Microscopic concavity and fluctuation bounds in a class of deposition processes

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Abstract. We prove fluctuation bounds for the particle current in totally asymmetric zero range processes in one dimension with nondecreasing, concave jump rates whose slope decays exponentially. Fluctuations in the characteristic directions have order of magnitude $t^{1/3}$. This is in agreement with the expectation that these systems lie in the same KPZ universality class as the asymmetric simple exclusion process. The result is via a robust argument formulated for a broad class of deposition-type processes. Besides this class of zero range processes, hypotheses of this argument have also been verified in the authors’ earlier papers for the asymmetric simple exclusion and the constant rate zero range processes, and are currently under development for a bricklayers process with exponentially increasing jump rates.

Résumé. Nous démontrons des bornes sur les fluctuations du courant de particules pour des processus de zero-range unidimensionnels totalement asymétriques avec des taux de sauts concaves dont la pente décroit exponentiellement. Les fluctuations dans la direction des caractéristiques sont de l’ordre $t^{1/3}$ en accord avec les prédictions de la classe d’universalité de KPZ. Notre résultat est obtenu par un raisonnement robuste qui est formulé pour une classe importante de processus de déposition. Au-delà du processus de zero-range, les hypothèses de notre argument ont aussi été vérifiées dans des articles antérieurs pour le processus d’exclusion simple asymétrique et le processus de zero-range avec taux constants. Ces hypothèses sont en cours de développement pour un processus de déposition avec des taux de sauts dont la croissance est exponentielle.

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1. Introduction

This paper studies anomalous current fluctuations of attractive interacting systems in one dimension with one conserved quantity. The family of models considered includes the asymmetric exclusion, the zero range, misanthrope-type and many other processes. In the asymmetric case (to be specified later) the Eulerian scaling of such a system leads to a (deterministic) hyperbolic conservation law with a hydrodynamic flux function $\mathcal{H}(\varrho)$. The characteristics of the
was treated in [13] and the totally asymmetric zero range process (TAZRP) with constant jump rate in [7]. The ASEP work [13] was the first to prove microscopic concavity. We rewrite our earlier proof for ASEP and constant rate TAZRP in a fairly general way, extract and formulate in a theorem. The development of a proof that works for particle systems: the ASEP is a generalization of the Hammersley process and in [22] for the last-passage model associated with TASEP. Current fluctuations for stationary TASEP were analyzed in [20]. Here is a selection of further results in this direction: [4, 14, 21, 23, 28]. Recently, the asymmetric simple exclusion (ASEP) also got within reach of these techniques [34].

Two types of mathematical work should be distinguished. The hypothesis of microscopic concavity consists of control of second class particles that is a microscopic counterpart of the macroscopic effect that concavity of \( \mathcal{H} \) has on characteristics. We make this technically precise in Section 2.6. Once the microscopic concavity assumption is made the proof works for the entire class of processes. This then is the sense in which we take a step toward universality. As a by-product, we also obtain superdiffusivity of the second class particle in the stationary process.

Earlier proofs of \( t^{1/3} \) fluctuations have been quite rigid in the sense that they work only for particular cases of the models where special combinatorial properties emerge as if through some fortuitous coincidences. There is basically no room for perturbing the rules of the process. By contrast, the proof given in the present paper works for the whole class of processes. The hypothesis of microscopic concavity that is required is certainly nontrivial. But it does not seem to rigidly exclude all but a handful of the processes in the broad class. The estimates that it requires can probably be proved in different ways for different subclasses of the processes. And the proof itself may evolve further and weaken the hypothesis required.

To summarize, we are currently able to verify the required hypothesis of microscopic concavity for the following three subclasses of processes.

(i) The asymmetric simple exclusion process (ASEP). Full details of this case are reported elsewhere [12] and we give a brief informal description in Section 2.8.1. This proof is somewhat simpler than the earlier one given in [13].

(ii) Totally asymmetric zero range processes (TAZRP) with a concave jump rate function whose slope decreases geometrically, and may be eventually constant. This example has been out of reach for existing methods, so it is completely new in this context. It is developed fully in the present paper. As a special case, the result of [7] for the constant rate TAZRP is also recovered.
(iii) The totally asymmetric bricklayers process with convex, exponential jump rate. This system satisfies the analogous microscopic convexity. Due to the fast growth of the jump rate function this example needs more preliminary work than was sensible to include in the present paper, and so the result will be published separately in the future.

We expect that a broader class of totally asymmetric concave zero range processes should be amenable to further progress because a key part of the hypothesis can be verified, and only a certain tail estimate is missing. We explain this in Section 2.8.2.

Interacting particle systems can naturally be given a surface growth representation where integrated particle current becomes the height of a surface and particle occupations become (negative) discrete gradients of this surface. We found this picture extremely helpful in visualizing currents and couplings, hence this is the way we introduce and handle the processes.

This paper has two parts. In the main part we prove the general fluctuation bound under the assumptions needed for membership in the class of processes and the assumption of microscopic concavity. The remainder of the paper shows that the assumptions required by the general result are satisfied by a class of zero range processes. Here is a section by section outline.

In Section 2 we define the general family of processes under consideration, describe the microscopic concavity property and other assumptions used, and state the general results. Partly as corollaries to the fluctuation bound along the characteristic we obtain a law of large numbers for a second class particle and limits that show how fluctuations in noncharacteristic directions on the diffusive scale come directly from fluctuations of the initial state. Section 2.8 describes two examples. Section 2.8.1 gives a brief description of how the asymmetric simple exclusion process (ASEP) satisfies the assumptions of our general theorem. (Full details for this example are reported in [12].) Section 2.8.2 describes a class of totally asymmetric zero range processes with concave jump rates that increase with exponentially decaying slope.

The general theorem is proved in two parts: the upper bound in Section 3 and the lower bound in Section 4. Section 5 proves a strong law for the second class particle, partly as a corollary of the main fluctuation bounds. We then return to the zero range example and give a complete proof for this class of processes in Section 6.

The three-part Appendix contains auxiliary computations for the stationary distribution and hydrodynamic flux function. In particular, if the jump rate function of a zero range process is concave and not linear then the hydrodynamic flux $H$ satisfies $H''(\rho) < 0$ for all densities $0 < \rho < \infty$.

**Notation.** We summarize here some notation for easy reference. $\mathbb{Z}^+ = \{0, 1, 2, \ldots\}$, $\mathbb{R}^+ = [0, \infty)$. Centering a random variable is denoted by $\tilde{X} = X - EX$. Constants $C$, $\alpha$ do not depend on time, but may depend on the density parameter $\rho$ and their values can change from line to line. The numbering of these constants is of no particular significance and is meant only to facilitate following the arguments.

2. Definitions and results

We define the class of processes studied in this paper, give a list of examples, and discuss some of basic properties. Then come the hypotheses and main results of this paper, followed by two examples of subclasses of processes for which the hypotheses can be verified.

2.1. A family of deposition processes

The family of processes we consider is the one described in [11], and we repeat the definition here. We start with the interface growth picture, but we end up using the height and particle languages interchangeably. For extended-integer-valued boundaries $-\infty \leq \omega_{\min} \leq 0$ and $1 \leq \omega_{\max} \leq \infty$ define the single-site state space

$$I := \{z \in \mathbb{Z}: \omega_{\min} - 1 < z < \omega_{\max} + 1\}$$

and the increment configuration space

$$\Omega := \{\omega = (\omega_i)_{i \in \mathbb{Z}}: \omega_i \in I\} = I^\mathbb{Z}.$$
At times it will be convenient to have notation for the increment configuration \( \delta_j \in \Omega \) with exactly one nonzero entry equal to 1:

\[
(\delta)_{j} = \begin{cases} 
1 & \text{for } i = j, \\
0 & \text{for } i \neq j.
\end{cases}
\]  

(2.1)

For each pair of neighboring sites \( i \) and \( i + 1 \) of \( \mathbb{Z} \) imagine a column of bricks over the interval \( (i, i + 1) \). The height \( h_i \) of this column is integer-valued. The components of a configuration \( \omega \in \Omega \) are the negative discrete gradients of the heights: \( \omega_i = h_{i-1} - h_i \in I \).

The evolution is described by jump processes whose rates \( p \) and \( q \) are nonnegative functions on \( I \times I \). Two types of moves are possible. A brick can be deposited:

\[
(\omega_i, \omega_{i+1}) \rightarrow (\omega_i - 1, \omega_{i+1} + 1) \quad \text{with rate } p(\omega_i, \omega_{i+1}),
\]  

(2.2)

or removed:

\[
(\omega_i, \omega_{i+1}) \rightarrow (\omega_i + 1, \omega_{i+1} - 1) \quad \text{with rate } q(\omega_i, \omega_{i+1}).
\]  

(2.3)

Conditionally on the present state, these moves happen independently at all sites \( i \). We can summarize this information in the formal infinitesimal generator \( L \) of the process \( \omega(\cdot) \):

\[
(L \varphi)(\omega) = \sum_{i \in \mathbb{Z}} p(\omega_i, \omega_{i+1}) \cdot \left[ \varphi(\ldots, \omega_i - 1, \omega_{i+1} + 1, \ldots) - \varphi(\omega) \right]
\]

\[
+ \sum_{i \in \mathbb{Z}} q(\omega_i, \omega_{i+1}) \cdot \left[ \varphi(\ldots, \omega_i + 1, \omega_{i+1} - 1, \ldots) - \varphi(\omega) \right].
\]  

(2.4)

\( L \) acts on bounded cylinder functions \( \varphi : \Omega \to \mathbb{R} \) (this means that \( \varphi \) depends only on finitely many \( \omega_i \)-values).

Thus we have a Markov process \( \{\omega(t) : t \in \mathbb{R}^+\} \) of an evolving increment configuration and a Markov process \( \{h(t) : t \in \mathbb{R}^+\} \) of an evolving height configuration. The initial increments \( \omega(0) \) specify the initial height \( h(0) \) up to a vertical translation. We shall always normalize the height process so that \( h_0(0) = 0 \).

In the particle picture the variable \( \omega_i(t) \) represents the number of particles at site \( i \) at time \( t \). Step (2.2) represents a rightward jump of a particle over the edge \((i, i + 1)\), while step (2.3) represents a leftward jump. (If negative \( \omega \)-values are permitted, one needs to consider particles and antiparticles, with antiparticles jumping in the opposite direction.) Figure 1 shows a configuration and a possible step with both walls and particles. It is in the particle guise that many of these processes appear in the literature: simple exclusion processes, zero range processes and misanthrope processes are examples included in the class studied in this paper.

It will be useful to see that

\[
h_i(t) = h_i(t) - h_i(0) = \text{the net number of particles that have passed, from left to right,}
\]

the straight-line space–time path that connects \((1/2, 0)\) to \((i + 1/2, t)\).

(2.5)

In particular, height increment \( h_i(t) - h_i(0) \) is the cumulative net particle current across the edge \((i, i + 1)\) during time \((0, t)\).

We impose the following four assumptions (2.6)–(2.9) on the rates.

- The rates \( p, q : I \times I \to \mathbb{R}^+ \) must satisfy

\[
p(\omega^{\text{min}}, \cdot) = p(\cdot, \omega^{\text{max}}) \equiv q(\omega^{\text{max}}, \cdot) \equiv q(\cdot, \omega^{\text{min}}) \equiv 0
\]  

whenever either \( \omega^{\text{min}} \) or \( \omega^{\text{max}} \) is finite. Either both \( p \) and \( q \) are strictly positive in all other cases, or one of them is identically zero. The process is called totally asymmetric if either \( q \equiv 0 \) or \( p \equiv 0 \).
The dynamics has a smoothing effect when we assume the following monotonicity:

\[ p(z + 1, y) \geq p(z, y), \quad p(y, z + 1) \leq p(y, z), \]
\[ q(z + 1, y) \leq q(z, y), \quad q(y, z + 1) \geq q(y, z) \]

(2.7)

for \( y, z, z + 1 \in I \). Under this property the higher the neighbors of a column, the faster it grows and the longer it waits for a brick removal, on average. This is the notion of attractivity.

The next two assumptions guarantee the existence of translation-invariant product-form stationary measures. (Similar assumptions were employed by Cocozza-Thivent [16].)

- For any \( x, y, z \in I \)

\[ p(x, y) + p(y, z) + p(z, x) + q(x, y) + q(y, z) + q(z, x) = p(x, z) + p(z, y) + p(y, x) + q(x, z) + q(z, y) + q(y, x). \]

(2.8)

- There are symmetric functions \( s_p \) and \( s_q \) on \( I \times I \), and a function \( f \) on \( I \) such that \( f(\omega_{\min}) = 0 \) whenever \( \omega_{\min} \) is finite, \( f(z) > 0 \) for \( z > \omega_{\min} \), and for any \( y, z \in I \),

\[ p(y, z) = s_p(y, z + 1) f(y) \quad \text{and} \]
\[ q(y, z) = s_q(y + 1, z) f(z). \]

(2.9)

(Interpret \( s_p(y, z) = s_q(y, z) = 0 \) if \( y \) or \( z > \omega_{\max} \).) Condition (2.7) implies that \( f \) is nondecreasing on \( I \).

An attempt at covering this broad class of processes raises the uncomfortable point that there is no unified existence proof for this entire class. Different constructions in the literature place various boundedness or growth conditions on \( p \) and \( q \) and the space \( I \), and result in various degrees of regularity for the semigroup. (Among key references are Liggett’s monograph [27], and articles [1,9] and [26].) These existence matters are beyond the scope of this paper. Yet we wish to give a general proof for fluctuations that in principal works for all processes in the family, subject to
the more serious assumptions we explain in Section 2.6. To avoid extraneous technical issues we make the following blanket assumptions on the rates \( p \) and \( q \) to be considered.

- We assume that the increment process \( \omega(t) \), and the corresponding height process \( h(t) \) with normalization \( h_0(0) = 0 \), that obey Poisson rates \( p \) and \( q \) as described by (2.2) and (2.3), can be constructed with cadlag paths in a subspace \( \mathcal{W} \) of tempered increment configurations (i.e. configurations that obey some restrictive growth conditions).
- The subspace \( \mathcal{W} \) is of full measure under the invariant distributions \( \mu^\theta \) defined in Section 2.4.
- It is also possible to construct jointly several versions of the process with initial configurations from the space \( \mathcal{W} \) and with joint evolution obeying basic coupling (described in Section 2.3).
- Rates \( p \) and \( q \) have all moments under the invariant distributions \( \mu^\theta \). In fact arguments like Lemma C.2 of the Appendix provide this when \( f \) does not grow faster than exponential on \( \mathbb{Z}^+ \) and does not decrease faster to zero than exponential on \( \mathbb{Z}^- \).

The reader will see that our proofs in Sections 3–6 do not make any analytic demands on the semigroup and its relation to the generator. We only use couplings, counting of particle currents and simple Poisson bounds.

Two identities from article [11] play a key role in this paper, given as (2.19) and (2.20) in Section 2.5. These identities hold for all processes in the family under study. The proofs given in [11] use generator calculations which may not be justified for all these processes. However, these identities can also be proved by counting particles and taking limits of finite-volume processes ([12] contains an example). Such a proof should be available with any reasonable construction of a process. Hence we shall not hesitate to use the results of [11].

2.2. Examples

To give concrete meaning to the general formulation of the previous section we describe some basic examples. The type of state space \( I \) distinguishes three cases that we call generalized exclusion, misanthrope and bricklayers processes. In all cases there are two parameters \( 0 \leq p, q \leq 1 \) such that \( p + q = 1 \). Asymmetric processes have \( p \neq q \).

These are the processes for which our results are relevant.

1. Generalized exclusion processes. These are the cases where both \( \omega_{\min} \) and \( \omega_{\max} \) are finite.

- The asymmetric simple exclusion process (ASEP) introduced by F. Spitzer [33] is defined by \( \omega_{\min} = 0, \omega_{\max} = 1 \), \( f(z) = 1 \{ z = 1 \} \), \( s_p(y,z) = p \cdot 1 \{ y = z = 1 \} \) and \( s_q(y,z) = q \cdot 1 \{ y = z = 1 \} \). This produces the familiar rates

\[
p(y,z) = p \cdot 1 \{ y = 1, z = 0 \} \quad \text{and} \quad q(y,z) = q \cdot 1 \{ y = 0, z = 1 \}.
\]

Here \( \omega_i \in \{ 0, 1 \} \) is the occupation number for site \( i \), \( p(\omega_i, \omega_{i+1}) \) is the rate for a particle to jump from site \( i \) to \( i + 1 \), and \( q(\omega_i, \omega_{i+1}) \) is the rate for a particle to jump from site \( i + 1 \) to \( i \). These rates have values \( p \) and \( q \), respectively, whenever there is a particle to perform the above jumps, and there is no particle on the terminal site of the jumps. Conditions (2.7) and (2.8) are also satisfied by these rates.

- Particle–antiparticle exclusion process. Let \( \omega_{\min} = -1, \omega_{\max} = 1 \). Take \( f(-1) = 0, f(0) = c \) (creation), \( f(1) = a \) (annihilation) where \( c \) and \( a \) are positive rates with \( c \leq a/2 \),

\[
s_p(0, 1) = s_p(1, 0) = p, \quad s_p(0, 0) = \frac{pa}{2c}, \quad s_p(1, 1) = \frac{p}{2},
\]

\[
s_q(0, 1) = s_q(1, 0) = q, \quad s_q(0, 0) = \frac{qa}{2c}, \quad s_q(1, 1) = \frac{q}{2}
\]

and \( s_p, s_q \) zero in all other cases. These result in rates

\[
p(0, 0) = pc, \quad p(0, -1) = p(1, 0) = \frac{pa}{2}, \quad p(1, -1) = pa,
\]

\[
q(0, 0) = qc, \quad q(-1, 0) = q(0, 1) = \frac{qa}{2}, \quad q(-1, 1) = qa
\]

and zero in all other cases. If \( \omega_i \) is the number of particles at site \( i \), with \( \omega_i = -1 \) meaning the presence of an antiparticle, then this model describes an asymmetric exclusion process of particles and antiparticles with annihilation and particle–antiparticle pair creation. These rates also satisfy our conditions.
One can imagine other generalizations with bounded numbers of particles and/or antiparticles per site.

2. **Generalized misanthrope processes** have \( \omega^{\min} > -\infty, \omega^{\max} = \infty \).

   - **Zero range process.** Take \( \omega^{\min} = 0, \omega^{\max} = \infty \), an arbitrary nondecreasing function \( f : \mathbb{Z}^+ \to \mathbb{R}^+ \) such that \( f(0) = 0 \),
     \[
     s_p(y, z) \equiv p \quad \text{and} \quad s_q(y, z) \equiv q, \quad p(y, z) = pf(y) \quad \text{and} \quad q(y, z) = qf(z).
     \]
     Again, \( \omega_i \) represents the number of particles at site \( i \). Depending on this number, a particle jumps from \( i \) to the right with rate \( pf(\omega_i) \), and to the left with rate \( qf(\omega_i) \). These rates trivially satisfy conditions (2.7) and (2.8).

3. **General deposition processes** have \( \omega^{\min} = -\infty \) and \( \omega^{\max} = \infty \). The height differences between adjacent columns can be arbitrary integers. Antiparticles are needed for a particle representation of the process.

   - **Bricklayers process.** Let \( f : \mathbb{Z} \to \mathbb{R}^+ \) be nondecreasing and satisfy
     \[ f(z) \cdot f(1 - z) = 1 \quad \text{for all} \quad z \in \mathbb{Z}. \]
     The values of \( f \) for positive \( z \)'s thus determine the values for nonpositive \( z \)'s. Let
     \[
     s_p(y, z) = p + \frac{p}{f(y)f(z)} \quad \text{and} \quad s_q(y, z) = q + \frac{q}{f(y)f(z)},
     \]
     which results in
     \[
     p(y, z) = pf(y) + pf(-z) \quad \text{and} \quad q(y, z) = qf(-y) + qf(z).
     \]
     The following picture motivates the name bricklayers process. At each site \( i \) stands a bricklayer who lays a brick on the column to his left at rate \( pf(-\omega_i) \) and on the column to his right at rate \( pf(\omega_i) \). Each bricklayer also removes a brick from his left at rate \( qf(\omega_i) \) and from his right at rate \( qf(-\omega_i) \). Conditions (2.7) and (2.8) hold for the rates.

     These were examples for which our theorem holds, provided the hypotheses on microscopic concavity to be described below can be verified.

2.3. **Basic coupling**

In **basic coupling** the joint evolution of \( n \) processes \( \omega^{m}(\cdot), m = 1, \ldots, n \), is defined in such a manner that the processes “jump together as much as possible.” The joint rates are determined as follows, given the current configurations \( \omega^1, \omega^2, \ldots, \omega^n \in \mathcal{L}^2 \). Consider a step of type (2.2) over the edge \((i, i+1)\). Let \( m \mapsto \ell(m) \) be a permutation that orders the rates of the individual processes for this move:

\[
r(m) \equiv p(\omega_i^{\ell(m)}, \omega_{i+1}^{\ell(m)}) \leq p(\omega_i^{\ell(m+1)}, \omega_{i+1}^{\ell(m+1)}) \equiv r(m+1), \quad 1 \leq m < n.
\]

Set also the dummy value \( r(0) = 0 \). Now the rule is that independently for each \( m = 1, \ldots, n \), at rate \( r(m) - r(m - 1) \), precisely processes \( \omega^{\ell(m)}, \omega^{\ell(m+1)}, \ldots, \omega^{\ell(n)} \) execute move (2.2), and processes \( \omega^{(1)}, \omega^{(2)}, \ldots, \omega^{(m-1)} \) do not. The combined effect of these joint rates creates the correct marginal rates, that is, process \( \omega^{\ell(m)} \) executes this move with rate \( r(m) \).

Notice also that, due to (2.7), a jump of \( \omega^a \) without \( \omega^b \) can only occur if \( p(\omega_i^a, \omega_{i+1}^b) < p(\omega_i^b, \omega_{i+1}^a) \) which implies \( \omega_i^a > \omega_i^b \) or \( \omega_{i+1}^a < \omega_{i+1}^b \). The result of this step (2.2) then cannot increase the number of discrepancies between the two processes, hence the name **attractivity** for (2.7). In particular, a sitewise ordering \( \omega_i^a \leq \omega_i^b \) \( \forall i \in \mathbb{Z} \) is preserved by the basic coupling.

One can check that moves of type (2.3) with rates \( q \) obey the same attractivity property.

The differences between two processes are called **second class particles**. Their number is nonincreasing. In particular, if \( \omega_i^a \geq \omega_i^b \) for each \( i \in \mathbb{Z} \), then the second class particles are conserved. In view of (2.5), in this case the net
number of second class particles that pass from left to right across the straight-line space–time path from \((1/2, 0)\) to \((i + 1/2, t)\) equals the growth difference
\[
(h_i^a(t) - h_0^a(0)) - (h_i^b(t) - h_0^b(0)) = h_i^a(t) - h_i^b(t)
\] (2.10)
between the two processes \(\omega^a(\cdot)\) and \(\omega^b(\cdot)\).

A special case that is of key importance to us is the situation where only one second class particle is present between two processes.

### 2.4. Translation invariant stationary product distributions

The results of this paper concern stationary processes with particular product-form marginal distributions that we define in this section. For many cases it has been proved that these measures are the only extremal translation-invariant stationary distributions. Following some ideas in Cocozza-Thivent [16], we first consider the nondecreasing function \(f\) whose existence was assumed in (2.9). For \(I \ni z > 0\) define
\[
f(z) := \prod_{y=1}^{z} f(y),
\]
while for \(I \ni z < 0\) let
\[
f(z) := \frac{1}{\prod_{y=z+1}^{0} f(y)},
\]
and then \(f(0) := 1\). This definition satisfies \(f(z)! \cdot f(z+1) = f(z+1)!\) for all \(z \in I\). Let
\[
\bar{\theta} := \begin{cases} 
\log \left( \liminf_{z \to \infty} \left( f(z) \right)^{1/z} \right) = \lim_{z \to \infty} \log(f(z)), & \text{if } \omega_{\max} = \infty \\
\infty, & \text{else}
\end{cases}
\]
and
\[
\bar{\theta} := \begin{cases} 
\log \left( \limsup_{z \to \infty} \left( f(-z) \right)^{-1/z} \right) = \lim_{z \to \infty} \log(f(-z)), & \text{if } \omega_{\min} = -\infty \\
-\infty, & \text{else}
\end{cases}
\]
By monotonicity of \(f\), we have \(\bar{\theta} \geq \bar{\theta}\). The case \(\bar{\theta} = \bar{\theta}\) would imply that \(\omega_{\min} = -\infty, \omega_{\max} = \infty\), and \(f\) is a constant. Notice that (2.7) and (2.9) imply that \(s_p\) is nonincreasing in its variables, but \(p\) is nondecreasing in its first variable. Hence a constant \(f\) results in an \(s_p\) that does not depend on its first variable. But then by its symmetric property it does not depend on its second variable either, and we conclude that a constant \(f\) implies constant rates \(p\) (and, similarly, \(q\)). We exclude this uninteresting case by postulating
\[
\text{assume } f \text{ to be such that } \bar{\theta} < \bar{\theta}. \quad (2.11)
\]
For \(\theta \in (\bar{\theta}, \bar{\theta})\) define the state sum
\[
Z(\theta) := \sum_{z \in I} \frac{e^{\theta z}}{f(z)!} < \infty. \quad (2.12)
\]
Let the product-distribution \(\mu^\theta\) on \(\Omega = I^\mathbb{Z}\) have marginals
\[
\mu^\theta(z) = \mu^\theta(\omega; \omega_i = z) := \frac{1}{Z(\theta)} \cdot \frac{e^{\theta z}}{f(z)!} \quad (z \in I). \quad (2.13)
\]
Assumptions (2.6), (2.7), (2.8), (2.9) imply that for \(\theta \in (\bar{\theta}, \bar{\theta})\) the product distribution \(\mu^\theta\) is stationary for the process generated by (2.4) (see [11]; notice that the top display on page 443 of [11] is incorrect, to get the correct identity,
multiply with the cylinder functions and take expectation). For some calculations in the Appendix it will be convenient to note that the family \([\mu^\epsilon]\) can be obtained by exponentially weighting a probability measure \(\mu^{\theta_0}\) for a fixed value \(\theta_0 \in (\tilde{\theta}, \tilde{\theta})\).

\(P^\theta, E^\theta, \text{Var}^\theta, \text{Cov}^\theta\) will refer to laws of a process evolving in this stationary distribution. In the Appendix we show that the density

\[q(\theta) := E^\theta(\omega)\]

is a strictly increasing, infinitely differentiable function of the parameter \(\theta\) that maps the interval \((\tilde{\theta}, \tilde{\theta})\) onto the interval \((\omega_{\min}, \omega_{\max})\). (The following point should cause no confusion: the single-site state space \(I\) consists of the integers between \(\omega_{\min}\) and \(\omega_{\max}\), including endpoints if finite, but for density values the interval \((\omega_{\min}, \omega_{\max})\) is an interval of real numbers.) For most cases we shall use the density \(q\), rather than \(\theta\), for parameterizing the stationary distributions. Accordingly, \(\mu^\epsilon, P^\epsilon, E^\epsilon, \text{Var}^\epsilon, \text{Cov}^\epsilon\) will refer to laws of a density \(q\) stationary process.

2.5. Hydrodynamics and some exact identities

The hydrodynamic flux is defined as

\[\mathcal{H}(q) := E^\theta(p(\omega_0, \omega_1) - q(\omega_0, \omega_1)).\quad (2.14)\]

\(\mathcal{H}(q)\) is the expected net rate at which a given column grows, or at which particles pass any fixed lattice edge from left to right in a stationary density-\(q\) process. We show smoothness of \(\mathcal{H}\) in Section C of the Appendix. It is expected, and in many instances proved, that asymmetric members of our class satisfy the conservation law

\[\partial_T q(T, X) + \partial_X \mathcal{H}(q(T, X)) = 0\]

in the Eulerian-scaled time and space variables \(T\) and \(X\), see e.g. Rezakanlou [31] or Bahadoran, Guiol, Ravishankar and Saada [2]. The characteristic speed is the velocity with which small perturbations of the solution of this PDE propagate, and is given by

\[V^\epsilon := \mathcal{H}'(q).\quad (2.15)\]

A particular expectation we shall need several times is

\[E^\theta(h_i(t)) = \mathcal{H}(q)t - q_i, \quad t \geq 0, i \in \mathbb{Z}.\quad (2.16)\]

For \(i = 0\) this follows from (2.5), and in general from the \(i = 0\) case together with \(\omega_j(t) = h_{j-1}(t) - h_j(t)\).

When a stationary process is perturbed by adding a second class particle at the origin at time zero, we obtain two processes, \(\omega^- (\cdot)\) and \(\omega^+ (\cdot)\). It is not a priori clear what the initial joint distribution of the occupation variables \(\omega^- (0), \omega^+ (0)\) should be. For ASEP there is no ambiguity due to the simplicity of the single-site state space: the only way to have a discrepancy is to set \(\omega^- (0) = 0, \omega^+ (0) = 1\). A useful generalization of this distribution to the broader class of processes involves the following family of probability measures on \(I\) introduced in [11]:

\[\tilde{\mu}^\epsilon(y) := \frac{1}{\text{Var}^\epsilon(\omega_0)} \sum_{\omega = y + 1}^{\omega_{\max}} (z - q) \mu^\epsilon(z), \quad y \in I.\quad (2.17)\]

An empty sum is zero by convention and so if \(\omega_{\max} < \infty, \tilde{\mu}^\epsilon(\omega_{\max}) = 0\). Consequently there is room for an additional particle under the \(\tilde{\mu}^\epsilon\) distribution, in the sense that if \(\omega \sim \tilde{\mu}^\epsilon\) then also \(\omega + 1 \in I\).

To our knowledge these distributions \(\tilde{\mu}^\epsilon\) do not possess any invariance properties. Their virtue is that they make identities (2.19) and (2.20) below true. We show in Section B of the Appendix that both \(\mu^\epsilon\) and \(\tilde{\mu}^\epsilon\) are stochastically monotone in the density \(q\). (There is, however, no stochastic domination between \(\mu^\epsilon\) and \(\tilde{\mu}^\epsilon\) in general.)

Denote by \(E\) the expectation w.r.t. the evolution of a pair \((\omega^-, \omega^+)\) started with initial data (recall (2.1))

\[\omega^- (0) = \omega(0) - \delta_0 \sim \left(\bigotimes_{i \neq 0} \mu^\epsilon\right) \otimes \tilde{\mu}^\epsilon,\quad (2.18)\]
and evolving under the basic coupling. This pair will always have a single second class particle whose position is denoted by \( Q(t) \). In other words, \( \omega^-(t) = \omega(t) - \delta_{\bar{Q}(t)} \). Corollaries 2.4 and 2.5 of [11] state that

\[
\text{Var}^\theta(h_i(t)) = \text{Var}^\theta(\omega) \cdot E \left| Q(t) - i \right|
\]

(2.19)

and

\[
E(\omega(t)) = V^\theta \cdot t
\]

(2.20)

for any \( i \in \mathbb{Z} \) and \( t \geq 0 \). Note in particular that in (2.19) the variances are taken in a stationary process, while the expectation of \( Q(t) \) is taken in the coupling with initial distribution (2.18). These two identities follow from the definition of our models together with translation invariance and the product structure of the stationary distribution.

2.6. Microscopic concavity

From now on fix the jump rates \( p,q : I \times I \to \mathbb{R}^+ \) that define the process in question, assumed to satisfy all the assumptions discussed thus far. The \( t^{1/3} \) current or height fluctuations are expected when the hydrodynamic flux \( H(\varrho) \) is strictly concave or convex. In this paper we discuss only the strictly concave case. This implies that the characteristic speed \( V^\varrho = H'(\varrho) \) is a decreasing function of density \( \varrho \):

\[
\lambda < \varrho \implies V^\lambda > V^\varrho.
\]

(2.21)

The microscopic counterpart of a characteristic is the motion of a second class particle. Our key assumption that we term microscopic concavity is that the ordering (2.21) can also be realized at the particle level as an ordering between two second class particles introduced into two processes at densities \( \lambda \) and \( \varrho \). Since this is now a probabilistic notion, there are several possible formulations, ranging from almost sure (\( Q^\lambda(t) \geq Q^\varrho(t) \) in a coupling) to distributional formulations. Assumption 2.1 below gives the precise technical form in which this paper utilizes this notion of microscopic concavity. It stipulates that the ordering of second class particles is achieved by processes that evolve on the labels of auxiliary second class particles, and also requires some control of the tails of these random labels.

We do not imagine that this precise formulation will be the right one for all processes. We take it as a starting point and future work may lead to alternative formulations. Assumption 2.1 has the virtue that its requirements can be verified for some interesting processes.

Let \( \lambda < \varrho \) be two densities. Proposition B.4 in the Appendix gives the stochastic domination \( \hat{\mu}^\lambda \leq \hat{\mu}^\varrho \). Define \( \hat{\mu}^\varrho + 1 \) as the measure that gives weight \( \hat{\mu}^\varrho(z - 1) \) to an integer \( z \) such that \( \omega_{\text{min}} < z < \omega_{\text{max}} + 1 \). Let \( \hat{\mu}^{\lambda,\varrho} \) be a coupling measure with marginals \( \hat{\mu}^\lambda \) and \( \hat{\mu}^\varrho + 1 \) and with the property

\[
\hat{\mu}^{\lambda,\varrho}\{(y,z) : \omega_{\text{min}} - 1 < y < z < \omega_{\text{max}} + 1\} = 1.
\]

(2.22)

Let also \( \mu^{\lambda,\varrho} \) be a coupling measure of site-marginals \( \mu^\lambda \) and \( \mu^\varrho \) of the invariant distributions, with

\[
\mu^{\lambda,\varrho}\{(y,z) : \omega_{\text{min}} - 1 < y \leq z < \omega_{\text{max}} + 1\} = 1,
\]

(2.23)

this is possible by Corollary B.3 of the Appendix. Note the distinction that under \( \hat{\mu}^{\lambda,\varrho} \) the second coordinate is strictly above the first.

To have notation for inhomogeneous product measures on \( I^\mathbb{Z} \), let \( \lambda = (\lambda_i)_{i \in \mathbb{Z}} \) and \( \varrho = (\varrho_i)_{i \in \mathbb{Z}} \) denote sequences of density values, with \( \lambda_i \) and \( \varrho_i \) assigned to site \( i \). The product distribution with marginals \( \hat{\mu}^{\lambda_0,\varrho_0} \) at the origin and \( \mu^{\lambda_i,\varrho_i} \) at other sites is denoted by

\[
\hat{\mu}^{\lambda,\varrho} := \left(\bigotimes_{i \neq 0} \mu^{\lambda_i,\varrho_i}\right) \otimes \hat{\mu}^{\lambda_0,\varrho_0}.
\]

(2.24)

Measure \( \hat{\mu}^{\lambda,\varrho} \) gives probability one to the event

\[
\{(\eta(0), \omega(0)) : \eta_0(0) < \omega_0(0) \text{, and } \eta_i(0) \leq \omega_i(0) \text{ for } 0 \neq i \in \mathbb{Z}\}.
\]
The initial configuration \((\eta(0), \omega(0))\) will always be assumed a member of this set, and the pair process \((\omega(t), \omega(t))\) evolves in basic coupling. In general \(\tilde{\mathcal{L}}_{\omega-\eta}\) is not stationary for this joint evolution.

The discrepancies between these two processes are called the \(\omega - \eta\) (second class) particles. The number of such particles at site \(i\) at time \(t\) is \(\omega_i(t) - \eta_i(t)\). In the basic coupling the \(\omega - \eta\) particles are conserved, in the sense that none are created or annihilated. We label the \(\omega - \eta\) particles with integers, and let \(X_m(t)\) denote the position of particle \(m\) at time \(t\). The initial labeling is chosen to satisfy

\[
\cdots \leq X_{-1}(0) \leq X_0(0) = 0 < X_1(0) \leq \cdots.
\]

We can specify that \(X_0(0) = 0\) because under \(\tilde{\mathcal{L}}_{\omega-\eta}\) there is an \(\omega - \eta\) particle at site 0 with probability 1. During the evolution we keep the positions \(X_i(t)\) of the \(\omega - \eta\) particles ordered. To achieve this we stipulate that

whenever an \(\omega - \eta\) particle jumps from a site, if the jump is to the right the highest label moves, and if the jump is to the left the lowest label moves.

(2.25)

Here is the precise form of microscopic concavity for this paper. The assumption states that a certain joint construction of processes (that is, a coupling) can be performed for a range of densities in a neighborhood of a fixed density \(\varrho\). Recall (2.1) for the definition of the configuration \(\hat{\delta}\).

**Assumption 2.1.** Given a density \(\varrho \in (\omega_{\text{min}}, \omega_{\text{max}})\), there exists \(\gamma_0 > 0\) such that the following holds. For any \(\lambda\) and \(\varrho\) such that \(\varrho - \gamma_0 \leq \lambda_i \leq \varrho \leq \varrho + \gamma_0\) for all \(i \in \mathbb{Z}\), a joint process \((\omega(t), \omega(t), y(t), z(t))_{t \geq 0}\) can be constructed with the following properties.

- Initially \((\eta(0), \omega(0))\) is \(\tilde{\mathcal{L}}_{\omega-\eta}\)-distributed and the joint process \((\eta(\cdot), \omega(\cdot))\) evolves in basic coupling.
- Processes \(y(\cdot)\) and \(z(\cdot)\) are integer-valued. Initially \(y(0) = z(0) = 0\). With probability one
  \[
y(t) \leq z(t) \quad \text{for all} \ t \geq 0.
  \]
- Define the processes
  \[
  \omega^- (t) := \omega(t) - \hat{\delta}X_y(t) \quad \text{and} \quad \eta^+(t) := \eta(t) + \hat{\delta}X_z(t).
  \]

Then both pairs \((\eta, \eta^+)\) and \((\omega^-, \omega)\) evolve marginally in basic coupling.

- For each \(\gamma \in (0, \gamma_0)\) and large enough \(t \geq 0\) there exists a probability distribution \(v^{\varrho, \gamma}(t)\) on \(\mathbb{Z}^+\) satisfying the tail bound
  \[
v^{\varrho, \gamma}(t)\{ y: \ y \geq \gamma_0 \} \leq Ct^{\kappa - 1} 2^{2\gamma - 3\gamma_0 - \kappa}
  \]
  for some fixed constants \(3/2 \leq \kappa < 3\) and \(C < \infty\), and such that if \(\varrho - \gamma \leq \lambda_i \leq \varrho + \gamma\) for all \(i \in \mathbb{Z}\), then we have the stochastic bounds
  \[
y(t) \overset{d}{\leq} v^{\varrho, \gamma}(t) \quad \text{and} \quad z(t) \overset{d}{\geq} -v^{\varrho, \gamma}(t).
  \]

Let us clarify some of the details in this assumption.

Equation (2.27) says that \(Q^\gamma(t) := X_z(t)(t)\) is the single second class particle between \(\eta\) and \(\eta^+\), while \(Q(t) := X_y(t)(t)\) is the one between \(\omega^-\) and \(\omega\). The first three bullets say that it is possible to construct jointly four processes \((\eta, \eta^+, \omega^-, \omega)\) with the specified initial conditions and so that each pair \((\eta, \omega), (\eta, \eta^+)\) and \((\omega^-, \omega)\) has the desired marginal distribution, and most importantly so that

\[
Q^\gamma(t) = X_z(t)(t) \geq X_y(t)(t) = Q(t).
\]

(2.30)

This is a consequence of (2.26) because the \(\omega - \eta\) particles \(X_i(t)\) stay ordered.
The tail bound (2.28) is formulated in this somewhat complicated fashion because this appears to be the weakest form our present proof allows. In our currently available examples $\nu^{\phi,\gamma}(t)$ is actually a fixed geometric distribution. However, we expect that other examples will require more complicated bounds and so including this generality is sensible.

The assumptions made imply $\eta(t) \leq \omega(t)$ a.s., and by (2.27)

$$\eta(t) \leq \eta^+(t) \leq \omega(t) \quad \text{and} \quad \eta(t) \leq \omega^-(t) \leq \omega(t) \quad \text{a.s.}$$

In our actual constructions of the processes $\eta, \eta^+, \omega^-, \omega$ for ASEP (Section 2.8.1 and [12]), for a class of totally asymmetric zero range processes (Section 6) and for the totally asymmetric bricklayers process with exponential rates (future work) it turns out that the triples $(\eta, \eta^+, \omega)$ and $(\eta, \omega^-, \omega)$ evolve also in basic coupling, but the full joint evolution $(\eta, \eta^+, \omega^-, \omega)$ does not.

As already explained, the microscopic concavity idea is contained in inequality (2.26). There is also a sense in which the tail bounds (2.29) relate to concavity of the flux. Consider the situation $\lambda_i \equiv \lambda < \varrho \equiv \varrho_i$. We would expect the $\omega - \eta$ particle $X_0(\cdot)$ to have average and long-term velocity

$$R(\lambda, \varrho) = H(\varrho) - H(\lambda) = \frac{\rho - \lambda}{\rho^2},$$

the Rankine–Hugoniot or shock speed. By concavity $H'(\varrho) = V^\varrho \leq R(\lambda, \varrho) = V^\lambda = H'(\lambda)$. A strict microscopic counterpart would be $y(t) \leq 0 \leq z(t)$. But this condition is overly restrictive. The only cases we know to satisfy it are the totally asymmetric simple exclusion process and the totally asymmetric zero range process with constant rate. The distributional bounds (2.29) are natural relaxations of $y(t) \leq 0 \leq z(t)$.

By the same token, perhaps the way to covering more examples with our approach involves a similar distributional weakening of (2.26), but this seems less straightforward.

2.7. Results

We need a few more assumptions and then we can state the main result. Constants $C, \alpha$ will not depend on time, but might depend on the density parameter $\varrho$, and their values can change from line to line. We are now working with a fixed member of the class of processes described in Section 2.1 with rate functions $p, q : I \times I \to \mathbb{R}^+$. Recall that $H$ is the hydrodynamic flux defined in (2.14). In the Appendix we show $H'$ is infinitely differentiable under the restrictions on the rates placed in Section 2.1.

**Assumption 2.2.** The rates $p, q$ and density $\varrho \in (\omega_{\min}, \omega_{\max})$ have the following properties.

- The jump rate functions $p$ and $q$ satisfy assumptions (2.6), (2.7), (2.8), (2.9) and (2.11) discussed in Sections 2.1 and 2.4.
- $H''(\varrho) < 0$.
- Let $(\omega^-, \omega)$ be a pair of processes in basic coupling, started from distribution (2.18), with second class particle $Q(t)$. Then there exist constants $0 < \alpha_0, C_0 < \infty$ such that

$$P\{|Q(t)| > K\} \leq C_0 \cdot \frac{t^2}{K^3} \quad (2.31)$$

whenever $K > \alpha_0 t$ and $t$ is large enough.

As mentioned, our results are valid only for asymmetric processes. The assumption of asymmetry is implicitly contained in $H''(\varrho) < 0$. Symmetric processes have $H(\varrho) \equiv 0$. Exponential tail bounds for $|Q(t)|$ that imply assumption (2.31) hold automatically if the rates $p, q$ have bounded increments because the rates for $Q$ come from these increments of $p$ and $q$. Here is the main result.
Theorem 2.3. Let Assumptions 2.1 and 2.2 hold for density \( \varrho \). Let the processes \((\omega^-(t), \omega(t))\) evolve in basic coupling with initial distribution (2.18) and let \( Q(t) \) be the position of the second class particle between \( \omega^-(t) \) and \( \omega(t) \). Then there is a constant \( C_1 = C_1(\varrho) \in (0, \infty) \) such that for all \( 1 \leq m < 3 \),

\[
\frac{1}{C_1} < \liminf_{t \to \infty} \frac{\mathbb{E} |Q(t) - \varrho(t)|^m}{t^{2m/3}} \leq \limsup_{t \to \infty} \frac{\mathbb{E} |Q(t) - \varrho(t)|^m}{t^{2m/3}} < \frac{C_1}{3-m}. \tag{2.32}
\]

Diffusive fluctuations are characterized by a variance of order \( t \). The estimates above show that the second class particle has variance of order \( t^{2/3} \), this is called superdiffusivity.

Next some corollaries. Notation \([X]\) stands for the lower integer part of \( X \).

Corollary 2.4 (Current variance). Under Assumptions 2.1 and 2.2, there is a constant \( C_1 = C_1(\varrho) > 0 \), such that

\[
\frac{1}{C_1} < \liminf_{t \to \infty} \frac{\text{Var}^\varrho(h_{[V\varrho]})}{t^{2/3}} \leq \limsup_{t \to \infty} \frac{\text{Var}^\varrho(h_{[V\varrho]})}{t^{2/3}} < C_1.
\]

This follows from (2.19) with the choice \( m = 1 \).

Corollary 2.5 (Law of Large Numbers for the second class particle). Under Assumptions 2.1 and 2.2, the Weak Law of Large Numbers holds in a density-\( \varrho \) stationary process:

\[
Q(t) \xrightarrow{d} \varrho.
\tag{2.33}
\]

If the rates \( p \) and \( q \) have bounded increments, then almost sure convergence also holds in (2.33) (Strong Law of Large Numbers).

The Weak Law is a simple consequence of Theorem 2.3. The Strong Law will be proved in Section 5.

Corollary 2.6 (Dependence of current on the initial configuration). Under Assumptions 2.1 and 2.2, for any \( V \in \mathbb{R} \) and \( \alpha > 1/3 \) the following limit holds in the \( L^2 \) sense for a density-\( \varrho \) stationary process:

\[
\lim_{t \to \infty} \frac{h_{[V\varrho]}(t) - h_{[V\varrho]-[V\varrho]}(0)}{t^{\alpha}} = 0.
\tag{2.34}
\]

Recall that

\[
h_{[V\varrho]-[V\varrho]}(0) = \begin{cases} \sum_{i=[V\varrho]-[V\varrho]+1}^{\infty} \omega_i(0) & \text{if } V < \varrho, \\ 0 & \text{if } V = \varrho, \\ -\sum_{i=1}^{[V\varrho]-[V\varrho]} \omega_i(0) & \text{if } V > \varrho, \end{cases}
\tag{2.35}
\]

only depends on a finite segment of the initial configuration. Limit (2.34) shows that on the diffusive time scale \( t^{1/2} \) only fluctuations from the initial distribution are visible: these fluctuations are translated rigidly at the characteristic speed \( \varrho \). Proof of (2.34) follows by translating \( h_{[V\varrho]}(t) - h_{[V\varrho]-[V\varrho]}(0) \) to \( h_{[V\varrho]}(t) - h_0(0) = h_{[V\varrho]}(t) \) and by applying Corollary 2.4. From (2.34), (2.35) and the i.i.d. initial \( \{\omega_i\} \) follow a limit for the variance and a central limit theorem (CLT), which we record in our final corollary.

Corollary 2.7 (Central Limit Theorem for the current). Under Assumptions 2.1 and 2.2, for any \( V \in \mathbb{R} \) in a density-\( \varrho \) stationary process

\[
\lim_{t \to \infty} \frac{\text{Var}^\varrho(h_{[V\varrho]}(t))}{t} = \text{Var}^\varrho(\omega) \cdot |V - V| =: D,
\tag{2.36}
\]

and the Central Limit Theorem also holds: the centered and normalized height \( \tilde{h}_{[V\varrho]}(t)/\sqrt{t \cdot D} \) converges in distribution to a standard normal.
For ASEP the CLT, the limiting variance (2.36) and the appearance of initial fluctuations on the diffusive scale were proved by P. A. Ferrari and L. R. G. Fontes [18]. For convex rate zero range and bricklayers processes Corollary 2.7 was proved by M. Balázs [5].

**Remark on the convex case.** Our results and proofs work in the analogous way in the case where the flux is convex and the corresponding microscopic convexity is assumed.

2.8. Two examples that satisfy microscopic concavity

Presently we have verified all the hypotheses of Theorem 2.3 for two classes of processes.

2.8.1. The asymmetric simple exclusion process

The asymmetric simple exclusion process (ASEP) was the first example described in Section 2.2. It has two parameters $0 \leq p \neq q \leq 1$ such that $p + q = 1$. To be specific let us take $p > q$ so that on average particles prefer to drift to the right. The invariant measure $\mu^p$ is the Bernoulli distribution with parameter $0 \leq \varrho \leq 1$, while $\bar{\mu}^p$ is concentrated on zero for any $\varrho$. The hydrodynamic flux is strictly concave: $\hat{\eta}(\varrho) = (p - q)\varrho(1 - \varrho)$.

The detailed construction of the processes $y(t)$ and $z(t)$ needed for Assumption 2.1 can be found in [12]. Here it is in a nutshell.

Given the background process $(\eta(t), \varphi(t))$ and the second class particles $\{X_m(t)\}$ between them, the processes $y(\cdot)$ and $z(\cdot)$ are nearest-neighbor random walks on the labels $\{m\}$ with rates $p$ and $q$. Walk $y(\cdot)$ has bias to the left (rate $p$ to the left, rate $q$ to the right) and walk $z(\cdot)$ has bias to the right (rate $p$ to the right, rate $q$ to the left). Their jumps are restricted so that jumps between labels $m$ and $m + 1$ are permitted only when $X_m$ and $X_{m+1}$ are adjacent. The clocks governing these jumps are coupled so that the ordering $y \leq z$ is preserved.

Since a second class particle in ASEP is bounded by a rate one Poisson process, (2.31) holds.

We gave an earlier proof of Theorem 2.3 for ASEP in [13]. The present general proof evolved from that earlier one.

2.8.2. Totally asymmetric zero range process with jump rates that increase with exponentially decaying slope

This class is completely new in the sequence of models for which $t^{1/3}$-scaling of current fluctuations have been verified. Models in this class have a richer behavior than either ASEP or the totally asymmetric zero range process (TAZRP) with constant rate. As explained in Section 2.2, in a TAZRP one particle is moved from site $i$ to site $i + 1$ at rate $f(\omega_i)$, and no particle jumps to the left (our convention for total asymmetry is $p = 1 - q = 1$). The jump rate $f : \mathbb{Z}^+ \rightarrow \mathbb{R}^+$ is nondecreasing, $f(0) = 0$, and $f(z) > 0$ for $z > 0$. Assume further that $f$ is concave.

As we shall see later in Section 6, one aspect of microscopic concavity, namely the ordering of second class particles, can be achieved for any TAZRP with a nondecreasing concave jump rate. Indeed, up to Lemma 6.2 in Section 6 we only use monotonicity and concavity of the rates $f$. Thus for concave TAZRP only the tail control (2.28) and (2.29) of the label processes remains to be provided. For this part we currently need a stronger hypothesis, detailed in the next assumption.

**Assumption 2.8.** Let $p = 1 - q = 1$. Assume the jump rate function $f$ of a totally asymmetric zero range process has these properties:

- $f(0) = 0 < f(1)$,
- $f$ is nondecreasing: $f(z + 1) \geq f(z)$,
- $f$ is concave with an exponentially decreasing slope: there is an $0 < r < 1$ such that for each $z \geq 1$ such that $f(z) - f(z - 1) > 0$,

$$\frac{f(z + 1) - f(z)}{f(z) - f(z - 1)} \leq r.$$  \hspace{1cm} (2.37)

The case where $f$ becomes constant above some $z_0$ is included.

**Theorem 2.9.** Under Assumption 2.8, a stationary totally asymmetric zero range process satisfies the conclusions of Theorem 2.3, and the conclusions of Corollaries 2.4, 2.5, 2.6 and 2.7.
A class of examples of rates that satisfy Assumption 2.8 are

\[ f(z) = 1 - \exp(-\beta z^0), \quad \beta > 0, \vartheta \geq 1. \]

Another example is the most basic, constant rate TAZRP with \( f(z) = 1(z > 0) \). For this last case a proof has already been given in [7].

To prove Theorem 2.9 we need to check Assumptions 2.1 and 2.2 of Theorem 2.3. The construction of the label processes \( y(t) \) and \( z(t) \) and verification of Assumption 2.1 are done in Section 6. Assumption 2.2 requires only a few comments. The properties of the rates required in the first bullet of Assumption 2.2 are straightforward. Since \( f \) is concave and cannot be linear due to (2.37), Proposition C.1 in the Appendix implies that \( \mathcal{H}''(\varrho) < 0 \) for each \( \varrho > 0 \). Concavity of \( f \) implies bounded jump rates for the second class particle \( Q(t) \), hence a simple Poisson bound gives (2.31).

The remainder of the paper is devoted to proofs. The next two sections prove Theorem 2.3, after that we prove the Strong Law for the second class particle, and then we return to finish the proof of Theorem 2.9.

### 3. Upper bound of the main theorem

In this section we prove the upper bound of (2.32). We first give a sketch of the proof. As in Section 2.6 on microscopic concavity, we consider the second class particle \( Q(t) \) in a pair of processes \( (\omega^-, \omega) \) at density \( \varrho \). Additionally there is a positive density of other second class particles that arise from a coupling of \( (\omega^-, \omega) \) with a third process \( \eta \) at density \( \lambda \in (\varrho - \gamma_0, \varrho) \). We emphasize that the coupled \( (\eta, \omega) \) is not stationary. This is not only because we modified the marginals at the origin to \( \hat{\mu}^0 \) and \( \hat{\mu}^\lambda \) from (2.17), but more fundamentally because i.i.d. product measures are not stationary for the coupled evolution. Nevertheless, the marginal processes \( \omega \) and \( \eta \) are close enough to their stationary product distributions so that we can calculate conveniently.

The \( \omega - \eta \) second class particles are conserved during the evolution, and their current is the difference between the currents (heights) of the \( \omega \) and the \( \eta \) processes. Careful coupling makes it possible to compare \( Q(t) \) with the position of a tagged \( \omega - \eta \) second class particle \( X_0(t) \). Fluctuation bounds for \( Q(t) \) are derived through several steps: a deviation of \( Q(t) \) implies a similar deviation for \( X_0(t) \), which results in a deviation of height differences \( h^\omega - h^\eta \). The probability of this is bounded by Chebyshev’s inequality which brings in variances of the currents \( h^\xi \) and \( h^\eta \). These variances are further turned into the first moment of \( Q(t) \) essentially via (2.19) and (2.20). Now the loop is closed, as deviations of \( Q(t) \) are bounded by the centered first absolute moment of \( Q(t) \). Along the way we see that the sharpest bound is obtained with \( \lambda = \varrho - c \cdot u/\tau \) for a constant \( c \). We also mention in advance that the critical part of our estimate comes from the order of magnitude \( u \sim t^{2/3} \), thus \( \varrho - \lambda \sim t^{-1/3} \). With this choice the means for currents and second class particle velocities that we use for centering provide factors of just the right order for successful completion of the estimation.

Density \( \varrho \) is fixed. Let \( \lambda \in (\varrho, \varrho - \gamma_0) \) and apply Assumption 2.1 with constant sequences \( \varrho_i \equiv \varrho \) and \( \lambda_i \equiv \lambda \) for all \( i \in \mathbb{Z} \). Notations \( \mathbf{P}, \mathbf{E}, \text{Var}, \text{Cov} \) will refer to the coupled four-process evolution described in Assumption 2.1, while \( \mathbf{P}^\rho, \mathbf{E}^\rho, \text{Var}^\rho, \text{Cov}^\rho \) will refer to a density \( \varrho \) stationary process. Abbreviate

\[ \Psi(t) := \mathbf{E} \left[ Q(t) - \left\lfloor V^\rho t \right\rfloor \right]. \tag{3.1} \]

The requirement that \( (\omega^-, \omega) \) obey the basic coupling was included in Assumption 2.1. Consequently \( \Psi(t) \) is the \( m = 1 \) expectation of (2.32).

The following lemma does the main work towards the upper bound. We keep \( \mathcal{H}''(\varrho) \) explicitly in the estimates, because its nonvanishing is the key feature behind the \( t^{1/3} \)-fluctuations.

**Lemma 3.1.** There exist positive constants \( \alpha_1, \alpha_2, t_0 \) such that for each \( t > t_0 \) and integer \( u \) such that \( \alpha_2 \sqrt{t} < u < \alpha_1 t \),

\[ \mathbf{P}\{Q(t) > \left\lfloor V^\rho t \right\rfloor + u\} \leq C_5 \frac{t^2 \mathcal{H}''(\varrho)^2}{u^4} \{\Psi(t) + u\} + C_4 \frac{t^2}{u^3}. \tag{3.2} \]
Proof. We start with an integer $u > 0$, and write
\[
P\{Q(t) > \lceil V^0 t \rceil + u\} \leq P\{y(t) \geq k\} + \mathbf{P}\{X_k(t) \geq Q(t) > \lceil V^0 t \rceil + u\}. \tag{3.3}
\]

The event \{\(X_k(t) > \lceil V^0 t \rceil + u\)\} implies that among the \(X_m\)'s at most particles \(X_1, \ldots, X_{k-1}\) have passed the path \(s(\lceil V^0 t \rceil + u + 1/2)_{0 \leq s \leq 1}\) from right to left. Each such passing decreases \(h^0_{\lceil V^0 t \rceil + u}(t) - h^0_{\lceil V^0 t \rceil + u}(t)\) by one (recall the statement around (2.10)). Hence we can bound the probability in (3.3) by
\[
P\{y(t) \geq k\} + \mathbf{P}\{h^0_{\lceil V^0 t \rceil + u}(t) - h^0_{\lceil V^0 t \rceil + u}(t) > -k\}.
\]

We introduce two more processes: \(\eta^\text{eq}\) is a stationary process started with initial data \(\eta^i(0) = \eta_i(0)\) for \(i \neq 0\), while \(\eta^0(0)\) is \(\mu^\lambda\) distributed independently of everything. \(\omega^\text{eq}\) is a stationary process started with \(\omega^i(0) = \omega_i(0)\) for \(i \neq 0\), and \(\omega^0(0)\) is \(\mu^0\) distributed independently of everything. Include these in the basic coupling of \((\eta, \omega)\) and write
\[
h^0_{\lceil V^0 t \rceil + u}(t) - h^0_{\lceil V^0 t \rceil + u}(t) = h^\omega_{\lceil V^0 t \rceil + u}(t) - h^\omega_{\lceil V^0 t \rceil + u}(t) + h^\omega_{\lceil V^0 t \rceil + u}(t)
\]
\[
- h^\omega_{\lceil V^0 t \rceil + u}(t) - h^\omega_{\lceil V^0 t \rceil + u}(t) + h^\omega_{\lceil V^0 t \rceil + u}(t).
\]

Basic coupling implies
\[
h^\omega_{\lceil V^0 t \rceil + u}(t) - h^\omega_{\lceil V^0 t \rceil + u}(t) \leq |\omega_0(0) - \omega_0(0)| \leq |\omega_0(0)| + |\omega_0(0)| \quad \text{and}
\]
\[
h^\omega_{\lceil V^0 t \rceil + u}(t) - h^\omega_{\lceil V^0 t \rceil + u}(t) \leq |\eta^\text{eq}(0) - \eta_0(0)| \leq |\eta^\text{eq}(0)| + |\eta_0(0)|.
\]

We bound the stationary expectations using (2.16), (2.15) and Taylor’s formula. This is a crucial computation, which shows that on the characteristic position (that would be case \(u = 0\)), expectation of the height difference is only \(O(\varrho - \lambda)^2\), without constant and first-order expression of the densities.
\[
E^\omega h^\omega_{\lceil V^0 t \rceil + u}(t) - E^\lambda h^\omega_{\lceil V^0 t \rceil + u}(t)
\]
\[
= \mathcal{H}(\varrho) t - \lceil V^0 t \rceil + u)\varrho - \mathcal{H}(\lambda) t + \lceil V^0 t \rceil + u)\lambda
\]
\[
\leq t(\mathcal{H}(\varrho) - \mathcal{H}(\lambda) + \mathcal{H}'(\varrho)(\varrho - \lambda)) + u(\lambda - \varrho) + C_1
\]
\[
\leq -\frac{t}{2} \mathcal{H}''(\varrho)(\varrho - \lambda)^2 + u(\lambda - \varrho) + C_2 t(\varrho - \lambda)^3 + C_1.
\]

\(\mathcal{H}\) can be differentiated arbitrarily many times, as we show in Section C of the Appendix. Constant \(C_1\) above bounds errors from discarded integer parts. Recall that tilde stands for the centered random variable. Collecting terms we continue from (3.3) as follows.
\[
P\{Q(t) > \lceil V^0 t \rceil + u\} \leq P\{y(t) \geq k\} + \mathbf{P}\{\tilde{h}^\omega_{\lceil V^0 t \rceil + u}(t) - \tilde{h}^\omega_{\lceil V^0 t \rceil + u}(t) > -k + \frac{t}{2} \mathcal{H}''(\varrho)(\varrho - \lambda)^2 + u(\varrho - \lambda) - C_2 t(\varrho - \lambda)^3 - C_1\}
\]
\[
= P\{y(t) \geq k\} + \mathbf{P}\{\tilde{h}^\omega_{\lceil V^0 t \rceil + u}(t) - \tilde{h}^\omega_{\lceil V^0 t \rceil + u}(t) > -k + \frac{t}{2} \mathcal{H}''(\varrho)(\varrho - \lambda)^2 + \frac{u}{2}(\varrho - \lambda)\}
\]
\[
+ \mathbf{P}\{|\eta_0(0)| + |\eta^\text{eq}(0)| + |\omega_0(0)| + |\omega^\text{eq}(0)|
\]
\[
> -k + \frac{u}{2}(\varrho - \lambda) - C_2 t(\varrho - \lambda)^3 - C_1\}.
\]
From now on we use the specific assumption $\mathcal{H}''(\varrho) < 0$. We maximize the terms on the right-hand side of the probability of $\tilde{h}$'s by the choice

$$ \varrho - \lambda = \frac{-u}{2t\mathcal{H}''(\varrho)}. $$

To stay within the range of densities covered by Assumption 2.1 we must ensure that $\lambda > \varrho - \gamma_0$. So we introduce a small constant $\alpha_1 > 0$ and restrict our calculations to the case $u < \alpha_1 t$. Then

$$ P\{Q(t) > V^o t + u\} \leq P\{y(t) \geq k\} + P\{\tilde{h}_{[V^o t + u]}^\varrho(t) - \tilde{h}_{[V^o t + u]}^\varrho(t) > \frac{-u^2}{8t\mathcal{H}''(\varrho)}\} $$

$$ + P\{\eta_0(0) + |\eta_0^\varrho(0)| + |\omega_0(0)| + |\omega_0^\varrho(0)| > -k - \frac{1}{4\mathcal{H}''(\varrho)} \cdot \frac{u^2}{t} + \frac{C_2}{\mathcal{H}''(\varrho)^2} \cdot \frac{u^3}{t^2} - C_1\}. $$

Now we set

$$ k = \left\lfloor -\frac{1}{8\mathcal{H}''(\varrho)} \cdot \frac{u^2}{t} \right\rfloor, $$

and assume $\alpha_2\sqrt{t} < u < \alpha_1 t$ for a possibly smaller $\alpha_1$ and a large enough $\alpha_2$. That allows us to unify the right-hand side of the inequality in the last line. Thus for all large $u$ and $t$ with $\alpha_2\sqrt{t} < u < \alpha_1 t$

$$ P\{Q(t) > V^o t + u\} \leq P\{y(t) \geq \left\lfloor -\frac{1}{8\mathcal{H}''(\varrho)} \cdot \frac{u^2}{t} \right\rfloor\} + P\{\tilde{h}_{[V^o t + u]}^\varrho(t) - \tilde{h}_{[V^o t + u]}^\varrho(t) > \frac{-u^2}{8t\mathcal{H}''(\varrho)}\} $$

$$ + P\{\eta_0(0) + |\eta_0^\varrho(0)| + |\omega_0(0)| + |\omega_0^\varrho(0)| > C_3 \frac{u^2}{t}\}. $$

Assumption (2.28) allows us to bound the first probability on the right by $C_4 t^2/u^3$ (take $\gamma = \varrho - \lambda$). Apply Chebyshev’s inequality on the second line and Markov’s inequality on the third one:

$$ P\{Q(t) > V^o t + u\} \leq \frac{64}{u^4} \frac{t^2\mathcal{H}''(\varrho)^2}{\mathcal{H}''(\varrho)^2} \cdot \text{Var}(h_{[V^o t + u]}^\varrho(t) - h_{[V^o t + u]}^\varrho(t)) + \frac{C_3}{u^2} + \frac{C_4}{u^3} $$

$$ \leq 128 \frac{t^2\mathcal{H}''(\varrho)^2}{u^4} \left\{\text{Var}(h_{[V^o t + u]}^\varrho(t)) + \text{Var}(h_{[V^o t + u]}^\varrho(t))\right\} + \frac{C_4}{u^3}. $$

The term $C_3 t/u^2$ was subsumed under $C_4 t^2/u^3$ due to the condition $u < \alpha_1 t$. The variances here are taken under the stationary distributions of the processes $\eta^\varrho$ and $\omega^\varrho$. That allows us to apply (2.19), whose right-hand side takes us back to the four-process coupling under measure $P$. Recall (3.1).

$$ P\{Q(t) > V^o t + u\} \leq C_5 \frac{t^2\mathcal{H}''(\varrho)^2}{u^4} \left\{E|Q(t) - [V^o t] - u| + E|Q^o(t) - [V^o t] - u|\right\} + \frac{C_4}{u^3} $$

$$ \leq C_5 \frac{t^2\mathcal{H}''(\varrho)^2}{u^4} \left\{E|Q(t) - [V^o t]| + E|Q^o(t) - [V^o t]| + 2u\right\} + \frac{C_4}{u^3} $$

$$ = C_5 \frac{t^2\mathcal{H}''(\varrho)^2}{u^4} \left\{\psi(t) + 2u + E|Q^o(t) - [V^o t]|\right\} + \frac{C_4}{u^3}. $$

The variable $Q^o(t)$ above is the location of a single discrepancy between the process $\bar{\eta}$ and one started initially with $\eta^+ = \eta(0) + \delta_0$. 
It remains to relate $E|Q^n(t) - [V^{\varrho}t]|$ to $\Psi(t)$. This is where part (2.30) of Assumption 2.1 is a key point. Compute now in the four-process coupling of $\eta, \eta^+, \omega^-, \omega$ described in Assumption 2.1. Use (2.30) and Taylor expansion of $H$ again:

$$E|Q^n(t) - [V^{\varrho}t]| \leq E(Q^n(t) - Q(t)) + \Psi(t)$$

$$= (H'(\lambda) - H'(\varrho))t + \Psi(t)$$

$$\leq H''(\varrho) \cdot (\lambda - \varrho) + C_6(\varrho - \lambda)^2t + \Psi(t)$$

$$= \frac{u}{2} + C_6\frac{u^2}{t} + \Psi(t) \leq \left(\frac{1}{4} + C_6 \alpha^1\right)u + \Psi(t). \quad (3.4)$$

The last inequality used $u < \alpha_1 t$. Substitute this back into the previous display and rename constants. This finishes the proof of (3.2) and completes the lemma. □

Completely analogous arguments lead to the same upper bound for the lower tail of $Q(t)$, and together we get the following bound on the tail of the absolute deviation, still for $\alpha_2 \sqrt{t} < u < \alpha_1 t$:

$$P(|Q(t) - [V^{\varrho}t]| > u) \leq C_5 \frac{t^2H''(\varrho)^2}{u^3} \{\Psi(t) + u\} + C_4 \frac{t^2}{u^3}. \quad (3.5)$$

Next we relax the restriction to integral $u$ and the upper limit on it:

**Lemma 3.2.** There are positive constants $\alpha_2, t_0$ such that for all $t > t_0$ and all real $u > \alpha_2 \sqrt{t}$,

$$P(|Q(t) - [V^{\varrho}t]| > u) \leq C_5 \frac{t^2H''(\varrho)^2}{u^3} \{\Psi(t) + u\} + C_4 \frac{t^2}{u^3}. \quad (3.5)$$

**Proof.** Any $u \geq 1$ is less than twice its integer part. Hence by simply increasing the constants $C_i$, for all large $t$ and all real $u \in (\alpha_2 \sqrt{t}, \alpha_1 t)$,

$$P(|Q(t) - [V^{\varrho}t]| > u) \leq C_5 \frac{t^2H''(\varrho)^2}{u^3} \{\Psi(t) + u\} + C_4 \frac{t^2}{u^3}. \quad (3.5)$$

Recall (2.31). When $\alpha_1 < \alpha_0 + 2|V^{\varrho}| + 2$, assume $\alpha_1 t \leq u < (\alpha_0 + 2|V^{\varrho}| + 2)t$. Then $\alpha_2 \sqrt{t} < u \cdot \alpha_1 / (\alpha_0 + 2|V^{\varrho}| + 2) < \alpha_1 t$ for large enough $t$, and (3.5) still holds for $u$ replaced by $u \cdot \alpha_1 / (\alpha_0 + 2|V^{\varrho}| + 2)$:

$$P(|Q(t) - [V^{\varrho}t]| > u) \leq P\left(|Q(t) - [V^{\varrho}t]| > u \cdot \frac{\alpha_1}{\alpha_0 + 2|V^{\varrho}| + 2}\right)$$

$$\leq C_5 \frac{t^2H''(\varrho)^2}{u^3} \{\Psi(t) + u\} + C_4 \frac{t^2}{u^3}$$

via modifying the constants by factors of $\alpha_1 / (\alpha_0 + 2|V^{\varrho}| + 2)$.

Finally, the case $u \geq (\alpha_0 + 2|V^{\varrho}| + 2)t$ will not be relevant for us hence, due to the fact that $u - |[V^{\varrho}t]| > \alpha_0 t$, we can use (2.31):

$$P(|Q(t) - [V^{\varrho}t]| > u) \leq P(|Q(t)| > u - |[V^{\varrho}t]|) \leq C_7 \frac{t^2}{u^3} \leq C_8 \frac{t^2}{u^3}.$$
Fluctuation bounds

Proof of the upper bound of Theorem 2.3. We now fix $r > 0$, $1 \leq m < 3$, and write
\[
E(\|Q(t) - [V^0 t]\|^m) = \int_0^\infty P(\|Q(t) - [V^0 t]\|^m > v) \, dv
\]
\[
\leq r^m t^{(2/3)m} + m \int_{r t^{2/3}}^\infty \left( C_5 \frac{t^2 H''(\eta)}{u^4} \{\Psi(t) + u\} + C_4 \frac{t^2}{u^3}\right) u^{-m-1} \, du
\]
\[
= r^m t^{(2/3)m} + \frac{mC_5 H''(\eta)}{4-m} r^{m-4} t^{(2/3)m-2/3} \Psi(t) + \frac{mC_5 H''(\eta)}{3-m} r^{m-3} t^{(2/3)m}.
\]
First choose $m = 1$ and $r$ large enough to get $\Psi(t) \leq C t^{2/3}$. Then insert this bound back into the last line of the display to get the bound for general $1 \leq m < 3$. □

4. Lower bound of the main theorem

We begin again with an informal preview of the proof. The proof of the lower bound of (2.32) uses similar ideas as the upper bound proof but with an extra twist. The starting point is a pair of processes $(\xi, \xi^+)$ at density $\lambda$ with one second class particle $Q^{(-n)}(t)$ between them started from position $-n$. Coupled to this pair is a process $\xi \geq \xi^+$ that is mostly in density $\varrho > \lambda$, except that we set $\xi = \xi^+$ on the interval $-n + 1, -n + 2, \ldots, 0$. The position $-n$ is chosen so that the $\lambda$-characteristic $-n + V^\lambda t$ started from $-n$ satisfies $V^\lambda t - n = V^\varrho t - u$ for large enough $u > 0$ so that the upper bound makes the event $Q^{(-n)}(t) < V^\varrho t$ likely. Reasoning as we did for the upper bound, from this event we can deduce an inequality for the current difference between the $\xi$ and the $\xi^+$ processes. In order to turn this inequality into a deviation that can be bounded by Chebyshev’s inequality as in the upper bound proof, we change the $\xi$ process into a stationary process by introducing the appropriate Radon–Nikodym density for the initial distribution. As in the upper bound proof, the useful perturbation of density is of the order $\varrho - \lambda = bt^{-1/3}$.

Density $\varrho$ is fixed again, and $\lambda \in (\varrho - \gamma_0, \varrho)$ is a varying auxiliary density. We let the jointly defined four processes $(\eta, \eta^+, \omega^-, \omega)$ be exactly as defined in the upper bound proof of Section 3, namely, as given by Assumption 2.1 with constant densities $\lambda_i \equiv \lambda$ and $\varrho_i \equiv \varrho$. The initial distribution of $(\eta, \omega)$ is $\hat{\mu}^\lambda \varrho$ of (2.24). Two second class particles start from the origin: $Q^\varrho$ between processes $\eta$ and $\eta^+$, and $Q$ between processes $\omega^-$ and $\omega$. The quantity of primary interest is abbreviated, as before, by $\Psi(t) = E[Q(t) - [V^\varrho t]]$.

To prove the lower bound of (2.32) it suffices, by Jensen’s inequality, to prove the case $m = 1$. This means showing that $\Psi(t) \geq C t^{2/3}$ for large $t$ and a constant $C > 0$.

4.1. Perturbing a segment initially

For this proof we need to introduce another coupled system and invoke Assumption 2.1 once more. By concavity of the flux characteristic speeds $V^\varrho = H'(\varrho)$ and $V^\lambda = H'(\lambda)$ satisfy $V^\varrho \leq V^\lambda$. Throughout this section $u > 0$ denotes a positive integer, and
\[
n = \left[ V^\lambda t \right] - \left[ V^\varrho t \right] + u.
\]
Recall definitions (2.22) and (2.23) of the single-site coupling measures. Let $(\xi(\cdot), \xi^+(\cdot))$ be a pair of processes that obeys the basic coupling, and whose initial distribution is the product measure
\[
\left( \bigotimes_{i < -n} \mu^\lambda \varrho \right) \left( \bigotimes_{i = -n} \hat{\mu}^\lambda \varrho \right) \left( \bigotimes_{-n < i \leq 0} \mu^\lambda \varrho \right) \left( \bigotimes_{0 < i} \mu^\lambda \varrho \right).
\]
This initial measure complies with the pattern in (2.24), but translated $n$ sites to the left so that $\hat{\mu}^\lambda \varrho$ is the distribution at site $-n$ instead of the origin. A few points about this initial state: $\xi(0)$ has the stationary density-$\lambda$ product distribution except at site $-n$ where it is $\hat{\mu}^\lambda$-distributed. $\xi^+(0)$ has the product distribution with marginals $\mu^\varrho$, except at sites
where the parameter \( \varrho \) switches to \( \lambda \), and at site \(-n\) where it has distribution \( \hat{\mu}^0 + 1 \). At sites \(-n < i \leq 0\) \( \mu^{i,\lambda} \) forces \( \xi_i(0) = \xi_i \).

We add a second class particle to the process \( \tilde{\xi}(\cdot) \), start it at site \(-n\) and denote its position at time \( t \) by \( Q^{(-n)}(t) \).

Let \( \tilde{\xi}^+(t) := \tilde{\xi}(t) + \delta_{Q^{(-n)}(t)} \).

As described in Section 2.6 the \( \tilde{\xi} - \xi \) second class particles are labeled and their ordered positions denoted by \( \{X_m(t)\} \). The labeling is chosen to satisfy initially

\[
\cdots \leq X_{-1}(0) \leq X_0(0) = -n < 0 < X_1(0) \leq X_2(0) \leq \cdots .
\]

Thus initially \( X_0(0) = -n = Q^{(-n)}(0) \). We invoke Assumption 2.1 to have a label process \( z(t) \) with tail bound (2.29) such that \( Q^{(-n)}(t) = X_{z(t)}(t) \). (Here \( \tilde{\xi} \) plays the role of \( \eta \) and \( \xi \) plays the role of \( \omega \) of Assumption 2.1.)

As before, the heights (or currents, recall (2.5)) of the processes \( \tilde{\xi}(\cdot) \) and \( \xi(\cdot) \) are denoted by \( h^\xi_{[V_t]} \) and \( h^\xi_{[V_t]} \), respectively. The first observation is that \( Q^{(-n)} \) gives one-sided control over the difference of these currents.

**Lemma 4.1.** For any \( i \in \mathbb{Z} \)

\[
Q^{(-n)}(t) \leq i \quad \text{implies} \quad h^\xi_i(t) - h^\xi_i(t) \leq -z(t).
\]

**Proof.** Recall again, from (2.5) and the statement around (2.10), that the height difference \( h^\xi_i(t) - h^\xi_i(t) \) equals the net number of second class particle passings of the path \( (si + 1/2)_0 \leq s \leq 1 \) from left to right. That is, each left-to-right passing increases \( h^\xi_i(t) - h^\xi_i(t) \) while each right-to-left passing decreases it.

Suppose \( z(t) \leq 0 \). Then (4.1) and \( X_{z(t)}(t) = Q^{(-n)}(t) \leq i \) imply that only those second class particles with labels \( z(t) + 1, z(t) + 2, \ldots, 0 \) could have crossed the path \( (si + 1/2)_0 \leq s \leq 1 \) from left to right. The claim follows.

If \( z(t) > 0 \), then \( X_{z(t)}(t) = Q^{(-n)}(t) \leq i \) implies that at least those second class particles with labels \( 1, 2, \ldots, z(t) \) have crossed the path \( (si + 1/2)_0 \leq s \leq 1 \) from right to left. Again the claim follows. \(\square\)

Let \( \tilde{\omega}(\cdot) \) be a process started from the product distribution \( (\bigotimes_{i \neq -n} \mu^0) \otimes (\hat{\mu}^0 + 1) \). The next lemma compares the initial distributions of \( \xi \) and \( \tilde{\omega} \). No coupling of \( \xi \) and \( \tilde{\omega} \) is proposed or required.

**Lemma 4.2.** There exist constants \( \gamma = \gamma(\varrho) > 0 \) and \( C_1(\varrho) < \infty \) such that for all \( \lambda \in (\varrho - \gamma, \varrho) \) and all events \( A \) the following inequality holds:

\[
P[\xi \in A] \leq P[\tilde{\omega} \in A]^{1/2} \cdot \exp\{C_1(\varrho)n(\varrho - \lambda)^2\}.
\]

**Proof.** We use the Cauchy–Schwarz inequality below to perform a change of measure on the distribution of the \( \xi \) process. First we condition on the initial \( \tilde{\xi} \)-configuration at sites \( \{-n + 1, -n + 2, \ldots, -1, 0\} \).

\[
P[\xi \in A] = \sum_{z_{-n+1}, \ldots, z_1, z_0} P[\xi \in A|z_{-n+1}(0) = z_{-n+1}, \ldots, z_0(0) = z_0] \times \left[ \prod_{i=-n+1}^{0} \mu^0(z_i) \right]^{1/2} \prod_{i=-n+1}^{0} \frac{\mu^\lambda(z_i)}{[\mu^0(z_i)]^{1/2}}
\]

\[
\leq \left[ \sum_{z_{-n+1}, \ldots, z_0} P[\xi \in A|z_{-n+1}(0) = z_{-n+1}, \ldots, z_0(0) = z_0] \right]^{2} \prod_{i=-n+1}^{0} \mu^0(z_i)
\]

\[
\times \sum_{z_{-n+1}, \ldots, z_0} \prod_{i=-n+1}^{0} \frac{[\mu^\lambda(z_i)]^2}{\mu^0(z_i)} \right]^{1/2}
\]

\[
\leq \left[ \sum_{z_{-n+1}, \ldots, z_0} P[\xi \in A|z_{-n+1}(0) = z_{-n+1}, \ldots, z_0(0) = z_0] \prod_{i=-n+1}^{0} \mu^\lambda(z_i) \right]^{1/2}
\]
The last inequality came from dropping the square. For the last equality note that the distributions of the initial configurations \( \{\hat{\omega}(0)\} \) and \( \{\zeta(0)\} \) are product-form and agree outside the interval \( \{-n+1,-n+2,\ldots,-1,0\} \). Thus conditioned on the initial values in \( \{-n+1,-n+2,\ldots,-1,0\} \) these processes have identical conditional probabilities.

To complete the proof we bound the last factor in brackets. Recall formulas (2.12) and (2.13) for the state sum and the site-marginals. Without the power 1/2 the factor in brackets equals

\[
\sum_{z_{-n+1},\ldots,z_0} \left( \frac{Z(\theta(q))}{Z(\theta(\lambda))^2} \right)^n \prod_{i=-n+1}^0 \frac{e^{(2\theta(\lambda)-\theta(q))z_i}}{f(z_i)!} = \left( \frac{Z(2\theta(\lambda)-\theta(q))Z(\theta(q))}{Z(\theta(\lambda))^2} \right)^n.
\]

In the Appendix we show that \( \log Z(\theta) \) and \( \theta(q) \) are infinitely differentiable. Let \( \varepsilon = \theta(q) - \theta(\lambda) \). By local Lipschitz continuity of the function \( \theta(q) \), the interval \( (\theta(\lambda) - \varepsilon, \theta(\lambda) + \varepsilon) \) is in \( (\hat{\theta}, \bar{\theta}) \) with a small enough choice of \( \gamma \). There exists some \( \bar{\theta} \in (\theta(\lambda) - \varepsilon, \theta(\lambda) + \varepsilon) \) such that

\[
\log \left( \frac{Z(2\theta(\lambda)-\theta(q))Z(\theta(q))}{Z(\theta(\lambda))^2} \right) = \log Z(\theta(\lambda) - \varepsilon) + \log Z(\theta(\lambda) + \varepsilon) - 2\log Z(\theta(\lambda))
\]

\[
= \frac{1}{2} \frac{d^2}{d\theta^2} \log Z(\theta) \varepsilon^2 \leq C_1(q) \cdot (q - \lambda)^2.
\]

Thus we get the bound

\[
\left( \frac{Z(2\theta(\lambda)-\theta(q))Z(\theta(q))}{Z(\theta(\lambda))^2} \right)^n \leq \exp \{ C_1(q) \cdot n(q - \lambda)^2 \}.
\]

\[\square\]

4.2. Completion of the proof of the lower bound

The gist of the proof is to get upper bounds on the complementary probabilities \( P(Q^{(n)}(t) > [V^\theta t]) \) and \( P(Q^{(n)}(t) \leq [V^\theta t]) \). As stated \( u \) is an arbitrary but positive integer and \( n = [V^\lambda t] - [V^\theta t] + u \).

**Lemma 4.3.**

\[
P\{Q^{(n)}(t) > [V^\theta t]\} \leq \frac{\Psi(t)}{u} + \frac{C_2 t(q - \lambda)}{u} + \frac{2}{u}.
\]

**Proof.** Distributionwise the system \( (\xi, \xi^+, Q^{(n)}) \) is a translate of \( (\eta, \eta^+, Q^n) \), and so

\[
P\{Q^{(n)}(t) > [V^\theta t]\} = P\{Q^{(n)}(t) + n - [V^\lambda t] > u\}
\]

\[
= P\{Q^n(t) - [V^\lambda t] > u\} \leq \frac{E(|Q^n(t) - [V^\lambda t]|)}{u} \leq \frac{E(|Q^n(t) - Q(t)|)}{u} + \frac{E(|Q(t) - [V^\theta t]|)}{u} + \frac{[V^\lambda t] - [V^\theta t]}{u}.
\]

Use (2.30) precisely as was done in (3.4) to conclude that the first term equals

\[
u^{-1}E(Q^n(t) - Q(t)) = u^{-1}t(H'(\lambda) - H'(q)) = -u^{-1}H''(v)t(q - \lambda)
\]
for some \( v \in (\lambda, \rho) \). The second term is \( \Psi(t)/u \), and the third term is similarly estimated by \( -u^{-1}H''(\xi) + 2/u \), the last part coming from discarded integer parts. Setting \( C_2 := 2 \max_{v \in [\xi, \rho]} -H''(v) \) finishes the proof. \( \square \)

Notice that \( H''(\rho) < 0 \) was crucial in the previous proof, as well as in the following lemma, and the final proof thereafter. These points show where the proof fails for symmetric systems — recall that these would have lower-order current fluctuations on the characteristics.

**Lemma 4.4.** Let \( K = K(t) \) satisfy \( 0 < K < -\frac{1}{2} t H''(\rho) (\rho - \lambda)^2 \). Then for small enough \( \gamma > 0 \), large enough \( t \), and \( \lambda \in (\rho - \gamma, \rho) \),

\[
\mathbb{P}\left\{ Q^{(n)}(t) \leq [V^n t] \right\} \leq \frac{\text{Var}^d(\eta_0))^{1/2} \Psi(t)^{1/2}}{-(1/3)t H''(\rho) (\rho - \lambda)^2 - K} \cdot e^{C_1 n(\rho - \lambda)^2} \\
+ \frac{C_4}{-t H''(\rho) (\rho - \lambda)^2 - C_3 t (\rho - \lambda)^3 - \rho} \cdot e^{C_1 n(\rho - \lambda)^2} \\
+ \frac{\text{Var}^d(\eta_0) \Psi(t)}{K^2/4} + \frac{C_6 t (\rho - \lambda)}{K^2} + \frac{C_5}{K - 4|\lambda|} + C t^{k-1} \gamma^{2k-3} K^{-k}.
\]

**Proof.** Lemma 4.1 leads to

\[
\mathbb{P}\left\{ Q^{(n)}(t) \leq [V^n t] \right\} \leq \mathbb{P}\left\{ h_{\tilde{\omega} V}^\xi(t) - h_{\tilde{\omega} V}^\xi(t) \leq -z(t) \right\}
\leq \mathbb{P}\{-z(t) \geq K/4\} \quad (4.2)
+ \mathbb{P}\left\{ h_{\tilde{\omega} V}^\xi(t) \leq K + t(H(\lambda) - \lambda H'(\rho)) \right\} \quad (4.3)
+ \mathbb{P}\left\{ h_{\tilde{\omega} V}^\xi(t) > 3K/4 + t(H(\lambda) - \lambda H'(\rho)) \right\}. \quad (4.4)
\]

To bound (4.2) we use the assumed distribution bound (2.28) on \( z(t) \) and get

\[
\mathbb{P}\{-z(t) \geq K/4\} \leq C t^{k-1} \gamma^{2k-3} K^{-k}.
\]

Apply Lemma 4.2 to line (4.3) to bound it by the probability of the process \( \tilde{\omega} \):

\[
(4.3) \leq \left[ \mathbb{P}\left\{ h_{\tilde{\omega} V}^\xi(t) \leq K + t(H(\lambda) - \lambda H'(\rho)) \right\} \right]^{1/2} \cdot e^{C_1 n(\rho - \lambda)^2}.
\]

As in the proof of Lemma 3.1 we compare with a coupled stationary processes to get precise bounds:

\[
h_{\tilde{\omega} V}^\xi(t) = \tilde{h}_{\tilde{\omega} V}^\xi(t) + \left[ h_{\tilde{\omega} V}^\xi(t) - h_{\tilde{\omega} V}^\xi(t) \right]
+ \left[ \mathbb{E} h_{\tilde{\omega} V}^\xi(t) - t(H(\rho) - \rho H'(\rho)) \right] + t(H(\rho) - \rho H'(\rho))
\geq \tilde{h}_{\tilde{\omega} V}^\xi(t) - |\tilde{\omega}_{-n}(0)| - |\omega_{-n}^\xi(0)| - |\rho| + t(H(\rho) - \rho H'(\rho)).
\]

After the equality sign, the absolute value of the first term in brackets is not larger than \( |\tilde{\omega}_{-n}(0) - \omega_{-n}^\xi(0)| \leq |\tilde{\omega}_{-n}(0)| + |\omega_{-n}^\xi(0)| \). The second term in brackets is between \( -|\rho| \) and \( |\rho| \) due to the integer part in \([V^n t]\). Consequently

\[
h_{\tilde{\omega} V}^\xi(t) \leq K + t(H(\lambda) - \lambda H'(\rho))
\]

implies

\[
\tilde{h}_{\tilde{\omega} V}^\xi(t) - |\tilde{\omega}_{-n}(0)| - |\omega_{-n}^\xi(0)| \leq K + t\left[ H(\lambda) - H(\rho) + H'(\rho)(\rho - \lambda) \right] + |\rho|
\leq K + \frac{1}{2} t H''(\rho)(\rho - \lambda)^2 + C_3 t (\rho - \lambda)^3 + |\rho|.
\]
Then, we cut the event into two parts according to the value of $|\tilde{\omega}_{-n}(0)| + |\omega_{-n}^{\text{eq}}(0)|$ and we use (2.19) to bound the variance of $\text{Var}[h_{\{\text{Vert}l\}}^{\text{eq}}(t)]$ by the function $\Psi(t)$.

\begin{equation}
(4.3) \leq \left[ \mathbf{P}\left\{ \tilde{\omega}_{\text{Vert}l}^{\text{eq}}(t) \leq K + \frac{1}{3} t H''(\varrho)(\varrho - \lambda)^2 \right\} \right]^{1/2} \cdot C_{1n}(\varrho - \lambda)^2
\end{equation}

\begin{align*}
&+ \left[ \mathbf{P}\left\{ |\tilde{\omega}_{-n}(0)| + |\omega_{-n}^{\text{eq}}(0)| > \frac{1}{6} t H''(\varrho)(\varrho - \lambda)^2 - C_3 t(\varrho - \lambda)^3 - |\varrho| \right\} \right]^{1/2} \cdot C_{1n}(\varrho - \lambda)^2 \\
&\leq \frac{\text{Var}^G(h_{\text{Vert}l}^{\text{eq}}(t))^{1/2}}{-\sigma^2} \cdot C_{1n}(\varrho - \lambda)^2 + \frac{[\mathbf{E}(|\tilde{\omega}_{-n}(0)| + |\omega_{-n}^{\text{eq}}(0)|)^2]^{1/2}}{-(1/3) t H''(\varrho)(\varrho - \lambda)^2} \cdot C_4 \\
&\leq \frac{\text{Var}^G(\omega_0)^{1/2} \psi(t)^{1/2}}{-\sigma^2} \cdot C_{1n}(\varrho - \lambda)^2 + \frac{[\mathbf{E}(|\tilde{\omega}_{-n}(0)| + |\omega_{-n}^{\text{eq}}(0)|)^2]^{1/2}}{-(1/3) t H''(\varrho)(\varrho - \lambda)^2} \cdot C_4.
\end{align*}

Now we turn to (4.4). To reduce $h_{\{\text{Vert}l\}}^{\xi}$ to the current of the density-$\lambda$ equilibrium process $h_{\{\text{Vert}l\}}^{\text{eq}}$ and to get rid of the integer part errors we argue as before.

\[ h_{\{\text{Vert}l\}}^{\xi} = h_{\{\text{Vert}l\}}^{\text{eq}} + [h_{\{\text{Vert}l\}}^{\xi} - h_{\{\text{Vert}l\}}^{\text{eq}}] + [\mathbf{E}_t h_{\{\text{Vert}l\}}^{\text{eq}} - t(H(\lambda) - \lambda H'(\varrho)) + t(H(\lambda) - \lambda H'(\varrho)). \]

$h_{\{\text{Vert}l\}}^{\xi}(t)$ differs by at most $|\xi_{-n}(0) - \eta_{-n}^{\text{eq}}(0)|$ from $h_{\{\text{Vert}l\}}^{\text{eq}}(t)$. Taking integer parts again into account, giving another error term $|\varrho|$, line (4.4) is bounded from above by

\[ \mathbf{P}\left\{ |\tilde{\omega}_{-n}(0)| + |\omega_{-n}^{\text{eq}}(0)| + |\varrho| \geq 3K/4 \right\}. \]

Then, we cut the event into two parts and use Markov’s inequality in the second one:

\begin{equation}
(4.4) \leq \mathbf{P}\left\{ \tilde{\omega}_{\text{Vert}l}^{\text{eq}}(t) \geq K/2 \right\} + \mathbf{P}\left\{ |\xi_{-n}(0)| + |\eta_{-n}^{\text{eq}}(0)| > K/4 - |\varrho| \right\} \leq \frac{\text{Var}^G(h_{\{\text{Vert}l\}}^{\text{eq}})}{K^2/4} + \frac{C_5}{K - 4|\varrho|}.
\end{equation}

We can use (2.19) again to continue with

\begin{equation}
(4.4) \leq \frac{\text{Var}^G(\xi_0)\mathbf{E}(|Q^\varrho(t) - [Q^\varrho_t]|)}{K^2/4} + \frac{C_5}{K - 4|\varrho|}.
\end{equation}

Repeating the first two steps of calculation (3.4) we can write

\[ \mathbf{E}(|Q^\varrho(t) - [Q^\varrho_t]|) \leq \mathbf{E}(|Q^\varrho(t) - Q(t)|) + \mathbf{E}(|Q(t) - [Q^\varrho_t]|) \leq C t(\varrho - \lambda) + \Psi(t). \]

So, we finally get

\begin{equation}
(4.4) \leq \frac{\text{Var}^G(\eta_0)\psi(t)}{K^2/4} + \frac{C_6 t(\varrho - \lambda)}{K^2} + \frac{C_5}{K - 4|\varrho|}.
\end{equation}

\[ \square \]

**Proof of the lower bound of Theorem 2.3.** As observed in the beginning of this section, it suffices to prove that

\[ \liminf_{t \to \infty} t^{-2/3} \psi(t) > 0. \]

In the last two lemmas take

\[ u = \left\lfloor bt^{2/3} \right\rfloor, \quad \varrho - \lambda = bt^{1/3} \quad \text{and} \quad K = bt^{1/3}. \]
where $h$ and $b$ are large, in particular $b$ large enough to have $b < -\frac{1}{3} H''(\varrho)b^2$ so that $K$ satisfies the assumption of Lemma 4.4. Then
\[
 n = \lfloor V^\lambda t \rfloor - \lfloor V^\varrho t \rfloor + u \leq (H'(\lambda) - H'(\varrho))t + u + 2 = -H''(\varrho)(\varrho - \lambda)t + u + C_7 t(\varrho - \lambda)^2 + 2 \\
 \leq (-H''(\varrho)b + h)t^{2/3} + C_7 b^2 t^{1/3} + 3 \leq C_8 t^{2/3}
\]
for large enough $t$. With these definitions we can simplify the outcomes of Lemma 4.3 and Lemma 4.4 to the inequalities
\[
 P\{Q^{(-n)}(t) > \lfloor V^\varrho t \rfloor \} \leq C \frac{\Psi(t)}{t^{2/3}} + \frac{C_2 b}{h} + \frac{2}{ht^{2/3}} \tag{4.6}
\]
and
\[
 P\{Q^{(-n)}(t) \leq \lfloor V^\varrho t \rfloor \} \leq C \left( \frac{\Psi(t)}{t^{2/3}} \right)^{1/2} + C \frac{\Psi(t)}{t^{2/3}} + \frac{C_6}{b} + \frac{C_5}{bt^{1/3}} + C b^{\kappa - 3}. \tag{4.7}
\]
The new constant $C$ depends on $b$ and $h$.

The lower bound (4.5) now follows because the left-hand sides of (4.6) and (4.7) add up to 1 for each fixed $t$, while we can fix $b$ large enough and then $h$ large enough so that $C_2 b/h + C_6/b + C b^{\kappa - 3} < 1$ (recall $\kappa < 3$). Then $t^{-2/3} \Psi(t)$ must have a positive lower bound for all large enough $t$. This completes the proof of Theorem 2.3.

5. Strong Law of Large Numbers for the second class particle

This section proves the Strong Law of Large Numbers (Corollary 2.5). We assume that the jump rates of the second class particle are bounded, i.e.,
\[
 p(y + 1, z) - p(y, z), \quad p(y, z) - p(y, z + 1), \quad q(y, z + 1) - q(y, z), \quad q(y, z) - q(y + 1, z) \leq C \forall \omega_{\min} \leq y, z < \omega_{\max}. \tag{5.1}
\]
This means that the second class particle has at most rate $C$ to jump to the right and to the left, respectively, implying that starting at any time $t$, it can be bounded by rate $C$ Poisson processes that start from its position $Q(t)$.

**Proof of Corollary 2.5.** Let $\varepsilon, \delta > 0$. Define the events
\[
 A_n := \left\{ \left| \frac{Q(n^{1+\delta})}{n^{1+\delta}} - V^\varrho \right| > \varepsilon/2 \right\}
\]
for $n \in \mathbb{N}$. Then, Markov’s inequality and Theorem 2.3 imply, for $1 \leq m < 3$ and large $n$,
\[
 P\{A_n\} = P\left\{ \left| \frac{Q(n^{1+\delta})}{n^{1+\delta}} - V^\varrho n^{1+\delta} \right|^m > (\varepsilon/2)^m n^{(1+\delta)m} \right\} \\
 \leq \frac{1}{(\varepsilon/2)^m n^{(1+\delta)m}} \cdot E\left[ \left| \frac{Q(n^{1+\delta})}{n^{1+\delta}} - V^\varrho n^{1+\delta} \right|^m \right] \\
 \leq \frac{C_1}{(3 - m)(\varepsilon/2)^m} \cdot \frac{1}{n^{m(1+\delta)/3}},
\]
which is summable if $(1 + \delta)m > 3$. Here $\delta$ can be chosen arbitrarily small by taking $m$ close to 3. By the Borel–Cantelli Lemma there exists a.s. $n_0 \in \mathbb{N}$ such that
\[
 \forall n \geq n_0 \left| \frac{Q(n^{1+\delta})}{n^{1+\delta}} - V^\varrho \right| < \varepsilon/2. \tag{5.2}
\]
Using this we show that a.s. there exists \( n_1 \in \mathbb{N} \) such that
\[
\left| \frac{Q(t)}{t} - V^0 \right| < \varepsilon \quad \text{for all } t \geq n_1^{(1+\delta)}.
\] (5.3)

Let \( n \geq n_0 \) and suppose there exists some \( t \in [n^{1+\delta}, (n + 1)^{1+\delta}) \) such that (5.3) fails: \( |Q(t) - V^0 t| \geq \varepsilon t \). Together with (5.2) we have, if \( n \) is large,
\[
|Q(t) - Q(n^{1+\delta})| \geq |Q(t) - V^0 t| - |Q(n^{1+\delta}) - V^0 n^{1+\delta}| - |V^0 t - V^0 n^{1+\delta}|
\geq \varepsilon t - \varepsilon/2 \cdot n^{1+\delta} - |V^0|(t - n^{1+\delta}) \geq \frac{\varepsilon}{4} n^{1+\delta}.
\] (5.4)

The jump rates (5.1) (both left and right) of \( Q \) are bounded by \( C \). However, the event (5.4) implies that at least \( \lfloor \frac{\varepsilon}{4} n^{1+\delta} \rfloor \) many left jumps or this many right jumps happen in the time interval \( [n^{1+\delta}, (n + 1)^{1+\delta}) \). For large \( n \), the length of this interval is smaller than \( 2(1+\delta)n^5 \). Let \( N(\cdot) \) be a rate \( C \) Poisson process. Then for large \( n \) the probability of the event (5.4) is bounded from above by
\[
2\mathbb{P}\left\{ N\left(2(1+\delta)n^5\right) \geq \frac{\varepsilon}{4} n^{1+\delta}\right\} \leq 2\mathbb{P}\left\{ e^{N(2(1+\delta)n^5)} \geq e^{\varepsilon/4 n^{1+\delta}}\right\}
\leq 2e^{-\varepsilon/4 n^{1+\delta}} \mathbb{E}[e^{N(2(1+\delta)n^5)}]
= 2e^{-\varepsilon/4 n^{1+\delta}} e^{(e-1)2C(1+\delta)n^5}.
\]
This quantity is summable over \( n \), so the Borel–Cantelli Lemma implies that a.s. (5.3) holds eventually. Since this is true for each \( \varepsilon > 0 \), the Strong Law of Large Numbers holds. \( \square \)

6. Microscopic concavity for a class of totally asymmetric concave exponential zero range processes

In this section we verify that Assumption 2.1 can be satisfied under Assumption 2.8, and thereby complete the proof of Theorem 2.9.

The task is to construct the processes \( y(t) \) and \( z(t) \) with the requisite properties. First let the processes \( (\eta(t), \omega(t)) \) evolve in the basic coupling so that \( \eta_i(t) \leq \omega_i(t) \) for all \( i \in \mathbb{Z} \) and \( t \geq 0 \). We consider as a background process this pair with the labeled and ordered \( \omega - \eta \) second class particles \( \cdots \leq X_{-2}(t) \leq X_{-1}(t) \leq X_0(t) \leq X_1(t) \leq X_2(t) \leq \cdots \).

At each time \( t \geq 0 \) this background induces a partition \( \{\mathcal{M}_i(t)\} \) of the label space \( \mathbb{Z} \) into intervals indexed by sites \( i \in \mathbb{Z} \), with partition intervals given by
\[
\mathcal{M}_i(t) := \{m: X_m(t) = i\}.
\]
(For simplicity we assumed infinitely many second class particles in both directions, but no problem arises in case we only have finitely many of them.) \( \mathcal{M}_i(t) \) contains the labels of the second class particles that reside at site \( i \) at time \( t \), and can be empty. The labels of the second class particles that are at the same site as the one labeled \( m \) form the set \( \mathcal{M}_{X_m}(t) := \{a^m(t), a^m(t) + 1, \ldots, b^m(t)\} \). The processes \( a^m(t) \) and \( b^m(t) \) are always well defined and satisfy \( a^m(t) \leq m \leq b^m(t) \).

Let us clarify these notions by discussing the ways in which \( a^m(t) \) and \( b^m(t) \) can change.

- A second class particle jumps from site \( X_m(t) - 1 \) to site \( X_m(t) \). Then this one necessarily has label \( a^m(t) - 1 \), and it becomes the lowest labeled one at site \( X_m(t) = X_m(t) \) after the jump. Hence \( a^m(t) = a^m(t) - 1 \).
- A second class particle, different from \( X_m \), jumps from site \( X_m(t) - 1 \) to site \( X_m(t) + 1 \). Then this one is necessarily labeled \( b^m(t) - 1 \), and it leaves the site \( X_m(t) \), hence \( b^m(t) = b^m(t) - 1 \).
- The second class particle \( X_m \) is the highest labeled on its site, that is, \( m = b^m(t) \), and it jumps to site \( X_m(t) + 1 \). Then this particle becomes the lowest labeled in the set \( \mathcal{M}_{X_m(t)+1} = \mathcal{M}_{X_m(t)} \), hence \( a^m(t) = m \). In this case \( b^m(t) \) can be computed from \( b^m(t) - a^m(t) + 1 = \omega X_m(t) - \eta X_m(t) \), the number of second class particles at the site of \( X_m \) after the jump.
We fix initially \( y(0) = z(0) = 0 \). The evolution of \((y, z)\) is superimposed on the background evolution \((\eta, \omega, \{X_m\})\) following the general rule below: Immediately after every move of the background process that involves the site where \( y \) resides before this move, \( y \) picks a new value from the labels on the site where it resides after the move. Thus \( y \) itself jumps only within partition intervals \( M_i \). But \( y \) joins a new partition interval whenever it is the highest \( X \)-label on its site and its “carrier” particle \( X_y \) is forced to move to the next site on the right. This is the situation when \( y(t^-) = b^{y(t^-)}(t^-) - 1 \) and at time \( t \) an \( \omega - \eta \) move from this site happens. (Recall that the choice of \( X \)-particle to move is determined by rule (2.25). In the present case there is only one type of \( \omega - \eta \) move: the highest label from a site moves to the next site on the right.) All this works for \( z \) exactly the same way.

Next we specify the probabilities that \( y \) and \( z \) use to refresh their values. When \( y \) and \( z \) reside at separate sites, they refresh independently. When they are together in the same partition interval, they use the joint distribution in the third bullet below.

- Whenever any change occurs in either \( \omega \) or \( \eta \) at site \( X_{y(t^-)}(t^-) \) and, as a result of the jump, \( a^{y(t^-)}(t) \neq a^{z(t^-)}(t) \), that is, \( y(t^-) \) and \( z(t^-) \) belong to different parts after the jump then, independently of everything else,
  \[
y(t) := \begin{cases} 
a^{y(t^-)}(t), & \text{with pr.} \frac{f(\omega X_{y(t^-)}(t)) - f(\eta X_{y(t^-)}(t))}{f(\omega X_{y(t^-)}(t)) - f(\eta X_{y(t^-)}(t))}, \\
b^{y(t^-)}(t), & \text{with pr.} \frac{f(\omega X_{y(t^-)}(t)) - f(\eta X_{y(t^-)}(t) + 1)}{f(\omega X_{y(t^-)}(t)) - f(\eta X_{y(t^-)}(t))}
\end{cases}
\]
when the denominator is nonzero, and \( y(t) := a^{y(t^-)}(t) \) when the denominator is zero.

- Whenever any change occurs in either \( \omega \) or \( \eta \) at site \( X_{z(t^-)}(t^-) \) and, as a result of the jump, \( a^{y(t^-)}(t) \neq a^{z(t^-)}(t) \), that is, \( y(t^-) \) and \( z(t^-) \) belong to different parts after the jump then, independently of everything else,
  \[
z(t) := \begin{cases} 
b^{z(t^-)}(t) - 1, & \text{with pr.} \frac{f(\omega X_{z(t^-)}(t)) - f(\eta X_{z(t^-)}(t) + 1)}{f(\omega X_{z(t^-)}(t)) - f(\eta X_{z(t^-)}(t))}, \\
b^{z(t^-)}(t), & \text{with pr.} \frac{f(\omega X_{z(t^-)}(t)) - f(\eta X_{z(t^-)}(t))}{f(\omega X_{z(t^-)}(t)) - f(\eta X_{z(t^-)}(t))}
\end{cases}
\]
when the denominator is nonzero, and \( z(t) := b^{z(t^-)}(t) \) when the denominator is zero. When \( \omega X_{z(t^-)}(t) = \eta X_{z(t^-)}(t) + 1, b^{z(t^-)}(t) - 1 \) is not an admissible value but in this case the probability in the first line is zero.

- Whenever any change occurs in either \( \omega \) or \( \eta \) at sites \( X_{y(t^-)}(t^-) \) or \( X_{z(t^-)}(t^-) \) and, as a result of the jump, \( a^{y(t^-)}(t) = a^{z(t^-)}(t) \), that is, \( y(t^-) \) and \( z(t^-) \) belong to the same part after the jump, that is, \( X_{y(t^-)}(t) = X_{z(t^-)}(t) \) then, independently of everything else,
  \[
\begin{pmatrix} y(t) \\ z(t) \end{pmatrix} := \begin{pmatrix} a^{y(t^-)}(t) \\ b^{y(t^-)}(t) - 1 \end{pmatrix}, \begin{pmatrix} a^{x(t^-)}(t) \\ b^{x(t^-)}(t) \end{pmatrix}, \begin{pmatrix} a^{y(t^-)}(t) \\ b^{y(t^-)}(t) \end{pmatrix}, \begin{pmatrix} a^{y(t^-)}(t) \\ b^{y(t^-)}(t) \end{pmatrix},
\]
when the denominator is nonzero, and
\[
(\omega(t^-), \eta(t^-)) := (a^{y(t^-)}(t), b^{y(t^-)}(t))
\]
when the denominator is zero. When \( \omega X_{z(t^-)}(t) = \eta X_{z(t^-)}(t) + 1, b^{z(t^-)}(t) - 1 \) is not an admissible value but in this case the probability in the first line is zero.

The fact that the numbers on the right hand-sides are probabilities follows from \( \omega_i(t) > \eta_i(t) \) on the sites \( i \) in question, and from the monotonicity and concavity of \( f \). The above moves for \( y \) and \( z \) always occur within labels at a given site. This determines whether the particle \( Q(t) := X_{y(t)}(t) \) or \( Q^\psi(t) := X_{z(t)}(t) \) is the one to jump if the next move out of the site is an \( \omega - \eta \) move.
We prove that the above construction has the properties required in Assumption 2.1.

**Lemma 6.1.** The pair \((\omega^-, \omega) := (\omega - \delta_X, \omega)\) obeys basic coupling, as does the pair \((\eta, \eta^+) := (\eta, \eta + \delta_X)\).

**Proof.** We write the proof for \((\omega^-, \omega)\). We need to show that, given the configuration \((\eta, \omega, \{X_m\}, y)\), the jump rates of \((\omega^-, \omega)\) are the ones prescribed in basic coupling (Section 2.3) and by (2.2). Leftward jumps of type (2.3) do not happen in the system under discussion. Since the jump rate function \(p\) depends only on its first argument, jumps out of sites \(i \neq Q\) happen for \(\omega^-\) and \(\omega\) with the same rate \(p(\omega_i^-, \omega_{i+1}) = f(\omega_i^-) = f(\omega_i) = p(\omega_i, \omega_{i+1})\). The only point to consider is jumps out of site \(i = Q\).

Since the last time any change occurred at site \(i\), \(y\) chose values according to (6.1) or (6.3). Notice that (6.1) and (6.3) give the same marginal probabilities for this choice. Hence

\[
y \text{ took on value } a^y \text{ with probability } \frac{f(\omega_i - 1) - f(\eta_i)}{f(\omega_i) - f(\eta_i)} \quad (6.4)
\]

and

\[
y \text{ took on value } b^y \text{ with probability } \frac{f(\omega_i) - f(\omega_i - 1)}{f(\omega_i) - f(\eta_i)} \quad (6.5)
\]
as given in (6.1), or \(y\) took on value \(a^y\) in the case \(f(\omega_i) = f(\eta_i)\). According to the basic coupling of \(\eta\) and \(\omega\), the following jumps can occur over the edge \((i, i + 1)\):

- With rate \(p(\omega_i, \omega_{i+1}) - p(\eta_i, \eta_{i+1}) = f(\omega_i) - f(\eta_i)\), when positive, \(\omega\) jumps without \(\eta\). The highest labeled second class particle, \(X_{b^y}\) jumps from site \(i\) to site \(i + 1\).

  - With probability (6.5) \(X_y = Q\) jumps with \(X_{b^y}\). In this case
    \[
    \omega_i^-(t) = \omega_i(t) - 1 = \omega_i(t) = \omega_i^-(t)
    \]
since the difference \(Q\) disappears from site \(i\). Also,
    \[
    \omega_i^-(t) = \omega_i(t) - 1 = \omega_i(t) = \omega_i^-(t),
    \]
since the difference \(Q\) appears at site \(i + 1\). So in this case \(\omega\) undergoes a jump but \(\omega^-\) does not, and the rate is
    \[
    \left[f(\omega_i) - f(\eta_i)\right] \cdot \frac{f(\omega_i) - f(\omega_i - 1)}{f(\omega_i) - f(\eta_i)} = f(\omega_i) - f(\omega_i^-).
    \]

  - With probability (6.4) \(X_y = Q\) does not jump with \(X_{b^y}\), since it has label \(a^y\) and not \(b^y\) (this probability is zero if \(\omega_i = \eta_i + 1\)). In this case \(\omega^-\) and \(\omega\) perform the same jump and it occurs with rate
    \[
    \left[f(\omega_i) - f(\eta_i)\right] \cdot \frac{f(\omega_i) - f(\eta_i)}{f(\omega_i) - f(\eta_i)} = f(\omega_i^-) - f(\eta_i).
    \]

- With rate \(p(\eta_i, \eta_{i+1}) = f(\eta_i)\), both \(\eta\) and \(\omega\) jump over the edge \((i, i + 1)\). No change occurs in the \(\omega - \eta\) particles, hence no change occurs in \(Q\). This implies that the process \(\omega^-\) jumps as well.

Summarizing we see that the rate for \((\omega^-, \omega)\) to jump together over \((i, i + 1)\) is \(f(\omega_i^-)\), and the rate for \(\omega\) to jump without \(\omega^-\) is \(f(\omega_i) - f(\omega_i^-)\). This is exactly what basic coupling requires.

A very similar argument can be repeated for \((\eta, \eta^+)\).

**Lemma 6.2.** Inequality (2.26) \(y \leq z\) holds in the above construction.

**Proof.** Since no jump of \(y\) or \(z\) moves one of them into a new partition interval, the only situation that can jeopardize (2.26) is the simultaneous refreshing of \(y\) and \(z\) in a common partition interval. But this case is governed by step (6.3) which by definition ensures that \(y \leq z\).
So far in this section everything is valid for a general zero range process with nondecreasing concave jump rate. Now we use the special convexity requirement (2.37). With \( r \in (0, 1) \) from (2.37), define the geometric distribution

\[
v(m) := \begin{cases} 
(1 - r)r^m, & m \geq 0, \\
0, & m < 0.
\end{cases}
\]  

(6.6)

**Lemma 6.3.** Conditioned on the process \((\eta, \omega)\), the bounds \( y(t) \leq v \) and \( z(t) \geq -v \) hold for all \( t \geq 0 \).

The proof of this lemma is achieved in three steps.

**Lemma 6.4.** Let \( Y \) be a random variable with distribution \( v \), and fix integers \( a \leq b \) and \( \eta < \omega \) so that \( \omega - \eta = b - a + 1 \). Apply the following operation to \( Y \):

(i) if \( a \leq Y \leq b \), apply the probabilities from (6.1) (equivalently, (6.4) and (6.5)) with parameters \( a, b, \eta, \omega \) to pick a new value for \( Y \);

(ii) if \( Y < a \) or \( Y > b \) then do not change \( Y \).

Then the resulting distribution \( v^* \) is stochastically dominated by \( v \).

**Proof.** There is nothing to prove when \( b = a \), hence we assume \( b > a \) or, equivalently, \( \omega - \eta = b - a + 1 \geq 2 \). It is also clear that \( v^*(m) = v(m) \) for \( m < a \) or \( m > b \). We need to prove, in view of the distribution functions,

\[
\sum_{\ell=a}^{m} v^*(\ell) \geq \sum_{\ell=a}^{m} v(\ell) \quad \text{or, equivalently,} \quad \sum_{\ell=m}^{b} v^*(\ell) \leq \sum_{\ell=m}^{b} v(\ell)
\]

for all \( a \leq m \leq b \). Notice that \( v^* \) gives zero weight on values \( a < m < b \) (if any), therefore the left hand-side of the second inequality equals \( v^*(b) \) for \( a < m \leq b \). Hence the above display is proved once we show

\[
v^*(b) \leq v(b), \quad \text{that is,} \quad \frac{f(\omega) - f(\omega - 1)}{f(\omega) - f(\eta)} \cdot \sum_{\ell=a}^{b} v(\ell) \leq v(b),
\]  

(6.7)

see (6.1). When \( f(\omega) = f(\omega - 1) \), there is nothing to prove. Hence assume \( f(\omega) > f(\omega - 1) \) which by concavity implies that \( f \) has positive increments on \( \{\eta, \ldots, \omega\} \). If \( b < 0 \) then both sides are zero. If \( b \geq 0 \) then we have, by (2.37),

\[
v(\ell) \leq v(b) \cdot r^{\ell-b} \leq v(b) \cdot \prod_{z=\omega-b+\ell}^{\omega-1} \frac{f(z) - f(z-1)}{f(z+1) - f(z)} = v(b) \cdot \frac{f(\omega - b + \ell) - f(\omega - b + \ell - 1)}{f(\omega) - f(\omega - 1)}
\]

for each \( \ell \leq b \). The first inequality also takes into account possible \( v(\ell) = 0 \) values for negative \( \ell \)'s. With this we can write

\[
\sum_{\ell=a}^{b} v(\ell) \leq v(b) \cdot \frac{f(\omega) - f(\omega - b + a - 1)}{f(\omega) - f(\omega - 1)}
\]

which becomes (6.7) via \( \omega - \eta = b - a + 1 \).

We repeat the lemma for \( z(t) \).

**Lemma 6.5.** Let \( Z \) be a random variable of distribution \(-v\), and fix integers \( a \leq b \), \( \eta < \omega \) so that \( \omega - \eta = b - a + 1 \). Operate on \( Z \) as was done for \( Y \) in Lemma 6.4, but this time use the probabilities from (6.2) with parameters \( a, b, \eta, \omega \). Let \(-v^*\) be the resulting distribution. Then \( v^* \) is stochastically dominated by \( v \).
Fluctuation bounds

Proof. Again, we assume $b > a$ or, equivalently, $\omega - \eta = b - a + 1 \geq 2$. It is also clear that $v^*(-m) = v(-m)$ for $m < a$ or $m > b$. We need to prove

$$
\sum_{\ell=a}^{m} v^*(-\ell) \leq \sum_{\ell=a}^{m} v(-\ell)
$$

for all $a \leq m \leq b$. Notice that $-v^*$ gives zero weight on values $a \leq \ell < b - 1$ (if any), therefore the left hand-side of the inequality equals 0 for $a \leq m < b - 1$, $v^*(b - 1)$ for $m = b - 1$, and agrees to the right hand-side for $m = b$. Hence the above display is proved once we show

$$
v^*(-b) \geq v(-b), \quad \text{that is,}
$$

$$
\frac{f(\eta + 1) - f(\eta)}{f(\omega) - f(\eta)} \cdot \sum_{\ell=a}^{b} v(-\ell) \geq v(-b),
$$

(6.8)

see (6.2). We have, by (2.37),

$$
v(-\ell) \geq v(-b) \cdot r^{b-\ell} \geq v(-b) \cdot \prod_{z=\eta+1}^{\eta+b-\ell} \frac{f(z+1) - f(z)}{f(z) - f(z-1)} = v(-b) \cdot \frac{f(\eta + 1 + b - \ell) - f(\eta + b - \ell)}{f(\eta + 1) - f(\eta)}
$$

for each $\ell \leq b$. The first inequality also takes into account possible $v(-b) = 0$ values for positive $b$’s. With this we can write

$$
\sum_{\ell=a}^{b} v(-\ell) \geq v(-b) \cdot \frac{f(\eta + 1 + b - a) - f(\eta)}{f(\eta + 1) - f(\eta)}
$$

which becomes (6.8) via $\omega - \eta = b - a + 1$.

Lemma 6.6. The dynamics defined by (6.1) or (6.2) is attractive.

Proof. Following the same realizations of (6.1), we see that two copies of $y(\cdot)$ under a common environment can be coupled so that whenever they get to the same part $M_i$, they move together from that moment. The same holds for $z(\cdot)$.

Proof of Lemma 6.3. Initially $y(0) = 0$ by definition, which is clearly a distribution dominated by $v$ of (6.6). Now we argue recursively: by time $t$ the distribution of $y(t)$ was a.s. only influenced by finitely many jumps of the environment, which resulted in distributions $v_1$, then $v_2$, then $v_3$, etc. Suppose $v_k \overset{d}{\leq} v$, and let $v^*$ be the distribution that would result from $v$ by the $(k + 1)$st jump. Then $v_{k+1} \overset{d}{\leq} v^*$ by $v_k \overset{d}{\leq} v$ and Lemma 6.6, while $v^* \overset{d}{\leq} v$ by Lemma 6.4. A similar argument proves the lemma for $z(\cdot)$.

Appendix A: Convexity and total positivity

This section derives a general convexity result for exponentially tilted measures. Let $v$ be a nondegenerate probability measure on $\mathbb{R}$ and assume that for some open interval $\mathcal{I} \subseteq \mathbb{R}$,

$$
Y(\theta) = \int e^{\theta x} v(dx) < \infty \quad \text{for all } \theta \in \mathcal{I}.
$$

(A.1)
For \( \theta \in \mathcal{I} \) define the exponentially tilted measures \( \nu^\theta \) by
\[
\int g \, d\nu^\theta = Y(\theta)^{-1} \int g(x) e^{\theta x} \nu(dx)
\]
(for bounded Borel test functions \( g \)). The nondegeneracy assumption (that \( \nu \) is not supported on a single point) and (A.1) guarantee that
\[
\varrho(\theta) = \int x \nu^\theta(dx)
\]
is a finite, continuous, strictly increasing function that maps \( \mathcal{I} \) onto a nontrivial open interval \( \mathcal{J} \). For \( \varrho \in \mathcal{J} \) the inverse function is denoted by \( \theta(\varrho) \).

Let \( \psi \) be a measurable function on \( \mathbb{R} \), and assume (by shrinking \( \mathcal{I} \) if necessary) that
\[
\int |\psi| \, d\nu^\theta < \infty \quad \text{for all } \theta \in \mathcal{I}.
\]
Since \(|x|^k \leq k! e^{-k} (e^x + e^{-x})\) for any \( \varepsilon > 0 \) and \( \mathcal{I} \) is an open interval, it follows that \( \int |\psi||x|^k \, d\nu^\theta < \infty \) for all \( k \geq 0 \) and \( \theta \in \mathcal{I} \). Consequently as a function of \( \theta \) the integral \( \int \psi \, d\nu^\theta \) has derivatives of all orders.

A particular case is \( \psi(x) = x \) which gives the infinite differentiability of \( \varrho(\theta) \). Let us also note the infinite differentiability of the inverse function \( \theta(\varrho) \). Since \( \varrho'(\theta) \) is the variance of the distribution \( \nu^\theta \), \( \varrho'(\theta) > 0 \) by the nondegeneracy of \( \nu \), and so directly from the definition of the derivative \( \theta'(\varrho) = 1/\varrho'(\theta(\varrho)) \). Repeated use of basic differentiation rules produces all derivatives \( \theta^{(n)}(\varrho) \). Notice that this argument shows a uniform lower and upper bound of \( \varrho'(\theta) \), that is, Lipschitz continuity of both \( \varrho(\theta) \) and \( \theta(\varrho) \) on bounded closed intervals.

Define
\[
\Psi(\varrho) = \int \psi \, d\nu^{\theta(\varrho)}.
\]
\( \Psi \) is also infinitely differentiable as a composite of two such functions.

**Theorem A.1.** Assume \( \psi \) is a convex function on \( \mathbb{R} \). Then \( \Psi \) is convex on \( \mathcal{J} \). Assume furthermore that no linear function \( g(x) = ax + b \) satisfies \( \psi = g \ \nu \text{-a.e.} \) Then \( \Psi''(\varrho) > 0 \) for all \( \varrho \in \mathcal{J} \) and in particular \( \Psi \) is strictly convex on \( \mathcal{J} \).

**Proof.** The proof can be reduced to the theory of total positivity. In what follows, citations and terminology are from Karlin’s monograph [24]. The claims made in our Theorem A.1 follow from applying Theorem 3.5(a)–(c) from p. 285 of [24] to the operator
\[
T\psi(\varrho) = \int \psi \, d\nu^{\theta(\varrho)} = \int_{\mathbb{R}} K(\varrho,x)\psi(x) \nu(dx), \quad \varrho \in \mathcal{J},
\]
where the kernel is defined by \( K(\varrho,x) = Y(\varrho(\theta))^{-1} e^{\theta \varrho x} \). The property of the kernel \( K \) that gives the result is extended total positivity (ETP) of order 3. This is the requirement of strict positivity on certain types of determinants of partials of dimensions up to \( 3 \times 3 \): for all \( (\varrho,x) \in \mathcal{J} \times \mathbb{R} \),
\[
K^*\left( \begin{array}{c} \varrho, \ldots, \varrho \\ x, \ldots, x \end{array} \right) = \det_{1 \leq i,j \leq n} \frac{\partial^{i+j} - 2}{\partial \varrho^{i-1} \partial x^{j-1}} K(\varrho,x) > 0 \quad \text{for } n = 1, 2, 3.
\]
(A.2)

We argue this in stages.

We first observe that the kernel \( L(\theta,x) = Y(\theta)^{-1} e^{\theta x} \) on \( \mathcal{I} \times \mathbb{R} \) is ETP of all orders. Recall that the Wronskian of \( n \) functions \( f_1, \ldots, f_n \) is the \( n \times n \) determinant
\[
W[f_1, \ldots, f_n](x) = \det_{1 \leq i,j \leq n} f_i^{(j-1)}(x).
\]
If \( u \) is another function, the Wronskian satisfies the identity

\[
W[uf_1, \ldots, uf_n](x) = u(x)^n W[f_1, \ldots, f_n](x).
\]  

(A.3)

To justify (A.3), Leibniz’s rule

\[
(u f_i)^{(j-1)} = \sum_{k=1}^{j} \binom{j-1}{k-1} f_i^{(k-1)} u^{(j-k)} \quad (1 \leq j \leq n)
\]

implies that the matrix \( A = [(u f_i)^{(j-1)}(x)]_{1 \leq i, j \leq n} \) is the product of the matrices

\[
B = [(f_i)^{(k-1)}(x)]_{1 \leq i, k \leq n} \quad \text{and} \quad C = \left[ \binom{j-1}{k-1} u^{(j-k)}(x) \mathbf{1}[k \leq j] \right]_{1 \leq k, j \leq n}.
\]

By upper-triangularity \( \det C = u(x)^n \). Then the corresponding determinant identity \( \det(A) = \det(B) \cdot \det(C) \) is precisely (A.3).

Now we can verify the ETP property of kernel \( L \), utilizing (A.3):

\[
\det_{1 \leq i, j \leq n} \left[ \frac{\partial^{i+j-2}}{(\partial x^{i-1}) \partial \theta^{j-1}} L(\theta, x) \right] = \det_{1 \leq i, j \leq n} \left[ \frac{\partial^{j-1}}{(\partial \theta^{j-1})} \left\{ Y^{-1}(\theta)^{-1} Yx \right\} \right]
\]

\[
= Y^{n} e^{\theta x} W[1, \theta, \ldots, \theta^{n-1}]
\]

\[
= Y^{n} e^{\theta x} \prod_{j=1}^{n-1} j! > 0.
\]

To go from \( L(\theta, x) \) to \( K(\varphi, x) = L(\theta(\varphi), x) \), consider the \( 3 \times 3 \) determinant that appears in (A.2), apply the chain rule and a row operation:

\[
\left| \begin{array}{ccc}
K_{x} & K_{xx} & K_{x}\theta \\
K_{\theta} & K_{\theta x} & K_{\theta xx} \\
K_{\theta \theta} & K_{\theta \theta x} & K_{\theta \theta xx}
\end{array} \right| = \left| \begin{array}{ccc}
L_{x} & L_{xx} & L_{\theta x} \\
L_{\theta x} & L_{\theta xx} & L_{\theta x} \\
L_{\theta x} & L_{\theta xx} & L_{\theta x}
\end{array} \right| > 0.
\]

The last inequality is by the ETP property of kernel \( L \) and the strict positivity \( \theta > 0 \) of the derivative. The \( 1 \times 1 \) and \( 2 \times 2 \) determinants in (A.2) are principal minors of the determinant above and are positive by the same reasoning.

We have shown that the kernel \( K \) has the ETP property of order 3. In addition to ETP, Theorem 3.5 from p. 285 of [24] requires the hypotheses

\[
\int K(\varphi, x) \nu(dx) = 1 \quad \text{and} \quad \int K(\varphi, x) x \nu(dx) = a \varphi + b
\]

for some \( a > 0 \) and \( b \in \mathbb{R} \). The first one is true by virtue of the normalization \( Y(\theta)^{-1} \), and the second one with \( a = 1 \) and \( b = 0 \) by the definition of \( \varphi(\theta) \). The proof is now completed by an appeal to Theorem 3.5 from p. 285 of [24].

These convexity properties can also be proved in an elementary way by developing suitable correlation inequalities. Such a proof is given in the note [10]. We are indebted to an anonymous referee of that note for pointing out the connection with total positivity.

Subsequent sections of the appendix extract from Theorem A.1 consequences for the processes we study.
Appendix B: Monotonicity of measures

In this part of the appendix we show that the measures $\mu^\theta$ and $\hat{\mu}^\theta$ defined in (2.13) and (2.17), respectively, are stochastically monotone as functions of $\varrho$. We start with a simple lemma.

**Lemma B.1.** Fix a function $\varphi(\omega)$ on $\mathbb{Z}$, bounded by a polynomial. Then $E^\theta(\varphi(\omega))$ is differentiable in $\theta$ on $(\theta, \tilde{\theta})$, and
\[
\frac{d}{d\theta}E^\theta(\varphi(\omega)) = \text{Cov}^\theta(\varphi(\omega), \omega).
\]

**Proof.** Convergence of the series involved in $E^\theta(\varphi(\omega))$ can be verified via the ratio test, even after differentiating the terms. Since $\mu^\theta$ is the exponentially weighted version of $\mu^\theta_0$ for some $\theta_0$, we have
\[
\frac{d}{d\theta}E^\theta_{\omega}(\varphi(\omega) \cdot e^{(\theta-\theta_0)\omega}) = \frac{E^{\theta_0}(\varphi(\omega) \cdot e^{(\theta-\theta_0)\omega}) - E^{\theta_0}(\varphi(\omega) \cdot e^{(\theta-\theta_0)\omega})}{E^{\theta_0}(e^{(\theta-\theta_0)\omega})^2} = \text{Cov}^\theta(\varphi(\omega), \omega).
\]

**Corollary B.2.** For any $\theta < \tilde{\theta}$, the state sum (2.12) satisfies
\[
\frac{d}{d\theta} \log Z(\theta) = \frac{1}{Z(\theta)} \sum_{\omega = \omega_{\text{min}}}^{\omega_{\text{max}}} \text{argmax}_{z \in I} \frac{e^{\theta z}}{f(z)!} = E^\theta(\omega) =: \varrho(\theta), \quad (B.1)
\]
\[
\frac{d^2}{d\theta^2} \log Z(\theta) = \frac{d}{d\theta} \varrho(\theta) = \text{Var}^\theta(\omega). \quad (B.2)
\]
The function $\varrho(\theta)$ is strictly increasing and maps $(\theta, \tilde{\theta})$ onto $(\omega_{\text{min}}, \omega_{\text{max}})$.

**Proof.** Everything is already covered except the last surjectivity statement. Due to the monotonicity and continuity one only needs to show convergence at the boundaries $\theta, \tilde{\theta}$ to $\omega_{\text{min}}, \omega_{\text{max}}$. First let us consider the case when $\tilde{\theta} < \infty$. Then $\omega_{\text{max}} = \infty$ and Fatou’s lemma implies
\[
\lim_{\theta \to \tilde{\theta}} \log Z(\theta) = \lim_{\theta \to \tilde{\theta}} \sum_{z \in I} \frac{e^{\theta z}}{f(z)!} = \sum_{z \in I} \lim_{\theta \to \tilde{\theta}} \frac{e^{\theta z}}{f(z)!} = \sum_{z \in I} \frac{e^{\tilde{\theta} z}}{f(z)!} = \infty
\]
since for $z > 0$
\[
\frac{e^{\tilde{\theta} z}}{f(z)!} = \prod_{y=1}^{z} \frac{e^{\tilde{\theta}}}{f(y)} \geq 1
\]
by definition of $\tilde{\theta}$ and $f$ being nondecreasing. This shows that $\log Z(\theta)$ takes on arbitrarily large values as $\theta \to \tilde{\theta}$. We also know that it is a smooth and convex function on $(\theta, \tilde{\theta})$ (see (B.2)). This implies that its derivative (B.1) is not bounded from above i.e., arbitrarily large $\varrho$ values can be achieved. The same reasoning works in case $\theta > -\infty$ for arbitrarily large negative $\varrho$ values.

When $\tilde{\theta} = \infty$ then, regardless whether $\omega_{\text{max}}$ is finite or infinite, fix any $0 \leq y < \omega_{\text{max}}$ and write
\[
\varrho(\theta) = E^\theta(\omega \cdot 1_{\{|\omega > y\}}) + E^\theta([|\omega|]_+ \cdot 1_{\{|\omega \leq y\}}) - E^\theta([|\omega|]_- \cdot 1_{\{|\omega \leq y\}})
\]
\[
\geq (y+1) \cdot P^\theta(\omega > y) - E^\theta([|\omega|]_- \cdot 1_{\{|\omega \leq y\}})
\]
\[
\geq (y+1) - (y+1) \cdot P^\theta(\omega \leq y) - \sqrt{E^\theta([|\omega|]_-^2)} \cdot \sqrt{P^\theta(\omega \leq y)}
\]
\[
\geq (y+1) - (y+1) \cdot P^\theta(\omega \leq y) - \sqrt{E^{\theta_0}([|\omega|]_-^2)} \cdot \sqrt{P^\theta(\omega \leq y)} \quad (B.3)
\]
for a fixed $\theta < \theta_0 < \theta$. The last inequality follows by monotonicity of $\mu^\theta$ in $\theta$ and $([\omega^-])^2$ being a nonincreasing function of $\omega$. For any $\omega_{\min}^\theta - 1 < y \leq \omega$ and $\theta > \bar{\theta}$,

$$\frac{\mu^\theta(z)}{\mu^\theta(y + 1)} = \prod_{x = z}^y \frac{\mu^\theta(x)}{\mu^\theta(x + 1)} = \prod_{x = z}^y \frac{f(x + 1)}{e^\theta} \leq \left( \frac{f(y + 1)}{e^\theta} \right)^{y-z+1}.$$ 

Given $0 \leq y < \omega_{\max}^\theta$ and $1 > \varepsilon > 0$, there is a large enough $\theta$ which makes the last fraction smaller than $\varepsilon$. With such a choice we have

$$P^\theta(\omega \leq y) = \sum_{z = \omega_{\min}^\theta}^y \mu^\theta(z) \leq \mu^\theta(y + 1) \sum_{z = \omega_{\min}^\theta}^y \varepsilon^{y-z+1} \leq \frac{1 - \varepsilon^{y-\omega_{\min}^\theta+1}}{1 - \varepsilon}.$$ 

Therefore, for the case of a finite $\omega_{\max}^\theta$, choosing $y = \omega_{\max}^\theta - 1$ and large $\theta$ makes (B.3) arbitrarily close to $\omega_{\max}^\theta$. When $\omega_{\max}^\theta = \infty$, the argument shows that $\bar{\varphi}(\theta) \geq y + 1$ can be achieved for any $y \geq 0$. A similar computation demonstrates that any density towards $\omega_{\min}^\theta$ can be reached when $\bar{\theta} = -\infty$.  

**Corollary B.3.** The measures $\mu^\theta$ are stochastically nondecreasing in $\varphi$.

**Proof.** Since $\varphi$ and $\theta$ are strictly increasing functions of each other, it is equivalent to show monotonicity of $\mu^\theta$. This follows if we can show $0 \leq \frac{d}{d\varphi} \mathbb{E}^\theta(\varphi(\omega))$ for an arbitrary bounded nondecreasing function $\varphi$. Lemma B.1 transforms this derivative into the covariance of $\varphi(\omega)$ and $\omega$, which is nonnegative due to $\varphi$ being nondecreasing.

Monotonicity of $\hat{\mu}^\theta$ requires somewhat more of a convexity argument.

**Proposition B.4.** The family of measures $\hat{\mu}^\theta$, defined in (2.17), is stochastically nondecreasing in $\varphi$.

**Proof.** Start by rewriting the definition:

$$\hat{\mu}^\theta(y) = \frac{\mathbb{E}^\theta([\omega - \varphi] \cdot 1[\omega > y])}{\text{Var}^\theta(\omega)} = \frac{\text{Cov}^\theta(\omega, 1[\omega > y])}{\text{Cov}^\theta(\omega, \omega)} = \left. \frac{\frac{d}{d\varphi} P^\theta(\omega > y)}{\frac{d}{d\varphi} \varphi(\omega)} \right|_{\varphi = \varphi(\theta)} = \frac{d}{d\varphi} P^\theta(\omega > y).$$

Let us denote the $\hat{\mu}^\theta$-expectation by $\hat{\mathbb{E}}^\theta$. Fix a bounded nondecreasing function $\varphi$. We need to show

$$0 \leq \frac{d}{d\varphi} \hat{\mathbb{E}}^\theta(\varphi(\omega)).$$

We compute a different expression for this derivative. Passing the derivative through the sum in the third equality below is justified because the series involved are dominated by certain geometric series, uniformly over $\theta$ in small open neighborhoods. This follows from the definitions of $\bar{\theta}$ and $\bar{\varphi}$ and the assumption $\bar{\theta} < \theta(\varphi) < \bar{\varphi}$.

$$\hat{\mathbb{E}}^\theta(\varphi(\omega)) = \sum_{y = \omega_{\min}^\theta}^{\omega_{\max}^\theta} \varphi(y) \cdot \frac{d}{d\varphi} P^\theta(\omega > y) = \sum_{y = \omega_{\min}^\theta}^{\omega_{\max}^\theta} \varphi(y) \cdot \left[ P^\theta(\omega > y) - 1[0 \geq y] \right]$$

$$= \frac{d}{d\varphi} \sum_{y = \omega_{\min}^\theta}^{\omega_{\max}^\theta} \varphi(y) \cdot \left[ 1[0 > y] - 1[0 \geq y] \right] = \frac{d}{d\varphi} \mathbb{E}^\theta \sum_{y = \omega_{\min}^\theta}^{\omega_{\max}^\theta} \varphi(y) \cdot \left[ 1[0 > y] - 1[0 \geq y] \right]$$

$$= \frac{d}{d\varphi} \mathbb{E}^\theta \sum_{y = \omega_{\min}^\theta}^{\omega_{\max}^\theta} \varphi(y) \cdot [1[0 > y] - 1[0 \geq y]] = \frac{d}{d\varphi} \mathbb{E}^\theta \left[ \sum_{y = 1}^{\omega_{\max}^\theta} \varphi(y) - \sum_{y = 0}^{\omega_{\min}^\theta} \varphi(y) \right] = \frac{d}{d\varphi} \Phi(\omega).$$
Above we introduced the function
\[ \Phi(x) = x - \sum_{y = x}^{x-1} \varphi(y) - \sum_{y = x}^{0} \varphi(y), \]
with the convention that empty sums are zero. To conclude the proof, notice that \( \Phi(x + 1) - \Phi(x) = \varphi(x) \). Thus a nondecreasing function \( \varphi \) determines a (nonstrictly) convex function \( \Phi \) with \( \Phi(1) = 0 \), and vice-versa. Hence Theorem A.1 establishes that
\[ \frac{d}{d\varrho} \mathbb{E}^\varrho \varphi(\omega) = \frac{d^2}{d\varrho^2} \mathbb{E}^\varrho \Phi(\omega) \geq 0. \]

□

Appendix C: Regularity properties of the hydrodynamic flux function

For the zero range process defined among the examples in Section 2.2, the hydrodynamic (macroscopic) flux function \( \mathcal{H} : \mathbb{R}^+ \to \mathbb{R}^+ \) of (2.14) is given by
\[ \mathcal{H}(\varrho) = \mathbb{E}^\varrho f(\omega). \]
The results of Section A for \( f \) now read as follows:

**Proposition C.1.** If the jump rate \( f \) of the zero range process is convex (or concave), then the flux \( \mathcal{H} \) is also convex (or concave, respectively). Moreover, in this case \( \mathcal{H}''(\varrho) > 0 \) (or \( \mathcal{H}''(\varrho) < 0 \), respectively) for all \( \varrho > 0 \) if and only if \( f \) is not a linear function.

Parts of this proposition were proved with coupling methods in [5].

Next we show in the general case that \( \mathcal{H}(\varrho) \) is well defined, and is infinitely differentiable. (We use third derivatives in the proof of Theorem 2.3.) The function \( \mathcal{H}(\varrho) \) is, in general, the expected net growth rate w.r.t. \( \mu^\varrho \) as defined in (2.14). We show that the series making up this expectation is finite, even after differentiating its terms. This will then lead to smoothness of \( \mathcal{H}(\varrho) \).

**Lemma C.2.** Let \( g(y, z) \geq 0 \) be any function on \( \mathbb{Z} \times \mathbb{Z} \), bounded by a polynomial in \( |y| \) and \( |z| \). Then for any \( \theta < \theta' < \bar{\theta} \),
\[ \mathbb{E}^\theta \left[ \left( p(\omega_0, \omega_1) + q(\omega_0, \omega_1) \right) g(\omega_0, \omega_1) \right] < \infty. \]

**Proof.** We deal with the first part that contains \( p \), the one with \( q \) can be treated analogously. The sum we are looking at is
\[ \sum_{y = \omega_{\min} + 1}^{\omega_{\max}} \sum_{z = \omega_{\min}}^{\omega_{\max} - 1} p(y, z) \cdot g(y, z) \cdot \frac{e^{\theta(y+z)}}{f(y)! \cdot f(z)!} \cdot \frac{1}{Z(\theta)^2}. \]
These sums are certainly convergent if \( \omega_{\min} \) and \( \omega_{\max} \) are both finite. When this is not the case we split both summations at zero, and convergence is established on the four quadrants of the plane. We use (2.7) and the corollary
\[ p(y, z) = p(z + 1, y - 1) \cdot \frac{f(y)}{f(z + 1)} \quad \text{for} \quad \omega_{\min} < y \leq \omega_{\max} \quad \text{and} \quad \omega_{\min} \leq z < \omega_{\max} \]
of (2.9), and we consider empty sums to be zero.
• \( y > 0, z > 0 \): In this case
\[
p(y, z) \leq p(y, 0) = p(1, y - 1) \cdot \frac{f(y)}{f(1)} \leq p(1, 0) \cdot \frac{f(y)}{f(1)},
\]
and the corresponding part of the summation is bounded by
\[
p(1, 0) \cdot \sum_{y=0}^{\omega_{\max}} \sum_{z=0}^{\omega_{\max} - 1} g(y, z) \cdot \frac{e^{\theta(y+z)}}{f(y)! \cdot f(z)!} \cdot \frac{1}{Z(\theta)^2}.
\]

• \( y \leq 0, z > 0 \): In this case
\[
p(y, z) \leq p(1, 0),
\]
and the corresponding part of the summation is bounded by
\[
p(1, 0) \cdot \sum_{y=\omega_{\min}}^{0} \sum_{z=\omega_{\min} + 1}^{\omega_{\max} - 1} g(y, z) \cdot \frac{e^{\theta(y+z)}}{f(y)! \cdot f(z)!} \cdot \frac{1}{Z(\theta)^2}.
\]

• \( y \leq 0, z \leq 0 \): In this case
\[
p(y, z) \leq p(1, z) = p(z + 1, 0) \cdot \frac{f(1)}{f(z + 1)} \leq p(1, 0) \cdot \frac{f(1)}{f(z + 1)},
\]
and the corresponding part of the summation is bounded by
\[
p(1, 0) f(1) \cdot \sum_{y=\omega_{\min}}^{0} \sum_{z=\omega_{\min}}^{0} g(y, z) \cdot \frac{e^{\theta(y+z)}}{f(y)! \cdot f(z)!} \cdot \frac{1}{Z(\theta)^2}.
\]

• \( y > 0, z \leq 0 \): In this case
\[
p(y, z) = p(z + 1, y - 1) \cdot \frac{f(y)}{f(z + 1)} \leq p(1, 0) \cdot \frac{f(y)}{f(z + 1)},
\]
and the corresponding part of the summation is bounded by
\[
p(1, 0) \cdot \sum_{y=\omega_{\max}}^{0} \sum_{z=\omega_{\min}}^{0} g(y, z) \cdot \frac{e^{\theta(y+z)}}{f(y)! \cdot f(z)!} \cdot \frac{1}{Z(\theta)^2}.
\]

Convergence of each of these bounds for \( \theta < \bar{\theta} < \bar{\theta} \) is established e.g. by the ratio test.

Notice that a similar argument gives finite higher moments of the rates when \( \log(f) \) is at most linear in both directions on \( \mathbb{Z} \).

**Corollary C.3.** \( \mathcal{H}(\rho) \) is infinitely differentiable at all \( \rho \in (\omega_{\min}, \omega_{\max}) \).

**Proof.** By the previous lemma the series
\[
F(\theta) := \mathcal{H}(\rho(\theta)) = \frac{1}{Z(\theta)^2} \cdot \sum_{y, z=\omega_{\min}}^{\omega_{\max}} (p(y, z) - q(y, z)) \frac{e^{\theta(y+z)}}{f(y)! \cdot f(z)!},
\]
is convergent and infinitely differentiable. Since \( \mathcal{H}(\rho) = F(\theta(\rho)) \) and \( \rho \mapsto \theta(\rho) \) is infinitely differentiable as well, the claim follows.
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References

