16\chi^{-2} \rceil \leq 18\chi^{-2}$. Since $\epsilon \leq 1$, the last displayed condition is thus implied by

$$\epsilon \leq (\lceil 16\chi^{-2} \rceil!)^{-1} (22)^{-27\chi^{-2}/2} D^{-27\chi^{-2}/2}$$

since $D \geq 1$. Since $\lceil 16\chi^{-2} \rceil \leq 18\chi^{-2}$ and $\ell! \leq \ell^\ell$, the last is implied by

$$\epsilon \leq (18\chi^{-2})^{-18\chi^{-2}} (22)^{-27\chi^{-2}/2} D^{-27\chi^{-2}/2} = \left(\chi^{36} \cdot (18)^{-18} (22)^{-27/2} \cdot D^{27/2}\right)^{-\chi^{-2}}$$

which since $D \geq 1$ and $(18)^{18}22^{27/2} = 5.21 \dots \times 10^{40}$ is implied by our hypothesis that $\epsilon \leq (10^{-41}\chi^{36}D^{-14})^{\chi^{-2}}$. \square

Analysing the $[\epsilon]$ list. We now analyse the $[\epsilon]$ list, aiming to produce a simplified list of conditions that collectively imply all conditions in the $[\epsilon]$ list.

We begin by noting that Lemma B.1 provides an upper bound on the quantity $\tau > 0$ specified in $[\tau]$. Indeed, since $\tau = 4D(\log \epsilon^{-1})^{4/3} + 2\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil$, we see that, when $D \geq 1$ and $\epsilon \leq e^{-1}$, the quantity τ is at most $22D(\log \epsilon^{-1})^{4/3}$. When the hypotheses of the lemma hold, then, we see that

$$\tau \leq \epsilon^{-\chi^2/6}.$$

Introducing

$$[\epsilon 24] \epsilon \leq 2^{-4\chi^{-2}},$$

we next note that

$$[\epsilon 23, D1, \chi 1, \epsilon 24] \rightarrow [\epsilon 7].$$

Indeed, the conditions $[\epsilon 23]$, $[D1]$ and $[\chi 1]$ imply the hypotheses of Lemma B.1. $[\epsilon 7]$ is under these circumstances implied by $4^{-1}\epsilon^{-\chi} \geq \epsilon^{-\chi^2/2}$ or equivalently $\epsilon^{\chi^2/2-\chi} \geq 4$. Thus, it is implied by $[\chi 1, \epsilon 24]$. Thus, $[\epsilon 23, D1, \chi 1, \epsilon 24] \rightarrow [\epsilon 7]$, as we claimed.

Analysing $[\epsilon 20]$. We introduce some further conditions in order to replace $[\epsilon 20]$:

$$[\epsilon 25] \epsilon \leq \exp \left\{ -\beta\chi^{-1/2}(3480)^{1/2} \right\}$$

$$[\epsilon 26] \epsilon \leq e^{-e}$$

$$[\epsilon 27] \epsilon \leq (2\hat{H})^{-3480\chi^{-1}(\log \beta)^2}$$

$$[\epsilon 28] \chi \leq 2^{-1}(1 + 500(\log \beta)^2)^{-1}$$

and

$$[\epsilon 29] \epsilon \leq \exp \left\{ -2^{12} \right\}.$$

Lemma B.2. *We have that*

$$[D1, \chi 1, \epsilon 23, \epsilon 25, \epsilon 26, \epsilon 27, \epsilon 28, \epsilon 29] \rightarrow [\epsilon 20]$$

Proof. From $[D1, \epsilon 26]$, we have $D(\log \epsilon^{-1})^{4/3} \geq 1$ and thus $\lceil 8D(\log \epsilon^{-1})^{4/3} \rceil \leq 9D(\log \epsilon^{-1})^{4/3}$. We see then that, in these circumstances, $[\epsilon 20]$ is implied by

$$\begin{aligned} & \epsilon^{-3+\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2} \\ & \leq 2^{-1}\hat{H}^{-1}(18D(\log \epsilon^{-1})^{4/3})^{-(\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288-3/2} \exp \left\{ -2(\log \epsilon^{-1})^{11/12} \right\}, \end{aligned}$$

or equivalently

$$\begin{aligned} & -3 + \chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2 \\ & \geq \log(2\hat{H})(\log \epsilon^{-1})^{-1} \\ & \quad + ((\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288 + 3/2) \log \left(18D(\log \epsilon^{-1})^{4/3} \right) (\log \epsilon^{-1})^{-1} + 2(\log \epsilon^{-1})^{-1/12} \end{aligned}$$

This condition is satisfied provided that each of the following conditions is met:

$$\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2 \geq 6$$

$$\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2 \geq 6 \log(2\hat{H})(\log \epsilon^{-1})^{-1}$$

$\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2 \geq 6((\log \beta)^{-2}(\log \log \epsilon^{-1})^2/288+3/2) \log \left(18D(\log \epsilon^{-1})^{4/3}\right)(\log \epsilon^{-1})^{-1}$
and

$$\chi(580)^{-1}(\log \beta)^{-2}(\log \log \epsilon^{-1})^2 \geq 12(\log \epsilon^{-1})^{-1/12}$$

During the proof of Lemma B.2, we will call these last four bounds [1], [2], [3] and [4].

Recall that the quantity β is at least 4. Thus, $\beta \geq 1$, so that [1] is implied by

$$\log \log \epsilon^{-1} \geq \chi^{-1/2}(580)^{1/2}(\log \beta)6^{1/2}$$

and thus by [ε25].

$$\epsilon \leq \exp \left\{ -\beta^{\chi^{-1/2}(3480)^{1/2}} \right\}$$

[2] is implied by

$$\chi(580)^{-1}(\log \beta)^{-2} \geq 6 \log(2\hat{H})(\log \epsilon^{-1})^{-1}$$

and [ε26] in the form $(\log \log \epsilon^{-1})^2 \geq 1$. The last display is

$$\log \epsilon^{-1} \geq 3480\chi^{-1}(\log \beta)^2 \log(2\hat{H})$$

and thus is [ε27].

When [ε26], i.e. $(\log \log \epsilon^{-1})^2 \geq 1$, holds, [3] is implied by

$$\chi(580)^{-1}(\log \beta)^{-2} \geq 6((\log \beta)^{-2}/288 + 3/2) \log \left(18D(\log \epsilon^{-1})^{4/3}\right)(\log \epsilon^{-1})^{-1}$$

and thus by

$$\chi(580)^{-1} \geq 6(1/288 + 3(\log \beta)^2/2) \log \left(18D(\log \epsilon^{-1})^{4/3}\right)(\log \epsilon^{-1})^{-1}$$

By Lemma B.1, $22D(\log \epsilon^{-1})^{4/3} \leq \epsilon^{-\chi^2/6}$ is implied by [ε23, D1, χ1]. The last display is thus implied by

$$\chi(580)^{-1} \geq 6(1/288 + 3(\log \beta)^2/2) \log \left(\epsilon^{-\chi^2/6}\right)(\log \epsilon^{-1})^{-1}$$

which is

$$(580)^{-1} \geq \chi(1/288 + 3(\log \beta)^2/2)$$

which is implied by

$$288(580)^{-1} \geq \chi(1 + 500(\log \beta)^2)$$

which is implied by [ε28].

When [1] holds, [4] is implied by

$$6 \geq 12(\log \epsilon^{-1})^{-1/12}$$

which is [ε29].

Thus, the hypotheses of Lemma B.2 imply [1, 2, 3, 4]. This completes the proof of the lemma. \square

Analysing [ε21]. We now add a condition to the [D] list

$$[D2] \ D \geq c^{-1/3}2^5$$

and another to the [ε] list:

$$[\epsilon30] \ \epsilon \leq 2^{-1}C^{-1}$$

Lemma B.3. *We have that*

$$[D1, D2, \chi1, \epsilon23, \epsilon30] \rightarrow [\epsilon21].$$

Proof. By Lemma B.1, $22D(\log \epsilon^{-1})^{4/3} \leq \epsilon^{-\chi^2/6}$ is implied by $[\epsilon 23, D1, \chi 1]$. Since $[\epsilon 23]$ implies that $\epsilon \leq e^{-1}$, we see that, under these circumstances, the condition $[\epsilon 21]$ is implied by $[D2]$ in the guise $2^{-13}cD^3 - 2 \geq 2$ alongside

$$\epsilon^2 \leq 2^{-1}\epsilon^{\chi^2/6}C^{-1}.$$

Since $\chi \leq 1$ by $[\chi 1]$, the last display is implied by $[\epsilon 30]$. □

he lemma. □

Analysing $[\epsilon 21]$. We now add conditions

$$[D3] \quad D \geq 5^{4/3} \cdot 10^{44/3}c_1^{-4/3}$$

and

$$[\epsilon 31] \quad \epsilon \leq 8^{-1}c_1C^{-1}$$

Lemma B.4. *We have that*

$$[D1, D3, \chi 1, \epsilon 23, \epsilon 31] \rightarrow [\epsilon 22].$$

Proof. By Lemma B.1, $22D(\log \epsilon^{-1})^{4/3} \leq \epsilon^{-\chi^2/6}$ is implied by $[\epsilon 23, D1, \chi 1]$. The condition $[\epsilon 22]$ is thus under these circumstances implied by

$$\epsilon^{10^{-11}c_1D^{3/4}-3} \leq 8^{-1}c_1C^{-1}\epsilon^{\chi^2/6}$$

which in turn is implied by $[D3]$ in the guise

$$10^{-11}c_1D^{3/4} - 3 \geq 2$$

and $[\epsilon 31]$. □

Since $c_1 \leq 1$, $[\epsilon 31] \rightarrow [\epsilon 30]$. Note that $[\epsilon 29] \rightarrow [\epsilon 26]$.

Summary of the analysis of $[\epsilon 20, 21, 22]$. To obtain $[\epsilon 20, 21, 22]$ using the preceding three lemmas, we invoke the hypotheses $[D1, D2, D3, \chi 1, \epsilon 23, \epsilon 25, \epsilon 26, \epsilon 27, \epsilon 28, \epsilon 29, \epsilon 30, \epsilon 31]$. As we have just noted, however, $[\epsilon 26, 30]$ are redundant. Thus,

$$[D1, D2, D3, \chi 1, \epsilon 23, \epsilon 25, \epsilon 27, \epsilon 28, \epsilon 29, \epsilon 31] \rightarrow [\epsilon 20, 21, 22].$$

We now introduce some further conditions:

$$[\epsilon 32] \quad \epsilon \leq (4k_0)^{-\chi^{-1}}$$

$$[\epsilon 33] \quad \epsilon \leq \exp \{ -4^3c^{-9} \}$$

$$[\epsilon 34] \quad \epsilon \leq \exp \left\{ -D^{-3} \left(78\Psi_1 \vee 4((\Psi_2 + 1)^2 + \Psi_3)^{1/2} \right)^3 \right\}$$

$$[\epsilon 35] \quad D(\log \epsilon^{-1})^{4/3} \geq 10^9c_1^{-4/5}C^{1/2}$$

$$[\epsilon 36] \quad \epsilon \leq \exp \left\{ -D^{-3/4}10^{27/4}c_1^{-3/5}C^{3/8} \right\}$$

$$[\epsilon 37] \quad \epsilon \leq 10^{-1}D^{-4}$$

Lemma B.5. *We have the following inferences.*

$$(1) \quad [\epsilon 8] \rightarrow [\epsilon 1]$$

$$(2) \quad [\epsilon 29] \rightarrow [\epsilon 2, 3]$$

$$(3) \quad [\epsilon 5, 28] \rightarrow [\epsilon 4]$$

- (4) $[\epsilon 6] = [\epsilon 32]$
- (5) $[\epsilon 29, D1] \rightarrow [\epsilon 8]$
- (6) $[\epsilon 13, \epsilon 23, D1, \chi 1, \epsilon 37] \rightarrow [\epsilon 9]$
- (7) $[\epsilon 29, D1] \rightarrow [\epsilon 13]$
- (8) $[\epsilon 33, D1] \rightarrow [\epsilon 14]$
- (9) $[\epsilon 34, D1, \epsilon 29] \rightarrow [\epsilon 15]$
- (10) $[\epsilon 11] \rightarrow [\epsilon 16]$
- (11) $[\epsilon 35] \rightarrow [\epsilon 17, 18]$
- (12) $[\epsilon 36] = [\epsilon 35]$
- (13) $[\epsilon 29] \rightarrow [\epsilon 19]$

Proof: (1). Due to $[\epsilon 1]$ being equivalent to $\epsilon \leq \exp \{ -D^{-3} \}$.

(2). Trivial.

(3). $[\epsilon 28]$ implies that $\chi \leq 1/2$. We also use $\beta \geq e$, which follows from the definition of β .

(4,5). Trivial.

(6). $[\epsilon 9]$ is implied by $[\epsilon 13]$ and the bound

$$\epsilon \leq 4^{-2} \left(4/3 \cdot D (\log \epsilon^{-1})^{1/3} \right)^{-4}.$$

This display is equivalent to

$$\epsilon \leq 4^{-6} 3^4 D^{-4} (\log \epsilon^{-1})^{4/3}.$$

By Lemma B.1, $[\epsilon 23, D1, \chi 1]$ imply that the last display is implied by

$$\epsilon \leq 4^{-6} 3^4 D^{-4} \cdot 22 D \epsilon^{\chi^2/6}.$$

Invoking $[\chi 1]$ (which is $\chi \leq 1$), we see that the last is implied by

$$\epsilon \leq (22 \cdot 4^{-6} 3^4)^{6/5} D^{-18/5}.$$

Since $(22 \cdot 4^{-6} 3^4)^{6/5} = 0.368 \dots$, we see that, using $D \geq 1$ (i.e., $[D1]$), the last condition is implied by $[\epsilon 37]$. This completes the derivation of **(6)**.

(7). $[\epsilon 13]$ is equivalent to $\epsilon \leq \exp \{ -6^3 D^{-3} \}$ which is implied by $[\epsilon 29, D1]$.

(8). $[\epsilon 14]$ is equivalent to $\epsilon \leq \exp \{ -(3c^{-3} + 1)^3 D^{-3} \}$. Using $D \geq 1$ (i.e., $[D1]$) and $c \leq 1$, this is implied by $[\epsilon 33]$.

(9). Note first that the bound $1 \leq 2^{-1} D (\log \epsilon^{-1})^{1/3}$ is implied by $[D1, \epsilon 29]$.

Thus, when $[D1, \epsilon 29]$ holds, $[\epsilon 15]$ is implied by

$$D (\log \epsilon^{-1})^{1/3} \geq 78 \Psi_1 \vee 4 ((\Psi_2 + 1)^2 + \Psi_3)^{1/2}.$$

This display is equivalent to $[\epsilon 34]$.

(10). $[\epsilon 11] \rightarrow [\epsilon 16]$ since $c_3 \leq c_1$.

(11). $[\epsilon 35] \rightarrow [\epsilon 17]$ due to $C \geq 1$. $[\epsilon 35] \rightarrow [\epsilon 18]$ due to $c_1 \leq 1$.

(12). Trivial.

(13). $[\epsilon 19]$ is implied by $\log \epsilon^{-1} \geq 11\epsilon^{1/2}$. Since $\epsilon \leq 1$ by $[\epsilon 29]$, the last inequality is implied by $\epsilon \leq e^{-11}$, which itself is implied by $[\epsilon 29]$. \square

A summary of progress so far. We are working to simplify the $[\epsilon]$ -list, which consists of the conditions $[\epsilon 1, \epsilon 2, \dots, \epsilon 20]$. Drawing together our inferences made thus far, we see that this collection of conditions is implied by

$$[\epsilon 5, \epsilon 10, \epsilon 11, \epsilon 12, \epsilon 23, \epsilon 24, \epsilon 25, \epsilon 27, \epsilon 28, \epsilon 29, \epsilon 31, \epsilon 32, \epsilon 33, \epsilon 34, \epsilon 36, \epsilon 37, D1, D2, D3, \chi 1].$$

A little further simplification will now be made of these conditions, and so it is convenient to restate them all now.

$$\begin{aligned} [\epsilon 5] \quad \epsilon &\leq \exp \{ -\chi^{-2} \} \\ [\epsilon 10] \quad \epsilon &\leq (\eta_0)^{72} \\ [\epsilon 11] \quad \epsilon &\leq 10^{-1342} c_3^{44} \\ [\epsilon 12] \quad \epsilon &\leq \exp \{ -2C^{3/8} \} \\ [\epsilon 23] \quad \epsilon &\leq (10^{-41} \chi^{36} D^{-14}) \chi^{-2}, \\ [\epsilon 24] \quad \epsilon &\leq 2^{-4\chi^{-2}}, \end{aligned}$$

$$\begin{aligned} [\epsilon 25] \quad \epsilon &\leq \exp \{ -\beta \chi^{-1/2} (3480)^{1/2} \} \\ [\epsilon 27] \quad \epsilon &\leq (2\hat{H})^{-3480\chi^{-1}(\log \beta)^2} \\ [\epsilon 28] \quad \chi &\leq 2^{-1} (1 + 500(\log \beta)^2)^{-1} \end{aligned}$$

and

$$\begin{aligned} [\epsilon 29] \quad \epsilon &\leq \exp \{ -2^{12} \}. \\ [\epsilon 31] \quad \epsilon &\leq 8^{-1} c_1 C^{-1} \\ [\epsilon 32] \quad \epsilon &\leq (4k_0)^{-\chi^{-1}} \\ [\epsilon 33] \quad \epsilon &\leq \exp \{ -4^3 c^{-9} \} \\ [\epsilon 34] \quad \epsilon &\leq \exp \left\{ -D^{-3} \left(78\Psi_1 \vee 4((\Psi_2 + 1)^2 + \Psi_3)^{1/2} \right)^3 \right\} \\ [\epsilon 36] \quad \epsilon &\leq \exp \left\{ -D^{-3/4} 10^{27/4} c_1^{-3/5} C^{3/8} \right\} \\ [\epsilon 37] \quad \epsilon &\leq 10^{-1} D^{-4} \\ [D1] \quad D &\geq 1 \\ [D2] \quad D &\geq c^{-1/3} 2^5 \\ [D3] \quad D &\geq 5^{4/3} \cdot 10^{44/3} c_1^{-4/3} \\ [\chi 1] \quad \chi &\leq 1 \end{aligned}$$

Some further inferences.

We now introduce a further condition:

$$[\epsilon 38] \quad \epsilon \leq \exp \left\{ -10^7 c_1^{-9} C^{3/8} \right\}$$

Note that $[\epsilon 38] \rightarrow [\epsilon 12]$ since $c_1 \leq 1$. Since $c_1 \leq 1$, it is easily checked that $[\epsilon 38] \rightarrow [\epsilon 31]$. Since $c_1 \leq c$ and $C \geq 1$, $[\epsilon 38] \rightarrow [\epsilon 33]$. Note also that $[\epsilon 38, D1] \rightarrow [\epsilon 36]$ since $c_1 \leq 1$.

Note that $[\epsilon 23, D1, \chi 1] \rightarrow [\epsilon 5, \epsilon 24, \epsilon 37]$.

Note also that $[\epsilon 28] \rightarrow [\chi 1]$.

Introducing

$$[D4] \quad D \geq 10^{16} c_1^{-4/3}$$

we see that, since $5^{4/3} 10^{44/3} = 3.96 \dots \times 10^{15}$, $[D4] \rightarrow [D3]$. We also have $[D4] \rightarrow [D1, 2]$ since $c_1 \leq 1$ and $c_1 \leq c$.

Introducing

$$[\epsilon 39] \quad e \leq 10^{-1779} c_3^{44},$$

we see, since $\exp \{2^{12}\} = [10^{1778}, 10^{1779}]$ and $c_3 \leq 1$, $[\epsilon 39] \rightarrow [\epsilon 29]$. We also have $[\epsilon 39] \rightarrow [\epsilon 11]$.

Concluding the analysis of the $[\epsilon]$ list.

In summary, then, the original $[\epsilon]$ -list, consisting of $[\epsilon 1, \epsilon 2, \dots, \epsilon 20]$, is implied by

$$[\epsilon 10, \epsilon 23, \epsilon 25, \epsilon 27, \epsilon 28, \epsilon 32, \epsilon 34, \epsilon 38, \epsilon 39, D4]. \quad (29)$$

These conditions collectively may be expressed as:

$$\epsilon \leq \max \left\{ (\eta_0)^{72}, (10^{-41} \chi^{36} D^{-14})^{\chi^{-2}}, \exp \left\{ -\beta \chi^{-1/2} (3480)^{1/2} \right\}, (2\hat{H})^{-3480 \chi^{-1} (\log \beta)^2}, (4k_0)^{-\chi^{-1}}, \right. \\ \left. \exp \left\{ -D^{-3} \left(78 \Psi_1 \vee 4((\Psi_2 + 1)^2 + \Psi_3)^{1/2} \right)^3 \right\}, \exp \left\{ -10^7 c_1^{-9} C^{3/8} \right\}, 10^{-1779} c_3^{44} \right\},$$

as well as

$$\chi \leq 2^{-1} (1 + 500 (\log \beta)^2)^{-1}$$

and

$$D \geq 10^{16} c_1^{-4/3}.$$

Analysing the $[n]$ list.

We begin this analysis by presenting some further conditions. Since the conditions concern ϵ , we label them as such.

$$[\epsilon 40] \quad \epsilon \leq \exp \left\{ -(12)^{-4} (\log K_0)^4 \right\}$$

$$[\epsilon 41] \quad \epsilon \leq \exp \{-8!\}$$

$$[\epsilon 42] \quad \epsilon \leq \exp \left\{ -2K_0^2 \right\}$$

$$[\epsilon 43] \quad K_0^{(12)^{-2} (\log \log (2^{-1} \epsilon^{-\chi}))^2} \leq \epsilon^{-1}$$

$$[\epsilon 44] \quad (\log (2^{-1} \epsilon^{-\chi}))^{K_0} \leq \epsilon^{-1}$$

$$[\epsilon 45] \quad (\log \epsilon^{-1})^{K_0} \leq \epsilon^{-1}$$

The next lemma will be used to simplify the $[n]$ list.

Lemma B.6. (1) $[\epsilon 40, \epsilon 41, \chi 1] \rightarrow [\epsilon 43]$

(2) $[\epsilon 42] \rightarrow [\epsilon 45]$

(3) $[\epsilon 45, \chi 1] \rightarrow [\epsilon 44]$

Proof: (1). Note that $[\epsilon 43]$ is equivalent to

$$(\log \log(2^{-1}\epsilon^{-\chi}))^2 (12)^{-2} \log K_0 \leq \log \epsilon^{-1}$$

and thus is implied by

$$(12)^{-2} \log K_0 \leq (\log \epsilon^{-1})^{1/2}$$

and

$$(\log \log(2^{-1}\epsilon^{-\chi}))^2 \leq (\log \epsilon^{-1})^{1/2}.$$

These two displayed equations will here be called [1] and [2].

Note that [1] is equivalent to $[\epsilon 40]$.

Set $h = \log \epsilon^{-1}$. Note that [2] is equivalent to

$$(\log(\chi h - \log 2))^4 \leq h$$

When $[\chi 1]$ holds, the last is implied by $(\log h)^4 \leq h$ or $h \leq \exp\{h^{1/4}\}$ which is implied by $h \geq 8!$ or equivalently $[\epsilon 41]$.

(2). Again we set $h = \log \epsilon^{-1}$. Note that $[\epsilon 45]$ is equivalent to $h \leq \exp\{hK_0^{-1}\}$ which is implied by $h \leq h^2K_0^{-2}/2$ or $h \geq 2K_0^2$ which is $[\epsilon 42]$.

(3). Trivial. □

We now introduce some further conditions expressed in terms of n .

$$[n14] \ n \geq 2\epsilon^{-7/2}$$

$$[n15] \ n \geq a_0^{-9}\epsilon^{-3}$$

$$[n16] \ n \geq 10^{186}c^{-36}\epsilon^{-3/2-466\chi}$$

$$[n17] \ n \geq 10^{618}D^{36}c_3^{-84}\epsilon^{-234}$$

$$[n18] \ n \geq 2^{16}a_0^{-9}D^9\epsilon^{-3}$$

$$[n19] \ n \geq 10^{186}c^{-36}\epsilon^{-75/2-466\chi}$$

$$[n20] \ n \geq 10^{35}c^{-18}D^{36}\epsilon^{-50}$$

$$[n21] \ n \geq 10^{35}c^{-18}D^{36}\epsilon^{-198}$$

$$[n22] \ n \geq 10^{618}D^{36}c_3^{-84}a_0^{-9}\epsilon^{-504}$$

Lemma B.7. (1) $[n16] \rightarrow [n1]$

(2) $[n14, \epsilon 43, \epsilon 44] \rightarrow [n2]$

(3) $[n15, \epsilon 23, D1, \chi 1] \rightarrow [n3]$

(4) $[n16, \epsilon 23, D1, \chi 1] \rightarrow [n4]$

(5) $[n17, \epsilon 8] \rightarrow [n6]$

(6) $[n18, \epsilon 8] \rightarrow [n7]$

(7) $[n16] \rightarrow [n9]$

- (8) $[n19] \rightarrow [n10, 16]$
- (9) $[n20, \epsilon 8] \rightarrow [n11]$
- (10) $[n20, D1] \rightarrow [n12]$
- (11) $[n21] \rightarrow [n13, n20]$
- (12) $[n22] \rightarrow [n14, 15, 16, 17, 18, 19, 20, 21]$

Proof: (1). $[n1]$ is the condition that $n \geq 8$ which is implied by $[n16]$ since $c \leq 1$ and $\epsilon \leq 1$.

(2). When $[e43, 44]$ hold, $[n2]$ is implied by $n\epsilon^{3/2} \geq 2\epsilon^{-2}$ and thus by $[n14]$.

(3). By Lemma B.1 and the comments that follow it, $[\epsilon 23, D1, \chi 1]$ imply that $\tau \leq \epsilon^{-\chi^2/6}$. Under these circumstances, $[n3]$ is implied by $n\epsilon^{3/2} \geq a_0^{-9}\epsilon^{-3\chi^2/2}$. Since $\chi \leq 1$, it is also implied by $n \geq a_0^9\epsilon^{-3}$.

(4). For the same reason, $[\epsilon 23, D1, \chi 1]$ imply that $[n4]$ is implied by

$$n\epsilon^{3/2} \geq 10^{325}c^{-36}(2^{-1}\epsilon^{-\chi})^{465} \max\{1, \epsilon^{-\chi^{12}}\}$$

Since $\epsilon \leq 1$ and $\chi \leq 1$, this condition is implied by

$$n\epsilon^{3/2} \geq 10^{325}c^{-36}2^{-465}\epsilon^{-466\chi}$$

which since $2^{465} = 9.52 \dots \times 10^{139}$ is implied by $[n16]$.

(5). Note that $2 \leq D(\log \epsilon^{-1})^{1/3}$ holds when $\epsilon \leq \exp\{-8D^{-3}\}$ occurs and that the latter bound in $[\epsilon 8]$. Thus, when $[\epsilon 8]$ occurs, $[n6]$ is implied by

$$n \geq 2 \cdot 10^{606}c_3^{-84}\epsilon^{-222}(2D(\log \epsilon^{-1})^{1/3})^{36}$$

Since $\log \epsilon^{-1} \leq \epsilon^{-1}$, the last bound is implied by

$$n \geq 2^{37}10^{606}D^{36}c_3^{-84}\epsilon^{-234}$$

Since $2^{37} = 1.37 \times 10^{11}$, the last is implied by $[n17]$.

(6). For the same reason, $[n7]$ is when $[\epsilon 8]$ occurs implied by

$$n \geq 2a_0^{-9}(2D(\log \epsilon^{-1})^{1/3})^9 2^6.$$

This is implied by $[n18]$.

(7,8). This is due to $\epsilon \leq 1$.

(9). When $[\epsilon 8]$ holds, $[n11]$ is implied by

$$n \geq 10^{24}c^{-18}\epsilon^{-75/2}(2D(\log \epsilon^{-1})^{1/3})^{36}$$

and thus by

$$n \geq 10^{24}2^{36}c^{-18}\epsilon^{-75/2-12}D^{36}.$$

Since $2^{36} = 6.87 \dots \times 10^{10}$ and $\epsilon \leq 1$, this is implied by

$$n \geq 10^{24}2^{36}c^{-18}\epsilon^{-50}D^{36}$$

and thus by $[n20]$.

(10). $[n12]$ is implied by

$$D(\log \epsilon^{-1})^{1/3} \leq 6^{-1}cn^{1/9}$$

or equivalently

$$n \geq 6^9 c^{-9} D^9 (\log \epsilon^{-1})^3$$

When $[D1]$ holds, this is implied by $[n20]$ since $\epsilon \leq 1$.

(11). $[n13]$ is implied by

$$D^{36} (\log \epsilon^{-1})^{48} 3^{-36} \epsilon^{-150} \leq n$$

and thus by

$$n \geq D^{36} 3^{-36} \epsilon^{-198}.$$

This last is implied by $[n21]$ since $c \leq 1$. Note that $[n21] \rightarrow [n20]$ since $\epsilon \leq 1$.

(12). This is due to $D \geq 1$, $c_3 \leq 1$, $a_0 \leq 1$, $\epsilon \leq 1$, $c_3 \leq c$ and $\chi \leq 1$. (Recall that the condition $a_0 \leq 1$ was imposed when this parameter was introduced in Theorem 4.1.) \square

Simplifying the $[n]$ list. We learn from Lemma B.7 that the original $[n]$ list, consisting of conditions $[n1, 2, \dots, 13]$ is implied by

$$[n5, n8, n22, \epsilon8, \epsilon23, \epsilon43, \epsilon44, D1, \chi1].$$

Recall that our simplified $[\epsilon]$ list (29) consists of $[\epsilon10, \epsilon23, \epsilon25, \epsilon27, \epsilon28, \epsilon32, \epsilon34, \epsilon38, \epsilon39, D4]$. Since we are assuming these conditions, we may remove from the displayed list those conditions that are implied by them: these include $[D1]$, $[\chi1]$, $[\epsilon8]$ and $[\epsilon23]$. (Note that $[\epsilon8]$ is implied because it is on the original $[\epsilon]$ list, $[\epsilon1, 2, \dots, 22]$, and the simplified $[\epsilon]$ list.) Moreover, by Lemma B.6, we may replace $[\epsilon43, 44]$ by $[\epsilon40, 41, 42]$. Thus, our original $[n]$ list is, when the simplified $[\epsilon]$ list is assumed, implied by

$$[n5, n8, n22, \epsilon40, \epsilon41, \epsilon42].$$

We may drop $[\epsilon41]$ because it is implied by $[\epsilon11]$ since $c_3 \leq 1$.

Finally, we will drop $[\epsilon40]$ by arguing that it is implied by $[\epsilon42]$. To argue this, recall that the parameter K_0 , which originates in Theorem 4.1, is supposed to satisfy $K_0 \geq 1$. If $[\epsilon42]$ fails to imply $[\epsilon40]$, then $2K_0^2 < (12)^{-4} (\log K_0)^4$ or equivalently $\exp \{2^{1/4} 12 K_0^{1/2}\} \leq K_0$. This last is not valid when $(4!)^{-1} (2^{1/4} 12)^4 K_0 \geq 1$. Since $K_0 \geq 1$, this condition is not met, so $[\epsilon42] \rightarrow [\epsilon40]$.

Thus, we choose our simplified $[n]$ list to be

$$[n5, n8, n22, \epsilon42].$$

Conclusion. The original collection of conditions, on the $[\epsilon]$, $[n]$, $[D]$ and $[\chi]$ lists, are implied by the simplified $[\epsilon]$ and $[n]$ lists, which consist of the conditions:

$$[\epsilon10, \epsilon23, \epsilon25, \epsilon27, \epsilon28, \epsilon32, \epsilon34, \epsilon38, \epsilon39, D4, n5, n8, n22, \epsilon42].$$

These conditions collectively may be expressed as:

$$\begin{aligned} \epsilon \leq & \max \left\{ (\eta_0)^{72}, (10^{-41} \chi^{36} D^{-14}) \chi^{-2}, \exp \left\{ -\beta \chi^{-1/2} (3480)^{1/2} \right\}, \right. \\ & (2\hat{H})^{-3480 \chi^{-1} (\log \beta)^2}, (4k_0)^{-\chi^{-1}}, \exp \left\{ -2K_0^2 \right\}, 10^{-1779} c_3^{44}, \\ & \left. \exp \left\{ -D^{-3} \left(78\Psi_1 \vee 4((\Psi_2 + 1)^2 + \Psi_3)^{1/2} \right)^3 \right\}, \exp \left\{ -10^7 c_1^{-9} C^{3/8} \right\} \right\}, \\ n \geq & \max \left\{ 4(K_0)^9 (\log \epsilon^{-1})^{K_0}, c^{-18} \max \left\{ (\Psi_2 + 1)^9, 10^{23} \Psi_1^9 \right\}, 10^{618} D^{36} c_3^{-84} a_0^{-9} \epsilon^{-504} \right\}, \end{aligned}$$

as well as

$$\chi \leq 2^{-1}(1 + 500(\log \beta)^2)^{-1}$$

and

$$D \geq 10^{16} c_1^{-4/3}.$$

Since these are the hypotheses of Lemma 8.1, we have completed the calculational derivation of this result.

REFERENCES

- [BDJ99] Jinho Baik, Percy Deift, and Kurt Johansson. On the distribution of the length of the longest increasing subsequence of random permutations. *J. Amer. Math. Soc.*, 12(4):1119–1178, 1999.
- [BFPS07] Alexei Borodin, Patrik L. Ferrari, Michael Prähofer, and Tomohiro Sasamoto. Fluctuation properties of the TASEP with periodic initial configuration. *J. Stat. Phys.*, 129(5-6):1055–1080, 2007.
- [BSS16] Riddhipratim Basu, Vidas Sidoravicius, and Allan Sly. Last passage percolation with a defect line and the solution of the slow bond problem. *arXiv:1408.3464*, 2016.
- [BSS17] Riddhipratim Basu, Sourav Sarkar, and Allan Sly. Invariant measures for TASEP with a slow bond. *arXiv:1704.07799*, 2017.
- [CH14] Ivan Corwin and Alan Hammond. Brownian Gibbs property for Airy line ensembles. *Invent. Math.*, 195(2):441–508, 2014.
- [Cor12] Ivan Corwin. The Kardar-Parisi-Zhang equation and universality class. *Random Matrices Theory Appl.*, 1(1):1130001, 76, 2012.
- [CP15] Eric Cator and Leandro P. R. Pimentel. On the local fluctuations of last-passage percolation models. *Stochastic Process. Appl.*, 125(2):538–551, 2015.
- [Ede61] Murray Eden. A two-dimensional growth process. In *Proc. 4th Berkeley Sympos. Math. Statist. and Prob., Vol. IV*, pages 223–239. Univ. California Press, Berkeley, Calif., 1961.
- [Häg08] Jonas Hägg. Local Gaussian fluctuations in the Airy and discrete PNG processes. *Ann. Probab.*, 36(3):1059–1092, 2008.
- [Ham17a] Alan Hammond. Brownian regularity for the Airy line ensemble, and multi-polymer watermelons in Brownian last passage percolation. 2017. math.berkeley.edu/~alanmh/papers/BrownianReg.pdf.
- [Ham17b] Alan Hammond. Modulus of continuity of polymer weight profiles in Brownian last passage percolation. 2017. math.berkeley.edu/~alanmh/papers/ModCon.pdf.
- [Ham17c] Alan Hammond. On the rarity of several disjoint polymers in Brownian last passage percolation. 2017. math.berkeley.edu/~alanmh/papers/NonIntPolymer.pdf.
- [Joh00] Kurt Johansson. Transversal fluctuations for increasing subsequences on the plane. *Probab. Theory Related Fields*, 116(4):445–456, 2000.
- [Joh03] Kurt Johansson. Discrete polynuclear growth and determinantal processes. *Comm. Math. Phys.*, 242(1-2):277–329, 2003.
- [MFQR13] Gregorio Moreno Flores, Jeremy Quastel, and Daniel Remenik. Endpoint distribution of directed polymers in $1 + 1$ dimensions. *Comm. Math. Phys.*, 317(2):363–380, 2013.
- [MQR17] Konstantin Matetski, Jeremy Quastel, and Daniel Remenik. The KPZ fixed point. *arXiv:1701.00018*, 2017.
- [OY02] Neil O’Connell and Marc Yor. A representation for non-colliding random walks. *Electron. Comm. Probab.*, 7:1–12 (electronic), 2002.
- [Pim14] Leandro P. R. Pimentel. On the location of the maximum of a continuous stochastic process. *J. Appl. Probab.*, 51(1):152–161, 2014.
- [Pim17] Leandro Pimentel. Local behavior of Airy processes. *arXiv:1704.01903*, 2017.
- [PS02] Michael Prähofer and Herbert Spohn. Scale invariance of the PNG droplet and the Airy process. *J. Statist. Phys.*, 108(5-6):1071–1106, 2002. Dedicated to David Ruelle and Yasha Sinai on the occasion of their 65th birthdays.
- [Sas05] T. Sasamoto. Spatial correlations of the 1D KPZ surface on a flat substrate. *J. Phys. A*, 38(33):L549–L556, 2005.

A. HAMMOND, DEPARTMENT OF MATHEMATICS AND STATISTICS, U.C. BERKELEY, EVANS HALL, BERKELEY, CA, 94720-3840, U.S.A.

E-mail address: alanmh@berkeley.edu