

Open ASEP in the Weakly Asymmetric Regime

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Abstract

We consider ASEP on a bounded interval and on a half-line with sources and sinks. On the full line, Bertini and Giacomin in 1997 proved convergence under weakly asymmetric scaling of the height function to the solution of the KPZ equation. We prove here that under similar weakly asymmetric scaling of the sources and sinks as well, the bounded interval ASEP height function converges to the KPZ equation on the unit interval with Neumann boundary conditions on both sides (different parameter for each side), and likewise for the half-line ASEP to KPZ on a half-line. This result can be interpreted as showing that the KPZ equation arises at the triple critical point (maximal current / high density / low density) of the open ASEP. © 2018 Wiley Periodicals, Inc.

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1 Introduction

The open asymmetric simple exclusion process (ASEP) is a default paradigm in statistical physics for studying transport in systems in contact with reservoirs that keep the two ends of the system at different local densities. When particles are driven in one direction, the system approaches a nonequilibrium steady state. In other words, while there is an invariant measure, there is also a net flux of particles flowing through the system, and there is no time-reversal symmetry. Consequently, standard notions of equilibrium statistical mechanics do not apply. Open ASEP enjoys certain exact relations (such as the matrix product ansatz, or solvability

through the Bethe ansatz) which has made it a profitable model through which to develop new physical predictions and theories for a broader class of models that share the same type of steady state behaviors.

In this paper we demonstrate how the stochastic heat equation (SHE) / Kardar-Parisi-Zhang (KPZ) equation / stochastic Burgers equation (SBE) on an interval with various types of boundary conditions arise from the open ASEP under a certain precise weakly asymmetric scaling limit of the model and its parameters (we also study the half-line versions of these equations with boundary conditions at the origin).

While, to our knowledge, this connection between the KPZ equation and the open ASEP is new to both the mathematics and physics literature, it is certainly not without precedent. For ASEP on the whole line, the seminal work of Bertini and Giacomin [7] demonstrated convergence of ASEP under a similar scaling to the KPZ equation on \mathbb{R} via a discrete Cole-Hopf transformation introduced by Gärtner [36]. They assumed near-equilibrium initial data, and narrow wedge initial data was treated in [1]. There has since been great progress in expanding this weak universality of the KPZ equation on \mathbb{R} using methods similar to those of Bertini and Giacomin [15, 16, 19, 57], as well as using energy solutions (introduced by Assing [2] as well as Jara and Gonçalves [38] and proved to be unique by Gubinelli and Perkowski [46]) [26, 38–40, 45, 47, 48], and Hairer’s regularity structures (introduced in [50]) [51, 52]. These works have dealt entirely with the KPZ equation on \mathbb{R} or the torus ($[0, 1]$ with periodic boundary conditions).

We deal here with the KPZ equation with Neumann boundary condition on $[0, 1]$ (with generically different derivatives on the two ends). For the SHE this corresponds with Robin boundary conditions and for the SBE with Dirichlet boundary conditions. None of these boundary conditions make a priori sense (solutions of SHE and KPZ are not differentiable, and SBE does not take values). This is resolved through defining a mild solution and equivalent martingale problem for the SHE with boundary conditions.

In our work, we tried to follow the scheme of Bertini and Giacomin, though were quickly forced to overcome some significant new complexities. First we apply the Gärtner (or microscopic Hopf-Cole) transform [27, 36], which yields a microscopic SHE for the exponentiated ASEP height function. Parametrizing the open ASEP boundary conditions by two effective density parameters (one for each side of the interval) leads to Robin boundary conditions on the microscopic SHE. The challenge then becomes to prove convergence to the corresponding continuum SHE. This is done by first showing tightness and then identifying (through a martingale problem formulation) the limit. Tightness reduces to fine estimates about the heat kernel (and its derivatives and various weighted integrals) for the microscopic SHE, which, in our case, become much more involved and nonstandard than in the periodic or full-line case (for instance, they require us to invoke Sturm-Liouville theory and a delicate method of images estimate). In identifying the limiting martingale problem, Bertini and Giacomin discovered a key identity (proposition 4.8

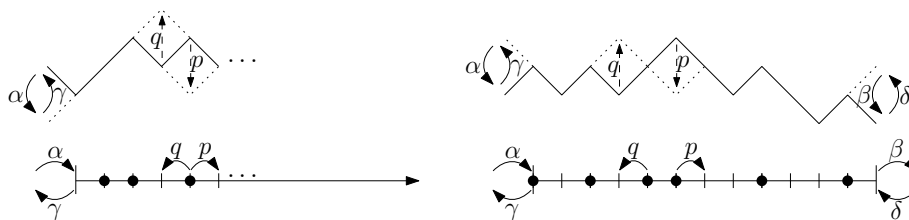


FIGURE 1.1. The half-line and open ASEP (left and right, respectively) drawn in terms of particles and height functions (bottom and top, respectively). In open ASEP, particles inside jump left and right (i.e., according to clocks of exponential distribution with this rate) at rates q and p ; particles are inserted (provided the destination is empty) from the left (respectively, right) at rate α (respectively, δ); particles are removed (provided the removal site is occupied) from the left (respectively, right) at rate γ (respectively, β). In half-line ASEP, there is only a boundary on the left, and the right goes on infinitely.

and lemma A.1 in [7]) for the quadratic martingale that identifies the white noise. In our case of bounded intervals, that identity does not hold. Instead we find (via a new method using Green's functions) an approximate version which suffices—see Lemma 5.8 and Proposition 5.1.

Presently the other above-mentioned KPZ equation convergence methods have not been developed into the context we consider here. These other methods are less reliant upon the exact structure of the underlying particle system, so it would be nice to see them developed so as to prove universality of the type of convergence we have shown here for the particular model of open ASEP. The energy solutions method only applies in stationarity (or for initial data with finite entropy with respect to the stationary measure) and relies upon certain inputs from hydrodynamic theory that may be more complicated in this setting since the open ASEP generally lacks product-invariant measures. However, in a special one-parameter case (see Remark 2.9) the invariant measure reduces to product Bernoulli. Upon sending this present paper out for comments, we learned that [42] are completing a work in which they develop the energy solutions approach to study the KPZ limit of open ASEP in the special subcase when the invariant measure is product Bernoulli, and the effective density is exactly $\frac{1}{2}$. This is a special point in the two-dimensional family of parameters we consider here. For generic choices of our parameters, the invariant measure is not of product form.

1.1 Existing Open ASEP Results

Before defining the open/half-line ASEP and stating our main convergence theorem, let us try to put our work into the context of known results. Much of this discussion involves results from the physics literature that have not received a mathematically rigorous treatment.

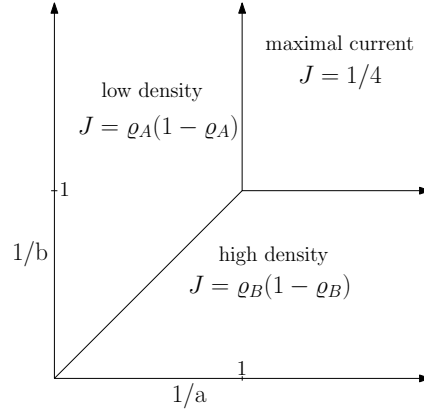


FIGURE 1.2. The phase diagram for open ASEP current.

The open ASEP is a test tube for developing predictions for the behavior of systems with nonequilibrium steady states. The model is illustrated in Figure 1.1 and defined mathematically in Definition 2.1. The matrix product ansatz [21, 24, 43, 59, 67, 69] is the main method that has been utilized in studying the open ASEP steady state (see also Derrida's ICM proceedings [20] for further references). Open ASEP has a highly nontrivial steady state, though it is still possible to compute various asymptotics about stationary expectations, such as that of the current. Let us give the phase diagram (see Figure 1.2 for an illustration), which was first discovered and proved by Liggett [61] in the special parametrization we take below in Definition 2.8.

Consider open ASEP on an interval of length N and with boundary parameters $\alpha, \beta, \gamma, \delta$ (see Figure 1.1). Define the average current J_N to be $(p - q)^{-1}$ times the expected value in the steady state of the net number of particles to enter the system from the left in a unit of time. As $N \rightarrow \infty$, this approaches a current J that can be calculated through the matrix product ansatz. It depends on two density parameters¹ \mathbf{a} and \mathbf{b} given by the formulas (see, for instance, [43])

$$(1.1) \quad \begin{aligned} \mathbf{a} &= \frac{p - q - \alpha + \gamma + \sqrt{(p - q - \alpha + \gamma)^2 + 4\alpha\gamma}}{2\alpha}, \\ \mathbf{b} &= \frac{p - q - \beta + \delta + \sqrt{(p - q - \beta + \delta)^2 + 4\beta\delta}}{2\beta}. \end{aligned}$$

We assume here (and throughout the paper) that $q < p$. Note that \mathbf{a} only includes the left source/sink parameters (α and γ) and \mathbf{b} those on the right (β and δ). The

¹In the main text of the paper we will work with centered occupation variables taking values of ± 1 instead of uncentered occupancy variables that take values 1 (particle) and 0 (hole). The notion of density is with respect to uncentered occupancy variables (though it is easily affinely transformed into centered ones) and can be thought of as the expected value of the uncentered occupancy variable.

effective left and right particle densities are given (respectively) by

$$(1.2) \quad \varrho_A = 1/(1 + \mathbf{a}) \quad \text{and} \quad \varrho_B = b/(1 + \mathbf{b}).$$

The low-density phase arises when $\frac{1}{\mathbf{a}} < \frac{1}{\mathbf{b}} \wedge 1$ (we use $1/\mathbf{a}$ and $1/\mathbf{b}$ to match the traditional appearance of the phase diagram) and corresponds to a phase with a current $J = \varrho_A(1 - \varrho_A)$. In this phase, there is a low ($< 1/2$) effective density of particles near the left boundary, and the bulk dynamics and the right boundary easily transport this density through the system. The high-density phase arises when $\frac{1}{\mathbf{b}} < \frac{1}{\mathbf{a}} \wedge 1$ and corresponds to a phase with a current $J = \varrho_B(1 - \varrho_B)$. In this phase, there is a high ($> 1/2$) effective density of particles near the right boundary, and this causes an overall jamming of the system so that the bulk dynamics and left boundary cannot operate at their maximum efficiency. The maximal current phase, when $1/\mathbf{a}, 1/\mathbf{b} > 1$, corresponds with flux $J = 1/4$ and arises when the system operates at maximum efficiency. The effective densities at the left boundary are more than $1/2$ and at the right boundary are less than $1/2$. Since the density in the bulk that propagates particles at the maximal speed is $1/2$, these boundaries do not inhibit transport.

The above phase diagram describes the law of large number behavior for current. On the other hand, less is understood about the finer scale nature of fluctuations in currents (and in associated height functions) for open ASEP.

The main result of our paper proves that the KPZ equation arises at the triple point between the low density, high density, and maximal current phases. In particular, for system size N , we scale $q = 1/2 - 1/(2\sqrt{N})$ and $p = 1/2 + 1/(2\sqrt{N})$ (approximately) and tune $\alpha, \beta, \gamma, \delta$ in such a way that the left and right densities also equal $1/2 + O(1/\sqrt{N})$ (equivalently, \mathbf{a} and \mathbf{b} equal $1 + O(1/\sqrt{N})$). Then, in time of order N^2 and scaling fluctuations by order $1/\sqrt{N}$ we come upon the KPZ equation on $[0, 1]$ with Neumann boundary conditions. This result is stated as Theorem 2.18 (see also Theorem 2.17 for the half-line version of this result).

Returning to an earlier result, the hydrodynamic limit (i.e., shape theorem or law of large numbers) has been proved for open ASEP under the symmetric and weakly asymmetric scaling limits. In both cases, the boundary parameters $\alpha, \beta, \gamma, \delta$ are fixed. In the symmetric case the internal left and right jump rate parameters q and p are chosen to be equal. Weakly asymmetric scaling in the hydrodynamic limit literature means something different than above—one takes $q = 1/2 - 1/N$ and $p = 1/2 + 1/N$. Under this type of weakly asymmetric scaling, the arguments in [18, 36] can prove convergence of the open ASEP density field to the viscous Burgers equation on $[0, 1]$ with fixed boundary density. (In fact, the proof presented in [18, 36] is on the entire line, while another paper [55] is on the torus; however, the arguments can be adapted to the boundary driven case; see [30, 31, 41, 54] for hydrodynamic limit results of different boundary driven models.) To our knowledge, no results have been proved regarding the totally asymmetric (or partially asymmetric) open ASEP in which one might expect (as in the full-line case) the inviscid Burgers equation.

The fluctuations of the particle density field around its limiting behavior is understood in the weakly asymmetric case. When started out of equilibrium, [41] and [58] prove convergence in the (respectively) weakly asymmetric and symmetric cases to generalized Ornstein-Uhlenbeck processes (see also [22] for an earlier physics prediction of these results). In the symmetric case, and with extremely depressed boundary rates, [34] has also considered the limiting nonequilibrium fluctuations.

In the stationary state, the (single time) spatial fluctuations for the weakly asymmetric system are Gaussian [22], while those of the totally asymmetric case are known and non-Gaussian [23]. In both cases, the stationary state fluctuations are given in terms of the sum of two processes; however, there seems to be no way to scale the weakly asymmetric limit process to the totally asymmetric limit process. It is compelling to speculate that, as in the full-line case [1], the KPZ equation limit we consider here provides the mechanism for crossing over between these two types of steady state behaviors. There is presently no description for the steady state of the KPZ equation on $[0, 1]$ with Neumann boundary condition, and it would be interesting to understand whether the matrix product ansatz (or the combinatorial methods used in [12, 13]) has a meaningful limit under the scalings considered in this paper.

Returning to the current, the variance of the actual number of particles to enter the system in time t (centered by the expected value tJ_N) grows like $\Delta_N t$ where Δ_N is called the diffusion coefficient. [25] argues that in the high- and low-density phases, the diffusion coefficient has a positive limit as $N \rightarrow \infty$, whereas in the maximal current regime, it decays like $1/\sqrt{N}$ —a fact that they say is consistent with the scaling exponents for the KPZ universality class. There do not seem to be any further results developing this perspective. So, in many ways the fluctuation profile for the various portions of the phase diagram (assuming they are dictated solely by these three phases) remains quite mysterious.

There is a special case of open ASEP in which all boundary parameters are set to 0 and the system becomes closed with reflection at the boundaries. In this case, [57] recently studied various hydrodynamic and fluctuation limit theorems. That version of ASEP admits a reversible invariant measure due to the particle conservation. When started very far from equilibrium, the fluctuations of the system's height function under Bertini-Giacomin-style weak asymmetry scaling converges to the KPZ equation on the *full line*. That convergence holds for a finite amount of time and then suddenly breaks once the invariant measure is reached. This should be contrasted to our present results in which the KPZ equation is on $[0, 1]$ and remains valid for all time. At a technical level, [57] follows the Bertini-Giacomin approach. The reflecting boundary conditions for ASEP correspond with the Dirichlet boundary condition for the associated microscopic heat equation, for which heat kernel estimates are easily accessible through comparison with the analogous full-line heat kernel.

It is also worth mentioning that besides the law of large numbers and fluctuation behaviors, the large-deviation principle is well-studied for open ASEP. A combination of matrix product ansatz and Bethe ansatz methods have led to a detailed description of the rate function and minimizers for open ASEP in various parameter regimes—see, for example, [6, 8, 10, 17, 29, 43] and references therein.

1.2 Existing Half-Line ASEP Results

A special case of the half-line ASEP was introduced by Liggett in [60] where the creation rate $\alpha = p\lambda$ and the annihilation rate $\gamma = q(1 - \lambda)$. Here, λ controls the overall density at the origin, which equals $(1 - \lambda)/\lambda$. Liggett showed that the long-term behavior of this process has a phase transition: if the rate $\lambda < 1/2$ and the original density is below a critical value, the stationary measure is a product measure with density λ ; otherwise, if $\lambda > 1/2$ the stationary measure is spatially correlated and behaves like product measure with density $1/2$ at infinity. Grosskinsky [44] studied, using the matrix product ansatz, the correlations of these stationary measures, and [28] also employed the matrix product ansatz to study the large deviations for this system.

Using methods from Bethe ansatz and symmetric functions, there has also been some work on studying asymptotics of the fluctuations of the number of particles inserted into the system. In the totally asymmetric case ($p = 1, \alpha > 0, q = \gamma = 0$) [3] proved a KPZ universality class limit theorem with either square root / cube root fluctuations and Gaussian / GUE Tracy-Widom-type statistics (depending on the exact strength of α). This result was predated by work of Baik and Rains [4], which demonstrated a similar phenomenon for the polynuclear growth model. In the partially asymmetric reflecting case ($\alpha = \gamma = 0$) Tracy-Widom derived explicit formulas for the configuration probabilities in [68], though no asymptotics have been accessible from these formulas as of yet (one generally expects from the universality belief that the same sort of dichotomy between square root and cube root fluctuations exists for the general partially asymmetric case. For the KPZ equation itself, [11, 49] have employed the nonrigorous replica Bethe ansatz methods to derive a similar set of KPZ universality class limit theorems as was shown in the TASEP. There are also some formulas derived for the half-space log-gamma polymer in [64], though they have yet to produce any asymptotic results.

After completing the present paper, [5] proved an exact one-point distribution formula for half-line ASEP under the parametrization considered herein, when $A = -1/2$. While that value of A is outside the class we consider (where $A, B \geq 0$), [65] has been able to extend our analysis to all real A, B . Combining those two works proves the first one-point distribution formula and cube root fluctuations for the half-space KPZ equation in the special case when $A = -1/2$.

Outline

The introduction above contains a brief survey of the open and half-line ASEP, as well as a number of relevant works in the literature. Section 2 contains the

definitions of the models, the choices of parameters and scalings, assumptions on initial data, and finally the main results of this paper (Theorems 2.17 and 2.18), which prove convergence to the KPZ equation with Neumann boundary conditions. Section 3 contains the Gärtner transform of open and half-line ASEP, and the resulting microscopic SHE on the half-line and a bounded interval with Robin boundary condition. This simple boundary condition occurs on a two-dimensional subspace of the four boundary parameters $\alpha, \beta, \gamma, \delta$ when they are parametrized by two density parameters. Section 4 spends a significant amount of space developing a variety of bounds for the discrete heat kernel (and its discrete derivative) associated with Laplacian with Robin boundary conditions. Some of these bounds require us to derive a generalized method of images, as well as to apply methods from Sturm-Liouville theory. This section then uses the heat kernel bounds to prove tightness of the rescaled microscopic SHE. Section 5 contains the identification of all limit points with solutions to the martingale problems for the continuum SHE with Robin boundary conditions (and then shows the uniqueness of the martingale problem and identifies it with the mild solution, thus proving our main theorems). The section also contains a crucial cancellation in the form of Proposition 5.1 and Lemma 5.8, without which the identification of the limiting quadratic martingale would not be evident.

2 Definitions of the Models and the Main Results

The aim of this paper is to study the *weakly* asymmetric simple exclusion processes (ASEP) with open boundaries and their KPZ equation limits. We will discuss two cases: the first case is an ASEP model on the *half-line* (i.e., positive integers) with an open boundary at the origin, which will be called ASEP-H; the second case is an ASEP model on a *bounded interval* with open boundaries at the two ends, which will be called ASEP-B.

We start with the definition of the asymmetric simple exclusion process (ASEP) on the positive integers. See Figure 1.1 for an illustration of these definitions.

DEFINITION 2.1 (Half line ASEP or “ASEP-H”). Fix four nonnegative real parameters $p, q, \alpha, \gamma \geq 0$. Let $\eta(x) \in \{\pm 1\}$ denote the centered occupation variable at site $x \in \mathbb{Z}_{>0} = \{1, 2, 3, \dots\}$. The site $x \in \mathbb{Z}_{>0}$ is said to be occupied by a particle if $\eta(x) = 1$, and empty if $\eta(x) = -1$. The half-line ASEP is a continuous-time Markov process on the state space $\{(\eta(x))_{x \in \mathbb{Z}_{>0}} \in \{\pm 1\}^{\mathbb{Z}_{>0}}\}$. The state η_t at time t evolves according to the following dynamics (we could state the generator, though we do not make explicit use of it and hence avoid writing it): at any given time $t \in [0, \infty)$ and $x \in \mathbb{Z}_{>0}$ a particle jumps from the site x to the site $x + 1$ at exponential rate

$$(2.1) \quad c^R(\eta_t, x) = \frac{p}{4}(1 + \eta_t(x))(1 - \eta_t(x + 1))$$

and from site $x + 1$ to site x at exponential rate

$$(2.2) \quad c^L(\eta_t, x) = \frac{q}{4}(1 - \eta_t(x))(1 + \eta_t(x + 1));$$

and if $x = 1$, a particle is created or annihilated at the site 1 at exponential rates

$$r_A^+(\eta_t(1)) = \frac{\alpha}{2}(1 - \eta_t(1)) \quad \text{and} \quad r_A^-(\eta_t(1)) = \frac{\gamma}{2}(1 + \eta_t(1)),$$

respectively. All these events of jumps and creations/annihilations are independent of each other.

Such a process can be constructed by the standard procedures as in [62]. Note that the rate $r_A^\pm(\eta(1))$ is such that particles are not allowed to be created at 1 if $x = 1$ is already occupied, and not allowed to be annihilated from 1 if there is no particle at $x = 1$. Also, the number of particles is not preserved. One may also imagine a pile of infinitely many particles at site 0, and particles are allowed to jump between 0 and 1 at rates $r_A^\pm(\eta(1))$.

We now define ASEP on the lattice with two-sided open boundaries.

DEFINITION 2.2 (ASEP on bounded interval or ‘‘ASEP-B’’). Fix six nonnegative real parameters $p, q, \alpha, \beta, \gamma, \delta \geq 0$. Let $\eta(x) \in \{\pm 1\}$ denote the centered particle occupation variable at site $x \in \Lambda_N \stackrel{\text{def}}{=} \{1, 2, \dots, N\}$; x is occupied by a particle if $\eta(x) = 1$ and empty if $\eta(x) = -1$. The ASEP on the bounded interval Λ_N is a continuous-time Markov process on the state space $\{(\eta(x))_{x \in \Lambda_N} \in \{\pm 1\}^{\Lambda_N}\}$. The state η_t at time t evolves according to the following dynamics: at any given time $t \in [0, \infty)$ and $x \in \Lambda_N$, a particle jumps from site x to site $x + 1$ (respectively, from site $x + 1$ to site x) at exponential rate $c^R(\eta_t, x)$ (respectively, $c^L(\eta_t, x)$) where $c^R(\eta_t, x)$ and $c^L(\eta_t, x)$ are defined in (2.1) and (2.2).

In addition to the jumps, a particle is created (respectively, annihilated) at the boundary site $x = 1$ at exponential rates r_A^+ (respectively, r_A^-), and a particle is created (respectively, annihilated) at the boundary site $x = N$ at exponential rates r_B^+ (respectively, r_B^-), where

$$\begin{aligned} r_A^+(\eta(1)) &= \frac{\alpha}{2}(1 - \eta(1)), & r_B^+(\eta(N)) &= \frac{\delta}{2}(1 - \eta(N)), \\ r_A^-(\eta(1)) &= \frac{\gamma}{2}(1 + \eta(1)), & r_B^-(\eta(N)) &= \frac{\beta}{2}(1 + \eta(N)). \end{aligned}$$

All these events of jumps and creations/annihilations are independent of each other.

DEFINITION 2.3 (Height functions). We define the *height function* $h_t(x)$ for $x \in \mathbb{Z}_{\geq 0} = \{0, 1, 2, \dots\}$ associated with the ASEP-H and the *height function* $h_t(x)$ for $x \in \Lambda_N \cup \{0\}$ associated with the ASEP-B as follows:

- Let $h_t(0)$ be two times the *net* number of particles that are *removed* (i.e., the number of particles annihilated minus the number of particles created)

from the site $x = 1$ during the time interval $[0, t]$. (In particular $h_0(0) = 0$.)

- For $x > 0$, let

$$h_t(x) \stackrel{\text{def}}{=} h_t(0) + \sum_{y=1}^x \eta_t(y).$$

According to this definition we have $\nabla^+ h_t(x) = \eta_t(x+1)$ for every $t \geq 0$ and $x \geq 0$ (respectively, $0 \leq x < N$) for ASEP-H (respectively, ASEP-B). Here the forward and backward discrete gradients are defined as

$$(2.3) \quad \nabla^+ f(x) \stackrel{\text{def}}{=} f(x+1) - f(x), \quad \nabla^- f(x) \stackrel{\text{def}}{=} f(x-1) - f(x).$$

Let us also record our convention for the discrete Laplacian,

$$(2.4) \quad \Delta f(x) \stackrel{\text{def}}{=} f(x-1) - 2f(x) + f(x+1).$$

Note that the above definition of $h_t(x)$ is such that for both models, creating or annihilating particles at $x = 1$ only affects the value of $h_t(0)$ and none of the values of $h_t(x)$ for $x > 0$. Also, for ASEP-B, creating or annihilating particles at $x = N$ only affects the value of $h_t(N)$ but none of the values of $h_t(x)$ for $x < N$. Finally, the internal particle jumps can only affect the values of $h_t(x)$ for x in the “bulk”, i.e., $x > 0$ for ASEP-H or $0 < x < N$ for ASEP-B. We can in fact equivalently consider the following interface growth model. See Figure 1.1 for an illustration of this connection.

DEFINITION 2.4 (Solid on solid (SOS) models with moving boundaries). The SOS model on $\mathbb{Z}_{\geq 0}$ with a moving boundary is a jump Markov process defined on the state space

$$\{h \in \mathbb{Z}^{\mathbb{Z}_{\geq 0}} : |\nabla^+ h(x)| = 1, \forall x \in \mathbb{Z}_{\geq 0}\}.$$

The dynamics for the height function h_t at time t are as follows. For each $x \in \mathbb{Z}_{> 0}$, if $\Delta h_t(x) = 2$ then $h_t(x)$ increases by 2 at exponential rate q , and if $\Delta h_t(x) = -2$ then $h_t(x)$ decreases by 2 at exponential rate p . Moreover, if $\nabla^+ h_t(0) = 1$, then $h_t(0)$ increases by 2 at exponential rate γ , and if $\nabla^+ h_t(0) = -1$ then $h_t(0)$ decreases by 2 at exponential rate α . All the jumps are independent.

The SOS model on $\Lambda_N \cup \{0\}$ with moving boundaries is a jump Markov process defined on the state space

$$\{h \in \mathbb{Z}^{\Lambda_N \cup \{0\}} : |\nabla^+ h(x)| = 1, \forall x \in \{0, 1, \dots, N-1\}\}.$$

For $x \in \{0, 1, \dots, N\}$, the $h_t(x)$ increases or decreases according to the same rule as for SOS on $\mathbb{Z}_{\geq 0}$ above, together with the rule that if $\nabla^- h_t(N) = 1$ then $h_t(N)$ increases by 2 at exponential rate β and if $\nabla^- h_t(N) = -1$ then $h_t(N)$ decreases by 2 at exponential rate δ .

In this article, we will show via the Cole-Hopf/Gärtner transformation that under the weak asymmetry scaling, the height function converges to the Cole-Hopf solution to the KPZ equation:

$$(2.5) \quad \partial_T H = \frac{1}{2} \Delta H + \frac{1}{2} (\partial_X H)^2 + \dot{W},$$

with Neumann (generically inhomogeneous) boundary conditions, where \dot{W} is the space-time white noise (formally, $\mathbf{E}(\dot{W}_T(X)\dot{W}_S(X')) = \delta(T-S)\delta(X-X')$). Here, the Cole-Hopf solution to (2.5) with generically inhomogeneous Neumann boundary condition is defined by $H_T(X) = \log \mathcal{Z}_T(X)$ where $\mathcal{Z} \in C([0, \infty), C(\mathbb{R}_+))$ for ASEP-H or $\mathcal{Z} \in C([0, \infty), C([0, 1]))$ for ASEP-B is the mild solution (defined immediately below) to the stochastic heat equation (SHE)

$$(2.6) \quad \partial_T \mathcal{Z} = \frac{1}{2} \Delta \mathcal{Z} + \mathcal{Z} \dot{W},$$

with Robin boundary conditions.

DEFINITION 2.5. We say that $\mathcal{Z}_T(X)$ is a *mild solution* to SHE (2.6) on \mathbb{R}_+ starting from initial data $\mathcal{Z}_0(\bullet) \in C(\mathbb{R}_+)$ satisfying a Robin boundary condition with parameter $A \in \mathbb{R}$

$$(2.7) \quad \partial_X \mathcal{Z}_T(X)|_{X=0} = A \mathcal{Z}_T(0) \quad \forall T > 0$$

if \mathcal{Z}_T is adapted to the filtration $\sigma\{\mathcal{Z}_0, W|_{[0, T]}\}$ and

$$(2.8) \quad \begin{aligned} \mathcal{Z}_T(X) &= \int_0^\infty \mathcal{P}_T^R(X, Y) \mathcal{Z}_0(Y) dY \\ &+ \int_0^T \int_0^\infty \mathcal{P}_{T-S}^R(X, Y) \mathcal{Z}_S(Y) dW_S(dY) \end{aligned}$$

where the last integral is the Itô integral with respect to the cylindrical Wiener process W , and \mathcal{P}^R is the heat kernel satisfying the Robin boundary condition

$$\partial_X \mathcal{P}_T^R(X, Y)|_{X=0} = A \mathcal{P}_T^R(0, Y) \quad \forall T > 0, Y > 0.$$

We say that $\mathcal{Z}_T(X)$ is a *mild solution* to SHE (2.6) on $[0, 1]$ starting from initial data $\mathcal{Z}_0(\bullet) \in C([0, 1])$ with Robin boundary conditions with parameters (A, B)

$$(2.9) \quad \partial_X \mathcal{Z}_T(X)|_{X=0} = A \mathcal{Z}_T(0), \quad \partial_X \mathcal{Z}_T(X)|_{X=1} = -B \mathcal{Z}_T(1) \quad \forall T > 0,$$

if $\mathcal{Z}_T(X)$ is adapted to the filtration $\sigma\{\mathcal{Z}_0, W|_{[0, T]}\}$ and satisfies (2.8) with the integration domain $[0, \infty)$ replaced by $[0, 1]$ and \mathcal{P}^R satisfying

$$\partial_X \mathcal{P}_T^R(X, Y)|_{X=0} = A \mathcal{P}_T^R(0, Y), \quad \partial_X \mathcal{P}_T^R(X, Y)|_{X=1} = -B \mathcal{P}_T^R(1, Y),$$

for all $T > 0, Y \in (0, 1)$.

Remark 2.6. At least on the formal level, the Robin boundary condition for SHE (2.6) corresponds to (via the Cole-Hopf transformation $H_T(X) = \log \mathcal{Z}_T(X)$) the inhomogeneous Neumann boundary condition for KPZ (2.5). For instance, if

$\partial_X \mathcal{Z}_T(0) = A \mathcal{Z}_T(0)$, then $\partial_X H_T(0) = A$. In fact, this also corresponds to the inhomogeneous Dirichlet boundary condition for the stochastic Burgers equation

$$(2.10) \quad \partial_T u = \frac{1}{2} \partial_X^2 u + \frac{1}{2} \partial_X (u^2) + \partial_X \dot{W}$$

where $u \stackrel{\text{def}}{=} \partial_X H$ with boundary condition $u_T(0) = A$. Also, $\partial_X \mathcal{Z}_T(1) = -B \mathcal{Z}_T(1)$ corresponds to $\partial_X H_T(1) = -B$ and $u_T(1) = -B$. This is only formal because the KPZ equation is not differentiable and the stochastic Burgers equation does not take function-valued solutions. This is why we have defined the Robin boundary condition in terms of the heat kernel and not the derivative of \mathcal{Z} . After completing our present paper, [37] provided solution theories to a class of singular SPDEs with boundary conditions, which includes the KPZ equation with Neumann boundary conditions.

PROPOSITION 2.7. *Let $A, B \geq 0$, $\bar{T} > 0$, and I be the interval $[0, 1]$ or \mathbb{R}_+ . Given an initial data $\mathcal{Z}_0 \in C(I)$ satisfying*

$$\sup_{X \in I} e^{-bX} \mathbf{E}(\mathcal{Z}_0(X)^p) < \infty$$

for some $b = b_p$ for all $p > 0$, there exists a mild solution to SHE (2.6) in $C([0, \bar{T}], C(I))$ with Robin condition (2.7) if $I = \mathbb{R}_+$ or (2.9) if $I = [0, 1]$. The mild solution is unique in the class of adapted processes satisfying

$$(2.11) \quad \sup_{T \in [0, \bar{T}]} \sup_{X \in I} e^{-aX} \mathbf{E}(\mathcal{Z}_T(X)^2) < \infty$$

for some $a > 0$. Finally, assuming that \mathcal{Z}_0 is given by a nonnegative measure, $\mathcal{Z}_T(X)$ is almost surely strictly positive for all $X \in I$ and all $T \in [0, \bar{T}]$.

PROOF. Existence and uniqueness of mild solutions to the SHE (2.6) was shown in [70, thm. 3.2] in the case of a bounded interval $[0, 1]$ with Neumann boundary condition. The case of bounded interval $[0, 1]$ or half-line \mathbb{R}_+ with Robin condition follows *mutatis mutandis*, so we only sketch the proof here.

First of all, we note that although [70, thm. 3.2] stated that the ‘‘weak solution’’ (in the PDE sense, i.e., the equation (2.6) when integrating both sides against smooth test functions with corresponding boundary condition) exists and is unique, the proof there actually showed the existence and uniqueness of the *mild* solution.

The proof in [70] only used the following properties of the (Neumann) heat kernel \mathcal{P}^N : for each $\bar{T} > 0$ there exists a constant $C(\bar{T})$ such that for all $T < \bar{T}$,

$$\mathcal{P}_T^N(X, Y) \leq \frac{C(\bar{T})}{\sqrt{T}} e^{-\frac{|Y-X|^2}{4T}}$$

(see [70, (3.7)], which actually also had a factor e^{-T} due to an extra mass term considered in the equation therein, but it is never used as long as one does not care about integrability at $T = \infty$.) For the Robin heat kernel \mathcal{P}^R , we can still prove the above bound: see Lemma 4.3. Note that our bound has $2T$ in the exponent

instead of $4T$ because our diffusive term is $\frac{1}{2}\Delta\mathcal{Z}$. Therefore the existence and uniqueness follows for $I = [0, 1]$. For $I = \mathbb{R}_+$, the only tweak in the proof is that one needs the following bound:

$$\int_0^\infty \mathcal{P}_T^R(X, Y) e^{aY} dY \leq C(\bar{T}) e^{aX}, \quad X > 0,$$

which is easy to show.

The positivity result follows exactly as in Mueller's full-line proof [63], and we do not reproduce it here. \square

We turn now to describe how the above continuum equations arise from our discrete ASEP models. The following parametrizations of the boundary parameters will be assumed throughout the remainder of this paper. The reason we make this choice of parametrization is that it enables us to write the Cole-Hopf/Gärtner transform of ASEP in terms of a discrete SHE with Robin boundary conditions.

DEFINITION 2.8 (Boundary parametrization). For ASEP-B we parametrize α and γ by a single parameter $\mu_A \in [\sqrt{q/p}, \sqrt{p/q}]$ and β and δ by a single parameter $\mu_B \in [\sqrt{q/p}, \sqrt{p/q}]$ as follows:

$$(2.12) \quad \begin{aligned} \alpha &= \frac{p^{3/2}(\sqrt{p} - \mu_A \sqrt{q})}{p - q}, & \beta &= \frac{p^{3/2}(\sqrt{p} - \mu_B \sqrt{q})}{p - q}, \\ \gamma &= \frac{q^{3/2}(\sqrt{q} - \mu_A \sqrt{p})}{q - p}, & \delta &= \frac{q^{3/2}(\sqrt{q} - \mu_B \sqrt{p})}{q - p}. \end{aligned}$$

For ASEP-H we parametrize using the above formulas for α and γ .

Remark 2.9. The condition on $\mu_A, \mu_B \in [\sqrt{q/p}, \sqrt{p/q}]$ is necessary and sufficient for all rates to be positive. These choices of rates satisfy the simple relations

$$\frac{\alpha}{p} + \frac{\gamma}{q} = 1, \quad \frac{\beta}{p} + \frac{\delta}{q} = 1.$$

In fact, these relations are exactly the relations assumed by Liggett in [60,61] when considering ASEP on the half-line and bounded interval. In the open ASEP case, plugging these parametrizations into (1.1) yields simple formulas for

$$a = \frac{\mu_A \sqrt{pq} - q}{p - \mu_A \sqrt{pq}}, \quad b = \frac{\mu_B \sqrt{pq} - q}{p - \mu_B \sqrt{pq}}.$$

By (1.2), we may compute the effective densities $\rho_A = \alpha/p$ and $\rho_B = \delta/q$. In the special one-parameter subcase when the effective densities on the left and right are equal, i.e., $\rho_A = \rho_B$ in (1.2), then one may check (see, e.g., [42]) that the invariant measure for open ASEP is a product of Bernoulli random variables with density ρ_A . The equality $\rho_A = \rho_B$ amounts to the relationships between μ_A and μ_B that

$$\mu_A = \frac{p + q - \mu_B \sqrt{pq}}{\sqrt{pq}}.$$

The special case when $\varrho_A = \varrho_B = 1/2$ (and the invariant measure is product Bernoulli with parameter $1/2$) arises when $\mu_A = \mu_B = \frac{p+q}{2\sqrt{pq}}$.

We now define the weakly asymmetric scaling that will be assumed throughout the remainder of this paper (whether stated explicitly or not). All functions, e.g., $Z_t(x)$, and all parameters will implicitly depend on ε throughout (though our notation will generally not make this explicit).

DEFINITION 2.10 (Weakly asymmetric scaling). Let $\varepsilon > 0$ be a small enough so that all parameters defined below are positive. Define, for $\varepsilon > 0$

$$(2.13) \quad q = \frac{1}{2}e^{-\sqrt{\varepsilon}} \quad \text{and} \quad p = \frac{1}{2}e^{\sqrt{\varepsilon}}.$$

In other words the particles jump to the right at a higher rate $p > 1/2$ than to the left at rate $q < 1/2$ (which is opposite the convention in [7]). For the ASEP-H, the above definition is sufficient. For the ASEP-B, we further assume that $\varepsilon = 1/N$ where N is the size of the bounded interval. The boundary parameters are likewise weakly scaled so that $\mu_A = 1 - \varepsilon A$ and $\mu_B = 1 - \varepsilon B$ for some constants $A, B \geq 0$. See Remark 3.2 for more on the restriction that $A, B \geq 0$.

Remark 2.11. Given the weakly asymmetric scaling, we have the following expansions (in ε small) for our model parameters:

$$\begin{aligned} p &= \frac{1}{2} + \frac{1}{2}\sqrt{\varepsilon} + O(\varepsilon), & q &= \frac{1}{2} - \frac{1}{2}\sqrt{\varepsilon} + O(\varepsilon), \\ \mu_A &= 1 - A\varepsilon, & \mu_B &= 1 - B\varepsilon, \\ \alpha &= \frac{1}{4} + \left(\frac{3}{8} + \frac{1}{4}A\right)\sqrt{\varepsilon} + O(\varepsilon), & \beta &= \frac{1}{4} + \left(\frac{3}{8} + \frac{1}{4}B\right)\sqrt{\varepsilon} + O(\varepsilon), \\ \gamma &= \frac{1}{4} - \left(\frac{3}{8} + \frac{1}{4}A\right)\sqrt{\varepsilon} + O(\varepsilon), & \delta &= \frac{1}{4} - \left(\frac{3}{8} + \frac{1}{4}B\right)\sqrt{\varepsilon} + O(\varepsilon). \end{aligned}$$

We also have expansions of the open ASEP phase diagram parameters and the associated effective densities:

$$\begin{aligned} a &= 1 - (1 + 2A)\sqrt{\varepsilon} + O(\varepsilon), & b &= 1 - (1 + 2B)\sqrt{\varepsilon} + O(\varepsilon), \\ \varrho_A &= \frac{1}{2} + \left(\frac{1}{4} + \frac{1}{2}A\right)\sqrt{\varepsilon} + O(\varepsilon^{3/2}), & \varrho_B &= \frac{1}{2} - \left(\frac{1}{4} + \frac{1}{2}B\right)\sqrt{\varepsilon} + O(\varepsilon^{3/2}). \end{aligned}$$

From this, one sees that we are exactly tuning into a $\varepsilon^{1/2}$ -scale window around the triple critical point in the open ASEP phase diagram in Figure 1.2.

We now define the microscopic Cole-Hopf/Gärtner transformed process [36] defined as follows:

DEFINITION 2.12 (Gärtner transformation). For $x \in \mathbb{Z}_{\geq 0}$ in the case of ASEP-H, or $x \in \{0\} \cup \Lambda_N$ in the case of ASEP-B, let

$$(2.14) \quad Z_t(x) \stackrel{\text{def}}{=} e^{-\lambda h_t(x) + \nu t}$$

where

$$(2.15) \quad \lambda \stackrel{\text{def}}{=} \frac{1}{2} \log \frac{q}{p} \quad \nu \stackrel{\text{def}}{=} p + q - 2\sqrt{pq}.$$

Note that $\lambda < 0$. We extend $Z_t(x)$ by linear interpolation to all $x \in \mathbb{R}_+$ for ASEP-H, and to all $x \in [0, N]$ for ASEP-B, so that $Z \in D([0, \infty), C(I))$ with $I = \mathbb{R}_+$ or $I = [0, N]$, which is the space of $C(I)$ -valued, right-continuous-with-left-limits processes.

Remark 2.13. We remark that [7] adopted the convention that $q > p$ and the height profile grows upward, while we follow the convention here that $q < p$ and the height profile grows downward. In [7] the height profile *subtracting* a positive drift term scales to the KPZ equation (2.5) with a negative sign in front of the nonlinearity, while in our case here since $-\lambda > 0$, $\nu > 0$, the height profile *plus* a positive drift term scales to the KPZ equation (2.5) with a positive sign in front of the nonlinearity.

Let $\|f_t(x)\|_n \stackrel{\text{def}}{=} (\mathbf{E}|f_t(x)|^n)^{1/n}$ denote the L^n norm. Following [7], we consider the following near-equilibrium initial conditions for ASEP-H. For the purpose of the assumption we include a superscript ε in h_t and Z_t to explicitly identify the ε -dependence of the model. This ε will generally be suppressed in what follows.

ASSUMPTION 2.14 (Near equilibrium initial conditions for ASEP-H). *We assume that the sequence of ε -indexed initial height functions $\{h_0^\varepsilon(\cdot)\}_\varepsilon$ or the Gärtner transformed functions $\{Z_0^\varepsilon(\cdot)\}_\varepsilon$ (defined via h_0^ε as in (2.14)) associated with the ASEP-H is “near equilibrium,” namely for any $\alpha \in (0, \frac{1}{2})$ and every $n \in \mathbb{N}$ there exist finite constants C and a such that for every $x, x' \in \mathbb{R}_+$ and every $\varepsilon > 0$ one has*

$$(2.16) \quad \|Z_0^\varepsilon(x)\|_n \leq C e^{a\varepsilon x},$$

$$(2.17) \quad \|Z_0^\varepsilon(x) - Z_0^\varepsilon(x')\|_n \leq C(\varepsilon|x - x'|)^\alpha e^{a\varepsilon(x+x')}.$$

For ASEP-B we do not need to impose any growth condition at infinity, but we still have the following bound that is uniform in ε . Recall that ε and N are related by $\varepsilon = 1/N$.

ASSUMPTION 2.15 (Initial condition for ASEP-B). *For ASEP-B on $\{0\} \cup \Lambda_N$, we assume that a sequence of ε -indexed initial height functions $\{h_0^\varepsilon(\cdot)\}_\varepsilon$ or the Gärtner transformed functions $\{Z_0^\varepsilon(\cdot)\}_\varepsilon$ (defined via h_0^ε as in (2.14)) associated with the ASEP-B is such that, for any $\alpha \in (0, \frac{1}{2})$ and every $n \in \mathbb{N}$, there exists a finite constant C so that for every $x, x' \in [0, N]$ and every $\varepsilon > 0$ one has*

$$(2.18) \quad \|Z_0^\varepsilon(x)\|_n \leq C,$$

$$(2.19) \quad \|Z_0^\varepsilon(x) - Z_0^\varepsilon(x')\|_n \leq C(\varepsilon|x - x'|)^\alpha.$$

Given $\bar{T} > 0$, we endow the space $D([0, \bar{T}], C(\mathbb{R}_+))$ and $D([0, \bar{T}], C([0, 1]))$ with the Skorokhod topology and the space $C(\mathbb{R}_+)$ with the topology of uniform convergence on compact sets, and use \Rightarrow to denote weak convergence of probability laws. When processes converge in these topologies, we refer to it as convergence as a space-time process.

DEFINITION 2.16 (Scaled processes). The scaled processes \mathcal{Z} are defined by

$$(2.20) \quad \mathcal{Z}_T^\varepsilon(X) \stackrel{\text{def}}{=} Z_{\varepsilon^{-2}T}(\varepsilon^{-1}X) \quad T \in [0, \bar{T}]$$

where Z (which depends on ε through the model parameters) is defined in (2.14). For ASEP-H, (2.20) is defined for all $X \in \mathbb{R}_+$ and $\mathcal{Z}^\varepsilon \in D([0, \bar{T}], C(\mathbb{R}_+))$; for ASEP-B, (2.20) is defined for all $X \in [0, 1]$ and $\mathcal{Z}^\varepsilon \in D([0, \bar{T}], C([0, 1]))$.

We state our main theorems for the convergence of these scaled processes. Recall the definition of the mild solution to the SHE, the choice of ASEP parameters, weakly asymmetric scaling, and scaled processes given in the above definitions.

THEOREM 2.17. *Given any initial conditions Z_0^ε satisfying Assumption 2.14 such that $Z_0^\varepsilon \Rightarrow \mathcal{Z}^{ic}$ as $\varepsilon \rightarrow 0$, where $\mathcal{Z}^{ic} \in C(\mathbb{R}_+)$, then $\mathcal{Z}^\varepsilon \Rightarrow \mathcal{Z}$ as a space-time process as $\varepsilon \rightarrow 0$, where \mathcal{Z} is the unique mild solution to the SHE (2.6) on \mathbb{R}_+ from the initial data \mathcal{Z}^{ic} satisfying the Robin boundary condition with parameter A .*

THEOREM 2.18. *Given any initial conditions Z_0^ε satisfying Assumption 2.15 such that $Z_0^\varepsilon \Rightarrow \mathcal{Z}^{ic}$ as $\varepsilon \rightarrow 0$, where $\mathcal{Z}^{ic} \in C([0, 1])$, then $\mathcal{Z}^\varepsilon \Rightarrow \mathcal{Z}$ is a space-time process, as $\varepsilon \rightarrow 0$, where \mathcal{Z} is the unique mild solution to SHE (2.6) on $[0, 1]$ from the initial data \mathcal{Z}^{ic} satisfying the Robin boundary condition with parameters (A, B) .*

3 Hopf-Cole/Gärtner Transform and the Microscopic SHE

We derive the microscopic SHE satisfied by the Hopf-Cole/Gärtner transformation [27, 36] of ASEP-H and ASEP-B (recall Definition 2.12). For open ASEP, this transformation and the associated boundary parametrization that yields Robin boundary conditions was given recently in [41] (though we were unaware of that work until it was kindly pointed out to us by Gonçalves).

We will assume the choices of parameters given in Definition 2.8 and the weakly asymmetric scaling given in Definition 2.10.

LEMMA 3.1. *For ASEP-H, $Z_t(x)$ defined in (2.14) satisfies*

$$(3.1) \quad dZ_t(x) = \frac{1}{2} \Delta Z_t(x) dt + dM_t(x)$$

for all $x \in \mathbb{Z}_{\geq 0}$, with Robin boundary condition at $x = 0$:

$$(3.2) \quad \mu_A(Z(-1) - Z(0)) + (1 - \mu_A)Z(-1) = 0, \quad \text{i.e., } Z(-1) = \mu_A Z(0),$$

where $M_\bullet(x)$ for $x \in \mathbb{Z}_{\geq 0}$ are martingales with bracket processes

$$\frac{d}{dt} \langle M(x), M(y) \rangle_t = 0, \quad x \neq y,$$

$$\begin{aligned} \frac{d}{dt} \langle M(x), M(x) \rangle_t = \\ \left(\left(\frac{q}{p} - 1 \right)^2 c^R(\eta_t, x) + \left(\frac{p}{q} - 1 \right)^2 c^L(\eta_t, x) \right) Z_t(x)^2, \quad x > 0, \end{aligned}$$

$$\frac{d}{dt} \langle M(0), M(0) \rangle_t = \left(\left(\frac{q}{p} - 1 \right)^2 r_A^+(\eta_t(1)) + \left(\frac{p}{q} - 1 \right)^2 r_A^-(\eta_t(1)) \right) Z_t(0)^2.$$

For ASEP-B, (3.1) holds for all $x \in \Lambda_N$, with Robin boundary conditions at $x = 0$ and $x = N$:

$$(3.3) \quad Z(-1) = \mu_A Z(0), \quad Z(N+1) = \mu_B Z(N),$$

where $M_\bullet(x)$ for $x \in \Lambda_N$ are martingales with bracket processes the same as in the ASEP-H case for $x \in \Lambda_N \setminus \{N\}$ while at the boundaries $x = N$,

$$\begin{aligned} \frac{d}{dt} \langle M(N), M(N) \rangle_t = & \left(\left(\frac{q}{p} - 1 \right)^2 r_B^-(\eta_t(N)) \right. \\ & \left. + \left(\frac{p}{q} - 1 \right)^2 r_B^+(\eta_t(N)) \right) Z_t(N)^2. \end{aligned}$$

PROOF. We first consider ASEP-H. Since only the jumps can affect the value of $h_t(x)$ for $x > 0$, and only the creation/annihilation at $x = 1$ can affect the value of $h_t(0)$, we have, by definition (2.14) of Z and definition of ASEP-H,

$$dZ_t(x) = \Omega(x)Z_t(x)dt + dM_t(x)$$

where the drift term (we suppress the dependence in t to simplify our notation)

$$\Omega(x) = \begin{cases} \nu + (e^{-2\lambda} - 1)c^L(\eta, x) + (e^{2\lambda} - 1)c^R(\eta, x) & \text{if } x > 0, \\ \nu + (e^{-2\lambda} - 1)r_A^-(\eta(1)) + (e^{2\lambda} - 1)r_A^+(\eta(1)) & \text{if } x = 0, \end{cases}$$

and $M_\bullet(x)$, $x \in \mathbb{Z}_{\geq 0}$, are martingales with the desired bracket process (this can be checked easily).

The relation $\Omega(x)Z(x) = \frac{D}{2} \Delta Z(x)$ for $x > 0$ can be achieved by setting

$$\lambda = \frac{1}{2} \log \frac{q}{p}, \quad \nu = p + q - 2\sqrt{pq}, \quad D = 2\sqrt{pq},$$

as in the case of the standard ASEP without boundary; see, for instance, [14]. Note that with the choice (2.13) we have $D = 1$ and $\lambda < 0$; however, in the following calculations we will represent all the quantities in terms of p, q without specifying p, q as in (2.13).

With the constants λ, ν, D determined as above, for $x = 0$ we have

$$(3.4) \quad \Omega(0) = \begin{cases} p + q - 2\sqrt{pq} + \left(\frac{q}{p} - 1\right)\alpha & \text{if } \eta(1) = -1, \\ p + q - 2\sqrt{pq} + \left(\frac{p}{q} - 1\right)\gamma & \text{if } \eta(1) = 1. \end{cases}$$

We want to match $\Omega(0)Z(0)$ with $\frac{D}{2}\Delta Z(0)$, with certain “outer” boundary conditions imposed on $Z(-1)$. Recall that by definition $Z(0) = e^{-\lambda h(0)+vt}$ and $Z(1) = e^{-\lambda h(0)-\lambda\eta(1)+vt}$.

For the Robin boundary condition

$$\mu_A(Z(-1) - Z(0)) + (1 - \mu_A)Z(-1) = 0, \quad \text{i.e., } Z(-1) - \mu_A Z(0) = 0$$

(where the real parameter μ_A interpolates the Dirichlet condition $Z(-1) = 0$ and the Neumann condition $Z(-1) - Z(0) = 0$), we have

$$\Delta Z(0) = Z(1) - (2 - \mu_A)Z(0) = (e^{-\lambda\eta(1)} - 2 + \mu_A)Z(0).$$

The condition $\Omega(0)Z(0) = \frac{D}{2}\Delta Z(0)$ is satisfied for $\eta(1) = -1$ if

$$p + q - 2\sqrt{pq} + \left(\frac{q}{p} - 1\right)\alpha = \sqrt{pq}\left(\sqrt{\frac{q}{p}} - 2 + \mu_A\right),$$

and for $\eta(1) = 1$ if

$$p + q - 2\sqrt{pq} + \left(\frac{p}{q} - 1\right)\gamma = \sqrt{pq}\left(\sqrt{\frac{p}{q}} - 2 + \mu_A\right).$$

Solving these equations leads precisely to the choices of α, γ in Definition 2.8.

For ASEP-B, the proof is analogous. The martingale bracket processes are easily computed. The derivation of our α, γ parametrization at the left boundary is the same, and one likewise readily checks that the same derivation with respect to the right boundary condition leads to the forms of β, δ in Definition 2.8. \square

Remark 3.2. The range of $\mu_A, \mu_B \in [\sqrt{q/p}, \sqrt{p/q}]$ assumed in Definition 2.8 was necessary to ensure the nonnegativity of the boundary rates. This limits the types of boundary conditions that can arise. For instance, the Neumann boundary condition for the SHE is accessible when $\mu_A = \mu_B = 1$ (i.e., $A = B = 0$). On the other hand, Dirichlet boundary conditions would require $\mu_A = \mu_B = 0$, which is certainly out of range.

In Definition 2.10, we assumed further that $\mu_A, \mu_B \leq 1$. This is only a technical restriction as it simplifies heat kernel bounds by disallowing for exponential modes. In a work in preparation, [65] provides the necessary heat kernel bounds to control possible exponential modes and extends our proof to $A, B < 0$ (i.e., $\mu_A, \mu_B > 1$) as well as to delta function initial data.

Utilizing the weakly asymmetric scalings from Definition 2.10—in particular their asymptotic expansions in Remark 2.11—leads to the following.

LEMMA 3.3. *For ASEP-H one has*

$$(3.5) \quad \frac{d}{dt}\langle M(x), M(x) \rangle_t = \begin{cases} \varepsilon Z_t(x)^2 - \nabla^+ Z_t(x)\nabla^- Z_t(x) + o(\varepsilon)Z_t(x)^2 & \text{if } x > 0, \\ \varepsilon Z_t(x)^2 + o(\varepsilon)Z_t(x)^2 & \text{if } x = 0, \end{cases}$$

where $o(\varepsilon)$ is a term uniformly bounded by constant C_ε and $C_\varepsilon/\varepsilon \rightarrow 0$.

For ASEP-B the first estimate in (3.5) holds for every $x \in \{1, 2, \dots, N-1\}$, and the second estimate in (3.5) holds for $x \in \{0, N\}$.

PROOF. The proof for the bulk (the first equation in (3.5)) is standard; see [15, prop. 2.1(b)] with the parameter j therein set as $j = 1/2$. We only remark that the term $\nabla^+ Z_t(x) \nabla^- Z_t(x)$ arises from the quadratic term $\eta(x)\eta(x+1)$ in the definition of the rates c^L, c^R . At $x = 0$, we use the expansions in Remark 2.11 to immediately conclude the desired behavior in ε . The proof for the case of ASEP-B is analogous. \square

4 Tightness

4.1 Estimates of Robin Heat Kernels

Before proceeding with the proof of tightness of the sequence \mathcal{Z}^ε , we need some estimates on discrete heat kernels with Robin boundary conditions.

Elastic Brownian Motion and Random Walk

A (discrete time) random walk \tilde{R}_H^e on $\mathbb{Z}_{\geq 0}$ with elastic boundary at -1 jumps in the same way as the standard random walk when it is at $x > 0$; if it is at 0 , then with probability $1/2$ it jumps to the site 1 , with probability $\mu_A/2$ it stays at the site 0 (i.e., “reflected back by the wall -1 ”), and finally with probability $(1 - \mu_A)/2$ it is killed (i.e., “absorbed by the wall -1 ”). The *continuous-time walk* R_H^e on $\mathbb{Z}_{\geq 0}$ with *elastic boundary* is then defined via \tilde{R}_H^e by imposing exponential holding time before each jump in the usual way. The heat kernel \mathbf{p}^R satisfying the Robin condition (3.2) is the transition probability for R_H^e , which is represented as

$$(4.1) \quad \mathbf{p}_t^R(x, y) = e^{-t} \sum_{n=0}^{\infty} \frac{t^n}{n!} p_n^e(x, y)$$

where $p_n^e(x, y)$ is the transition probability of the discrete time random walk \tilde{R}_H^e . A (discrete time) random walk \tilde{R}_B^e on $\Lambda_N = \{0, 1, \dots, N\}$ with elastic boundary at both ends is defined in the same way, except that if it is at 0 (respectively, at N), then with probability $\mu_A/2$ (respectively, at $\mu_B/2$) it stays there, and with probability $(1 - \mu_A)/2$ (respectively, $(1 - \mu_B)/2$) it is killed. The *continuous-time random walk* R_B^e on Λ_N with *elastic boundary* is then defined via \tilde{R}_B^e analogously as for the half-line case. The heat kernel \mathbf{p}^R satisfying the Robin condition (3.3) is the transition probability for R_B^e . See, for instance, [66].

Remark 4.1. Here and in what follows we will use \mathbf{p}^R to represent the discrete heat kernel in both the $\mathbb{Z}_{\geq 0}$ and Λ_N cases. Likewise we will use \mathcal{P}_T^R (introduced below) for the continuous heat kernel in both the \mathbb{R}_+ and $[0, 1]$ cases.

A *Brownian motion* B_H^e on \mathbb{R}_+ with *elastic boundary* at 0 is defined as $B_H^e(t) = |B(t)|$ if $t < \mathfrak{m}$ and killed at time \mathfrak{m} where $\mathfrak{m} \stackrel{\text{def}}{=} \mathfrak{t}^{-1}(\varepsilon/A)$, ε is an independent exponential random variable of rate 1 , and \mathfrak{t}^{-1} is the inverse function of the reflecting

Brownian local time

$$\mathfrak{t}(t) = \lim_{\varepsilon \rightarrow 0} (2\varepsilon)^{-1} \text{Leb}(s : |B(s)| < \varepsilon, 0 \leq s \leq t)$$

where Leb is the Lebesgue measure. Obviously if $A = 0$ then $B_H^\varepsilon(t)$ is simply the reflected Brownian motion $|B(t)|$, and if $A \rightarrow \infty$ one recovers the Brownian motion killed at the origin. The kernel \mathcal{P}^R that satisfies the Robin boundary condition with parameter A is the transition probability of B_H^ε . A *Brownian motion* B_B^ε on $[0, 1]$ with *elastic boundary at both ends* is defined as $B_B^\varepsilon(t) = B_H^\varepsilon(t)$ if $t < \mathfrak{n}$ and killed at time \mathfrak{n} where $\mathfrak{n} \stackrel{\text{def}}{=} \mathfrak{t}^{-1}(\varepsilon/B)$, ε is as above, and \mathfrak{t}^{-1} is the inverse function of the local time

$$\mathfrak{t}(t) = \lim_{\varepsilon \rightarrow 0} (2\varepsilon)^{-1} \text{Leb}(s : B_H^\varepsilon(s) > 1 - \varepsilon, 0 \leq s \leq t).$$

The kernel \mathcal{P}_t^R on $[0, 1]$ that satisfies the Robin boundary condition with parameter A, B is the transition probability of B_B^ε . See, for instance, [53].

From these probabilistic interpretations one immediately has the following simple properties.

LEMMA 4.2. *The above kernels \mathcal{P}_T^R and \mathbf{p}_t^R satisfy the semigroup properties and are both nonnegative at all space-time points. Also, the kernels \mathcal{P}_T^R on $I \in \{\mathbb{R}_+, [0, 1]\}$ and \mathbf{p}_t^R on $\Lambda \in \{\mathbb{Z}_{\geq 0}, \{0, 1, \dots, N\}\}$ satisfy*

$$\int_I \mathcal{P}_t^R(X, Y) dY \leq 1, \quad \sum_{y \in \Lambda} \mathbf{p}_t^R(x, y) \leq 1.$$

The inequalities become equalities if and only if $A = 0$ (or $\mu = 1$).

LEMMA 4.3. *The kernels \mathcal{P}_T^R on \mathbb{R}_+ or $[0, 1]$ satisfy the following upper bound. For each $\bar{T} > 0$ there exists a constant $C(\bar{T})$ such that for all $T < \bar{T}$,*

$$\mathcal{P}_T^R(X, Y) \leq \frac{C(\bar{T})}{\sqrt{T}} e^{-\frac{|Y-X|^2}{2T}}.$$

PROOF. For the reflected Brownian motion on $[0, 1]$, for which the Robin heat kernel \mathcal{P}^R reduces to the Neumann heat kernel \mathcal{P}^N , this is proved in [70, eq. (3.7)] (except that the equation considered therein has an extra linear damping term and thus the resulting bound has an extra factor e^{-T}). The proof of [70] relies on the explicit expression of \mathcal{P}^N via the image method. Since the transition probability of the elastic Brownian motion must be smaller than the reflected Brownian motion, one immediately obtains the claimed upper bound for \mathcal{P}^R . For the case of \mathbb{R}_+ , one can again prove the claimed upper bound for \mathcal{P}^N by the method of images and then the bound for \mathcal{P}^R follows from it. Note that this bound is also easily proved for \mathcal{P}^R directly by applying the formula in Lemma 4.4 below, which is derived by the image method. \square

A Generalized Image Method

We start by providing a formula for \mathcal{P}_T^R using a generalization of the image method (to the Robin boundary condition case). We will not make much use of this continuous-space formula, but include it since it motivates the more complicated discrete-image method formulas provided in Lemmas 4.5 and 4.7. In scanning the literature, we came upon some continuous-space generalizations of the image method on \mathbb{R} with some Robin boundary condition at the origin (see [32, 33, 35]). Discrete space and the boundary interval geometry lead to much more involved calculations.

LEMMA 4.4. *The half-line heat kernel $\mathcal{P}_T^R(X, Y)$ for $X, Y \geq 0$ that satisfies the Robin boundary condition $\partial_X \mathcal{P}_T^R(X, Y)|_{X=0} = A\mathcal{P}_T^R(0, Y)$ has the following representation:*

$$(4.2) \quad \mathcal{P}_T^R(X, Y) = P_T(X-Y) + P_T(X+Y) - 2A \int_{-\infty}^0 P_T(X+Y-Z)e^{AZ} dZ$$

where $P_T(X) = \frac{1}{\sqrt{2\pi T}}e^{-X^2/(2T)}$ is the standard continuous heat kernel for the heat operator $\partial_T - \frac{1}{2}\Delta$.

PROOF. We will prove (4.2) by showing that given any function $\phi(X)$ on $X \in \mathbb{R}_+$,

$$u(T, X) = \int_0^\infty \mathcal{P}_T^R(X, Y)\phi(Y)dY$$

solves the equation $\partial_T u = \frac{1}{2}\Delta u$ on $\mathbb{R}_+ \times \mathbb{R}_+$ with initial condition $u(0, X) = \phi(X)$ ($X \in \mathbb{R}_+$) and Robin boundary condition

$$\partial_X u(X, T)|_{X=0} = Au(0, T).$$

The above equation can be solved by extending the initial values ϕ so that $\phi' - A\phi$ is odd, namely, for $X < 0$ one should solve the ODE

$$(4.3) \quad \phi'(X) - A\phi(X) = -\phi'(-X) + A\phi(-X), \quad X < 0,$$

and then solve the heat equation on the entire line \mathbb{R}_+ with the extended ϕ as initial data. Indeed, $u(T, X) = \int_{\mathbb{R}} P_T(X-Y)\phi(Y)dY$ satisfies

$$\begin{aligned} \partial_X u(X, T)|_{X=0} - Au(0, T) &= \int_{\mathbb{R}} P_T'(-Y)\phi(Y)dY - A \int_{\mathbb{R}} P_T(-Y)\phi(Y)dY \\ &= \int_{\mathbb{R}} P_T(-Y)(\phi'(Y) - A\phi(Y))dY, \end{aligned}$$

which vanishes because P_T is even and $\phi' - A\phi$ is odd.

To solve (4.3), we use an integrating factor e^{-AX} and get

$$(e^{-AX}\phi(X))' = -e^{-AX}(\phi'(-X) - A\phi(-X)), \quad X < 0,$$

so

$$\phi(X) = Ce^{AX} + e^{AX} \int_X^0 e^{-As} (\phi'(-s) - A\phi(-s)) ds, \quad X < 0.$$

Choose C so that the extended function ϕ is continuous at 0, that is, $C = \phi(0)$. This can be simplified as

$$\phi(X) = \phi(-X) - 2Ae^{AX} \int_0^{-X} e^{As} \phi(s) ds, \quad X < 0.$$

so that

$$\begin{aligned} u(T, X) &= \int_0^\infty P_T(X - Y) \phi(Y) dY + \int_{-\infty}^0 P_T(X - Y) \phi(-Y) dY \\ &\quad - 2A \int_{-\infty}^0 P_T(X - Y) e^{AY} \int_0^{-Y} e^{As} \phi(s) ds dY \\ &= \int_0^\infty \left(P_T(X - Y) + P_T(X + Y) \right. \\ &\quad \left. - 2A \int_{-\infty}^0 P_T(X + Y - Z) e^{AZ} dZ \right) \phi(Y) dY. \end{aligned}$$

This shows that the Robin heat kernel is given by (4.2). \square

We turn now to the analogous result for the discrete space half-line heat kernel.

LEMMA 4.5. *The half-line discrete heat kernel $\mathbf{p}_t^R(x, y)$ for $x, y \in \mathbb{Z}_{\geq 0}$ that satisfies the Robin boundary condition $\mathbf{p}_t^R(-1, y) = \mu_A \mathbf{p}_t^R(0, y)$ has the following representation:*

$$(4.4) \quad \begin{aligned} \mathbf{p}_t^R(x, y) &= p_t(x - y) + \mu_A p_t(x + y + 1) \\ &\quad + (\mu_A^2 - 1) \sum_{z=-\infty}^{-2} p_t(x + y - z) \mu_A^{-z-2} \end{aligned}$$

where p is the transition probability of standard continuous-time random walk on \mathbb{Z} (i.e., jumps left and right by 1 at rate $1/2$).

PROOF. The proof is analogous to that of Lemma 4.4, but adapted to the discrete setting. Given ϕ on $\mathbb{Z}_{\geq 0}$, one wants to extend it to \mathbb{Z} such that for $x < 0$,

$$(4.5) \quad \phi(x - 1) - \mu_A \phi(x) = -(\phi(-x - 1) - \mu_A \phi(-x)).$$

The above relation is solved by the function

$$(4.6) \quad \phi(x) = \mu_A \phi(-x - 1) + (\mu_A^2 - 1) \sum_{k=0}^{-x-2} \mu_A^{-x-2-k} \phi(k), \quad x < 0$$

(if $-x - 2 < 0$ the sum over k is understood as 0). Therefore the solution to the heat equation is given by

$$\begin{aligned}
u(t, x) &= \sum_{y=-\infty}^{\infty} p_t(x-y)\phi(y) \\
&= \sum_{y=0}^{\infty} p_t(x-y)\phi(y) + \sum_{y=-\infty}^{-1} p_t(x-y)\mu_A\phi(-y-1) \\
&\quad + (\mu_A^2 - 1) \sum_{y=-\infty}^{-1} p_t(x-y) \sum_{k=0}^{-y-2} \mu_A^{-y-2-k} \phi(k) \\
&= \sum_{y=0}^{\infty} \left(p_t(x-y) + \mu_A p_t(x+y+1) \right. \\
&\quad \left. + (\mu_A^2 - 1) \sum_{z=-\infty}^{-2-y} p_t(x-z)\mu_A^{-z-y-2} \right) \phi(y).
\end{aligned}$$

To check that this u satisfies the Robin boundary condition $u_t(-1) = \mu_A u_t(0)$ for any t , observe that

$$\begin{aligned}
(4.7) \quad u_t(-1) - \mu_A u_t(0) &= \sum_{y=-\infty}^{\infty} p_t(-1-y)\phi(y) - \mu_A \sum_{y=-\infty}^{\infty} p_t(-y)\phi(y) \\
&= \sum_{y=-\infty}^{\infty} p_t(-y)(\phi(y-1) - \mu_A \phi(y)).
\end{aligned}$$

By (4.5) and $p_t(-y) = p_t(y)$ the above expression is 0. So the Robin heat kernel is given by (4.4). \square

From the explicit formulas in Lemma 4.4 and Lemma 4.5, one can see that

$$\mathcal{P}_t^R(x, y) = \mathcal{P}_t^R(y, x), \quad \mathbf{p}_t^R(x, y) = \mathbf{p}_t^R(y, x);$$

i.e., they are symmetric in x, y . In proving these lemmas we could have just checked that the stated formulas solve the relevant heat equations. We opted for the more detailed derivations since it is informative in attacking the below finite-interval case, where we are unable to provide a concise closed-form solution to check. Instead, in that case we provide bounds on the solution, which suffice for our applications.

To find the Robin heat kernel on the discrete finite interval Λ_N , we need to apply the above image method in a recursive way. It will be convenient to introduce the notation

$$\bar{N} = N + 1$$

so that $\mathbb{Z} = \bigcup_{k \in \mathbb{Z}} \{k\bar{N}, k\bar{N}+1, \dots, (k+1)\bar{N}-1\}$, a union of nonoverlapping sets, each consisting of \bar{N} points. We start with the first several steps of the extension in order to motivate the general case. Given ϕ on $\{0, 1, \dots, \bar{N}-1\}$, one first extends it to $\{-\bar{N}, -\bar{N}+1, \dots, -1\}$ such that for $x \in \{-\bar{N}, -\bar{N}+1, \dots, -1\}$,

$$(4.8) \quad \phi(x-1) - \mu_A \phi(x) = -(\phi(-x-1) - \mu_A \phi(-x)).$$

For this we have already obtained in (4.6) that

$$(4.9) \quad \phi(x) = \mu_A \phi(-x-1) + (\mu_A^2 - 1) \sum_{y=0}^{-x-2} \mu_A^{-x-2-y} \phi(y)$$

for $x \in \{-\bar{N}, -\bar{N}+1, \dots, -1\}$. We then extend ϕ to $\{\bar{N}, \dots, 2\bar{N}-1\}$ such that

$$(4.10) \quad \phi(x) - \mu_B \phi(x-1) = -(\phi(2\bar{N}-x) - \mu_B \phi(2\bar{N}-x-1))$$

for every $x \in \{\bar{N}, \bar{N}+1, \dots, 2\bar{N}-1\}$. Solving the above equation (4.10) (or directly checking the following result), one has

$$(4.11) \quad \phi(x) = \mu_B \phi(2\bar{N}-x-1) + (\mu_B^2 - 1) \sum_{y=2\bar{N}-x}^{\bar{N}-1} \mu_B^{y-2\bar{N}+x} \phi(y)$$

for $x \in \{\bar{N}, \bar{N}+1, \dots, 2\bar{N}-1\}$. We should then iterate this procedure to extend ϕ to increasingly larger domains, but at this point if we were to stop the iteration we would get the “leading-order terms” of our heat kernel

$$\begin{aligned} \mathbf{p}_t^R(x, y) &\approx p_t(x-y) + \mu_A p_t(x+y+1) + \mu_B p_t(x+y+1-2\bar{N}) \\ &\quad + (\mu_A^2 - 1) \sum_{z=-\bar{N}}^{-2-y} p_t(x-z) \mu_A^{-z-y-2} \\ &\quad + (\mu_B^2 - 1) \sum_{z=2\bar{N}-y}^{2\bar{N}-1} p_t(x-z) \mu_B^{z+y-2\bar{N}} \end{aligned}$$

for $x, y \in \{0, 1, \dots, N\}$ (recall that \mathbf{p}_t^R is of course always defined only on $x, y \in \{0, 1, \dots, N\}$). Here “ \approx ” means that we should actually take into account more terms on the right-hand side in order to obtain an equality (see the final result (4.21) below for comparison).

We now further extend ϕ to $\{-2\bar{N}, -2\bar{N}+1, \dots, -\bar{N}-1\}$ according to (4.8). This is possible because the values on $\{0, 1, \dots, 2\bar{N}\}$ have already been defined, and we get (4.9) for all $x \in \{-2\bar{N}, -2\bar{N}+1, \dots, -\bar{N}-1\}$. We then plug (4.11) in to get an expression only depending on the values of ϕ on $\{0, 1, \dots, \bar{N}-1\}$. This

eventually yields

$$\begin{aligned}
\phi(x) &= \mu_A \mu_B \phi(2\bar{N} + x) + (\mu_A^2 - 1) \sum_{y=0}^{\bar{N}-1} \mu_A^{-x-2-y} \phi(y) \\
&+ \sum_{y=2\bar{N}+x+1}^{\bar{N}-1} \left(\mu_A (\mu_B^2 - 1) \mu_B^{y-(2\bar{N}+x+1)} \right. \\
(4.12) \quad &\left. + \mu_B (\mu_A^2 - 1) \mu_A^{y-(2\bar{N}+x+1)} \right) \phi(y) \\
&+ (\mu_A^2 - 1) (\mu_B^2 - 1) \sum_{y=2\bar{N}+x+2}^{\bar{N}-1} \left(\sum_{\substack{i,j \geq 0 \\ i+j=y-(2\bar{N}+2+x)}} \mu_A^i \mu_B^j \right) \phi(y)
\end{aligned}$$

for every $x \in \{-2\bar{N}, -2\bar{N} + 1, \dots, -\bar{N} - 1\}$.

We can iterate this procedure and use the condition (4.8) and (4.10) *in turn* to extend ϕ to the entire \mathbb{Z} . Let u be the solution to the heat equation on \mathbb{Z} starting from this extended ϕ . Since (4.8) holds for all $x < 0$ by our construction, as in (4.7) we have $u_t(-1) - \mu_A u_t(0) = 0$. Also, since (4.10) holds for all $x \geq \bar{N}$ by construction, we have

$$\begin{aligned}
&u_t(\bar{N}) - \mu_B u_t(\bar{N} - 1) \\
&= \sum_{y=-\infty}^{\infty} p_t(\bar{N} - y) \phi(y) - \mu_B \sum_{y=-\infty}^{\infty} p_t(\bar{N} - 1 - y) \phi(y) \\
&= \sum_{y=-\infty}^{\infty} p_t(-y) (\phi(\bar{N} + y) - \mu_B \phi(\bar{N} - 1 + y)).
\end{aligned}$$

This vanishes because p_t is even, and $\phi(\bar{N} + y) - \mu_B \phi(\bar{N} - 1 + y)$ is odd in y by (4.10) (with x in (4.10) chosen as $x = \bar{N} + y$). Therefore u satisfies the desired boundary condition.

We have the following Lemma 4.6 to represent the extended ϕ in terms of the original given ϕ defined on Λ_N in a suitable form that is convenient for the following analysis.

For $x \in \{k\bar{N}, k\bar{N} + 1, \dots, k\bar{N} + \bar{N} - 1\}$, we define

$$(4.13) \quad x^\star = \begin{cases} (k+1)\bar{N} - x - 1 & \text{if } k \text{ is an odd integer,} \\ x - k\bar{N} & \text{if } k \text{ is an even integer.} \end{cases}$$

It is easy to check that one always has $0 \leq x^\star \leq \bar{N} - 1$. This is the preimage of x under reflection through the two sets of boundaries of Λ_N .

LEMMA 4.6. *Let ϕ be the function on \mathbb{Z} obtained by the above recursive extension procedure. There exists a constant C_0 that only depends on A, B such that for each $k \in \mathbb{Z}$,*

$$(4.14) \quad \phi(x) = I_k \phi(x^*) + \varepsilon \sum_{y=0}^{\bar{N}-1} E_k(x, y) \phi(y)$$

for all $x \in \{k\bar{N}, k\bar{N} + 1, \dots, k\bar{N} + \bar{N} - 1\}$, where $0 < I_k \leq 1$ and

$$\max_{\substack{k\bar{N} \leq x < (k+1)\bar{N} \\ 0 \leq y < \bar{N}}} |E_k(x, y)| \leq C_0^{|k|}.$$

One can check that (4.12) is indeed of the form (4.14) since $\mu_A^2 - 1 \sim \mu_B^2 - 1 \sim \varepsilon$ and the sum $\sum \mu_A^i \mu_B^j \sim N \sim \frac{1}{\varepsilon}$.

PROOF. The proof goes by induction. To begin, if $k = 0$ then $x^* = x$ so (4.14) holds with $I_0 = 1$ and $E_0 = 0$. Suppose that the statement of the lemma is true for $|k| \leq m$; we show it for $k = -m - 1$ and $k = m + 1$.

Consider the case $k = -m - 1$. Since ϕ is required to satisfy (4.8), as in (4.6) we have

$$(4.15) \quad \phi(x) = \mu_A \phi(-x - 1) + (\mu_A^2 - 1) \sum_{y=0}^{-x-2} \mu_A^{-x-2-y} \phi(y)$$

for $x \in [k\bar{N}, k\bar{N} + \bar{N} - 1] \cap \mathbb{Z}$, where $\phi(y)$ on the RHS has been defined as our induction assumption since $-x - 2 \leq (m + 1)\bar{N} - 2$. Noting that $m\bar{N} \leq -x - 1 \leq (m + 1)\bar{N} - 1$, the first term on the RHS of (4.15) is equal to, by inductive assumption,

$$(4.16) \quad \mu_A I_m \phi((-x - 1)^*) + \mu_A \varepsilon \sum_{y=0}^{\bar{N}-1} E_m(x, y) \phi(y).$$

By definition one can check that

$$(-x - 1)^* = (m + 1)\bar{N} + x = x^* \quad \text{if } m \text{ is odd,}$$

$$(-x - 1)^* = -x - 1 - m\bar{N} = x^* \quad \text{if } m \text{ is even.}$$

The second term on the RHS of (4.15) is equal to (noting that $-x - 2 \leq (m + 1)\bar{N} - 2$)

$$(\mu_A^2 - 1) \sum_{\ell=0}^m \sum_{y=\ell\bar{N}}^{(\ell\bar{N} + \bar{N} - 1) \wedge (-x - 2)} \mu_A^{-x-2-y} \phi(y).$$

Using the inductive assumption, this quantity is equal to

$$(4.17) \quad (\mu_A^2 - 1) \sum_{\ell=0}^m \sum_{y=\ell\bar{N}}^{(\ell\bar{N}+\bar{N}-1)\wedge(-x-2)} \mu_A^{-x-2-y} \left(I_\ell \phi(y^*) + \varepsilon \sum_{z=0}^{\bar{N}-1} E_\ell(y, z) \phi(z) \right).$$

We now deal with the terms with I_ℓ and the terms with E_ℓ separately. Regarding the terms with I_ℓ in (4.17), for each fixed ℓ , as y ranges from $\ell\bar{N}$ to $(\ell+1)\bar{N}-1$, y^* ranges over $\{0, 1, \dots, \bar{N}-1\}$, and this correspondence is one-to-one. So

$$(4.18) \quad \sum_{\ell=0}^m \sum_{y=\ell\bar{N}}^{(\ell\bar{N}+\bar{N}-1)\wedge(-x-2)} \mu_A^{-x-2-y} I_\ell \phi(y^*) = \sum_{y^*=0}^{\bar{N}-1} \left(\sum_{\ell=0}^m \mu_A^{-x-2-y} I_\ell \mathbf{1}_{y \leq -x-2} \right) \phi(y^*)$$

where the y in the parentheses on the RHS is determined by y^* via reversing (4.13), namely,

$$(4.19) \quad y = \iota(y^*; \ell) \stackrel{\text{def}}{=} \begin{cases} (\ell+1)\bar{N} - y^* - 1 & \text{if } \ell \text{ is an odd integer,} \\ y^* + \ell\bar{N} & \text{if } \ell \text{ is an even integer.} \end{cases}$$

Regarding the terms with E_ℓ in (4.17), we have

$$(4.20) \quad \sum_{\ell=0}^m \sum_{y=\ell\bar{N}}^{(\ell\bar{N}+\bar{N}-1)\wedge(-x-2)} \mu_A^{-x-2-y} \left(\varepsilon \sum_{z=0}^{\bar{N}-1} E_\ell(y, z) \phi(z) \right) = \varepsilon \sum_{z=0}^{\bar{N}-1} \left(\sum_{\ell=0}^m \sum_{y=\ell\bar{N}}^{(\ell\bar{N}+\bar{N}-1)\wedge(-x-2)} \mu_A^{-x-2-y} E_\ell(y, z) \right) \phi(z).$$

Summarizing the above formulas (4.16), (4.17), (4.18), and (4.20), we have

$$\phi(x) = I_{-m-1} \phi(x^*) + \varepsilon \sum_{y=0}^{\bar{N}-1} E_{-m-1}(x, y) \phi(y)$$

where

$$I_{-m-1} \stackrel{\text{def}}{=} \mu_A I_m,$$

$$\begin{aligned} E_{-m-1}(x, y) &\stackrel{\text{def}}{=} \mu_A E_m(x, y) + \varepsilon^{-1} (\mu_A^2 - 1) \sum_{\ell=0}^m \mu_A^{-x-2-\iota(y; \ell)} I_\ell \mathbf{1}_{\iota(y; \ell) \leq -x-2} \\ &+ (\mu_A^2 - 1) \sum_{\ell=0}^m \sum_{\bar{y}=\ell\bar{N}}^{\ell\bar{N}+\bar{N}-1} \mu_A^{-x-2-\bar{y}} \mathbf{1}_{\bar{y} \leq -x-2} E_\ell(\bar{y}, y). \end{aligned}$$

The coefficient I_{-m-1} clearly satisfies the desired bound since $0 < \mu_A \leq 1$ and $0 < I_m \leq 1$ by induction. Since $|\mu_A^2 - 1| \leq 3A\varepsilon$ for ε sufficiently small, and $N = 1/\varepsilon$, we have

$$\begin{aligned} |E_{-m-1}(x, y)| &\leq |E_m(x, y)| + 3Am + 3A \sum_{\ell=0}^m \max_{\ell\bar{N} \leq \bar{y} < (\ell+1)\bar{N}} |E_\ell(\bar{y}, y)| \\ &\leq C_0^m + 3Am + 3A \sum_{\ell=0}^m C_0^\ell, \end{aligned}$$

where we have applied the inductively assumed bounds on E_m, E_ℓ . It is easy to see that there exists C_0 that is independent of m such that

$$C_0^m + 3Am + 3A \cdot \frac{C_0^{m+1} - 1}{C_0 - 1} \leq C_0^{m+1}$$

for all $m > 0$. Indeed, one can divide both sides by C_0^{m+1} and see that it suffices to show that for sufficiently large C_0 one has $\frac{1}{C_0} + \frac{3A}{C_0-1} \leq 0.9$ or $\frac{1+3A}{C_0-1} \leq 0.9$. With this choice of C_0 the exponential bound $|E_{-m-1}| \leq C_0^{m+1}$ follows.

The case $k = m + 1$ can be shown in the same way. Therefore the inductive proof is complete. \square

The following lemma will be useful for the proofs of the Robin heat kernel estimates on bounded intervals. One can immediately see that the statement of the following lemma is true for the Neumann heat kernel on the bounded interval with $I_k = 1$ and $E_k = 0$.

LEMMA 4.7. *The discrete heat kernel \mathbf{p}^R on $\{0, 1, \dots, N\}$ that satisfies the Robin boundary condition $\mathbf{p}_t^R(-1, y) = \mu_A \mathbf{p}_t^R(0, y)$ and $\mathbf{p}_t^R(N+1, y) = \mu_B \mathbf{p}_t^R(N, y)$ has the following representation:*

$$\begin{aligned} \mathbf{p}_t^R(x, y) &= p_t(x - y) + \sum_{k \in \mathbb{Z}, k \neq 0} I_k p_t(x - \iota(y; k)) \\ (4.21) \quad &+ \varepsilon \sum_{k \in \mathbb{Z}, k \neq 0} \sum_{\bar{y}=k\bar{N}}^{(k+1)\bar{N}-1} p_t(x - \bar{y}) E_k(\bar{y}, y) \end{aligned}$$

where ι is defined in (4.19) and I_k and E_k satisfy the estimates in Lemma 4.6.

PROOF. The heat kernel \mathbf{p}^R is related to the standard continuous time discrete space random walk heat kernel p by

$$\sum_{y=0}^{\bar{N}-1} \mathbf{p}_t^R(x, y) \phi(y) = \sum_{y \in \mathbb{Z}} p_t(x, y) \phi(y) = \sum_{k \in \mathbb{Z}} \sum_{y=k\bar{N}}^{(k+1)\bar{N}-1} p_t(x - y) \phi(y).$$

Note that ϕ in the middle and on the right-hand side of the above equation is the extension to all of \mathbb{Z} of ϕ on Λ as described earlier in this subsection. By applying Lemma 4.6, the above quantity is equal to

$$\begin{aligned}
& \sum_{k \neq 0} \sum_{y=k\bar{N}}^{(k+1)\bar{N}-1} p_t(x-y) \left(I_k \phi(y^*) + \varepsilon \sum_{z=0}^{\bar{N}-1} E_k(y,z) \phi(z) \right) \\
& + \sum_{y=0}^{\bar{N}-1} p_t(x-y) \phi(y) \\
& = \sum_{y^*=0}^{\bar{N}-1} \left(\sum_{k \neq 0} I_k p_t(x - \iota(y^*; k)) \right) \phi(y^*) \\
& + \sum_{z=0}^{\bar{N}-1} \left(\varepsilon \sum_{k \neq 0} \sum_{y=k\bar{N}}^{(k+1)\bar{N}-1} p_t(x-y) E_k(y,z) \right) \phi(z) \\
& + \sum_{y=0}^{\bar{N}-1} p_t(x-y) \phi(y).
\end{aligned}$$

Therefore \mathbf{p}^R is given by (4.21). \square

Sturm-Liouville Theory

We need some results on the spectrum of the discrete operator $-\frac{1}{2}\Delta$ on the discrete interval $\{0, 1, \dots, N\} \subset \mathbb{Z}$ with Robin boundary condition

$$u(-1) - \mu_A u(0) = 0, \quad u(N+1) - \mu_B u(N) = 0.$$

We start by recalling some summation-by-parts formulas. We only need the formulas on the bounded intervals here in this subsection, but the formulas for the case of the half-line will also be used later. For the case of the half-line, we define

$$\mathbf{h}_+ \stackrel{\text{def}}{=} \{u : \{-1, 0\} \cup \mathbb{Z}_+ \rightarrow \mathbb{R} \mid \lim_{x \rightarrow \infty} u(x) = 0, \langle u, u \rangle_{\mathbf{h}_+} < \infty\}$$

where $\langle u, v \rangle_{\mathbf{h}_+} \stackrel{\text{def}}{=} \sum_{x=-1}^{\infty} (\nabla^+ u(x)) (\nabla^+ v(x))$. (Recall that ∇^\pm are defined in (2.3).) The following summation-by-parts formulas are facts of finite difference Laplacian Δ in general and has nothing to do with the particular type of boundary conditions; the proof is omitted since it is straightforward.

LEMMA 4.8. *For functions u, v on $\{-1, 0, \dots, N+1\}$, one has*

$$\begin{aligned}
(4.22) \quad \sum_{x=0}^N u(x) \Delta v(x) &= u(N+1) \nabla^+ v(N) + u(-1) \nabla^- v(-1) \\
&\quad - \sum_{x=-1}^N (\nabla^+ u(x)) (\nabla^+ v(x)),
\end{aligned}$$

$$\begin{aligned}
(4.23) \quad \sum_{x=0}^N u(x)\Delta v(x) &= \sum_{x=0}^N v(x)\Delta u(x) + u(N+1)\nabla^+ v(N) \\
&\quad + u(-1)\nabla^- v(0) - v(N+1)\nabla^+ u(N) \\
&\quad - v(-1)\nabla^- u(0).
\end{aligned}$$

For functions $u, v \in \mathbf{h}_+$, one has

$$(4.24) \quad \sum_{x=0}^{\infty} u(x)\Delta v(x) = \sum_{x=0}^{\infty} v(x)\Delta u(x) + u(-1)\nabla^- v(0) - v(-1)\nabla^- u(0).$$

First of all, we observe that if $\mu_A, \mu_B \leq 1$, the eigenvalues of the matrix $-\frac{1}{2}\Delta$ with Robin boundary condition are all nonnegative. This is because $-\frac{1}{2}\Delta$ is a self-adjoint Markov generator of the elastic random walk, but we can also see this more explicitly as follows. Note that if $u = (u(0), u(1), \dots, u(N))$ is an eigenvector with eigenvalue λ for the matrix $-\frac{1}{2}\Delta$ with Robin condition (that is, an $(N+1) \times (N+1)$ matrix), then, using (4.22), together with the Robin condition, we have

$$\begin{aligned}
2\lambda &= \frac{\langle u, (-\Delta)u \rangle}{|u|^2} = \frac{\sum_{x=0}^N u(x)(-\Delta u(x))}{\sum_{x=0}^N u^2(x)} \\
&= \frac{-u(N+1)^2(1 - \mu_B^{-1}) + u(-1)^2(\mu_A^{-1} - 1) + \sum_{x=-1}^N (\nabla^+ u(x))^2}{\sum_{x=0}^N u^2(x)} \geq 0.
\end{aligned}$$

Here, $\langle u, (-\Delta)u \rangle$ is a bilinear form on \mathbb{R}^{N+1} , while in $\sum_{x=0}^N u(x)(-\Delta u(x))$ the Δ is the usual finite difference Laplacian as in (4.22) where we need to specify two extra values $u(-1) \stackrel{\text{def}}{=} \mu_A u(0)$ and $u(N+1) = \mu_B u(N)$ using the Robin condition.

Let $0 \leq \lambda_0 \leq \lambda_1 \leq \dots \leq \lambda_N$ be the eigenvalues of $-\frac{1}{2}\Delta$ with the above boundary condition. Let ψ_k be the k^{th} normalized (i.e., $\sum_{x=0}^N \psi_k(x)^2 = 1$) eigenfunction associated with λ_k :

$$(4.25) \quad \psi_k(x) = C_{1,k} \cos(\omega_k x) + C_{2,k} \sin(\omega_k x).$$

Since they satisfy the Robin boundary condition we have

$$\begin{aligned}
(4.26) \quad C_{1,k} \cos(-\omega_k) + C_{2,k} \sin(-\omega_k) &= \mu_A C_{1,k}, \\
C_{1,k} \cos(\omega_k(N+1)) + C_{2,k} \sin(\omega_k(N+1)) \\
&= \mu_B (C_{1,k} \cos(\omega_k N) + C_{2,k} \sin(\omega_k N)).
\end{aligned}$$

From these we cancel out $C_{1,k}, C_{2,k}$ and we get, after simplification, the equation for ω_k :

$$(4.27) \quad \sin(\omega_k(N+2)) - (\mu_A + \mu_B) \sin(\omega_k(N+1)) + \mu_A \mu_B \sin(\omega_k N) = 0.$$

Note that $\omega = 0$ and $\omega = \pi$ are always solutions to (4.27). If $\mu_A = \mu_B = 1$, they correspond to a constant eigenvector. But if $(\mu_A, \mu_B) \neq (1, 1)$, $\omega = 0$ and $\omega = \pi$ do not correspond to nontrivial eigenvectors since $C_{1,k}$ in (4.26) must then be 0 and therefore $\psi = 0$.

Let us record here the spectral decomposition of the heat kernel:

$$(4.28) \quad \mathbf{p}_t^R(x, y) = \sum_{k=0}^N \psi_{\lambda_k}(x) \psi_{\lambda_k}(y) e^{-t\lambda_k}.$$

LEMMA 4.9. *Under the above setting, if $\mu_A = \mu_B = 1$, we have $\omega_k = \frac{k\pi}{N+1}$ and $\lambda_k = 1 - \cos(\frac{k\pi}{N+1})$. If $\mu_A + \mu_B < 2$ we have*

$$\lambda_k = 1 - \cos(\omega_k), \quad k = 0, 1, \dots, N,$$

where ω_k for $k = 0, 1, \dots, N$ are the $N + 1$ solutions of (4.27) in $(0, \pi)$ and are ordered as $\omega_0 \leq \omega_1 \leq \dots \leq \omega_N$. Furthermore, we have

$$(4.29) \quad \frac{k\pi}{N+1} \leq \omega_k \leq \frac{(k+1)\pi}{N+1}$$

for every $k = 0, 1, \dots, N$.

PROOF. For the case $\mu_A = \mu_B = 1$, one can directly check that $\omega_k = \frac{k\pi}{N+1}$ solve (4.27). So we focus on the case $\mu_A + \mu_B < 2$.

We compute the derivative of the LHS of (4.27) w.r.t. ω_k at $\omega_k = 0$, which is equal to

$$\begin{aligned} & (N+2) - (\mu_A + \mu_B)(N+1) + \mu_A \mu_B N \\ & \geq (N+2) - (2 - \varepsilon A - \varepsilon B)(N+1) + (1 - \varepsilon A - \varepsilon B)N \\ & = \varepsilon A + \varepsilon B \geq 0 \end{aligned}$$

so the LHS of (4.27) is increasing at 0. Then it is easy to prove that the LHS of (4.27) at $\frac{k\pi}{N+1}$ is negative if k is odd and positive if k is even, since at these values the LHS of (4.27) becomes

$$\sin\left(k\pi + \frac{k\pi}{N+1}\right) + \mu_A \mu_B \sin\left(k\pi - \frac{k\pi}{N+1}\right),$$

and the first term always dominates and determines the sign if $\mu_A \mu_B < 1$. The claim (4.29) follows immediately by continuity of the LHS of (4.27) in ω_k . \square

LEMMA 4.10. *Under the above setting, there exists a constant $C > 0$ such that for all sufficiently large N , $0 \leq k \leq N$, $0 \leq x \leq N$, one has $|\psi_k(x)| \leq \frac{C}{\sqrt{N}}$.*

PROOF. By an elementary trigonometric identity we rewrite ψ_k as $\psi_k(x) = C_k \sin(\omega_k x + \theta_k)$. When $k = 0$, $\omega_0 < \frac{\pi}{N+1}$, so it is easy to see that over the interval $\{0, 1, \dots, N\}$ (which is less than half of the period of ψ_0), the L^2 -normalized

function ψ_0 is bounded by $\frac{C}{\sqrt{N}}$. Since ψ_k is normalized for every k ,

$$\begin{aligned}
1 &= C_k^2 \sum_{x=0}^N \sin^2(\omega_k x + \theta_k) \\
(4.30) \quad &= C_k^2 \sum_{x=0}^N \frac{1}{2} (1 - \cos(2\omega_k x + 2\theta_k)), \\
3 &= C_k^2 \left(\frac{N+1}{2} - \frac{\sin((N+1)\omega_k) \cos(N\omega_k + 2\theta_k)}{\sin \omega_k} \right),
\end{aligned}$$

where we used a Lagrange trigonometric summation formula. For $0 < k \leq N/2$, by Lemma 4.9 we know that

$$\omega_k \in \left[\frac{k\pi}{N+1}, \frac{(k+1)\pi}{N+1} \right],$$

and therefore for N sufficiently large

$$\sin \omega_k \geq \mathbf{1}_{k=1} \frac{8\omega_k}{9} + \mathbf{1}_{k>1} \frac{\omega_k}{2} \geq \frac{8\pi}{9(N+1)},$$

so the above quantity (4.30) is larger than

$$C_k^2 \left(\frac{N+1}{2} - \frac{9(N+1)}{8\pi} \right),$$

and therefore $|C_k|$ and thus $|\psi_k|$ are bounded by $\frac{C}{\sqrt{N}}$. If $N/2 < k < N$, we can write $\sin \omega_k = \sin(\pi - \omega_k)$ and then follow the same argument. Finally, when $k = N$, if $\mu_A = \mu_B = 1$, then $\omega_N = \frac{N\pi}{N+1}$; otherwise one can check that the LHS of (4.27) at $w = \frac{N\pi}{N+1}$ and $w = \frac{(3N+1)\pi}{3N+3}$ have opposite signs. This implies that $\pi - \omega_N > \frac{2\pi}{3N+3}$. For N sufficiently large one has

$$\sin(\pi - \omega_N) \geq \frac{8}{9}(\pi - \omega_N) \geq \frac{16\pi}{9(3N+3)},$$

so (4.30) is larger than

$$C_k^2 \left(\frac{N+1}{2} - \frac{27(N+1)}{16\pi} \right),$$

and therefore $|C_k|$ and thus $|\psi_k|$ are bounded by $\frac{C}{\sqrt{N}}$. \square

Heat Kernel Estimates

With the techniques developed above, we now prove the various heat kernel estimates, which will be useful for the rest of the paper. Proposition 4.11 and Corollary 4.12 below are the estimates for the μ_A -Robin heat kernel on $\mathbb{Z}_{\geq 0}$, while

Proposition 4.13 and Corollary 4.14 below are the estimates for the (μ_A, μ_B) -Robin heat kernel on $\{0, 1, \dots, N\}$. The analogous estimates for the standard heat kernels on the entire \mathbb{Z} are known, for instance, [7, 19].

PROPOSITION 4.11. *Assume that \mathbf{p}^R is the heat kernel on $\mathbb{Z}_{\geq 0}$ with Robin boundary condition (3.2) with $\mu = 1 - \varepsilon A$ for a constant $A > 0$. Given any $b \geq 0$, for any $|n| \leq \lceil t^{1/2} \rceil$, $v \in [0, 1]$, $0 \leq t < t' < \varepsilon^{-2} \bar{T}$, and $x, y \in \mathbb{Z}_{\geq 0}$, we have*

$$(4.31) \quad \mathbf{p}_t^R(x, y) \leq e^{C(t'-t)} \mathbf{p}_{t'}^R(x, y),$$

$$(4.32) \quad |\mathbf{p}_{t'}^R(x, y) - \mathbf{p}_t^R(x, y)| \leq C(A) (1 \wedge t^{-\frac{1}{2}-v})(t' - t)^v,$$

$$(4.33) \quad \mathbf{p}_t^R(x, y) \leq C(A, b, \bar{T}) (1 \wedge t^{-\frac{1}{2}}) e^{-b|x-y|(1 \wedge t^{-1/2})},$$

$$(4.34) \quad |\nabla_n \mathbf{p}_t^R(x, y)| \leq C(A, b, \bar{T}) (1 \wedge t^{-\frac{1+v}{2}}) |n|^v e^{-b|x-y|(1 \wedge t^{-1/2})},$$

Here $\nabla_n f(x) \stackrel{\text{def}}{=} f(x+n) - f(x)$ acts on the first variable of the functions.

PROOF. To prove (4.31), we use (4.1) to get

$$\mathbf{p}_t^R(x, y) \leq e^{-t'} e^{t'-t} \sum_{n=0}^{\infty} \frac{t'^n}{n!} p_n^e(x, y) = e^{t'-t} \mathbf{p}_{t'}^R(x, y),$$

where $p_n^e(x, y)$ is the transition probability of the discrete time elastic random walk.

For the other bounds, we use the explicit formula obtained in Lemma 4.5:

$$(4.35) \quad \begin{aligned} \mathbf{p}_t^R(x, y) &= p_t(x-y) + \mu_A p_t(x+y+1) \\ &+ (\mu_A^2 - 1) \sum_{z=-\infty}^{-2} p_t(x+y-z) \mu_A^{-z-2}, \end{aligned}$$

and the existing estimates for the entire line kernel $p_t(x)$.

To prove (4.32), we make use of [19, (A.10)]:

$$(4.36) \quad |p_{t'}(x) - p_t(x)| \leq C(1 \wedge t^{-\frac{1}{2}-v})(t' - t)^v,$$

and this together with (4.35) yields

$$|\mathbf{p}_{t'}^R(x, y) - \mathbf{p}_t^R(x, y)| \leq C(1 \wedge t^{-\frac{1}{2}-v})(t' - t)^v \left(1 + \mu_A + (\mu_A^2 - 1) \sum_{z=-\infty}^{-2} \mu_A^{-z-2} \right).$$

The last factor is bounded by a constant independent of ε . Indeed, this is obvious if $A = 0$ (i.e., $\mu = 1$); and if $A > 0$, the sum over z yields a factor $\frac{1}{1-\mu}$, which multiplied by $\mu_A^2 - 1$ is bounded by a constant independent of ε .

To prove (4.33), we use [19, (A.12)]:

$$(4.37) \quad p_t(x) \leq C(b) (1 \wedge t^{-\frac{1}{2}}) e^{-b|x|(1 \wedge t^{-1/2})}.$$

Note that by $e^{-b(x+y)(1 \wedge t^{-1/2})} \leq e^{-b|x-y|(1 \wedge t^{-1/2})}$ the first two terms in (4.35) satisfy the desired bound. Since $\mu_A \leq 1$ and $x, y \in \mathbb{Z}_{\geq 0}$,

$$\begin{aligned} & \sum_{z=-\infty}^{-2} p_t(x+y-z) \mu_A^{-z-2} \\ & \leq C(b) \sum_{z=-\infty}^{-2} (1 \wedge t^{-1/2}) e^{-b(x+y-z)(1 \wedge t^{-1/2})} \\ & \leq C(b) (1 \wedge t^{-1/2}) e^{-b(x+y)(1 \wedge t^{-1/2})} \frac{e^{-2b(1 \wedge t^{-1/2})}}{1 - e^{-b(1 \wedge t^{-1/2})}} \\ & \leq C(b) e^{-b(x+y)(1 \wedge t^{-1/2})} \end{aligned}$$

where we summed over z and used

$$(4.38) \quad e^{-q}/(1 - e^{-q}) \leq 1/q$$

for any $q \geq 0$. By the assumption on μ_A and using $t \leq \varepsilon^{-2}\bar{T}$, we have

$$|\mu_A^2 - 1| \leq C(A)\varepsilon \leq C(A, \bar{T})(1 \wedge t^{-\frac{1}{2}}).$$

So the last term of (4.35) also satisfies the desired bound and thus we obtain (4.33).

To prove (4.34), we can use [19, (A.13)]:

$$(4.39) \quad |\nabla_n p_t(x)| \leq C(b)(1 \wedge t^{-\frac{1+v}{2}}) |n|^v e^{-b|x|(1 \wedge t^{-1/2})},$$

and proceed in the same way as in the proof for (4.33) to obtain (4.34). \square

COROLLARY 4.12. *Let \mathbf{p}^R be the heat kernel on $\mathbb{Z}_{\geq 0}$ as above. Given any $a \geq 0$, for any $t \in [0, \bar{T}\varepsilon^{-2}]$, $x \in \mathbb{Z}_{\geq 0}$, we have*

$$(4.40) \quad \sum_{y \geq 0} \mathbf{p}_t^R(x, y) e^{a\varepsilon y} e^{a|x-y|(1 \wedge t^{-1/2})} \leq C(a, A, \bar{T}) e^{a\varepsilon x},$$

$$(4.41) \quad \sum_{y \geq 0} \nabla_x \mathbf{p}_t^R(x, y) e^{a\varepsilon y} e^{a|x-y|(1 \wedge t^{-1/2})} \leq C(a, A, \bar{T}) e^{a\varepsilon x} t^{-1/2}.$$

PROOF. To prove (4.40), note that its LHS is bounded above by

$$(4.42) \quad e^{a\varepsilon x} \sum_{y \geq 0} \mathbf{p}_t^R(x, y) e^{a|x-y|(\varepsilon + (1 \wedge t^{-1/2}))},$$

and applying (4.33) this is further bounded above by

$$C(A, b, \bar{T}) e^{a\varepsilon x} \sum_{z \in \mathbb{Z}} (1 \wedge t^{-\frac{1}{2}}) e^{-b|z|(1 \wedge t^{-1/2})} e^{a|z|(\varepsilon + (1 \wedge t^{-1/2}))}$$

for any $b > 0$. Without loss of generality we assume $\varepsilon^2 < \bar{T}$, and since $t \leq \varepsilon^{-2}\bar{T}$, one has $\bar{T}^{1/2}(1 \wedge t^{-1/2}) > \varepsilon$. Choosing $b = 2a(\bar{T}^{1/2} + 1)$, the above quantity is

bounded above by

$$C(A, a, \bar{T}) e^{a\epsilon x} \sum_{z \in \mathbb{Z}} (1 \wedge t^{-\frac{1}{2}}) e^{-\frac{b}{2}|z|(1 \wedge t^{-1/2})}.$$

(4.40) then follows by performing the sum over z and applying (4.38).

The estimate (4.41) follows in the same way using (4.34) in place of (4.33). \square

PROPOSITION 4.13. *Assume that \mathbf{p}^R is the heat kernel on $\{0, 1, \dots, N\}$ with Robin boundary condition*

$$\begin{aligned} \mathbf{p}_t^R(-1, y) &= \mu_A \mathbf{p}_t^R(0, y), & \mathbf{p}_t^R(N+1, y) &= \mu_B \mathbf{p}_t^R(N, y) \\ \forall 0 \leq t \leq \varepsilon^{-2}\bar{T}, & 0 \leq y \leq N, \end{aligned}$$

with $\mu_A = 1 - \varepsilon A$ and $\mu_B = 1 - \varepsilon B$ for constants $A > 0, B > 0$. Given any $b \geq 0$, for any $|n| \leq \lceil t^{1/2} \rceil$, $v \in [0, 1]$, $0 \leq t < t' < \varepsilon^{-2}\bar{T}$, and $0 \leq x, y \leq N$, we have all the bounds stated in Proposition 4.11 where the constants depend on A and B now.

PROOF. The proof for (4.31) follows in the same way as in the half-line case.

Turning to (4.32), unlike in the half-line case, it does not seem to be easy to apply the standard heat kernel estimate (4.36) combined with (4.21) to reach the conclusion. This is because (4.21) involves summation over all periods while (4.36) does not capture any spatial decay. Instead of trying to improve upon estimates to follow this route, we observe below that (4.32) can be seen as a consequence of spectral properties of $-\frac{1}{2}\Delta$ as contained in Lemma 4.9.

First note that by (4.28)

$$\mathbf{p}_t^R(x, y) - \mathbf{p}_{t'}^R(x, y) = \sum_{k=1}^N \psi_{\lambda_k}(x) \psi_{\lambda_k}(y) (e^{-t\lambda_k} - e^{-t'\lambda_k}).$$

By Lemma 4.9 there exist constants C_1, C'_1, C_2, C'_2 such that

$$(4.43) \quad C'_1 \frac{k^2}{N^2} \leq C_1 w_k^2 \leq \lambda_k = 1 - \cos(\omega_k) \leq C_2 w_k^2 \leq C'_2 \frac{k^2}{N^2}.$$

Using the upper bound in (4.43), one has

$$|1 - e^{(t-t')\lambda_k}| \leq C |(t-t')\lambda_k|^v \leq C |t-t'|^v k^{2v} / N^{2v}$$

for $v \in [0, 1]$ and $\omega_k \leq \pi$. By Lemma 4.10 the eigenfunctions ψ_{λ_k} can be bounded by a constant times \sqrt{N} ; thus by the lower bound in (4.43)

$$\begin{aligned} |\mathbf{p}_t^R(x, y) - \mathbf{p}_{t'}^R(x, y)| &\leq \frac{C}{N} \sum_{k=1}^N e^{-t\lambda_k} |t-t'|^v \left(1 \wedge \frac{k^{2v}}{N^{2v}}\right) \\ &\leq \frac{C}{N} \sum_{k=1}^N e^{-C'_1 t k^2 / N^2} |t-t'|^v \left(1 \wedge \frac{k^{2v}}{N^{2v}}\right). \end{aligned}$$

Summing over k we obtain the desired bound (4.32).

To prove (4.33), we invoke Lemma 4.7 and use the standard heat kernel bound (4.37). Since the terms with $|k| \leq 2$ can be all bounded using (4.37), one only needs to deal with the terms with $|k| > 2$. The second term on the RHS of (4.21) can be bounded by

$$C(\tilde{b})(1 \wedge t^{-\frac{1}{2}}) \sum_{k>2} I_k e^{-\tilde{b}|x-t(y;k)|(1 \wedge t^{-1/2})} \leq \\ C(\tilde{b})(1 \wedge t^{-\frac{1}{2}}) e^{-\tilde{b}N(1 \wedge t^{-1/2})} \cdot \frac{e^{-2\tilde{b}N(1 \wedge t^{-1/2})}}{1 - e^{-\tilde{b}N(1 \wedge t^{-1/2})}}$$

since $|I_k| \leq 1$. Using $e^{-2q}/(1 - e^{-q}) \leq 1/q$ for any $q \geq 0$, the above expression is bounded by

$$C(\tilde{b})(1 \wedge t^{-\frac{1}{2}}) \frac{e^{-\tilde{b}N(1 \wedge t^{-1/2})}}{\tilde{b}N(1 \wedge t^{-1/2})} \leq C(\bar{T}, \tilde{b})(1 \wedge t^{-\frac{1}{2}}) e^{-\tilde{b}|x-y|(1 \wedge t^{-1/2})}$$

where we used $t \leq \varepsilon^{-2}\bar{T}$ and $\varepsilon = 1/N$.

By using (4.37) and the bound for E_k in Lemma 4.6, the last term on the RHS of (4.21) can be bounded by

$$\varepsilon C(\tilde{b}) \sum_{k>2} \sum_{\bar{y}=k\bar{N}}^{(k+1)\bar{N}-1} (1 \wedge t^{-\frac{1}{2}}) e^{-\tilde{b}|x-\bar{y}|(1 \wedge t^{-1/2})} C_0^{|k|} \leq \\ C(\tilde{b}, C_0)(1 \wedge t^{-\frac{1}{2}}) e^{-(\tilde{b}N(1 \wedge t^{-1/2}))} \sum_{k>0} e^{(-\tilde{b}N(1 \wedge t^{-1/2}) + \log C_0)|k|}$$

where we used $|x - \bar{y}| \geq (|k| - 1)N$ and replaced the sum over \bar{y} by a factor $N = 1/\varepsilon$. As in the proof of Corollary 4.12, we have $N\bar{T}^{1/2}(1 \wedge t^{-1/2}) > 1$. Choosing $\tilde{b} = 2 \log C_0 T^{1/2}$ and summing over k , the above quantity is bounded by (noting that $|x - y| \leq N$)

$$C(A, B, b, \bar{T})(1 \wedge t^{-\frac{1}{2}}) e^{-b|x-y|(1 \wedge t^{-1/2})}.$$

The proof of (4.33) is then completed.

The proof of (4.34) follows analogously by using the standard kernel estimates (4.39) together with (4.21). \square

COROLLARY 4.14. *Let \mathbf{p}^R be the heat kernel on $\{0, 1, \dots, N\}$ with Robin boundary condition as above. For any $t \in [0, \bar{T}\varepsilon^{-2}]$, $x \in \{0, 1, \dots, N\}$, we have*

$$(4.44) \quad \sum_{y=0}^N \mathbf{p}_t^R(x, y) \leq C(A, B, \bar{T}), \\ \sum_{y=0}^N \nabla_x \mathbf{p}_t^R(x, y) e^{a|x-y|(1 \wedge t^{-1/2})} \leq C(A, B, \bar{T}) t^{-\frac{1}{2}}.$$

PROOF. These are consequences of (4.33) and (4.34), and the proof is analogous with that of Proposition 4.12. \square

4.2 Proof of Tightness

The rest of this section is devoted to the proofs of Propositions 4.15 and 4.16, which will yield tightness of the rescaled process. Recall that

$$\|f_t(x)\|_n \stackrel{\text{def}}{=} (\mathbf{E}|f_t(x)|^n)^{\frac{1}{n}}$$

denotes the L^n norm. Recall also the definitions given in Section 2. The following two propositions provide vital continuity estimates from which the tightness follows by standard arguments.

PROPOSITION 4.15. *Let Z be the Gärtner transformed process for ASEP-H. Fix $\bar{T} < \infty$, $n \in \mathbb{N}$, $\alpha \in (0, \frac{1}{2})$, and some near-equilibrium initial conditions as in Assumption 2.14, with the corresponding finite constant a . Then there exists some finite constant C such that*

$$(4.45) \quad \|Z_t(x)\|_{2n} \leq C e^{a\epsilon x}$$

$$(4.46) \quad \|Z_t(x) - Z_t(x')\|_{2n} \leq C(\epsilon|x - x'|)^\alpha e^{a\epsilon(x+x')}$$

$$(4.47) \quad \|Z_t(x) - Z_{t'}(x)\|_{2n} \leq C\epsilon^\alpha (1 \vee |t' - t|^{\frac{\alpha}{2}}) e^{2a\epsilon x}$$

for all $t, t' \in [0, \epsilon^{-2}\bar{T}]$ and $x, x' \in \mathbb{R}_+$ with $|x - x'| \leq \epsilon^{-1}$.

PROPOSITION 4.16. *Let Z be the Gärtner transformed process for ASEP-B. Fix $\bar{T} < \infty$, $n \in \mathbb{N}$, $\alpha \in (0, \frac{1}{2})$, and some initial conditions as in Assumption 2.15. Then there exists a finite constant C such that*

$$(4.48) \quad \|Z_t(x)\|_{2n} \leq C,$$

$$(4.49) \quad \|Z_t(x) - Z_t(x')\|_{2n} \leq C(\epsilon|x - x'|)^\alpha,$$

$$(4.50) \quad \|Z_t(x) - Z_{t'}(x)\|_{2n} \leq C\epsilon^\alpha (1 \vee |t' - t|^{\frac{\alpha}{2}}),$$

for all $t, t' \in [0, \epsilon^{-2}\bar{T}]$ and $x, x' \in [0, N]$.

Recall the scaled process \mathcal{Z}^ϵ from Definition 2.16.

PROPOSITION 4.17. *For ASEP-H with near equilibrium initial conditions as in Assumption 2.14, the law of $\{\mathcal{Z}^\epsilon\}_\epsilon$ is tight in $D([0, \bar{T}]; C(\mathbb{R}_+))$ and any limit point of $\{\mathcal{Z}^\epsilon\}_\epsilon$ is in $C([0, \bar{T}]; C(\mathbb{R}_+))$. For ASEP-B with boundary conditions as in Assumption 2.15 the law of $\{\mathcal{Z}^\epsilon\}_\epsilon$ is tight in $D([0, \bar{T}]; C([0, 1]))$ and any limit point of $\{\mathcal{Z}^\epsilon\}_\epsilon$ is in $C([0, \bar{T}]; C([0, 1]))$.*

PROOF. Let $I = \mathbb{R}_+$ for ASEP-H and $I = [0, 1]$ for ASEP-B. First of all, the bounds for (4.45) and (4.46), or (4.48) and (4.49), imply tightness in $C(I)$; see, for instance, [9]. Then (4.47) or (4.50) and [7, prop. 4.9] (and the arguments below proposition 4.9 in [7] with $C(\mathbb{R})$ replaced by $C(I)$) combine to imply tightness in $D([0, \bar{T}]; C(I))$ and that the limiting points must lie in $C([0, \bar{T}]; C(I))$. The only necessary change in the arguments [7, prop. 4.9] is that the metric on $C(\mathbb{R}_+)$ is

defined as $\rho(f, g) = \sum_{n=1}^{\infty} \frac{1}{2^n} (1 \wedge \max_{x \in [0, n]} |f(x) - g(x)|)$, and for $C([0, 1])$ we can simply use the C^∞ norm. \square

The following technical result is useful for L^n estimates.

LEMMA 4.18. *Let $\Lambda \in \{\mathbb{Z}_{\geq 0}, \{0, 1, \dots, N\}\}$. Given any $n \in \mathbb{N}$, there exists a finite constant C such that, for any deterministic function $f_s(x, x'): [0, \infty) \times \Lambda^2 \rightarrow \mathbb{R}$ and any $t \leq t' \in [0, \infty)$ with $t' - t \geq 1$,*

$$(4.51) \quad \left\| \int_t^{t'} \sum_{x' \in \Lambda} f_s(x, x') dM_s(x') \right\|_{2n}^2 \leq C \varepsilon \int_t^{t'} \sum_{x' \in \Lambda} \bar{f}_s(x, x')^2 \|Z_s^2\|_n(x') ds$$

where

$$(4.52) \quad \bar{f}_s(x, x') \stackrel{\text{def}}{=} \sup_{|s' - s| \leq 1} |f_{s'}(x, x')|$$

and M is the martingale introduced in Lemma 3.1.

PROOF. This is essentially [19, lemma 3.1]; see also [15, lemma 3.1]. Fixing t and calling $R_{t'}(x)$ the integral on the left-hand side of (4.51), by the Burkholder-Davis-Gundy (BDG) inequality, one only needs to bound $\|[R_\bullet(x)]_{t'}\|_n$, where $[R_\bullet(x)]_{t'}$ is the optional quadratic variation, i.e., the sum of the squares of all the jumps of $R_\bullet(x)$ over time $(t, t']$. The only inputs from the martingale M to the proof in [19, lemma 3.1] are that a jump of $M(x')$ at time s equals

$$((q/p)^{\pm 1} - 1) Z_{s-}(x')^2 \leq C \varepsilon Z_{s-}(x')^2,$$

and that

$$(4.53) \quad \sup_{s \in (s_1, s_2]} Z_s(x') \leq e^{2\sqrt{\varepsilon} N_I(x')} Z_{s_1}(x')$$

where $s_1, s_2 \in I$ and $N_I(x')$ is the number of jumps occurring at x' during a unit time interval I that is stochastically bounded by a Poisson random variable with constant rate; these remain true in our case, and therefore the detailed proof is omitted. \square

The rest of this section is devoted to the proofs of Propositions 4.15 and 4.16. We rewrite the discrete SHE (3.1) in the following integrated form:

$$(4.54) \quad Z_t(x) = \sum_{x' \in \Lambda} \mathbf{p}_t^R(x, x') Z_0(x') + \int_0^t \sum_{x' \in \Lambda} \mathbf{p}_{t-s}^R(x, x') dM_s(x'),$$

with $\Lambda \in \{\mathbb{Z}_{\geq 0}, \{0, 1, \dots, N\}\}$ depending on the half-line or bounded interval case.

We start with the proof of Proposition 4.15.

PROOF OF PROPOSITION 4.15. In what follows, the value of constants may change from line to line (and within lines). Let I_1 and I_2 denote the first and second terms on the RHS of (4.54), respectively.

We begin by proving the *uniform bound* (4.45). First, by (4.40) we have the following bound on the Robin heat kernel:

$$(4.55) \quad \sum_{y \geq 0} \mathbf{p}_t^R(x, y) e^{a\epsilon y} \leq C e^{a\epsilon x} \quad \text{for } t \leq \varepsilon^{-2}\bar{T}.$$

For I_1 , by the triangle inequality we have

$$\|I_1(t, x)^2\|_n = \|I_1(t, x)\|_{2n}^2 \leq \left(\sum_{x'} \mathbf{p}_t^R(x, x') \|Z_0(x')\|_{2n} \right)^2.$$

Combining this with (4.55) and (2.16), we obtain

$$(4.56) \quad \|I_1(t, x)^2\|_n \leq C e^{2a\epsilon x}.$$

Turning to bounding I_2 , we assume $t \geq 1$ and apply Lemma 4.18 with $f_s(x, x') = \mathbf{p}_{t-s}^R(x, x')$ to obtain

$$\|I_2(t, x)^2\|_n \leq C \varepsilon \int_0^t \sum_{y \geq 0} \bar{\mathbf{p}}_{t-s}^2(x, y) \|Z_s^2\|_n(y) ds$$

where $\bar{\mathbf{p}}$ is the local supremum of \mathbf{p}^R defined as in (4.52). By (4.31) we have $\mathbf{p}_t^R \leq C \mathbf{p}_{t'}^R$ for $0 \leq t' - t \leq 1$, and by (4.33) we have the estimate $\mathbf{p}_t^R \leq C t^{-1/2}$. Therefore $\bar{\mathbf{p}}_{t-s}^2(x, y) \leq C(t-s)^{-1/2} \mathbf{p}_{t-s}^R(x, y)$ and hence

$$\|I_2(t, x)^2\|_n \leq C \varepsilon \int_0^t (t-s)^{-1/2} \left(\sum_{y \geq 0} \mathbf{p}_{t-s}^R(x, y) \|Z_s^2\|_n(y) \right) ds \quad \text{for } t \geq 1.$$

Combining this with (4.56) yields

$$(4.57) \quad \|Z_t^2(x)\|_n \leq C e^{2a\epsilon x} + C \varepsilon \int_0^t (t-s)^{-1/2} \left(\sum_{y \geq 0} \mathbf{p}_{t-s}^R(x, y) \|Z_s^2\|_n(y) \right) ds.$$

The bound (4.57) was derived for $t \geq 1$, but in fact it also holds true for $t \leq 1$. This is so because, by (4.45) and (4.53) with $(s_1, s_2] = (0, t]$, we already have $\|Z_t^2(x)\|_{2n} \leq C e^{2a\epsilon x}$ for $t \leq 1$. With this, iterating this inequality and using the semigroup property of the heat kernel \mathbf{p}^R and (4.55), we then arrive at

$$\|Z_t^2(x)\|_n \leq \left(C e^{2a\epsilon x} + \sum_{j=1}^{\infty} \frac{C^j}{j!} \left(\varepsilon \int_0^t s^{-1/2} ds \right)^j e^{2a\epsilon x} \right).$$

With $t \leq \varepsilon^{-2}\bar{T}$, the desired result (4.45) follows.

Now we turn to proving the *spatial Hölder estimate* (4.46). For this we extend Z_0 to the entire \mathbb{Z} (we still denote it by Z_0) such that $Z_0(x-1) - \mu_A Z_0(x)$ is odd in x :

$$(4.58) \quad Z_0(x-1) - \mu_A Z_0(x) = -(Z_0(-x-1) - \mu_A Z_0(-x)), \quad x < 0.$$

As in Section 4.1, this implies that

$$(4.59) \quad I_1(t, x) = \sum_{y \geq 0} \mathbf{p}_t^R(x, y) Z_0(y) = \sum_{y \in \mathbb{Z}} p_t(x - y) Z_0(y) = p_t * Z_0(x).$$

The advantage of the extension is that now $p_t(x - y)$ only depends on the difference between x and y . By the triangular inequality, we have

$$(4.60) \quad \|I_1(t, x) - I_1(t, x')\|_{2n}^2 \leq \left(\sum_{\bar{x} \in \mathbb{Z}} p_t(\bar{x}) \|Z_0(x - \bar{x}) - Z_0(x' - \bar{x})\|_{2n} \right)^2.$$

By (2.17) we have $\|Z_0(x) - Z_0(x')\|_n \leq C(\varepsilon|x - x'|)^\alpha e^{a\varepsilon(x+x')}$ for $x, x' \in \mathbb{Z}_{\geq 0}$, where $\alpha \in (0, \frac{1}{2})$. This is also true for the extended Z_0 with any x, x' . To see this, rewrite (4.58) as (recall that $\mu = 1 - \varepsilon A$)

$$Z_0(x-1) - Z_0(x) = -(Z_0(-x-1) - Z_0(-x)) - \varepsilon A(Z_0(x) + Z_0(-x)) \quad x < 0.$$

Then, for $x < x' < 0$, sum the above identity over the points between x and x' , which yields

$$(4.61) \quad Z_0(x) - Z_0(x') = Z_0(-x-1) - Z_0(-x'-1) - \varepsilon A \sum_{z=x'}^{x+1} (Z_0(z) + Z_0(-z)).$$

Therefore, by (4.61), (2.16), and (2.17),

$$\begin{aligned} \|Z_0(x) - Z_0(x')\|_n &\leq C(\varepsilon|x - x'|)^\alpha e^{a\varepsilon(x+x')} + C\varepsilon|x - x'| e^{a\varepsilon(x+x')} \\ &\leq C(\varepsilon|x - x'|)^\alpha e^{a\varepsilon(x+x')} \end{aligned}$$

since $|x - x'| \leq 1/\varepsilon$ and $\alpha \in (0, \frac{1}{2})$. For $x < 0 < x'$ we have the same bound by $|x|^\alpha + |x'|^\alpha \leq C(x' - x)^\alpha$. For the *standard* heat kernel p , $\sum_{\bar{x} \in \mathbb{Z}} p_t(\bar{x}) e^{a\varepsilon\bar{x}} \leq C$; thus the RHS of (4.60) is bounded by

$$(4.62) \quad \left(\sum_{\bar{x} \in \mathbb{Z}} p_t(\bar{x}) (\varepsilon|x - x'|)^\alpha e^{a\varepsilon(|x - \bar{x}| + |x' - \bar{x}|)} \right)^2 \leq C(\varepsilon|x - x'|)^{2\alpha} e^{2a\varepsilon(x+x')}.$$

For $\|I_2(t, x) - I_2(t, x')\|_{2n}^2$, we apply Lemma 4.18 with $f_s(x, \bar{x}) = \mathbf{p}_{t-s}^R(x', \bar{x}) - \mathbf{p}_{t-s}^R(x, \bar{x})$ and use the already-proved result (4.45) to bound $\|Z_s^2(x')\|_n$ by $e^{2a\varepsilon x'}$. We then use the fact that

$$\begin{aligned} (\mathbf{p}_{t-[s]}^R(x', \bar{x}) - \mathbf{p}_{t-[s]}^R(x, \bar{x}))^2 &\leq \\ &|\mathbf{p}_{t-[s]}^R(x', \bar{x}) - \mathbf{p}_{t-[s]}^R(x, \bar{x})| (\mathbf{p}_{t-[s]}^R(x', \bar{x}) + \mathbf{p}_{t-[s]}^R(x, \bar{x})), \end{aligned}$$

and for the first factor we apply the L^∞ gradient estimate (4.34), which implies

$$|\mathbf{p}_{t-[s]}^R(x', \bar{x}) - \mathbf{p}_{t-[s]}^R(x, \bar{x})| \leq C(1 \wedge (t-s)^{-\frac{1}{2}-\frac{\alpha}{2}}) |x - x'|^\alpha,$$

and for the second factor we use the L^1 bound (4.55) to integrate out the weight $e^{2a\varepsilon x'}$. This yields

$$\|I_2(t, x) - I_2(t, x')\|_{2n}^2 \leq C\varepsilon|x - x'|^\alpha e^{2a\varepsilon(x+x')} \int_0^t (t-s)^{-\frac{1}{2}-\frac{\alpha}{2}} ds.$$

Noting that $\int_0^t (t-s)^{-1/2-\alpha/2} ds \leq Ct^{(1-\alpha)/2} \leq C(\bar{T})\varepsilon^{\alpha-1}$, we arrive at the bound (4.46).

Next we prove the *temporal Hölder estimate* (4.47). Without loss of generality, we assume $t < t' - 1$. For I_1 , we first note that $I_1(t, x) \geq 0$ since both \mathbf{p}^R and Z_0 are positive. We then use the semigroup property of \mathbf{p}^R and $\sum_y \mathbf{p}_{t'-t}^R(x, y) \leq 1$ (Lemma 4.2) to get

$$\begin{aligned} I_1(t', x) - I_1(t, x) &= \sum_{z \geq 0} \mathbf{p}_{t'-t}^R(x, z) I_1(t, z) - I_1(t, x) \\ (4.63) \quad &\leq \sum_{z \geq 0} \mathbf{p}_{t'-t}^R(x, z) (I_1(t, z) - I_1(t, x)). \end{aligned}$$

By the spatial Hölder estimate (4.46), we have

$$\|I_1(t, z) - I_1(t, x)\|_{2n} \leq C(\varepsilon|x - z|)^\alpha e^{a\varepsilon(x+z)}.$$

Also, since $|r|^\alpha \leq Ce^{|r|}$ for any r , one has

$$\begin{aligned} &\sum_{z \geq 0} |x - z|^\alpha \mathbf{p}_{t'-t}^R(x, z) e^{a\varepsilon(x+z)} \\ (4.64) \quad &\leq C \sum_{z \geq 0} e^{a|x-z|(1 \wedge (t'-t)^{-1/2})} (1 \vee (t' - t)^{\frac{\alpha}{2}}) \mathbf{p}_{t'-t}^R(x, z) e^{a\varepsilon(x+z)} \\ &\leq Ce^{2a\varepsilon x} (1 \vee |t' - t|^{\frac{\alpha}{2}}) \end{aligned}$$

where we applied Corollary 4.12 in the last inequality. Therefore one obtains the desired bound $\|I_1(t, x) - I_1(t, x)\|_{2n} \leq C\varepsilon^\alpha |t' - t|^{\alpha/2} e^{a\varepsilon|x|}$.

Regarding the term I_2 , we write $I_2(t', x) - I_2(t, x) = J_1(t, t', x) + J_2(t, t', x)$ where

$$\begin{aligned} J_1(t, t', x) &= \int_t^{t'} \sum_{y \geq 0} \mathbf{p}_{t'-s}^R(x, y) dM_s(y), \\ J_2(t, t', x) &= \int_0^t \sum_{y \geq 0} (\mathbf{p}_{t'-s}^R - \mathbf{p}_{t-s}^R)(x, y) dM_s(y). \end{aligned}$$

For the term J_1 , applying Lemma 4.18, then proceeding as above with the uniform bound (4.45) on Z , the bound (4.33), and Corollary 4.12, we obtain

$$\|J_1(t, t', x)\|_n^2 \leq C(\varepsilon^\alpha |t' - t|^{\frac{\alpha}{2}} e^{a\varepsilon x})^2.$$

As for J_2 , applying Lemma 4.18 using

$$(\mathbf{p}_{t'-s}^R - \mathbf{p}_{t-s}^R)^2 \leq |\mathbf{p}_{t'-s}^R - \mathbf{p}_{t-s}^R| (\mathbf{p}_{t'-s}^R + \mathbf{p}_{t-s}^R)$$

followed by the estimate (4.32)

$$|\mathbf{p}_{t'-s}^R - \mathbf{p}_{t-s}^R| \leq C(1 \wedge t^{-\frac{1}{2}-\alpha})(t' - t)^\alpha$$

together with again the uniform bound (4.45) and Corollary 4.12, one obtains the desired bound $\|J_2\|_{2n}^2 \leq C\varepsilon^{2\alpha}|t' - t|^\alpha e^{2a\varepsilon|x|}$. Combining all these bounds concludes the proof of the proposition. \square

PROOF OF PROPOSITION 4.16. The proof follows in the same way as the proof of Proposition 4.15 except that we simply take $a = 0$ (erasing the weights $e^{a\varepsilon x}$) and replace the summation domain \mathbb{Z}_+ of the spatial variables by $\{0, 1, \dots, N\}$, and then apply Proposition 4.13 and Corollary 4.14 instead of Proposition 4.11 and Corollary 4.12 for all the heat kernel estimates. Also, regarding the bound for $\|I_1(t, x) - I_1(t, x')\|_{2n}^2$ in the proof of the spatial Hölder estimate (4.46), for two points $x < x' < 0$ with $|x - x'| \leq N$ we need (4.61), and for two points $0 < x < x'$ with $|x - x'| \leq N$ we then replace the constant A in (4.61) by B . \square

5 Proof of the Main Theorem

5.1 A Crucial Cancellation

Now that we have proved the tightness result, we would like to identify a limit as the mild solution to SHE (2.6). We want to show that the martingale M converges to $\mathcal{L}\dot{W}$, so the bracket of M should behave like $\varepsilon\mathcal{L}^2$ where the factor ε is the correct scaling factor we need. In view of (3.5), however, there is another term $\nabla^+ Z_t(x)\nabla^- Z_t(x)$ in the bracket. The “key estimate” in [7, sec. 4.2] shows that this term is actually small; in fact, if Z_t is replaced by a heat kernel p_t , then one has the following identity ([7, lemma A.1] or [15, lemma 4.2]):

$$\sum_{x \in \mathbb{Z}} \int_0^\infty \nabla^+ p_t(x)\nabla^- p_t(x) dt = 0,$$

which is a crucial cancellation around which the key estimate revolves. Note that the integrand $\nabla^+ p_t(x)\nabla^- p_t(x)$ can be also written as $-\nabla^+ p_t(x)\nabla^+ p_t(x-1)$. We show an analogue of this identity in Proposition 5.1 below. On \mathbb{R} , the identity can be proved via integration by parts since the heat kernel is just a function of the difference $x - y$. In our present situation, this approach fails (the heat kernel depends on the actual values of x and y , not just their difference), and we develop below a new method of proof using Green’s functions. It is worth noting that the half-line identity (5.2) is exactly the same as that for the full line, while the bounded interval identity (5.1) involves nonzero contribution for all choices of x and \bar{x} —a new feature.

PROPOSITION 5.1. *For the Robin heat kernel \mathbf{p}^R on $\{0, 1, \dots, N\}$, one has*

$$(5.1) \quad \sum_{y=0}^N \int_0^\infty \nabla_x^+ \mathbf{p}_t^R(x, y)\nabla_x^+ \mathbf{p}_t^R(\bar{x}, y) dt = \mathbf{1}_{x=\bar{x}}(1-c) - \mathbf{1}_{x \neq \bar{x}}c$$

for all $x, \bar{x} \in \{0, 1, \dots, N-1\}$, where the constant c is independent of x, \bar{x} and such that $0 \leq c \leq C\varepsilon$ for some constant $C > 0$, and $c = 0$ if $\mu_A = 1$ or $\mu_B = 1$. For the Robin heat kernel \mathbf{p}^R on $\mathbb{Z}_{\geq 0}$, one has

$$(5.2) \quad \sum_{y=0}^{\infty} \int_0^{\infty} \nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^+ \mathbf{p}_t^R(\bar{x}, y) dt = \mathbf{1}_{x=\bar{x}}$$

for all $x, \bar{x} \geq 0$.

PROOF. For the case of the finite intervals, using the spectral decomposition in (4.28) ($\{\lambda_k\}_{k=0}^N$ are the eigenvalues of $-\frac{1}{2}\Delta$ with (μ_A, μ_B) -Robin boundary condition and $\{\psi_k\}_{k=0}^N$ are the corresponding eigenfunctions), one has

$$(5.3) \quad \sum_{y=0}^N \int_0^{\infty} \nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^+ \mathbf{p}_t^R(\bar{x}, y) dt = \sum_{y=0}^N \int_0^{\infty} \sum_{k=0}^N \sum_{\bar{k}=0}^N \nabla_x^+ \psi_k(x) \psi_k(y) e^{-\lambda_k t} \nabla_x^+ \psi_{\bar{k}}(\bar{x}) \psi_{\bar{k}}(y) e^{-\lambda_{\bar{k}} t} dt.$$

The eigenfunctions ψ_k are orthonormal, i.e., $\sum_y \psi_k(y) \psi_{\bar{k}}(y) = \delta_{k\bar{k}}$, since the Laplacian with Robin boundary conditions is a finite symmetric matrix. Using this and performing the summation over y , we obtain that the above expression in (5.3) equals

$$(5.4) \quad \int_0^{\infty} \sum_{k=0}^N \nabla_x^+ \psi_k(x) \nabla_x^+ \psi_k(\bar{x}) e^{-2\lambda_k t} dt = \sum_{k=0}^N \frac{\nabla_x^+ \psi_k(x) \nabla_x^+ \psi_k(\bar{x})}{2\lambda_k} \stackrel{\text{def}}{=} F(x, \bar{x}).$$

By Lemma 5.2, it turns out that $F(x, x)$ does not depend on x . Consider $F(0, 0)$. Let G^R be the Green's function of $-\frac{1}{2}\Delta$ with the same Robin boundary condition. Note that G^R is symmetric: $G^R(x, y) = G^R(y, x)$. Since

$$(5.5) \quad \begin{aligned} -1 &= \frac{1}{2} \Delta_x G^R(x, 0) \Big|_{x=0} = \frac{1}{2} G^R(1, 0) - \frac{1}{2} (2 - \mu_A) G^R(0, 0), \\ 0 &= \frac{1}{2} \Delta_x G^R(x, 1) \Big|_{x=0} = \frac{1}{2} G^R(1, 1) - \frac{1}{2} (2 - \mu_A) G^R(0, 1), \end{aligned}$$

we have, using the identity (5.7) from Lemma 5.2 below,

$$(5.6) \quad 2F(0, 0) = G^R(0, 0) + G^R(1, 1) - 2G^R(1, 0) = (\mu_A - 1)^2 G^R(0, 0) + 2\mu_A$$

where in the last step we solved for $G^R(0, 1)$ and $G^R(1, 1)$ in terms of $G^R(0, 0)$ from (5.5).

Now if $\mu_A = 1$, then $F(0, 0) = 1$, and invoking (5.8) from Lemma 5.2 below, we obtain (5.1) with $c = 0$. We therefore assume $\mu_A < 1$, namely $A > 0$ below.

Invoking the formula (5.9) for $G^R(0, 0)$ in Lemma 5.3 below, and recalling that $\mu_A = 1 - \varepsilon A$ and $\mu_B = 1 - \varepsilon B$, we can straightforwardly check from (5.6) that

$$F(0, 0) = \frac{(\varepsilon A)^2(1 + N\varepsilon B)}{\varepsilon(A + B) + AB\varepsilon^2 N} + (1 - \varepsilon A) = \frac{A + B + AB\varepsilon(N - 1)}{A + B + AB\varepsilon N}.$$

With $N = 1/\varepsilon$, we have

$$F(0, 0) = \frac{A + B + AB - AB\varepsilon}{A + B + AB},$$

from which we immediately see that the constant c in (5.8) below satisfies $0 \leq c \leq C\varepsilon$ for some constants $C > 0$ that depend only on A, B . Therefore invoking (5.8) again one has $F(x, y) = -c$, where c is as above if $x \neq y$, and thus (5.1) is proved.

To prove (5.2) for the case of the half-line, we start with a Robin heat kernel \mathbf{p}^R on $\{0, 1, \dots, N\}$ with μ_A and, say, $\mu_B = 0$. The same arguments above lead us to (5.6), namely,

$$\sum_{y=0}^N \int_0^\infty \nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^+ \mathbf{p}_t^R(\bar{x}, y) dt = \frac{1}{2}(\mu_A - 1)^2 G^R(0, 0) + \mu_A - \mathbf{1}_{x \neq \bar{x}}.$$

Taking the limit $N \rightarrow \infty$ and applying (5.10) of Lemma 5.3 to the above equation, we obtain (5.2). \square

LEMMA 5.2. *Let $F(x, y)$ be defined as in (5.4). Then the function F can be represented as*

$$(5.7) \quad F(x, y) = \frac{1}{2}(G^R(x, y) + G^R(x+1, y+1) - G^R(x+1, y) - G^R(x, y+1)),$$

where G^R is the Green's function of $-\frac{1}{2}\Delta$ with Robin boundary condition. Furthermore, there exists a constant c (possibly depending on ε) such that

$$(5.8) \quad F(x, y) = \mathbf{1}_{x=y}(1 - c) - \mathbf{1}_{x \neq y}c.$$

In other words, F is equal to a constant on the diagonal and is equal to another constant off diagonal, and these two constants differ by 1.

PROOF. The proof of (5.7) is simple. Indeed, since

$$G^R(x, y) = \sum_k \lambda_k^{-1} \psi_k(x) \psi_k(y),$$

it is clear that

$$F(x, y) = \sum_k (2\lambda_k)^{-1} (\psi_k(x+1) - \psi_k(x))(\psi_k(y+1) - \psi_k(y))$$

is equal to RHS of (5.7). It is also clear that $F(x, y) = F(y, x)$.

To prove (5.8), note that for all $x \in \{0, 1, \dots, N-2\}$,

$$\begin{aligned} 2\nabla_x^+ F(x, y) &= (G^R(x+1, y) + G^R(x+2, y+1) - G^R(x+2, y) - G^R(x+1, y+1)) \\ &\quad - (G^R(x, y) + G^R(x+1, y+1) - G^R(x+1, y) - G^R(x, y+1)) \\ &= -\Delta_x G^R(x+1, y) + \Delta_x G^R(x+1, y+1) \\ &= 2(\mathbf{1}_{x+1=y} - \mathbf{1}_{x=y}). \end{aligned}$$

So F only changes values when crossing the diagonal:

$$\nabla_x^+ F(x, y) = \begin{cases} -1 & \text{if } x = y, \\ 1 & \text{if } x+1 = y, \\ 0 & \text{otherwise.} \end{cases}$$

This immediately yields (5.8). \square

LEMMA 5.3. Assume that $\mu_A < 1$ or $\mu_B < 1$. Let G^R be the Green's function of $-\frac{1}{2}\Delta$ with Robin boundary condition as above. On the finite interval, one has

$$(5.9) \quad G^R(0, 0) = 2 \cdot \frac{N+1 - N\mu_B}{N+2 - (N+1)(\mu_A + \mu_B) + N\mu_A\mu_B}.$$

On \mathbb{Z}_{\geq} one has

$$(5.10) \quad G^R(0, 0) = \frac{2}{1 - \mu_A}.$$

PROOF. If $N = 1$, then the operator $-\Delta$ on $\{0, 1\}$ with Robin boundary condition is the 2×2 matrix

$$\begin{bmatrix} 2 - \mu_A & -1 \\ -1 & 2 - \mu_B \end{bmatrix}_{2 \times 2}.$$

The first entry (i.e., the upper-left one) of its inverse matrix can be directly computed, which is exactly (5.9) with $N = 2$. In fact, the numerator $2 - \mu_B$ on the RHS of (5.9) is precisely the minor of the above matrix deleting the first row and the first column, and the denominator $3 - 2(\mu_A + \mu_B) + \mu_A\mu_B$ of the RHS of (5.9) is the determinant of the above 2×2 matrix.

We prove (5.9) for general N by induction and Cramer's rule. Suppose that for all numbers less than N , the determinant of $-\Delta$ is given by the denominator of the RHS of (5.9), and the cofactor of $2 - \mu_A$ (i.e., the minor by deleting the first row and the first column) is given by the numerator of the RHS of (5.9) (so in particular (5.9) holds for all numbers less than N), we show that this is also the case for N .

If we denote by M_N the cofactor of the first (i.e., upper-left) entry $2 - \mu_A$ (which is a determinant of an $N \times N$ matrix) of the matrix $-\Delta$, namely

$$-\Delta = \begin{bmatrix} 2 - \mu_A & (-1, 0, \dots, 0) \\ (-1, 0, \dots, 0)^\top & * \end{bmatrix}_{(N+1) \times (N+1)}, \quad \det(*) = M_N,$$

then it is easy to see that (M_0 is understood as 1)

$$M_N = 2 \times M_{N-1} - M_{N-2}.$$

Using the induction assumption,

$$M_N = 2(N - (N - 1)\mu_B) - (N - 1 - (N - 2)\mu_B) = (N + 1) - N\mu_B,$$

which is the desired numerator.

Turning to the determinant of $-\Delta$, we have

$$\det(-\Delta) = (2 - \mu_A) \times M_N - M_{N-1},$$

and again using the induction assumption together with the formula for M_N we have just proved, we have

$$\det(-\Delta) = (2 - \mu_A)(N + 1 - N\mu_B) - (N - (N - 1)\mu_B),$$

which is equal to the denominator of (5.9). Therefore (5.9) holds for all N .

The half-line case (5.10) follows immediately by taking $N \rightarrow \infty$. \square

COROLLARY 5.4. *Let $\bar{T} > 0$, $a \geq 0$. For the Robin heat kernel \mathbf{p}^R on $\mathbb{Z}_{\geq 0}$, there exist $\varepsilon_0 > 0$ and $c_\star < 1$ such that*

$$(5.11) \quad \sum_{y \geq 1} \int_0^{\varepsilon^{-2}\bar{T}} |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| e^{a\varepsilon|x-y|} dt \leq c_\star$$

for every $x \geq 1$, $\varepsilon < \varepsilon_0$. Moreover, for any $S \in [0, \bar{T}]$, there exists $C > 0$ such that with $s := \varepsilon^{-2}S$ one has

$$(5.12) \quad \sum_{y \geq 1} \int_0^s |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| e^{a\varepsilon|x-y|} (s-t)^{-\frac{1}{2}} dt \leq C\varepsilon$$

for every $x \geq 1$, $\varepsilon < \varepsilon_0$. For the Robin heat kernel \mathbf{p}^R on $\{0, 1, \dots, N\}$, the above bounds hold for every $x \in \{1, 2, \dots, N-1\}$ with $a = 0$ and y summing from 1 to $N-1$.

Before our proof, we remark that the proof of these estimates for the standard heat kernel on the entire line can be found in [7, lemmas A.2 and A.3].

PROOF. We first consider the finite interval case. To prove (5.11), using the Cauchy-Schwarz inequality, the LHS of (5.11) with $a = 0$ is strictly smaller than

$$(5.13) \quad \left(\sum_{y=0}^N \int_0^\infty (\nabla_x^+ \mathbf{p}_t^R(x, y))^2 dt \right)^{\frac{1}{2}} \left(\sum_{y=0}^N \int_0^\infty (\nabla_x^- \mathbf{p}_t^R(x, y))^2 dt \right)^{\frac{1}{2}},$$

which is bounded by 1 by Proposition 5.1 with $x = \bar{x}$; here the Cauchy-Schwarz inequality is strict because $\nabla_x^+ \mathbf{p}^R \neq \nabla_x^- \mathbf{p}^R$. This is the argument in [7] and is sufficient to show that the LHS of (5.11) is bounded by a constant strictly smaller than 1. However, to show that this constant c_\star is actually uniform in x , we need to

be more careful. We apply the Lagrange identity (which captures the sharpness of Cauchy-Schwarz inequality):

$$\begin{aligned}
& \left(\sum_{y=0}^N (\nabla_x^+ \mathbf{p}_t^R(x, y))^2 \right) \left(\sum_{y=0}^N (\nabla_x^- \mathbf{p}_t^R(x, y))^2 \right) \\
(5.14) \quad & - \left(\sum_{y=0}^N |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| \right)^2 \\
& = \sum_{y, \bar{y}=0}^N (|\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, \bar{y})| - |\nabla_x^+ \mathbf{p}_t^R(x, \bar{y}) \nabla_x^- \mathbf{p}_t^R(x, y)|)^2.
\end{aligned}$$

It is clear that if $\nabla_x^+ \mathbf{p}^R$ and $\nabla_x^- \mathbf{p}^R$ were equal the RHS of (5.14) would be 0. We claim that there exists $t_0 > 0$ and $\varepsilon_0 > 0$ such that for all $t \leq t_0$, all $x \in \{1, 2, \dots, N-1\}$, and all $\varepsilon < \varepsilon_0$, one has

$$(5.15) \quad \nabla_x^\pm \mathbf{p}_t^R(x, x) \leq -\frac{9}{10} \quad \text{and} \quad |\nabla_x^+ \mathbf{p}_t^R(x, x-1)| \vee |\nabla_x^- \mathbf{p}_t^R(x, x+1)| \leq \frac{1}{10}.$$

Here ∇_x^\pm only acts on the *first* variable of \mathbf{p}^R as before. (In fact, in the special case $t = 0$ the claim obviously holds since the first quantity is equal to -1 and the second equal to 0.) If we assume this claim, for $t \leq t_0$ the RHS of (5.14) is bounded from below by the term² with $y = x$ and $\bar{y} = x + 1$,

$$(|\nabla_x^+ \mathbf{p}_t^R(x, x) \nabla_x^- \mathbf{p}_t^R(x, x+1)| - |\nabla_x^+ \mathbf{p}_t^R(x, x+1) \nabla_x^- \mathbf{p}_t^R(x, x)|)^2 \geq \frac{1}{2},$$

where the last step used (5.15). Therefore for all t , (5.14) can be rewritten as

$$\begin{aligned}
& \left(\sum_{y=0}^N (\nabla_x^+ \mathbf{p}_t^R(x, y))^2 \right)^{\frac{1}{2}} \left(\sum_{y=0}^N (\nabla_x^- \mathbf{p}_t^R(x, y))^2 \right)^{\frac{1}{2}} \\
& - \sum_{y=0}^N |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| \\
(5.16) \quad & \geq \frac{\mathbf{1}_{t \leq t_0}}{2} \left(\prod_{\sigma \in \{\pm\}} \left(\sum_{y=0}^N (\nabla_x^\sigma \mathbf{p}_t^R(x, y))^2 \right)^{\frac{1}{2}} \right. \\
& \quad \left. + \sum_{y=0}^N |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| \right)^{-1} \\
& \geq \mathbf{1}_{t \leq t_0} \prod_{\sigma \in \{\pm\}} \left(\sum_{y=0}^N (\nabla_x^\sigma \mathbf{p}_t^R(x, y))^2 \right)^{-\frac{1}{2}}.
\end{aligned}$$

²We bound the sum over y, \bar{y} from below by only one term in the sum because the summand concentrates around the region where t is small and y, \bar{y} are close to x (but $y \neq \bar{y}$).

Now by (4.34) one has $|\nabla_x^\pm \mathbf{p}_t^N(x, y)| \leq C$, and by (4.44) one has

$$\sum_{y=0}^N \nabla_x^\pm \mathbf{p}_t^R(x, y) \leq C t^{-\frac{1}{2}};$$

thus for $t_0/2 \leq t \leq t_0$ the last line of (5.16) is bounded from below by a constant \tilde{c} (which only depends on t_0); in other words, it is bounded from below by $\mathbf{1}_{t_0/2 \leq t \leq t_0} \tilde{c}$ for all t .

Now we integrate over t from 0 to ∞ on both sides of (5.16); note that the time integral of the first term in the first line of (5.16) can be bounded from above by (5.13) using the Cauchy-Schwarz inequality, which is then bounded from above by 1 as mentioned above. Therefore we conclude that

$$1 - \int_0^\infty \sum_y |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| dt \geq \int_{t_0/2}^{t_0} \tilde{c} dt = \tilde{c} t_0/2,$$

which is the desired uniform bound (with $c_\star = 1 - \tilde{c} t_0/2$).

It thus remains to prove the claim (5.15). For this, we can first prove the similar estimates for the standard heat kernel on the entire line

$$\nabla^\pm p_t(0) \leq -\frac{19}{20} \quad \text{and} \quad |\nabla^+ p_t(1)| \vee |\nabla^- p_t(-1)| \leq \frac{1}{20}$$

for sufficiently small time t , which is easy to show by using the relation between p and the discrete time standard heat kernel (a relation as (4.1)) and the explicit formula for the latter kernel. We then apply Lemma 4.7 and exponential spatial decay of $\nabla^\pm p_t$ to obtain (5.15).

To prove (5.12), note that

$$\begin{aligned} \sum_{y=1}^N \int_0^{\frac{s}{2}} |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| (s-t)^{-\frac{1}{2}} dt &\leq \\ C \sum_{y=1}^N \int_0^{\frac{s}{2}} |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| s^{-\frac{1}{2}} dt &\leq \varepsilon C, \end{aligned}$$

where in the last step we used $s^{-1/2} = \varepsilon S^{-1/2}$ and the bound (5.11). On the other hand, by (4.34) with $n = 1$, $v = 1$, and (4.44) with $a = 0$,

$$\begin{aligned} \sum_{y=1}^N \int_{\frac{s}{2}}^s |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| (s-t)^{-1/2} dt &\leq \\ \leq C \int_{\frac{s}{2}}^s t^{-1} t^{-1/2} (s-t)^{-1/2} dt &\leq \\ \leq C (-2(s-t)^{1/2} s^{-1} t^{-1/2} \Big|_{t=\frac{s}{2}}^s) \leq C s^{-1} = C \varepsilon^2 S \leq C \varepsilon^2, \end{aligned}$$

so (5.12) is proved.

Regarding the kernel half-line case, note that both \mathbf{p}^R and $e^{a\varepsilon|x-y|}$ depend on ε . Denote by $L(\varepsilon, \bar{\varepsilon})$ the LHS of (5.11) with $e^{a\varepsilon|x-y|}$ replaced by $e^{a\bar{\varepsilon}|x-y|}$. We show that there exists $\varepsilon_0 > 0$ such that for all $\varepsilon, \bar{\varepsilon} < \varepsilon_0$, one has $L(\varepsilon, \bar{\varepsilon}) \leq c_\star < 1$; thus, in particular, $L(\varepsilon, \varepsilon) \leq c_\star < 1$.

By (4.34) one has $|\nabla_x^\pm \mathbf{p}_t^R(x, y)| \leq C(1 \wedge t^{-1})$, and as in the proof of Corollary 4.12 (namely, following the arguments starting from (4.42) which lead to the bound (4.40)) we have $\sum_{y \geq 0} \mathbf{p}_t^R(x, y) e^{a\varepsilon|x-y|} \leq C$. So

$$\int_0^\infty \sum_{y \geq 0} |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| e^{a\varepsilon|x-y|} dt \leq C \int_0^\infty (1 \wedge t^{-1}) t^{-1/2} dt \leq C.$$

With this integrability estimate at hand and noting that $e^{a\bar{\varepsilon}|x-y|}$ monotonically decreases to 1 as $\bar{\varepsilon} \rightarrow 0$, we can apply the dominated convergence theorem,

$$\lim_{\bar{\varepsilon} \rightarrow 0} L(\varepsilon, \bar{\varepsilon}) = \int_0^\infty \sum_{y \geq 0} |\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y)| dt.$$

The RHS can be bounded by a constant c_\star strictly smaller than 1, uniformly in $\varepsilon < \varepsilon_0$ and $x \geq 1$, following the same arguments as in the finite interval case. This implies that by slightly increasing c_\star and slightly decreasing ε_0 , one has $L(\varepsilon, \bar{\varepsilon}) \leq c_\star$ for all $\varepsilon, \bar{\varepsilon} < \varepsilon_0$, which is the desired bound.

Relation (5.12) for the half-line case then follows similarly as the finite interval case by cutting the integral into two parts $t < s/2$ and $t \geq s/2$ and applying (5.11). \square

5.2 Key Estimate and Identifying the Limit

In order to identify the limit of \mathcal{Z}^ε , we will use an equivalent formulation of the mild solution called the martingale solution.

DEFINITION 5.5. Let I be the interval $[0, 1]$ or $[0, \infty)$. Let

$$(5.17) \quad \begin{aligned} (\varphi, \psi) &\stackrel{\text{def}}{=} \int_I \varphi(X) \psi(X) dX, \\ C_A^\infty &\stackrel{\text{def}}{=} \{\phi \in C_c^\infty(\mathbb{R}) \mid \phi'(0) = A\phi(0)\}, \\ C_{A,B}^\infty &\stackrel{\text{def}}{=} \{\phi \in C_c^\infty(\mathbb{R}) \mid \phi'(0) = A\phi(0), \phi'(1) = -B\phi(1)\}, \end{aligned}$$

where $C_c^\infty(\mathbb{R})$ is the space of compacted supported smooth functions on \mathbb{R} . We say that a probability measure \mathcal{Q} on $C(\mathbb{R}_+, C(I))$ solves the martingale problem for SHE (2.6) with initial condition \mathcal{Z}^{ic} if it satisfies the following requirements. Letting \mathcal{Z} be the canonical coordinate in $C(\mathbb{R}_+, C(I))$ we have $\mathcal{Z}(0, \cdot) = \mathcal{Z}^{ic}$

in distribution, and for all $\varphi \in C_A^\infty$ if $I = \mathbb{R}_+$ or all $\varphi \in C_{A,B}^\infty$ if $I = [0, 1]$, the processes

$$(5.18) \quad N_T(\varphi) \stackrel{\text{def}}{=} (\mathcal{Z}_T, \varphi) - (\mathcal{Z}_0, \varphi) - \frac{1}{2} \int_0^T (\mathcal{Z}_S, \varphi'') dS,$$

$$(5.19) \quad Q_T(\varphi) \stackrel{\text{def}}{=} N_T(\varphi)^2 - \int_0^T (\mathcal{Z}_S^2, \varphi^2) dS,$$

are Q -local martingales. If $I = [0, \infty)$, we further require that for all $\bar{T} > 0$, there exists $a \geq 0$ such that

$$(5.20) \quad \sup_{T \in [0, \bar{T}]} \sup_{X \in \mathbb{R}_+} e^{-aX} \mathbf{E}(\mathcal{Z}_T(X)^2) < \infty.$$

Notation: Throughout the rest of this section, we denote $\Lambda = \{0, 1, \dots, N\}$ for ASEP-B and $\Lambda = \mathbb{Z}_{\geq 0}$ for ASEP-H, and $\Lambda_0 = \{1, 2, \dots, N-1\}$ for ASEP-B and $\Lambda_0 = \mathbb{Z}_{>0}$ for ASEP-H.

PROPOSITION 5.6. *For ASEP-H, consider any initial conditions Z_0^ε satisfying Assumption 2.14 such that $Z_0^\varepsilon \Rightarrow \mathcal{Z}^{ic}$ as $\varepsilon \rightarrow 0$, where $\mathcal{Z}^{ic} \in C(\mathbb{R}_+)$. Then any limit point of Z^ε solves the martingale problem on \mathbb{R}_+ in Definition 5.5 with initial data \mathcal{Z}^{ic} satisfying the Robin boundary condition with parameter A .*

For ASEP-B, consider any initial conditions Z_0^ε satisfying Assumption 2.15 such that $Z_0^\varepsilon \Rightarrow \mathcal{Z}^{ic}$ as $\varepsilon \rightarrow 0$, where $\mathcal{Z}^{ic} \in C([0, 1])$. Then any limit point of Z^ε solves the martingale problem on $[0, 1]$ in Definition 5.5 with initial data \mathcal{Z}^{ic} satisfying the Robin boundary condition with parameter (A, B) .

PROOF. By the uniform bound (4.45) in the case of ASEP-H, any limit point of the family Z^ε satisfies (5.20). Since $Z_0^\varepsilon \Rightarrow \mathcal{Z}^{ic}$, the initial condition of the martingale problem is also satisfied for any limit point. So it only remains to show that any limit point satisfies the conditions (5.18) and (5.19).

Define for all $t \in [0, \varepsilon^{-2}\bar{T}]$

$$(Z_t, \varphi)_\varepsilon \stackrel{\text{def}}{=} \varepsilon \sum_{x=0}^N \varphi(\varepsilon x) Z_t(x) \quad \varphi \in C_{A,B}^\infty$$

or

$$(Z_t, \varphi)_\varepsilon \stackrel{\text{def}}{=} \varepsilon \sum_{x=0}^\infty \varphi(\varepsilon x) Z_t(x) \quad \varphi \in C_A^\infty,$$

for the finite interval and the half-line case, respectively.

Consider the microscopic analogues of (5.18) and (5.19) as (Δ below is the discrete Laplacian with Robin boundary conditions)

$$(5.21) \quad N_T^\varepsilon(\varphi) \stackrel{\text{def}}{=} (Z_{\varepsilon^{-2}T}, \varphi)_\varepsilon - (Z_0, \varphi)_\varepsilon - \frac{1}{2} \int_0^{\varepsilon^{-2}T} (\Delta Z_s, \varphi)_\varepsilon ds,$$

$$Q_T^\varepsilon(\varphi) \stackrel{\text{def}}{=} N_T^\varepsilon(\varphi)^2 - \langle N_T^\varepsilon(\varphi) \rangle.$$

Indeed, by Lemma 3.1, $N_T^\varepsilon(\varphi)$ and hence $Q_T^\varepsilon(\varphi)$ are martingales. Now we would like to show that $N_T^\varepsilon(\varphi)$ can be rewritten as

$$(5.22) \quad \varepsilon \sum_{X \in \varepsilon\Lambda} \mathcal{Z}_T^\varepsilon(X)\varphi(X) - \varepsilon \sum_{X \in \varepsilon\Lambda} \mathcal{Z}_0^\varepsilon(X)\varphi(X) \\ - \frac{1}{2} \int_0^T \varepsilon \sum_{X \in \varepsilon\Lambda} \mathcal{Z}_S^\varepsilon(X)\varphi''(X)dS + \text{Err}^{(1)}$$

and that $Q_T^\varepsilon(\varphi)$ can be rewritten as

$$(5.23) \quad N_T^\varepsilon(\varphi)^2 - \int_0^T \varepsilon \sum_{X \in \varepsilon\Lambda} \mathcal{Z}_S^\varepsilon(X)^2 \varphi(X)^2 dS + \text{Err}^{(2)},$$

where the error terms $\text{Err}^{(1)}$, $\text{Err}^{(2)}$ vanish in probability as $\varepsilon \rightarrow 0$. Here $\varepsilon\Lambda = \{X \in \mathbb{R} \mid X/\varepsilon \in \Lambda\}$. Assuming this vanishing error, the proof is completed by passing to the limit along a converging subsequence and noting that $\varepsilon\Lambda \rightarrow I$ and the Riemann sums converge to the integrals by continuity of ϕ and \mathcal{Z}^ε and its limit \mathcal{Z} . Thus, we conclude that any limiting point \mathcal{Z} of the sequence \mathcal{Z}^ε satisfies (5.18) and (5.19). It remains to show that the error terms in (5.22) and (5.23) go to 0 as desired.

We start by arguing for (5.22). Consider the last term in the expression of $N_T^\varepsilon(\varphi)$. In the finite interval case, applying the summation-by-parts formula (4.23) with $u(\bullet) = \varphi(\varepsilon\bullet)$ and $v(\bullet) = Z_s(\bullet)$ (recall that $N = 1/\varepsilon$ so that, for instance, $u(N+1) = \varphi(1+\varepsilon)$), we find that

$$\begin{aligned} (\Delta Z_s, \varphi)_\varepsilon &= \varepsilon \sum_{x=0}^N Z_s(x)\Delta\varphi(\varepsilon x) + \varepsilon\nabla^+ Z_s(N)\varphi(1+\varepsilon) + \varepsilon\nabla^- Z_s(0)\varphi(-\varepsilon) \\ &\quad - \varepsilon Z_s(N+1)(\varphi(1+\varepsilon) - \varphi(1)) - \varepsilon Z_s(-1)(\varphi(-\varepsilon) - \varphi(0)) \\ &= \varepsilon \sum_{x=0}^N Z_s(x)\Delta\varphi(\varepsilon x) - \varepsilon^2 B Z_s(N)\varphi(1+\varepsilon) - \varepsilon^2 A Z_s(0)\varphi(-\varepsilon) \\ &\quad - \varepsilon(1 - \varepsilon B)Z_s(N)(\varphi(1+\varepsilon) - \varphi(1)) \\ &\quad - \varepsilon(1 - \varepsilon A)Z_s(0)(\varphi(-\varepsilon) - \varphi(0)) \end{aligned}$$

where $\Delta\varphi(\varepsilon x) = \varphi(\varepsilon(x+1)) + \varphi(\varepsilon(x-1)) - 2\varphi(\varepsilon x)$. Note that in the last equality we used the Robin boundary condition for Z_s , for instance, $\nabla^+ Z_s(N) =$

$(\mu_B - 1)Z_s(N) = -\varepsilon B Z_s(N)$. Therefore

$$\begin{aligned}
 & \int_0^{\varepsilon^{-2}T} (\Delta Z_s, \varphi)_\varepsilon ds \\
 (5.24) \quad & = \int_0^{\varepsilon^{-2}T} \left(\varepsilon \sum_{x=0}^N Z_s(x) \varphi''(\varepsilon x) - BZ_s(N)\varphi(1) - AZ_s(0)\varphi(0) \right. \\
 & \quad \left. - Z_s(N)\phi'(1) + Z_s(0)\phi'(0) + R_0^\varepsilon(\phi) \right) \varepsilon^2 ds
 \end{aligned}$$

where $\phi''(\varepsilon x)$ stands for the second continuous derivative of φ evaluated at εx and the error term

$$\begin{aligned}
 R_0^\varepsilon(\phi) & \stackrel{\text{def}}{=} (Z_s, \varepsilon^{-1} \Delta \phi(\varepsilon \bullet) - \phi''(\varepsilon \bullet))_\varepsilon \\
 & + (BZ_s(N)\varphi(1) - BZ_s(N)\varphi(1 + \varepsilon)) \\
 & + (AZ_s(0)\varphi(0) - AZ_s(0)\varphi(-\varepsilon)) \\
 & + \left(Z_s(N)\phi'(1) - (1 - \varepsilon B)Z_s(N) \frac{\varphi(1 + \varepsilon) - \varphi(1)}{\varepsilon} \right) \\
 & + \left(-Z_s(0)\phi'(0) - (1 - \varepsilon A)Z_s(0) \frac{\varphi(-\varepsilon) - \varphi(0)}{\varepsilon} \right).
 \end{aligned}$$

Since ϕ is smooth, by the uniform bound (4.48) on Z , we have $\mathbf{E}(R_0^\varepsilon(\phi)^2) \rightarrow 0$ as $\varepsilon \rightarrow 0$. Invoking the assumed Robin boundary condition (5.17) on ϕ , the four boundary terms on the RHS of (5.24) add up to 0.

Therefore by the changes of variables $X = \varepsilon x$ and $S = \varepsilon^2 s$ and the definition (2.20) of \mathcal{Z}^ε , we have cast $N_T^\varepsilon(\varphi)$ into the desired form (5.22) where the error $\text{Err}^{(1)} = -\frac{1}{2} \int_0^T R_0^\varepsilon(\phi) dS$ vanishes in probability. This completes the proof that any limiting point \mathcal{Z} of the sequence \mathcal{Z}^ε satisfies (5.18) for the finite interval case. The half-line case can be shown in the same way by applying the summation-by-parts formula (4.24) and invoking both the discrete boundary conditions for Z and the continuous boundary condition for the test function ϕ .

Turning to $Q_T^\varepsilon(\varphi)$ and (5.23), we apply Lemma 3.3 to calculate $\langle N_T^\varepsilon(\varphi) \rangle$ and obtain the following expression for $Q_T^\varepsilon(\varphi)$:

$$(5.25) \quad Q_T^\varepsilon(\varphi) \stackrel{\text{def}}{=} N_T^\varepsilon(\varphi)^2 - \varepsilon^2 \int_0^{\varepsilon^{-2}T} (Z_s^2, \varphi^2)_\varepsilon ds + R_1^\varepsilon(\varphi) + R_2^\varepsilon(\varphi)$$

where

$$R_1^\varepsilon(\varphi) \stackrel{\text{def}}{=} o(\varepsilon^2) \int_0^{\varepsilon^{-2}T} (Z_s^2, \varphi^2)_\varepsilon ds,$$

$$R_2^\varepsilon(\varphi) \stackrel{\text{def}}{=} \varepsilon^2 \int_0^{\varepsilon^{-2}T} \sum_{x \in \Lambda_0} \nabla^- Z_s(x) \nabla^+ Z_s(x) \varphi(x)^2 ds,$$

where in the definition of R_2^ε , x is summed in the “bulk” Λ_0 . Since the second term on the RHS of (5.25) is equal to the second term in (5.23) after passing to the macroscopic variables T, X , it now suffices to prove that $\mathbf{E}(R_i^\varepsilon(\varphi))^2 \rightarrow 0$ for $i = 1, 2$. By the uniform bound (4.45) or (4.48) on Z , we clearly have $\mathbf{E}(R_1^\varepsilon(\varphi)^2) \rightarrow 0$. To control $R_2^\varepsilon(\varphi)^2$, we follow [7] by using the “key estimate” as in Lemma 5.7 in the following. Letting $\mathcal{F}_t \stackrel{\text{def}}{=} \sigma(Z_s(x) : x \in \mathbb{Z}_{\geq 0}, s \leq t)$, we denote the canonical filtration, and for $y \in \Lambda_0$ letting

$$(5.26) \quad U^\varepsilon(y, s, s') \stackrel{\text{def}}{=} \mathbf{E}(\nabla^- Z_s(y) \nabla^+ Z_{s'}(y) \mid \mathcal{F}_{s'}),$$

we have that

$$\mathbf{E}(R_2^\varepsilon(\varphi)^2) = \varepsilon^4 \int_0^{\varepsilon^{-2}T} ds \int_0^s ds' \sum_{x, y \in \Lambda_0} \varphi(\varepsilon x)^2 \varphi(\varepsilon y)^2 \cdot \mathbf{E}(\nabla^- Z_{s'}(x) \nabla^+ Z_{s'}(x) U^\varepsilon(y, s, s')).$$

Here x, y are both summed in the bulk Λ_0 as above.

With $|\nabla^\pm Z_t(x)| \leq C \varepsilon^{1/2} Z_t(x)$, we further obtain

$$(5.27) \quad \mathbf{E}(R_2^\varepsilon(\varphi)^2) \leq C \varepsilon^5 \int_0^{\varepsilon^{-2}T} ds \int_0^s ds' \sum_{x, y \in \Lambda_0} \varphi(\varepsilon x)^2 \varphi(\varepsilon y)^2 \cdot \mathbf{E}(Z_{s'}(x)^2 U^\varepsilon(y, s, s')).$$

Note if we were to simply use $|\nabla^\pm Z_t(x)| \leq C \varepsilon^{1/2} Z_t(x)$ to “brutally” bound $|U^\varepsilon(y, s, s')|$ by $\varepsilon C Z_s^2(y)$, the resulting bound on $\mathbf{E}(R_2^\varepsilon(\varphi)^2)$ would be of order $O(1)$ (since the change of time and space variables to macroscopic variables gives ε^{-6}), which would be insufficient for our purpose. To show $\mathbf{E}(R_2^\varepsilon(\varphi)^2) \rightarrow 0$, we divide the time integrals in (5.27) into two parts. The first part consists of the region with $s' < \varepsilon^{-3/2}$, where the above brutal estimate gives a bound of $O(\varepsilon^{1/2})$. For the second part, which is the rest of the integral, we utilize the key Lemma 5.7 below, which provides the important decay factor $(t-s)^{-1/2}$ that is typically $O(\varepsilon)$, thus together with the factor $\varepsilon^{1/2-\delta}$ improving the above brutal bound. Indeed, as

in the proof of proposition 4.11 in [7], in the case of ASEP-H, conditioning on $\{Z_{s'}(x) \leq \bar{K}\}$, (5.27) can be bounded by

$$C \bar{K}^2 \varepsilon^3 e^{a\varepsilon(x+y)} \int_0^{\varepsilon^{-2}T} \int_0^s \varepsilon^{\frac{1}{2}-\delta} (s-s')^{-\frac{1}{2}} ds' ds \leq C e^{a\varepsilon(x+y)} \varepsilon^{\frac{1}{2}-\delta} \bar{K}^2,$$

while by conditioning on $\{Z_{s'}(x) > \bar{K}\}$ (5.27) can be bounded using the Chebyshev inequality by $C e^{a\varepsilon(x+y)} \bar{K}^{-2}$. Letting $\varepsilon \rightarrow 0$ and then $\bar{K} \rightarrow \infty$, we obtain $\mathbf{E}(R_2^\varepsilon(\varphi)^2) \rightarrow 0$. For ASEP-B the argument is the same with $a = 0$.

Therefore we have shown that $Q_T^\varepsilon(\varphi)$ as in (5.25) can be indeed cast into the form (5.23) with error term $\text{Err}^{(2)} = R_1^\varepsilon(\varphi) + R_2^\varepsilon(\varphi)$, which vanishes in probability. By the arguments below (5.23), we conclude that any limit \mathcal{Z}_ε satisfies the conditions in Definition 5.5 and thus solves the martingale problem. \square

LEMMA 5.7. *Assume the above setting, and consider both the ASEP-H and the ASEP-B models. For all $\bar{T} > 0$, $\delta > 0$, there are constants $a, C > 0$ such that*

$$(5.28) \quad \sup_{x \in \Lambda_0} e^{-ax} \mathbf{E}|U^\varepsilon(x, t, s)| \leq C \varepsilon^{\frac{1}{2}-\delta} (t-s)^{-\frac{1}{2}}$$

for all $\varepsilon^{-3/2} \leq s < t \leq \varepsilon^{-2}\bar{T}$ and all $\varepsilon > 0$. Here, for ASEP-B, the constant a may be taken to be 0.

PROOF OF LEMMA 5.7. The proof follows a similar argument as that for [7, lemma 4.8]. Throughout the proof we assume $a = 0$ in the case of ASEP-B. Let $Z_t(x) = I_t(x) + N_t^t(x)$ where

$$I_t(x) = \sum_{y \in \Lambda} \mathbf{p}_t^R(x, y) Z_0(y), \quad N_s^t(x) \stackrel{\text{def}}{=} \int_0^s \sum_{y \in \Lambda} \mathbf{p}_{t-\tau}^R(x, y) dM_\tau(y).$$

Note that $N_s^t(x)$ is a martingale in s . For $s \leq r \leq t$, one has

$$(5.29) \quad \begin{aligned} & \mathbf{E}(\nabla^- N_r^t(x) \nabla^+ N_r^t(x) \mid \mathcal{F}_s) \\ &= \nabla^- N_s^t(x) \nabla^+ N_s^t(x) \\ &+ \mathbf{E}\left(\int_s^r \sum_{y \in \Lambda} K_{t-\tau}(x, y) d\langle M(y), M(y) \rangle_\tau \mid \mathcal{F}_s\right) \end{aligned}$$

where

$$(5.30) \quad K_t(x, y) \stackrel{\text{def}}{=} \nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^- \mathbf{p}_t^R(x, y).$$

With $U^\varepsilon(x, t, s)$ defined as in (5.26) and with $\mathbf{E}(N_r^t(x)|\mathcal{F}_s) = N_s^t(x)$, one has by (5.29)

$$(5.31) \quad \begin{aligned} U^\varepsilon(x, t, s) &= \nabla^- I_t(x) \nabla^+ I_t(x) + \nabla^- I_t(x) \nabla^+ N_s^t(x) \\ &\quad + \nabla^- N_s^t(x) \nabla^+ I_t(x) + \nabla^- N_s^t(x) \nabla^+ N_s^t(x) \\ &\quad + \mathbf{E} \left(\int_s^t \sum_{y \in \Lambda} K_{t-\tau}(x, y) d \langle M(y) \rangle_\tau \middle| \mathcal{F}_s \right). \end{aligned}$$

We bound the L^1 norms (i.e., $\mathbf{E}|\cdot|$) of the terms on the RHS. For the first four terms, by the Cauchy-Schwarz inequality one only needs to show

$$(5.32) \quad \mathbf{E}(\nabla^\pm I_t(x))^2, \mathbf{E}(\nabla^\pm N_s^t(x))^2 \leq C \varepsilon^{\frac{1}{2}} (t-s)^{-\frac{1}{2}} e^{2a\varepsilon x}.$$

Estimates for $\mathbf{E}(\nabla^\pm I_t(x))^2$. We use (2.16) to obtain

$$(5.33) \quad \begin{aligned} \mathbf{E}((\nabla^\pm I_t(x))^2) &= \sum_{y, y'} \nabla^\pm \mathbf{p}_t^R(x, y) \nabla^\pm \mathbf{p}_t^R(x, y') \mathbf{E}(Z_0(y) Z_0(y')) \\ &\leq C \left(\sum_{y \in \Lambda} \nabla^\pm \mathbf{p}_t^R(x, y) e^{a\varepsilon y} \right)^2. \end{aligned}$$

By Corollaries 4.12 or 4.14, the above quantity can be bounded by $C e^{2a\varepsilon x} t^{-1}$. Further expressing t^{-1} as $t^{-1/2} t^{-1/2}$, and applying $t^{-1/2} < (t-s)^{-1/2}$ and $t^{-1/2} \leq \varepsilon^{3/4}$ (since we assume $t \geq \varepsilon^{-3/2}$), we thereby obtain the desired bound on $\mathbf{E}(\nabla^\pm I_t(x))^2$ as in (5.32).

Estimates for $\mathbf{E}(\nabla^\pm N_s^t(x))^2$. One has

$$\begin{aligned} \mathbf{E}((\nabla^\pm N_s^t(x))^2) &= \mathbf{E} \int_0^s \sum_{y \in \Lambda} (\nabla^\pm \mathbf{p}_{t-\tau}^R)^2(x, y) d \langle M \rangle_\tau(y) \\ &\leq C \int_0^s \left(\sup_{y \in \Lambda} |\nabla^+ \mathbf{p}_{t-\tau}^R(x, y)| \right. \\ &\quad \cdot \left. \left(\sum_{y \in \Lambda} |\nabla^- \mathbf{p}_{t-\tau}^R(x, y)| \mathbf{E} \left| \frac{d}{d\tau} \langle M \rangle_\tau(y) \right| \right) d\tau. \end{aligned}$$

By combining Lemma 3.3, the bound $\nabla^\pm Z_t(y) \leq C \varepsilon^{1/2} Z_t(y)$, and the uniform bounds (4.45) or (4.48), one has $\mathbf{E} \left| \frac{d}{d\tau} \langle M(y) \rangle_\tau \right| \leq C \varepsilon e^{a\varepsilon y}$; we then apply the same reasoning as used above to bound (5.33), together with (4.34) for $v = 1$, to obtain

$$\mathbf{E}((\nabla^\pm N_s^t(x))^2) \leq C \varepsilon e^{2ax} \int_0^s (t-\tau)^{-\frac{3}{2}} ds.$$

Upon integrating over τ , we obtain the desired bound on $\mathbf{E}(\nabla^\pm N_s^t(x))^2$ as in (5.32).

Estimates for the last term on the RHS of (5.31). We use the explicit expression of the predictable quadratic variation given in Lemma 3.3 to rewrite the last term on the RHS of (5.31) as $\tilde{I}_1 + \tilde{I}_2$ where

$$\begin{aligned}\tilde{I}_1(s, t, x) &\stackrel{\text{def}}{=} (\varepsilon + o(\varepsilon)) \sum_{y \in \Lambda} \int_s^t K_{t-\tau}(x, y) \mathbf{E}(Z_\tau(y)^2 \mid \mathcal{F}_s) d\tau, \\ \tilde{I}_2(s, t, x) &\stackrel{\text{def}}{=} - \sum_{y \in \Lambda_0} \int_s^t K_{t-\tau}(x, y) \mathbf{E}(\nabla^- Z_\tau(y) \nabla^+ Z_\tau(y) \mid \mathcal{F}_s) d\tau.\end{aligned}$$

By Lemma 5.8 below, we can bound $\mathbf{E}|\tilde{I}_1(s, t, x)|$ by $C\varepsilon^{1/2-\delta}e^{a\varepsilon x}(t-s)^{-1/2}$. So we have obtained the desired bound on all the terms coming from the RHS of (5.31) except for the term \tilde{I}_2 , but \tilde{I}_2 contains the same conditional expectation on the LHS of (5.31), which means that one can bound this conditional expectation in terms of an expression involving this conditional expectation itself. Indeed, collecting the bounds for the terms in (5.31), one has

$$\begin{aligned}(5.34) \quad & e^{-a\varepsilon x} \mathbf{E}|U^\varepsilon(x, t, s)| \\ & \leq C\varepsilon^{\frac{1}{2}-\delta}(t-s)^{-\frac{1}{2}} \\ & \quad + e^{-a\varepsilon x} \sum_{y \in \Lambda_0} \int_s^t |K_{t-\tau}(x, y)| \cdot \mathbf{E}|U^\varepsilon(y, \tau, s)| d\tau \\ & \leq C\varepsilon^{\frac{1}{2}-\delta}(t-s)^{-\frac{1}{2}} \\ & \quad + \sum_{y \in \Lambda_0} \int_s^t |K_{t-\tau}(x, y)| e^{a\varepsilon|x-y|} \cdot e^{-a\varepsilon y} \mathbf{E}|U^\varepsilon(y, \tau, s)| d\tau.\end{aligned}$$

We find that $e^{-a\varepsilon x} \mathbf{E}|U^\varepsilon(x, t, s)|$ is bounded by

$$\begin{aligned}& C\varepsilon^{\frac{1}{2}-\delta}(t-s)^{-\frac{1}{2}} \\ & \quad + \sum_{y_1 \in \Lambda_0} \int_s^t |K_{t-\tau_1}(x, y_1)| e^{a\varepsilon|x-y_1|} \left(C\varepsilon^{\frac{1}{2}-\delta}(\tau_1-s)^{-\frac{1}{2}} \right. \\ & \quad \left. + \sum_{y_2 \in \Lambda_0} \int_s^{\tau_1} |K_{\tau_1-\tau_2}(y_1, y_2)| e^{a\varepsilon|y_1-y_2|} \cdot e^{-a\varepsilon y_2} \mathbf{E}|U^\varepsilon(y_2, \tau_2, s)| d\tau_2 \right) d\tau_1.\end{aligned}$$

by iterating the bound (5.34) as in [7, lemma 4.8]. Expanding the parentheses we then obtain three terms, with the second term (a single summation-integration over space-time) given by A_1 defined below, and the third term (a double summation-integration over space-time) again involving the expectation of U^ε . Repeating this

iteration, we eventually get

$$e^{-a\epsilon x} \mathbf{E}|U^\epsilon(x, t, s)| \leq C \epsilon^{\frac{1}{2}-\delta} (t-s)^{-\frac{1}{2}} + \sum_{n \geq 1} A_n(x, t, s)$$

where, with $y_0 := x$, $\tau_0 := t$,

$$A_n(x, t, s) = \int \sum_{y_1, 2, \dots, y_n \in \Lambda_0} C \epsilon^{\frac{1}{2}-\delta} (\tau_n - s)^{-\frac{1}{2}} \cdot \prod_{i=1}^n |K_{\tau_{i-1}-\tau_i}(y_{i-1}, y_i)| e^{a\epsilon|y_{i-1}-y_i|} \prod_{i=1}^n d\tau_i,$$

where the time integrals are over $s \leq \tau_n \leq \tau_{n-1} \leq \dots \leq \tau_1 \leq t$. To estimate A_n for each n , we first integrate over $\tau_n \in [s, \tau_{n-1}]$ and sum over $y_n \in \Lambda_0$ in the following way: by a change of variable $\tau_n = \tau_{n-1} - r$ we instead integrate over $r \in [0, \tau_{n-1} - s]$, and we are in the scope of applying (5.12) and obtaining a factor $C\epsilon$. We then integrate and sum over the other space-time variables using (5.11), which yields a bound by c_\star^{n-1} where c_\star was introduced in (5.11). Together with the factor $C\epsilon^{1/2-\delta}$ on the right-hand side of the above definition of A_n , we obtain that $|A_n(x, t, s)| \leq C\epsilon^{3/2-\delta} c_\star^{n-1}$. Since $c_\star < 1$ one obtains a convergent series and the sum over n yields a bound by $C\epsilon^{3/2-\delta}$ that is smaller than $C\epsilon^{1/2-\delta} (t-s)^{-1/2}$, where the last constant C depends on \bar{T} . Therefore the bound (5.28) follows. \square

LEMMA 5.8. *Let \mathcal{F}_s be the filtration defined above, and $K_t(x, y)$ be the deterministic function defined in (5.30). There exist $a, C > 0$ such that for any $\delta > 0$, $0 \leq s \leq t \leq \epsilon^{-2}\bar{T}$, $x \in \Lambda_0$, and $\epsilon > 0$ sufficiently small, we have*

$$(5.35) \quad \mathbf{E} \left[\left| \epsilon \sum_{y \in \Lambda} \int_s^t K_{t-\tau}(x, y) \mathbf{E}(Z_\tau(y)^2 | \mathcal{F}_s) d\tau \right| \right] \leq C \epsilon^{\frac{1}{2}-\delta} e^{a\epsilon x} (t-s)^{-\frac{1}{2}}.$$

For ASEP-B, we can take $a = 0$.

PROOF. To start, we note that by definition

$$K_t(x, y) = -\nabla_x^+ \mathbf{p}_t^R(x, y) \nabla_x^+ \mathbf{p}_t^R(x-1, y).$$

Using Proposition 5.1 (with the variable \bar{x} therein taken as $\bar{x} = x-1$) we have $\sum_y \int_0^\infty K_\tau(x, y) d\tau = O(\epsilon)$. The quantity in the L^1 norm $\mathbf{E}|\cdot|$ on LHS of (5.35) can be then rewritten as

$$(5.36) \quad \begin{aligned} & \epsilon \sum_{y \in \Lambda} \int_s^t K_{t-\tau}(x, y) \mathbf{E}(Z_\tau(y)^2 - Z_t(x)^2 | \mathcal{F}_s) d\tau \\ & + \epsilon \mathbf{E}(Z_t(x)^2 | \mathcal{F}_s) \sum_{y \in \Lambda} \int_{t-s}^\infty K_\tau(x, y) d\tau + O(\epsilon^2) \mathbf{E}(Z_t(x)^2 | \mathcal{F}_s) \\ & =: J_1 + J_2 + J_3 \end{aligned}$$

where we have labeled the three terms as J_1, J_2, J_3 . Here we have simply subtracted and added a term $\sum_y \int_s^t K_{t-\tau}(x, y) \mathbf{E}(Z_t(x)^2 | \mathcal{F}_s) d\tau$ and made a change of variable $t-\tau \rightarrow \tau$ and applied Proposition 5.1; the expression $O(\varepsilon^2)$ stands for ε times the constant c in (5.1) in the case of ASEP-B, and is actually 0 in the case of ASEP-H. We estimate the L^1 norm of these three terms separately as follows. Throughout this proof we set $a = 0$ for ASEP-B.

Estimates of J_3 . Note that $(t-s)^{-1/2} \geq \varepsilon \bar{T}^{-1/2} > \varepsilon^{3/2}$ for $\varepsilon > 0$ sufficiently small. Using this and the uniform bound (4.45) or (4.48) on Z , which states that $\|Z_t(x)\|_2 \leq C e^{(1/2)a\varepsilon x}$ for a constant a , we concluded that $\mathbf{E}|J_3|$ is bounded by the RHS of (5.35).

Estimates of J_2 . Recall that K is a product of two factors $\nabla^+ \mathbf{p}^R$ and $\nabla^- \mathbf{p}^R$. Applying (4.34) with $n = v = 1$ to one of the factors $\nabla^\pm \mathbf{p}^R$ in K , and (4.41) or (4.44) to the other factor $\nabla^\mp \mathbf{p}^R$, we can bound $\sum_y |K_\tau(x, y)|$ by $C(1 \wedge \tau^{-3/2})e^{a\varepsilon x}$. Using this and the uniform bound (4.45) or (4.48), we obtain the desired bound on J_3 as

$$(5.37) \quad \mathbf{E}|J_3(s, t, x)| \leq C \varepsilon e^{a\varepsilon x} \int_{t-s}^{\infty} \tau^{-\frac{3}{2}} d\tau \leq C \varepsilon e^{a\varepsilon x} (t-s)^{-\frac{1}{2}}.$$

Estimates of J_1 . The idea of controlling J_1 is to use the fact that $K_{t-\tau}(x, y)$ concentrates on values of (τ, y) that are close to (t, x) , and that, thanks to the Hölder estimates (4.46)–(4.47), or (4.49)–(4.50), $|Z_\tau(y)^2 - Z_t(x)^2|$ is small when $(\tau, y) \approx (t, x)$. More precisely, with

$$|Z_\tau(y)^2 - Z_t(x)^2| \leq (Z_\tau(y) + Z_t(x))(|Z_\tau(y) - Z_t(y)| + |Z_t(y) - Z_t(x)|),$$

we use the Cauchy-Schwarz inequality and the Hölder estimates (4.46)–(4.47) or (4.49)–(4.50) for $\alpha = \frac{1}{2} - \delta$ to obtain

$$\mathbf{E}|Z_\tau(y)^2 - Z_t(x)^2| \leq C \varepsilon^{\frac{1}{2}-\delta} e^{a\varepsilon(x+y)} (|y-x|^{\frac{1}{2}-\delta} + (|t-\tau| \vee 1)^{\frac{1}{4}-\frac{\delta}{2}}).$$

Therefore, invoking (4.34) with $n = v = 1$

$$\begin{aligned} & \mathbf{E}|J_1(s, t, x)| \\ & \leq C \varepsilon^{\frac{3}{2}-\delta} e^{a\varepsilon x} \int_0^{\varepsilon^{-2}\bar{T}} ((t-\tau)^{-1} \wedge 1) \\ & \quad \times \left(\sum_y |\nabla^- \mathbf{p}_{t-\tau}(x, y)| e^{a\varepsilon y} (|y-x|^{\frac{1}{2}-\delta} + (|t-\tau| \vee 1)^{\frac{1}{4}-\delta}) \right) d\tau. \end{aligned}$$

Arguing as in (4.64), together with (4.41) or (4.44), we obtain

$$\begin{aligned} & \mathbf{E}|J_1(s, t, x)| \leq \\ & C \varepsilon^{\frac{3}{2}-\delta} e^{a\varepsilon x} \int_0^{\varepsilon^{-2}\bar{T}} (1 \wedge (t-\tau)^{-1}) ((t-\tau)^{\frac{1}{4}-\delta} \vee 1) (t-\tau)^{-\frac{1}{2}} d\tau. \end{aligned}$$

Here the last factor $(t-\tau)^{-1/2}$ arises from the RHS of (4.41) or (4.44). The integral is bounded by a constant because as $\tau \rightarrow t$ the integrand behaves as $(t-\tau)^{-1/2}$, and as $\tau \rightarrow \infty$ the integrand behaves as $\tau^{-5/4}$. So $\mathbf{E}|J_1(s, t, x)|$ is bounded by $C\varepsilon^{3/2-\delta}e^{a\varepsilon x}$. With $(t-s)^{-1/2} \geq t^{-1/2} \geq \bar{T}^{-1/2}\varepsilon$, the desired bound

$$\mathbf{E}|J_1(s, t, x)| \leq C\varepsilon^{\frac{1}{2}-\delta}e^{a\varepsilon x}(t-s)^{-\frac{1}{2}}$$

follows. This concludes the proof of the lemma. \square

We show the uniqueness for the martingale problem and its equivalence with the mild solution to the SHE.

PROPOSITION 5.9. *Let I be the interval $[0, 1]$ or $[0, \infty)$, and $\mathcal{Z}^{ic} \in C(I)$ be a random function. If $I = [0, \infty)$ we assume that for each $p \geq 1$ there exists a $a > 0$ such that*

$$\sup_{X \in \mathbb{R}_+} e^{-aX} \mathbf{E}(\mathcal{Z}^{ic}(X))^p < \infty.$$

Then the martingale problem (Definition 5.5) has a unique solution whose law coincides with that of the mild solution to SHE with initial condition \mathcal{Z}^{ic} .

PROOF. We only need to show the uniqueness, since the existence follows immediately from the convergence along subsequences of \mathcal{Z}^ε provided in Proposition 5.6. To prove uniqueness of the solution to the martingale problem, we need to show a martingale representation theorem which, possibly by extending the probability space and the filtration, represents the local martingale N in (5.18) as a stochastic integral of \mathcal{Z} against a Wiener process, essentially following [7, 56]. By Definition 5.5 there exists a sequence of stopping times $\{\tau_N\}$ such that $\lim_{N \rightarrow \infty} \tau_N = +\infty$ \mathcal{Q} -a.s., $N^{\tau_N}(\phi)$ is a square-integrable martingale, and $\Lambda^{\tau_N}(\phi)$ is a martingale. The associated orthogonal martingale measure (see [70] for a definition) $N(dS dX)$ has quadratic variation measure $\mathcal{Z}_S(X)^2 dS dX$.

We introduce a cylindrical Wiener process \bar{W} on $L^2(I)$ (possibly by extending the probability space and the filtration) that is independent of \mathcal{Z} . Let \mathcal{Q}' be the probability measure on the extended space. Define a process

$$\begin{aligned} \tilde{W}_T^N(\phi) &\stackrel{\text{def}}{=} \int_0^{T \wedge \tau_N} \int_I \frac{1}{\mathcal{Z}_S(X)} \mathbf{1}_{\{\mathcal{Z}_S(X) \neq 0\}} \phi(X) N(dS dX) \\ &\quad + \int_0^{T \wedge \tau_N} (\mathbf{1}_{\{\mathcal{Z}_S(X) = 0\}} \phi, d\bar{W}_S). \end{aligned}$$

It is easy to see by Lévy characterization that \tilde{W}_T^N is simply a cylindrical Wiener process on $L^2(I)$ (which we denote by W_T) stopped at τ_N , and that one has the martingale representation $N_T^{\tau_N}(\phi) = \int_0^{T \wedge \tau_N} (\mathcal{Z}_S \phi, dW_S)$. Using this representation together with (5.18), one has

$$(\mathcal{Z}_T, \phi) - (\mathcal{Z}_0, \phi) = \int_0^{T \wedge \tau_N} (\mathcal{Z}_S \phi, dW_S) + \frac{1}{2} \int_0^{T \wedge \tau_N} (\mathcal{Z}_S, \phi'') dS$$

\mathcal{Q}' -a.s. Sending $N \rightarrow \infty$ we obtain that \mathcal{Z} is the “weak solution” (in the PDE sense) defined in [70, eq. 3.9]. In order to identify \mathcal{Z} as the mild solution, as in [70, eq. 3.10], we can actually show that for all smooth functions $\psi(T, X)$ such that $\psi'(T, X)|_{X=0} = A\psi(T, 0)$, as well as $\psi'(T, X)|_{X=1} = -B\psi(T, 1)$ if $I = [0, 1]$, one has

$$\begin{aligned} (\mathcal{Z}_T, \psi(T)) - (\mathcal{Z}_0, \psi(0)) &= \int_0^T (\mathcal{Z}_S \psi(S), dW_S) \\ &\quad + \frac{1}{2} \int_0^T \left(\mathcal{Z}_S, \psi''(S) + \frac{\partial \psi}{\partial S}(S) \right) dS. \end{aligned}$$

Choose $\psi(S, Y) = \int_I \mathcal{P}_{T-S}^R(Y, U) \phi(U) dU$, which clearly satisfies the above Robin boundary condition since \mathcal{P}^R does, and is also such that $\psi'' + \frac{\partial \psi}{\partial S} = 0$. Sending $\phi(\cdot) \rightarrow \delta_X(\cdot)$ and using the symmetric property $\mathcal{P}_S^R(X, Y) = \mathcal{P}_S^R(Y, X)$, we see that $\psi(S, \cdot) \rightarrow \mathcal{P}_{T-S}^R(X, \cdot)$, that $\psi(T, \cdot) \rightarrow \delta_X(\cdot)$, and that $\psi(0, \cdot) \rightarrow \mathcal{P}_T^R(X, \cdot)$. Therefore \mathcal{Z} is the mild solution defined in Definition 2.5 and in particular the martingale solution is unique. \square

PROOF OF THEOREMS 2.17 AND 2.18. The rescaled processes are tight according to Proposition 4.17, and by Proposition 5.6 any limiting point solves the martingale problem with the desired boundary conditions and initial condition. According to Proposition 5.9 the law of the martingale solution coincides with that of the mild solution. Therefore Theorems 2.17 and 2.18 follow immediately. \square

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