

Active spanning trees with bending energy on planar maps and SLE-decorated Liouville quantum gravity for $\kappa > 8$

Ewain Gwynne
MIT

Adrien Kassel
ENS Lyon

Jason Miller
Cambridge

David B. Wilson
Microsoft Research

Abstract

We introduce a two-parameter family of probability measures on spanning trees of a planar map. One of the parameters controls the activity of the spanning tree and the other is a measure of its bending energy. When the bending parameter is 1, we recover the active spanning tree model, which is closely related to the critical Fortuin–Kasteleyn model. A random planar map decorated by a spanning tree sampled from our model can be encoded by means of a generalized version of Sheffield’s hamburger-cheeseburger bijection. Using this encoding, we prove that for a range of parameter values (including the ones corresponding to maps decorated by an active spanning tree), the infinite-volume limit of spanning-tree-decorated planar maps sampled from our model converges in the peanosphere sense, upon rescaling, to an SLE_κ -decorated γ -Liouville quantum cone with $\kappa > 8$ and $\gamma = 4/\sqrt{\kappa} \in (0, \sqrt{2})$.

Contents

1	Introduction	2
1.1	Overview	2
1.2	Basic notation	7
1.3	Generalized burger model	7
1.4	Active spanning trees with bending energy	11
1.5	Statement of main results	18
1.6	Outline	19
2	Scaling limit when all orders are stale	22
3	Scaling limit with stale orders and duplicate burgers	25
3.1	Comparison of expected lengths of reduced words	25
3.2	Bound on the number of unidentified symbols	30
3.3	Renewal times in the word	33
3.4	Variance of the discrepancy between burger types	40
3.5	Expected length of the reduced word	41
3.6	Tail bound for the length of the reduced word	43
3.7	Convergence to correlated Brownian motion	46
4	Open problems	48
A	Basic properties of the burger model	49

1 Introduction

1.1 Overview

We study a family of probability measures on spanning-tree-decorated rooted planar maps, which we define in Section 1.3, using a generalization of the Sheffield hamburger-cheeseburger model [She16]. This family includes as special cases maps decorated by a uniform spanning tree [Mul67], planar maps together with a critical Fortuin–Kasteleyn (FK) configuration [She16], and maps decorated by an active spanning tree [KW16]. These models converge in a certain sense (described below) to Liouville quantum gravity (LQG) surfaces decorated by Schramm–Loewner evolution (SLE_κ) [Sch00], and any value of $\kappa > 4$ corresponds to some measure in the family. Although our results are motivated by SLE and LQG, our proofs are entirely self-contained, requiring no knowledge beyond elementary probability theory.

Consider a spanning-tree-decorated rooted planar map (M, e_0, T) , where M is a planar map, e_0 is an oriented root edge for M , and T is a spanning tree of M . Let M^* be the dual map of M and let T^* be the dual spanning tree, which consists of the edges of M^* which do not cross edges of T . Let Q be the quadrangulation whose vertex set is the union of the vertex sets of M and M^* , obtained by identifying each vertex of M^* with a point in the corresponding face of M , then connecting it by an edge (in Q) to each vertex of M on the boundary of that face. Each face of Q is bisected by either an edge of T or an edge of T^* (but not both). Let e_0 be the oriented edge of Q with the same initial endpoint as e_0 and which is the first edge in the clockwise direction from e_0 among all such edges. As explained in, e.g., [She16, § 4.1], there is a path λ consisting of edges of (the dual of) Q which snakes between the primal tree T and dual tree T^* , starts and ends at e_0 , and hits each edge of Q exactly once. This path λ is called the *Peano curve* of (M, e_0, T) . See Figure 1 for an illustration.

For Euclidean lattices, Lawler, Schramm, and Werner [LSW04] showed that the uniform spanning tree Peano curve converges to SLE_8 . For random tree-decorated planar maps with suitable weights coming from the critical Fortuin–Kasteleyn model, Sheffield [She16] proved a convergence result which, when combined with the continuum results of [DMS14], implies that the Peano curve converges in a certain sense to a space-filling version of SLE_κ with $4 < \kappa \leq 8$ on an LQG surface. The measures on tree-decorated planar maps we consider generalize these, and converge in this same sense to SLE_κ with $4 < \kappa < \infty$.

For the measures on tree-decorated planar maps which we consider in this paper, the conjectured scaling limit of the Peano curve λ is a *whole-plane space-filling* SLE_κ from ∞ to ∞ for an appropriate value of $\kappa > 4$. In the case when $\kappa \geq 8$, SLE_κ is space-filling [RS05], and whole-plane space-filling SLE_κ from ∞ to ∞ is just a whole-plane variant of chordal SLE_κ (see [DMS14, footnote 9]). It is characterized by the property that for any stopping time τ for the curve, the conditional law of the part of the curve which has not yet been traced is that of a chordal SLE_κ from the tip of the curve to ∞ . Ordinary SLE_κ for $\kappa \in (4, 8)$ is not space-filling [RS05]. In this case, whole-plane space-filling SLE_κ from ∞ to ∞ is obtained from a whole-plane variant of ordinary chordal SLE_κ by iteratively filling in the “bubbles” disconnected from ∞ by the curve. The construction of space-filling SLE_κ in this case is explained in [MS17, § 1.2.3 and 4.3]. For $\kappa > 4$, whole-plane space-filling SLE_κ is the Peano curve of a certain tree of $SLE_{16/\kappa}$ -type curves, namely the set of all flow lines (in the sense

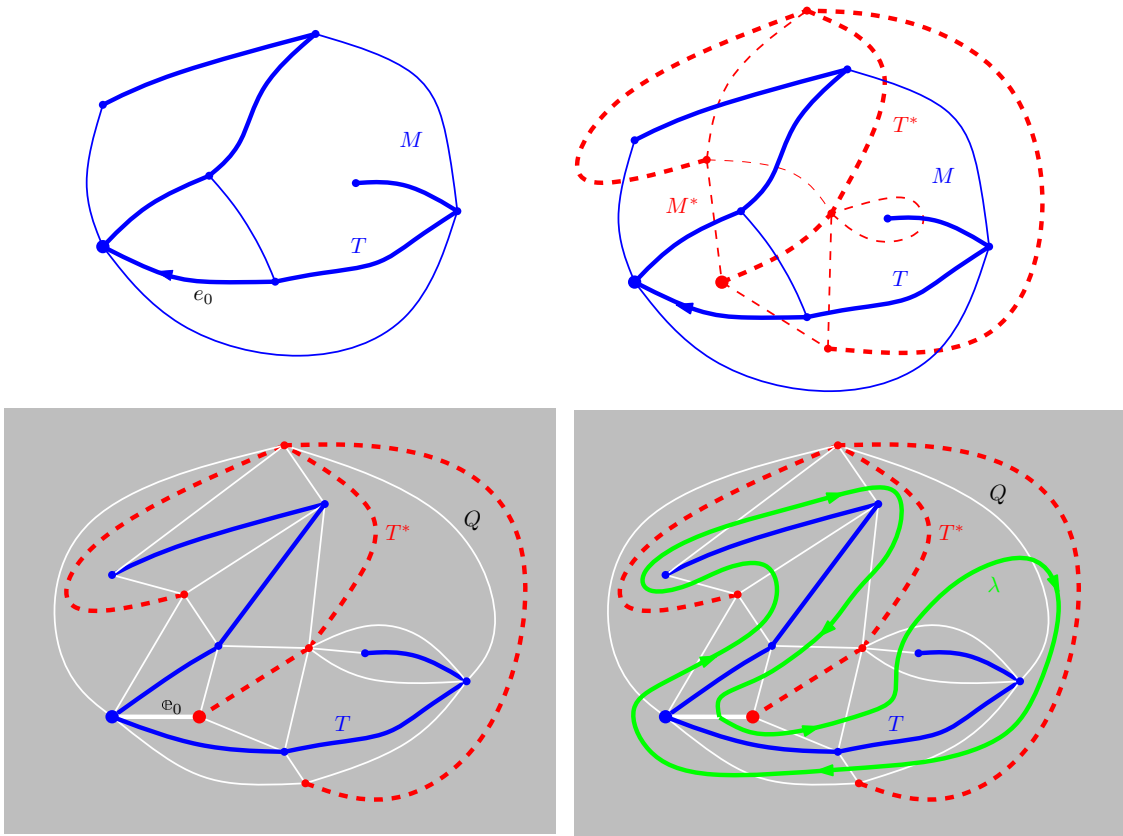


Figure 1: Top left: a rooted map (M, e_0) (in blue) with a spanning tree T (heavier blue lines). Top right: the dual map M^* (dashed red) with the dual spanning tree T^* (heavier dashed red lines). Bottom left: the quadrangulation Q (in white) whose vertices are the vertices of M and M^* . Bottom right: the Peano curve λ (in green), exploring clockwise. Formally, λ is a cyclic ordering of the edges of Q with the property that successive edges share an endpoint. The triple (M, e_0, T) can be encoded by means of a two-dimensional simple walk excursion in the first quadrant with $2n$ steps, equivalently a word consisting of elements of the set Θ_0 defined below which reduces to the empty word; see Figure 3.

of [MS16d, MS16e, MS16a, MS17]) of a whole-plane Gaussian free field (GFF) started from different points but with a common angle.

There are various ways to formulate the convergence of spanning-tree-decorated planar maps toward space-filling SLE_κ -decorated LQG surfaces. One can embed the map M into \mathbb{C} (e.g. via circle packing or Riemann uniformization) and conjecture that the Peano curve of T (resp. the measure which assigns mass $1/n$ to each vertex of M) converges in the Skorokhod metric (resp. the weak topology) to the space-filling SLE_κ (resp. the volume measure associated with the γ -LQG surface). Alternatively, one can first try to define a metric on an LQG surface (which has so far been accomplished only in the case when $\gamma = \sqrt{8/3}$ [MS16f, MS15c, MS15a, MS15b, MS16b, MS16c], in which case it is isometric to some variant of the Brownian map [Le 14, Mie09]), and then try to show that the graph metric on M (suitably rescaled) converges in the Hausdorff metric to an LQG surface. Convergence in the former sense has only recently been shown for “mated-CRT maps” using the Tutte (harmonic or barycentric) embedding [GMS17b]. It has not yet been proved for any other

random planar map model, and convergence in the latter (metric) sense has been established only for uniform planar maps and slight variants thereof (which correspond to $\gamma = \sqrt{8/3}$) [Le13, Mie13].

Here we consider a different notion of convergence, called convergence in the *peanosphere sense*, which we now describe (see Figure 2). This notion of convergence is based on the work [DMS14], which shows how to encode a γ -quantum cone (a certain type of LQG surface parametrized by \mathbb{C} , obtained by zooming in near a point sampled from the γ -LQG measure induced by a GFF [DMS14, § 4.3]) decorated by an independent whole-plane space-filling SLE_κ curve η with $\kappa = 16/\gamma^2$ in terms of a correlated two-dimensional Brownian motion Z , with correlation $-\cos(4\pi/\kappa)$. Recall that the contour function of a discrete, rooted plane tree is the function one obtains by tracing along the boundary of the tree starting at the root and proceeding in a clockwise manner and recording the distance to the root from the present vertex. The two coordinates of the Brownian motion Z are the contour functions of the $\text{SLE}_{16/\kappa}$ tree (whose Peano curve is η) and that of the corresponding dual tree (consisting of GFF flow lines whose angles differ from the angles of the flow lines in the original tree by π). Here, the distance to the root is measured using γ -LQG length. On the discrete side, the entire random planar map is determined by the pair of trees. One non-obvious fact established in [DMS14] is that the corresponding statement is true in the continuum: the entire γ -quantum cone and space-filling SLE turn out to be almost surely determined by the Brownian motion Z . We say that the triple (M, e_0, T) converges in the scaling limit (in the peanosphere sense) to a γ -quantum cone decorated by an independent whole-plane space-filling SLE_κ if the joint law of the contour functions (or some slight variant thereof) of the primal and dual trees T and T^* converges in the scaling limit to the joint law of the two coordinates of Z .

The present paper is a generalization of [She16], which was the first work to study peanosphere convergence. The paper [She16] considered rooted critical FK planar maps. For $n \in \mathbb{N}$ and $q \geq 0$, a *rooted critical FK planar map with parameter q and size n* is a triple (M, e_0, S) consisting of a planar map M with n edges, a distinguished oriented root edge e_0 for M , and a set S of edges of M , sampled with probability proportional to $q^{K(S)/2}$, where $K(S)$ is the number of connected components of S plus the number of complementary connected components of S . The conditional law of S given M is that of the self-dual FK model on M [FK72]. An *infinite-volume rooted critical FK planar map with parameter q* is the infinite-volume limit of rooted critical FK planar maps of size n in the sense of Benjamini–Schramm [BS01].

There is a natural (but not bijective) means of obtaining a spanning tree T of M from the FK edge set S , which depends on the choice of e_0 ; see [Ber08b, She16]. It is conjectured [She16, DMS14] that the triple (M, e_0, T) converges in the scaling limit to an LQG sphere with parameter γ decorated by an independent whole-plane space-filling SLE_κ with parameters satisfying

$$\sqrt{q} = -2 \cos\left(\frac{4\pi}{\kappa}\right), \quad \gamma = \frac{4}{\sqrt{\kappa}}. \quad (1.1)$$

In [She16, Thm. 2.5], this convergence is proven in the peanosphere sense in the case of infinite-volume FK planar maps. This is accomplished by means of a bijection, called the *Sheffield hamburger-cheeseburger bijection*, between triples (M, e_0, S) consisting of a rooted

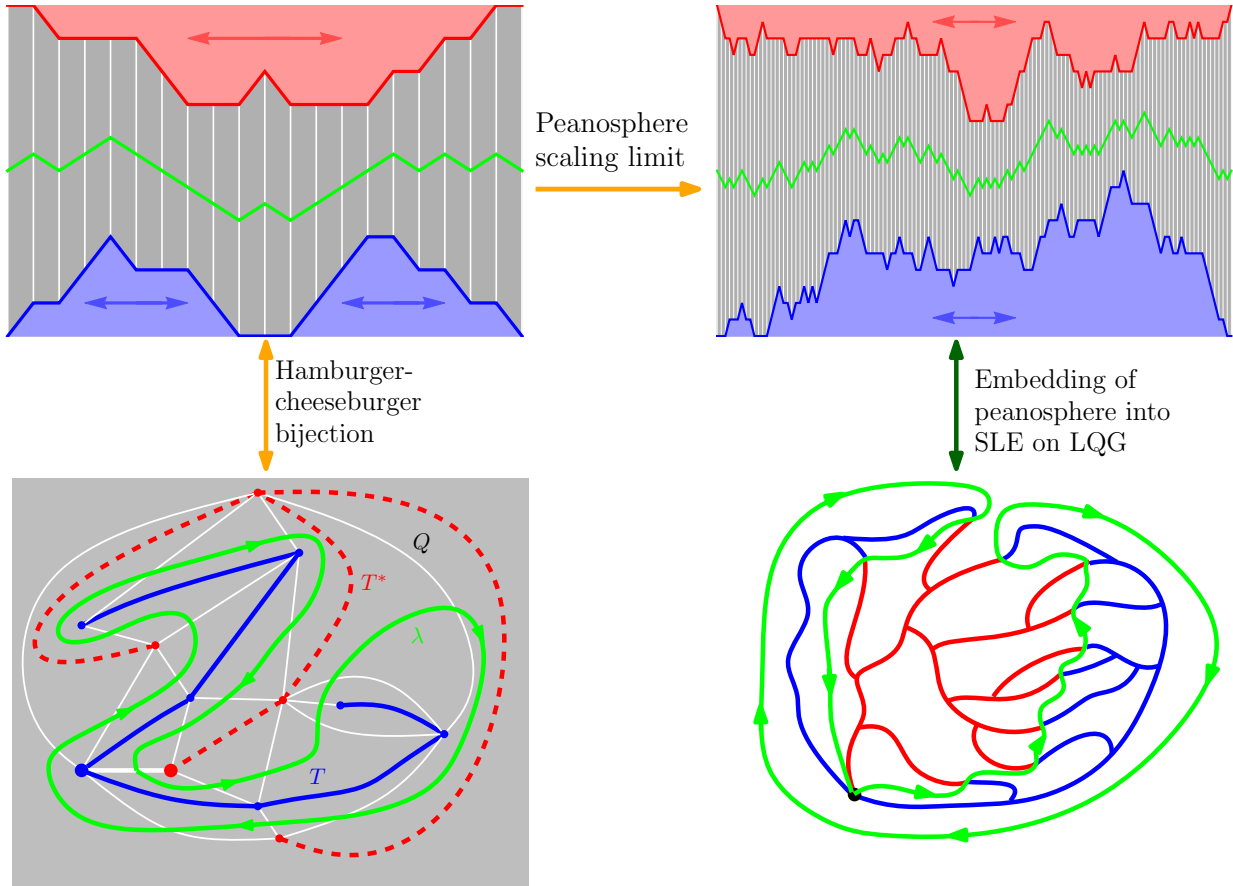


Figure 2: Shown on the top left are the contour functions for the discrete primal tree (blue) and dual tree (red) for the tree-decorated planar map on the bottom left using Sheffield’s hamburger-cheeseburger bijection [She16]. Vertices of the tree and dual tree correspond to blue and red horizontal segments in the contour representation; edges of the tree and dual tree correspond to matching up and down steps. The white boundaries between quadrangles in the quadrangulation correspond to the white vertical segments between the blue and red contour functions; the bold white boundary in the quadrangulation, which marks the starting point of the Peano curve, corresponds to the left and right edges in the contour diagram. The main contribution of the current paper is to establish an infinite volume version of the scaling limit result indicated by the orange horizontal arrow on the top. That is, if one first takes a limit as the size of the map tends to infinity, then the contour functions for the infinite discrete pair of trees converge to a two-dimensional correlated Brownian motion which encode a pair of infinite continuum random trees (CRTs) — this is convergence in the so-called *peanosphere sense*. The main result of [DMS14] implies that these two infinite CRTs glued together as shown (i.e. contracting the vertical white segments in addition to gluing along the horizontal arrows) determine their embedding into an SLE-decorated LQG surface. That is, if one observes the two contour functions on the top right, then one can measurably recover the LQG surface decorated with an SLE indicated on the bottom right and conversely if one observes the SLE-decorated LQG surface on the bottom right then one can measurably recover the contour functions on the top right. This allows us to interpret our scaling limit result as a convergence result to SLE-decorated LQG.

planar map of size n and a distinguished edge set S ; and certain words in an alphabet of five symbols (representing two types of “burgers” and three types of “orders”). This bijection is essentially equivalent for a fixed choice of M to the bijection of [Ber08b]. The word associated with a triple (M, e_0, S) gives rise to a walk on \mathbb{Z}^2 whose coordinates are (roughly speaking) the contour function of the spanning tree T of M naturally associated with S (under the mapping mentioned in the previous paragraph) and the contour function of the dual spanning tree T^* of the dual map M^* . There is also an infinite-volume version of Sheffield’s bijection which is a.s. well defined for infinite-volume FK planar maps. See [Che17] for a detailed exposition of this version of the bijection.

Various strengthenings of Sheffield’s scaling limit result (including an analogous scaling limit result for finite-volume FK planar maps) are proven in [GMS17a, GS17, GS15]. See also [Che17, BLR17] for additional results on FK planar maps and [GM17] for a scaling limit result in a stronger topology which is proven using the above peanosphere scaling limit results.

In [KW16], a new family of probability measures on spanning-trees of (deterministic) rooted planar maps, which generalizes the law arising from the self-dual FK model, was introduced. As explained in that paper, the law on trees T of a rooted map (M, e_0) arising from a self-dual FK model is given by the distribution on all spanning trees of M weighted by $y^{a(T)}$, where $y = \sqrt{q} + 1$ and $a(T) = a(T, e_0) \in \mathbb{N}$ is the “embedding activity” of T (which depends on the choice of root e_0 ; we will remind the reader of the definition later). It also makes sense to consider the probability measure on trees T weighted by $y^{a(T)}$ for $y \in (0, 1)$, so that trees with a lower embedding activity are more likely. The (unifying) discrete model corresponding to any $y \geq 0$ is called a y -active spanning tree.

In the context of the current paper, it is natural to look at a joint law on the triple (M, e_0, T) such that the marginal on (M, e_0) is the measure which weights a rooted planar map by the partition function of active spanning trees. Indeed, as we explain later, with this choice of law, exploring the tree respects the Markovian structure of the map. We call a random triple sampled from this law a *random rooted active-tree-decorated planar map with parameter $y \geq 0$ and size $n \in \mathbb{N}$* . The limiting case $y = 0$ corresponds to a spanning tree conditioned to have the minimum possible embedding activity, which is equivalent to a bipolar orientation on M for which the source and sink are adjacent [Ber08a] (see [KMSW15] for more on random bipolar-oriented planar maps).

It is conjectured in [KW16] that for $y \in [0, 1)$ the scaling limit of a random spanning tree T on large subgraphs of a two-dimensional lattice sampled with probability proportional to $y^{a(T)}$ is an SLE_κ with $\kappa \in (8, 12]$ determined by

$$\frac{y - 1}{2} = -\cos\left(\frac{4\pi}{\kappa}\right). \tag{1.2}$$

It is therefore natural to expect that the scaling limit of a rooted active-tree-decorated planar map is a γ -LQG surface decorated by an independent space-filling SLE_κ with $\kappa \in (8, 12]$ as in (1.2) and $\gamma = 4/\sqrt{\kappa}$.

We introduce in Section 1.3 a two-parameter family of probability measures on words in an alphabet of 8 symbols which generalizes the hamburger-cheeseburger model of [She16]. Under the bijection of [She16], each of these models corresponds to a probability measure on spanning-tree-decorated planar maps. One parameter in our model corresponds to the

parameter y of the active spanning tree, and the other, which we call z , controls the extent to which the tree T and its corresponding dual tree T^* are “tangled together”. This second parameter can also be interpreted in terms of some form of *bending energy* of the Peano curve which separates the two trees, in the sense of [BBG12, DGK00]; see Remark 1.10. We prove an analogue of [She16, Thm. 2.5] for our model which in particular implies that active-tree-decorated planar maps for $0 \leq y < 1$ converge in the scaling limit to γ -quantum cones decorated by SLE_κ in the peanosphere sense for $\kappa \in (8, 12]$ as in (1.2) and $\gamma = 4/\sqrt{\kappa}$. If we also vary z , the other parameter of our model, we obtain tree-decorated random planar maps which converge in the peanosphere sense to $4/\sqrt{\kappa}$ -quantum cones decorated by space-filling SLE_κ for any value of $\kappa > 8$.

Remark 1.1. When $y = 0$, an active-tree-decorated planar map is equivalent to a uniformly random bipolar-oriented planar map [Ber08b]. In [KMSW15], the authors use a bijective encoding of bipolar-oriented planar maps, which is not equivalent to the one used in this paper, to show that random bipolar-oriented planar maps with certain face degree distributions converge in the peanosphere sense to an SLE_{12} -decorated $\sqrt{4/3}$ -LQG surface, both in the finite-volume and infinite-volume cases (see also [GHS16] for a stronger convergence result). In the special case when $z = 1$, our Theorem 1.14 implies convergence of infinite-volume uniform bipolar-oriented planar maps in the peanosphere sense, but with respect to a different encoding of the map than the one used in [KMSW15]. More precisely, bipolar-oriented maps are encoded in [KMSW15] by a random walk in \mathbb{Z}^2 with a certain step distribution. The encoding of bipolar-oriented maps by the generalized hamburger-cheeseburger bijection corresponds to a random walk in $\mathbb{Z}^2 \times \{0, 1\}$ with a certain step distribution. Both of these walks converge in law to a correlated Brownian motion (ignoring the extra bit in the hamburger-cheeseburger bijection), and the correlations are the same, so we say that they both converge in the peanosphere sense.

1.2 Basic notation

We write \mathbb{N} for the set of positive integers.

For $a < b \in \mathbb{R}$, we define the discrete intervals $[a, b]_{\mathbb{Z}} := [a, b] \cap \mathbb{Z}$ and $(a, b)_{\mathbb{Z}} := (a, b) \cap \mathbb{Z}$.

If a and b are two quantities, we write $a \preceq b$ (resp. $a \succeq b$) if there is a constant C (independent of the parameters of interest) such that $a \leq Cb$ (resp. $a \geq Cb$). We write $a \asymp b$ if $a \preceq b$ and $a \succeq b$.

1.3 Generalized burger model

We now describe the family of words of interest to us in this paper. These are (finite or infinite) words which we read from left to right and which consist of letters representing burgers and orders which are matched to one another following certain rules. Several basic properties of this model are proved in Appendix A. Let

$$\Theta_0 := \{\mathbf{h}, \mathbf{c}, \mathbf{H}, \mathbf{C}\}, \tag{1.3}$$

and let $\mathcal{W}(\Theta_0)$ be the set of all finite words consisting of elements of Θ_0 . The alphabet Θ_0 generates a semigroup whose elements are words in $\mathcal{W}(\Theta_0)$ modulo the relations

$$\begin{aligned} cC = hH = \emptyset & \quad (\text{order fulfillment}) \\ cH = Hc, \quad hC = Ch. \end{aligned} \tag{1.4}$$

Following Sheffield [She16], we think of h, c, H, C as representing a hamburger, a cheeseburger, a hamburger order, and a cheeseburger order, respectively. A hamburger order is fulfilled by the freshest available hamburger (i.e., the rightmost hamburger which has not already fulfilled an order) and similarly for cheeseburger orders. We say that an order and a burger which cancel out via the first relation of (1.4) have been *matched*, and that the order has *consumed* the burger. See Fig. 3 (a) for a diagram representing matchings in an example.

We enlarge the alphabet by defining

$$\Theta := \Theta_0 \cup \{d, e, F, S\}, \tag{1.5}$$

and let $\mathcal{W}(\Theta)$ be the set of all finite words consisting of elements of Θ . The alphabet Θ generates a semigroup whose elements are finite words consisting of elements of Θ modulo the relations (1.4) and the additional relations

$$\begin{aligned} hF = hH = \emptyset & \quad cF = cC = \emptyset \\ hS = hC & \quad cS = cH \\ hd = hh & \quad cd = cc \\ he = hc & \quad ce = ch. \end{aligned} \tag{1.6}$$

In the language of burgers, the symbol F represents a “flexible order” which requests the freshest available burger. The symbol S represents a “stale order” which requests the freshest available burger of the type *opposite* the freshest available burger. The symbol d represents a “duplicate burger” which acts like a burger of the same type as the freshest available burger. The symbol e represents an “opposite burger” which acts like a burger of the type opposite the freshest available burger. The model of [She16] includes the flexible order F but no other elements of $\Theta \setminus \Theta_0$.

If a symbol in $\{d, e, F, S\}$ has been replaced by a symbol in Θ_0 via one of the relations in (1.6), we say that this symbol is *identified by* the earlier symbol in the relation; and *identified as* the symbol in Θ_0 with which it has been replaced.

Given a word $x = x_1 \cdots x_n \in \mathcal{W}(\Theta)$, we write $|x| = n$ for the number of symbols in x .

Definition 1.2. A word in $\mathcal{W}(\Theta)$ is called *reduced* if all of its orders, d ’s, and e ’s lie to the left of all of its h ’s and c ’s. In Lemma A.1 we show that for any finite word x , there is a unique reduced word which can be obtained from x by applying the relations (1.4) and (1.6), which we call the reduction of x , and denote by $\mathcal{R}(x)$.

An important property of the reduction operation (proved in Lemma A.2) is

$$\mathcal{R}(xy) = \mathcal{R}(\mathcal{R}(x)\mathcal{R}(y)).$$

Note that for any $x \in \mathcal{W}(\Theta)$, we have $|\mathcal{R}(x)| \leq |x|$.

Definition 1.3. We write $x' = \mathcal{I}(x)$ (the *identification* of x) for the word with $|x'| = |x|$ obtained from x as follows. For each $i \in \{1, \dots, |x|\}$, if $x_i \in \Theta_0$, we set $x'_i = x_i$. If $x_i \in \{\mathbf{F}, \mathbf{S}\}$ and x_i is replaced by a hamburger order (resp. cheeseburger order) via (1.6) when we pass to the reduced word $\mathcal{R}(x)$, we set $x'_i = \mathbf{H}$ (resp. $x'_i = \mathbf{C}$). If $x_i \in \{\mathbf{d}, \mathbf{e}\}$ and x_i is replaced with a hamburger (resp. cheeseburger) via (1.6) when we pass to the reduced word, we set $x'_i = \mathbf{h}$ (resp. $x'_i = \mathbf{c}$). Otherwise, we set $x'_i = x_i$. We say that a symbol x_i is *identified in the word* x if x'_i is an element of Θ_0 , and *unidentified in the word* x otherwise.

For example,

$$\begin{aligned}\mathcal{R}(\mathbf{cFdhS}) &= \mathbf{dCh} \\ \mathcal{I}(\mathbf{cFdhS}) &= \mathbf{cCdHc}.\end{aligned}$$

Note that $\mathcal{R}(\mathcal{I}(x)) = \mathcal{R}(x)$. Note also that any symbol x_i which has a match when we pass to $\mathcal{R}(x)$ is necessarily identified, but identified symbols are not necessarily matched. Indeed, symbols in Θ_0 are always identified, and there may be \mathbf{S} , \mathbf{d} , and/or \mathbf{e} symbols in x which are identified, but do not have a match.

Definition 1.4. For $\theta \in \Theta$ and a finite word x consisting of elements of Θ , we write

$$\begin{aligned}\mathcal{N}_\theta(x) &:= \text{number of } \theta\text{-symbols in } x \\ \mathcal{N}_{\theta_1|\dots|\theta_k}(x) &:= \mathcal{N}_{\theta_1}(x) + \dots + \mathcal{N}_{\theta_k}(x)\end{aligned}$$

We also define

$$\begin{aligned}\mathcal{B}(x) &:= \mathcal{N}_{\mathbf{h}|\mathbf{c}|\mathbf{d}|\mathbf{e}}(x) = \text{number of burgers in } x \\ \mathcal{O}(x) &:= \mathcal{N}_{\mathbf{H}|\mathbf{C}|\mathbf{F}|\mathbf{S}}(x) = \text{number of orders in } x \\ \mathcal{C}(x) &:= \mathcal{B}(x) - \mathcal{O}(x)\end{aligned}$$

and

$$\begin{aligned}d(x) &:= \mathcal{N}_{\mathbf{h}}(x) - \mathcal{N}_{\mathbf{H}}(x) \\ d^*(x) &:= \mathcal{N}_{\mathbf{c}}(x) - \mathcal{N}_{\mathbf{C}}(x) \\ \vec{d}(x) &:= (d(x), d^*(x)) \\ \mathcal{D}(x) &:= d(x) - d^*(x).\end{aligned}$$

The reason for the notation d and d^* is that these quantities represent distances to the root edge in the primal and dual trees, respectively, in the construction of [She16, § 4.1] (see the discussion just below). Note that these quantities are still defined even if x has some symbols in $\{\mathbf{d}, \mathbf{e}, \mathbf{F}, \mathbf{S}\}$.

Fig. 3 (b) shows a random-walk representation of \vec{d} computed on increasing prefixes of a finite (identified) word. This process will later be our main object of study.

If x is a finite word consisting of elements of Θ with $\mathcal{R}(x) = \emptyset$, then the bijection described in [She16, § 4.1] applied to $\mathcal{I}(x)$ uniquely determines a rooted spanning-tree-decorated map (M, e_0, T) associated with x .

We now describe the probability measure on words which gives rise to the law on spanning-tree-decorated planar maps which we are interested in. Let

$$\mathcal{P} := \{(p_{\mathbf{F}}, p_{\mathbf{S}}, p_{\mathbf{d}}, p_{\mathbf{e}}) \in [0, 1]^4 : p_{\mathbf{F}} + p_{\mathbf{S}} \leq 1 \quad \text{and} \quad p_{\mathbf{d}} + p_{\mathbf{e}} < 1\}.$$

For a vector $\vec{p} = (p_F, p_S, p_d, p_e) \in \mathcal{P}$, we define a probability measure $\mathbb{P} = \mathbb{P}_{\vec{p}}$ on Θ by

$$\begin{aligned} \mathbb{P}(F) &= \frac{p_F}{2}, & \mathbb{P}(S) &= \frac{p_S}{2}, & \mathbb{P}(H) &= \mathbb{P}(C) = \frac{1 - p_F - p_S}{4} \\ \mathbb{P}(d) &= \frac{p_d}{2}, & \mathbb{P}(e) &= \frac{p_e}{2}, & \mathbb{P}(h) &= \mathbb{P}(c) = \frac{1 - p_d - p_e}{4}. \end{aligned} \quad (1.7)$$

Let $X = \cdots X_{-1}X_0X_1 \cdots$ be a bi-infinite word whose symbols are i.i.d. samples from the probability measure (1.7). The identification procedure extends naturally to bi-infinite words, and we show in Appendix A that a.s. the bi-infinite identified word $X' = \mathcal{I}(X)$ exists and contains only elements of Θ_0 . Furthermore, a.s. each order in X consumes a burger and each burger in X is consumed by an order. That is, each symbol X_i in X has a match $X_{\phi(i)}$ which cancels it out, so that in effect the reduced bi-infinite word $\mathcal{R}(X)$ is a.s. empty.

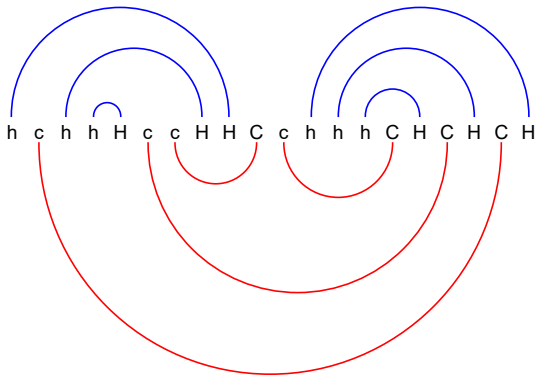
Definition 1.5. We write $X' = \cdots X'_{-1}X'_0X'_1 \cdots$ for the identification of the bi-infinite word X .

Definition 1.6. For $i \in \mathbb{Z}$, we write $\phi(i) \in \mathbb{Z}$ for the index of the symbol matched to X_i in the word X . (From the above property, a.s. ϕ is an involution of \mathbb{Z} .)

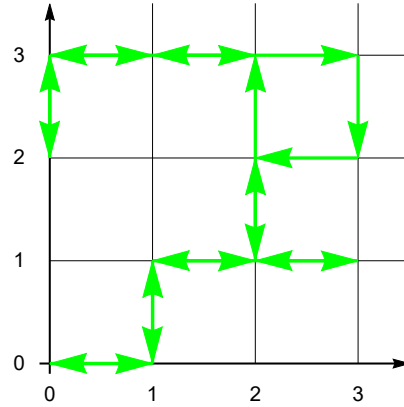
For $a < b \in \mathbb{R}$, we write

$$X(a, b) := \mathcal{R}(X_{\lfloor a \rfloor} \cdots X_{\lfloor b \rfloor}) \quad \text{and} \quad X'(a, b) := \mathcal{R}(X'_{\lfloor a \rfloor} \cdots X'_{\lfloor b \rfloor}). \quad (1.8)$$

The aforementioned results of Appendix A allow us to use the infinite-volume version of Sheffield's bijection [She16] (which is described in full detail in [Che17]) to construct an infinite-volume rooted spanning-tree-decorated planar map $(M^\infty, e_0, T^\infty)$ from the identified word X' of Definition 1.5.



(a) The word associated to the decorated map of Fig. 1. The chords represent the matchings between orders and burgers that fulfill them.



(b) The trace of the walk $(\vec{d}_i)_{0 \leq i \leq |x|}$ corresponding to the \vec{d} vector of increasing prefixes of the word $x = h c h h H c c H H C c h h h C H C H C H$. The walk gives the number of available hamburgers and cheeseburgers as a function of time.

Figure 3

The set \mathcal{P} describes a four-parameter family of probability measures on Θ , and hence a four-parameter family of probability measures on triples $(M^\infty, e_0, T^\infty)$. However, as we will see in Corollary 1.13 below, the law of X' (and hence also the law of $(M^\infty, e_0, T^\infty)$) depends only on the two parameters $p_F - p_S$ and $p_d - p_e$ (equivalently the parameters y and z defined in (1.11)).

Remark 1.7. The model described above includes three special symbols which are natural generalizations of the special order F included in [She16]: the order S has the opposite behavior as the order F , and the burgers d and e behave in the same way as S and F but with burgers in place of orders. As we will see in Section 1.4, each of these symbols has a natural topological interpretation in terms of the spanning-tree-decorated rooted planar maps encoded by words consisting of elements of Θ .

Remark 1.8. As we will see, the words we consider in this paper can behave in very different ways from the words considered in [She16], which do not include the symbols S , d , or e . For example, in the setting of Section 3, where we allow S 's and d 's but not F 's or e 's, the net hamburger/cheeseburger counts $d(X(1, n))$ and $d^*(X(1, n))$ in a reduced word tend to be *negatively* correlated (Theorem 1.15) and the reduced word $X(1, n)$ tends to have *more* symbols than the corresponding reduced word in the case when $p_F = p_S = p_d = p_e = 0$ (Lemma 3.1). The opposite is true in the setting of [She16]. As another example, in the setting of Section 3 we expect, but do not prove, that the infinite reduced word $X(1, \infty)$ a.s. contains only finitely many unidentified S 's and d 's, whereas $X(1, \infty)$ a.s. contains infinitely many unidentified F 's in the setting of [She16] (Remark 3.7).

1.4 Active spanning trees with bending energy

Let (M, e_0) be a (deterministic) planar map with n edges with oriented root edge e_0 . Let M^* be the dual map of M and let (Q, e_0) be the associated rooted quadrangulation (as described at the beginning of the introduction). In this subsection we introduce a probability measure on spanning trees of M which is encoded by the model of Section 1.3.

There is a bijection between spanning trees on M and *noncrossing Eulerian cycles* on the *medial graph* of M , which is the planar dual graph of Q . (An Eulerian cycle is a cycle which traverses each edge exactly once, vertices may be repeated.) To describe this bijection, let λ be a noncrossing Eulerian cycle on the dual of Q starting and ending at e_0 . By identifying an edge of Q^* with the edge of Q which crosses it, we view λ as a function from $[1, 2n]_{\mathbb{Z}}$ to the edge set of Q . Each quadrilateral of Q is bisected by one edge of M and one edge of M^* , and λ crosses each such quadrilateral exactly twice (one such quadrilateral is shown in gray in Figure 4). Hence λ crosses each edge of M and each edge of M^* either 0 or 2 times. The set T of edges of M which are not crossed by λ is a spanning tree of M whose discrete Peano curve is λ and the set T^* of edges of M^* not crossed by λ is the corresponding dual spanning tree of M^* . Each quadrilateral of Q is bisected by an edge of either T or T^* (but not both). This establishes a one-to-one correspondence between noncrossing Eulerian cycles on the dual of Q starting and ending at e_0 and spanning trees of M .

Now fix a noncrossing Eulerian cycle λ as above. For $i \in [1, 2n]_{\mathbb{Z}}$ we let \bar{e}_i be the edge of $T \cup T^*$ which bisects the last quadrilateral of Q crossed by λ exactly once at or before time i , if such a quadrilateral exists. Let e be an edge of $T \cup T^*$, and let $j, k \in [1, 2n]_{\mathbb{Z}}$ be the first

and second times respectively that λ crosses the quadrilateral of Q bisected by e . Observe that if e and \bar{e}_{k-1} both belong to M or both belong to M^* , then in fact $e = \bar{e}_{k-1}$. In this case, we say that e is of *active type*; this definition coincides with “embedding activity”, as illustrated in Figure 4. If \bar{e}_{j-1} exists and e and \bar{e}_{j-1} either both belong to M or both belong to M^* , then we say that e is of *duplicate type*; duplicate edges are illustrated in Figure 5, and Remark 1.10 below discusses their relevance. Figure 6 shows the active and duplicate edges from Figure 1. An edge can be of both active and duplicate type, or of neither active nor duplicate type.

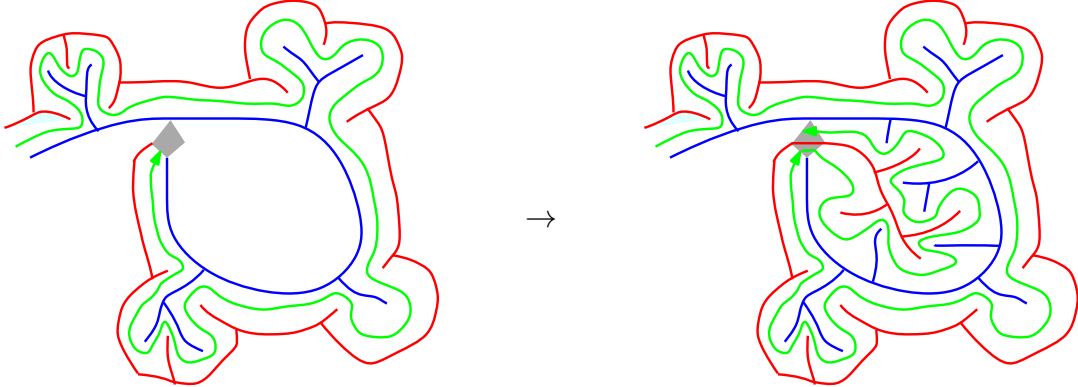


Figure 4: The Peano exploration process with the Peano path λ in green, primal tree T in blue, and dual tree T^* in red. When the gray quadrilateral is first encountered (left panel), the dual edge e is forced to be present (otherwise there would be a primal cycle). This means that e is “embedding active”, in the sense of [Ber08b] (see also [Cou14]). The Peano curve then explores the map in the region enclosed by the blue near-cycle and exits through the same (gray) quadrilateral (right panel). Just before the second time the gray quadrilateral is encountered, the most recent quadrilateral encountered exactly once is the gray quadrilateral, so $\bar{e}_{k-1} = e$, so e is of active type as defined above. This characterization of the embedding activity was explained in [She16].

Following [Ber08b, She16], a noncrossing Eulerian cycle λ based at e_0 can be encoded by means of a word x of length $2n$ consisting of elements of Θ_0 with reduced word $\mathcal{R}(x) = \emptyset$. The symbol h (resp. H) corresponds to the first (resp. second) time that λ crosses an edge of M , and the symbol c (resp. C) corresponds to the first (resp. second) time that λ crosses an edge of M^* . The two times that λ crosses a given quadrilateral of Q correspond to a burger and the order which consumes it. With \bar{e}_i as above, the burger corresponding to the quadrilateral bisected by \bar{e}_i is the same as the rightmost burger in the reduced word $\mathcal{R}(x_1 \cdots x_i)$; the edge \bar{e}_i is undefined if and only if this reduced word is empty. Therefore edges of active type correspond to orders which consume the most recently added burger that has not yet been consumed, and edges of duplicate type correspond to burgers which are the same type as the the most recently added burger that has not yet been consumed.

For a spanning tree T of M rooted at e_0 , we let $a(T)$ be the number of active edges and $d(T)$ the number of duplicate edges of its Peano curve λ . These quantities depend on

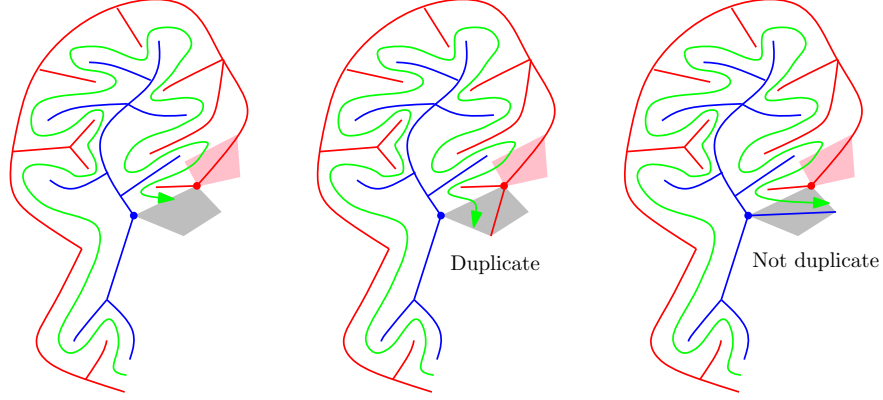


Figure 5: Left: the two trees T and T^* and the Peano curve λ (in green) run up until step $i - 1$. The pink quadrilateral is the most recent one which has been crossed exactly once by λ by time $i - 1$ and \bar{e}_{i-1} is the red edge which bisects this quadrilateral. The vertices v_{i-1}^0 and v_{i-1}^1 discussed in Remark 1.10 are shown in red and blue, respectively. At step i , λ will either bend away from the red vertex (middle) or toward the red vertex (right). In the former case the edge which bisects the grey quadrilateral belongs to the same tree as \bar{e}_{i-1} , so the edge $\lambda(i)$ is of duplicate type.

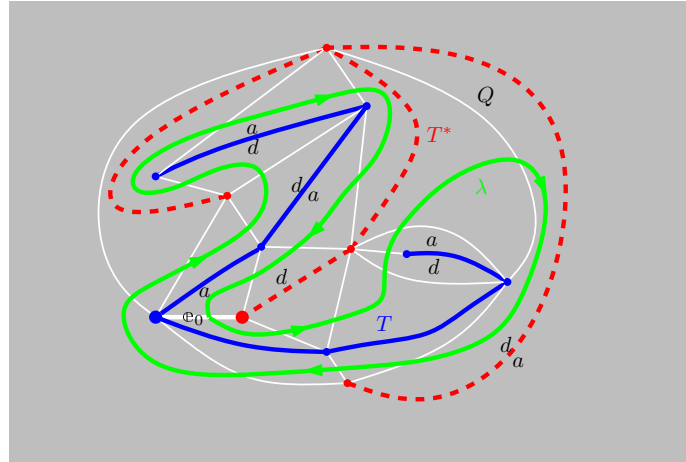


Figure 6: The quadrangulation Q , the trees T and T^* , and the Peano curve λ constructed from the triple (M, e_0, T) of Figure 1 with active (resp. duplicate) edges of $T \cup T^*$ indicated with an a (resp. d). Edges can be both active and duplicate. The root edge e_0 is indicated by a thicker white line. If we allow symbols in Θ (rather than just Θ_0), the triple (M, e_0, T) can be encoded by many different words of length $2n$; more precisely, it can be encoded by any word whose identification is the word shown in Figure 3. The word corresponding to (M, e_0, T) with the smallest possible number of elements of Θ_0 is **heedFedSSFceddSFSFSF**. In this word, **F** (resp. **S**) symbols correspond to the second time λ crosses a quadrilateral of Q bisected by an active (resp. inactive) edge and **d** (resp. **e**) symbols correspond to the first time λ crosses a quadrilateral of Q bisected by a duplicate (resp. non-duplicate) edge. The **h** and **c** symbols correspond to times i for which the edge \bar{e}_i is not defined.

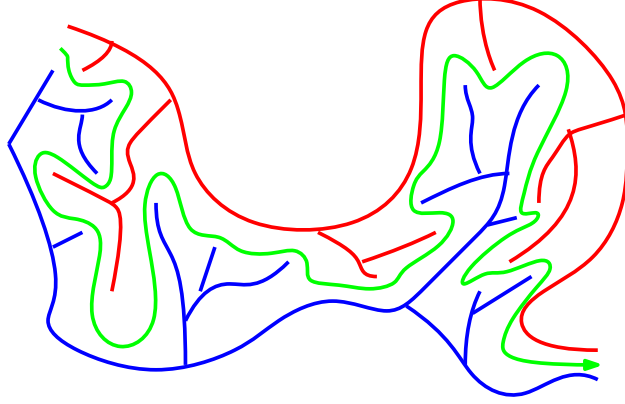


Figure 7: Given an initial portion of the exploration process λ , the set of active and duplicate edges in the remainder of the graph only depends on the initial segment through its boundary. Consequently, the law of the decorated random map conditional on the part already drawn only depends on the white region with the boundary components consisting in the red and blue curves only visited on one side by the green curve.

the choice of e_0 . We define the partition function

$$\mathcal{Z}(M, e_0, y, z) = \sum_{\text{spanning tree } T} y^{a(T)} z^{d(T)}, \quad (1.9)$$

which gives rise, when $y, z \geq 0$, to a probability measure

$$\mathbb{P}[T] = \frac{y^{a(T)} z^{d(T)}}{\mathcal{Z}(M, e_0, y, z)}, \quad (1.10)$$

on the set of spanning trees T of M . This distribution on spanning trees satisfies a domain Markov property: for $i \in [1, 2n]_{\mathbb{Z}}$, the conditional law of $\lambda|_{[i+1, 2n]_{\mathbb{Z}}}$ given $\lambda|_{[1, i]_{\mathbb{Z}}}$ depends only on the set of quadrilaterals and half-quadrilaterals not yet visited by λ together with the starting and ending points of the path $\lambda([1, i]_{\mathbb{Z}})$. See Figure 7 for an illustration of the Markov property of the random decorated map. We call a spanning tree sampled from the above distribution an *active spanning tree with bending energy*, for reasons which are explained in the remarks below.

Remark 1.9. There are other notions of “active edge”, each of which gives rise to the same Tutte polynomial

$$T_M(x, y) = \sum_{\text{spanning trees } t \text{ of } M} x^{\# \text{ internally active edges of } t} y^{\# \text{ externally active edges of } t}.$$

The embedding activity illustrated in Figure 4 differs from Tutte’s original definition, but is more natural in this context because it has the domain Markov property, and has a simple characterization in terms of the hamburger-cheesburger model. The embedding activity is similar to Bernardi’s definition [Ber08b, § 3.1, Def. 3], but with “maximal” in place of “minimal”. The partition function $\mathcal{Z}(M, e_0, y, 1) = T_M(y, y)$ is the Tutte polynomial of M

evaluated at (y, y) . In this case ($z = 1$), the partition function is that of the active spanning tree model of [KW16], which when $y \geq 1$ coincides with the partition function of the self-dual Fortuin–Kasteleyn (FK) model with parameter $q = (y - 1)^2$.

Remark 1.10. To our knowledge, the notion of edges of duplicate type does not appear elsewhere in the literature. However, this notion can be viewed as a variant of the notion of *bending energies* studied in [BBG12] and initially introduced in a different guise in [DGK00]. Suppose $(\mathcal{T}, \mathfrak{v})$ is a rooted triangulation and ℓ is a non-self-crossing oriented loop in the dual of \mathcal{T} , viewed as a cyclically ordered sequence of distinct triangles in \mathcal{T} . For each triangle t hit by loop ℓ , there is a single edge of t which is not shared by the triangles hit by ℓ immediately before and after t . We say that t points outward (resp. inward) if this edge is on the same (resp. opposite) side of the loop ℓ as the root vertex \mathfrak{v} . The *bending* of ℓ is the number of pairs of consecutive triangles which either both point outward or both face inward. Such a pair of triangles corresponds to a time when loop ℓ “bends around” a vertex. If we view the Peano curve λ considered above as a loop in the triangulation whose edges are the union of the edges of the quadrangulation Q and the trees T and T^* , then the bending of λ in the sense of [BBG12] is the number of consecutive pairs of symbols of one of the forms hh, HH, hH, Hh, cc, CC, cC, or Cc in the identified word which encodes the triple (M, e_0, T) under Sheffield’s bijection.

The loops considered in [BBG12] are those arising from variants of the $O(n)$ model, so are expected to be non-space-filling in the limit (in fact they are conjectured to converge to CLE_κ loops for $\kappa \in (8/3, 8)$ [She09]). For space-filling loops (such as the Peano curve λ), it is natural to keep track of times when the loop returns to a triangle which shares a vertex with one it hits previously, and then bends toward the set of triangles which it has hit more recently.

Let us now be more precise about what this means. It is easy to see from Sheffield’s bijection (and is explained in [Che17, § 4.2]) that two edges $\lambda(i)$ and $\lambda(j)$ for $i, j \in [1, 2n]_{\mathbb{Z}}$ share a primal (resp. dual) endpoint if and only if the rightmost hamburger (resp. cheeseburger) in the reduced words $\mathcal{R}(x_1 \cdots x_i)$ and $\mathcal{R}(x_1 \cdots x_j)$ both correspond to the same burger in the original word x , or if these reduced words both have no hamburgers (resp. cheeseburgers). Consequently, an edge of duplicate type can be equivalently defined as an edge $\lambda(i)$ such that λ crosses a quadrilateral of Q for the first time at time i and the following is true. Let v_{i-1}^0 and v_{i-1}^1 be the endpoints of $\lambda(i-1)$, enumerated in such a way that λ hits an edge which shares the endpoint v_{i-1}^0 for the first time before it hits an edge which shares the endpoint v_{i-1}^1 for the first time. Then λ turns toward v_{i-1}^1 at time i (cf. Figure 5). From this perspective, a time when λ crosses a quadrilateral bisected by an edge of duplicate type can be naturally interpreted as a time when λ “bends away from the set of triangles which it has hit more recently”. Hence our model is a probability measure on planar maps decorated by an active spanning tree (in the sense of [KW16]), weighted by an appropriate notion of the bending of the corresponding Peano curve.

The generalized burger model of Section 1.3 encodes a random planar map decorated by an active spanning tree with bending energy. The correspondence between the probability vector $\vec{p} = (p_{\text{F}}, p_{\text{S}}, p_{\text{d}}, p_{\text{e}}) \in \mathcal{P}$ and the pair of parameters (y, z) is given by

$$y = \frac{1 + p_{\text{F}} - p_{\text{S}}}{1 - p_{\text{F}} + p_{\text{S}}} \quad \text{and} \quad z = \frac{1 + p_{\text{d}} - p_{\text{e}}}{1 - p_{\text{d}} + p_{\text{e}}}, \quad (1.11)$$

i.e.

$$p_{\mathbb{F}} - p_{\mathbb{S}} = \frac{y-1}{1+y} \quad \text{and} \quad p_{\mathbb{d}} - p_{\mathbb{e}} = \frac{z-1}{1+z}. \quad (1.12)$$

To see why this is the case, let \dot{X} be a random word of length $2n$ sampled from the conditional law of $X_1 \cdots X_{2n}$ given $\{X(1, 2n) = \emptyset\}$, where X is the bi-infinite word from Section 1.3 (in the case when $p_{\mathbb{S}} = 1$, we allow the last letter of \dot{X} to be a flexible order, since a word whose orders are all \mathbb{S} 's cannot reduce to the empty word). Let $\dot{X}' := \mathcal{I}(\dot{X})$ and let (M, e_0, T) be the rooted spanning-tree-decorated planar map associated with \dot{X}' under the bijection of [She16, § 4.1].

Lemma 1.11. *1. The law of (M, e_0, T) is that of the uniform measure on edge-rooted, spanning-tree decorated planar maps weighted by $y^{a(T)} z^{d(T)}$, with y and z as in (1.11).*

2. The conditional law of T given (M, e_0) is given by the law (1.10); and when $z = 1$, the law of (M, e_0, T) is that of an active-tree-decorated planar map (as defined in the introduction).

3. If $(M^\infty, e_0^\infty, T^\infty)$ is the infinite-volume rooted spanning-tree-decorated planar map associated with X (by the infinite-volume version of Sheffield's bijection, see the discussion just after (1.8)), then $(M^\infty, e_0^\infty, T^\infty)$ has the law of the Benjamini-Schramm limit [BS01] of the law of (M, e_0, T) as $n \rightarrow \infty$.

Proof. Throughout the proof we write $a \propto b$ if a/b is a constant depending only on n and \vec{p} . Let $x \in \mathcal{W}(\Theta)$ be a word of length $2n$ which satisfies $\mathcal{R}(x) = \emptyset$. Note that x must contain n burgers and n orders. Then in the notation of Definition 1.4,

$$\mathbb{P}(\dot{X} = x) \propto \left(\frac{2p_{\mathbb{F}}}{1 - p_{\mathbb{F}} - p_{\mathbb{S}}} \right)^{\mathcal{N}_{\mathbb{F}}(x)} \left(\frac{2p_{\mathbb{S}}}{1 - p_{\mathbb{F}} - p_{\mathbb{S}}} \right)^{\mathcal{N}_{\mathbb{S}}(x)} \left(\frac{2p_{\mathbb{d}}}{1 - p_{\mathbb{d}} - p_{\mathbb{e}}} \right)^{\mathcal{N}_{\mathbb{d}}(x)} \left(\frac{2p_{\mathbb{e}}}{1 - p_{\mathbb{d}} - p_{\mathbb{e}}} \right)^{\mathcal{N}_{\mathbb{e}}(x)}. \quad (1.13)$$

Let $A^{\mathbb{H}}$ (resp. $\tilde{A}^{\mathbb{H}}$) be the set of $i \in [1, 2n]_{\mathbb{Z}}$ for which \dot{X}'_i is a hamburger order matched to a hamburger which is (resp. is not) the rightmost burger in $\dot{X}'(1, i-1)$ (notation as in (1.8)). Let $D^{\mathbb{H}}$ (resp. $\tilde{D}^{\mathbb{H}}$) be the set of $i \in [2, 2n]_{\mathbb{Z}}$ for which \dot{X}'_i is a hamburger, $\dot{X}'(1, i-1) \neq \emptyset$, and the rightmost burger in $\dot{X}'(1, i-1)$ is a hamburger (resp. cheeseburger). Define $A^{\mathbb{C}}$, $\tilde{A}^{\mathbb{C}}$, $D^{\mathbb{C}}$, and $\tilde{D}^{\mathbb{C}}$ similarly but with hamburgers and cheeseburgers interchanged. Then

$$a(T) = \#A^{\mathbb{H}} + \#A^{\mathbb{C}} \quad \text{and} \quad d(T) = \#D^{\mathbb{H}} + \#D^{\mathbb{C}}.$$

If we condition on \dot{X}' , then we can re-sample \dot{X} as follows. For each $i \in A^{\mathbb{H}}$, independently sample $\dot{X}_i \in \{\mathbb{H}, \mathbb{F}\}$ from the probability measure $\mathbb{P}(\mathbb{H}) = (1 - p_{\mathbb{F}} - p_{\mathbb{S}})/(1 + p_{\mathbb{F}} - p_{\mathbb{S}})$, $\mathbb{P}(\mathbb{F}) = 2p_{\mathbb{F}}/(1 + p_{\mathbb{F}} - p_{\mathbb{S}})$. For each $i \in \tilde{A}^{\mathbb{H}}$, independently sample $\dot{X}_i \in \{\mathbb{H}, \mathbb{S}\}$ from the probability measure $\mathbb{P}(\mathbb{H}) = (1 - p_{\mathbb{F}} - p_{\mathbb{S}})/(1 - p_{\mathbb{F}} + p_{\mathbb{S}})$, $\mathbb{P}(\mathbb{S}) = 2p_{\mathbb{S}}/(1 - p_{\mathbb{F}} + p_{\mathbb{S}})$. For each $i \in D^{\mathbb{H}}$, independently sample $\dot{X}_i \in \{\mathbb{h}, \mathbb{d}\}$ from the probability measure $\mathbb{P}(\mathbb{h}) = (1 - p_{\mathbb{d}} - p_{\mathbb{e}})/(1 + p_{\mathbb{d}} - p_{\mathbb{e}})$, $\mathbb{P}(\mathbb{d}) = 2p_{\mathbb{d}}/(1 + p_{\mathbb{d}} - p_{\mathbb{e}})$. For each $i \in \tilde{D}^{\mathbb{H}}$, independently sample $\dot{X}_i \in \{\mathbb{h}, \mathbb{e}\}$ from the probability measure $\mathbb{P}(\mathbb{h}) = (1 - p_{\mathbb{d}} - p_{\mathbb{e}})/(1 - p_{\mathbb{d}} + p_{\mathbb{e}})$,

$\mathbb{P}(e) = 2p_e/(1 - p_d + p_e)$. Then do the same for A^C , \tilde{A}^C , D^C , and \tilde{D}^C but with hamburgers and cheeseburgers interchanged.

The above resampling rule implies that with x as above,

$$\begin{aligned} \mathbb{P}(\dot{X} = x \mid \dot{X}' = \mathcal{I}(x)) &\propto \left(\frac{1 - p_F + p_S}{1 + p_F - p_S}\right)^{a(T)} \left(\frac{1 - p_d + p_e}{1 + p_d - p_e}\right)^{d(T)} \left(\frac{2p_F}{1 - p_F - p_S}\right)^{\mathcal{N}_F(x)} \\ &\quad \times \left(\frac{2p_S}{1 - p_F - p_S}\right)^{\mathcal{N}_S(x)} \left(\frac{2p_d}{1 - p_d - p_e}\right)^{\mathcal{N}_d(x)} \left(\frac{2p_e}{1 - p_d - p_e}\right)^{\mathcal{N}_e(x)}. \end{aligned} \quad (1.14)$$

By dividing (1.13) by (1.14), we obtain

$$\mathbb{P}(\dot{X}' = \mathcal{I}(x)) \propto y^{a(T)} z^{d(T)}.$$

Therefore, the probability of any given realization of (M, e_0, T) is proportional to $y^{a(T)} z^{d(T)}$, which gives assertion 1. Assertion 2 is an immediate consequence of assertion 1. Assertion 3 follows from the same argument used in [She16, § 4.2] together with the results of Appendix A. \square

Remark 1.12. The model described in Lemma 1.11 is self dual in the sense that the law of (M, e_0, T) is the same as the law of (M^*, e_0^*, T^*) , where M^* is the dual map of M , e_0^* is the edge of M^* which crosses e_0 , and T^* is the dual spanning tree (consisting of edges of M^* which do not cross edges of T). This duality corresponds to the fact that the law of the inventory accumulation model of Section 1.3 is invariant under the replacements $\mathbf{h} \leftrightarrow \mathbf{c}$ and $\mathbf{C} \leftrightarrow \mathbf{H}$. It may be possible to treat non-self dual variants of this model in our framework by relaxing the requirement that $\mathbb{P}(\mathbf{h}) = \mathbb{P}(\mathbf{c})$ and $\mathbb{P}(\mathbf{C}) = \mathbb{P}(\mathbf{H})$ in (1.7), but we do not investigate this. We remark that there are bijections and Brownian motion scaling limit results analogous to the ones in this paper for other random spanning-tree-decorated map models which do not possess this self duality; see, e.g., [KMSW15, LSW17].

We end by recording the following corollary of Lemma 1.11, which says that the law of the identification of the word X (and therefore the law of the associated tree-decorated map) depends on the parameter \vec{p} only via the quantities y and z of (1.11).

Corollary 1.13. *Suppose $\vec{p} = (p_F, p_S, p_d, p_e)$ and $\tilde{\vec{p}} = (\tilde{p}_F, \tilde{p}_S, \tilde{p}_d, \tilde{p}_e)$ are two vectors in \mathcal{P} which satisfy $p_F - p_S = \tilde{p}_F - \tilde{p}_S$ and $p_d - p_e = \tilde{p}_d - \tilde{p}_e$. Let $X = \cdots X_{-1} X_0 X_1 \cdots$ (resp. $\tilde{X} = \cdots \tilde{X}_{-1} \tilde{X}_0 \tilde{X}_1 \cdots$) be a bi-infinite word such that $\{X_i\}_{i \in \mathbb{N}}$ (resp. $\{\tilde{X}_i\}_{i \in \mathbb{N}}$) is a collection of i.i.d. samples from the probability measure (1.7) with probabilities \vec{p} (resp. with $\tilde{\vec{p}}$). Then the identifications $\mathcal{I}(X)$ and $\mathcal{I}(\tilde{X})$ agree in law.*

Proof. It follows from Lemma 1.11 that the infinite-volume tree-decorated planar maps $(M^\infty, e_0^\infty, T^\infty)$ and $(\tilde{M}^\infty, \tilde{e}_0^\infty, \tilde{T}^\infty)$ associated with $\mathcal{I}(X)$ and $\mathcal{I}(\tilde{X})$ agree in law. Since these maps uniquely determine $\mathcal{I}(X)$ and $\mathcal{I}(\tilde{X})$, respectively, via the same deterministic procedure, we infer that $\mathcal{I}(X) \stackrel{d}{=} \mathcal{I}(\tilde{X})$. \square

1.5 Statement of main results

Fix $\vec{p} = (p_F, p_S, p_d, p_e) \in \mathcal{P}$ and let X be the bi-infinite word from Section 1.3, whose symbols are i.i.d. samples from the probability measure 1.7. Also let $X' = \cdots X'_{-1} X'_0 X'_1 \cdots = \mathcal{I}(X)$ be the identification of X , as in Definition 1.5 and recall the notation (1.8). For $i \in \mathbb{Z}$, define (in the notation of Definition 1.4)

$$d(i) := \begin{cases} d(X'(1, i)) & i \geq 1 \\ 0 & i = 0 \\ d(X'(i+1, 0)) & i \leq -1 \end{cases} \quad \text{and} \quad d^*(i) := \begin{cases} d^*(X'(1, i)) & i \geq 1 \\ 0 & i = 0 \\ d^*(X'(i+1, 0)) & i \leq -1. \end{cases}$$

We extend d and d^* to \mathbb{R} by linear interpolation, and define $\vec{d}(t) := (d(t), d^*(t))$.

For $n \in \mathbb{N}$ and $t \in \mathbb{R}$, let

$$U^n(t) := n^{-1/2} d(nt), \quad V^n(t) := n^{-1/2} d^*(nt), \quad Z^n(t) := (U^n(t), V^n(t)). \quad (1.15)$$

It is an immediate consequence of [She16, Thm. 2.5] that in the case where $p_d = p_e = p_S = 0$ and $p_F \in (0, 1/2)$, the random path Z^n converges in law as $n \rightarrow \infty$ in the topology of uniform convergence on compact intervals to a two-sided two-dimensional correlated Brownian motion $Z = (U, V)$ with $Z(0) = 0$ and

$$\text{Var}(U(t)) = \text{Var}(V(t)) = \frac{1}{y+1} |t| \quad \text{and} \quad \text{Cov}(U(t), V(t)) = \frac{y-1}{2(y+1)} |t|, \quad \forall t \in \mathbb{R} \quad (1.16)$$

with y as in (1.11). In the case when $p_d = p_e = p_S = 0$ and $p_F \in [1/2, 1]$, the coordinates of Z^n instead converge in law to two identical two-sided Brownian motions with variance 1/4.

In light of Corollary 1.13 above, the above implies that if $(p_F, p_S, p_d, p_e) \in \mathcal{P}$ with $p_d = p_e = 0$ and $p_F - p_S \geq 0$ (equivalently $y \geq 1$ and $z = 1$), then Z^n converges in law as $n \rightarrow \infty$ to a Brownian motion as in (1.16) (resp. a pair of identical Brownian motions with variance 1/4) if $1 \leq y < 3$ (resp. $y \geq 3$ and $z = 1$). Our main contribution is to prove that the path Z^n converges to a correlated Brownian motion for additional values of y and z .

Theorem 1.14. *Let $\vec{p} = (p_F, p_S, p_d, p_e) \in \mathcal{P}$ with $p_F = 0$ and $p_S = 1$ (equivalently, in the notation (1.11), $y = 0$ and $z \geq 0$ is arbitrary). Then with Z^n as in (1.15), we have $Z^n \rightarrow Z$ in law in the topology of uniform convergence on compacts, where $Z = (U, V)$ is a two-sided correlated Brownian motion with $Z(0) = 0$ and*

$$\text{Var}(U(t)) = \text{Var}(V(t)) = \frac{(1+z)|t|}{2} \quad \text{and} \quad \text{Cov}(U(t), V(t)) = -\frac{z|t|}{2}, \quad \forall t \in \mathbb{R}. \quad (1.17)$$

We prove Theorem 1.14 in Section 2.

Theorem 1.15. *Let $\vec{p} = (p_F, p_S, p_d, p_e) \in \mathcal{P}$ with $p_F - p_S \leq 0$ and $p_e - p_d \leq 0$ (equivalently, with y and z as in (1.11), we have $y \in [0, 1]$ and $z \in [1, \infty)$). There is a parameter $\chi \in (1, \infty)$, depending only on y and z , such that with Z^n as in (1.15), Z^n converges in law (in the*

topology of uniform convergence on compacts) to a two-sided correlated Brownian motion $Z = (U, V)$ with $Z(0) = 0$ and

$$\begin{aligned} \text{Var}(U(t)) = \text{Var}(V(t)) &= \frac{1}{2} \left(1 + \frac{(z-y)\chi}{(y+1)(z+1)} \right) |t| \quad \text{and} \\ \text{Cov}(U(t), V(t)) &= -\frac{(z-y)\chi}{2(y+1)(z+1)} |t|, \quad \forall t \in \mathbb{R}. \end{aligned} \tag{1.18}$$

In the case when $z = 1$, we have $\chi = 2$. When $y = 0$, we have $\chi = z + 1$.

Figure 8 illustrates the range of parameter values for which Theorems 1.14 and 1.15 (and their analogues elsewhere in the literature) apply. The value of χ when $y = 0$ follows from Theorem 1.14. The value of χ when $z = 1$ will be obtained in the course of proving Theorem 1.15. It remains an open problem to compute χ in the case when $z \neq 1$ and $y \neq 0$ or to obtain any scaling limit result at all in the case when $z \in [0, 1)$ and $y > 0$ or when $z \neq 1$ and $y \geq 1$.

Theorem 1.15 combined with [DMS14, Thm. 1.13] and [GHMS17, Thm. 1.1] tells us that the infinite-volume rooted spanning-tree-decorated random planar map $(M^\infty, e_0^\infty, T^\infty)$ converges in the peanosphere sense, upon rescaling, to a γ -quantum cone decorated by an independent whole-plane space-filling SLE_κ with $\gamma = 4/\sqrt{\kappa}$ for some $\kappa \geq 8$. Furthermore, since we know the value of χ when $z = 1$, Theorem 1.15 together with [She16, Thm. 2.5] and Lemma 1.11 below imply the following.

Corollary 1.16. *Suppose $\vec{p} \in \mathcal{P}$ is such that $z = 1$ and $y \in [0, 3)$. Then Z^n converges in law to a correlated Brownian motion $Z = (U, V)$ with*

$$\text{Var}(U(t)) = \text{Var}(V(t)) = \frac{1}{y+1} |t| \quad \text{and} \quad \text{Cov}(U(t), V(t)) = \frac{y-1}{2(y+1)} |t|, \quad \forall t \in \mathbb{R}.$$

Hence the scaling limit of an infinite-volume active-tree-decorated planar map with parameter $y \in [0, 3)$ in the peanosphere sense is a γ -quantum cone decorated by an independent whole-plane space-filling SLE_κ with

$$\frac{y-1}{2} = -\cos\left(\frac{4\pi}{\kappa}\right), \quad \gamma = \frac{4}{\sqrt{\kappa}}, \quad \kappa > 4.$$

1.6 Outline

The remainder of this paper is structured as follows. In Section 2, we prove Theorem 1.14. The key observation in the proof is that if every order in X is an \mathbf{S} , then the most recently added burger which has not yet been consumed is the same as the most recently added burger. This allows us to break up the word X into i.i.d. blocks of geometric size corresponding to increments of X between the times when the type of the most recently added burger changes. Donsker's theorem applied to the change of \vec{d} over each of the blocks then concludes the proof.

The proof of Theorem 1.15, which is given in Section 3, is much more involved than that of Theorem 1.14. Section 3 is independent from Section 2.

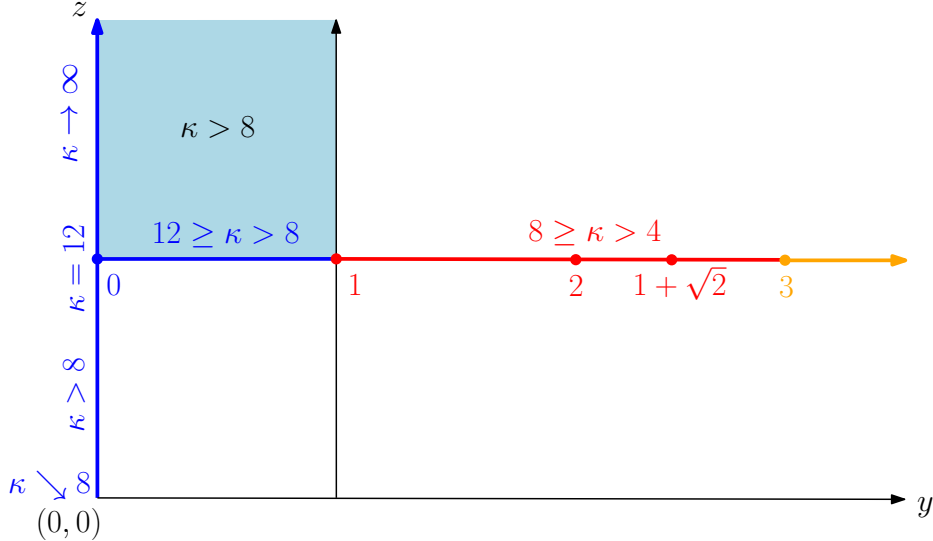


Figure 8: A graph of the range of parameter values for which peanosphere scaling limit results for spanning-tree-decorated random planar maps are known, along with the corresponding values of κ . On the red and orange segments, the path Z^n converges to a non-negatively correlated Brownian motion [She16]. On the orange segment the correlation is 1 and the maps are not conjectured to converge to SLE_κ -decorated LQG for any $\kappa > 4$. On the red segment, which corresponds to critical FK planar maps for $q \in [0, 4)$, peanosphere scaling limit results are known both in the infinite-volume and finite-volume cases [She16, GMS17a, GS17, GS15], and several additional results are known [Che17, BLR17, GM17]. The blue and light blue regions are treated in this paper, and give negatively correlated Brownian motions in the scaling limit. The blue line segments are values for which an infinite-volume peanosphere scaling limit result is known and the exact correlation of the limiting Brownian motion (equivalently the limiting values of γ and κ) is known. The horizontal blue segment corresponds to active-tree-decorated planar maps with parameter $y \in [0, 1]$ and the vertical segment corresponds to various laws on bipolar-oriented planar maps. The light blue region is the set of parameter values for which the path Z^n is known to converge to a negatively correlated Brownian motion but the exact correlation is unknown. Special parameter values are shown with dots. The case when $(y, z) = (0, 1)$ corresponds to a uniform bipolar-oriented random planar map, as studied in [KMSW15]. The case when $(y, z) = (1, 1)$ corresponds to a random planar map decorated by a uniform spanning tree. The case $(y, z) = (2, 1)$ corresponds to the uniform distribution on the underlying planar map M , and is the only case where metric scaling limit results are known. The case $(y, z) = (1 + \sqrt{2}, 1)$ corresponds to the FK Ising model.

The proof of Theorem 1.15 uses many of the same ideas as the proof of [She16, Thm. 2.5]. However, the argument used in [She16] does not suffice for our purposes. One of the key inputs in the proof [She16, Thm. 2.5] is a tail bound for the law of the length of the reduced word $|X(1, n)|$ (see [She16, Lem. 3.13]). This tail bound is deduced from the fact that changing a single symbol in the word $X_1 \cdots X_n$ changes the value of $\mathcal{D}(X(1, n))$, defined as in Definition 1.4, by at most 2 (this fact implies that a certain martingale has bounded increments and allows one to apply Azuma's inequality). When we consider words with stale

orders and/or duplicate burgers, the above Lipschitz property does not hold. For example, the reduction of the word hhhcSSS consists of a single c , but if we change the c to an h , the reduced word has length 7. We still obtain an analogue of [She16, Lem. 3.13] in the setting of Theorem 1.15 (see Proposition 3.21 below), but our proof of this result requires analogues of most of the other lemmas in [She16, § 3] as well as some additional estimates.

Section 3 is structured as follows. In Section 3.1, we prove a monotonicity result (Lemma 3.1) which says that for a general choice of p_{S} and p_{d} , the expected number of burgers and the expected number of orders in the reduced word $X(1, n)$ is greater than or equal to the corresponding expectation under the law where $p_{\text{S}} = p_{\text{d}} = p_{\text{F}} = p_{\text{e}} = 0$. Under this latter law, the process \vec{d} of Definition 1.4 is a simple random walk on \mathbb{Z}^2 . In fact, this monotonicity holds even if we condition on an event E which depends only on the one-dimensional simple random walk $i \mapsto \mathcal{C}(X(1, i))$ for $i \in [1, n]_{\mathbb{Z}}$ (Definition 1.3). The proof proceeds by way of a careful analysis of how the length of the reduction of a finite word changes when we replace the rightmost symbol among all of the S and d symbols by an element of Θ_0 .

In Section 3.2, we prove a result to the effect that the number of unidentified d 's and S 's in $X(1, n)$ is typically negligible in comparison to the number of unmatched H 's or C 's (Lemma 3.4). Since the d 's and S 's in the reduced word are the only thing which prevents the walk \vec{d} of Definition 1.4 from having independent increments, this result tells us that macroscopic increments of \vec{d} are in some sense “close” to being independent. This fact will be used frequently in the later subsections. To prove Lemma 3.4, we use the monotonicity lemma from Section 3.1 to show that the expected number of unmatched H 's added to the word between successive times that unidentified d 's and S 's are added is infinite.

In Section 3.3, we study the time J which is the smallest $j \in \mathbb{N}$ such that $X(-j, -1)$ contains an h or c . The analogue of the time J also plays a key role in [She16, GMS17a, GS17, GS15]. The importance of J in our setting is that the burger X_{-J} determines the identification of the symbol X_0 . We will prove a number of facts about J , the most important of which are Proposition 3.11 (which shows that $\chi := \mathbb{E}(|X(-J, -1)|) < \infty$) and Lemma 3.14 (which shows that the expected number of burgers and the expected number of orders in $X(-J, -1)$ are the same) and Lemma 3.16 (a uniform integrability result for $|X(-n, -1)|$ on the event $\{J > n\}$).

Section 3.4 contains the calculation which leads to the formula for the variances and covariances of the limiting Brownian motions in Theorem 1.15. This calculation is based on the results of Section 3.3 and is similar to [She16, § 3.1].

Section 3.5 shows that $\mathbb{E}(|X(1, n)|) \asymp n^{1/2}$. The upper bound follows from an analysis of the times at which burgers of a given type are added when we read the word backwards. The upper bound for the number of d 's and S 's in $X(1, n)$ from Lemma 3.4 plays an important role in the proof of this estimate since it allows us to avoid worrying about such unmatched symbols. The proof of the corresponding lower bound uses a comparison to a simple random walk on \mathbb{Z}^2 based on Lemma 3.1.

In Section 3.6, we build on the results of Section 3.5 to prove an exponential upper tail bound for $n^{-1/2}|X(1, n)|$ analogous to [She16, Lem. 3.13] (Proposition 3.21).

In Section 3.7, we use this tail bound to deduce tightness of the law of the re-scaled random walk Z^n in the local uniform topology, then conclude the proof of Theorem 1.15 by

using our upper bound for the number of **d**'s and **S**'s in $X(1, n)$ to show that any subsequential limiting law must have independent, stationary increments.

Section 4 contains some open problems related to the model studied in this paper. Appendix A proves some basic facts about the reduction operation \mathcal{R} and the bi-infinite word X .

Acknowledgements. We thank the Isaac Newton Institute in Cambridge, UK, where this work was started, for its hospitality. Part of this work was completed while E.G. was an intern with the Microsoft Research Theory group. E.G. was partially supported by the U.S. Department of Defense via an NDSEG fellowship. When this project was completed, A.K. was supported by ETH Zürich and was part of NCCR SwissMAP of the Swiss National Science Foundation. J.M. was supported by NSF grant DMS-1204894. We thank two anonymous referees for helpful comments on an earlier version of this paper.

2 Scaling limit when all orders are stale

In this section we prove Theorem 1.14, which yields the scaling limit of the law of the walk Z^n when all orders are **S**. Throughout this section we use the notation of Sections 1.3 and 1.5 with $p_F = 0$ and $p_S = 1$ and to lighten notation, we set

$$p := p_d \quad \text{and} \quad q := p_e.$$

We recall in particular the bi-infinite word X and its identification $X' = \mathcal{I}(X)$.

The idea of the proof of Theorem 1.14 is to break up the word X into independent and (almost) identically distributed blocks of random size such that, within each block, the identifications of the symbols **d**, **e**, and **S** are determined. We then apply Donsker's invariance principle to a certain random walk obtained by summing over the blocks.

Let ι_0 be the smallest $i \geq 0$ for which $X_i = \mathbf{h}$. Inductively, if $k \in \mathbb{N}$ and ι_{k-1} has been defined, let ι_k be the smallest $i \geq \iota_{k-1} + 1$ for which

$$\begin{cases} X_i \in \{\mathbf{e}, \mathbf{c}\} & k \text{ odd,} \\ X_i \in \{\mathbf{e}, \mathbf{h}\} & k \text{ even.} \end{cases}$$

(In other words, the sequence $(\iota_k)_{k \geq 0}$ is the sequence of nonnegative indices which correspond to alternation in the type of burger produced.)

Let

$$\xi_k = (\xi_k^H, \xi_k^C) := \begin{cases} (\mathcal{N}_{\mathbf{h|d|e}}(X_{\iota_{k-1}} \cdots X_{\iota_k-1}), -\mathcal{N}_{\mathbf{S}}(X_{\iota_{k-1}} \cdots X_{\iota_k-1})), & k \text{ odd} \\ (-\mathcal{N}_{\mathbf{S}}(X_{\iota_{k-1}} \cdots X_{\iota_k-1}), \mathcal{N}_{\mathbf{c|d|e}}(X_{\iota_{k-1}} \cdots X_{\iota_k-1})), & k \text{ even.} \end{cases} \quad (2.1)$$

(There are no **e** symbols in the subword $X_{\iota_{k-1}} \cdots X_{\iota_k-1}$ except possibly for $X_{\iota_{k-1}}$.) Let

$$\Xi_k = (\Xi_k^H, \Xi_k^C) := \sum_{j=1}^k \xi_j. \quad (2.2)$$

Lemma 2.1. *In the setting described just above, we have the following.*

1. For each $k \in \mathbb{N}$, we have $\vec{d}(\iota_k - 1) - \vec{d}(\iota_0 - 1) = \Xi_k$.
2. For each odd (resp. even) $k \in \mathbb{N}$, we have $\iota_k - \iota_{k-1} = \xi_k^{\text{H}} - \xi_k^{\text{C}}$ (resp. $\iota_k - \iota_{k-1} = \xi_k^{\text{C}} - \xi_k^{\text{H}}$).
3. The random variables ξ_k for $k \in \mathbb{N}$ are independent.
4. For each $k \in \mathbb{N}$, the law of $\iota_k - \iota_{k-1}$ is geometric with success probability $(1 - p + q)/4$. If k is odd (resp. even), then given $\iota_k - \iota_{k-1}$ the symbols of $X_{\iota_{k-1}+1} \cdots X_{\iota_k}$ are i.i.d., and each is a burger with probability $(1 + p - q)/(3 + p - q)$. In particular, the conditional law of $\xi_k^{\text{H}} - 1$ (resp. $\xi_k^{\text{C}} - 1$) given $\iota_k - \iota_{k-1}$ is the binomial distribution with parameters $\iota_k - \iota_{k-1} - 1$ and $(1 + p - q)/(3 + p - q)$.

Proof. Since the only orders are of type S, for any $i \in \mathbb{Z}$ the most recently added burger which hasn't yet been consumed is the same as the most recently added burger. By the definition of the times ι_* , if $\iota_{k-1} \leq i < \iota_k$, then the top burger is of type h if k is odd and of type c if k is even. For simplicity we assume throughout the rest of the proof that k is odd; the case when k is even is symmetric.

For $\iota_{k-1} \leq i < \iota_k$, if X_i is a burger, then $X'_i = \text{h}$, and if X_i is an order, then $X'_i = \text{C}$, which implies $\vec{d}(\iota_k - 1) - \vec{d}(\iota_{k-1} - 1) = \xi_k$. Summing this relation and the analogous relation in the case when k is even gives assertion 1. Since k is assumed to be odd, the total number of burgers and orders in $X_{\iota_{k-1}} \cdots X_{\iota_k}$ is $\xi_k^{\text{H}} - \xi_k^{\text{C}}$, which implies assertion 2.

Since ι_{k-1} for $k \in \mathbb{N}$ is a stopping time for the filtration generated by $X_1 \cdots X_n$ for $n \in \mathbb{N}$, the strong Markov property implies $X_{\iota_{k-1}+1} \cdots X_{\iota_k}$ is independent of $X_1 \cdots X_{\iota_{k-1}}$, which implies assertion 3.

In view of the strong Markov property (and again recalling that k is assumed to be odd), we see that $X_{\iota_{k-1}+1} \cdots X_{\iota_k}$ is a string of i.i.d. symbols terminated at the first c or e. By (1.7), the terminating symbol occurs with probability $\mathbb{P}(\text{e}) + \mathbb{P}(\text{c}) = p_e/2 + (1 - p_e - p_d)/4 = (1 - p + q)/4$, which implies the geometric law for $\iota_k - \iota_{k-1}$. Given the length of the string $X_{\iota_{k-1}+1} \cdots X_{\iota_k}$, each symbol except the last is a burger independently with probability

$$\frac{\mathbb{P}(\text{h}) + \mathbb{P}(\text{d})}{\mathbb{P}(\text{h}) + \mathbb{P}(\text{d}) + \mathbb{P}(\text{S})} = \frac{(1 + p - q)/4}{(3 + p - q)/4},$$

which finishes proving assertion 4. □

Proposition 2.2. *For odd $k \in \mathbb{N}$,*

$$\begin{aligned} \mathbb{E}(\iota_k - \iota_{k-1}) &= \frac{4}{1 - p + q}, & \text{Var}(\iota_k - \iota_{k-1}) &= \frac{4(3 + p - q)}{(1 - p + q)^2}, \\ \mathbb{E}(\xi_k^{\text{H}}) &= \frac{2}{1 - p + q}, & \text{Var}(\xi_k^{\text{H}}) &= \frac{2(1 + p - q)}{(1 - p + q)^2}, \\ \mathbb{E}(\xi_k^{\text{C}}) &= -\frac{2}{1 - p + q}, & \text{Var}(\xi_k^{\text{C}}) &= \frac{2(3 - p + q)}{(1 - p + q)^2}, \\ & & \text{Cov}(\xi_k^{\text{H}}, \xi_k^{\text{C}}) &= -\frac{2(1 + p - q)}{(1 - p + q)^2}. \end{aligned} \tag{2.3}$$

For even $k \in \mathbb{N}$, the same holds with ξ_k^{H} and ξ_k^{C} interchanged.

Proof. Let Z_i be the indicator random variable for the word $X_{\iota_{k-1}+1} \cdots X_{\iota_k-1}$ having length at least i and having a burger in position i , and let Z_i^S be the indicator variable for the word having length at least i and having an order in position i . For odd k , $\xi_k^H = 1 + \sum_{i=1}^{\infty} Z_i$ and $\xi_k^C = -\sum_{i=1}^{\infty} Z_i^S$, and vice versa for even k . Assertions 2 and 4 of Lemma 2.1 yield $\mathbb{E}[Z_i]$, $\mathbb{E}[Z_i^S]$, $\mathbb{E}[Z_i Z_j]$, $\mathbb{E}[Z_i^S Z_j^S]$, and $\mathbb{E}[Z_i Z_j^S]$, from which (2.3) follows from a short calculation. \square

Proof of Theorem 1.14. For $k \in \mathbb{N} \cup \{0\}$ let Ξ_k^H and Ξ_k^C be as in (2.2). Extend Ξ^H and Ξ^C from $\mathbb{N} \cup \{0\}$ to $[0, \infty)$ by linear interpolation. For $t \geq 0$ and $n \in \mathbb{N}$, let

$$\widehat{U}^n(t) := n^{-1/2} \Xi_{2nt}^H, \quad \widehat{V}^n(t) := n^{-1/2} \Xi_{2nt}^C, \quad \text{and} \quad \widehat{Z}^n(t) := \left(\widehat{U}^n(t), \widehat{V}^n(t) \right).$$

It follows from (2.3) that for each $k \in \mathbb{N}$,

$$\begin{aligned} \mathbb{E}(\xi_{2k-1}^H + \xi_{2k}^H) &= \mathbb{E}(\xi_{2k-1}^C + \xi_{2k}^C) = 0, \\ \text{Var}(\xi_{2k-1}^H + \xi_{2k}^H) &= \text{Var}(\xi_{2k-1}^C + \xi_{2k}^C) = \frac{8}{(1-p+q)^2}, \\ \text{Cov}(\xi_{2k-1}^H + \xi_{2k}^H, \xi_{2k-1}^C + \xi_{2k}^C) &= -\frac{4(1+p-q)}{(1-p+q)^2}. \end{aligned}$$

By Lemma 2.1, the pairs $(\xi_{2k-1}^H + \xi_{2k}^H, \xi_{2k-1}^C + \xi_{2k}^C)$ for each $k \in \mathbb{N}$ are i.i.d. By Donsker's invariance principle (see [Whi02, Thm. 4.3.5] for a statement in general dimension), \widehat{Z}^n converges in law as $n \rightarrow \infty$ in the topology of uniform convergence on compacts to a pair $\widehat{Z} = (\widehat{U}, \widehat{V})$ of correlated Brownian motions with $\widehat{Z}(0) = 0$

$$\text{Var}(\widehat{U}(t)) = \text{Var}(\widehat{V}(t)) = \frac{8t}{(1-p+q)^2} \quad \text{and} \quad \text{Cov}(\widehat{U}(t), \widehat{V}(t)) = -\frac{4(1+p-q)t}{(1-p+q)^2}, \quad \forall t \geq 0. \quad (2.4)$$

By the law of large numbers, a.s.

$$\lim_{k \rightarrow \infty} k^{-1} \iota_{[tk]} = \frac{4t}{1-p+q}, \quad \forall t \in \mathbb{Q}. \quad (2.5)$$

By the Skorokhod representation theorem, we can find a coupling of a sequence of words (X^n) , each with the same law as X , with the correlated Brownian motion \widehat{Z} such that (with \widehat{Z}^n and ι_k^n defined with respect to the word X^n) we a.s. have $\widehat{Z}^n \rightarrow \widehat{Z}$ and $k^{-1} \iota_{[tk]}^n \rightarrow 4t/(1-p+q)$ for each $t \in \mathbb{Q}$. Combining (2.5) with the fact that each coordinate of Z^n is monotone between subsequent renewal times, and the continuity of Brownian motion, we obtain that

$$(t \mapsto Z^n(t)) \xrightarrow{n \rightarrow \infty} \left(t \mapsto \widehat{Z} \left(\frac{1-p+q}{8} t \right) \right)$$

in the topology of uniform convergence on compacts of $[0, \infty)$. By (2.4), $t \mapsto \widehat{Z} \left(\frac{1-p+q}{8} t \right)$ is a Brownian motion with variances and covariances as in (1.17). We thus obtain $Z^n|_{[0, \infty)} \rightarrow Z|_{[0, \infty)}$ in law, with Z as in the theorem statement. Since the law of the bi-infinite word X is translation invariant, we also have that $(Z^n - Z^n(s_0))|_{[s_0, \infty)} \rightarrow (Z - Z(s_0))|_{[s_0, \infty)}$ in law for each $s_0 \in \mathbb{R}$. Since $Z^n(0) = Z(0) = 0$ for each $n \in \mathbb{N}$, for $s_0 < 0$ and $t \geq s_0$,

$$Z^n(t) = (Z^n(t) - Z^n(s_0)) - (Z^n(0) - Z^n(s_0))$$

and the analogous relation holds for Z . From this we infer that $Z^n|_{[s_0, \infty)} \rightarrow Z|_{[s_0, \infty)}$ in law for each $s_0 \in \mathbb{R}$. Since s_0 can be made arbitrarily negative, we infer that $Z^n \rightarrow Z$ in law in the topology of uniform convergence on compact subsets of \mathbb{R} . \square

3 Scaling limit with stale orders and duplicate burgers

In this section we prove Theorem 1.15. Since the paths Z^n are deterministic functions of the identified word $X' = \mathcal{I}(X)$, Corollary 1.13 implies that we only need to prove Theorem 1.15 in the case when $p_F = p_e = 0$.

Throughout this section, we fix $p \in [0, 1)$ and $q \in [0, 1)$ and let $\mathbb{P}^{p,q}$ denote the law of the bi-infinite word X whose symbols are i.i.d. samples from the law (1.7) with $p_F = p_e = 0$, $p_S = p$, and $p_d = q$. Let $\mathbb{E}^{p,q}$ denote the corresponding expectation. When there is no ambiguity (i.e. only one pair (p, q) is under consideration) we write $\mathbb{P} = \mathbb{P}^{p,q}$ and $\mathbb{E} = \mathbb{E}^{p,q}$.

Since we think of p_S and p_d as being fixed, we abuse notation and allow “constants” to depend on p_S and p_d , including the implicit constants in asymptotic notation.

3.1 Comparison of expected lengths of reduced words

The following lemma is one of our main tools for estimating expectations of quantities associated with the word X .

Lemma 3.1. *Suppose we are in the setting described at the beginning of this section. Let $n \in \mathbb{N}$ and let E be an event which is measurable with respect to the σ -algebra generated by $\{\mathcal{C}(X(1, i))\}_{i \in [1, n]_{\mathbb{Z}}}$ (where here \mathcal{C} is as in Definition 1.4, i.e., $\mathcal{C} = \mathcal{B} - \mathcal{O}$). For each $n \in \mathbb{N}$ and $(p, q) \in [0, 1] \times [0, 1)$, we have (in the notation of Definition 1.4)*

$$\mathbb{E}^{p,q}(\mathcal{B}(X(1, n))\mathbb{1}_E) \geq \mathbb{E}^{0,0}(\mathcal{B}(X(1, n))\mathbb{1}_E) \quad (3.1)$$

and

$$\mathbb{E}^{p,q}(\mathcal{O}(X(1, n))\mathbb{1}_E) \geq \mathbb{E}^{0,0}(\mathcal{O}(X(1, n))\mathbb{1}_E). \quad (3.2)$$

The intuitive reason why we expect Lemma 3.1 to be true is that it is “harder” for a d or S to find a match than it is for an element of Θ_0 to find a match, since the d or S has to be identified, then matched. So, replacing d ’s and S ’s by elements of Θ_0 should tend to reduce the number of burgers and orders in the word.

To prove the lemma, we will iteratively replace the rightmost symbol amongst all of the d ’s and S ’s in $X_1 \dots X_n$ by an h or c with equal probability (if it is a d) or by an H or C with equal probability (if it is an S) and argue that each of these replacements reduces the expected number of burgers and orders in $X(1, n)$. The key tool in the proof is Lemma 3.3 below.

Remark 3.2. The proof of Lemma 3.1 is based on an argument of Linxiao Chen which appears in the proof of Lemma 5 in the original arXiv version of [Che17]. Chen’s argument does not in fact yield the stochastic domination statement claimed in his Lemma 5, but does prove the analogue of Lemma 3.1 in the setting where $p_S = p_d = p_e = 0$ and $p_F \in [0, 1]$.

Define an involution $\theta \mapsto \theta^\dagger$ on Θ by

$$\begin{aligned} \mathfrak{h}^\dagger &= \mathfrak{c} & \mathfrak{c}^\dagger &= \mathfrak{h} & \mathfrak{H}^\dagger &= \mathfrak{C} & \mathfrak{C}^\dagger &= \mathfrak{H} \\ \mathfrak{d}^\dagger &= \mathfrak{d} & \mathfrak{e}^\dagger &= \mathfrak{e} & \mathfrak{F}^\dagger &= \mathfrak{F} & \mathfrak{S}^\dagger &= \mathfrak{S}. \end{aligned} \tag{3.3}$$

For a word $x = x_1 \cdots x_{|x|}$ consisting of elements of Θ , we write $x^\dagger = x_1^\dagger \cdots x_{|x|}^\dagger$.

For such a word x , we write $s(x)$ for the index of the rightmost symbol among all of the \mathfrak{d} or \mathfrak{S} symbols in x (or $s(x) = 0$ if no such x exists). We define

$$x^{\text{Hc}} = \begin{cases} \text{word obtained from } x \text{ by replacing } x_{s(x)} \text{ with } \mathfrak{H} & s(x) > 0 \text{ and } x_{s(x)} = \mathfrak{S} \\ \text{word obtained from } x \text{ by replacing } x_{s(x)} \text{ with } \mathfrak{c} & s(x) > 0 \text{ and } x_{s(x)} = \mathfrak{d} \\ x & s(x) = 0, \end{cases}$$

and we define x^{Ch} similarly but with \mathfrak{C} and \mathfrak{h} in place of \mathfrak{H} and \mathfrak{c} .

We write $r(x)$ for the largest $k \in [1, s(x) - 1]_{\mathbb{Z}}$ for which $x_k = \mathfrak{h}$ or $x_k = \mathfrak{c}$ and x_k has no match in $x_1 \cdots x_{s(x)-1}$ (or $r(x) = 0$ if no such k exists). We define an involution

$$\Psi(x) = \begin{cases} (x_1 \cdots x_{r(x)-1})^\dagger x_{r(x)} \cdots x_{s(x)} (x_{s(x)+1} \cdots x_{|x|})^\dagger & r(x) > 0 \\ x_1 \cdots x_{s(x)} (x_{s(x)+1} \cdots x_{|x|})^\dagger & r(x) = 0. \end{cases}$$

We make the following elementary observations about the above operations.

1. Involution commutes with reduction, i.e. $\mathcal{R}(x^\dagger) = \mathcal{R}(x)^\dagger$ for all words x .
2. $s(\Psi(x)) = s(x)$ and $r(\Psi(x)) = r(x)$ for all words x , and hence $\Psi(\Psi(x)) = x$.

Lemma 3.3. *Let $x = x_1 \cdots x_{|x|}$ be a word consisting of elements of $\Theta_0 \cup \{\mathfrak{d}, \mathfrak{S}\}$. If*

$$\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) > \mathcal{B}(\mathcal{R}(x)) \tag{3.4}$$

then

$$\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) = \mathcal{B}(\mathcal{R}(x)) + 1 \quad \text{and} \quad \mathcal{B}(\mathcal{R}(\Psi(x)^{\text{Hc}})) = \mathcal{B}(\mathcal{R}(\Psi(x))) - 1. \tag{3.5}$$

To prove Lemma 3.3, we first explain why (3.4) implies that $x_{s(x)}$ is identified in x (i.e., $r(x) > 0$) and that the word $\mathcal{R}(x_{r(x)} \cdots x_{s(x)-1})$ must take the form $\mathfrak{C}^n \mathfrak{h}^m$ for some $m \geq 1$ and $n \geq 0$ (where here \mathfrak{C}^n denotes the word which is a concatenation of n \mathfrak{C} 's, etc.); see (3.6). By means of (3.7), we then reduce to the case when $\mathcal{R}(x_1 \cdots x_{r(x)-1})$ (resp. $\mathcal{R}(x_{s(x)} \cdots x_{|x|})$) contains only \mathfrak{h} 's and \mathfrak{c} 's (resp. \mathfrak{H} 's and \mathfrak{C} 's). This reduction together with (3.6) will allow us to write down explicit expressions for the quantities in (3.5) in terms of n and m . Comparing these expressions will yield (3.5).

Proof of Lemma 3.3. If (3.4) holds, then $x^{\text{Hc}} \neq x$, so the word x contains at least one \mathfrak{d} or \mathfrak{S} . Since $x_{s(x)}$ is the rightmost \mathfrak{d} or \mathfrak{S} in the word x , replacing $x_{s(x)}$ by \mathfrak{c} or \mathfrak{H} does not change the identification of any symbol in $x_{s(x)+1} \cdots x_{|x|}$.

We first argue that (3.4) implies that $x_{s(x)}$ is identified in x , and hence that $r(x) > 0$. Indeed, suppose $x_{s(x)}$ is not identified in the word x . Then the reduced word $x(1, s(x) - 1)$ contains no \mathfrak{h} or \mathfrak{c} symbols, since the presence of any such symbol would identify $x_{s(x)}$

(recall (1.6)). In this case, the reduced words $\mathcal{R}(x)$ and $\mathcal{R}(x^{\text{Hc}})$ would have the same set of symbols except the symbol coming from position $s(x)$, and possibly an order in $x_{s(x)+1} \cdots x_{|x|}$ which may consume $x_{s(x)}^{\text{Hc}}$ if it is a burger. But then $\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) \leq \mathcal{B}(\mathcal{R}(x))$, contradicting (3.4).

Henceforth assume that (3.4) holds, which implies (by the preceding paragraph) that $x_{s(x)}$ is identified in x . If $r(x) < k < s(x)$ and x_k is a burger, then by definition of $r(x)$, either $x_k \in \{\mathbf{h}, \mathbf{c}\}$ but is consumed by an order in $x_{r(x)} \cdots x_{s(x)}$, or else $x_k = \mathbf{d}$. If $x_k = \mathbf{d}$ and is identified as $x_{r(x)}^\dagger$, consider the first such k . By definition of $r(x)$, the burger that identifies x_k is consumed in $x_{r(x)} \cdots x_{s(x)}$, at a time by which x_k must therefore also have been consumed. Thus the only burgers in $\mathcal{R}(x_{r(x)} \cdots x_{s(x)})$ are \mathbf{d} 's that are identified as $x_{r(x)}$ and the burger $x_{r(x)}$ itself.

If $x_{s(x)} = \mathbf{S}$, then since $\mathcal{R}(x^{\text{Hc}}) \neq \mathcal{R}(x)$, it must be that $x_{s(x)}$ corresponds to a \mathbf{C} symbol in the identification $\mathcal{I}(x)$, which in turn implies $x_{r(x)} = \mathbf{h}$. If on the other hand $x_{s(x)} = \mathbf{d}$, then $x_{s(x)}$ must be identified by an \mathbf{h} in the word x , which again implies $x_{r(x)} = \mathbf{h}$, since from the previous paragraph we know that all potential intermediate burgers would be \mathbf{d} 's. Since the burger $x_{r(x)} = \mathbf{h}$ is not consumed in $\mathcal{R}(x_{r(x)} \cdots x_{s(x)})$, any order in this reduced word is identified and must be of type \mathbf{C} . Regardless of $x_{s(x)}$,

$$\mathcal{R}(x_{r(x)} \cdots x_{s(x)-1}) = \mathbf{C}^n \mathbf{h}^m \quad \text{with } m \geq 1 \text{ and } n \geq 0. \quad (3.6)$$

Write $x(1, r(x) - 1) = Uu$ and $x(s(x) + 1, |x|) = Vv$, where U and V are words consisting of only orders and \mathbf{d} 's, and u and v are words consisting of only \mathbf{h} 's and \mathbf{c} 's. By definition of $s(x)$, V contains no \mathbf{S} or \mathbf{d} . Let α denote the identification of $x_{s(x)}$ in $x_{r(x)} \cdots x_{s(x)}$, which is either \mathbf{C} or \mathbf{h} . By the relation $\mathcal{R}(\mathcal{R}(x)\mathcal{R}(y)) = \mathcal{R}(xy)$ (Lemma A.2) and the commutativity of \mathbf{h} with \mathbf{C} ,

$$\begin{aligned} \mathcal{R}(x) &= U\mathcal{R}(uh^m\mathbf{C}^n\alpha V)v, & \mathcal{R}(x^{\text{Hc}}) &= U\mathcal{R}(uh^m\mathbf{C}^n\alpha^\dagger V)v, \\ \mathcal{R}(\Psi(x)) &= U^\dagger\mathcal{R}(u^\dagger\mathbf{h}^m\mathbf{C}^n\alpha V^\dagger)v^\dagger, & \mathcal{R}(\Psi(x)^{\text{Hc}}) &= U^\dagger\mathcal{R}(u^\dagger\mathbf{h}^m\mathbf{C}^n\alpha^\dagger V^\dagger)v^\dagger. \end{aligned} \quad (3.7)$$

From (3.7) we see that changing U and v while leaving the other words fixed does not change $\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) - \mathcal{B}(\mathcal{R}(x))$ or $\mathcal{B}(\mathcal{R}(\Psi(x)^{\text{Hc}})) - \mathcal{B}(\mathcal{R}(\Psi(x)))$, so we assume without loss of generality that $U = v = \emptyset$.

Under this assumption, the words $\mathcal{R}(x)$ and $\mathcal{R}(\Psi(x))$ both take the form $\mathcal{R}(yY)$, where y is a word with only hamburgers and cheeseburgers and Y is a word with only hamburger orders and cheeseburger orders. If α is an order, then $\mathcal{R}(x^{\text{Hc}})$ and $\mathcal{R}(\Psi(x)^{\text{Hc}})$ also take the form $\mathcal{R}(yY)$, but if α is a burger and $n > 0$, then $\mathcal{R}(x^{\text{Hc}})$ and $\mathcal{R}(\Psi(x)^{\text{Hc}})$ take the form $\mathcal{R}(y\mathbf{C}^n\mathbf{c}Y)$ (where in both cases, as above, y denotes a word with only \mathbf{c} 's and \mathbf{h} 's, and Y a word with only \mathbf{C} 's and \mathbf{H} 's).

For convenience we define

$$\begin{aligned} \Delta_{\mathbf{h}} &:= \mathcal{N}_{\mathbf{h}}(u) - \mathcal{N}_{\mathbf{H}}(V) \\ \Delta_{\mathbf{c}} &:= \mathcal{N}_{\mathbf{c}}(u) - \mathcal{N}_{\mathbf{C}}(V). \end{aligned}$$

Suppose first $\alpha = \mathbf{C}$. From (3.7) we see

$$\begin{aligned} \mathcal{B}(\mathcal{R}(x)) &= (\Delta_{\mathbf{h}} + m) \vee 0 + (\Delta_{\mathbf{c}} - n - 1) \vee 0 \\ \mathcal{B}(\mathcal{R}(x^{\text{Hc}})) &= (\Delta_{\mathbf{h}} + m - 1) \vee 0 + (\Delta_{\mathbf{c}} - n) \vee 0 \end{aligned}$$

Since $\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) > \mathcal{B}(\mathcal{R}(x))$ it follows that $\Delta_{\text{h}} \leq -m$ and $\Delta_{\text{c}} \geq n+1$, and hence $\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) = \mathcal{B}(\mathcal{R}(x)) + 1$, as claimed. From (3.7) together with $\Delta_{\text{h}} \leq -1$ and $\Delta_{\text{c}} \geq 1$, we see

$$\begin{aligned}\mathcal{B}(\mathcal{R}(\Psi(x))) &= (\Delta_{\text{c}} + m) \vee 0 + (\Delta_{\text{h}} - n - 1) \vee 0 = (\Delta_{\text{c}} + m) + 0 \\ \mathcal{B}(\mathcal{R}(\Psi(x^{\text{Hc}}))) &= (\Delta_{\text{c}} + m - 1) \vee 0 + (\Delta_{\text{h}} - n) \vee 0 = (\Delta_{\text{c}} + m - 1) + 0\end{aligned}$$

so $\mathcal{B}(\mathcal{R}(\Psi(x^{\text{Hc}}))) = \mathcal{B}(\mathcal{R}(\Psi(x))) - 1$, as claimed.

Suppose next $\alpha = \text{h}$. From (3.7) we see

$$\begin{aligned}\mathcal{B}(\mathcal{R}(x)) &= (\Delta_{\text{h}} + m + 1) \vee 0 + (\Delta_{\text{c}} - n) \vee 0 \\ \mathcal{B}(\mathcal{R}(x^{\text{Hc}})) &= (\Delta_{\text{h}} + m) \vee 0 + [(\mathcal{N}_{\text{c}}(u) - n) \vee 0 + 1 - \mathcal{N}_{\text{c}}(V)] \vee 0\end{aligned}$$

The nested- \vee expression arises because $\mathcal{R}(x^{\text{Hc}})$ takes the form $\mathcal{R}(y\text{C}^n\text{c}Y)$. Since $\Delta_{\text{c}} = \mathcal{N}_{\text{c}}(u) - \mathcal{N}_{\text{c}}(V)$ and $\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) > \mathcal{B}(\mathcal{R}(x))$, it follows (by a short argument by contradiction due to the nested- \vee expression) that $\Delta_{\text{h}} \leq -m - 1$ and

$$(\mathcal{N}_{\text{c}}(u) - n) \vee 0 \geq \mathcal{N}_{\text{c}}(V),$$

which in turn implies either $\mathcal{N}_{\text{c}}(V) = 0$ or $\Delta_{\text{c}} \geq n$. In either case, $\Delta_{\text{c}} \geq 0$. We also see $\mathcal{B}(\mathcal{R}(x^{\text{Hc}})) = \mathcal{B}(\mathcal{R}(x)) + 1$, as claimed. Referring to (3.7) again, and using from above that $\Delta_{\text{h}} \leq -2$ and $\Delta_{\text{c}} \geq 0$, we see

$$\begin{aligned}\mathcal{B}(\mathcal{R}(\Psi(x))) &= (\Delta_{\text{c}} + m + 1) \vee 0 + (\Delta_{\text{h}} - n) \vee 0 = (\Delta_{\text{c}} + m + 1) + 0 \\ \mathcal{B}(\mathcal{R}(\Psi(x^{\text{Hc}}))) &= (\Delta_{\text{c}} + m) \vee 0 + [(\mathcal{N}_{\text{h}}(u) - n) \vee 0 + 1 - \mathcal{N}_{\text{h}}(V)] \vee 0 \\ &= (\Delta_{\text{c}} + m) + [((\Delta_{\text{h}} - n) \vee (-\mathcal{N}_{\text{h}}(V))) + 1] \vee 0\end{aligned}$$

Since $\mathcal{N}_{\text{h}}(u) - \mathcal{N}_{\text{h}}(V) = \Delta_{\text{h}} \leq -m - 1 \leq -2$, it follows that $\mathcal{N}_{\text{h}}(V) \geq 2$, and so

$$\mathcal{B}(\mathcal{R}(\Psi(x^{\text{Hc}}))) = \Delta_{\text{c}} + m,$$

so in this case as well $\mathcal{B}(\mathcal{R}(\Psi(x^{\text{Hc}}))) = \mathcal{B}(\mathcal{R}(\Psi(x))) - 1$, as claimed. \square

Proof of Lemma 3.1. The law of $n \mapsto \mathcal{C}(X(1, n))$ is that of one-dimensional simple random walk, regardless of p and q . Therefore $\mathbb{P}^{p,q}(E) = \mathbb{P}^{0,0}(E)$, so to prove (3.1) it suffices to show

$$\mathbb{E}^{p,q}(\mathcal{B}(X(1, n)) | E) \geq \mathbb{E}^{0,0}(\mathcal{B}(X(1, n)) | E), \quad \forall n \in \mathbb{N}. \quad (3.8)$$

To this end, let $X^0 = X_1^0 \cdots X_n^0$ be a word whose law is that of $X_1 \cdots X_n$ under $\mathbb{P}^{p,q}$. Let $\{\xi_k\}_{k \in [1, n]_{\mathbb{Z}}}$ be i.i.d. Bernoulli random variables with parameter $1/2$, independent from X^0 . For $k \in [1, n]_{\mathbb{Z}}$ inductively define

$$X^k = \begin{cases} (X^{k-1})^{\text{Hc}} & \text{if } \xi_k = 0 \\ (X^{k-1})^{\text{Ch}} & \text{if } \xi_k = 1. \end{cases}$$

Since $\mathcal{N}_{\text{d}|S}(X^k) = 0 \vee (\mathcal{N}_{\text{d}|S}(X^{k-1}) - 1)$, and the word X^n is obtained from X^0 by replacing each d symbol in X^0 with an independent random symbol which is uniformly distributed

on $\{\mathfrak{h}, \mathfrak{c}\}$ and each \mathbf{S} symbol in X with an independent random symbol which is uniformly distributed on $\{\mathbf{H}, \mathbf{C}\}$, the law of X^n is that of $X_1 \cdots X_n$ under $\mathbb{P}^{0,0}$.

We next argue that

$$\Psi(X^k) \stackrel{d}{=} X^k \quad \forall k \in [0, n]_{\mathbb{Z}}. \quad (3.9)$$

To see this, let $k \in [1, n]_{\mathbb{Z}}$ and let j_k be the k^{th} largest $j \in [1, n]_{\mathbb{Z}}$ for which $X_j^0 \in \{\mathfrak{d}, \mathbf{S}\}$, or $j_k = 0$ if no such j exists. Also let j'_k be the largest $j \in [1, j_k - 1]_{\mathbb{Z}}$ for which the reduced word $X^0(j, j_k - 1)$ contains an \mathfrak{h} or \mathfrak{c} , or $j'_k = 0$ if no such j exists. Then j_k and j'_k are stopping times for the filtration generated by X^0 , read from right to left. By the strong Markov property, the conditional law of $X_1^0 \cdots X_{j'_k-1}^0$ given $X_{j'_k}^0 \cdots X_n^0$ is a string of $(j'_k - 1) \vee 0$ i.i.d. symbols sampled from the law $\mathbb{P}^{p,q}$. Hence given j'_k , $X_1^0 \cdots X_{j'_k-1}^0$ is conditionally independent from $X_{j_k+1}^0 \cdots X_n^0$ and $X_{j'_k}^0 \cdots X_{j_k}^0$.

By the above description of the conditional law of $X_1^0 \cdots X_{j'_k-1}^0$ given j'_k and $X_{j'_k}^0 \cdots X_n^0$ and the symmetry between hamburgers and cheeseburgers, we infer that this conditional law is invariant under involution. Since the definition of j_k is invariant under involution, we infer that also the conditional law of $X_{j_k+1}^0 \cdots X_n^0$ given j_k is invariant under involution. Since j_k is a stopping time for X^0 , read backwards, it follows that the joint conditional law of $X_1^0 \cdots X_{j'_k-1}^0$ and $X_{j_k+1}^0 \cdots X_n^0$ given j_k, j'_k and $X_{j'_k}^0 \cdots X_{j_k}^0$ is invariant under involution. In particular,

$$X^0 \stackrel{d}{=} \left(X_1^0 \cdots X_{j'_k-1}^0 \right)^\dagger X_{j'_k}^0 \cdots X_{j_k}^0 \left(X_{j_k+1}^0 \cdots X_n^0 \right)^\dagger. \quad (3.10)$$

The word X^k (resp. $\Psi(X^k)$) is obtained from the word on the left (resp. right) side of (3.10) by replacing its k rightmost \mathfrak{d} or \mathbf{S} symbols with independent random symbols sampled uniformly from $\{\mathfrak{h}, \mathfrak{c}\}$ or $\{\mathbf{H}, \mathbf{C}\}$ respectively. We thus obtain (3.9).

Now let E be an event as in the statement of the lemma, defined with the word $X_1^0 \cdots X_n^0$ in place of the word $X_1 \cdots X_n$. The operations $x \mapsto x^{\mathfrak{Hc}}$, $x \mapsto x^{\mathfrak{Ch}}$, and $x \mapsto \Psi(x)$ replace burgers with burgers and orders with orders in the word x , so the sequence $\mathcal{C}(x(1, i))_{i=1, \dots, n}$ is the same for each $x \in \{X^k, \Psi(X^k)\}$ and $k \in [0, n]_{\mathbb{Z}}$. Since the event E is determined by $\mathcal{C}(X^0(1, i))_{i=1, \dots, n}$, we see that the definition of E is unaffected if we replace X^0 with X^k or $\Psi(X^k)$ for any $k \in [0, n]_{\mathbb{Z}}$. From this observation, we deduce the following:

1. The conditional law of X^0 given E is the same as the conditional law of $X_1 \cdots X_n$ given E under $\mathbb{P}^{p,q}$.
2. The conditional law of X^n given E is the same as the conditional law of $X_1 \cdots X_n$ given E under $\mathbb{P}^{0,0}$.
3. E is independent from the Bernoulli random variables $\{\xi_k\}_{k \in [1, n]_{\mathbb{Z}}}$.
4. By (3.9), for each $k \in [1, n]_{\mathbb{Z}}$, the conditional laws of X^k and $\Psi(X^k)$ given E agree.

By combining these observations with Lemma 3.3, we find that for each $k \in [1, n]_{\mathbb{Z}}$,

$$\begin{aligned}
& \mathbb{P}(\mathcal{B}(\mathcal{R}(X^k)) > \mathcal{B}(\mathcal{R}(X^{k-1})) \mid E) \\
&= \frac{1}{2} \mathbb{P}(\mathcal{B}(\mathcal{R}((X^{k-1})^{\text{Hc}})) > \mathcal{B}(\mathcal{R}(X^{k-1})) \mid E) + \frac{1}{2} \mathbb{P}(\mathcal{B}(\mathcal{R}((X^{k-1})^{\text{Ch}})) > \mathcal{B}(\mathcal{R}(X^{k-1})) \mid E) \\
&\leq \frac{1}{2} \mathbb{P}(\mathcal{B}(\mathcal{R}(\Psi(X^{k-1})^{\text{Hc}})) = \mathcal{B}(\mathcal{R}(\Psi(X^{k-1}))) - 1 \mid E) \\
&\quad + \frac{1}{2} \mathbb{P}(\mathcal{B}(\mathcal{R}(\Psi(X^{k-1})^{\text{Ch}})) = \mathcal{B}(\mathcal{R}(\Psi(X^{k-1}))) - 1 \mid E) \\
&= \mathbb{P}(\mathcal{B}(\mathcal{R}(X^k)) = \mathcal{B}(\mathcal{R}(X^{k-1})) - 1 \mid E). \tag{3.11}
\end{aligned}$$

We used observation 3 above in the first equality and observation 4 in the last equality. Lemma 3.3 implies that $\mathcal{B}(\mathcal{R}(X^k)) = \mathcal{B}(\mathcal{R}(X^{k-1})) + 1$ whenever $\mathcal{B}(\mathcal{R}(X^k)) > \mathcal{B}(\mathcal{R}(X^{k-1}))$, so (3.11) implies

$$\begin{aligned}
& \mathbb{E}(\mathcal{B}(\mathcal{R}(X^k)) - \mathcal{B}(\mathcal{R}(X^{k-1})) \mid E) \\
&\leq \mathbb{P}(\mathcal{B}(\mathcal{R}(X^k)) > \mathcal{B}(\mathcal{R}(X^{k-1})) \mid E) - \mathbb{P}(\mathcal{B}(\mathcal{R}(X^k)) = \mathcal{B}(\mathcal{R}(X^{k-1})) - 1 \mid E) \leq 0,
\end{aligned}$$

whence

$$\mathbb{E}(\mathcal{B}(\mathcal{R}(X^k)) \mid E) \leq \mathbb{E}(\mathcal{B}(\mathcal{R}(X^{k-1})) \mid E) \quad \forall k \in [1, n]_{\mathbb{Z}}.$$

Therefore

$$\mathbb{E}(\mathcal{B}(\mathcal{R}(X^n)) \mid E) \leq \mathbb{E}(\mathcal{B}(\mathcal{R}(X^0)) \mid E). \tag{3.12}$$

By observations 1 and 2 above, we obtain (3.1).

The bound (3.2) follows observations 1 and 2 above, (3.12) and

$$\mathcal{B}(\mathcal{R}(X^0)) - \mathcal{O}(\mathcal{R}(X^0)) = \mathcal{B}(\mathcal{R}(X^n)) - \mathcal{O}(\mathcal{R}(X^n)). \quad \square$$

3.2 Bound on the number of unidentified symbols

In the next three subsections we prove analogues of various results found in [She16, § 3] in the setting of Theorem 1.15. Throughout, we assume we are in the setting described just above the statement of Theorem 1.15 for fixed $(p, q) \in [0, 1] \times [0, 1)$.

The main purpose of this section is to prove the following more quantitative analogue of [She16, Lem. 3.7].

Lemma 3.4. *For each $\varepsilon > 0$, there are positive numbers $c_0, c_1 > 0$ such that, for each $n \in \mathbb{N}$ and $A > 0$, the event*

$$F_n(\varepsilon, A) := \left\{ \frac{\mathcal{N}_{\text{d|S}}(X(1, n))}{\mathcal{N}_{\text{H}}(X(1, n)) \vee A} \geq \varepsilon \right\} \tag{3.13}$$

occurs with probability

$$\mathbb{P}(F_n(\varepsilon, A)) \leq c_0 e^{-c_1 A}. \tag{3.14}$$

Lemma 3.4 will be an important tool in what follows since it allows us in many cases to ignore the (potentially quite complicated) manner in which the d 's and S 's are identified. When we apply the lemma, we will typically take ε to be a small fixed parameter and A to be a small positive power of n (so that $\mathbb{P}(F_n(\varepsilon, A))$ decays faster than any negative power of n). We expect that an even stronger statement than Lemma 3.4 is true, namely, that

$\mathcal{N}_{d|S}(X(1, \infty)) < \infty$ a.s. and that $\mathcal{N}_{d|S}(X(1, \infty))$ is stochastically dominated by a geometric distribution. The reason for this is explained in Remark 3.7.

To prove Lemma 3.4, we first observe that if $i \in [1, n]_{\mathbb{Z}}$ is such that X_i is a d or S which is not identified in $X_1 \dots X_n$, then the word $X(1, i-1)$ must contain no hamburgers or cheeseburgers (such a hamburger or cheeseburger would identify X_i). We will prove that the expected number of unmatched H 's added to the word between the successive times when $X(1, j)$ contains no burgers is infinite (Lemma 3.6). By Hoeffding's inequality and the fact that the increments of the word X between these successive times are i.i.d., this will tell us that the number of d 's and S 's in $X(1, n)$ is typically negligible compared to the number of H 's.

To start off, we consider the time

$$K = \min\{i \in \mathbb{N} : \mathcal{C}(X(1, i)) = -1\} \quad (3.15)$$

(here $i \mapsto \mathcal{C}(X(1, i))$ is the simple random walk as in Definition 1.4).

Lemma 3.5. *We have*

$$\mathbb{E}(|X(1, K)|) = \infty. \quad (3.16)$$

Furthermore, if we let P be the smallest $j \in \mathbb{N}$ for which $\mathcal{C}(X(-j, -1)) = 1$, then

$$\mathbb{E}(|X(-P, -1)|) = \infty. \quad (3.17)$$

Proof. For each $n \in \mathbb{N}$, the event $\{K = n\}$ depends only on $\mathcal{C}(X(1, i))$ for $i \in [1, n]_{\mathbb{Z}}$. By Lemma 3.1, we find

$$\mathbb{E}(|X(1, K)| \times \mathbb{1}_{(K=n)}) \geq \mathbb{E}^{0,0}(|X(1, K)| \times \mathbb{1}_{(K=n)}), \quad \forall n \in \mathbb{N}$$

where here $\mathbb{E}^{0,0}$ denotes the law of X with $p = q = 0$. By summing over all n , we obtain

$$\mathbb{E}(|X(1, K)|) \geq \mathbb{E}^{0,0}(|X(1, K)|). \quad (3.18)$$

By standard estimates for one-dimensional simple random walk, $\mathbb{P}^{0,0}(K = n) \asymp n^{-3/2}$. Under $\mathbb{P}^{0,0}$, if we condition on $\{K = n\}$, then the conditional law of the walk $\vec{d} = (d, d^*)$ restricted to $[0, n]_{\mathbb{Z}}$ is that of a two-dimensional simple random walk conditioned to first exit the diagonal half plane $\{x + y \geq 0\}$ at time n . With uniformly positive probability under this conditioning, it holds that

$$d(n) - \inf_{i \in [0, n]_{\mathbb{Z}}} d(i) \geq n^{1/2},$$

in which case $\mathcal{N}_h(X(1, n)) \geq n^{1/2}$. Therefore,

$$\mathbb{E}^{0,0}(|X(1, K)| \times \mathbb{1}_{(K=n)}) \succeq n^{-1}.$$

By summing over all $n \in \mathbb{N}$ we obtain $\mathbb{E}^{0,0}(\mathcal{B}(X(1, K))) = \infty$ and hence (3.16).

We similarly obtain (3.17). □

Lemma 3.6. *Let I_1 be the smallest $i \in \mathbb{N}$ for which $X(1, i)$ contains no hamburgers or cheeseburgers. Then $\mathbb{E}(\mathcal{N}_H(X(1, I_1))) = \infty$ (here we take $\mathcal{N}_H(X(1, I_1)) = \infty$ if $I_1 = \infty$).*

Remark 3.7. It is possible that $I_1 = \infty$ with positive probability. In fact, we expect (but do not prove) that this is the case since the coordinates of the re-scaled walk Z^n in (1.15) should be close to attaining a simultaneous running infimum at time I_1 ; and the coordinates of the negatively correlated Brownian motion Z in Theorem 1.15 a.s. do not have any simultaneous running infima (this follows by applying a linear transformation and using that an uncorrelated two-dimensional Brownian motion a.s. has no θ -cone times for $\theta < \pi/2$ [Shi85, Eva85]). Note that if $I_1 = \infty$ with positive probability, then a.s. there are only finitely many times in \mathbb{N} for which $X(1, i)$ contains no h's or c's, and hence only finitely many unidentified d's and S's in $X(1, \infty)$. We note, by way of comparison, that in the setting when $p_S = p_d = p_e = 0$ and $p_F \in (0, 1]$, the word $X(1, \infty)$ a.s. contains infinitely many F's; see [She16, Lemma 3.7] in the case $p_F > 1/2$ and [GMS17a, Proposition 3.5] in the case $p_F < 1/2$ (the same proof works for $p_F = 1/2$).

Proof of Lemma 3.6. The statement of the lemma is obvious if $I_1 = \infty$ with positive probability, so we can assume that $I_1 < \infty$ a.s. If $I_1 > 1$ and $X(1, I_1)$ contains a d or S symbol, then $X(1, i)$ would have to contain no hamburgers or cheeseburgers for some $i \leq I_1 - 1$ (corresponding to the index of the d or S in question), which contradicts the definition of I_1 . Thus either $I_1 = 1$ or the word $X(1, I_1)$ contains no unidentified d's or S's.

If $I_1 > 1$, since every burger in $X(1, I_1)$ is identified, by definition of I_1 , it must be that $X(1, I_1)$ contains no burgers. Thus if $I_1 > 1$, the word $X(2, I_1)$ contains more orders than burgers.

Now let K_2 be the smallest $i \geq 2$ for which $\mathcal{C}(X(2, i)) \leq -1$. Then $X_2 \cdots X_{K_2}$ is independent from X_1 and agrees in law with $X_1 \cdots X_K$. On the event $\{X_1 = \mathbf{h}\}$, we have $I_1 \geq K_2$. Therefore, every order appearing in $X(2, K_2)$ except possibly one also appears in $X(1, I_1)$. It follows that

$$\mathbb{E}(\mathcal{N}_{\mathbf{H}|\mathbf{C}}(X(1, I_1))) \geq \frac{1-q}{4} \mathbb{E}(|X(1, K)| - 1) = \infty.$$

By symmetry between H and C, we also have $\mathbb{E}(\mathcal{N}_{\mathbf{H}}(X(1, I_1))) = \infty$. □

Proof of Lemma 3.4. Let $I_0 = 0$ and for $m \in \mathbb{N}$, let I_m be the m^{th} smallest $i \in \mathbb{N}$ for which $X(1, i)$ contains no hamburgers or cheeseburgers. The definition of I_1 is the same as that given in Lemma 3.6. Furthermore, if $i \in \mathbb{N}$ and X_i is a d or a S which is not identified in $X_1 X_2 \cdots$, then i must be one of the times I_m for $m \in \mathbb{N}$.

For each $m \in \mathbb{N}$, the time I_m is a stopping time for the filtration generated by X , read forward. Furthermore, for $m \in \mathbb{N}$ and $i \geq I_{m-1} + 1$, the word $X(1, i)$ contains no hamburgers or cheeseburgers if and only if $X(I_{m-1} + 1, i)$ contains no hamburgers or cheeseburgers. By the strong Markov property, the words $X_{I_{m-1}+1} \cdots X_{I_m}$ for $m \in \mathbb{N}$ are i.i.d.

For $m \in \mathbb{N}$, let

$$\xi_m := \mathcal{N}_{\mathbf{H}}(X(I_{m-1} + 1, I_m)),$$

so that the random variables ξ_m for $m \in \mathbb{N}$ are i.i.d. None of the H's in $X(I_{m-1} + 1, I_m)$ have a match in $X_1 X_2 \cdots$, so for each $m \in \mathbb{N}$

$$\mathcal{N}_{\mathbf{H}}(X(1, I_m)) = \sum_{k=1}^m \xi_k.$$

By Lemma 3.6, for each $\varepsilon > 0$ we can find an $R > 0$ such that

$$\mathbb{E}(\xi_1 \wedge R) \geq 2\varepsilon^{-1}.$$

By Hoeffding's inequality for sums of i.i.d. bounded random variables, for each $m \in \mathbb{N}$,

$$\begin{aligned} \mathbb{P}(\mathcal{N}_H(X(1, I_m)) \leq \varepsilon^{-1}m) &\leq \mathbb{P}\left(\frac{1}{m} \sum_{k=1}^m (\xi_k \wedge R) \leq \varepsilon^{-1}\right) \\ &\leq \exp\left(-\frac{2m}{\varepsilon^2 R^2}\right). \end{aligned} \quad (3.19)$$

Given $n \in \mathbb{N}$, let M_n be the largest $m \in \mathbb{N}$ for which $I_m \leq n$. Then $\mathcal{N}_H(X(1, n)) \geq \mathcal{N}_H(X(1, I_{M_n}))$ and $\mathcal{N}_{\text{d|S}}(X(1, n)) \leq M_n$. By (3.19),

$$\begin{aligned} \mathbb{P}(F_n(\varepsilon, A)) &\leq \mathbb{P}\left(\frac{M_n}{\mathcal{N}_H(X(1, I_{M_n}))} \geq \varepsilon, M_n \geq \varepsilon A\right) \\ &= \sum_{m=\lceil \varepsilon A \rceil}^{\infty} \mathbb{P}(\mathcal{N}_H(X(1, I_m)) \leq \varepsilon^{-1}m, M_n = m) \\ &\leq \sum_{m=\lceil \varepsilon A \rceil}^{\infty} \exp\left(-\frac{2m}{\varepsilon^2 R^2}\right) \end{aligned}$$

so we take $c_1 = 2/(\varepsilon R^2)$ and $c_0 = 1/(1 - e^{-c_1/\varepsilon})$. \square

3.3 Renewal times in the word

For the bi-infinite word X , let J be the age of the freshest (unconsumed) non-duplicate burger, as seen from the present:

$$J := \min \{j \in \mathbb{N} : \mathcal{N}_{\text{h|c}}(X(-j, -1)) > 0\}, \quad (3.20)$$

and more generally we define a sequence of backward renewal times J_m by

$$J_m := \begin{cases} 0 & m = 0 \\ \min \{j \in \mathbb{N} : \mathcal{N}_{\text{h|c}}(X(-j, -J_{m-1} - 1)) > 0\} & m \in \mathbb{N}. \end{cases} \quad (3.21)$$

We also define

$$\chi := \mathbb{E}(|X(-J, -1)|), \quad (3.22)$$

In the case $p = q = 0$, $\mathbb{E}[J] = \infty$, so *a priori* we could have $\chi = \infty$, but we will prove that χ is finite in Proposition 3.11 below.

In this subsection we carry out a careful study of the time J and related quantities. These results are needed for the variance calculation in the next subsection. We start by recording some basic properties of J (which follow easily from the definition) in Lemma 3.8 and an alternative definition of J_m in Lemma 3.9. In Lemma 3.10, we show that J has finite moments up to order 1/2. The idea of the proof is to bound J above by a time associated with the simple

random walk $j \mapsto \mathcal{C}(X(-j, -1))$. Using this and Lemma 3.4, we prove in Proposition 3.11 that $\chi := \mathbb{E}(|X(-J, -1)|)$ is finite and that $\mathbb{E}(\mathcal{C}(X(-J, -1)))$ is non-negative. We then show that in fact this latter expectation is 0 using a generalization of the proof of [She16, Lem. 3.5]. Since $|X(-J, -1)| = 2 - \mathcal{C}(X(-J, -1))$ whenever $X(-J, -1)$ contains no \mathbf{d} 's, this shows in particular that $\chi = 2$ when $q = 0$, i.e., $z = 1$ (which is why we get an exact expression for the variances and covariances in Theorem 1.15 in this case). The last main result of this subsection is Lemma 3.16, which shows that $\mathbb{E}(|X(-n, -1)|\mathbb{1}_{(J < n)}) \rightarrow 0$ as $n \rightarrow \infty$, and is an easy consequence of the earlier results in this subsection and a dominated convergence argument.

Lemma 3.8. *With J as in (3.20),*

1. J is a.s. finite.
2. $X_{-J} \in \{\mathbf{h}, \mathbf{c}\}$.
3. The symbol X_{-J} does not have a match in $X_{-J} \cdots X_{-1}$.
4. The reduced word $X(-J, -1)$ consists of only hamburgers and cheeseburger orders (if $X_{-J} = \mathbf{h}$) or cheeseburgers and hamburger orders (if $X_{-J} = \mathbf{c}$).

Proof. Assertion 1 follows from Lemma A.5. By definition of J , the word $X(-J + 1, -1)$ contains no \mathbf{h} or \mathbf{c} symbols, so assertion 2 follows from Lemma A.2 (applied with $x = X_{-J}$ and $y = X_{-J+1} \cdots X_{-1}$).

Suppose $k \in [1, J - 1]_{\mathbb{Z}}$. By definition of J , the word $X(-k, -1)$ contains no \mathbf{h} or \mathbf{c} . If $X(-J, -k - 1)$ contained no burger, then $\mathcal{R}(X(-J, -k - 1)X(-k, -1)) = X(-J, 1)$ would contain no \mathbf{h} or \mathbf{c} , contrary to the definition of J . So $X(-J, -k)$ contains a burger.

We argue by induction on $J - k \in [0, J - 1]_{\mathbb{Z}}$ that each symbol in $X_{-J} \cdots X_{-k}$ is identified in this word. Since $X_{-J} \in \{\mathbf{h}, \mathbf{c}\}$, this is true for $k = J$. If the claim is true for k , then since $X(-J, -k)$ contains a burger, each of which by induction is identified, it follows that X_{-k+1} is identified in $X_{-J} \cdots X_{-k+1}$, completing the induction.

Every burger in $X(-J + 1, -1)$ is a \mathbf{d} . Since each burger in $X(-J, -1)$ is identified, it must be that they are identified to X_{-J} .

Suppose that X_{-J} is matched to an order $X_{\phi(-J)}$ for $\phi(-J) \in [-J + 1, -1]_{\mathbb{Z}}$. We assume without loss of generality that $X_{-J} = \mathbf{h}$. Consequently, $X(-J, -1)$ contains no \mathbf{c} . Since $X_{-J} = \mathbf{h}$ is consumed, the reduced word $X(-J, \phi(-J))$ consists of only \mathbf{c} 's and \mathbf{C} 's. Since $X(\phi(-J) + 1, -1)$ contains no \mathbf{h} or \mathbf{c} , each \mathbf{d} in $X(\phi(-J) + 1, -1)$ is identified by a \mathbf{c} in $X_{-J} \cdots X_{-1}$. Consequently, $X(-J, -1)$ contains no \mathbf{h} . We have already shown above that $X(-J, -1)$ contains no \mathbf{c} , so we contradict the definition of J . We thus obtain assertion 3.

Since each burger in $X(-J, -1)$ is identified to X_{-J} , and X_{-J} is not consumed, it must be that each order in $X(-J, -1)$ is for the opposite burger type, which proves assertion 4. \square

Our next lemma is an analogue of [GMS17a, Lem. A.7] in the setting where we read the word backward, rather than forward, and is proven in a similar manner.

Lemma 3.9. *The time J_m from (3.21) is the m^{th} smallest $j \in \mathbb{N}$ such that $\mathcal{N}_{\mathbf{h}|\mathbf{c}}(X(-j, -k)) > 0$ for all $k \in [1, j]_{\mathbb{Z}}$.*

Proof. Let $\tilde{J}_0 = 0$ and for $m \in \mathbb{N}$, let \tilde{J}_m be the m^{th} smallest $j \in \mathbb{N}$ such that $X(-j, -k)$ contains a hamburger or a cheeseburger for each $k \in [1, j]_{\mathbb{Z}}$. We show by induction that $\tilde{J}_m = J_m$ for each $m \in \mathbb{N}$. The base case $m = 0$ is trivial. Suppose $m \in \mathbb{N}$ and $\tilde{J}_{m-1} = J_{m-1}$. By assertion 3 of Lemma 3.8, the word $X(-J_m, -k)$ contains a hamburger or a cheeseburger (namely X_{-J_m}) for each $k \in [\tilde{J}_{m-1} + 1, J_m]_{\mathbb{Z}}$. By definition of \tilde{J}_{m-1} , the word $X(-\tilde{J}_{m-1}, -k)$ (and hence the word $X(-J_m, -k)$) contains a hamburger or a cheeseburger for each $k \in [1, \tilde{J}_{m-1}]_{\mathbb{Z}}$. Thus J_m is one of the $\tilde{J}_{m'}$'s, and hence $J_m \geq \tilde{J}_m$. On the other hand, the word $X(-\tilde{J}_m, -J_{m-1} - 1)$ contains a hamburger or cheeseburger by the inductive hypothesis and the definition of \tilde{J}_m , so $J_m \leq \tilde{J}_m$, so in fact $\tilde{J}_m = J_m$. \square

We next prove that J has finite moments up to order $1/2$ (actually we prove something a little stronger, which will be needed for technical reasons below).

Lemma 3.10. *Let M be the smallest $m \in \mathbb{N}$ for which $\mathcal{C}(X(-J_m, -1)) \geq 1$. Almost surely $M < \infty$, and for each $\zeta \in (0, 1/2)$, we have $\mathbb{E}(J_M^\zeta) < \infty$.*

Proof. Let $P_0 = 0$ and for $m \in \mathbb{N}$, let P_m be the smallest $j \in \mathbb{N}$ for which $\mathcal{C}(X(-j, -1)) = m$, as in Lemma A.3. Also let \tilde{M} be the smallest $m \in \mathbb{N}$ for which $X_{-P_m} \in \{\mathbf{h}, \mathbf{c}\}$. By Lemma A.3, the word $X(-P_{\tilde{M}}, -n)$ contains either a hamburger or a cheeseburger for each $n \in [1, P_{\tilde{M}}]_{\mathbb{Z}}$. Therefore, Lemma 3.9 implies that $P_{\tilde{M}} = J_{\tilde{m}}$ for some $\tilde{m} \in \mathbb{N}$. Since $\mathcal{C}(X(-P_{\tilde{M}}, -1)) = \tilde{M} \geq 1$, we have $M \leq \tilde{m}$. Therefore $J_M \leq P_{\tilde{M}}$.

For $\zeta \in (0, 1/2)$, the function $t \mapsto t^\zeta$ is concave, hence subadditive. Thus, for $m \in \mathbb{N}$

$$P_m^\zeta \leq \sum_{k=1}^m (P_k - P_{k-1})^\zeta.$$

Since $j \mapsto \mathcal{C}(X(-j, -1))$ is a simple random walk, $\mathbb{E}(P_1^\zeta) < \infty$ for $\zeta \in (0, 1/2)$. By the strong Markov property, for each $m \in \mathbb{N}$, it holds with conditional probability $1 - q$ given $X_{-P_{m-1}} \cdots X_{-1}$ that $X_{-P_m} \in \{\mathbf{h}, \mathbf{c}\}$. Therefore, the law of \tilde{M} is geometric with success probability $1 - q$, and in particular $\mathbb{E}(\tilde{M}) < \infty$. By Wald's equation, it holds for each $\zeta \in (0, 1/2)$ that $\mathbb{E}(P_{\tilde{M}}^\zeta) < \infty$, and hence also $\mathbb{E}(J_M^\zeta) < \infty$. \square

We are now ready to prove that the quantity χ of (3.22) is finite.

Proposition 3.11.

$$\chi = \mathbb{E}(|X(-J, -1)|) < \infty \tag{3.23}$$

and

$$\mathbb{E}(\mathcal{C}(X(-J, -1))) \geq 0. \tag{3.24}$$

Proof. Fix $\varepsilon, \zeta \in (0, 1/2)$ and for $n \in \mathbb{N}$, let $F_n = F_n(\varepsilon, n^\zeta)$ be defined as in (3.13) but with $X(-n, -1)$ in place of $X(1, n)$. Let

$$\Xi := \sum_{n=1}^{\infty} n \mathbb{1}_{F_n}. \tag{3.25}$$

By Lemma 3.4 and translation invariance, $\mathbb{E}(\Xi) < \infty$.

For $n \in \mathbb{N}$, if F_n occurs, then

$$\mathcal{C}(X(-n, -1)) \leq |X(-n, -1)| \leq n \leq \Xi. \quad (3.26)$$

For $n \in \mathbb{N}$, if $n < J$ then every burger in $X(-n, -1)$ is a **d**. If $n < J$ and furthermore F_n does not occur, then $\mathcal{N}_{\mathbf{d}|\mathbf{S}}(X(-n, -1)) \leq \varepsilon \mathcal{N}_{\mathbf{H}|\mathbf{C}}(X(-n, -1)) + \varepsilon n^\zeta$ since $\varepsilon < 1$, so

$$\begin{aligned} \mathcal{C}(X(-n, -1)) &= \mathcal{N}_{\mathbf{d}}(X(-n, -1)) - \mathcal{N}_{\mathbf{H}|\mathbf{C}|\mathbf{S}}(X(-n, -1)) \\ &\leq \mathcal{N}_{\mathbf{d}|\mathbf{S}}(X(-n, -1)) - \varepsilon \mathcal{N}_{\mathbf{H}|\mathbf{C}}(X(-n, -1)) \leq \varepsilon n^\zeta. \end{aligned} \quad (3.27)$$

For $n \in \mathbb{N} \cup \{0\}$, let

$$Y_n = \mathcal{C}(X(-(J \wedge n), -1)).$$

Whether or not $F_{(J \wedge n)-1}$ occurs, from (3.26) and (3.27) applied to $(J \wedge n) - 1$, we have

$$Y_n \leq 1 + \mathcal{C}(X(-(J \wedge n) + 1, -1)) \leq 1 + \varepsilon J^\zeta + \Xi. \quad (3.28)$$

Note that the Ξ comes from the possibility that $F_{(J \wedge n)-1}$ does not occur. Since $\mathcal{C}(X(-n, -1))$ is a martingale, the optional stopping theorem implies $\mathbb{E}[Y_n] = 0$. Let $R = 1 + J^\zeta + \Xi$. By Lemma 3.10 (note that $J \leq J_M$) and since $\mathbb{E}(\Xi) < \infty$, we have $\mathbb{E}(R) < \infty$. Since $0 \leq R - Y_n$ and $Y_n \rightarrow \mathcal{C}(X(-J, -1))$, Fatou's lemma implies

$$\mathbb{E}(R - \mathcal{C}(X(-J, -1))) \leq \liminf_n \mathbb{E}(R - Y_n) = \mathbb{E}(R). \quad (3.29)$$

This in particular implies $\mathbb{E}(\mathcal{C}(X(-J, -1))) \geq 0$, i.e., (3.24).

Since every burger in $X(-n, -1)$ is a **d** when $n < J$,

$$|X(-n, -1)| \mathbb{1}_{\{n < J\}} = -\mathcal{C}(X(-n, -1)) + 2\mathcal{N}_{\mathbf{d}}(X(-n, -1)). \quad (3.30)$$

If $n < J$ and F_n does not occur, then

$$\begin{aligned} \mathcal{N}_{\mathbf{d}}(X(-n, -1)) &\leq \varepsilon \mathcal{N}_{\mathbf{H}|\mathbf{C}}(X(-n, -1)) + \varepsilon n^\zeta \\ (1 - \varepsilon)\mathcal{N}_{\mathbf{d}}(X(-n, -1)) &\leq -\varepsilon \mathcal{C}(X(-n, -1)) + \varepsilon n^\zeta. \end{aligned}$$

Note that in the second inequality, we use that $\mathcal{N}_{\mathbf{d}}(X(-n, -1)) \leq \varepsilon n^\zeta$ and that $-\varepsilon \mathcal{N}_{\mathbf{d}}(X(-n, -1)) \leq -\varepsilon \mathcal{C}(X(-n, -1))$ since every burger in $X(-n, -1)$ is a **d**. Combining the above inequalities with (3.30) gives

$$|X(-n, -1)| \mathbb{1}_{\{n < J\} \cap F_n^c} \leq -\left(1 + \frac{2\varepsilon}{1 - \varepsilon}\right) \mathcal{C}(X(-n, -1)) + \frac{2\varepsilon}{1 - \varepsilon} n^\zeta. \quad (3.31)$$

We combine (3.26) and (3.31), applied to $n = J - 1$, to obtain

$$|X(-J, -1)| \leq 1 - \left(1 + \frac{2\varepsilon}{1 - \varepsilon}\right) \mathcal{C}(X(-J, -1)) + \frac{2\varepsilon}{1 - \varepsilon} J^\zeta + \Xi.$$

Since the expectation of each term on the right side of this last inequality is finite, we obtain (3.23). \square

Lemma 3.12. *With M as in Lemma 3.10, we have $\mathbb{E}(\mathcal{C}(X(-J_M, -1))) < \infty$.*

Proof. By definition of M and the times J_m ,

$$1 \leq \mathcal{C}(X(-J_M, -1)) \leq \mathcal{C}(X(-J_M, -J_{M-1} - 1)) \leq \mathcal{N}_d(X(-J_M + 1, -J_{M-1} - 1)) + 1;$$

in the second inequality, we use that $\mathcal{C}(X(-J_{M-1}, -1)) \leq 0$. Since every burger in $X(-J_M + 1, -J_{M-1} - 1)$ is a \mathbf{d} , and

$$\mathcal{C}(X(-J_M + 1, -J_{M-1} - 1)) \geq \mathcal{C}(X(-J_M + 1, -1)) \geq \mathcal{C}(X(-J_M, -1)) - 1 \geq 0,$$

we have

$$\mathcal{N}_d(X(-J_M + 1, -J_{M-1} - 1)) \geq \mathcal{N}_{\text{H|C}}(X(-J_M + 1, -J_{M-1} - 1)).$$

Now fix $\zeta \in (0, 1/2)$, and for $m \in \mathbb{N}$ let

$$E_m := \{\mathcal{N}_d(X(-J_m + 1, -J_{m-1} - 1)) \geq \mathcal{N}_{\text{H|C}}(X(-J_m + 1, -J_{m-1} - 1)) \vee m^\zeta\}.$$

Either $\mathcal{N}_d(X(-J_m + 1, -J_{m-1} - 1)) < M^\zeta$ or E_m occurs. Therefore,

$$\begin{aligned} \mathcal{C}(X(-J_M, -1)) &\leq \mathcal{N}_d(X(-J_M + 1, -J_{M-1} - 1)) + 1 \\ &\leq M^\zeta + \mathcal{N}_d(X(-J_M + 1, -J_{M-1} - 1)) \mathbb{1}_{E_M} + 1 \\ &\leq J_M^\zeta + \sum_{m=1}^{\infty} \mathcal{N}_d(X(-J_m + 1, -J_{m-1} - 1)) \mathbb{1}_{E_m} + 1. \end{aligned}$$

By Lemma 3.10 we know $\mathbb{E}(J_M^\zeta) < \infty$, so to complete the proof it suffices to show

$$\sum_{m=1}^{\infty} \mathbb{E}(\mathcal{N}_d(X(-J_m + 1, -J_{m-1} - 1)) \mathbb{1}_{E_m}) < \infty. \quad (3.32)$$

Recall that the words $X_{-J_m} \cdots X_{-J_{m-1}-1}$ are i.i.d. with the same law as $X_{-J} \cdots X_{-1}$. For $B > 0$, Lemma 3.4 and a union bound over all $n \in [1, B]_{\mathbb{Z}}$ yields

$$\mathbb{P}(\mathcal{N}_d(X(-J + 1, -1)) \geq \mathcal{N}_{\text{H|C}}(X(-J + 1, -1)) \vee A, J \leq B) \leq c_0 B e^{-c_1 A}$$

for constants $c_0, c_1 > 0$ depending only on p and q . Lemma 3.10 and the Chebyshev inequality together imply that $\mathbb{P}(J > B) = \mathbb{P}(J^\zeta > B^\zeta) \leq \mathbb{E}(J^\zeta) B^{-\zeta}$. Thus

$$\mathbb{P}(\mathcal{N}_d(X(-J + 1, -1)) \geq \mathcal{N}_{\text{H|C}}(X(-J + 1, -1)) \vee A) \leq c_0 B e^{-c_1 A} + \mathbb{E}(J^\zeta) B^{-\zeta},$$

and since B was arbitrary, we choose $B = \exp[c_1 A / (1 + \zeta)]$. Then

$$\begin{aligned} &\mathbb{E}(\mathcal{N}_d(X(-J_m + 1, -J_{m-1} - 1)) \mathbb{1}_{E_m}) \\ &= \mathbb{E}(\mathcal{N}_d(X(-J + 1, -1)) \mathbb{1}\{\mathcal{N}_d(X(-J + 1, -1)) \geq \mathcal{N}_{\text{H|C}}(X(-J + 1, -1)) \vee m^\zeta\}) \\ &= \sum_{k \geq m^\zeta} k \times \mathbb{P}(k = \mathcal{N}_d(X(-J + 1, -1)) \geq \mathcal{N}_{\text{H|C}}(X(-J + 1, -1)) \vee m^\zeta) \\ &\leq \sum_{k \geq m^\zeta} k \times \mathbb{P}(\mathcal{N}_d(X(-J + 1, -1)) \geq \mathcal{N}_{\text{H|C}}(X(-J + 1, -1)) \vee k) \\ &\leq \sum_{k \geq m^\zeta} k \times (c_0 + \mathbb{E}[J^\zeta]) \times \exp[-(c_1 \zeta / (1 + \zeta))k] \\ &\leq (m^\zeta + \text{const}) \times \text{const} \times \exp[-\text{const} \times m^\zeta], \end{aligned}$$

which is summable in m , establishing (3.32). \square

The next two lemmas correspond to [She16, Lem. 3.5]. However, slightly more work is needed to prove Lemma 3.14 below in our setting because the word $X(-J, -1)$ can contain more than one burger, so with J_M as in Lemma 3.10, we might have $\mathcal{C}(X(-J_M, -1)) > 1$.

Lemma 3.13. *Let M be the smallest $m \in \mathbb{N}$ for which $\mathcal{C}(X(-J_m, -1)) \geq 1$, as in Lemma 3.10. Then $\mathbb{E}[M] = \infty$.*

Proof. With P as in Lemma 3.5, i.e., the smallest $j \in \mathbb{N}$ for which $\mathcal{C}(X(-j, -1)) = 1$,

$$|X(-P, -1)| = 2\mathcal{N}_{\text{h|c|d}}(X(-P, -1)) + 1. \quad (3.33)$$

For $\varepsilon, \zeta \in (0, 1/2)$ and the events $F_n = F_n(\varepsilon, n^\zeta)$ and the random variable Ξ in (3.25),

$$\mathcal{N}_{\text{d|S}}(X(-P, -1)) \leq \begin{cases} \varepsilon|X(-P, -1)| + \varepsilon P^\zeta & \text{if } F_P \text{ does not occur} \\ \Xi & \text{if } F_P \text{ occurs.} \end{cases}$$

By this and (3.33),

$$|X(-P, -1)| \leq 2\mathcal{N}_{\text{h|c}}(X(-P, -1)) + 2\varepsilon P^\zeta + 2\varepsilon|X(-P, -1)| + 2\Xi + 1. \quad (3.34)$$

Since $\mathcal{C}(X(-J_M, -1)) \geq 1$, we have $P \leq J_M$. Since $\mathbb{E}(\Xi) < \infty$ and $\mathbb{E}(P^\zeta) \leq \mathbb{E}(J_M^\zeta) < \infty$, $\mathbb{E}(|X(-P, -1)|) = \infty$ by Lemma 3.5, and $\varepsilon < 1/2$, we deduce from (3.34) that

$$\mathbb{E}(\mathcal{N}_{\text{h|c}}(X(-P, -1))) = \infty.$$

Since $P \leq J_M$,

$$\mathcal{N}_{\text{h|c}}(X(-P, -1)) \leq \mathcal{N}_{\text{h|c}}(X(-J_M, -1)).$$

Since each symbol in $X(-J_m, -J_{m-1} - 1)$ is identified,

$$\mathcal{N}_{\text{h|c}}(X(-J_M, -1)) \leq \sum_{m=1}^M \mathcal{N}_{\text{h|c}}(X(-J_m, -J_{m-1} - 1)).$$

The summands are i.i.d., and have finite expectation by Proposition 3.11. But the left hand side has infinite expectation, so by Wald's equation, $\mathbb{E}[M] = \infty$. \square

Lemma 3.14.

$$\mathbb{E}(\mathcal{C}(X(-J, -1))) = 0.$$

Proof. Write $\alpha = \mathbb{E}(\mathcal{C}(X(-J, -1)))$. Observe that by Proposition 3.11,

$$0 \leq \alpha \leq \mathbb{E}(|\mathcal{C}(X(-J, -1))|) \leq \mathbb{E}(|X(-J, -1)|) < \infty.$$

The strong Markov property implies that the words $X_{-J_m} \cdots X_{-J_{m-1}-1}$ for $m \in \mathbb{N}$ are i.i.d., and each has the same law as $X_{-J} \cdots X_{-1}$. By Lemma 3.8, none of the reduced words $X(-J_m, -J_{m-1} - 1)$ contains an unidentified d or S. By definition of α , we find that

$$A_m := \mathcal{C}(X(-J_m, -1)) - \alpha m = \sum_{k=1}^m \mathcal{C}(X(-J_k, -J_{k-1} - 1)) - \alpha m$$

is a martingale in m .

Let M be the smallest $m \in \mathbb{N}$ for which $\mathcal{C}(X(-J_m, -1)) \geq 1$, as in Lemma 3.10. By the optional stopping theorem, for each $n \in \mathbb{N}$ we have $\mathbb{E}(A_{M \wedge n}) = 0$. Since $A_{M \wedge n} \leq \mathcal{C}(X(-J_M, -1))$ and the latter quantity has finite expectation by Lemma 3.12, it follows from Fatou's lemma that

$$0 \leq \mathbb{E}(A_M) \leq \mathbb{E}(\mathcal{C}(X(-J_M, -1))).$$

In particular $\mathbb{E}(A_M) \geq 0$ implies

$$\alpha \mathbb{E}(M) \leq \mathbb{E}(\mathcal{C}(X(-J_M, -1))).$$

By Lemma 3.12 $\mathbb{E}(\mathcal{C}(X(-J_M, -1))) < \infty$ and by Lemma 3.13 $\mathbb{E}(M) = \infty$, so $\alpha \leq 0$. We already showed in Proposition 3.11 that $\alpha \geq 0$, so in fact $\alpha = 0$. \square

The following corollary is the reason why we know the variance and covariance of Z in Theorem 1.15 in the case when $z = 1$.

Corollary 3.15. *If $q = 0$ then $\chi = 2$.*

Proof. When $q = 0$ the word $X(-J, -1)$ contains exactly one burger. Hence in this case $|X(-J, -1)| = 2 - \mathcal{C}(X(-J, -1))$. Therefore Lemma 3.14 implies $\chi = 2$ in this case. \square

Lemma 3.16.

$$\lim_{n \rightarrow \infty} \mathbb{E}(|X(-n, -1)| \times \mathbb{1}_{(J > n)}) = 0.$$

Proof. By the optional stopping theorem, for each $n \in \mathbb{N}$,

$$0 = \mathbb{E}(\mathcal{C}(X(-J \wedge n, -1))) = \mathbb{E}(\mathcal{C}(X(-J, -1))\mathbb{1}_{(J \leq n)}) + \mathbb{E}(\mathcal{C}(X(-n, -1))\mathbb{1}_{(J > n)}). \quad (3.35)$$

Since

$$|\mathcal{C}(X(-J, -1))\mathbb{1}_{(J \leq n)}| \leq |\mathcal{C}(X(-J, -1))| \leq |X(-J, -1)|,$$

and by Proposition 3.11 $\mathbb{E}(|X(-J, -1)|) < \infty$, by dominated convergence (and Lemma 3.14),

$$\lim_{n \rightarrow \infty} \mathbb{E}(\mathcal{C}(X(-J, -1))\mathbb{1}_{(J \leq n)}) = \mathbb{E}(\mathcal{C}(X(-J, -1))) = 0.$$

It therefore follows from (3.35) that

$$\lim_{n \rightarrow \infty} \mathbb{E}(\mathcal{C}(X(-n, -1))\mathbb{1}_{(J > n)}) = 0. \quad (3.36)$$

Now fix $\varepsilon, \zeta \in (0, 1/2)$ and let $F_n = F_n(\varepsilon, n^\zeta)$ be as in Lemma 3.4 with $X_{-n} \cdots X_{-1}$ in place of $X_1 \cdots X_n$, as in the proof of Proposition 3.11. By (3.31) and since $|X(-n, -1)| \leq n$,

$$|X(-n, -1)|\mathbb{1}_{(J > n)} \leq -\left(1 + \frac{2\varepsilon}{1 - \varepsilon}\right)\mathcal{C}(X(-n, -1))\mathbb{1}_{(J > n)} + \frac{2\varepsilon}{1 - \varepsilon}n^\zeta\mathbb{1}_{(J > n)} + n\mathbb{1}_{F_n}. \quad (3.37)$$

By (3.36), the expectation of the first term on the right in (3.37) tends to 0 as $n \rightarrow \infty$. By Lemma 3.4, $\lim_{n \rightarrow \infty} n\mathbb{P}(F_n) = 0$. By Lemma 3.12, for each $\zeta' \in (\zeta, 1/2)$ we have $\mathbb{E}(J^{\zeta'}) \leq \mathbb{E}(J_M^{\zeta'}) < \infty$, so by Chebyshev's inequality $\mathbb{P}(J > n) \leq \mathbb{E}(J_M^{\zeta'})/n^{\zeta'}$. By combining these observations with (3.37), we obtain the statement of the lemma. \square

3.4 Variance of the discrepancy between burger types

In this subsection we obtain an asymptotic formula for $\text{Var } \mathcal{D}(X'(1, n))$, where here \mathcal{D} is as in Definition 1.4 and X' is as in Definition 1.5. This formula will be used to obtain the variance and covariance for the limiting Brownian motion in Theorem 1.15. In particular, we prove Proposition 3.17 below. The proof is similar to the argument found in [She16, § 3.1], but unlike in [She16, § 3.1], all of the assumptions needed to make the argument work have already been proven. Recall from Proposition 3.11 that χ is finite.

Proposition 3.17. *Let χ be as in (3.22). Then*

$$\lim_{n \rightarrow \infty} n^{-1} \text{Var}(\mathcal{D}(X'(-n, -1))) = 1 + (p + q)\chi.$$

Proof. By Lemma 3.8, the word $X(-J, -1)$ is equal to $X'(-J, -1)$ and consists of either h's and C's (if $X_{-J} = \mathbf{h}$) or c's and H's (if $X_{-J} = \mathbf{c}$). Therefore,

$$\mathcal{D}(X(-J, -1)) = \mathcal{D}(X'(-J, -1)) = \pm |X(-J, -1)| \quad (3.38)$$

where the sign is positive if $X_{-J} = \mathbf{h}$ and negative if $X_{-J} = \mathbf{c}$. We observe that X_0 is independent from $X(-J, -1)$, and that X'_0 is determined by X_0 on the event $\{X_0 \notin \{\mathbf{d}, \mathbf{S}\}\}$. Therefore,

$$\mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X(-J, -1))\mathbb{1}_{(X_0 \notin \{\mathbf{d}, \mathbf{S}\})}) = 0.$$

If on the other hand $X_0 \in \{\mathbf{d}, \mathbf{S}\}$, if $X_{-J} = \mathbf{h}$ then $X'_0 \in \{\mathbf{h}, \mathbf{C}\}$, and if $X_{-J} = \mathbf{c}$ then $X'_0 \in \{\mathbf{c}, \mathbf{H}\}$. Therefore, if $X_0 \in \{\mathbf{d}, \mathbf{S}\}$ then $\mathcal{D}(X'_0)$ has the same sign as $\mathcal{D}(X(-J, -1))$, so

$$\begin{aligned} \mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X(-J, -1))) &= \mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X(-J, -1))\mathbb{1}_{(X_0 \in \{\mathbf{d}, \mathbf{S}\})}) \\ &= \mathbb{P}(X_0 \in \{\mathbf{d}, \mathbf{S}\})\mathbb{E}(|X(-J, -1)|) = \frac{\chi(p + q)}{2}. \end{aligned} \quad (3.39)$$

We next observe that X'_0 is determined by $X_{-J} \cdots X_{-1}$ and X_0 , so by the strong Markov property, for each $n \in \mathbb{N}$ it holds that X'_0 is conditionally independent from $X'_{-n} \cdots X'_{-J-1}$ given $X'_{-J} \cdots X'_{-1}$ (here we set $X(-n, -J-1) = \emptyset$ if $n \leq J$, so that the assertion holds vacuously in this case). By symmetry $\mathcal{D}(X'(-n, -J-1))$ has zero conditional mean given $X_{-J} \cdots X_{-1}$, so

$$\mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X'(-n, -J-1)) \mid X_{-J} \cdots X_{-1}) = 0. \quad (3.40)$$

Therefore,

$$\mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X'(-n, -1))) = \mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X(-J, -1))\mathbb{1}_{J \leq n}) + \mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X'(-n, -1))\mathbb{1}_{J > n}). \quad (3.41)$$

By (3.38), (3.39), and dominated convergence (with $|X(-J, -1)|$ as the dominator; recall Proposition 3.11) we find that the first term on the right in (3.41) tends to $\chi(p + q)/2$ as $n \rightarrow \infty$. The absolute value of the second term is at most $\mathbb{E}(|X(-n, -1)|\mathbb{1}_{(J > n)})$, which

tends to 0 by Lemma 3.16. By translation invariance, we therefore have

$$\begin{aligned}
\text{Var}(\mathcal{D}(X'(1, n))) &= \mathbb{E}(\mathcal{D}(X'(-n, -1))^2) \\
&= \sum_{i=1}^n \mathbb{E}(\mathcal{D}(X'_i)^2) + 2 \sum_{i=2}^n \mathbb{E}(\mathcal{D}(X'_i)\mathcal{D}(X'(1, i-1))) \\
&= n + 2 \sum_{i=2}^n \mathbb{E}(\mathcal{D}(X'_0)\mathcal{D}(X'(-i+1, -1))) \\
&= (1 + \chi(p+q))n + o(n). \quad \square
\end{aligned}$$

3.5 Expected length of the reduced word

In this subsection we estimate the expectations of several quantities related to the reduced words $X(1, n)$ and $X'(1, n)$ for $n \in \mathbb{N}$ (recall (1.8)). As one might expect due to the diffusive scaling for Z^n in (1.15), these quantities will typically be of order $n^{1/2}$. We first prove in Lemma 3.18 an upper bound for the length of the latter word, which may be shorter than $|X(1, n)|$ since there could be \mathbf{d} 's in $X_1 \dots X_n$ which are identified by burgers in $\dots X_{-1}X_0$ but matched to orders in $X_1 \dots X_n$. In Lemma 3.19, we transfer this to an upper bound for $|X(1, n)|$ using Lemma 3.4. We then use a comparison to simple random walk on \mathbb{Z}^2 (via Lemma 3.1) to prove a corresponding lower bound for the expected number of burgers and orders in $X(1, n)$ (Lemma 3.20).

Lemma 3.18. *For $n \in \mathbb{N}$, we have (using the notation \preceq from Section 1.2),*

$$\mathbb{E}(|X'(1, n)|) \preceq n^{1/2}.$$

Proof. By the symmetry between hamburgers and cheeseburgers, $\mathbb{E}(\mathcal{D}(X'(1, n))) = 0$, so by Proposition 3.17 and translation invariance, for each $n \in \mathbb{N}$ we have $\mathbb{E}(\mathcal{D}(X'(1, n))^2) = \text{Var}(\mathcal{D}(X'(-n, -1))) \preceq n$. Since $n \mapsto \mathcal{C}(X'(1, n))$ is a simple random walk, $\mathbb{E}(\mathcal{C}(X'(1, n))^2) = n$. With $d(X'(1, n))$ as in Definition 1.4, $d(X'(1, n)) = \frac{1}{2}(\mathcal{D}(X'(1, n)) + \mathcal{C}(X'(1, n)))$. By a union bound and the Chebyshev inequality, we infer

$$\mathbb{P}(|d(X'(1, n))| \geq k) \preceq n/k^2, \quad \forall n, k \in \mathbb{N}. \quad (3.42)$$

For $k \in \mathbb{N}$, let K_k be the smallest $i \in \mathbb{N}$ for which $X(-i, -1)$ contains at least k hamburgers. Then X_{-K_k} is a \mathbf{h} without a match in $X_{-K_k} \dots X_{-1}$, so each \mathbf{d} or \mathbf{S} in $X(-K_k + 1, -1)$ must be identified and there are no hamburger orders in $X(-K_k + 1, -1)$. Consequently, the word $X(-K_k, -1)$ contains at least k hamburgers, no unidentified \mathbf{d} 's or \mathbf{S} 's, and no orders other than cheeseburger orders. Therefore,

$$d(X(-K_k, -1)) = d(X'(-K_k, -1)) \geq k.$$

It follows that if $K_k \leq n$, then either

$$d(X'(-n, -K_k - 1)) \leq -k/2 \quad \text{or} \quad d(X'(-n, -1)) \geq k/2.$$

Since K_k is a backward stopping time for the word X , we infer from the strong Markov property and translation invariance that the conditional law of $d(X'(-n, -K_k - 1))$ given $X_{-K_k} \cdots X_{-1}$ is the same as the law of $d(X'(1, n - K_k))$. By (3.42) and the union bound,

$$\mathbb{P}(K_k \leq n) \preceq n/k^2,$$

and hence

$$\mathbb{P}(\mathcal{N}_h(X'(-n, -1)) \geq k) \preceq n/k^2, \quad \forall k, n \in \mathbb{N}. \quad (3.43)$$

By combining (3.42) and (3.43) and noting that $\mathcal{N}_{h|H}(x) \leq 2\mathcal{N}_h(x) + |d(x)|$ for every word x , we get

$$\mathbb{P}(\mathcal{N}_{h|H}(X'(-n, -1)) \geq k) \preceq n/k^2, \quad \forall k, n \in \mathbb{N}.$$

By symmetry, the analogous estimate holds with \mathfrak{c} and \mathfrak{C} in place of h and H . Since $|X'(-n, -1)| = \mathcal{N}_{h|\mathfrak{c}|H|\mathfrak{C}}(X'(-n, -1))$, a union bound therefore implies

$$\mathbb{P}(|X'(-n, -1)| \geq k) \preceq n/k^2, \quad \forall k, n \in \mathbb{N}.$$

Hence

$$\mathbb{E}(|X'(-n, -1)|) = \sum_{k=1}^{\infty} \mathbb{P}(|X'(-n, -1)| \geq k) \preceq \int_1^{\infty} (1 \wedge (n/k^2)) dk \preceq n^{1/2},$$

which finishes the proof in view of translation invariance. \square

We now estimate the expectation of $|X(1, n)|$, which may be larger than the expectation of $|X'(1, n)|$ since some duplicate burgers with no match in $X_1 \cdots X_n$ may correspond to hamburgers or cheeseburgers in X' which have a match in $X'_1 \cdots X'_n$.

Lemma 3.19.

$$\mathbb{E}(|X(1, n)|) \preceq n^{1/2}. \quad (3.44)$$

and

$$\mathbb{E}(\mathcal{N}_{d|S}(X(1, n))) = o(n^{1/2}), \quad (3.45)$$

as $n \in \mathbb{N}$ tends to infinity.

Proof. If $i \in [1, n]_{\mathbb{Z}}$ is such that X_i does not have a match in $X_1 \cdots X_n$ but X'_i has a match in $X'_1 \cdots X'_n$, then either $X_i = \mathfrak{d}$ or X_i is matched to a \mathfrak{d} in the word $X_1 \cdots X_n$. Therefore,

$$|X'(1, n)| \leq |X(1, n)| \leq |X'(1, n)| + 2\mathcal{N}_{\mathfrak{d}}(X(1, n)). \quad (3.46)$$

Now fix $\varepsilon, \zeta \in (0, 1/2)$ and for $n \in \mathbb{N}$ let $F_n = F_n(\varepsilon, n^\zeta)$ be the event defined in (3.13). On the event F_n^c , we have

$$\mathcal{N}_{d|S}(X(1, n)) \leq \varepsilon |X(1, n)| + \varepsilon n^\zeta \leq \varepsilon |X'(1, n)| + 2\varepsilon \mathcal{N}_{d|S}(X(1, n)) + \varepsilon n^\zeta,$$

where we used (3.46) in the second inequality. After re-arranging this inequality, and considering also the possibility that F_n occurs, we get

$$\mathcal{N}_{d|S}(X(1, n)) \leq \frac{\varepsilon}{1 - 2\varepsilon} |X'(1, n)| + \frac{\varepsilon}{1 - 2\varepsilon} n^\zeta + n \mathbb{1}_{F_n}. \quad (3.47)$$

Combining (3.47), the bound $\mathbb{E}(|X'(1, n)|) \preceq n^{1/2}$ from Lemma 3.18, the exponential decay of $\mathbb{E}(n \mathbb{1}_{F_n})$ from Lemma 3.4, and the fact that $\varepsilon > 0$ can be made arbitrarily small, we easily obtain (3.45). We obtain (3.44) from (3.45), (3.46) and Lemma 3.18. \square

Lemma 3.20. For $n \in \mathbb{N}$,

$$\mathbb{E}(\mathcal{N}_h(X(1, n))) \asymp \mathbb{E}(\mathcal{N}_H(X(1, n))) \asymp n^{1/2}. \quad (3.48)$$

Proof. The upper bounds for both expectations in (3.48) follow from Lemma 3.19, so we only need to prove the lower bounds.

Recall that $\mathbb{P}^{0,0}$ denotes the law of X with $p = q = 0$ and $\mathbb{E}^{0,0}$ is the corresponding expectation. By Lemma 3.1,

$$\begin{aligned} \mathbb{E}(\mathcal{N}_{h|c|d}(X(1, n))) &\geq \mathbb{E}^{0,0}(\mathcal{N}_{h|c}(X(1, n))) \\ \mathbb{E}(\mathcal{N}_{H|C|S}(X(1, n))) &\geq \mathbb{E}^{0,0}(\mathcal{N}_{H|C}(X(1, n))). \end{aligned} \quad (3.49)$$

If all symbols in $X_{-n} \cdots X_{-1}$ are identified, then

$$\begin{aligned} \mathcal{N}_h(X(-n, -1)) &= \max_{1 \leq i \leq n} d(X(-i, -1)) \\ \mathcal{N}_H(X(-n, -1)) &= \max_{1 \leq i \leq n+1} -d(X(-n, -i)). \end{aligned}$$

Under $\mathbb{P}^{0,0}$, the maps $i \mapsto \vec{d}(X(-i, -1))$ and $i \mapsto \vec{d}(X(-n, -i))$ are two-dimensional simple random walks, so we deduce (using e.g., Donsker's invariance principle and Fatou's lemma together with the fact that Brownian motion has a well-defined running supremum process which is positive at any given time)

$$\begin{aligned} \mathbb{E}^{0,0}(\mathcal{N}_h(X(-n, -1))) &\succeq n^{1/2} \\ \mathbb{E}^{0,0}(\mathcal{N}_H(X(-n, -1))) &\succeq n^{1/2}. \end{aligned} \quad (3.50)$$

By symmetry $\mathbb{E}(\mathcal{N}_h(X(1, n))) = \mathbb{E}(\mathcal{N}_c(X(1, n)))$ and $\mathbb{E}(\mathcal{N}_H(X(1, n))) = \mathbb{E}(\mathcal{N}_C(X(1, n)))$, and by (3.45) of Lemma 3.19 $\mathbb{E}(\mathcal{N}_{d|S}(X(1, n))) = o(n^{1/2})$, which combined with (3.49) and (3.50) gives the lower bounds $\mathbb{E}(\mathcal{N}_h(X(1, n))) \succeq n^{1/2}$ and $\mathbb{E}(\mathcal{N}_H(X(1, n))) \succeq n^{1/2}$. \square

3.6 Tail bound for the length of the reduced word

In this subsection we prove the following analogue of [She16, Lem. 3.13], which will be used to prove tightness of the sequence of paths Z^n defined in (1.15) in the proof of Theorem 1.15.

Proposition 3.21. *There are constants $a_0, a_1 > 0$ such that for each $n \in \mathbb{N}$ and $r > 0$,*

$$\mathbb{P}\left(\max_{\substack{i, j \in [1, n]_{\mathbb{Z}} \\ 1 \leq i \leq j \leq n}} |X(i, j)| > rn^{1/2}\right) \leq a_0 e^{-a_1 r}. \quad (3.51)$$

To prove Proposition 3.21, we will study the times at which unmatched hamburgers are added when we read the word backwards. The increments of X between these times are i.i.d., and the number of d 's which are identified at each of these times (some of which also correspond to unmatched hamburgers in our word) can be bounded using Lemma 3.4 (c.f. Lemma 3.23). Using a lower bound for the probability that a reduced word of length n contains no hamburgers (Lemma 3.22) and Chernoff's inequality, we get an upper tail bound

for the number of hamburgers in $X(-n, -1)$. By symmetry, we also have an analogous bound for the number of cheeseburgers in $X(-n, -1)$. Since the difference $\mathcal{C}(X(-n, -1))$ between the number of burgers and the number of orders in $X(-n, -1)$ evolves as a simple random walk on \mathbb{Z} and by another application of Lemma 3.4, this will be enough to prove Proposition 3.21.

Lemma 3.22. *Let J^h be the smallest $j \in \mathbb{N}$ for which $X(-j, -1)$ contains a hamburger. Then*

$$\mathbb{P}(J^h > n) \asymp n^{-1/2} \quad (3.52)$$

with the implicit constant depending only on p .

Proof. For $n \in \mathbb{N} \cup \{0\}$, let E_n be the event that $X(1, n)$ contains no hamburgers (recall that $X(1, 0) = \emptyset$). By translation invariance,

$$\mathbb{P}(E_n) = \mathbb{P}(J^h > n). \quad (3.53)$$

In particular, $n \mapsto \mathbb{P}(E_n)$ is non-increasing.

Suppose $i \in [1, n]_{\mathbb{Z}}$. If X_i identifies to \mathbf{H} in $X_1 \cdots X_n$ and has no match in $X_1 \cdots X_{i-1}$, then E_{i-1} occurs and $X_i \in \{\mathbf{H}, \mathbf{S}\}$. On the other hand, if E_{i-1} occurs, then by independence of the symbols of X , it holds with conditional probability $\frac{1-p}{4}$ that $X_i = \mathbf{H}$, in which case X_i does not have a match in $X_1 \cdots X_i$. Therefore,

$$\mathbb{E}(\mathcal{N}_{\mathbf{H}}(X(1, n))) \leq \sum_{i=0}^{n-1} \mathbb{P}(E_i) \leq \frac{4}{1-p} \mathbb{E}(\mathcal{N}_{\mathbf{H}}(X(1, n))).$$

By Lemma 3.20 we can find a constant $C > 1$ such that for each $n \in \mathbb{N}$

$$C^{-1}n^{1/2} \leq \sum_{i=1}^{n-1} \mathbb{P}(E_i) \leq Cn^{1/2}. \quad (3.54)$$

By monotonicity of $\mathbb{P}(E_n)$, we immediately obtain

$$n\mathbb{P}(E_n) \leq \sum_{i=1}^{n-1} \mathbb{P}(E_i) \leq Cn^{1/2}.$$

Furthermore,

$$4C^2n\mathbb{P}(E_n) \geq \sum_{i=n}^{\lceil 4C^2n \rceil - 1} \mathbb{P}(E_i) \geq 2Cn^{1/2} - Cn^{1/2} = Cn^{1/2}.$$

Combining these two relations with (3.53) yields (3.52). \square

Lemma 3.23. *Let J^h be the smallest $j \in \mathbb{N}$ for which $X(-j, -1)$ contains a hamburger. There are constants $a_0, a_1 > 0$ depending only on p such that for each $m \in \mathbb{N}$, we have*

$$\mathbb{P}(\mathcal{N}_{\mathbf{d}|\mathbf{S}}(X(-J^h + 1, -1)) > m) \leq a_0 e^{-a_1 m}. \quad (3.55)$$

Proof. We observe that $\mathcal{N}_d(X(-J^h + 1, -1)) \geq \mathcal{N}_h(X(-J^h + 1, -1))$; indeed, otherwise it is not possible for all of the H's in $X(-J^h + 1, -1)$ to be fulfilled in $X_{-J^h} \dots X_{-1}$ while still leaving a leftover h. Now let $c_0, c_1 > 0$ be as in Lemma 3.4 with $\varepsilon = 1$. By that lemma and a union bound,

$$\mathbb{P}(\mathcal{N}_{d|S}(X(-J^h + 1, -1)) \geq m, J^h \leq e^{c_1 m/2}) \leq c_0 e^{-c_1 m/2}.$$

On the other hand, by Lemma 3.22 we have

$$\mathbb{P}(J^h > e^{c_1 m/2}) \leq e^{-c_1 m/4}.$$

Combining these estimates yields (3.55). \square

Proof of Proposition 3.21. Let $J_0^h = 0$ and for $m \in \mathbb{N}$ inductively let J_m^h be the smallest $j \geq J_{m-1}^h$ for which $X(-j, -J_{m-1}^h - 1)$ contains a hamburger. Then J_1^h is the same as the time J^h from Lemma 3.22 and by the strong Markov property the increments $X_{-J_m^h} \dots X_{-J_{m-1}^h - 1}$ for $m \in \mathbb{N}$ are i.i.d. For $m \in \mathbb{N}$, let

$$H_m := \mathcal{N}_h(X(-J_m^h, -J_{m-1}^h - 1)) = 1 + \mathcal{N}_d(X(-J_m^h + 1, -J_{m-1}^h - 1)).$$

Since none of the reduced words $X(-J_m^h, -J_{m-1}^h - 1)$ contain H's,

$$\mathcal{N}_h(X(-J_m^h, -1)) = \sum_{k=1}^m H_k. \quad (3.56)$$

By Lemma 3.23, for some positive number $\beta > 0$ (depending only on p) $\mathbb{E}(e^{\beta H_k}) < \infty$, and since the H_k 's are i.i.d., Chernoff's bound implies that there are positive numbers $\tilde{c}_0, \tilde{c}_1 > 0$ such that for each $m \in \mathbb{N}$,

$$\mathbb{P}\left(\sum_{k=1}^m H_k \geq \tilde{c}_0 m\right) \leq e^{-\tilde{c}_1 m}. \quad (3.57)$$

By Lemma 3.22, we can find a constant $c > 0$ such that for each $n, m \in \mathbb{N}$,

$$\mathbb{P}(J_m^h - J_{m-1}^h > n) \geq cn^{-1/2}.$$

Since the increments $J_m^h - J_{m-1}^h$ are i.i.d., we infer that for each $n, m \in \mathbb{N}$,

$$\mathbb{P}(J_m^h \leq n) \leq \mathbb{P}(J_k^h - J_{k-1}^h \leq n, \forall k \leq m) \leq (1 - cn^{-1/2})^m \leq \exp[-cm/n^{1/2}]. \quad (3.58)$$

Recall that $\mathcal{N}_h(X(-j, -1))$ is monotone increasing in j . If $J_m^h \geq n$ and $\mathcal{N}_h(X(-J_m^h, -1)) \leq \tilde{c}_0 m$, then $\mathcal{N}_h(X(-j, -1)) \leq \tilde{c}_0 m$ for each $j \in [1, n]_{\mathbb{Z}}$. By taking $m = \lfloor rn^{1/2}/\tilde{c}_0 \rfloor$ and applying (3.56), (3.57), and (3.58), we find that for each $n \in \mathbb{N}$,

$$\mathbb{P}\left(\max_{j \in [1, n]_{\mathbb{Z}}} \mathcal{N}_h(X(-j, -1)) > rn^{1/2}\right) \leq c_0 e^{-c_1 r} \quad (3.59)$$

for appropriate $c_0, c_1 > 0$ independent of r and n . By symmetry, the analogous estimate holds with c in place of h .

Since $j \mapsto \mathcal{C}(X(-j, -1))$ is a simple random walk, we have (see e.g. [LL10, Prop. 2.1.2b])

$$\mathbb{P}\left(\max_{j \in [1, n]_{\mathbb{Z}}} |\mathcal{C}(X(-j, -1))| > rn^{1/2}\right) \leq b_0 e^{-b_1 r^2} \quad (3.60)$$

for universal constants $b_0, b_1 > 0$. By Lemma 3.4 (applied with $\varepsilon = \frac{1}{2}$ and $A = \text{const} \times rn^{1/2}$) and the union bound, except on an event of probability $\leq \exp(-\Theta(r))$,

$$\mathcal{N}_d(X(-j, -1)) \leq \frac{1}{2} \text{const} \times rn^{1/2} - \frac{1}{2} \mathcal{C}(X(-j, -1)) + \frac{1}{2} \mathcal{N}_{\text{h|c|d}}(X(-j, -1)), \quad \forall j \in [1, n]_{\mathbb{Z}}. \quad (3.61)$$

Re-arranging gives

$$\mathcal{N}_d(X(-j, -1)) \leq \text{const} \times rn^{1/2} - \mathcal{C}(X(-j, -1)) + \mathcal{N}_{\text{h|c}}(X(-j, -1)), \quad \forall j \in [1, n]_{\mathbb{Z}}. \quad (3.62)$$

By writing $|X(-j, -1)| = 2\mathcal{N}_d(X(-j, -1)) + 2\mathcal{N}_{\text{h|c}}(X(-j, -1)) - \mathcal{C}(X(-j, -1))$, using the bound (3.62), and the bounds (3.59) and (3.60), we obtain

$$\mathbb{P}\left(\max_{j \in [1, n]_{\mathbb{Z}}} |X(-j, -1)| > rn^{1/2}\right) \leq \text{const} \times e^{-\text{const} \times r}. \quad (3.63)$$

We now observe that for $1 \leq i \leq j \leq n$, each order and each unidentified **S** or **d** in $X(i, j)$ also appears in $X(i, n)$; and each **h** or **c** in $X(i, j)$ either appears in $X(i, n)$ or is consumed by a unique order in $X(j+1, n)$. Thus

$$|X(i, j)| \leq |X(i, n)| + |X(j+1, n)|. \quad (3.64)$$

This bound (3.64) together with (3.63) imply (3.51). \square

3.7 Convergence to correlated Brownian motion

We are now ready to conclude the proof of Theorem 1.15. We first establish tightness.

Lemma 3.24. *Suppose we are in the setting of Theorem 1.15. The sequence of laws of the paths Z^n for $n \in \mathbb{N}$ is tight in the topology of uniform convergence on compacts of \mathbb{R} .*

Proof. Fix $T \geq 1$ and $\varepsilon > 0$. For $N \in \mathbb{N}$, we cover the time interval $[0, T]$ by N blocks of the form $[kT/N, (k+2)T/N]$ for $k \in [0, N-1]_{\mathbb{Z}}$. Note that successive blocks overlap. Within each block, the path Z^n has (up to rounding error) $2nT/N$ steps. Any pair of times $s, t \in [0, T]$ with $|s-t| < T/N$ lie in some common block, and if $s, t \in \mathbb{Z}/n$ and $s < t$, $\|Z^n(s) - Z^n(t)\|_1$ is bounded by $|X(ns, nt)|$. Thus Proposition 3.21 together with the union bound implies that there exist constants $a_0, a_1 > 0$, such that for any $n \geq N$ (here we take $n \geq N$ to avoid worrying about rounding error),

$$\mathbb{P}\left(\sup_{\substack{s, t \in [0, T] \\ |s-t| \leq T/N}} \|Z^n(t) - Z^n(s)\|_1 \geq 2^{-m}\right) \leq 2Na_0 \exp(-a_1 T^{-1/2} N^{1/2} 2^{-m}).$$

By choosing $N = N_{T,\varepsilon,m}$ sufficiently large, depending on T , ε , and m , we can make this probability at most $\varepsilon 2^{-m}$ for all $n \geq N_{T,\varepsilon,m}$. By starting with $\delta_m = T/N_{T,\varepsilon,m}$, and then possibly shrinking δ_m , we can arrange that

$$\mathbb{P}\left(\sup_{\substack{s,t \in [0,T] \\ |s-t| \leq \delta_m}} \|Z^n(t) - Z^n(s)\|_1 \geq 2^{-m}\right) \leq \varepsilon 2^{-m}$$

for all $n \in \mathbb{N}$, not just $n \geq N_{T,\varepsilon,m}$. By the union bound, we obtain that for each $n \in \mathbb{N}$, it holds except on an event of probability at most ε that, whenever $m \in \mathbb{N}$ and $s, t \in [0, T]$ with $|t - s| \leq \delta_m$, we have $\|Z^n(t) - Z^n(s)\|_1 < 2^{-m}$. By the Arzelá-Ascoli theorem, we obtain tightness of the paths $Z^n|_{[0,\infty)}$ in the topology of uniform convergence on compacts. Tightness of the sequence of the full processes (defined on \mathbb{R}) follows from translation invariance. \square

Proof of Theorem 1.15. By Lemma 3.24 and Prokhorov's theorem, for any sequence of n 's tending to infinity, there exists a subsequence n_k and a random continuous path $Z = (U, V) : \mathbb{R} \rightarrow \mathbb{R}^2$ such that, as k tends to infinity, $Z^{n_k}|_{[0,\infty)}$ converges to Z in law in the topology of uniform convergence on compacts.

Next we show that the law of Z is uniquely determined (independently of the subsequence). Consider any subsequence n_k for which $Z^{n_k}|_{[0,\infty)}$ converges in law (in the topology of uniform convergence on compacts). By the Skorokhod representation theorem, we can find a coupling of a sequence of random words (X^{n_k}) , each with the law of X , such that if we define Z^{n_k} with X^{n_k} in place of X , then a.s. as k tends to infinity, Z^{n_k} converges to Z uniformly on compact subsets of $[0, \infty)$.

Fix real numbers $t_0 < t_2 < \dots < t_N$. For $j \in [1, N]_{\mathbb{Z}}$ and $k \in \mathbb{N}$, let

$$\Upsilon_j^{n_k} := n_k^{-1/2} \vec{d}(X^{n_k}(\lfloor t_{j-1} n_k \rfloor + 1, \lfloor t_j n_k \rfloor)).$$

Observe that $\Upsilon_j^{n_k}$ differs from $Z^{n_k}(t_j) - Z^{n_k}(t_{j-1})$ in either coordinate by at most

$$2n_k^{-1/2} + n_k^{-1/2} \mathcal{N}_{\text{d|S}}(X^{n_k}(\lfloor t_{j-1} n_k \rfloor + 1, \lfloor t_j n_k \rfloor)).$$

By Lemma 3.4 and Proposition 3.21, the latter quantity tends to 0 in probability as k tends to infinity, and since by the Skorokhod coupling $Z^{n_k} \rightarrow Z$, in fact $\Upsilon_j^{n_k} \rightarrow Z(t_j) - Z(t_{j-1})$ a.s. for each $j \in [1, N]_{\mathbb{Z}}$. The random variables $(\Upsilon_j^{n_k} : j \in [1, N]_{\mathbb{Z}})$ are independent, and by translation invariance of the law of X together with our above observation about $\Upsilon_j^{n_k}$, the law of each $\Upsilon_j^{n_k}$ converges as k tends to infinity to the law of $Z(t_j) - Z(t_{j-1})$. Hence the increments $Z(t_j) - Z(t_{j-1})$ are independent and each has the same law as $Z(t_j - t_{j-1})$, i.e., Z has independent stationary increments.

By Proposition 3.21 and the Vitali convergence theorem, we find that for each $t \geq 0$, the first and second moments of the coordinates of $Z^{n_k}(t)$ converge to the corresponding quantities for $Z(t)$. Convergence of the expectations implies that $\mathbb{E}(Z(t)) = 0$ for each $t \geq 0$, and convergence of variances implies with Proposition 3.21 implies $Z(t)$ has finite variance. Thus Z is a continuous Lévy process with independent stationary mean-zero increments, so Z must be a two-dimensional Brownian motion with $Z(0) = 0$ and some variances, covariance, and zero drift.

Since $\mathcal{C}(X'(1, n))$ is a simple random walk,

$$\lim_{k \rightarrow \infty} \text{Var} \left(n_k^{-1/2} \mathcal{C}(X'(1, n_k)) \right) = 1,$$

and by Proposition 3.17,

$$\lim_{k \rightarrow \infty} \text{Var} \left(n_k^{-1/2} \mathcal{D}(X'(1, n_k)) \right) = 1 + (p + q)\chi.$$

Furthermore, the conditional law of X given $\mathcal{C}(X(1, m))$ for all $m \in \mathbb{N}$ is invariant under the involution operation (3.3) and this operation changes the sign of $\mathcal{D}(X'(1, m))$, so

$$\text{Cov}(\mathcal{C}(X'(1, m)), \mathcal{D}(X'(1, m))) = 0, \quad \forall m \in \mathbb{N}.$$

Equivalently,

$$\begin{aligned} \text{Var}(U(t) + V(t)) &= 1, \\ \text{Var}(U(t) - V(t)) &= 1 + (p + q)\chi, \\ \text{Cov}(U(t) + V(t), U(t) - V(t)) &= 0. \end{aligned}$$

Recalling the formula (1.11), this implies that Z must be as in (1.18).

If the full sequence $\{Z^n\}_{n \in \mathbb{N}}$ failed to converge uniformly on compact subsets of $[0, \infty)$ to the law of Z , then there would be a subsequence bounded away from the law of Z . But by Prokhorov's theorem and the argument above, there would be a subsubsequence converging in law to Z , a contradiction. Hence the full sequence $\{Z^n\}_{n \in \mathbb{N}}$ converges uniformly on compact subsets of $[0, \infty)$, and thus on compact subsets of \mathbb{R} by translation invariance.

The statement that $\chi = 2$ when $q = 0$ is established in Lemma 3.14. We thus obtain the statement of the theorem when $p_S = p$, $p_d = q$, and $p_F = p_e = 0$. By Corollary 1.13 we obtain the statement of the theorem in general. \square

4 Open problems

Here we list some open problems related to the model studied in this paper, some of which were mentioned in the text.

1. Compute the value of the constant χ in Theorem 1.15 when $p_d - p_e \neq 0$ ($z \neq 1$). Figure 9 shows computer simulations of the value of χ and the corresponding value of κ in terms of y and z .
2. Prove an infinite-volume peanosphere scaling limit result similar to Theorems 1.14 and 1.15 in the case when $p_S \neq 1$ and $p_d - p_e < 0$ ($y > 0$ and $z \in (0, 1)$) or when $p_F - p_S > 0$ and $p_d - p_e \neq 0$ ($y > 1$ and $z \neq 1$).
3. Prove a scaling limit result for the walk $Z^n|_{[0, 2]}$ conditioned on the event that the reduced word $X(1, 2n) = \emptyset$ (which encodes a finite-volume spanning-tree-decorated random planar map), possibly just in the case when $p_d = p_e = p_F = 0$. See [GS15, Thm. 1.8] for an analogous result in the case when $p_d = p_e = p_S = 0$ and $p_F \in [0, 1/2)$.

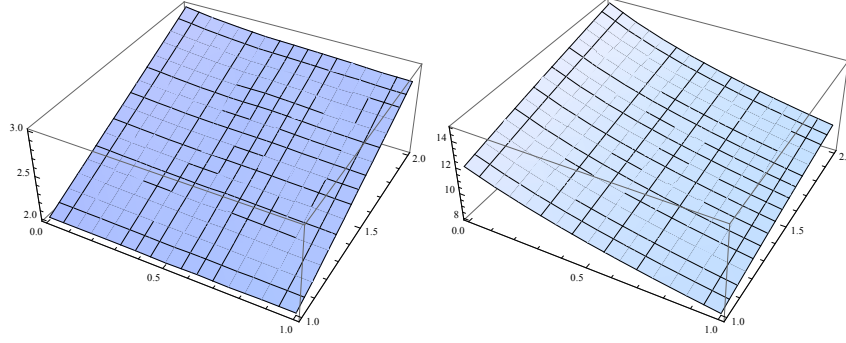


Figure 9: Experimental plots for χ and κ as a function of $(y, z) \in [0, 1] \times [1, 2]$.

4. Prove a scaling limit result for the bending loop model of Remark 1.10. In particular is there an encoding of this model in terms of a model on words analogous to the one studied in this paper?
5. For many statistical mechanics models on random planar maps which converge in the scaling limit to SLE_κ -decorated LQG for some value of $\kappa > 0$, it is expected that the same model on a Euclidean lattice converges in the scaling limit to SLE_κ . Recall that for peanosphere scaling limit results, the correlation of the Brownian motion is given by $-\cos(4\pi/\kappa)$. In light of Lemma 1.11 and Theorems 1.14 and 1.15, it is therefore natural to make the following conjecture, which expands the conjecture in [KW16].

Conjecture 4.1. *Let Λ be either the triangular, hexagonal, or square lattice and suppose that either $y = 0$ and $z > 0$ or $y \in [0, 1]$ and $z \in [1, \infty)$. Let T be a spanning tree on Λ sampled according to the law (1.10) (defined, e.g., by taking a limit of the law (1.10) on large finite sub-graphs of Λ) and let λ be its associated Peano curve. Then λ converges in law in the scaling limit to SLE_κ , where $\kappa \geq 8$ is chosen so*

$$-\cos\left(\frac{4\pi}{\kappa}\right) = \begin{cases} -\frac{z}{1+z}, & y = 0 \\ -\frac{(z-y)\chi}{(y+1)(z+1) + (z-y)\chi} & (y, z) \in [0, 1] \times [1, \infty), \end{cases}$$

where χ (depending on y and z) is as in Theorem 1.15.

Prove this conjecture. The case when $y = z = 1$ corresponds to the uniform spanning tree and has been treated in [LSW04]. The case $(y, z) = (1 + \sqrt{2}, 1)$ corresponds to the FK–Ising model and has recently been addressed in [KS15].

A Basic properties of the burger model

Recall that a word in $\mathcal{W}(\Theta)$ is called *reduced* if all of its orders (i.e., elements of $\{\mathbf{C}, \mathbf{H}, \mathbf{F}, \mathbf{S}\}$), \mathbf{d} 's, and \mathbf{e} 's lie to the left of all of its \mathbf{h} 's and \mathbf{c} 's.

Lemma A.1. *The reduction operation of Definition 1.2 is well-defined, i.e., for every finite word $x \in \mathcal{W}(\Theta)$ there is a unique reduced word $\mathcal{R}(x)$ which is equivalent to x modulo the relations (1.4) and (1.6).*

Proof. The proof follows the same argument as [She16, Prop. 2.1]. To prove existence, we define $\mathcal{R}(x)$ by induction on the length $|x|$ of the word, as follows. We set $\mathcal{R}(x) = \emptyset$ if $|x| = 0$ (i.e., $x = \emptyset$). Suppose now that $n \in \mathbb{N}$, a reduced word $\mathcal{R}(x)$ equivalent to x has been constructed for all words \tilde{x} with $|\tilde{x}| \leq n - 1$, and we are given a word $x = x_1 \dots x_n$ with $|x| = n$.

If we set $\tilde{x} = x_1 \dots x_{n-1}$, then the reduced word $\mathcal{R}(\tilde{x})$ consists of a (possibly empty) word U consisting of orders, d's, and e's followed by a (possibly empty) word u consisting of h's and c's. If either $u = \emptyset$ or $x_n \in \{\mathbf{h}, \mathbf{c}\}$, then we set $\mathcal{R}(x) = \mathcal{R}(\tilde{x})x_n$ to get a reduced word equivalent to x . If $u \neq \emptyset$ and $x_n \in \{\mathbf{d}, \mathbf{e}\}$, we replace x_n by an h or c using (1.6), then append this burger to the end of $\mathcal{R}(\tilde{x})$ to get $\mathcal{R}(x)$. If $u \neq \emptyset$ and $x_n = \mathbf{H}$, then using (1.4), we either match x_n to the rightmost h in u or (if u has no h's) we move x_n to the position between U and u to form a reduced word $\mathcal{R}(x)$ equivalent to x . We define $\mathcal{R}(x)$ similarly if $x_n = \mathbf{C}$. Finally, if $u \neq \emptyset$ and $x_n \in \{\mathbf{F}, \mathbf{S}\}$, we replace x_n by an H or C (depending on the rightmost burger in u) using (1.4) then proceed as in the case $x_n \in \{\mathbf{H}, \mathbf{C}\}$ to once again get a reduced word equivalent to x .

To prove uniqueness, we observe that $\mathcal{R}(x)$, as defined above, equals x if x is already reduced. The relations (1.4) and (1.6) list several pairs (w, w') of two-letter or zero-letter words which are defined to be equivalent. If (w, w') is one of these pairs of words (e.g., $(w, w') = (\mathbf{hS}, \mathbf{hC})$) then the above construction shows that replacing an instance of the word w which appears anywhere in the word x with the word w' in the same position will have no effect on the reduced word $\mathcal{R}(x)$. If x and x' are two words which are equivalent modulo the relations (1.4) and (1.6), then x' can be obtained from x by performing finitely many replacement operations of the above form. Consequently, $\mathcal{R}(x) = \mathcal{R}(x')$ whenever x and x' are equivalent. In particular, if x and x' are both reduced, then $x = x'$. \square

We make frequent use of the following consequence of uniqueness:

Lemma A.2. *For any finite words $x, y \in \mathcal{W}(\Theta)$,*

$$\mathcal{R}(\mathcal{R}(x)\mathcal{R}(y)) = \mathcal{R}(xy). \quad (\text{A.1})$$

Proof. If $x, y, x', y' \in \mathcal{W}(\Theta)$ are finite words such that x is equivalent to x' and y is equivalent to y' modulo the relations (1.4) and (1.6), then xy is equivalent to $x'y'$ modulo these relations (since one can apply the relations to convert x to x' , then apply the relations to convert y to y' separately). Consequently, $\mathcal{R}(x)\mathcal{R}(y)$ and hence also $\mathcal{R}(\mathcal{R}(x)\mathcal{R}(y))$ is equivalent to xy modulo our relations. Since $\mathcal{R}(\mathcal{R}(x)\mathcal{R}(y))$ is a reduced word, (A.1) follows from the uniqueness statement in Lemma A.1. \square

The above lemmas were purely combinatorial, but the following ones also use the stochastic properties of the burger model. Fix $\vec{p} = (p_{\mathbf{F}}, p_{\mathbf{S}}, p_{\mathbf{d}}, p_{\mathbf{e}}) \in \mathcal{P}$ and let X be a bi-infinite word whose symbols $\{X_i\}_{i \in \mathbb{Z}}$ are i.i.d. with the law (1.7), as in Section 1.3. We prove that a.s. each symbol in X has a match and that the bi-infinite identification $X' = \mathcal{I}(X)$ is well-defined and contains only symbols in Θ_0 , i.e., that the objects of Definitions 1.5 and 1.6 are well-defined.

Lemma A.3. *Let $m \in \mathbb{N}$ and let P_m be the smallest $j \in \mathbb{N}$ for which the net burger count satisfies $\mathcal{C}(X(-j, -1)) = m$. If $X_{-P_m} \in \{\mathbf{h}, \mathbf{c}\}$, then for any $n \in [1, P_m]_{\mathbb{Z}}$, every symbol in $X_{-P_m} \cdots X_{-n}$ is identified, and the reduced word $X(-P_m, -n)$ contains at least one burger.*

Proof. A.s. P_m is well-defined and finite. Suppose $n \in [1, P_m]_{\mathbb{Z}}$. By definition of P_m , $\mathcal{C}(X(-P_m, -1)) = m$ and $\mathcal{C}(X(-n+1, -1)) \leq m-1$, but we also have

$$\begin{aligned} \mathcal{C}(X(-P_m, -1)) &= \mathcal{C}(X_{-P_m} \cdots X_{-1}) \\ &= \mathcal{C}(X_{-P_m} \cdots X_{-n}) + \mathcal{C}(X_{-n+1} \cdots X_{-1}) \\ &= \mathcal{C}(X(-P_m, -n)) + \mathcal{C}(X(-n+1, -1)), \end{aligned}$$

so $\mathcal{C}(X(-P_m, -n)) \geq 1$ and hence the reduced word $X(-P_m, -n)$ contains at least one burger.

We now show by induction on $P_m - n \in [0, P_m - 1]_{\mathbb{Z}}$ that every symbol in $X_{-P_m} \cdots X_{-n}$ is identified. Since $X_{-P_m} \in \{\mathbf{h}, \mathbf{c}\}$, the claim is true for $n = P_m$. If the claim is true for $P_m \geq n \geq 2$, then since $X(-P_m, -n)$ contains a burger, each of which is by induction identified, the next symbol X_{-n+1} becomes identified, completing the induction. \square

Corollary A.4. *Almost surely each symbol in X is identified, i.e., the bi-infinite identified word $X' = \mathcal{I}(X)$ is well-defined and contains only symbols in Θ_0 .*

Proof. For $m \in \mathbb{N}$, by the strong Markov property, conditional on the stopping time P_m , the word $X_{-P_{m+1}} \cdots X_{-P_m-1}$ is independent of $X_{-P_m} \cdots X_{-1}$, and in particular $\mathbb{P}(X_{-P_{m+1}} \in \{\mathbf{h}, \mathbf{c}\}) = 1 - p_d - p_e > 0$. Almost surely for some m this event occurs, in which case, by Lemma A.3, the symbol X_0 is identified. By translation invariance, a.s. every symbol in X is identified. \square

Lemma A.5. *Almost surely,*

$$\lim_{n \rightarrow \infty} \mathcal{N}_{\mathbf{h}}(X(-n, -1)) = \lim_{n \rightarrow \infty} \mathcal{N}_{\mathbf{c}}(X(-n, -1)) = \infty. \quad (\text{A.2})$$

Proof. Suppose $r \in \mathbb{N}$. Let $\ell_0 = 0$, and for $k \in \mathbb{N}$ inductively define

$$\ell_k := \min \{t \in \mathbb{N} : \mathcal{C}(X_{-t} \cdots X_{-\ell_{k-1}-1}) = 2\ell_{k-1} + 2r\}.$$

Since $j \mapsto \mathcal{C}(X(-j, -1))$ is a simple random walk, a.s. each of the times ℓ_k is finite.

By the strong Markov property, conditional on ℓ_{k-1} , the word $X_{-\ell_k} \cdots X_{-\ell_{k-1}-1}$ is independent of $X_{-\ell_{k-1}} \cdots X_{-1}$. With probability $1 - p_d - p_e$ the symbol $X_{-\ell_k}$ is \mathbf{h} or \mathbf{c} , in which case, by Lemma A.3, each symbol of $X_{-\ell_k} \cdots X_{-\ell_{k-1}-1}$ is identified. Conditional on this event, since $X_{-\ell_k} \cdots X_{-\ell_{k-1}-1}$ contains at least $2\ell_{k-1} + 2r$ burgers, by symmetry it must be that with probability at least $1/2$ it contains at least $\ell_{k-1} + r$ hamburgers. But $X(-\ell_{k-1}, -1)$ contains no more than ℓ_{k-1} hamburger orders, so $X(-\ell_k, -1)$ contains at least r hamburgers. So regardless of $X_{-\ell_{k-1}} \cdots X_{-1}$, with probability at least $(1 - p_d - p_e)/2$, the reduced word $X(-\ell_k, -1)$ contains at least r hamburgers. Almost surely this event will occur for some k . Since $\mathcal{N}_{\mathbf{h}}(X(-n, -1))$ is monotone increasing in n , almost surely $\liminf_{n \rightarrow \infty} \mathcal{N}_{\mathbf{h}}(X(-n, -1)) \geq r$. Since $r \in \mathbb{N}$ was arbitrary, (A.2) follows. \square

Lemma A.6. *Almost surely, for each $i \in \mathbb{Z}$ there is a unique $\phi(i) \in \mathbb{Z}$, called the match of i , such that $X_{\phi(i)}$ cancels out X_i in the reduced word $X(\phi(i), i)$ if $\phi(i) < i$ or $X(i, \phi(i))$ if $i < \phi(i)$.*

Proof. If X'_i is an order, from Lemma A.5 we see that a.s. it has a match in $X_{-n} \cdots X_i$ for some sufficiently large n (using translation invariance).

Next we use the Mass-Transport Principle (see e.g. [LP16, Chapter 8]). We let each order send one unit of mass to the burger that it consumes. Since each order consumes a burger by the above paragraph and each letter is an order with probability $1/2$, the expected mass out of a letter is $1/2$. Since this mass transport rule is translation invariant, by the Mass-Transport Principle, the expected mass into a letter is $1/2$. Thus any given letter is with probability $1/2$ a burger that is consumed. But it is a burger with probability $1/2$, so a.s. all burgers are consumed. \square

References

- [BBG12] G. Borot, J. Bouttier, and E. Guitter. More on the $O(n)$ model on random maps via nested loops: loops with bending energy. *J. Phys. A* 45(27):275206, 32, 2012, 1202.5521. MR2947230
- [Ber08a] O. Bernardi. A characterization of the Tutte polynomial via combinatorial embeddings. *Ann. Comb.* 12(2):139–153, 2008, math/0608057. MR2428901
- [Ber08b] O. Bernardi. Tutte polynomial, subgraphs, orientations and sandpile model: new connections via embeddings. *Electron. J. Combin.* 15(1):Research Paper 109, 53, 2008, math/0612003. MR2438581
- [BLR17] N. Berestycki, B. t. Laslier, and G. Ray. Critical exponents on Fortuin-Kasteleyn weighted planar maps. *Comm. Math. Phys.* 355(2):427–462, 2017, 1502.00450. MR3681382
- [BS01] I. Benjamini and O. Schramm. Recurrence of distributional limits of finite planar graphs. *Electron. J. Probab.* 6:no. 23, 13 pp. (electronic), 2001, 0011019. MR1873300
- [Che17] L. Chen. Basic properties of the infinite critical-FK random map. *Ann. Inst. Henri Poincaré D* 4(3):245–271, 2017, 1502.01013. MR3713017
- [Cou14] J. Courtiel. A general notion of activity for the Tutte polynomial. *ArXiv e-prints*, December 2014, 1412.2081.
- [DGK00] P. Di Francesco, E. Guitter, and C. Kristjansen. Integrable 2D Lorentzian gravity and random walks. *Nuclear Physics B* 567(3):515–553, 2000. hep-th/9907084.
- [DMS14] B. Duplantier, J. Miller, and S. Sheffield. Liouville quantum gravity as a mating of trees. *ArXiv e-prints*, September 2014, 1409.7055.

- [Eva85] S. N. Evans. On the Hausdorff dimension of Brownian cone points. *Math. Proc. Cambridge Philos. Soc.* 98(2):343–353, 1985. [MR795899](#)
- [FK72] C. M. Fortuin and P. W. Kasteleyn. On the random-cluster model. I. Introduction and relation to other models. *Physica* 57:536–564, 1972. [MR0359655](#)
- [GHMS17] E. Gwynne, N. Holden, J. Miller, and X. Sun. Brownian motion correlation in the peanosphere for $\kappa > 8$. *Ann. Inst. Henri Poincaré Probab. Stat.* 53(4):1866–1889, 2017, [1510.04687](#). [MR3729638](#)
- [GHS16] E. Gwynne, N. Holden, and X. Sun. Joint scaling limit of a bipolar-oriented triangulation and its dual in the peanosphere sense. *ArXiv e-prints*, March 2016, [1603.01194](#).
- [GM17] E. Gwynne and J. Miller. Convergence of the topology of critical Fortuin-Kasteleyn planar maps to that of CLE_κ on a Liouville quantum surface. In preparation, 2017.
- [GMS17a] E. Gwynne, C. Mao, and X. Sun. Scaling limits for the critical Fortuin-Kasteleyn model on a random planar map I: cone times. *Annales de l’Institut Henri Poincaré*, 2017, [1502.00546](#).
- [GMS17b] E. Gwynne, J. Miller, and S. Sheffield. The Tutte embedding of the mated-CRT map converges to Liouville quantum gravity. *ArXiv e-prints*, May 2017, [1705.11161](#).
- [GS15] E. Gwynne and X. Sun. Scaling limits for the critical Fortuin-Kasteleyn model on a random planar map III: finite volume case. *ArXiv e-prints*, October 2015, [1510.06346](#).
- [GS17] E. Gwynne and X. Sun. Scaling limits for the critical Fortuin-Kasteleyn model on a random planar map II: local estimates and empty reduced word exponent. *Electronic Journal of Probability* 22:Paper No. 45, 1–56, 2017, [1505.03375](#).
- [KMSW15] R. Kenyon, J. Miller, S. Sheffield, and D. B. Wilson. Bipolar orientations on planar maps and SLE_{12} . *ArXiv e-prints*, November 2015, [1511.04068](#).
- [KS15] A. Kemppainen and S. Smirnov. Conformal invariance of boundary touching loops of FK Ising model. *ArXiv e-prints*, September 2015, [1509.08858](#).
- [KW16] A. Kassel and D. B. Wilson. Active spanning trees and Schramm–Loewner evolution. *Phys. Rev. E* 93:062121, Jun 2016, [1512.09122](#).
- [Le 13] J.-F. Le Gall. Uniqueness and universality of the Brownian map. *Ann. Probab.* 41(4):2880–2960, 2013, [1105.4842](#). [MR3112934](#)
- [Le 14] J.-F. Le Gall. Random geometry on the sphere. *Proceedings of the ICM*, 2014, [1403.7943](#).

- [LL10] G. F. Lawler and V. Limic. *Random walk: a modern introduction*, volume 123 of *Cambridge Studies in Advanced Mathematics*. Cambridge University Press, Cambridge, 2010. [MR2677157](#)
- [LP16] R. Lyons and Y. Peres. *Probability on Trees and Networks*, volume 42 of *Cambridge Series in Statistical and Probabilistic Mathematics*. Cambridge University Press, New York, 2016. Available at <http://pages.iu.edu/~rdlyons/>. [MR3616205](#)
- [LSW04] G. F. Lawler, O. Schramm, and W. Werner. Conformal invariance of planar loop-erased random walks and uniform spanning trees. *Ann. Probab.* 32(1B):939–995, 2004, [math/0112234](#). [MR2044671](#)
- [LSW17] Y. Li, X. Sun, and S. S. Watson. Schnyder woods, SLE(16), and Liouville quantum gravity. *ArXiv e-prints*, May 2017, [1705.03573](#).
- [Mie09] G. Miermont. Random maps and their scaling limits. In *Fractal geometry and stochastics IV*, volume 61 of *Progr. Probab.*, pages 197–224. Birkhäuser Verlag, Basel, 2009. [MR2762678](#)
- [Mie13] G. Miermont. The Brownian map is the scaling limit of uniform random plane quadrangulations. *Acta Math.* 210(2):319–401, 2013, [1104.1606](#). [MR3070569](#)
- [MS15a] J. Miller and S. Sheffield. An axiomatic characterization of the Brownian map. *ArXiv e-prints*, June 2015, [1506.03806](#).
- [MS15b] J. Miller and S. Sheffield. Liouville quantum gravity and the Brownian map I: The QLE(8/3,0) metric. *ArXiv e-prints*, July 2015, [1507.00719](#).
- [MS15c] J. Miller and S. Sheffield. Liouville quantum gravity spheres as matings of finite-diameter trees. *ArXiv e-prints*, June 2015, [1506.03804](#).
- [MS16a] J. Miller and S. Sheffield. Imaginary geometry III: reversibility of SLE $_{\kappa}$ for $\kappa \in (4, 8)$. *Ann. Math.* 184(2):455–486, 2016, [1201.1498](#).
- [MS16b] J. Miller and S. Sheffield. Liouville quantum gravity and the Brownian map II: geodesics and continuity of the embedding. *ArXiv e-prints*, May 2016, [1605.03563](#).
- [MS16c] J. Miller and S. Sheffield. Liouville quantum gravity and the Brownian map III: the conformal structure is determined. *ArXiv e-prints*, August 2016, [1608.05391](#).
- [MS16d] J. Miller and S. Sheffield. Imaginary geometry I: interacting SLEs. *Probab. Theory Related Fields* 164(3-4):553–705, 2016, [1201.1496](#). [MR3477777](#)
- [MS16e] J. Miller and S. Sheffield. Imaginary geometry II: Reversibility of SLE $_{\kappa}(\rho_1; \rho_2)$ for $\kappa \in (0, 4)$. *Ann. Probab.* 44(3):1647–1722, 2016, [1201.1497](#). [MR3502592](#)
- [MS16f] J. Miller and S. Sheffield. Quantum Loewner evolution. *Duke Math. J.* 165(17):3241–3378, 2016, [1312.5745](#). [MR3572845](#)

- [MS17] J. Miller and S. Sheffield. Imaginary geometry IV: interior rays, whole-plane reversibility, and space-filling trees. *Probab. Theory Related Fields* 169(3-4):729–869, 2017, [1302.4738](#). [MR3719057](#)
- [Mul67] R. C. Mullin. On the enumeration of tree-rooted maps. *Canad. J. Math.* 19:174–183, 1967. [MR0205882](#)
- [RS05] S. Rohde and O. Schramm. Basic properties of SLE. *Ann. of Math. (2)* 161(2):883–924, 2005, [math/0106036](#). [MR2153402](#)
- [Sch00] O. Schramm. Scaling limits of loop-erased random walks and uniform spanning trees. *Israel J. Math.* 118:221–288, 2000, [math/9904022](#). [MR1776084](#)
- [She09] S. Sheffield. Exploration trees and conformal loop ensembles. *Duke Math. J.* 147(1):79–129, 2009, [math/0609167](#). [MR2494457](#)
- [She16] S. Sheffield. Quantum gravity and inventory accumulation. *Ann. Probab.* 44(6):3804–3848, 2016, [1108.2241](#). [MR3572324](#)
- [Shi85] M. Shimura. Excursions in a cone for two-dimensional Brownian motion. *J. Math. Kyoto Univ.* 25(3):433–443, 1985. [MR807490](#)
- [Whi02] W. Whitt. *Stochastic-process limits*. Springer Series in Operations Research. Springer-Verlag, New York, 2002. An introduction to stochastic-process limits and their application to queues. [MR1876437](#)