



The Inert Drift Atlas Model

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Abstract: Consider a massive (inert) particle impinged from above by N Brownian particles that are instantaneously reflected upon collision with the inert particle. The velocity of the inert particle increases due to the influence of an external Newtonian potential (e.g. gravitation) and decreases in proportion to the total local time of collisions with the Brownian particles. This system models a semi-permeable membrane in a fluid having microscopic impurities (Knight in *Probab Theory Relat Fields* 121:577–598, 2001). We study the long-time behavior of the process (V, \mathbf{Z}) , where V is the velocity of the inert particle and \mathbf{Z} is the vector of gaps between successive particles ordered by their relative positions. The system is not hypoelliptic, not reversible, and has singular form interactions. Thus the study of stability behavior of the system requires new ideas. We show that this process has a unique stationary distribution that takes an explicit product form which is Gaussian in the velocity component and exponential in the other components. We also show that convergence in total variation distance to the stationary distribution happens at an exponential rate. We further obtain certain law of large numbers results for the particle locations and intersection local times.

1. Introduction

1.1. Motivation and model description. In this work we study the long-time behavior of an interacting particle system comprising a massive (inert) particle that moves under the combined influence of an external Newtonian potential (eg. gravitation) and a non-Newtonian ‘inert drift’ resulting from collisions with many microscopic (Brownian) particles. This serves as a simplified model for the motion of a semi-permeable membrane in a fluid having microscopic impurities (see [30]). The membrane, which allows fluid molecules to pass but is impermeable to the impurities, plays the role of the inert particle.

Mathematically, this model consists of N -Brownian particles in \mathbb{R} , with state processes denoted as $\{X_i(t), t \geq 0\}_{1 \leq i \leq N}$, interacting with the inert particle, with state

process $X_0(t)$, according to the following system of equations: For $t \geq 0$,

$$\begin{aligned} X_0(t) &= x_0 + \int_0^t V(s)ds, \quad V(t) = v_0 + gt - \sum_{i=1}^N \ell_i(t), \\ X_i(t) &= x_i + W_i(t) + \ell_i(t), \quad 1 \leq i \leq N. \end{aligned} \quad (1.1)$$

Here $x_0 \leq x_1 \leq \dots \leq x_N$ denote the initial positions of the $N+1$ particles, v_0 the initial velocity of the inert particle, $\{W_i, 1 \leq i \leq N\}$ are mutually independent standard real Brownian motions, $g \in (0, \infty)$ denotes the gravitation constant and ℓ_i is the collision local time between the i -th particle and the inert particle which, in particular, satisfies $\ell_i(t) = \int_0^t 1_{\{X_i(s)=X_0(s)\}} d\ell_i(s)$ for $1 \leq i \leq N$ and $t \geq 0$. The local time interactions model the cumulative transfer of momentum when a Brownian particle collides with the inert particle ‘infinitely often’ on finite time intervals, with each collision resulting in an infinitesimal momentum transfer. Such interactions lie at the heart of this model and interesting long time behavior results from the combined effect of the ‘soft’ gravitational potential and ‘hard’ collisions.

It follows from [8] (see Theorem 2.5 and Proposition 2.10 therein) that there is a strong solution to the system of equations in (1.1) and the solution satisfies $X_0(t) \leq X_i(t)$ for all $t \geq 0$ and $1 \leq i \leq N$ a.s. Using Gronwall’s lemma and the Lipschitz property of the Skorohod map it is easy to verify that in fact the system of equations in (1.1) has a unique strong solution. Given this unique solution process $\{X_i(t), t \geq 0\}_{0 \leq i \leq N}$ of (1.1) it will be convenient to consider the ordered particle system:

$$X_{(0)}(t) \leq X_{(1)}(t) \leq \dots \leq X_{(N)}(t), \quad t \geq 0,$$

where $\{X_{(i)}(t) : t \geq 0\}$ denotes the state process of the i -th particle from the bottom (note that the lowest particle, which we call the 0-th particle from the bottom, is the inert particle, in particular, $X_{(0)}(\cdot) = X_0(\cdot)$). By an application of Tanaka’s formula it is easy to verify that this ranked particle system satisfies the following system of equations: For $t \geq 0$,

$$\begin{aligned} X_{(0)}(t) &= x_0 + \int_0^t V(s)ds, \quad V(t) = v_0 + gt - L_1(t), \\ X_{(1)}(t) &= x_1 + B_1(t) - \frac{1}{2}L_2(t) + L_1(t), \\ X_{(i)}(t) &= x_i + B_i(t) - \frac{1}{2}L_{i+1}(t) + \frac{1}{2}L_i(t), \quad 2 \leq i \leq N. \end{aligned} \quad (1.2)$$

where $x_0 \leq x_1 \leq \dots \leq x_N$, $\{B_i, 1 \leq i \leq N\}$ are standard independent Brownian motions and for $1 \leq i \leq N$, L_i denotes the collision local time between the i -th and the $(i-1)$ -th ranked particle which satisfies $L_i(t) = \int_0^t 1_{\{X_{(i)}(s)=X_{(i-1)}(s)\}} dL_i(s)$ and $L_{N+1}(t) = 0$ for all $t \geq 0$.

We are interested in the time asymptotic behavior of the *velocity and gap processes* associated with this system. Namely, denoting $Z_i(t) \doteq X_{(i)}(t) - X_{(i-1)}(t)$, the object of interest is the stochastic process

$$(V(t), Z_1(t), \dots, Z_N(t)).$$

This process is given by the system of equations

$$\begin{aligned}
V(t) &= v_0 + gt - L_1(t), \\
Z_1(t) &= z_1 + B_1(t) - \int_0^t V(s)ds - \frac{1}{2}L_2(t) + L_1(t), \\
Z_2(t) &= z_2 + B_2(t) - B_1(t) - \frac{1}{2}L_3(t) + L_2(t) - L_1(t), \\
Z_i(t) &= z_i + B_i(t) - B_{i-1}(t) - \frac{1}{2}L_{i+1}(t) + L_i(t) - \frac{1}{2}L_{i-1}(t), \quad 3 \leq i \leq N.
\end{aligned} \tag{1.3}$$

The model described by equations (1.2) (with gaps evolving as in (1.3)), which we call the *inert drift Atlas model*, lies at the interface of two well-studied classes of interacting particle systems: *inert drift models* and *rank-based diffusions*, which we summarize below.

1.2. Previous work. The case where $N = 1$ (namely the two particle system) with $g = 0$ was analyzed in [30], which initiated the study of inert drift models. It was shown there that the inert particle progressively gains momentum from the local time interactions and eventually escapes the Brownian particle (no further collisions). When $g > 0$, [4] showed that the two particles never escape each other. Among other results, the paper showed that the velocity of the inert particle and the gap between the two particles jointly converge in total variation distance to an explicit stationary distribution having a product form density (no rates of convergence were obtained). The two particle model with gravitation and fluid viscosity was investigated in [2]. In [10], an inert drift model was considered where a particle moves as a diffusion process inside a bounded smooth domain and acquires inert drift when it hits the boundary of the domain. It was shown that the position of the particle and the cumulative inert drift have a product form stationary measure, which is unique under suitable conditions. A variety of related inert drift models have been studied in [12, 13, 39]. When the term $\sum_{i=1}^N \ell_i(t)$ in (1.1) is replaced by $N^{-1} \sum_{i=1}^N \ell_i(t)$ (mean field type interaction), the asymptotic behavior as $N \rightarrow \infty$ has been analyzed in [8, 9] where results on hydrodynamic limits and propagation of chaos have been obtained. Recently, unexpected connections have appeared between inert drift models and diffusion limits of load balancing systems like the Join-the-shortest-queue policy in heavy traffic [5, 6, 24]. More precisely, the joint evolution of the diffusion-scaled number of idle servers and busy servers converges in distribution to a diffusion that resembles the two particle inert drift system with linear drift. Consequently, there are several common themes at the technical level between [2, 4] and [5, 6]. Brownian particle systems of the form studied in the current work also arise as diffusion approximations of certain types of queuing systems in which each queue has the same finite capacity which is dynamically controlled in a manner that the increase in capacity is proportional to net job loss due to capacity constraints. In this model, currently under investigation, the individual queues play the role of Brownian particles whereas the dynamically changing queue capacity threshold represents the massive inert particle.

In a somewhat different vein, inspired by problems in mathematical finance, the study of rank-based diffusions [3, 7, 16, 34–37] have gained a lot of attention in recent years. These models consist of a collection of particles on the real line which evolve as diffusion processes where the drift and diffusivity of each particle is a function of its relative rank in the system. Closest in spirit to our model is the *Atlas model* where the lowest ranked

particle at any time moves as a Brownian motion with constant upward drift while the remaining particles evolve as standard Brownian motions (with zero drift).

1.3. Analytical challenges. The Atlas model and the model considered here are examples of particle systems with topological interactions in the terminology of [14]. In such particle systems, interactions between particles are determined by their relative positions. In particular, in both the Atlas model and in the particle system considered here, the lowest particle has different dynamical properties. Specifically, in the Atlas model the lowest particle gets a constant upward drift whereas in the model considered here the lowest particle experiences an *inert drift*. However there are some important differences between the two models. Unlike the Atlas model, where the collision local time of the lowest two particles enters directly in the position evolution of the lowest particle, here this local time impacts the velocity of the lowest particle. Indeed, this collision local time is the source of the inert drift of the lowest particle. Furthermore, there is no Brownian noise in the equation for $X_{(0)}$ in (1.2), unlike in the Atlas model. This results in the deterministic evolution of the velocity process in time periods with no collisions, making the full system, whose long-time behavior is of interest, non-elliptic (in fact, the driving diffusion process in the interior of the domain is not even hypoelliptic). More precisely, the law of $(V(t), Z_1(t), \dots, Z_N(t))$ for any $t > 0$ does not have a density with respect to Lebesgue measure, for general initial conditions. Also, we find that, unlike the Atlas model, the system considered here is not reversible. Hence, standard techniques for studying ergodicity behavior of elliptic diffusion processes cannot be applied, and one needs new methods. As noted above, inert two-particle systems have been studied in several previous works, however the current work is the first to study the ergodicity properties of a general N -particle system. There are fundamental differences in system behavior as one goes from $N = 1$ to $N > 1$ which make the study of ergodicity behavior significantly more demanding. In particular, as is crucially exploited in [2, 4], in the $N = 1$ case, there is a basic *regenerative structure* arising from the fact that at points of decrease of the velocity process, the remaining state coordinate, namely the one corresponding to Z_1 , is fully determined (in fact equal to 0). In the general N -particle system there is no such simple regenerative structure since, although the first gap coordinate Z_1 is once again 0 at points of decrease of V , the remaining coordinates, namely Z_2, \dots, Z_N can be arbitrary.

1.4. Main contributions. We now briefly describe the main contributions of this work. Since the system is not hypoelliptic, one cannot apply standard existing theory to argue uniqueness of invariant measures. Our first main result says that the Markov process $(V, \mathbf{Z}) = (V, Z_1, \dots, Z_N)$ admits at most one stationary distribution. We then produce an explicit stationary distribution for the system and together the two results (see Theorems 2.3 and 2.4) prove existence and uniqueness of stationary distributions of (V, \mathbf{Z}) . We in fact show that the unique stationary distribution takes a product form whose first component (corresponding to the velocity coordinate) is Gaussian and remaining are Exponential (see Theorem 2.4 for the precise form). In the case $N = 1$, a Gaussian-Exponential product form stationary distribution has appeared in previous works [4, 10, 39]; however, this is the first work that finds such a product form structure for a general N -particle system. This stationary distribution also has striking similarities with the Atlas model where the stationary distribution is a product of exponentials with rates decreasing with the ranks of the particles (see, for example, [34, Theorem 8]).

We next study the rate of convergence to stationarity. In Theorem 2.5, we show that the distribution of $(V(t), Z_1(t), \dots, Z_N(t))$ converges to equilibrium exponentially fast (exponential ergodicity) as $t \rightarrow \infty$. To the best of our knowledge, this is the first result on exponential ergodicity for any type of non-hypoelliptic reflected diffusion in dimensions higher than 2.

Finally in Theorem 2.6 we establish some law of large numbers type results. In particular, it is shown that the whole system ‘drifts’ to infinity at speed g/N . Although this is an intuitive result to expect, our proof crucially hinges on the rather technical result on exponential moments of return times to certain compact sets that form the basis of the exponential ergodicity proof. We also find, somewhat surprisingly, that the intensity of collisions when $N \geq 3$ is maximum, in a certain sense, between the first two Brownian particles (rather than between the inert and the first Brownian particle); see Remark 2.7.

1.5. Approach. A common approach to proving ergodicity or exponential rates of convergence to stationarity for diffusions in domains is by constructing a suitable Lyapunov function by analyzing the interplay between the “interior drift vector field” and the reflection vector field (cf. [1, 11, 21]). For example, in polyhedral domains with constant (oblique) reflection on each face of the boundary, the key insight in the construction of a Lyapunov function is that the drift vector field for stable systems must lie in the interior of the cone generated by the negatives of the reflection directions. Note that \mathbf{Z} is a reflected diffusion in the positive orthant \mathbb{R}_+^N with constant oblique reflection at each face. The interior drift of this process is $V(t)\mathbf{e}_1$, where \mathbf{e}_1 is the unit vector with 1 in the first coordinate. Due to the complicated dynamics of V , that includes in particular the local time for the first gap process Z_1 , its behavior in relation to the reflection field seems hard to analyze which makes a direct construction of an explicit form Lyapunov function (as in the above cited works) hard.

In this work we instead take a pathwise approach. The stability in the particle system studied here arises as a result of interplay between the intersection local times for the various particles in the system. This interplay is distilled in Lemma 7.10 which identifies a stabilizing ‘singular’ drift that prevents the gaps between the particles from being too large. This key lemma allows us to prove the finiteness of exponential moments of hitting times to certain compact sets by analyzing excursions of the process between suitably chosen stopping times (see Sects. 7.3–7.6). In conjunction with results of [18] (see Proposition 7.16 (a)), this analysis furnishes a general abstract form Lyapunov function, given in terms of exponential moments of these hitting times, which is key in the proof of exponential ergodicity. Another important ingredient in our proofs is establishing a certain minorization estimate (see Proposition 7.16 (b)). For hypoelliptic diffusions such an estimate follows readily from the existence of a density for the process at each time $t > 0$. However, in our case, establishing a suitable minorization bound involves substantial work and a careful exploitation of the properties of the collision local times of the particles in the system. The proof of this estimate, which uses an intricate and novel pathwise analysis, is the topic of Sect. 4.

1.6. Future directions. The current work is the first step in our program of analyzing high-dimensional reflected diffusions with inert drift type interactions. The natural next step will be to investigate ergodicity properties of the infinite-dimensional analogue of our model. The corresponding vector of velocity and gap processes is expected to have at

least one stationary distribution, given by the $N \rightarrow \infty$ limit of (2.3) below. It is unclear if this is the unique stationary distribution. Analogy with the Atlas model suggests infinitely many stationary distributions, each with a non-trivial domain of attraction [3, 16, 36]. Another interesting question concerns the study of hydrodynamic limits of empirical occupation measures of the system and relate them to the path asymptotics of the bottom k particles for $k \in \mathbb{N}$ (see [17] for related results on the Atlas model). Both these directions are currently under investigation.

1.7. Notation and preliminaries. The following notation will be used. For $d \in \mathbb{N}$ and $T > 0$, we denote by $\mathcal{C}([0, T] : \mathbb{R}^d)$ (resp. $\mathcal{C}([0, \infty) : \mathbb{R}^d)$) the space of continuous functions on $[0, T]$ (resp. $[0, \infty)$) with values in \mathbb{R}^d , equipped with the topology of uniform convergence (resp. local uniform convergence). The spaces $\mathcal{C}([0, T] : \mathbb{R}_+^d)$ (resp. $\mathcal{C}([0, \infty) : \mathbb{R}_+^d)$) of continuous functions with values in the nonnegative orthant \mathbb{R}_+^d are defined similarly. For $t \in [0, \infty)$ and $f \in \mathcal{C}([0, \infty) : \mathbb{R}^d)$, we define $\|f\|_t \doteq \sup_{0 \leq s \leq t} |f(s)|$, where $|\cdot|$ is the Euclidean norm on \mathbb{R}^d . Borel σ -fields on a metric space S will be denoted as $\mathcal{B}(S)$. Inequalities for vectors and vector-valued random variables are understood to be coordinatewise. An open set $G \subset \mathbb{R}^d$ is said to have a \mathcal{C}^2 boundary if each point in ∂G has a neighborhood in which ∂G is the graph of a \mathcal{C}^2 function of $d - 1$ of the coordinates (cf. [25, Section 6.2]). Throughout λ will denote the Lebesgue measure on a subset of a Euclidean space whose dimension will be clear from the context.

The following elementary estimate will be used several times. Suppose for $m \in \mathbb{N}$, $\tilde{B}_1, \dots, \tilde{B}_m$ are mutually independent Brownian motions and $\alpha_1, \dots, \alpha_m \in \mathbb{R}_+$. Let $\tilde{B}_i^*(t) \doteq \sup_{0 \leq s \leq t} |\tilde{B}_i(s)|$. Then there are $\varrho_1, \varrho_2 \in (0, \infty)$, such that

$$\mathbb{E} \left(e^{u \sum_{i=1}^m \alpha_i \tilde{B}_i^*(t)} \right) \leq \varrho_1 e^{\varrho_2 u^2 t} \text{ for all } t \geq 0 \text{ and } u \geq 0. \quad (1.4)$$

The dependence of the constants ϱ_1, ϱ_2 on m and α_i will usually be suppressed from the notation.

In the next section, we outline our main results. The organization of the paper is summarized at the end of the section.

2. Main Results

Define the $N \times N$ matrix

$$R \doteq \begin{pmatrix} 1 & -\frac{1}{2} & 0 & 0 & \cdots & 0 \\ -1 & 1 & -\frac{1}{2} & 0 & \cdots & 0 \\ 0 & -\frac{1}{2} & 1 & -\frac{1}{2} & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & \cdots & \cdots & -\frac{1}{2} & 1 \end{pmatrix}.$$

It is easily checked that the matrix $U = I - R$ has the property that U^T is a transient, substochastic matrix and thus has spectral radius strictly less than 1. Consequently, R is invertible and $W = R^{-1}$ can be written as an infinite sum of matrices with nonnegative entries. The Skorohod problem associated with such matrices R has been well studied and the following result is well known cf. [19, 20, 26].

We denote by $\mathcal{C}_0([0, \infty) : \mathbb{R}^N)$ the space of continuous functions $f : [0, \infty) \rightarrow \mathbb{R}^N$ such that $f(0) \geq 0$.

Proposition 2.1. *To each $x \in \mathcal{C}_0([0, \infty) : \mathbb{R}^N)$ there is a unique pair $(\eta, y) \in \mathcal{C}([0, \infty) : \mathbb{R}_+^N) \times \mathcal{C}([0, \infty) : \mathbb{R}_+^N)$ such that,*

- (i) for all $t \geq 0$, $y(t) = x(t) + R\eta(t)$,
- (ii) For each $i \in \{1, \dots, N\}$, (a) $\eta_i(0) = 0$, (b) $\eta_i(t)$ is non-decreasing in t , (c) $\int_0^\infty y_i(t) d\eta_i(t) = 0$.

The pair (η, y) is called the solution to the Skorokhod problem for x with respect to R . The map $\Gamma : \mathcal{C}_0([0, \infty) : \mathbb{R}^N) \rightarrow \mathcal{C}([0, \infty) : \mathbb{R}_+^N) \times \mathcal{C}([0, \infty) : \mathbb{R}_+^N)$ given by

$$\Gamma(x) = (\eta, y) = (\Gamma_1(x), \Gamma_2(x))$$

is Lipschitz in the sense that there is a $c_\Gamma \in (0, \infty)$ such that for $x, x' \in \mathcal{C}_0([0, \infty) : \mathbb{R}^N)$ and $t < \infty$,

$$\|\Gamma_1(x) - \Gamma_1(x')\|_t + \|\Gamma_2(x) - \Gamma_2(x')\|_t \leq c_\Gamma \|x - x'\|_t.$$

For $x \in \mathcal{C}_0([0, \infty) : \mathbb{R}^N)$, we occasionally write $\Gamma_1(x) = (\Gamma_{11}(x), \dots, \Gamma_{N1}(x))$.

The following result gives strong existence and uniqueness for the system of equations in (1.3). Proof is given in Sect. 5. Let

$$A \doteq \begin{pmatrix} 1 & 0 & 0 & \cdots & 0 & 0 \\ -1 & 1 & 0 & \cdots & 0 & 0 \\ 0 & -1 & 1 & \cdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \ddots & 1 & 0 \\ 0 & 0 & 0 & \cdots & -1 & 1 \end{pmatrix}.$$

Theorem 2.2. *Let $(\bar{\Omega}, \bar{\mathcal{F}}, \{\bar{\mathcal{F}}_t\}_{t \geq 0}, \bar{\mathbb{P}})$ be a filtered probability space on which are given N mutually independent standard real $\bar{\mathcal{F}}_t$ -Brownian motions B_1, \dots, B_N and $\bar{\mathcal{F}}_0$ -measurable random variables V^0 and $\mathbf{Z}^0 = (Z_1^0, \dots, Z_N^0)$ with values in \mathbb{R} and \mathbb{R}_+^N respectively. Then there is a continuous, $\bar{\mathcal{F}}_t$ -adapted, stochastic process $(V(t), Z_1(t), \dots, Z_N(t))_{0 \leq t < \infty}$ with values in $\mathbb{R} \times \mathbb{R}_+^N$ such that, for all $t \geq 0$,*

$$\begin{aligned} V(t) &= V^0 + gt - L_1(t), \\ \mathbf{Z}(t) &= \Gamma_2 \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^\cdot V(s) ds + \mathbf{A}\mathbf{B}(\cdot) \right) (t), \\ L_1(t) &= \Gamma_{11} \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^\cdot V(s) ds + \mathbf{A}\mathbf{B}(\cdot) \right) (t), \end{aligned} \tag{2.1}$$

where $\mathbf{B} = (B_1, \dots, B_N)'$ and $\mathbf{Z} = (Z_1, \dots, Z_N)'$. Furthermore, if $(\tilde{V}(t), \tilde{Z}_1(t), \dots, \tilde{Z}_N(t))$ is another such process then

$$(\tilde{V}(t), \tilde{Z}_1(t), \dots, \tilde{Z}_N(t)) = (V(t), Z_1(t), \dots, Z_N(t)) \text{ for all } t \geq 0, \text{ a.s.}$$

We remark that, with \mathbf{Z} , and V as in the theorem, letting

$$\mathbf{L}(t) = (L_1(t), \dots, L_N(t)) = \Gamma_1 \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^t V(s) ds + \mathbf{A}\mathbf{B}(\cdot) \right) (t) \text{ and } L_{N+1}(t) = 0,$$

we have that the following system of equations holds:

$$\begin{aligned} V(t) &= V^0 + gt - L_1(t), \\ Z_1(t) &= Z_1^0 + B_1(t) - \int_0^t V(s) ds - \frac{1}{2}L_2(t) + L_1(t), \\ Z_2(t) &= Z_2^0 + B_2(t) - B_1(t) - \frac{1}{2}L_3(t) + L_2(t) - L_1(t), \\ Z_i(t) &= Z_i^0 + B_i(t) - B_{i-1}(t) - \frac{1}{2}L_{i+1}(t) + L_i(t) - \frac{1}{2}L_{i-1}(t), \quad 3 \leq i \leq N. \end{aligned} \quad (2.2)$$

Consider the path space $\Omega^* = \mathcal{C}([0, \infty) : \mathbb{R}^N \times \mathbb{R} \times \mathbb{R}_+^N)$, \mathcal{F}^* the corresponding Borel σ -field on Ω^* . We also consider the space $(\Omega, \mathcal{F}) \doteq (\mathcal{C}([0, \infty) : \mathbb{R} \times \mathbb{R}_+^N), \mathcal{B}(\mathcal{C}([0, \infty) : \mathbb{R} \times \mathbb{R}_+^N)))$. On these two measurable spaces we denote, for $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, by $\mathbb{P}_{(v, \mathbf{z})}^*$ [resp. $\mathbb{P}_{(v, \mathbf{z})}$], the probability measures induced by $(\mathbf{B}, V, \mathbf{Z})$ [resp. (V, \mathbf{Z})] where (V, \mathbf{Z}) is the solution of (2.1) when $(V^0, \mathbf{Z}^0) = (v, \mathbf{z})$ a.s. Then from the unique solvability in the above theorem it follows that $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$ defines a strong Markov family. The next result concerns the stationary distribution of this Markov family.

Theorem 2.3. *There is a unique stationary distribution for the Markov family $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$.*

In fact this unique stationary distribution takes an explicit product form as given by the theorem below. Consider the probability measure π on $\mathbb{R} \times \mathbb{R}_+^N$ given by the formula:

$$\pi(dv, dz_1, \dots, dz_N) \doteq c_\pi e^{-(v - \frac{g}{N})^2} \prod_{i=1}^N e^{-2g \left(\frac{N-i+1}{N} \right) z_i} dv dz_1 \dots, dz_N, \quad (2.3)$$

where c_π is the normalization constant.

Theorem 2.4. *The probability measure π defined in (2.3) is the unique stationary distribution of $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$.*

Note that while Theorem 2.4 implies Theorem 2.3, we proceed by first showing that there exists at most one stationary distribution (Theorem 5.2). The existence and explicit form of the stationary distribution is subsequently exhibited (in Sect. 6) by proving that the density of π solves the partial differential equation (with boundary conditions) arising from the basic adjoint relationship [see (6.2)–(6.5)]. We have therefore separated out these results for clarity of exposition.

Our third result gives exponential ergodicity of the Markov process. Write an element $\omega \in \Omega^*$ [resp. $\omega \in \Omega$] as $\omega = (\beta, v, \zeta)$ [resp. $\omega = (v, \zeta)$], where $\beta \in \mathcal{C}([0, \infty) : \mathbb{R}^N)$, $v \in \mathcal{C}([0, \infty) : \mathbb{R})$ and $\zeta \in \mathcal{C}([0, \infty) : \mathbb{R}_+^N)$. For $t \in [0, \infty)$, abusing notation, denote the coordinate processes $\mathbf{B}(t)$, $V(t)$ and $\mathbf{Z}(t)$ on $(\Omega^*, \mathcal{F}^*)$ [resp. $V(t)$ and $\mathbf{Z}(t)$ on (Ω, \mathcal{F})] by the formulae

$$\mathbf{B}(t)(\omega) = \beta(t), \quad V(t)(\omega) = v(t), \quad \mathbf{Z}(t)(\omega) = \zeta(t), \quad t \geq 0.$$

Also, we will write $B_i(t)$ and $Z_i(t)$ respectively for the projections of $\mathbf{B}(t)$ and $\mathbf{Z}(t)$ onto their i^{th} coordinates. Consider the transition probability kernel of the Markov family $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$ defined as

$$\mathbb{P}^t((v, \mathbf{z}), A) \doteq \mathbb{P}_{(v, \mathbf{z})}((V(t), \mathbf{Z}(t)) \in A), \quad t \geq 0, (v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N, A \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N).$$

Also, for a bounded and measurable $\phi : \mathbb{R} \times \mathbb{R}_+^N \rightarrow \mathbb{R}$ we write

$$\mathbb{P}^t((v, \mathbf{z}), \phi) \doteq \int_{\mathbb{R} \times \mathbb{R}_+^N} \phi(\tilde{v}, \tilde{\mathbf{z}}) \mathbb{P}^t((v, \mathbf{z}), d\tilde{v} \times d\tilde{\mathbf{z}}).$$

Similarly, for ϕ as above, $\pi(\phi) \doteq \int \phi(\tilde{v}, \tilde{\mathbf{z}}) \pi(d\tilde{v} \times d\tilde{\mathbf{z}})$. The following theorem shows the convergence of the transition probability kernel to the stationary distribution in the total variation distance at an exponential rate. Denote by BM_1 the class of all measurable $\phi : \mathbb{R} \times \mathbb{R}_+^N \rightarrow \mathbb{R}$ such that $\sup_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N} |\phi(v, \mathbf{z})| \leq 1$.

Theorem 2.5. *There is a $\gamma \in (0, 1)$ and, for every $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, $\kappa(v, \mathbf{z}) \in (0, \infty)$, such that for all $t \geq 0$,*

$$\sup_{\phi \in \text{BM}_1} |\mathbb{P}^t((v, \mathbf{z}), \phi) - \pi(\phi)| \leq \kappa(v, \mathbf{z}) \gamma^t.$$

We note here that the proof of exponential ergodicity proceeds through establishing finiteness of exponential moments of certain hitting times. This, in turn, provides the tightness required to furnish an independent proof of existence of a stationary distribution.

Finally, we prove a strong law of large numbers type result for the system. Recall the ranked particle system $\{X_{(i)}(\cdot)\}_{0 \leq i \leq N}$ from (1.2). This process can be constructed on $(\Omega^*, \mathcal{F}^*, \mathbb{P}_{(v, \mathbf{z})}^*)$ for any $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ by solving the system of equations in (2.1) (or equivalently (2.2)), whose unique pathwise solutions are guaranteed by Theorem 2.2, and then defining $X_{(i)}$ by the right side of (1.2).

Theorem 2.6. *For any $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, the following limits hold $\mathbb{P}_{(v, \mathbf{z})}^*$ -almost surely:*

$$\lim_{t \rightarrow \infty} \frac{X_{(i)}(t)}{t} = \frac{g}{N}, \quad 0 \leq i \leq N, \quad (2.4)$$

$$\lim_{t \rightarrow \infty} \frac{L_1(t)}{t} = g, \quad (2.5)$$

$$\lim_{t \rightarrow \infty} \frac{L_i(t)}{t} = \frac{2(N-i+1)g}{N}, \quad 2 \leq i \leq N. \quad (2.6)$$

Remark 2.7. It is natural to expect that the gaps become larger in some sense as one moves away from the inert particle. This heuristic is quantified in the stochastic monotonicity of the stationary gaps displayed in (2.3). From this, it might appear that the growth rate of the local time $L_i(t)$ (which quantifies the intensity of collisions between the $(i-1)^{\text{th}}$ and i^{th} particle) with t should decrease as i increases from 1 to N . However, Theorem 2.6 shows that for $N \geq 3$, L_2 grows at a faster rate than L_1 and the expected decrease in rates holds from $i = 2$ onwards. Hence, perhaps surprisingly, particles indexed 1 and 2 collide ‘more often’ than particles 0 and 1 as time progresses.

2.1. Organization. The rest of the paper is organized as follows. In Sect. 3, we provide the proof of Theorem 2.2. In Sect. 4, we show a technical estimate which will be integral to the proofs of our main results. In Sect. 5, we show that there is at most one stationary distribution (Theorem 5.2). In Sect. 6, we prove Theorem 2.4. Together, these two results also establish Theorem 2.3. In Sect. 7, we give the proof of Theorem 2.5. Proofs of several technical results stated in Sect. 7 (without proof) are provided in Sect. 8. In Sect. 9, we establish Theorem 2.6.

3. Existence and Uniqueness of the Process

In this section, we prove Theorem 2.2. The proof uses the Lipschitz property in Proposition 2.1, and a standard Picard iteration scheme. We provide a sketch. Fix $T < \infty$. Let (V^0, \mathbf{Z}^0) be as in the statement of the theorem. Define, for $n \in \mathbb{N}_0$, continuous $\tilde{\mathcal{F}}_t$ -adapted $\mathbb{R} \times \mathbb{R}_+^N \times \mathbb{R}_+^N$ valued processes $\{(V^{(n)}(t), \mathbf{Z}^{(n)}(t), \mathbf{L}^{(n)}(t))\}_{0 \leq t \leq T}$, recursively, as follows. Let

$$V^{(0)}(t) \doteq V^0, \quad \mathbf{Z}^{(n)}(t) \doteq \mathbf{Z}^0, \quad \mathbf{L}^{(n)}(t) = 0, \quad 0 \leq t \leq T.$$

Having defined $\{(V^{(k)}(t), \mathbf{Z}^{(k)}(t), \mathbf{L}^{(k)}(t))\}_{0 \leq t \leq T}$ for $k = 0, \dots, n-1$, define

$$\begin{aligned} \mathbf{Z}^{(n)}(t) &= \Gamma_2 \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^t V^{(n-1)}(s) ds + \mathbf{A}\mathbf{B}(\cdot) \right) (t), \\ \mathbf{L}^{(n)}(t) &= \Gamma_1 \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^t V^{(n-1)}(s) ds + \mathbf{A}\mathbf{B}(\cdot) \right) (t), \\ V^{(n)}(t) &= V^0 + gt - L_1^{(n)}(t), \end{aligned} \tag{3.1}$$

where $L_1^{(n)}(t)$ is the first coordinate of $\mathbf{L}^{(n)}(t)$.

From the Lipschitz property in Proposition 2.1 it follows that, for any $n \geq 2$, and $t \in [0, T]$,

$$\|\mathbf{Z}^{(n)} - \mathbf{Z}^{(n-1)}\|_t + \|\mathbf{L}^{(n)} - \mathbf{L}^{(n-1)}\|_t \leq c_\Gamma \int_0^t \|V^{(n-1)} - V^{(n-2)}\|_s ds$$

and

$$\begin{aligned} \|V^{(n)} - V^{(n-1)}\|_t &= \|L_1^{(n)} - L_1^{(n-1)}\|_t \leq \|\mathbf{L}^{(n)} - \mathbf{L}^{(n-1)}\|_t \\ &\leq c_\Gamma \int_0^t \|V^{(n-1)} - V^{(n-2)}\|_s ds. \end{aligned}$$

Letting

$$\Delta_n(t) \doteq \|\mathbf{Z}^{(n)} - \mathbf{Z}^{(n-1)}\|_t + \|\mathbf{L}^{(n)} - \mathbf{L}^{(n-1)}\|_t + \|V^{(n)} - V^{(n-1)}\|_t,$$

we have for $n \geq 2$ and $t \in [0, T]$, $\Delta_n(t) \leq c_\Gamma \int_0^t \Delta_{n-1}(s) ds$. Now a standard argument shows that, a.s., $(V^{(n)}, \mathbf{Z}^{(n)}, \mathbf{L}^{(n)})$ is a Cauchy sequence in $\mathcal{C}([0, T] : \mathbb{R} \times \mathbb{R}_+^N \times \mathbb{R}_+^N)$.

Let $(V, \mathbf{Z}, \mathbf{L})$ denote the limit. It is easy to verify that this is a $\tilde{\mathcal{F}}_t$ -adapted process. Furthermore, sending $n \rightarrow \infty$ in (3.1) we see that $(V, \mathbf{Z}, \mathbf{L})$ solve, for $0 \leq t \leq T$,

$$\begin{aligned} \mathbf{Z}(t) &= \Gamma_2 \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^\cdot V(s) ds + A\mathbf{B}(\cdot) \right) (t), \\ \mathbf{L}(t) &= \Gamma_1 \left(\mathbf{Z}^0 - \mathbf{e}_1 \int_0^\cdot V(s) ds + A\mathbf{B}(\cdot) \right) (t), \\ V(t) &= V^0 + gt - L_1(t), \end{aligned} \quad (3.2)$$

where $L_1(t)$ is the first coordinate of $\mathbf{L}(t)$. In particular, (V, \mathbf{Z}) is a solution of (2.1). Since $T > 0$ is arbitrary this proves the first part of the theorem.

Now suppose that $(V, \mathbf{Z}, \mathbf{L})$ and $(\tilde{V}, \tilde{\mathbf{Z}}, \tilde{\mathbf{L}})$ are two continuous $\mathbb{R} \times \mathbb{R}_+^N \times \mathbb{R}_+^N$ valued $\tilde{\mathcal{F}}_t$ -adapted processes that solve (3.2). Then, for $t \in [0, T]$,

$$\begin{aligned} \|\mathbf{Z} - \tilde{\mathbf{Z}}\|_t + \|\mathbf{L} - \tilde{\mathbf{L}}\|_t &\leq c_\Gamma \int_0^t \|V - \tilde{V}\|_s ds = c_\Gamma \int_0^t \|L_1 - \tilde{L}_1\|_s ds \\ &\leq c_\Gamma \int_0^t (\|\mathbf{Z} - \tilde{\mathbf{Z}}\|_s + \|\mathbf{L} - \tilde{\mathbf{L}}\|_s) ds. \end{aligned}$$

Using Grönwall's lemma, it then follows that $\mathbf{Z}(t) = \tilde{\mathbf{Z}}(t)$ and $\mathbf{L}(t) = \tilde{\mathbf{L}}(t)$ for all $t \in [0, T]$ a.s. which also says that $V(t) = \tilde{V}(t)$ for all $t \in [0, T]$ a.s. The result follows. \square

4. A Minorization Estimate

In this section we will establish a minorization estimate for the transition probability kernel $\mathbb{P}^t((v, \mathbf{z}), A)$ introduced in the last section. This estimate will be a key ingredient in the proofs of Theorems 2.3 and 2.5. The deterministic motion of the bottom (inert) particle when $Z_1 > 0$ results in very singular behavior of our diffusion process manifested, in particular, by the lack of a density of $(V(t), \mathbf{Z}(t))$ with respect to Lebesgue measure for any $t > 0$ when the initial condition satisfies $Z_1(0) > 0$. Hence, one cannot use standard techniques for establishing a minorization condition for elliptic (or hypoelliptic) diffusions. We take a pathwise approach here by analyzing a suitable collection of driving Brownian paths to obtain a sub-density of the form described in Theorem 4.1. This is done by first ‘removing the drift’ by applying Girsanov’s Theorem and analyzing the simpler system given by gaps between N ordered Brownian motions and the local time at zero of the bottom particle. This, along with an appropriate control of the Radon-Nikodym derivative, yields the desired result.

Let

$$\varsigma \doteq \frac{1}{128}, \quad \varsigma^* \doteq \varsigma + \left(\frac{1}{63} - \frac{1}{64} \right).$$

Theorem 4.1. *Let $C = [0, \frac{g}{128}] \times [\frac{g}{2}, g]^N$. There exists $D \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ such that $\lambda(D \cap C) > 0$, and such that for each $(v, \mathbf{z}) \in [0, \frac{g}{128}] \times (0, \infty) \times \mathbb{R}_+^{N-1}$, there is a $K_{(v, \mathbf{z})} \in (0, \infty)$ so that*

$$\inf_{t \in [\varsigma, \varsigma^*]} \mathbb{P}^t((v, \mathbf{z}), S) \geq K_{(v, \mathbf{z})} \lambda(S \cap D) \text{ for every } S \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N). \quad (4.1)$$

Moreover, the map $(v, \mathbf{z}) \mapsto K_{(v, \mathbf{z})}$ is measurable and for any $0 \leq a_i < b_i < \infty$, $1 \leq i \leq N$, $a_1 > 0$, with $\bar{A} = [0, \frac{g}{128}] \times [a_1, b_1] \times \cdots \times [a_N, b_N]$,

$$\bar{K}_{\bar{A}} \doteq \inf_{(v, \mathbf{z}) \in \bar{A}} K_{(v, \mathbf{z})} > 0.$$

In proving the above it will be convenient to introduce a probability measure $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ that is mutually absolutely continuous to $\mathbb{P}_{(v, \mathbf{z})}^*$ and which is somewhat simpler to analyze. This measure corresponds to the law of the processes $(\mathbf{B}, V, \mathbf{Z})$ given as in (2.2) but with V on the right side of equation for Z_1 replaced by the 0 process. Recall the path space $(\Omega^*, \mathcal{F}^*)$ and the coordinate processes $(\mathbf{B}, V, \mathbf{Z})$ given on this space. Let $\{\mathcal{F}_t^*\}_{t \geq 0}$ be the filtration generated by these coordinate processes. For $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ let $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ be the probability measure on $(\Omega^*, \mathcal{F}^*)$ such that under $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ the following hold:

- (i) \mathbf{B} is the standard N -dimensional \mathcal{F}_t^* -Brownian motion.
- (ii) For each $t \in [0, \infty)$, with $L(t) = \Gamma_1(\mathbf{z} + A\mathbf{B}(\cdot))(t)$,

$$\mathbf{Z}(t) = \mathbf{z} + A\mathbf{B}(t) + RL(t), \quad V(t) = v + gt - L_1(t). \quad (4.2)$$

4.1. Outline of proof. The proof of Theorem 4.1 is organized as follows. In Lemma 4.2, we establish a version of Novikov's criterion which allows us to relate $\mathbb{P}_{(v, \mathbf{z})}^*$ to $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ via Girsanov's Theorem. In Corollary 4.3, we use the preceding lemma to invoke Girsanov's Theorem and make explicit the relation between the two measures.

We next prove a number of technical results in support of Theorem 4.1. In Lemma 4.4, we establish a minorization condition for a 'killed' version of \mathbf{Z} under law $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$, when $\mathbf{Z}(0)$ lies in a certain compact set F . In Lemma 4.5, we prove the existence of a subdensity for the supremum of Brownian motion over a compact time interval under certain constraints on its infimum and final location. This supremum, in turn, is connected to the local time L_1 via the Skorohod map. As under law $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$, existence of a subdensity at a fixed time for (\mathbf{Z}, V) is implied by that for (\mathbf{Z}, L_1) (see (4.2)), the above two lemmas are crucial in proving Theorem 4.1. Lemmas 4.6 and 4.7 provide a version of the 'support theorem' where a tractable event in terms of the driving Brownian motions is constructed under which the gap process \mathbf{Z} at a prescribed time $\varsigma/4$ lies in F almost surely under $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$.

Lastly, we prove Theorem 4.1. Using the strong Markov property, we analyze the process pathwise between appropriately chosen stopping times. We first let Z_1 hit zero at time τ_1 after which, under the event on the driving Brownian motions described in Lemma 4.7, the local time L_1 lies in a given Borel set and the gaps \mathbf{Z} lie in the set F introduced in Lemma 4.4 at time $\tau_1 + \varsigma/4$. Theorem 4.1 now follows upon combining this and the minorization condition on the killed gap process obtained in Lemma 4.4, which is used in analyzing the subsequent process path.

4.2. Proof of Theorem 4.1. In order to relate $\mathbb{P}_{(v, \mathbf{z})}^*$ with $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ we establish the following integrability property which will be used to verify a variation of Novikov's criterion. In the following, $\tilde{\mathbb{E}}_{(v, \mathbf{z})}^*$ denotes the expectation under the probability measure $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$. Under

$\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$, the local times L_i , $1 \leq i \leq N$ (and with $L_{N+1} = 0$) have the following pathwise representation:

$$\begin{aligned} L_1(t) &= \sup_{s \leq t} (-z_1 + \frac{1}{2} L_2(s) - B_1(s))^+, \\ L_2(t) &= \sup_{s \leq t} (-z_2 + \frac{1}{2} L_3(s) + L_1(s) + B_1(s) - B_2(s))^+, \\ L_i(t) &= \sup_{s \leq t} (-z_i + \frac{1}{2} (L_{i+1}(s) + L_{i-1}(s)) + B_{i-1}(s) - B_i(s))^+, \quad i = 3, \dots, N. \end{aligned} \quad (4.3)$$

Lemma 4.2. *For every $c \in (0, \infty)$ and $r \in \mathbb{N}$, there is a $m \in \mathbb{N}$ such that with $t_k = k/m$, $k = 0, 1, \dots, rm - 1$, for each $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,*

$$\tilde{\mathbb{P}}_{(v, \mathbf{z})}^* e^{\frac{c}{2} \int_{t_k}^{t_{k+1}} V(s)^2 ds} < \infty.$$

Proof. Fix $c \in (0, \infty)$ and $r \in \mathbb{N}$. Also, fix $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$. All equalities and inequalities in the proof are almost sure with respect to the measure $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$.

Note that, for $t \geq 0$, by (4.3),

$$\begin{aligned} L_1(t) &\leq \frac{1}{2} L_2(t) + \sup_{s \leq t} (-B_1(s)), \\ L_2(t) &\leq \frac{1}{2} L_3(t) + L_1(t) + \sup_{s \leq t} (B_1(s) - B_2(s)), \\ L_i(t) &\leq \frac{1}{2} (L_{i+1}(t) + L_{i-1}(t)) + \sup_{s \leq t} (B_{i-1}(s) - B_i(s)), \quad i = 3, \dots, N. \end{aligned}$$

Define

$$\begin{aligned} B_1^*(t) &= \sup_{s \leq t} (-B_1(s)) \\ B_i^*(t) &= \sup_{s \leq t} (B_{i-1}(s) - B_i(s)), \quad \text{for } i = 2, \dots, N \\ \mathbf{B}^*(t) &= (B_1^*(t), \dots, B_N^*(t)). \end{aligned} \quad (4.4)$$

Recall the matrices $U = I - R$ and $W = (I - U)^{-1}$. Then, it is easy to verify that

$$U = \begin{pmatrix} 0 & \frac{1}{2} & 0 & 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & \frac{1}{2} & 0 & 0 & \dots & 0 & 0 \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} & 0 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & \dots & \dots & \dots & \frac{1}{2} & 0 \end{pmatrix}. \quad (4.5)$$

and so from the above inequalities we can write, for $t \geq 0$,

$$\mathbf{L}(t) \leq U\mathbf{L}(t) + \mathbf{B}^*(t).$$

In particular, recalling that W can be written as an infinite sum of matrices with non-negative entries, we have that

$$L_1(t) \leq (W\mathbf{B}^*(t))_1.$$

Now fix $m \in \mathbb{N}$ which will be chosen suitably below. Define $t_k = k/m, k = 0, 1, \dots, rm-1$. Then, for any k as above,

$$\begin{aligned} \int_{t_k}^{t_{k+1}} V(s)^2 ds &\leq m^{-1} \sup_{s \in [t_k, t_{k+1}]} V(s)^2 \\ &= m^{-1} \sup_{s \in [t_k, t_{k+1}]} (v + gs - L_1(s))^2 \leq 2m^{-1}(|v| + rg)^2 + 2m^{-1}(L_1(r))^2. \end{aligned}$$

It then follows

$$\max_{0 \leq k \leq rm-1} e^{\frac{c}{2} \int_{t_k}^{t_{k+1}} V(s)^2 ds} \leq c_1 e^{cm^{-1}(W\mathbf{B}^*(r))_1^2}$$

where $c_1 = e^{cm^{-1}(|v|+rg)^2}$. The expectation of the right side under $\mathbb{P}_{(v, \mathbf{z})}^*$ (which is independent of $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$) is finite for sufficiently large m . The result follows. \square

For $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ and $r \in \mathbb{N}$, with an abuse of notation, denote the projection of $\mathbb{P}_{(v, \mathbf{z})}^*$ [resp. $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$] on $\Omega^r \doteq \mathcal{C}([0, r] : \mathbb{R}^N \times \mathbb{R} \times \mathbb{R}_+^N)$ once more as $\mathbb{P}_{(v, \mathbf{z})}^*$ [resp. $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$]. Denote by \mathcal{F}^r the Borel σ -field on Ω^r . The coordinate processes $\mathbf{B}, V, \mathbf{Z}$ on $(\Omega^r, \mathcal{F}^r)$ and the canonical filtration $\{\mathcal{F}_t^r\}_{0 \leq t \leq r}$ are defined in an analogous manner. Denote by \mathbf{e}_1 the unit vector $(1, 0, 0, \dots, 0)'$ in \mathbb{R}^N .

Corollary 4.3. Fix $r \in \mathbb{N}$. Define for $t \in [0, r]$, real measurable maps $\mathcal{E}(t)$ on $(\Omega^r, \mathcal{F}^r)$ as

$$\mathcal{E}(t) \doteq e^{-\sum_{i=1}^N \int_0^t V(s)(A^{-1}\mathbf{e}_1)_i dB_i(s) - \frac{|A^{-1}\mathbf{e}_1|^2}{2} \int_0^t V(s)^2 ds}.$$

Then for every $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, $\tilde{\mathbb{E}}_{(v, \mathbf{z})}^*[\mathcal{E}(r)] = 1$ and for every $F \in \mathcal{B}(\mathcal{C}([0, r] : \mathbb{R} \times \mathbb{R}_+^N))$

$$\mathbb{P}_{(v, \mathbf{z})}^*((V, \mathbf{Z}) \in F) = \tilde{\mathbb{E}}_{(v, \mathbf{z})}^*[I_{\{(V, \mathbf{Z}) \in F\}} \mathcal{E}(r)].$$

Proof. Fix $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ and $r \in \mathbb{N}$. For $t \in [0, r]$, define

$$\tilde{\mathbf{B}}(t) \doteq \mathbf{B}(t) + \int_0^t V(s)A^{-1}\mathbf{e}_1 ds.$$

By Lemma 4.2 with $c = |A^{-1}\mathbf{e}_1|^2$ and (a slight modification of) [28, Corollary 3.5.14], it follows that $\{\mathcal{E}(t)\}_{0 \leq t \leq r}$ is a martingale with respect to the filtration $\{\mathcal{F}_t^r\}_{0 \leq t \leq r}$ under the probability measure $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$. Hence, from Girsanov's theorem, $\{\tilde{\mathbf{B}}(s)\}_{0 \leq s \leq r}$ is a Brownian motion under the probability measure $\mathbb{Q}_{(v, \mathbf{z})}^*$ defined by $d\mathbb{Q}_{(v, \mathbf{z})}^* \doteq \mathcal{E}(r)d\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$. Also, under the measure $\mathbb{Q}_{(v, \mathbf{z})}^*$ we have

$$\begin{aligned} \mathbf{Z}(t) &= \Gamma_2(\mathbf{z} + A\mathbf{B}(\cdot))(t) = \Gamma_2\left(\mathbf{z} + A\left(\tilde{\mathbf{B}}(\cdot) - \int_0^\cdot V(s)A^{-1}\mathbf{e}_1 ds\right)\right)(t) \\ &= \Gamma_2\left(\mathbf{z} + A\tilde{\mathbf{B}}(\cdot) - \int_0^\cdot V(s)ds \mathbf{e}_1\right)(t). \end{aligned}$$

By the unique solvability given in Theorem 2.2 and the definition of $\mathbb{P}_{(v, \mathbf{z})}^*$ it now follows that the law of (V, \mathbf{Z}) under $\mathbb{Q}_{(v, \mathbf{z})}^*$ is same as that under $\mathbb{P}_{(v, \mathbf{z})}^*$. The result follows. \square

We next prove several technical estimates that will be needed in the proof of Theorem 4.1.

Lemma 4.4. *Let*

$$F \doteq [\frac{g}{16}, \frac{g}{4}] \times [\frac{g}{10}, 2g] \times [\frac{3g}{4}, 2g]^{N-2}.$$

Let $G \subset (\mathbb{R}_+^N)^o$ be an open and bounded domain with \mathcal{C}^2 boundary such that

$$F \subset F_1 \doteq [\frac{g}{16}, g] \times [\frac{g}{10}, 2g] \times [\frac{3g}{4}, 2g]^{N-2} \subset G.$$

Let $\sigma_F \doteq \inf_{x \in F, y \in \partial G} |A^{-1}(x - y)|$ and choose $\epsilon > 0$ so that $G_1 \doteq \{x \in G : \inf_{y \in \partial G} |A^{-1}(x - y)| > \epsilon\}$ satisfies $G \supset G_1 \supset F_1 \supset F$. Define on $(\Omega^, \mathcal{F}^*)$, $\tau_G = \inf\{t \geq 0 : \mathbf{Z}(t) \notin G\}$. Also, fix a ‘cemetery point’ $\partial^* \in (\mathbb{R}_+^N)^c$ and define the ‘killed process’ $\{\mathbf{Z}^*(t)\}$ by*

$$\mathbf{Z}^*(t) \doteq \begin{cases} \mathbf{Z}(t) & \text{if } t < \tau_G \\ \partial^* & \text{if } t \geq \tau_G, \end{cases} \quad (4.6)$$

Then, there is a $c_G \in (0, \infty)$ such that for any $J \in \mathcal{B}(\mathbb{R}_+^N)$,

$$\inf_{s \in [\frac{\epsilon}{4}, \varsigma^*], (v, \mathbf{z}) \in \mathbb{R} \times F} \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J) \geq c_G \lambda(J \cap G_1).$$

Proof. Fix $s \in [\frac{\epsilon}{4}, \varsigma^*]$, $(v, \mathbf{z}) \in \mathbb{R} \times F$ and $J \in \mathcal{B}(\mathbb{R}_+^N)$ with $\lambda(J \cap G_1) > 0$. Since, under $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$, $\mathbf{Z}(t) = \mathbf{z} + \mathbf{A}\mathbf{B}(t)$ until the first time it has hit the boundary of the positive orthant,

$$\begin{aligned} \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J) &= \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J \cap G) \\ &= \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{z} + \mathbf{A}\mathbf{B}(s) \in J \cap G, \mathbf{z} + \mathbf{A}\mathbf{B}(u) \in G \text{ for all } u \leq s) \\ &= \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(A^{-1}\mathbf{z} + \mathbf{B}(s) \in A^{-1}(J \cap G), \\ &\quad A^{-1}\mathbf{z} + \mathbf{B}(u) \in A^{-1}(G), \text{ for all } u \leq s). \end{aligned}$$

Denote the transition probability density at time t of an N -dimensional standard Brownian motion in $A^{-1}G$, started from x and killed at the boundary of $A^{-1}G$, by $p_t(x, \cdot)$. Then from the above identities it follows

$$\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J) = \int_{A^{-1}(G \cap J)} p_s(A^{-1}\mathbf{z}, y) dy. \quad (4.7)$$

From [40, Theorem 1.1] we have that there exists $T > 0$ and $c_1, c_2 \in (0, \infty)$ such that for all $x, y \in A^{-1}G$:

$$\begin{aligned} p_t(x, y) &\geq \left(\frac{\rho(x)\rho(y)}{t} \wedge 1 \right) \frac{c_1}{t^{N/2}} e^{-c_2|x-y|^2/t}, \quad t \in [0, T] \\ p_t(x, y) &\geq c_1 \rho(x)\rho(y) e^{-c_2 t}, \quad t \in (T, \infty), \end{aligned}$$

where $\rho(x) = \inf_{r \in \partial G} |x - A^{-1}r|$.

Note that there is a $\eta \in (0, \infty)$ such that

$$\lambda(A^{-1}(C)) = \eta\lambda(C) \text{ for all } C \in \mathcal{B}(\mathbb{R}^N).$$

We now estimate the right side of (4.7). First suppose that $s > T$. Then since $\mathbf{z} \in F$ and $s \leq 1$,

$$\begin{aligned} \int_{A^{-1}(G \cap J)} p_s(A^{-1}\mathbf{z}, y) dy &\geq \int_{A^{-1}(G \cap J)} c_1 \rho(A^{-1}\mathbf{z}) \rho(y) e^{-c_2 s} dy \\ &\geq c_1 \sigma_F e^{-c_2} \int_{A^{-1}(G_1 \cap J)} \rho(y) dy \geq c_1 \sigma_F e^{-c_2} \eta \lambda(G_1 \cap J). \end{aligned}$$

Letting $c_{G,1} = c_1 \sigma_F e^{-c_2} \eta$, we have from (4.7), when $s > T$,

$$\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J) \geq c_{G,1} \lambda(J \cap G_1). \quad (4.8)$$

Now consider the case $s \leq T$. Then, a similar estimate shows,

$$\begin{aligned} \int_{A^{-1}(J \cap G)} p_s(A^{-1}\mathbf{z}, y) dy &\geq \int_{A^{-1}(J \cap G)} \left(\frac{\rho(A^{-1}\mathbf{z}) \rho(y)}{s} \wedge 1 \right) \frac{c_1}{s^{N/2}} e^{-c_2 |A^{-1}\mathbf{z} - y|^2/s} dy \\ &\geq c_{G,2} \lambda(J \cap G_1) \end{aligned}$$

where $c_{G,2} = \eta c_1 ((\epsilon \sigma_F) \wedge 1) e^{-4c_2(\zeta)^{-1} \sup_{x,y \in A^{-1}G} |x-y|^2}$. Thus, once more from (4.7), when $s \leq T$,

$$\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J) \geq c_{G,2} \lambda(J \cap G_1). \quad (4.9)$$

Setting $c_G = c_{G,1} \wedge c_{G,2}$, we have the result on combining (4.8) and (4.9). \square

For $z_1 \in \mathbb{R}$, let \mathcal{P}_{z_1} denote a probability measure on Ω^* under which the coordinate process $\{B_1(t)\}$ is a standard Brownian motion starting at z_1 . We will use similar notation for the corresponding expectation.

Lemma 4.5. *There exists a $\mathcal{K} \in (0, \infty)$ such that for every $I \in \mathcal{B}(\mathbb{R})$,*

$$\begin{aligned} \mathcal{P}_0 \left(\sup_{0 \leq u \leq \frac{\zeta}{4}} B_1(u) \in I, \inf_{0 \leq u \leq \frac{\zeta}{4}} B_1(u) > -\frac{6g}{10}, B_1\left(\frac{\zeta}{4}\right) \in \left[-\frac{g}{8}, -\frac{g}{16}\right] \right) \\ \geq \mathcal{K} \lambda(I \cap [0, \frac{g}{63}]). \end{aligned}$$

Proof. Let $I \in \mathcal{B}(\mathbb{R})$. We assume without loss of generality that, $I \subset [0, \frac{g}{63}]$ and I is of the form $I = [\beta_1, \beta_2] \subset \mathbb{R}_+$ for $0 \leq \beta_1 \leq \beta_2$ (the choice of \mathcal{K} will be independent of β_1, β_2). Let $\gamma \doteq \frac{g}{2}(-\frac{1}{8} - \frac{1}{16})$ be the midpoint of $[-\frac{g}{8}, -\frac{g}{16}]$. For a level $c \in \mathbb{R}$, let $\tau_c \doteq \inf\{t \geq 0 : B_1(t) = c\}$. Define $\sigma \doteq \tau_{-6g/10}$ and $\tau_i^\beta \doteq \tau_{\beta_i}$ for $i = 1, 2$. Then

$$\begin{aligned} \mathcal{P}_0(\sup_{u \leq \frac{\zeta}{4}} (B_1(u)) \in I, \inf_{u \leq \frac{\zeta}{4}} B_1(u) > -\frac{6g}{10}, B_1(\frac{\zeta}{4}) \in [-\frac{g}{8}, -\frac{g}{16}]) \\ = \mathcal{P}_0(\sup_{u \leq \frac{\zeta}{4}} (B_1(u)) \in I, \sigma > \frac{\zeta}{4}, B_1(\frac{\zeta}{4}) \in [-\frac{g}{8}, -\frac{g}{16}]). \end{aligned}$$

Using the strong Markov property of the Brownian motion, we obtain,

$$\begin{aligned}
& \mathcal{P}_0(\sup_{u \leq \frac{\zeta}{4}} (B_1(u)) \in I, \sigma > \frac{\zeta}{4}, B_1(\frac{\zeta}{4}) \in [-\frac{g}{8}, -\frac{g}{16}]) \\
& \geq \mathcal{P}_0(\tau_1^\beta \leq \frac{\zeta}{8} \wedge \sigma, \sup_{u \leq \frac{\zeta}{4}} (B_1(u)) \in I, \sigma > \frac{\zeta}{4}, B_1(\frac{\zeta}{4}) \in [-\frac{g}{8}, -\frac{g}{16}]) \\
& = \mathcal{P}_0(\mathbf{1}_{\{\tau_1^\beta \leq \frac{\zeta}{8} \wedge \sigma\}} \Theta(\tau_1^\beta)),
\end{aligned}$$

where, for $t \in [0, \frac{\zeta}{8}]$,

$$\Theta(t) \doteq \mathcal{P}_{\beta_1}(\sup_{u \leq \frac{\zeta}{4} - t} (B_1(u)) \leq \beta_2, \sigma > \frac{\zeta}{4} - t, B_1(\frac{\zeta}{4} - t) \in [-\frac{g}{8}, -\frac{g}{16}]).$$

By another application of the strong Markov property, for $t \in [0, \frac{\zeta}{8}]$,

$$\begin{aligned}
\Theta(t) & \geq \mathcal{P}_{\beta_1}(\tau_\gamma \leq \tau_2^\beta \wedge \frac{\zeta}{16}, B_1(s) \in [-\frac{g}{8}, -\frac{g}{16}] \text{ for all } s \in [\tau_\gamma, \frac{\zeta}{4} - t]) \\
& = \mathcal{P}_{\beta_1}(\mathbf{1}_{\{\tau_\gamma \leq \tau_2^\beta \wedge \frac{\zeta}{16}\}} \Theta'(\tau_\gamma))
\end{aligned}$$

where for $u \in [0, \frac{\zeta}{16}]$,

$$\Theta'(u) \doteq \mathcal{P}_\gamma(B_1(s) \in [-\frac{g}{8}, -\frac{g}{16}] \text{ for all } s \in [0, \frac{\zeta}{4} - t - u]).$$

Thus letting

$$\kappa_1 \doteq \mathcal{P}_\gamma(B_1(s) \in [-\frac{g}{8}, -\frac{g}{16}] \text{ for all } s \in [0, \frac{\zeta}{4}])$$

we have that, for $t \in [0, \frac{\zeta}{8}]$,

$$\Theta(t) \geq \kappa_1 \mathcal{P}_{\beta_1}(\tau_\gamma \leq \tau_2^\beta \wedge \frac{\zeta}{16}).$$

Also, by an application of the reflection principle,

$$\begin{aligned}
\mathcal{P}_{\beta_1}(\tau_\gamma \leq \tau_2^\beta \wedge \frac{\zeta}{16}) & = \mathcal{P}_{\beta_1}(\tau_\gamma \leq \frac{\zeta}{16}) - \mathcal{P}_{\beta_1}(\tau_2^\beta < \tau_\gamma \leq \frac{\zeta}{16}) \\
& = \mathcal{P}_{\beta_1}(\tau_\gamma \leq \frac{\zeta}{16}) - \mathcal{P}_{\beta_1 + 2(\beta_2 - \beta_1)}(\tau_\gamma \leq \frac{\zeta}{16}).
\end{aligned}$$

From the definition of the stopping times we see,

$$\begin{aligned}
\mathcal{P}_{\beta_1}(\tau_\gamma \leq \frac{\zeta}{16}) & = \mathcal{P}_0(\sup_{u \leq \frac{\zeta}{16}} (B_1(u)) \geq \beta_1 - \gamma), \\
\mathcal{P}_{\beta_1 + 2(\beta_2 - \beta_1)}(\tau_\gamma \leq \frac{\zeta}{16}) & = \mathcal{P}_0(\sup_{u \leq \frac{\zeta}{16}} (B_1(u)) \geq \beta_1 + 2(\beta_2 - \beta_1) - \gamma).
\end{aligned}$$

Using the explicit form for the probability density for the law of the maximum of a Brownian motion, we then obtain,

$$\begin{aligned} \mathcal{P}_{\beta_1}(\tau_\gamma \leq \frac{\varsigma}{16}) - \mathcal{P}_{\beta_1+2(\beta_2-\beta_1)}(\tau_\gamma \leq \frac{\varsigma}{16}) &= \int_{\beta_1-\gamma}^{\beta_1+2(\beta_2-\beta_1)-\gamma} \frac{4\sqrt{2}}{\sqrt{\pi\varsigma}} e^{-8z^2/\varsigma} \\ &\geq \frac{8\sqrt{2}}{\sqrt{\pi\varsigma}} \inf_{\beta_1-\gamma \leq z \leq \beta_1+2(\beta_2-\beta_1)-\gamma} e^{-8z^2/\varsigma} (\beta_2 - \beta_1). \end{aligned}$$

Since

$$\beta_1 + 2(\beta_2 - \beta_1) - \gamma \leq 2\beta_2 - \gamma \leq \frac{2g}{63} + \frac{3g}{32} \leq \frac{g}{4},$$

we have

$$\inf_{\beta_1-\gamma \leq z \leq 2(\beta_2-\beta_1)-\gamma} e^{-8z^2/\varsigma} \geq e^{-\frac{g^2}{2\varsigma}}.$$

Thus, for $t \in [0, \varsigma/8]$,

$$\Theta(t) \geq \kappa_1 \mathcal{P}_{\beta_1}(\tau_\gamma \leq \tau_2^\beta \wedge \frac{\varsigma}{16}) \geq \kappa_1 \frac{8\sqrt{2}}{\sqrt{\pi\varsigma}} e^{-\frac{g^2}{2\varsigma}} (\beta_2 - \beta_1).$$

Finally, observe that, as $I \subset [0, \frac{g}{63}]$,

$$\mathcal{P}_0(\tau_1^\beta \leq \frac{\varsigma}{8} \wedge \sigma) \geq \mathcal{P}_0(\sup_{u \leq \frac{\varsigma}{8}} (B_1(s)) > \frac{g}{63}, \inf_{u \leq \frac{\varsigma}{8}} (B_1(s)) > -\frac{6g}{10}) \doteq \kappa_2.$$

The result now follows on setting $\mathcal{K} = \kappa_1 \kappa_2 \frac{8\sqrt{2}}{\sqrt{\pi\varsigma}} e^{-\frac{g^2}{2\varsigma}}$. \square

For $0 \leq s \leq 1$ and $(z_2, z_3, \dots, z_N) \in \mathbb{R}_+^{N-1}$, define

$$\begin{aligned} \hat{L}_1(s) &= \sup_{u \leq s} (-B_1(u)), \quad \hat{Z}_1(s) = B_1(s) + \hat{L}_1(s) \\ \hat{L}_i(s) &= \sup_{u \leq s} (-z_i + B_{i-1}(u) - B_i(u))^+, \quad i = 2, \dots, N. \end{aligned}$$

Lemma 4.6. *Let $I \in \mathcal{B}(\mathbb{R})$ be such that $I \subset [0, \frac{g}{63}]$ and $(v, \mathbf{z}) \in \mathbb{R} \times \{0\} \times [g, \frac{3g}{2}]^{N-1}$. Let $H \in \mathcal{F}^*$. Then the following are equivalent:*

- (a) On H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s., (i) $L_1(\frac{\varsigma}{4}) \in I$, (ii) $L_i(\frac{\varsigma}{4}) \leq \frac{g}{6}$, for all $i = 2, \dots, N$, (iii) $\sup_{0 \leq u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}$, (iv) $\sup_{0 \leq u \leq \frac{\varsigma}{4}} |B_i(u)| < \frac{g}{8}$, for all $i = 2, \dots, N$, (v) $Z_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}]$.
- (b) On H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s., (i') $\hat{L}_1(\frac{\varsigma}{4}) \in I$, (ii') $\hat{L}_i(\frac{\varsigma}{4}) \leq \frac{g}{6}$, for all $i = 2, \dots, N$, (iii') $\sup_{0 \leq u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}$, (iv') $\sup_{0 \leq u \leq \frac{\varsigma}{4}} |B_i(u)| < \frac{g}{8}$, for all $i = 2, \dots, N$, (v') $\hat{Z}_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}]$.

Furthermore, under these equivalent conditions, on H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s., $L_1(\varsigma/4) = \hat{L}_1(\varsigma/4)$ and $L_i(\varsigma/4) = 0$ for $i = 2, \dots, N$.

Proof. Fix $(v, \mathbf{z}) \in \mathbb{R} \times \{0\} \times [g, \frac{3g}{2}]^{N-1}$. Noting that $z_i \geq g$ for $i = 2, \dots, N$, we see that, when conditions (i) – (v) hold, for all $u \leq \frac{\varsigma}{4}$,

$$\begin{aligned} -z_i + B_{i-1}(u) - B_i(u) + \frac{1}{2}(L_{i-1}(u) + L_{i+1}(u)) &\leq -g + \frac{g}{4} + \frac{g}{6} \leq 0, \quad i = 3, \dots, N, \\ -z_2 + B_1(u) - B_2(u) + \frac{1}{2}L_3(u) + L_1(u) &\leq -g + \frac{6g}{10} + \frac{g}{8} + \frac{g}{12} + \frac{g}{63} \leq 0. \end{aligned}$$

Hence, when conditions (i) – (v) hold on H , by (4.3), $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s., $L_i(\frac{\varsigma}{4}) = \hat{L}_i(\frac{\varsigma}{4}) = 0$ for $i = 2, \dots, N$ which in turn says that $L_1(\frac{\varsigma}{4}) = \hat{L}_1(\frac{\varsigma}{4})$ and $Z_1(\frac{\varsigma}{4}) = \hat{Z}_1(\frac{\varsigma}{4})$. Thus in this case (i') – (v') hold on H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s.

On the other hand, suppose that (i') – (v') hold on H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s. Consider the stopping times,

$$\begin{aligned} v_2 &= \inf\{t \geq 0 : -z_2 + B_1(t) - B_2(t) + \frac{1}{2}L_3(t) + L_1(t) \geq 0\} \\ v_i &= \inf\{t \geq 0 : -z_i + B_{i-1}(t) - B_i(t) + \frac{1}{2}(L_{i+1}(t) + L_{i-1}(t)) \geq 0\}, \quad i = 3, \dots, N \end{aligned}$$

and let $v = \min_{2 \leq i \leq N} v_i$.

Then, for $s \leq v$, $L_i(s) = \hat{L}_i(s) = 0$, for all $i = 2, \dots, N$ and so $L_1(s) = \hat{L}_1(s)$ and $Z_1(s) = \hat{Z}_1(s)$. Thus, if $v \leq \frac{\varsigma}{4}$,

$$\begin{aligned} -z_2 + B_1(v) - B_2(v) + \frac{1}{2}L_3(v) + L_1(v) &= -z_2 + B_1(v) - B_2(v) + \hat{L}_1(v) \\ &\leq -g + \frac{6g}{10} + \frac{g}{8} + \frac{g}{63} < 0, \end{aligned}$$

and, for $i = 3, \dots, N$,

$$\begin{aligned} -z_i + B_{i-1}(v) - B_i(v) + \frac{1}{2}(L_{i+1}(v) + L_{i-1}(v)) &= -z_i + B_{i-1}(v) - B_i(v) \\ &\leq -g + \frac{g}{8} + \frac{g}{8} < 0. \end{aligned}$$

This contradicts the definition of v and consequently we must have that $v > \frac{\varsigma}{4}$. Thus (i) – (v) hold on H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s., and the result follows. \square

Recall the set F introduced in Lemma 4.4.

Lemma 4.7. *Let $I \in \mathcal{B}(\mathbb{R})$ be such that $I \subset [0, \frac{g}{63}]$ and $(v, \mathbf{z}) \in \mathbb{R} \times \{0\} \times [g, \frac{3g}{2}]^{N-1}$. Let $H \in \mathcal{F}^*$ and suppose the equivalent conditions of Lemma 4.6 hold on H . Then, on H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s., $\mathbf{Z}(\frac{\varsigma}{4}) \in F$.*

Proof. Under assumptions of the lemma, on H , $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$ a.s.,

$$\begin{aligned} Z_2(\varsigma/4) &= z_2 + B_2(\varsigma/4) - B_1(\varsigma/4) - \frac{1}{2}L_3(\varsigma/4) + L_2(\varsigma/4) - L_1(\varsigma/4) \\ &= z_2 + B_2(\varsigma/4) - B_1(\varsigma/4) - L_1(\varsigma/4) \geq g - \frac{g}{8} - \frac{6g}{10} - \frac{g}{63} \geq \frac{g}{10}, \\ Z_i(\varsigma/4) &= z_i + B_i(\varsigma/4) - B_{i-1}(\varsigma/4) - \frac{1}{2}(L_{i-1}(\varsigma/4) + L_{i+1}(\varsigma/4)) + L_i(\varsigma/4) \\ &= z_i + B_i(\varsigma/4) - B_{i-1}(\varsigma/4) \geq \frac{3g}{4}, \quad i = 3, \dots, N. \end{aligned}$$

From Lemma 4.6, under the assumptions of the current lemma, $L_1(\varsigma/4) = \hat{L}_1(\varsigma/4)$ and so

$$\frac{g}{63} \geq L_1(\varsigma/4) = \hat{L}_1(\varsigma/4) = \sup_{u \leq \varsigma/4} (-B_1(u)) \geq -B_1(\varsigma/4).$$

Thus we have the upper bound,

$$Z_2(\varsigma/4) = z_2 + B_2(\varsigma/4) - B_1(\varsigma/4) - L_1(\varsigma/4) \leq \frac{3g}{2} + \frac{g}{8} + \frac{g}{63} \leq 2g,$$

Also, for $i = 3, \dots, N$,

$$Z_i(\varsigma/4) = z_i + B_i(\varsigma/4) - B_{i-1}(\varsigma/4) \leq \frac{3g}{2} + \frac{g}{8} + \frac{g}{8} \leq 2g.$$

Hence, $(Z_2(\varsigma/4), \dots, Z_N(\varsigma/4)) \in [\frac{g}{10}, 2g] \times [\frac{3g}{4}, 2g]^{N-2}$. Also, under the conditions of the lemma $Z_1 \in [\frac{g}{16}, \frac{g}{4}]$. Thus $\mathbf{Z}(\varsigma/4) \in F$ and the lemma is proved. \square

We can now complete the proof of Theorem 4.1.

Proof of Theorem 4.1 Recall F , G and G_1 from Lemma 4.4. We will prove the theorem with $D \doteq D_1 \times G_1$ where $D_1 = [0, g/128]$. Let $(v, \mathbf{z}) \in [0, \frac{g}{128}] \times (0, \infty) \times \mathbb{R}_+^{N-1}$. All equalities and inequalities of random quantities in the proof are under the measure $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$. Let $r \in [\varsigma, \varsigma^*]$ be given. It suffices to establish the estimate in (4.1) for $S \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ of the form $S = I \times J$, $I \in \mathcal{B}(\mathbb{R})$, $J \in \mathcal{B}(\mathbb{R}_+^N)$ with $I \subseteq D_1$ and $J \subseteq G_1$, for a choice of the constant $K_{(v, \mathbf{z})}$ independent of I, J . For such an S , letting $\tilde{B}(t) \doteq \sum_{i=1}^N (A^{-1})_{i,1} B_i(t)$, by Corollary 4.3,

$$\begin{aligned} \mathbb{P}^r((v, \mathbf{z}), S) &= \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{\{V(r) \in I, \mathbf{Z}(r) \in J\}} \mathcal{E}(1) = \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{\{V(r) \in I, \mathbf{Z}(r) \in J\}} \mathcal{E}(r) \\ &= \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{\{V(r) \in I, \mathbf{Z}(r) \in J\}} e^{-\int_0^r V(s) d\tilde{B}(s) - \frac{|A^{-1}\mathbf{e}_1|^2}{2} \int_0^r V(s)^2 ds}. \end{aligned} \quad (4.10)$$

On the set $\{V(r) \in I\}$, $L_1(r) = gr - V(r) + v \leq gr + v \leq gr + \frac{g}{128}$, so that by monotonicity, $L_1(s) \leq g(r + \frac{1}{128})$ for all $s \leq r$. This implies that, on this set, for $s \leq r$, $-2g \leq -g(r + \frac{1}{128}) \leq V(s) \leq gr + v \leq 2g$, i.e., $|V(s)| \leq 2g$. Using this estimate in (4.10) we get

$$\mathbb{P}^r((v, \mathbf{z}), S) \geq e^{-2|A^{-1}\mathbf{e}_1|^2 g^2} \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{\{V(r) \in I, \mathbf{Z}(r) \in J\}} e^{-\int_0^r V(s) d\tilde{B}(s)}. \quad (4.11)$$

By Itô's formula,

$$\begin{aligned} -\int_0^r V(s) d\tilde{B}(s) &= \int_0^r \tilde{B}(s) dV(s) - V(r)\tilde{B}(r) \\ &= -\int_0^r \tilde{B}(s) dL_1(s) + g \int_0^r \tilde{B}(s) ds - V(r)\tilde{B}(r). \end{aligned}$$

Define the stopping time

$$\tau_1 \doteq \inf\{t \geq 0 : Z_1(t) = 0\},$$

and let

$$H \doteq \left\{ \tau_1 \leq \frac{\varsigma}{4}, (Z_2(\tau_1), \dots, Z_N(\tau_1)) \in [g, \frac{3g}{2}]^{N-1}, L_1(\tau_1 + \frac{\varsigma}{4}) \in v + gr - I, \right. \\ \left. \mathbf{Z}(\tau_1 + \frac{\varsigma}{4}) \in F, \mathbf{Z}(s) > 0, \text{ for all } s \in [\tau_1 + \frac{\varsigma}{4}, r], \mathbf{Z}(r) \in J \right\}.$$

Note that

$$H \subset \{(V(r), \mathbf{Z}(r)) \in I \times J\}. \quad (4.12)$$

On H we have,

$$-V(r)\tilde{B}(r) \geq -2g|\tilde{B}(r)| \geq -2g|\tilde{B}(r) - \tilde{B}(\tau_1 + \frac{\varsigma}{4})| - 2g|\tilde{B}(\tau_1 + \frac{\varsigma}{4}) - \tilde{B}(\tau_1)| - 2g|\tilde{B}(\tau_1)|.$$

In addition, on H ,

$$g \int_0^r \tilde{B}(s)ds = g \int_0^{\tau_1} \tilde{B}(s)ds + g \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} (\tilde{B}(s) - \tilde{B}(\tau_1))ds + \frac{g\varsigma}{4} \tilde{B}(\tau_1) \\ + g \int_{\tau_1 + \frac{\varsigma}{4}}^r (\tilde{B}(s) - \tilde{B}(\tau_1 + \frac{\varsigma}{4}))ds + g\tilde{B}(\tau_1 + \frac{\varsigma}{4})(r - (\tau_1 + \frac{\varsigma}{4})) \\ = g \int_0^{\tau_1} \tilde{B}(s)ds + g \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} (\tilde{B}(s) - \tilde{B}(\tau_1))ds + g\tilde{B}(\tau_1)(r - \tau_1) \\ + g \int_{\tau_1 + \frac{\varsigma}{4}}^r (\tilde{B}(s) - \tilde{B}(\tau_1 + \frac{\varsigma}{4}))ds + g(\tilde{B}(\tau_1 + \frac{\varsigma}{4}) - \tilde{B}(\tau_1))(r - (\tau_1 + \frac{\varsigma}{4})).$$

Also, by the definition of τ_1 , on H ,

$$- \int_0^r \tilde{B}(s)dL_1(s) = - \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} \tilde{B}(s)dL_1(s) \\ = - \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} (\tilde{B}(s) - \tilde{B}(\tau_1))dL_1(s) - \tilde{B}(\tau_1)(L_1(\tau_1 + \frac{\varsigma}{4}) - L_1(\tau_1)) \\ \geq - \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} (\tilde{B}(s) - \tilde{B}(\tau_1))dL_1(s) - g|\tilde{B}(\tau_1)|.$$

Now let

$$U_1 \doteq g \int_{\tau_1 + \frac{\varsigma}{4}}^r (\tilde{B}(s) - \tilde{B}(\tau_1 + \frac{\varsigma}{4}))ds - 2g|\tilde{B}(r) - \tilde{B}(\tau_1 + \frac{\varsigma}{4})|, \\ U_2 \doteq g(\tilde{B}(\tau_1 + \frac{\varsigma}{4}) - \tilde{B}(\tau_1))(r - (\tau_1 + \frac{\varsigma}{4})) - 2g|\tilde{B}(\tau_1 + \frac{\varsigma}{4}) - \tilde{B}(\tau_1)| \\ - \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} (\tilde{B}(s) - \tilde{B}(\tau_1))dL_1(s) + g \int_{\tau_1}^{\tau_1 + \frac{\varsigma}{4}} (\tilde{B}(s) - \tilde{B}(\tau_1))ds, \\ U_3 \doteq -3g|\tilde{B}(\tau_1)| + g \int_0^{\tau_1} \tilde{B}(s)ds + g\tilde{B}(\tau_1)(r - \tau_1).$$

Then, by (4.12), we have the lower bound

$$\tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{\{V(r) \in I, \mathbf{Z}(r) \in J\}} e^{-\int_0^r V(s) d\tilde{B}(s)} \geq \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_H e^{-\int_0^r V(s) d\tilde{B}(s)} \geq \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_H e^{U_1+U_2+U_3}. \quad (4.13)$$

Recall the killed process \mathbf{Z}^* from (4.6). Define the sets

$$\begin{aligned} H_1(s) &= \{\mathbf{Z}^*(s) \in J\}, \quad 0 \leq s \leq 1, \\ H_2(v) &= \left\{ L_1\left(\frac{\varsigma}{4}\right) \in gr + v - I, L_i\left(\frac{\varsigma}{4}\right) \leq \frac{g}{6} \text{ for } 2 \leq i \leq N, \sup_{u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}, \right. \\ &\quad \left. \sup_{u \leq \frac{\varsigma}{4}} |B_i(u)| < \frac{g}{8} \text{ } i = 2, \dots, N, Z_1\left(\frac{\varsigma}{4}\right) \in \left[\frac{g}{16}, \frac{g}{4}\right] \right\}, \\ &\quad \text{where } v \in [0, g/128]. \\ H_3 &= \left\{ \tau_1 \leq \frac{\varsigma}{4}, (Z_2(\tau_1), \dots, Z_N(\tau_1)) \in [g, \frac{3g}{2}]^{N-1} \right\}. \end{aligned}$$

Also, set

$$\begin{aligned} U'_1(t) &\doteq g \int_0^t \tilde{B}(s) ds - 2g|\tilde{B}(t)|, \quad 0 \leq t \leq 1, \\ U'_2 &\doteq -3g|\tilde{B}\left(\frac{\varsigma}{4}\right)| - \int_0^{\frac{\varsigma}{4}} \tilde{B}(s) dL_1(s) + g \int_0^{\frac{\varsigma}{4}} \tilde{B}(s) ds. \end{aligned}$$

Applying the Strong Markov Property at $\tau_1 + \frac{\varsigma}{4}$ and then τ_1 , we have

$$\begin{aligned} \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_H e^{U_1+U_2+U_3} &\geq \inf_{(\tilde{v}, \tilde{\mathbf{z}}) \in \mathbb{R} \times F, \frac{\varsigma}{4} \leq s \leq r} \tilde{\mathbb{E}}_{(\tilde{v}, \tilde{\mathbf{z}})}^* \mathbf{1}_{H_1(s)} e^{U'_1(s)} \\ &\quad \times \inf_{(\hat{v}, \hat{\mathbf{z}}) \in \mathbb{R} \times [g, \frac{3g}{2}]^{N-1}} \tilde{\mathbb{E}}_{(\hat{v}, (0, \hat{\mathbf{z}}))}^* \mathbf{1}_{\{L_1(\frac{\varsigma}{4}) \in gr+v-I, \mathbf{Z}(\frac{\varsigma}{4}) \in F\}} e^{U'_2} \times \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{H_3} e^{U_3}. \end{aligned} \quad (4.14)$$

Note that since by assumption $I \subseteq [0, g/128]$, $r \in [\varsigma, \varsigma^*]$, and $v \in [0, g/128]$,

$$\tilde{I} \doteq gr + v - I \subseteq [0, g/63]. \quad (4.15)$$

Thus, using Lemma 4.7, we see

$$\begin{aligned} \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_H e^{U_1+U_2+U_3} &\geq \inf_{(\tilde{v}, \tilde{\mathbf{z}}) \in \mathbb{R} \times F, \frac{\varsigma}{4} \leq s \leq r} \tilde{\mathbb{E}}_{(\tilde{v}, \tilde{\mathbf{z}})}^* \mathbf{1}_{H_1(s)} e^{U'_1(s)} \\ &\quad \times \inf_{(\hat{v}, \hat{\mathbf{z}}) \in \mathbb{R} \times [g, \frac{3g}{2}]^{N-1}} \tilde{\mathbb{E}}_{(\hat{v}, (0, \hat{\mathbf{z}}))}^* \mathbf{1}_{H_2(v)} e^{U'_2} \times \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{H_3} e^{U_3}. \end{aligned}$$

For the final term, note that, on H_3 , $U_3 \geq -5g \sup_{0 \leq s \leq \varsigma^*} |\tilde{B}(s)|$. Now, for $M' > 0$, define

$$\begin{aligned} H'_3(M') &= \left\{ \sup_{0 \leq s \leq \varsigma^*} |\tilde{B}(s)| < M', Z_1(s) > 0 \text{ for all } s \leq \frac{\varsigma}{8}, \inf_{\frac{\varsigma}{8} \leq s \leq \frac{\varsigma}{4}} Z_1(s) = 0, \right. \\ &\quad \left. (Z_2(s), \dots, Z_N(s)) \in [g, \frac{3g}{2}]^{N-1} \text{ for all } s \in [\frac{\varsigma}{8}, \frac{\varsigma}{4}] \right\}. \end{aligned}$$

Clearly $H'_3(M') \subset H_3$. For any $(v, \mathbf{z}) \in [0, \frac{g}{128}] \times (0, \infty) \times \mathbb{R}_+^{N-1}$, one can construct suitable Brownian paths to obtain a measurable choice of $M' = M'(v, \mathbf{z})$ such that

$$\kappa_{(v, \mathbf{z})} \doteq \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(H'_3(M'(v, \mathbf{z}))) > 0.$$

This definition readily implies the measurability of $(v, \mathbf{z}) \mapsto \kappa_{(v, \mathbf{z})}$ through the measurability of the maps $(v, \mathbf{z}) \mapsto M'(v, \mathbf{z})$ and $(v, \mathbf{z}) \mapsto \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(A^\circ)$ for any $A^\circ \in \mathcal{F}$. Recall the set \bar{A} from the statement of Theorem 4.1. Using continuity properties of the transition kernel of Brownian motion in its starting point, the choice of $M'(v, \mathbf{z})$ can be made such that

$$\sup_{(v, \mathbf{z}) \in \bar{A}} M'(v, \mathbf{z}) < \infty, \quad \inf_{(v, \mathbf{z}) \in \bar{A}} \kappa_{(v, \mathbf{z})} > 0. \quad (4.16)$$

It now follows that,

$$\tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{H_3} e^{U_3} \geq \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{H'_3(M'(v, \mathbf{z}))} e^{U_3} \geq e^{-5gM'(v, \mathbf{z})\kappa_{(v, \mathbf{z})}}. \quad (4.17)$$

Now consider the term involving $H_1(s)$. Note that, on the set $H_1(s)$, $A\mathbf{B}(u) + z \in G$ for all $u \leq s$. Since G is bounded, we have that for some $\kappa_G \in (0, \infty)$, under $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*$, for all $(v, \mathbf{z}) \in \mathbb{R} \times F$

$$\sup_{0 \leq u \leq s} |\tilde{B}(u)| \leq \kappa_G \text{ on } H_1(s), \text{ for all } s \in [\zeta/4, r] \text{ and } r \in [\zeta, \zeta^*].$$

Thus, from Lemma 4.4,

$$\begin{aligned} \inf_{(v, \mathbf{z}) \in \mathbb{R} \times F, \frac{\zeta}{4} \leq s \leq r} \tilde{\mathbb{E}}_{(v, \mathbf{z})}^* \mathbf{1}_{H_1(s)} e^{U'_1(s)} &\geq e^{-3g\kappa_G} \inf_{(v, \mathbf{z}) \in \mathbb{R} \times F, \frac{\zeta}{4} \leq s \leq r} \tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(\mathbf{Z}^*(s) \in J) \\ &\geq e^{-3g\kappa_G} c_G \lambda(J \cap G_1). \end{aligned} \quad (4.18)$$

Consider finally the term involving $H_2(v)$. From Lemma 4.6 (and recalling (4.15)) it follows that, on $H_2(v)$, for $v \in [0, g/128]$ and $0 \leq s \leq \zeta/4$,

$$-B_1(s) \leq \sup_{u \leq \zeta/4} (-B_1(u)) = L_1(\zeta/4) \leq g/63.$$

Using this and other properties of the set $H_2(v)$, we see that with $c_A \doteq \frac{6g}{10} \sum_{i=1}^N |(A^{-1})_{i1}|$, on $H_2(v)$,

$$\sup_{0 \leq s \leq \zeta/4} |\tilde{B}(s)| \leq c_A.$$

It then follows that, on $H_2(v)$,

$$U'_2 \geq -3gc_A - \frac{g\zeta}{4}c_A - c_A L_1(\zeta/4) \geq -4gc_A.$$

Thus, we have

$$\inf_{(\hat{v}, \hat{\mathbf{z}}) \in \mathbb{R} \times [g, \frac{3g}{2}]^{N-1}} \tilde{\mathbb{E}}_{(\hat{v}, (0, \hat{\mathbf{z}}))}^* \mathbf{1}_{H_2(v)} e^{U'_2} \geq e^{-4gc_A} \inf_{(\hat{v}, \hat{\mathbf{z}}) \in \mathbb{R} \times [g, \frac{3g}{2}]^{N-1}} \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{\mathbf{z}}))}^*(H_2(v)). \quad (4.19)$$

Note that the conditions $\sup_{u \leq \varsigma/4} (-B_2(u)) \leq -13g/30 + \hat{z}_2$ and $\sup_{u \leq \frac{\varsigma}{4}} B_1(u) < 6g/10$ imply that $\hat{L}_2(\varsigma/4) \leq g/6$. Thus from Lemma 4.6, and using (4.15) again,

$$\begin{aligned}
& \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^*(H_2(v)) \\
&= \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* \left(\hat{L}_1(\frac{\varsigma}{4}) \in \tilde{I}, \hat{L}_i(\frac{\varsigma}{4}) \leq \frac{g}{6}, \sup_{u \leq \frac{\varsigma}{4}} |B_i(u)| < \frac{g}{8} \text{ for } 2 \leq i \leq N, \right. \\
&\quad \left. \sup_{u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}, \hat{Z}_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}] \right) \\
&\geq \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* \left(\hat{L}_1(\frac{\varsigma}{4}) \in \tilde{I}, \sup_{u \leq \frac{\varsigma}{4}} (-B_2(u)) \leq -\frac{13g}{30} + \hat{z}_2, \hat{L}_3(\frac{\varsigma}{4}) \leq \frac{g}{6}, \dots, \hat{L}_N(\frac{\varsigma}{4}) \leq \frac{g}{6}, \right. \\
&\quad \left. \sup_{u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}, \sup_{u \leq \frac{\varsigma}{4}} |B_i(u)| < \frac{g}{8} \text{ for } 2 \leq i \leq N, \hat{Z}_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}] \right) \\
&= K_{\hat{z}} \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* \left(\hat{L}_1(\frac{\varsigma}{4}) \in \tilde{I}, \sup_{u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}, \hat{Z}_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}] \right), \tag{4.20}
\end{aligned}$$

where

$$\begin{aligned}
K_{\hat{z}} &= \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* \left(\sup_{u \leq \frac{\varsigma}{4}} (-B_2(u)) \leq -\frac{13g}{30} + \hat{z}_2, \hat{L}_3(\frac{\varsigma}{4}) \leq \frac{g}{6}, \dots, \hat{L}_N(\frac{\varsigma}{4}) \right. \\
&\quad \left. \leq \frac{g}{6}, \sup_{u \leq \frac{\varsigma}{4}} |B_i(u)| < \frac{g}{8} \text{ for } 2 \leq i \leq N \right),
\end{aligned}$$

and in the last step we have used the independence of B_1 and (B_2, \dots, B_N) .

Note that $\hat{Z}_1(\frac{\varsigma}{4}) = B_1(\frac{\varsigma}{4}) + \hat{L}_1(\frac{\varsigma}{4})$, so if $\hat{L}_1(\frac{\varsigma}{4}) \in [0, \frac{g}{63}]$ and $B_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{8}]$, then $\hat{Z}_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}]$. Consequently,

$$\begin{aligned}
& \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* (\hat{L}_1(\frac{\varsigma}{4}) \in \tilde{I}, \sup_{u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}, \hat{Z}_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{4}]) \\
&\geq \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* (\sup_{u \leq \frac{\varsigma}{4}} (-B_1(u)) \in \tilde{I}, \sup_{u \leq \frac{\varsigma}{4}} B_1(u) < \frac{6g}{10}, B_1(\frac{\varsigma}{4}) \in [\frac{g}{16}, \frac{g}{8}]) \\
&= \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* (\sup_{u \leq \frac{\varsigma}{4}} (B_1(u)) \in \tilde{I}, \inf_{u \leq \frac{\varsigma}{4}} B_1(u) > -\frac{6g}{10}, B_1(\frac{\varsigma}{4}) \in [-\frac{g}{8}, -\frac{g}{16}]),
\end{aligned}$$

where in the last line we have used the fact that $\{B(s)\}_{s \leq \frac{\varsigma}{4}}$ is equal in distribution to $\{-B(s)\}_{s \leq \frac{\varsigma}{4}}$.

Applying Lemma 4.5 we have

$$\begin{aligned}
& \tilde{\mathbb{P}}_{(\hat{v}, (0, \hat{z}))}^* (\sup_{u \leq \frac{\varsigma}{4}} (B_1(u)) \in \tilde{I}, \inf_{u \leq \frac{\varsigma}{4}} B_1(u) \\
&\quad > -\frac{6g}{10}, B_1(\frac{\varsigma}{4}) \in [-\frac{g}{8}, -\frac{g}{16}]) \geq \mathcal{H}\lambda(\tilde{I} \cap [0, \frac{g}{63}]) \\
&= \mathcal{H}\lambda(\tilde{I}) = \mathcal{H}\lambda(I) = \mathcal{H}\lambda(I \cap D_1), \tag{4.21}
\end{aligned}$$

where for the last equality we have used that $I \subseteq [0, g/128] = D_1$. Thus, letting

$$\hat{K} \doteq \inf_{\mathbf{z} \in [g, \frac{3g}{2}]^{N-1}} K_{\mathbf{z}},$$

we have on combining estimates in (4.11), (4.13), (4.14), (4.17), (4.18), (4.19), (4.20), (4.21),

$$\begin{aligned} \mathbb{P}^r((v, \mathbf{z}), S) &\geq e^{-2|A^{-1}\mathbf{e}_1|^2 g^2} e^{-5gM'(v, \mathbf{z})} \kappa_{(v, \mathbf{z})} e^{-3g\kappa_G} c_G e^{-4gc_A} \hat{K} \mathcal{K} \lambda(J \cap G_1) \lambda(I \cap D_1) \\ &= K_{(v, \mathbf{z})} \lambda((I \times J) \cap D) \end{aligned}$$

where

$$K_{(v, \mathbf{z})} = e^{-2|A^{-1}\mathbf{e}_1|^2 g^2} e^{-5gM'(v, \mathbf{z})} \kappa_{(v, \mathbf{z})} e^{-3g\kappa_G} c_G e^{-4gc_A} \hat{K} \mathcal{K}.$$

This proves the first statement in the theorem. The second statement is immediate from the measurability of $(v, \mathbf{z}) \mapsto \kappa_{(v, \mathbf{z})}$ indicated earlier in the proof and (4.16). \square

5. Stationary Distribution: Uniqueness

In this section, we establish uniqueness of the stationary distribution by using the minorization estimate in Theorem 4.1 in conjunction with the following lemma. This lemma also plays a crucial role in establishing exponential ergodicity of the system.

Lemma 5.1. *For each $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, there exists $r_0 \doteq r_0(v, \mathbf{z}) \in \mathbb{N}$ such that*

$$\mathbb{P}^{r_0}((v, \mathbf{z}), R) > 0,$$

where

$$R \doteq (0, \frac{g}{128}) \times (0, \infty) \times \mathbb{R}_+^{N-1}.$$

Furthermore, if $v \geq g/128$, we can take $r_0 = 1$.

Proof. Let $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ be given. In view of Corollary 4.3 it suffices to show that for some $r_0 \in \mathbb{N}$

$$\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*((V(r_0), \mathbf{Z}(r_0)) \in R) > 0.$$

Consider first the case where $v < g/128$. Define

$$v_1 \doteq \begin{cases} \frac{g}{256} - v, & v < \frac{g}{256}, \\ \frac{1}{2}(\frac{g}{128} - v), & v \in [\frac{g}{256}, \frac{g}{128}). \end{cases}$$

Set $v_2 \doteq v + v_1$. Write $v_1 = g(k + t_1)$ with $k \in \mathbb{N}_0$ and $t_1 \in [0, 1)$. Let $r_0 \doteq k + 1$ and $t_2 \doteq (k + t_1)/2$. Let $v_3 \doteq gr_0 - v_1$. Fix $\delta \in (0, v_3)$ such that $[v_2 - \delta, v_2 + \delta] \subset (0, g/128)$. Consider the set $A_1 \in \mathcal{F}^*$ defined as

$$A_1 \doteq \{L_1(t_2) \in [v_3 - \delta, v_3 + \delta], Z_1(t) > 0 \text{ for all } t \in (t_2, r_0)\}.$$

Then on A_1 , $\mathbf{Z}(r_0) \in (0, \infty) \times \mathbb{R}_+^{N-1}$ and

$$\begin{aligned} V(r_0) &= v + (k + 1)g - L_1(r_0) = v + (k + 1)g - L_1(t_2) \in v + (k + 1)g - [v_3 - \delta, v_3 + \delta] \\ &= [v + (k + 1)g - v_3 - \delta, v + (k + 1)g - v_3 + \delta]. \end{aligned}$$

Also

$$v + (k+1)g - v_3 = v + (k+1)g - (g(k+1) - v_1) = v + v_1 = v_2$$

Thus on A_1 , $V(r_0) \in [v_2 - \delta, v_2 + \delta] \subset (0, g/128)$ and consequently $A_1 \subset \{(V(r_0), \mathbf{Z}(r_0)) \in R\}$. It is easily verified that $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(A_1) > 0$ which proves the result for the case $v < g/128$.

Now consider the case $v \geq g/128$. Let $v_1 \doteq v + g - g/256$ and fix $\delta \in (0, g/256)$. Consider the set $A_2 \in \mathcal{F}^*$ defined as

$$A_2 \doteq \{L_1(1/2) \in [v_1 - \delta, v_1 + \delta], Z_1(t) > 0 \text{ for all } t \in (1/2, 1]\}.$$

Then, with $r_0 = 1$, we see, on A_2 , $\mathbf{Z}(r_0) \in (0, \infty) \times \mathbb{R}_+^{N-1}$ and

$$\begin{aligned} V(r_0) &= v + g - L_1(r_0) = v + g - L_1(1/2) \in v + g - [v_1 - \delta, v_1 + \delta] \\ &= [v + g - v_1 - \delta, v + g - v_1 + \delta] = [g/256 - \delta, g/256 + \delta] \\ &= [v_2 - \delta, v_2 + \delta] \subset (0, g/128). \end{aligned}$$

Thus $A_2 \subset \{(V(r_0), \mathbf{Z}(r_0)) \in R\}$. Once again, it is easily verified that $\tilde{\mathbb{P}}_{(v, \mathbf{z})}^*(A_2) > 0$ proving the result for the case $v \geq g/128$ with $r_0 = 1$. \square

Theorem 5.2. *The Markov family $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$ has at most one stationary distribution.*

Proof. By Birkhoff's ergodic theorem, if there are multiple stationary distributions, then we can find two that are mutually singular [23, Chapter 4, Theorem 4.4 and Lemma 4.6]. Suppose that π, π' are mutually singular stationary distributions. Then there is a $A \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ such that $\pi(A) = \pi'(A^c) = 0$. Recall the set D from Theorem 4.1. Since $\lambda(D) > 0$, it follows that either $\lambda(D \cap A) > 0$ or $\lambda(D \cap A^c) > 0$. For specificity, suppose $\lambda(D \cap A) > 0$. We will now argue that $\pi(A) > 0$, arriving at a contradiction. By Theorem 4.1, with R as in Lemma 5.1, for every $(v, \mathbf{z}) \in R$, there is a $K_{(v, \mathbf{z})} > 0$ such that

$$\mathbb{P}^\zeta((v, \mathbf{z}), A) \geq K_{(v, \mathbf{z})} \lambda(A \cap D).$$

Define the transition probability kernel \mathbb{Q} on $\mathbb{R} \times \mathbb{R}_+^N$ as

$$\mathbb{Q}((\tilde{v}, \tilde{\mathbf{z}}), S) \doteq \sum_{i=1}^{\infty} \frac{1}{2^i} \mathbb{P}^{i+\zeta}((\tilde{v}, \tilde{\mathbf{z}}), S), \quad (\tilde{v}, \tilde{\mathbf{z}}) \in \mathbb{R} \times \mathbb{R}_+^N, \quad S \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N).$$

Since π is a stationary distribution, we have

$$\pi(A) = \int_{\mathbb{R} \times \mathbb{R}_+^N} \mathbb{Q}((\tilde{v}, \tilde{\mathbf{z}}), A) d\pi(\tilde{v}, \tilde{\mathbf{z}}). \quad (5.1)$$

Also, for any $(\tilde{v}, \tilde{\mathbf{z}}) \in \mathbb{R} \times \mathbb{R}_+^N$ and with $r_0 = r_0(\tilde{v}, \tilde{\mathbf{z}}) \in \mathbb{N}$ as in Lemma 5.1,

$$\begin{aligned} \mathbb{Q}((\tilde{v}, \tilde{\mathbf{z}}), A) &\geq 2^{-r_0} \mathbb{P}^{r_0+\zeta}((\tilde{v}, \tilde{\mathbf{z}}), A) \geq 2^{-r_0} \int_R \mathbb{P}^{r_0}((\tilde{v}, \tilde{\mathbf{z}}), (dv, d\mathbf{z})) \mathbb{P}^\zeta((v, \mathbf{z}), A) \\ &\geq 2^{-r_0} \lambda(A \cap D) \int_R \mathbb{P}^{r_0}((\tilde{v}, \tilde{\mathbf{z}}), (dv, d\mathbf{z})) K_{(v, \mathbf{z})} > 0. \end{aligned}$$

From (5.1) it now follows that $\pi(A) > 0$ which gives a contradiction and proves the result. \square

6. Product Form of Stationary Density

In this section, we prove Theorem 2.4. The proof relies on ‘guessing’ a product form for the stationary joint density and showing that it satisfies the partial differential equations (along with appropriate boundary conditions) that characterize such stationary densities. This guess is inspired by [4], where a product form joint density was obtained for the velocity and gap processes of the system comprising one inert particle and one Brownian particle.

Proof of Theorem 2.4. The generator of the process (V, \mathbf{Z}) given by (1.3) acts on any $f : \mathbb{R} \times \mathbb{R}_+^N \rightarrow \mathbb{R}$ that is continuously differentiable in v and twice continuously differentiable in (z_1, \dots, z_N) , and compactly supported in the interior of $\mathbb{R} \times \mathbb{R}_+^N$, by

$$\begin{aligned} \mathcal{L}f(v, \mathbf{z}) = & \frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} \frac{\partial f}{\partial z_i \partial z_j}(v, \mathbf{z}) + g \frac{\partial f}{\partial v}(v, \mathbf{z}) \\ & - v \frac{\partial f}{\partial z_1}(v, \mathbf{z}), \quad (v, \mathbf{z}) \in \mathbb{R} \times (0, \infty)^N, \end{aligned}$$

where $h_{11} = 1$, $h_{ii} = 2$ for $2 \leq i \leq N$, $h_{ij} = -1$ for $|i - j| = 1$, and $h_{ij} = 0$ otherwise. Moreover, from the pathwise existence and uniqueness (Theorem 2.2) it readily follows that the associated submartingale problem [27, Definition 2.1] for our process is well-posed. For $c_0, c_1, \dots, c_N, \phi > 0$, consider the function

$$\pi(v, \mathbf{z}) = c_\pi e^{-c_0(v+\phi)^2} \prod_{i=1}^N e^{-c_i z_i}, \quad (v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N, \quad (6.1)$$

where c_π is the normalization constant ensuring $\int_{\mathbb{R} \times \mathbb{R}_+^N} \pi(v, \mathbf{z}) dv d\mathbf{z} = 1$. Translating the conditions (1)–(3) of [27, Theorem 3], π is the density of a stationary distribution if π satisfies the interior condition

$$\begin{aligned} \mathcal{L}^* \pi(v, \mathbf{z}) \doteq & \frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} \frac{\partial \pi}{\partial z_i \partial z_j}(v, \mathbf{z}) - g \frac{\partial \pi}{\partial v}(v, \mathbf{z}) \\ & + \frac{\partial(v\pi)}{\partial z_1}(v, \mathbf{z}) = 0, \quad (v, \mathbf{z}) \in \mathbb{R} \times (0, \infty)^N, \end{aligned} \quad (6.2)$$

and boundary conditions

$$2v\pi(v, \mathbf{z}) + \frac{\partial \pi}{\partial z_1}(v, \mathbf{z}) - \frac{\partial \pi}{\partial z_2}(v, \mathbf{z}) + \frac{\partial \pi}{\partial v}(v, \mathbf{z}) = 0 \quad \text{if } z_1 = 0, \quad (6.3)$$

$$-\frac{\partial \pi}{\partial z_{i-1}}(v, \mathbf{z}) + 2\frac{\partial \pi}{\partial z_i}(v, \mathbf{z}) - \frac{\partial \pi}{\partial z_{i+1}}(v, \mathbf{z}) = 0 \quad \text{if } z_i = 0, \text{ for some } 2 \leq i \leq N-1, \quad (6.4)$$

$$-\frac{\partial \pi}{\partial z_{N-1}}(v, \mathbf{z}) + 2\frac{\partial \pi}{\partial z_N}(v, \mathbf{z}) = 0 \quad \text{if } z_N = 0. \quad (6.5)$$

We will solve for the constants $c_0, c_1, \dots, c_N, c_\pi$ to obtain a π satisfying the above conditions.

The conditions (6.4) and (6.5) applied to (6.1) yield that

$$\begin{aligned} c_{i-1} - 2c_i + c_{i+1} &= 0, \quad i = 2, \dots, N-1, \\ c_{N-1} - 2c_N &= 0. \end{aligned} \quad (6.6)$$

From these identities, we obtain that

$$\begin{aligned} c_{N-1} &= 2c_N \\ c_{N-2} &= 2c_{N-1} - c_N = 3c_N. \end{aligned}$$

Fix $j \in \{2, \dots, N-1\}$. Suppose that we have $c_i = (N-i+1)c_N$ for all $j \leq i \leq N$. Then, from (6.6),

$$\begin{aligned} c_{j-1} - 2c_j + c_{j+1} &= (2(N-j+1) - (N-j))c_N \\ &= (N-(j-1)+1)c_N. \end{aligned}$$

Hence, we have by induction that $c_i = (N-i+1)c_N$ for $i = 1, \dots, N$. Substituting this into (6.3), we see that

$$2v - c_1 + c_2 - 2c_0(v + \phi) = 0, \quad \text{for all } v \in \mathbb{R}.$$

Since this holds for all $v \in \mathbb{R}$, we must have $c_0 = 1$, and so

$$2\phi = c_2 - c_1 = (N-1)c_N - Nc_N = -c_N$$

and thus $c_N = -2\phi$. Next substituting (6.1) in (6.2),

$$\frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} c_i c_j + 2g c_0(v + \phi) - c_1 v = 0, \quad \text{for all } v \in \mathbb{R}. \quad (6.7)$$

Again, since this holds for all $v \in \mathbb{R}$, we must have,

$$2g = c_1 = Nc_N.$$

From the above relations, we obtain

$$c_0 = 1, \quad c_i = 2 \left(\frac{N-i+1}{N} \right) g, \quad i = 1, \dots, N, \quad \phi = -\frac{g}{N}. \quad (6.8)$$

To show that this choice of constants yields a valid density for some stationary distribution, it remains only to demonstrate that (6.2) holds for all $(v, \mathbf{z}) \in \mathbb{R} \times (0, \infty)^N$, or equivalently, from (6.7) and (6.8),

$$\frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} c_i c_j - 2 \frac{g^2}{N} = 0. \quad (6.9)$$

To see this, note that

$$\begin{aligned} \frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} c_i c_j &= \frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} \left(\frac{N-i+1}{N} \right) \left(\frac{N-j+1}{N} \right) 4g^2 \\ &= \frac{2g^2}{N^2} \sum_{1 \leq i, j \leq N} h_{ij} (N-i+1)(N-j+1) \\ &= \frac{2g^2}{N^2} \sum_{i=1}^N (N-i+1) \sum_{j=1}^N h_{ij} (N-j+1). \end{aligned}$$

From the formulae of $\{h_{ij}\}_{1 \leq i, j \leq N}$, it follows that

$$\sum_{j=1}^N h_{ij}(N-j+1) = \delta_{1,i}, \quad \text{for all } i = 1, \dots, N,$$

where $\delta_{1,i}$ is the Kronecker delta function. Hence,

$$\frac{1}{2} \sum_{1 \leq i, j \leq N} h_{ij} c_i c_j = \frac{2g^2}{N^2} \sum_{i=1}^N (N-i+1) \delta_{1,i} = \frac{2g^2}{N},$$

which proves (6.9). We have therefore shown that π with constants as in (6.8) is indeed the density for a stationary distribution of the process (V, \mathbf{Z}) . Uniqueness follows from Theorem 2.3. \square

7. Exponential Ergodicity

In this section we will prove Theorem 2.5. Since the main source of stability in our system is the local time interactions between particles, standard PDE techniques for constructing Lyapunov functions [22, 29, 31] for hypoelliptic diffusions are hard to implement. Furthermore, the singular nature of the dynamics arising from the motion of the inert particle, and the spatial dependence of the drift (which contains a V term), make it challenging to adapt the Lyapunov function constructions for reflected Brownian motions, which proceed via an analysis of the associated noiseless system [1, 21].

7.1. Outline of approach. Here, we take a different approach to exponential ergodicity by analyzing excursions of the process between appropriately chosen stopping times (see (7.14)) as the velocity of the inert particle ‘toggles’ between two levels. Control on the exponential moments of these stopping times is established in Sects. 7.3 and 7.4. In Sect. 7.5, it is shown that the intersection local time between the bottom two particles creates a ‘singular’ drift that results in a reduction of the function $\bar{Z}_2(t) \doteq \sum_{i=1}^{N-1} i Z_{N-i+1}(t)$ of the gaps when observed at successive stopping times. These estimates are combined in Sect. 7.6 to show that the distribution of return times of the process to an appropriately chosen compact set C^* has exponentially decaying tails. Finally, in Sect. 7.7, the exponential moments of this return time are used to construct a suitable Lyapunov function. The minorization estimate in Theorem 4.1 is utilized to show that C^* is a ‘petite’ (or small) set in the language of [18]. These facts together imply exponential ergodicity using the machinery developed in [18] (see Theorem 6.2 there). Proofs of some technical lemmas are deferred to Sect. 8 in order to make it easier to see the overall idea.

We note here that the connection between finiteness of exponential moments of certain hitting times, Lyapunov functions and exponential ergodicity is not new [18, 32]. The main work in this section is in establishing that exponential moments of associated hitting times are finite through a detailed pathwise analysis of the process. A general treatment of the above connection in the context of diffusion processes has been undertaken in [15, 33], among others. However, typically the diffusion processes are assumed to be hypoelliptic and/or reversible with respect to the stationary measure, neither of which apply to our setting.

7.2. An inequality for the local time. In this section we establish an estimate on local times which will be used several times. Recall the matrix W from Sect. 2 and the process \mathbf{B}^* from (4.4).

Lemma 7.1. *For any $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ and $t \geq 0$, the following inequality holds, $\mathbb{P}_{(v, \mathbf{z})}^*$ – a.s., for all $i = 1, 2, \dots, N$,*

$$L_i(t) \leq W_{i,1}t \sup_{0 \leq s \leq t} (V(s))^+ + \sum_{j=1}^N W_{i,j} B_j^*(t). \quad (7.1)$$

Moreover, with $\bar{Y}(t) \doteq \sum_{i=2}^N (N - i + 1) B_i^*(t)$,

$$L_2(t) \leq \frac{2(N-1)}{N} L_1(t) + \frac{2}{N} \bar{Y}(t), \quad t \geq 0. \quad (7.2)$$

Proof. Let $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ be given. All inequalities in the proof are a.s. under $\mathbb{P}_{(v, \mathbf{z})}^*$. For $1 \leq i \leq N$, the local times L_i are given as

$$\begin{aligned} L_1(t) &= \sup_{s \leq t} (-z_1 + \frac{1}{2} L_2(s) + \int_0^s V(u) du - B_1(s))^+ \\ L_2(t) &= \sup_{s \leq t} (-z_2 + \frac{1}{2} L_3(s) + L_1(s) + B_1(s) - B_2(s))^+ \\ L_i(t) &= \sup_{s \leq t} (-z_i + \frac{1}{2} (L_{i-1}(s) + L_{i+1}(s)) + B_{i-1}(s) - B_i(s))^+, \quad i = 3, \dots, N. \end{aligned} \quad (7.3)$$

Using these identities we see that

$$\begin{aligned} L_1(t) &\leq \sup_{0 \leq s \leq t} (V(s))^+ t + \frac{1}{2} L_2(t) + \sup_{s \leq t} (-B_1(s)) \\ L_2(t) &\leq \frac{1}{2} L_3(t) + L_1(t) + \sup_{s \leq t} (B_1(s) - B_2(s)) \\ L_i(t) &\leq \frac{1}{2} (L_{i+1}(t) + L_{i-1}(t)) + \sup_{s \leq t} (B_{i-1}(s) - B_i(s)), \quad i = 3, \dots, N. \end{aligned} \quad (7.4)$$

Recalling the matrix U from (4.5) the above inequalities can be written as

$$\mathbf{L}(t) \leq \sup_{0 \leq s \leq t} (V(s))^+ t \mathbf{e}_1 + U \mathbf{L}(t) + \mathbf{B}^*(t), \quad t \geq 0.$$

Using the fact that $W = (I - U)^{-1}$ is a matrix with nonnegative entries, we have,

$$\mathbf{L}(t) \leq \sup_{0 \leq s \leq t} (V(s))^+ t W \mathbf{e}_1 + W \mathbf{B}^*(t), \quad t \geq 0.$$

This proves the first statement in the lemma. For the second inequality, note that from (7.4) we have

$$\begin{aligned} &\sum_{i=3}^N (N - i + 1) (L_i(t) - \frac{1}{2} L_{i+1}(t) - \frac{1}{2} L_{i-1}(t)) + (N - 1) (L_2(t) - L_1(t) - \frac{1}{2} L_3(t)) \\ &\leq \sum_{i=2}^N (N - i + 1) B_i^*(t) = \bar{Y}(t). \end{aligned}$$

Simplifying the left side, we see,

$$\frac{N}{2}L_2(t) - (N-1)L_1(t) \leq \bar{Y}(t)$$

which proves the second statement. \square

7.3. Hitting time of an upper velocity level. For $c \in \mathbb{R}$, let $\hat{\tau}_c \doteq \inf\{t \geq 0 : V(t) = c\}$. The main result of this section is the following control on exponential moments of this hitting time.

Proposition 7.2. *There exists a $\gamma \in (0, \infty)$ such that*

$$\sup_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{E}_{(v, \mathbf{z})}^* e^{\gamma \hat{\tau}_{4g}} < \infty.$$

Proof of the proposition relies on the three lemmas given below. Proofs of these lemmas are given in Sect. 8.1. The proposition is proved after the statements of these lemmas.

Lemma 7.3. *There exists a $\beta \in (0, \infty)$ so that*

$$\sup_{\mathbf{z} \in \mathbb{R}_+^N} \mathbb{E}_{(0, \mathbf{z})}^* e^{\beta \hat{\tau}_{g/(2N)}} < \infty.$$

Lemma 7.4. *We have*

$$\inf_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^* (\hat{\tau}_{4g} < \hat{\tau}_0) \doteq p > 0.$$

Lemma 7.5. *There exists $\gamma_1 > 0$ so that*

$$\sup_{(v, \mathbf{z}) \in [0, 4g] \times \mathbb{R}_+^N} \mathbb{E}_{(v, \mathbf{z})}^* e^{\gamma_1 (\hat{\tau}_{4g} \wedge \hat{\tau}_0)} < \infty.$$

We now prove the main result of the section.

Proof of Proposition 7.2. Define $\tau_{-1} = \tau_0 \doteq 0$ and for $i \in \mathbb{N}_0$, define

$$\tau_{2i+1} \doteq \inf\{t \geq \tau_{2i} : V(t) = 4g \text{ or } 0\}, \quad \tau_{2i+2} \doteq \inf\{t \geq \tau_{2i+1} : V(t) = \frac{g}{2N} \text{ or } 4g\}.$$

Define

$$\mathcal{N} = \inf\{k \geq 0 : V(\tau_{2k+1}) = 4g\}.$$

From Lemma 7.4 it follows that

$$\sup_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^* (\mathcal{N} = k) \leq (1-p)^{k-1}, \quad k \geq 0.$$

Note that

$$\hat{\tau}_{4g} \leq \sum_{i=0}^{\mathcal{N}} (\tau_{2i+1} - \tau_{2i-1}) \leq \sum_{i=1}^{\mathcal{N}+1} (\tau_{2i} - \tau_{2i-2}). \quad (7.5)$$

By Lemmas 7.3 and 7.5 there are $c_1, c_2 \in (0, \infty)$ such that

$$\sup_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(\tau_2 \geq t) \leq c_1 e^{-c_2 t}.$$

It then follows that, for $0 < \alpha < c_2$,

$$\begin{aligned} \sup_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{E}_{(v, \mathbf{z})}^* e^{\alpha \tau_2} &\leq \int_{-\infty}^{\infty} \alpha e^{\alpha s} \sup_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(\tau_2 \geq s) ds \\ &\leq 1 + \alpha c_1 \int_0^{\infty} e^{(\alpha - c_2)s} ds = 1 + \frac{\alpha c_1}{c_2 - \alpha}. \end{aligned}$$

Choose $\delta \in (0, 1)$ such that $(1 + \delta)(1 - p) \doteq \kappa < 1$. Now choose $\alpha > 0$ sufficiently small such that

$$\sup_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{E}_{(v, \mathbf{z})}^* e^{2\alpha \tau_2} \leq (1 + \delta).$$

Applying Cauchy-Schwarz and the Strong Markov property we now see that, for any $(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N$,

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{\alpha \sum_{i=1}^{\mathcal{N}+1} (\tau_{2i} - \tau_{2i-2})} &= \sum_{k=0}^{\infty} \mathbb{E}_{(v, \mathbf{z})}^* e^{\alpha \sum_{i=1}^{k+1} (\tau_{2i} - \tau_{2i-2})} \mathbf{1}_{\{\mathcal{N}=k\}} \\ &\leq \sum_{k=0}^{\infty} (\mathbb{E}_{(v, \mathbf{z})}^* e^{2\alpha \sum_{i=1}^{k+1} (\tau_{2i} - \tau_{2i-2})})^{\frac{1}{2}} (\mathbb{P}_{(v, \mathbf{z})}^*(\mathcal{N} = k))^{\frac{1}{2}} \\ &\leq \frac{2}{(1 - p)^{1/2}} \sum_{k=0}^{\infty} (1 + \delta)^{k/2} (1 - p)^{\frac{k}{2}} \leq \sum_{k=0}^{\infty} \kappa^{k/2} < \infty. \end{aligned}$$

The result now follows on combining the above estimate with (7.5). \square

7.4. Hitting time of a lower velocity level. Let $\sigma_1 \doteq \hat{\tau}_{4g} = \inf\{t \geq 0 : V(t) = 4g\}$ and set $\sigma_2 \doteq \inf\{t \geq \sigma_1 : V(t) = 2g\}$. The main result of the section is the following.

Proposition 7.6. *There is $\gamma_2 > 0$ such that*

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\gamma_2 \sigma_2} < \infty.$$

The proof relies on the following three lemmas. Proofs of these lemmas are given in Sect. 8.2. The proposition is proved after the statements of the lemmas.

Lemma 7.7. *There is a $\gamma_3 > 0$ such that*

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\gamma_3 Z_1(\sigma_1)^{1/2}} < \infty.$$

Lemma 7.8. Define $\tau_0^{Z_1} \doteq \inf\{t \geq 0 : Z_1(t) = 0\}$. There is a $\gamma_4 > 0$ and $\kappa_1, \kappa_2 \in (0, \infty)$ such that for any $z_1 \in \mathbb{R}_+$ and $\gamma \in (0, \gamma_4]$

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(4g, z_1, \hat{\mathbf{z}})}^* e^{\gamma \tau_0^{Z_1}} \leq \kappa_1 e^{\kappa_2 \gamma z_1^{1/2}}.$$

Lemma 7.9. There exists a $\gamma_5 > 0$ and $\kappa'_1, \kappa'_2 \in (0, \infty)$ such that for all $\gamma \in (0, \gamma_5)$ and $v \in [2g, \infty)$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(v, 0, \hat{\mathbf{z}})}^* e^{\gamma \hat{\tau}_{2g}} \leq \kappa'_1 e^{\kappa'_2 \gamma v}.$$

We now prove the main result of this section.

Proof of Proposition 7.6 Let $\alpha \in (0, 1)$ be such that

$$\alpha < \gamma_5, \quad \alpha(1 + \kappa'_2 g) \leq \gamma_4, \quad 2\alpha(1 + \kappa'_2 g)\kappa_2 \leq \gamma_3, \quad 2\alpha(1 + \kappa'_2 g) \leq \gamma, \quad (7.6)$$

where γ_5 and κ'_2 are as in Lemma 7.9, κ_2 and γ_4 are as in Lemma 7.8, γ_3 is as in Lemma 7.7 and γ is as in Proposition 7.2.

Fix $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$. Define stopping time

$$\eta_1 \doteq \inf\{t \geq \sigma_1 : Z_1(t) = 0\}. \quad (7.7)$$

Note that $\sigma_2 = \inf\{t \geq \eta_1 : V(t) = 2g\}$. From the strong Markov property, Lemma 7.9, and recalling the first condition on α from (7.6),

$$\begin{aligned} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\alpha \sigma_2} &= \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* \left[\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* (e^{\alpha \sigma_2} \mid \mathcal{F}_{\eta_1}^*) \right] \leq \kappa'_1 \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\kappa'_2 \alpha V(\eta_1) + \alpha \eta_1} \\ &\leq \kappa'_1 \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\kappa'_2 \alpha (4g + g\eta_1) + \alpha \eta_1} \leq \kappa'_1 e^{4\kappa'_2 \alpha g} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\alpha(1 + \kappa'_2 g)\eta_1}. \end{aligned}$$

Thus, with $d_1 = \kappa'_1 e^{4\kappa'_2 \alpha g}$ and $d_2 = (1 + \kappa'_2 g)$,

$$\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\alpha \sigma_2} \leq d_1 \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\alpha d_2 \eta_1}. \quad (7.8)$$

Using the strong Markov property again,

$$\begin{aligned} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\alpha d_2 \eta_1} &= \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* \left[\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* (e^{\alpha d_2 \eta_1} \mid \mathcal{F}_{\sigma_1}^*) \right] \\ &\leq \kappa_1 \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\kappa_2 \alpha d_2 Z_1(\sigma_1)^{1/2} + \alpha d_2 \sigma_1} \\ &\leq \kappa_1 \left(\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{2\kappa_2 \alpha d_2 Z_1(\sigma_1)^{1/2}} \right)^{1/2} \left(\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{2\alpha d_2 \sigma_1} \right)^{1/2}, \end{aligned}$$

where the second inequality is from Lemma 7.8 and on recalling the second condition on α from (7.6), and the last line is from Cauchy-Schwarz inequality. Next, applying Lemma 7.7, and recalling the third condition on α from (7.6),

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{2\kappa_2 \alpha d_2 Z_1(\sigma_1)^{1/2}} \doteq d_3 < \infty.$$

Finally, applying Proposition 7.2 and recalling the fourth condition on α from (7.6) we have

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{2\alpha d_2 \sigma_1} \doteq d_4 < \infty.$$

Combining the above estimates we have

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\alpha \sigma_2} \leq d_1 \kappa_1 d_3^{1/2} d_4^{1/2} < \infty.$$

The result follows. \square

7.5. A negative singular drift property. For $\mathbf{z} \in \mathbb{R}_+^N$, define $\bar{z}_2 \doteq \sum_{i=2}^N (N-i+1)z_i$. Similarly, for \mathbb{R}_+^N valued process $\{\mathbf{Z}(t)\}$, we define for $t \geq 0$,

$$\bar{Z}_2(t) \doteq \sum_{i=2}^N (N-i+1)Z_i(t) = \sum_{i=1}^{N-1} iZ_{N-i+1}(t). \quad (7.9)$$

The main result of this section is Proposition 7.11, where we will show that if \bar{z}_2 is large, then the process $\bar{Z}_2(\cdot)$ decreases in expectation in the course of an appropriately large number of excursions of the velocity process between the levels $2g$ and $4g$ (see (7.14)).

The following lemma gives a key algebraic representation of $\bar{Z}_2(\cdot)$ in terms of L_1 , L_{k+1} for $k \in \{1, \dots, N-1\}$, and additional error terms. If \bar{z}_2 is large, then there exists $k \in \{1, \dots, N-1\}$ such that $Z_{k+1}(0) = z_{k+1}$ is large. Thus, it takes a long time for this gap to hit zero. Before this time, the lowest (inert) particle ‘pushes’ the bottom $k+1$ particles up and thereby reduces $\bar{Z}_2(\cdot)$, as captured by the L_1 term in the lemma. This ‘singular’ drift through local times results in stability and, in turn, exponential ergodicity, of the system.

Lemma 7.10. *Let $Y_1^{(1)}(t) = 0$ and $Y_k^{(1)}(t) \doteq \sum_{i=2}^k (k-i+1)B_i^*(t)$, $t \geq 0$, $2 \leq k \leq N$. Also define $M(t) \doteq \sum_{i=2}^N B_i(t) - (N-1)B_1(t)$, $t \geq 0$. Then for all $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, $\mathbb{P}_{(v, \mathbf{z})}^*$ a.s.,*

$$\bar{Z}_2(t) - \bar{z}_2 \leq M(t) + \frac{N}{k} Y_k^{(1)}(t) + \frac{N}{2k} L_{k+1}(t) - \frac{(N-k)}{k} L_1(t), \quad t \geq 0, \quad 1 \leq k \leq N, \quad (7.10)$$

with equality for $k = 1$.

Proof. Note that, for $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, under $\mathbb{P}_{(v, \mathbf{z})}^*$, for $t \geq 0$,

$$\begin{aligned}
 \bar{Z}_2(t) &= -\frac{(N-1)}{2}L_1(t) + \sum_{i=2}^N (N-i+1) \\
 &\quad \left[(B_i(t) - B_{i-1}(t) - \frac{1}{2}(L_{i+1}(t) + L_{i-1}(t)) + L_i(t) + z_i) \right] \\
 &= -\frac{(N-1)}{2}L_1(t) + \sum_{i=1}^{N-1} i \\
 &\quad \left[(B_{N-i+1}(t) - B_{N-i}(t) - \frac{1}{2}(L_{N-i+2}(t) + L_{N-i}(t)) + L_{N-i+1}(t) + z_{N-i+1}) \right] \\
 &= \bar{z}_2 - \frac{(N-1)}{2}L_1(t) + \sum_{i=1}^{N-1} i B_{N-i+1}(t) - \sum_{i=1}^{N-1} i B_{N-i}(t) + \sum_{i=1}^{N-1} i (L_{N-i+1}(t) \\
 &\quad - \frac{1}{2}(L_{N-i+2}(t) + L_{N-i}(t))).
 \end{aligned}$$

Also,

$$\begin{aligned}
 \sum_{i=1}^{N-1} i B_{N-i+1}(t) - \sum_{i=1}^{N-1} i B_{N-i}(t) &= \sum_{i=0}^{N-2} (i+1) B_{N-i}(t) - \sum_{i=1}^{N-1} i B_{N-i}(t) \\
 &= \sum_{i=2}^N B_i(t) - (N-1)B_1(t).
 \end{aligned}$$

Moreover,

$$\begin{aligned}
 &\sum_{i=1}^{N-1} i (L_{N-i+1}(t) - \frac{1}{2}(L_{N-i+2}(t) + L_{N-i}(t))) \\
 &= -\frac{1}{2} \sum_{i=1}^{N-1} i (L_{N-i+2}(t) - L_{N-i+1}(t)) + \frac{1}{2} \sum_{i=1}^{N-1} i (L_{N-i+1}(t) - L_{N-i}(t)) \\
 &= -\frac{1}{2} \sum_{i=0}^{N-2} (i+1) (L_{N-i+1}(t) - L_{N-i}(t)) + \frac{1}{2} \sum_{i=1}^{N-1} i (L_{N-i+1}(t) - L_{N-i}(t)) \\
 &= \frac{(N-1)}{2} (L_2(t) - L_1(t)) + \frac{1}{2} L_2(t).
 \end{aligned}$$

Hence,

$$\begin{aligned}\bar{Z}_2(t) &= \bar{z}_2 - \frac{(N-1)}{2}L_1(t) + \sum_{i=2}^N B_i(t) - (N-1)B_1(t) \\ &\quad + \frac{(N-1)}{2}(L_2(t) - L_1(t)) + \frac{1}{2}L_2(t) \\ &= \bar{z}_2 + \sum_{i=2}^N B_i(t) - (N-1)B_1(t) + \frac{N}{2}L_2(t) - (N-1)L_1(t).\end{aligned}$$

Consider the martingale $M(t) \doteq \sum_{i=2}^N B_i(t) - (N-1)B_1(t)$. Then

$$\bar{Z}_2(t) = \bar{z}_2 + M(t) + \frac{N}{2}L_2(t) - (N-1)L_1(t). \quad (7.11)$$

This proves (7.10) for $k = 1$. Now, we will use this along with some local time inequalities to prove (7.10) for $k \geq 2$. Note that, from (7.4),

$$\begin{aligned}L_2(t) &\leq L_1(t) + B_2^*(t) + \frac{1}{2}L_3(t) \\ L_i &\leq B_i^*(t) + \frac{1}{2}(L_{i+1}(t) + L_{i-1}(t)), \quad i = 3, \dots, N.\end{aligned} \quad (7.12)$$

From these identities it follows that, for $k \in \{3, \dots, N\}$,

$$\begin{aligned}\sum_{i=3}^k (k-i+1) \left[L_i(t) - \frac{1}{2}(L_{i+1}(t) + L_{i-1}(t)) \right] &+ (k-1)(L_2(t) - L_1(t) - \frac{1}{2}L_3(t)) \\ &\leq \sum_{i=2}^k (k-i+1)B_i^*(t) \doteq Y_k^{(1)}(t).\end{aligned}$$

On the other hand,

$$\begin{aligned}\sum_{i=3}^k (k-i+1) \left(L_i(t) - \frac{1}{2}(L_{i+1}(t) + L_{i-1}(t)) \right) &+ (k-1)(L_2(t) - L_1(t) - \frac{1}{2}L_3(t)) \\ &= \frac{k}{2}L_2(t) - (k-1)L_1(t) - \frac{1}{2}L_{k+1}(t).\end{aligned}$$

Combining the last two displays and multiplying through by $\frac{2}{k}$,

$$L_2(t) \leq \frac{2(k-1)}{k}L_1(t) + \frac{1}{k}L_{k+1}(t) + \frac{2}{k}Y_k^{(1)}(t), \quad k = 3, \dots, N. \quad (7.13)$$

The last display holds trivially for $k = 1$ and also for $k = 2$, as can be seen from (7.12).

Hence, for all $1 \leq k \leq N$, using (7.11),

$$\begin{aligned}\bar{Z}_2(t) - \bar{z}_2 &= M(t) + \frac{N}{2}L_2(t) - (N-1)L_1(t) \\ &\leq M(t) + \frac{N}{2} \left(\frac{2}{k}Y_k^{(1)}(t) + \frac{2(k-1)}{k}L_1(t) + \frac{1}{k}L_{k+1}(t) \right) - (N-1)L_1(t) \\ &= M(t) + \frac{N}{k}Y_k^{(1)}(t) + \frac{N}{2k}L_{k+1}(t) - \frac{(N-k)}{k}L_1(t).\end{aligned}$$

This proves the lemma. \square

Define the sequence of stopping times $\{\sigma_m\}_{m \geq 0}$ as $\sigma_0 = 0$, and for $i \geq 0$,

$$\sigma_{2i+1} \doteq \inf\{t \geq \sigma_{2i} : V(t) = 4g\}, \quad \sigma_{2i+2} \doteq \inf\{t \geq \sigma_{2i+1} : V(t) = 2g\}. \quad (7.14)$$

For $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$, abusing notation, write $\sum_{i=2}^N (N-i+1)\hat{z}_i$ as \bar{z}_2 .

Proposition 7.11. *There exists $\Delta_0 > 0$ so that, for every $\Delta \geq \Delta_0$, there is a $l \in \mathbb{N}$ such that*

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1} : \bar{z}_2 \geq \Delta} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}(\bar{Z}_2(\sigma_{2l}) - \bar{z}_2) < 0.$$

This proposition will be proven using the following two lemmas. Proofs of the lemmas are given in Sect. 8.3.

Lemma 7.12. *There exists an $l_0 \in \mathbb{N}$ and $c_2 > 0$, such that for all $1 \leq k < N$, $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$, and $l \geq l_0$,*

$$\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^*(\bar{Z}_2(\sigma_{2l}) - \bar{z}_2) \leq -c_2 l + \frac{N}{2k} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l}). \quad (7.15)$$

To complete the proof of Proposition 7.11 we will estimate, in the next lemma, the second term in the bound (7.15).

Take $\Delta > 0$ and suppose $\hat{\mathbf{z}} \in \mathcal{S}_\Delta \doteq \{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1} : \bar{z}_2 \geq \Delta\}$. Then there is a $k \in \{1, \dots, N-1\}$ so that

$$z_{k+1} \geq \frac{\Delta}{N^2}. \quad (7.16)$$

We will work with this k in the following.

Lemma 7.13. *For $\Delta > 0$ and $\hat{\mathbf{z}} \in \mathcal{S}_\Delta$, let $k = k(\Delta)$ satisfy (7.16). There exist positive constants Δ_1, D_1, D_2, D_3 such that for any $\Delta \geq \Delta_1$ and $l \in \mathbb{N}$,*

$$\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l}) \leq D_1 l^{5/2} \left(\sqrt{l} e^{-D_2 \sqrt{\Delta}/l} + e^{-D_3 \Delta^{3/2}} \right).$$

Proof of Proposition 7.11. With l_0 as in Lemma 7.12 and Δ_1 as in Lemma 7.13, let $\Delta'_0 \doteq \max\{\Delta_1, l_0^4\}$. Setting $l = l(\Delta) = \lfloor \Delta^{1/4} \rfloor + 1$, we use Lemma 7.12 and Lemma 7.13 to obtain positive constants c'_2, D'_1, D'_2 such that for all $\Delta \geq \Delta'_0$ and $\hat{\mathbf{z}} \in \mathcal{S}_\Delta$,

$$\mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^*(\bar{Z}_2(\sigma_{2l}) - \bar{z}_2) \leq -c'_2 \Delta^{1/4} + \frac{N D'_1}{4k} \Delta^{3/4} e^{-D'_2 \Delta^{1/4}}.$$

The result now follows upon taking $\Delta_0 \geq \Delta'_0$ such that the above bound is negative for all $\Delta \geq \Delta_0$. \square

The next proposition shows that $|\bar{Z}_2(\sigma_2) - \bar{Z}_2(0)|$ has a finite exponential moment.

Proposition 7.14. *There exists $\gamma_6 > 0$ so that*

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* e^{\gamma_6 |\bar{Z}_2(\sigma_2) - \bar{z}_2|} < \infty.$$

Proof. From (7.11) and (7.2), under $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$,

$$\begin{aligned} |\bar{Z}_2(\sigma_2) - \bar{z}_2| &\leq |M(\sigma_2)| + \frac{N}{2}L_2(\sigma_2) + (N-1)L_1(\sigma_2) \\ &\leq |M(\sigma_2)| + |\bar{Y}(\sigma_2)| + 2(N-1)L_1(\sigma_2). \end{aligned}$$

Also note that

$$L_1(\sigma_2) = 2g + g\sigma_2 - V(\sigma_2) = 2g + g\sigma_2 - 2g = g\sigma_2.$$

Thus

$$|\bar{Z}_2(\sigma_2) - \bar{z}_2| \leq 2(N-1)g\sigma_2 + |M(\sigma_2)| + |\bar{Y}(\sigma_2)|.$$

Hence, writing $Y^\circ(t) := |M(t)| + |\bar{Y}(t)|$, for any $\gamma > 0$, using Cauchy Schwarz inequality,

$$\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{\gamma|\bar{Z}_2(\sigma_2) - \bar{z}_2|} \leq \left(\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{4(N-1)g\gamma\sigma_2} \right)^{1/2} \left(\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\gamma|Y^\circ(\sigma_2)|} \right)^{1/2}. \quad (7.17)$$

Recall γ_2 from Proposition 7.6, and write $D := \sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{\gamma_2\sigma_2} < \infty$. Proceeding as in the proof of Lemma 7.7 (see Sect. 8.2.1), observe using (1.4), Proposition 7.6 and Markov's inequality that there exist $c, c' > 0$ such that for any $\gamma \in (0, \gamma_2/2)$ and any $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$,

$$\begin{aligned} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\gamma|Y^\circ(\sigma_2)|} &\leq \sum_{k=0}^{\infty} \left(\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{4\gamma \sup_{0 \leq s \leq k+1} |Y^\circ(s)|} \right)^{1/2} (\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(\sigma_2 \geq k))^{1/2} \\ &\leq c\sqrt{D} \sum_{k=0}^{\infty} e^{c'\gamma^2(k+1) - \gamma k}. \end{aligned}$$

The proposition follows from the above bound, (7.17) and Proposition 7.6 upon choosing $\gamma \in (0, \min\{\gamma_2/(4(N-1)g), \gamma_2/2\})$ small enough so that the sum on the right side in the above display is finite. \square

7.6. Hitting time of a compact set. Recall the sequence of stopping times $\{\sigma_j\}_{j \in \mathbb{N}_0}$ introduced in (7.14) and the process \bar{Z}_2 defined in (7.9). Also fix $\Delta \geq \Delta_0$ where Δ_0 is as in Proposition 7.11. Define

$$\begin{aligned} \Gamma' &\doteq \inf\{\sigma_{2k} \geq 0 : k \in \mathbb{N}, \bar{Z}_2(\sigma_{2k}) \leq \Delta\}, \\ \Gamma &\doteq \inf\{t \geq 0 : Z_1(t) = 0, \bar{Z}_2(t) \leq \Delta, V(t) = 2g\}. \end{aligned} \quad (7.18)$$

Recall that for $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$, we write $\sum_{i=2}^N (N-i+1)\hat{z}_i$ as \bar{z}_2 .

Proposition 7.15. *There exist $\gamma_7 > 0$ and $c, c' > 0$ such that for any $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$ with $\bar{z}_2 \geq \Delta$ and any $t \geq c'\bar{z}_2$,*

$$\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(\Gamma > t) \leq ce^{-\gamma_7 t}.$$

Proof. From the definition of the stopping times $\{\sigma_j\}_{j \in \mathbb{N}_0}$ it follows that, for each $k \in \mathbb{N}$, σ_{2k} is a point of decrease of the velocity process and, consequently, $Z_1(\sigma_{2k}) = 0$. Indeed, if this is not the case, one can produce an open interval containing σ_{2l} where the velocity is strictly increasing, leading to a contradiction to the definition of σ_{2k} . Since $Z_1(\sigma_{2k}) = 0$, it follows that $\Gamma' \geq \Gamma$. It therefore suffices to show the result with Γ replaced by Γ' .

By Proposition 7.11, we can obtain $l_* \in \mathbb{N}$ and $\mu_* > 0$ such that for any $j \in \mathbb{N}$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \left\{ \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}(\bar{Z}_2(\sigma_{2jl_*}) - \bar{Z}_2(\sigma_{2(j-1)l_*}) \mid \mathcal{F}_{\sigma_{2(j-1)l_*}}^*) + \mu_* \right\} \mathbf{1}_{\{\Gamma' > \sigma_{2(j-1)l_*}\}} \leq 0, \quad (7.19)$$

where $\mathcal{F}_{\sigma_{2(j-1)l_*}}^*$ denotes the filtration associated with the process stopped at time $\sigma_{2(j-1)l_*}$.

Write $\mathcal{X}_j := \bar{Z}_2(\sigma_{2jl_*}) - \bar{Z}_2(\sigma_{2(j-1)l_*})$, $\tilde{\mathcal{F}}_{j-1} := \mathcal{F}_{\sigma_{2(j-1)l_*}}^*$, $j \in \mathbb{N}$. By Proposition 7.14 and the strong Markov property,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \sup_{j \in \mathbb{N}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* \left(e^{\gamma_6 |\mathcal{X}_j|} \mid \tilde{\mathcal{F}}_{j-1} \right) < \infty.$$

This in particular says that $\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \sup_{j \in \mathbb{N}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* \left(|\mathcal{X}_j| \mid \tilde{\mathcal{F}}_{j-1} \right) < \infty$. From these observations and Markov's inequality, we conclude that there exist positive constants c_1, c_2 such that for any $j \in \mathbb{N}$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* \left(|\mathcal{X}_j - \mathbb{E}(\mathcal{X}_j \mid \tilde{\mathcal{F}}_{j-1})| \geq x \mid \tilde{\mathcal{F}}_{j-1} \right) \leq c_1 e^{-c_2 x}, \quad x \geq 0.$$

Hence, by [38, Theorem 2.2] and its proof, there exist non-negative numbers (ν, b) such that for any $j \in \mathbb{N}$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* \left(e^{\lambda(\mathcal{X}_j - \mathbb{E}(\mathcal{X}_j \mid \tilde{\mathcal{F}}_{j-1}))} \mid \tilde{\mathcal{F}}_{j-1} \right) \leq e^{\nu^2 \lambda^2 / 2}, \quad \text{for all } |\lambda| < 1/b.$$

Therefore, by [38, Theorem 2.3], there exist positive constants c_3, c_4 such that for any $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$ with $\bar{z}_2 \geq \Delta$ and any $t > \left(\frac{2}{\mu_*} + \frac{2}{\Delta} \right) \bar{z}_2$,

$$\begin{aligned} & \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (\Gamma' > \sigma_{2l_* \lfloor t \rfloor}) \\ &= \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* \left(\sum_{j=1}^{\lfloor t \rfloor} \mathcal{X}_j > \Delta - \bar{z}_2, \Gamma' > \sigma_{2l_* \lfloor t \rfloor} \right) \\ &\leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* \left(\sum_{j=1}^{\lfloor t \rfloor} (\mathcal{X}_j - \mathbb{E}(\mathcal{X}_j \mid \tilde{\mathcal{F}}_{j-1})) > \Delta - \bar{z}_2 + \mu_* \lfloor t \rfloor, \Gamma' > \sigma_{2l_* \lfloor t \rfloor} \right) \\ &\leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* \left(\sum_{j=1}^{\lfloor t \rfloor} (\mathcal{X}_j - \mathbb{E}(\mathcal{X}_j \mid \tilde{\mathcal{F}}_{j-1})) > \Delta + \mu_* t / 4 \right) \\ &\leq c_3 e^{-c_4 t}. \end{aligned} \quad (7.20)$$

In the above display the first inequality is from (7.19) while the second inequality is from the facts that due to our condition on \bar{z}_2 and t we have that $\mu_*(t-1) > 2\bar{z}_2$ and $t > 2$, which says that

$$\mu_* \lfloor t \rfloor - \bar{z}_2 = \frac{1}{2} \mu_* \lfloor t \rfloor - \bar{z}_2 + \frac{1}{2} \mu_* \lfloor t \rfloor \geq \frac{1}{2} (\mu_* \lfloor t \rfloor - 2\bar{z}_2) + \frac{1}{4} \mu_* t \geq \frac{1}{4} \mu_* t.$$

Now, Proposition 7.6 and the strong Markov property imply that there exists $A \geq 1$ such that

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g, 0, \hat{\mathbf{z}})}^* (e^{\gamma_2 \sigma_{2l_*} \lfloor t \rfloor}) \leq A^{\lfloor t \rfloor}, \quad t > 0.$$

Hence, taking $a > 0$ such that $e^{\gamma_2 a} > A$, we obtain positive constants c'_3, c'_4 such that for any $t > 0$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g, 0, \hat{\mathbf{z}})}^* (\sigma_{2l_*} \lfloor t \rfloor > at) \leq c'_3 e^{-c'_4 t}. \quad (7.21)$$

Using (7.20) and (7.21), we conclude that there exist positive constants c_5, c_6 such that for any $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$ with $\bar{z}_2 \geq \Delta$ and any $t > \left(\frac{2}{\mu_*} + \frac{2}{\Delta}\right) \bar{z}_2$,

$$\mathbb{P}_{(2g, 0, \hat{\mathbf{z}})}^* (\Gamma' > at) \leq \mathbb{P}_{(2g, 0, \hat{\mathbf{z}})}^* (\Gamma' > \sigma_{2l_*} \lfloor t \rfloor) + \sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g, 0, \hat{\mathbf{z}})}^* (\sigma_{2l_*} \lfloor t \rfloor > at) \leq c_5 e^{-c_6 t}.$$

The result follows upon taking $c = c_5$, $\gamma_7 = c_6$ and $c' = a \left(\frac{2}{\mu_*} + \frac{2}{\Delta}\right)$. \square

7.7. Completing the proof of exponential ergodicity. In this section, we will complete the proof of Theorem 2.5. We begin with the following proposition the proof of which will be completed in Sect. 8.4. Fix $\Delta \geq \Delta_0$ where Δ_0 is as in Proposition 7.11. Define

$$C^* \doteq \{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N : v = 2g, z_1 = 0, \bar{z}_2 \leq \Delta\}. \quad (7.22)$$

Let $\tau_{C^*}(1) \doteq \inf\{t \geq 1 : (V(t), \mathbf{Z}(t)) \in C^*\}$.

Proposition 7.16.

1. *There exists $\eta > 0$ such that*

$$\tilde{V}_0(v, \mathbf{z}) \doteq \mathbb{E}_{(v, \mathbf{z})}^* e^{\eta \tau_{C^*}(1)} < \infty, \quad \text{for all } (v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N.$$

Furthermore,

$$\sup_{(v, \mathbf{z}) \in C^*} \tilde{V}_0(v, \mathbf{z}) \doteq M < \infty.$$

2. *There exists a non-zero measure ν on $\mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ and $r_1 \in (0, \infty)$ such that, for all $(v, \mathbf{z}) \in C^*$,*

$$\mathbb{P}^{r_1}((v, \mathbf{z}), A) \geq \nu(A) \text{ for all } A \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N).$$

Part (2) of the above proposition shows that, in the terminology of Down, Meyn and Tweedie (cf. [18, Section 3], the set C^* is ν -petite (or small) for the Markov family $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$. Together with part (1) of the proposition, this shows that the conditions of [18, Theorem 6.2] are satisfied and consequently, the function V_0 defined as

$$V_0(v, \mathbf{z}) \doteq 1 - \frac{1}{\eta} + \frac{1}{\eta} \tilde{V}_0(v, \mathbf{z}), \quad (v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N, \quad (7.23)$$

satisfies the drift condition (\mathcal{D}_T) in [18, Section 5]. We will now like to apply [18, Theorem 5.2] to conclude the proof of exponential ergodicity. For this we show in the next two results that the Markov process $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$ is irreducible and aperiodic.

Recall the set D from Theorem 4.1.

Proposition 7.17. *Define the measure ψ on $\mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ as $\psi(A) \doteq \lambda(A \cap D)$, $A \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$. Then the Markov process $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$ is ψ -irreducible.*

Proof. Fix $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$. Let $B \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ be such that $\lambda(B \cap D) > 0$. To establish ψ -irreducibility it suffices to show

$$\mathbb{E}_{(v, \mathbf{z})}^* \int_0^\infty \mathbf{1}_{\{(V(t), \mathbf{Z}(t)) \in B\}} dt > 0.$$

From Theorem 4.1, for each $t \in [\zeta, \zeta^*]$ and $(v', \mathbf{z}') \in R = (0, \frac{\zeta}{128}) \times (0, \infty) \times \mathbb{R}_+^{N-1}$,

$$\mathbb{P}^t((v', \mathbf{z}'), B) \geq K_{(v', \mathbf{z}')} \lambda(B \cap D).$$

Also, from Lemma 5.1, for any $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, there exists $r_0 \doteq r_0(v, \mathbf{z}) \in \mathbb{N}$ such that

$$\mathbb{P}^{r_0}((v, \mathbf{z}), R) > 0. \quad (7.24)$$

Observe that for $t \in [r_0 + \zeta, r_0 + \zeta^*]$,

$$\begin{aligned} \mathbb{P}^t((v, \mathbf{z}), B) &= \int_{\mathbb{R} \times \mathbb{R}_+^N} \mathbb{P}^{t-r_0}((v', \mathbf{z}'), B) d\mathbb{P}^{r_0}((v, \mathbf{z}), dv', d\mathbf{z}') \geq \lambda(B \cap D) \\ &\quad \int_R K_{(v', \mathbf{z}')} d\mathbb{P}^{r_0}((v, \mathbf{z}), dv', d\mathbf{z}'). \end{aligned}$$

The latter expression is strictly positive in view of (7.24), the positivity of $K_{(v, \mathbf{z})}$ for $(v, \mathbf{z}) \in R$ and our assumption concerning B . Finally note that

$$\mathbb{E}_{(v, \mathbf{z})}^* \int_0^\infty \mathbf{1}_{\{(V(t), \mathbf{Z}(t)) \in B\}} dt = \int_0^\infty \mathbb{P}^t((v, \mathbf{z}), B) dt \geq \int_{r_0 + \zeta}^{r_0 + \zeta^*} \mathbb{P}^t((v, \mathbf{z}), B) dt > 0.$$

The result follows. \square

Proposition 7.18. *The Markov process $\{\mathbb{P}_{(v, \mathbf{z})}\}_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N}$ is aperiodic.*

Proof. Recall the set C and the constant $\bar{K}_{\bar{A}}$ from Theorem 4.1 and let $\bar{K} \doteq \bar{K}_C$. Define the measure ν on $\mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$ as $\nu(B) \doteq \bar{K}\lambda(B \cap D)$, for $B \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$. From Theorem 4.1 it follows that the set C in the statement of the theorem is ν -small. Hence, for aperiodicity, it suffices to show that, for some $t_0 > 0$

$$\mathbb{P}^t((v, \mathbf{z}), C) > 0, \quad \text{for all } t \geq t_0, \text{ and } (v, \mathbf{z}) \in C. \quad (7.25)$$

Since $\lambda(C \cap D) > 0$, we have that (7.25) holds for $t \in [\zeta, \zeta^*]$ and all $(v, \mathbf{z}) \in C$. Let $\delta = \zeta^* - \zeta$. We now claim that, for all $m \in \mathbb{N}$,

$$\mathbb{P}^t((v, \mathbf{z}), C) > 0, \quad \text{for all } t \in [m\zeta, m\zeta + m\delta] \text{ and } (v, \mathbf{z}) \in C.$$

Indeed, clearly the result is true with $m = 1$, and if the result is true with $m = k$ then it is also true for $m = k + 1$ since any $t \in [(k + 1)\zeta, (k + 1)\zeta + (k + 1)\delta]$ can be written as $t_1 + t_2$ with $t_1 \in [k\zeta, k\zeta + k\delta]$ and $t_2 \in [\zeta, \zeta + \delta]$, and

$$\mathbb{P}^t((v, \mathbf{z}), C) \geq \int_C P^{t_1}((v, \mathbf{z}), (d\tilde{v}, d\tilde{\mathbf{z}})) P^{t_2}((\tilde{v}, \tilde{\mathbf{z}}), C) > 0 \text{ for all } (v, \mathbf{z}) \in C.$$

Now choose $k_0 \in \mathbb{N}$ such that $k_0\delta \geq \zeta$. Then $\mathbb{P}^t((v, \mathbf{z}), C) > 0$ for all $(v, \mathbf{z}) \in C$ and $t \in [k\zeta, (k + 1)\zeta]$ for all $k \geq k_0$. We conclude that $\mathbb{P}^t((v, \mathbf{z}), C) > 0$ for all $(v, \mathbf{z}) \in C$ and for all $t \geq k_0\zeta$. The result follows. \square

We can now complete the proof of exponential ergodicity.

Proof of Theorem 2.5 As noted previously, Proposition 7.16 shows that the conditions of [18, Theorem 6.2] are satisfied and consequently, the function V_0 defined in (7.23) satisfies the drift condition (\mathcal{D}_T) in [18, Section 5]. Also from Propositions 7.17 and 7.18 the Markov process is ψ -irreducible and aperiodic. The result is now immediate from [18, Theorem 5.2].

8. Proofs of Some Results from Sect. 7

In this section we present proofs of some technical results stated without proof in Sect. 7.

8.1. Proofs of lemmas for Proposition 7.2. In this section we provide the proofs of Lemmas 7.3, 7.4, and 7.5 stated in Sect. 7.3 that were used in the proof of Proposition 7.2.

8.1.1. Proof of Lemma 7.3 Fix $\mathbf{z} \in \mathbb{R}_+^N$. All inequalities in the proof will be a.s. under $\mathbb{P}_{(0, \mathbf{z})}^*$. Using (7.1), we have that for $t \leq \hat{\tau}_{g/(2N)}$,

$$L_1(t) \leq \sum_{i=1}^N W_{1,i} B_i^*(t) + \frac{gW_{1,1}t}{2N}.$$

It can be verified that

$$W_{1,1} = N, \text{ and } W_{i,1} = 2N - 2(i - 1), \quad i = 2, \dots, N.$$

Using this and since $V(t) = gt - L_1(t)$, it follows that

$$V(t) \geq - \sum_{i=1}^N W_{1,i} B_i^*(t) + g(1 - \frac{W_{1,1}}{2N})t = - \sum_{i=1}^N W_{1,i} B_i^*(t) + \frac{gt}{2} \doteq Q(t).$$

Define $\hat{\sigma}_{g/(2N)} \doteq \inf\{t \geq 0 : Q(t) = g/(2N)\}$. Then the above inequality implies that $\hat{\sigma}_{g/(2N)} \geq \hat{\tau}_{g/(2N)}$. By a standard concentration bound (see (1.4)) it follows that there are $\varrho_1, \varrho_2 \in (0, \infty)$ such that

$$\mathbb{E}_{(0,\mathbf{z})}^* e^{\theta \sum_{i=1}^N W_{1,i} B_i^*(s)} \leq \varrho_1 e^{\varrho_2 \theta^2 s} \text{ for all } s \geq 0 \text{ and } \theta \in (0, \infty).$$

Then, for an arbitrary $\theta, \beta > 0$, we have

$$\begin{aligned} \mathbb{E}_{(0,\mathbf{z})}^* e^{\beta \hat{\tau}_{g/(2N)}} &= \int_0^\infty \mathbb{P}_{(0,\mathbf{z})}^*(\hat{\tau}_{g/(2N)} > \frac{\ln(s)}{\beta}) ds \\ &\leq \int_0^\infty \mathbb{P}_{(0,\mathbf{z})}^*(\hat{\sigma}_{g/(2N)} > \frac{\ln(s)}{\beta}) ds \\ &\leq \int_0^\infty \mathbb{P}_{(0,\mathbf{z})}^*(Q(\frac{\ln(s)}{\beta}) < \frac{g}{2N}) ds \\ &\leq 1 + \int_1^\infty \mathbb{P}_{(0,\mathbf{z})}^*(\frac{g \ln(s)}{2\beta} < \frac{g}{2N} + \sum_{i=1}^N W_{1,i} B_i^*(\frac{\ln(s)}{\beta})) ds \\ &\leq 1 + e^{\theta g/(2N)} \int_1^\infty e^{-\theta g \ln(s)/2\beta} \mathbb{E}_{(0,\mathbf{z})}^* e^{\theta \sum_{i=1}^N W_{1,i} B_i^*(\frac{\ln(s)}{\beta})} ds \\ &\leq 1 + \varrho_1 e^{\theta g/(2N)} \int_1^\infty s^{-\theta g/2\beta} s^{\theta^2 \varrho_2/\beta} ds. \end{aligned}$$

Now take

$$\theta \doteq \frac{g}{4\varrho_2}, \quad \beta \doteq \frac{\theta g}{8}.$$

Then

$$-\theta g/2\beta + \theta^2 \varrho_2/\beta = -2.$$

The result follows. \square

8.1.2. Proof of Lemma 7.4 We will first show that

$$\inf_{(v,\mathbf{z}) \in [\frac{g}{4N}, 4g] \times [1, \infty) \times \mathbb{R}_+^{N-1}} \mathbb{P}_{(v,\mathbf{z})}^*(\hat{\tau}_{4g} < \hat{\tau}_0) \doteq p_1 > 0. \quad (8.1)$$

Note that, for $t > 0$, on the set $\{\hat{\tau}_{4g} > t\}$, for $(v, \mathbf{z}) \in [\frac{g}{4N}, 4g] \times \mathbb{R}_+^N$, under $\mathbb{P}_{(v,\mathbf{z})}^*$,

$$\begin{aligned} L_1(t) &\leq \sup_{0 \leq s \leq t} (-z_1 - B_1(s) + \frac{1}{2}L_2(s) + 4gs)^+ \leq \sup_{0 \leq s \leq t} (-z_1 - B_1(s) + 4gs)^+ + \frac{1}{2}L_2(t) \\ &\leq \sup_{0 \leq s \leq t} (-z_1 - B_1(s) + 4gs)^+ + \frac{(N-1)}{N}L_1(t) + \frac{1}{N}\bar{Y}(t), \end{aligned}$$

where the last inequality uses (7.2). Thus

$$L_1(t) \leq N \sup_{0 \leq s \leq t} (-z_1 - B_1(s) + 4gs)^+ + \bar{Y}(t). \quad (8.2)$$

Consider the set $A_1 \in \mathcal{F}^*$ defined as

$$A_1 \doteq \{-B_1(s) + 4gs - 1 < 0 \text{ for all } s \in [0, 8] \text{ and } \bar{Y}(8) < g/8N\}.$$

Note that

$$\inf_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_1) \doteq p'_1 > 0.$$

Also, for $(v, \mathbf{z}) \in [\frac{g}{4N}, 4g] \times [1, \infty) \times \mathbb{R}_+^{N-1}$, under $\mathbb{P}_{(v, \mathbf{z})}^*$, on A_1 ,

$$L_1(8 \wedge \hat{\tau}_{4g}) \leq N \sup_{0 \leq s \leq 8 \wedge \hat{\tau}_{4g}} (-z_1 - B_1(s) + 4gs)^+ + \bar{Y}(8) = \bar{Y}(8) < g/8N < 4g.$$

So, in particular,

$$V(8 \wedge \hat{\tau}_{4g}) = V(\hat{\tau}_{4g})1_{\{\hat{\tau}_{4g} \leq 8\}} + V(8)1_{\{\hat{\tau}_{4g} > 8\}} \geq 4g1_{\{\hat{\tau}_{4g} \leq 8\}} + (8g - 4g)1_{\{\hat{\tau}_{4g} > 8\}} = 4g$$

and consequently $\hat{\tau}_{4g} \leq 8$. Also, under the same conditions, for $s < 8$,

$$V(s \wedge \hat{\tau}_{4g}) \geq v - L_1(s \wedge \hat{\tau}_{4g}) \geq v - L_1(8 \wedge \hat{\tau}_{4g}) > \frac{g}{4N} - \frac{g}{8N} > 0.$$

Thus we have

$$p_1 = \inf_{(v, \mathbf{z}) \in [\frac{g}{4N}, 4g] \times [1, \infty) \times \mathbb{R}_+^{N-1}} \mathbb{P}_{(v, \mathbf{z})}^*(\hat{\tau}_{4g} < \hat{\tau}_0) \geq \inf_{(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_1) = p'_1 > 0.$$

This proves (8.1).

Let $v_1 \doteq \inf\{t \geq 0 : Z_1(t) \geq 1\}$. In order to complete the proof, from the strong Markov property, it suffices to show that

$$\inf_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(v_1 \wedge \hat{\tau}_{4g} < \hat{\tau}_{g/4N}) \doteq p_2 > 0. \quad (8.3)$$

Fix $\delta \in (0, 1)$ such that

$$2gN\delta + \frac{1}{2}gN\delta^2 \leq \frac{g}{16N}.$$

Define $A_2 \in \mathcal{F}^*$ as

$$A_2 \doteq \{B_1(\delta) \geq 1 + 4gN\delta + gN\delta^2 + \frac{3g}{16N}, \bar{Y}(\delta) + NB_1^*(\delta) \leq \frac{g}{16N}\}.$$

It is easy to check that

$$\inf_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_2) \doteq p'_2 > 0.$$

The Inert Drift Atlas Model

Furthermore, as in (8.2), for $(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N$, under $\mathbb{P}_{(v, \mathbf{z})}^*$, on A_2 ,

$$L_1(\delta) \leq NB_1^*(\delta) + \bar{Y}(\delta) + N \int_0^\delta V^+(s)ds \leq NB_1^*(\delta) + \bar{Y}(\delta) + 2gN\delta + \frac{1}{2}gN\delta^2.$$

Also, under the same conditions, from (7.2),

$$L_2(\delta) \leq 2L_1(\delta) + \frac{2}{N}\bar{Y}(\delta) \leq 2NB_1^*(\delta) + 2\bar{Y}(\delta) + 4gN\delta + gN\delta^2 + \frac{2}{N}\bar{Y}(\delta).$$

Thus

$$\begin{aligned} Z_1(\delta) &= z_1 + B_1(\delta) + L_1(\delta) - \frac{1}{2}L_2(\delta) - \int_0^\delta V(s)ds \\ &\geq 1 + 4gN\delta + gN\delta^2 + \frac{3g}{16N} - NB_1^*(\delta) - \bar{Y}(\delta) - 2gN\delta - \frac{1}{2}gN\delta^2 \\ &\quad - \frac{1}{N}\bar{Y}(\delta) - 2g\delta - \frac{1}{2}g\delta^2 \\ &\geq 1. \end{aligned}$$

Again, under the same conditions, for $0 \leq s \leq \delta$,

$$\begin{aligned} V(s) &\geq \frac{g}{2N} - L_1(\delta) \geq \frac{g}{2N} - NB_1^*(\delta) - \bar{Y}(\delta) - 2gN\delta - \frac{1}{2}gN\delta^2 \\ &\geq \frac{g}{2N} - \frac{g}{16N} - 2gN\delta - \frac{1}{2}gN\delta^2 \geq \frac{g}{2N} - \frac{g}{16N} - \frac{g}{16N} > \frac{g}{4N}. \end{aligned}$$

It then follows

$$\begin{aligned} p_2 &= \inf_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(v_1 \wedge \hat{\tau}_{4g} < \hat{\tau}_{g/4N}) \\ &\geq \inf_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(v_1 < \hat{\tau}_{g/4N}) \\ &\geq \inf_{(v, \mathbf{z}) \in [\frac{g}{2N}, 2g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_2) = p_2' > 0. \end{aligned}$$

This proves (8.3) and completes the proof of the lemma. \square

8.1.3. Proof of Lemma 7.5 By the strong Markov property, it suffices to show that for some $m \in \mathbb{N}$

$$\inf_{(v, \mathbf{z}) \in [0, 4g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(\hat{\tau}_{4g} \wedge \hat{\tau}_0 \leq m) > 0. \quad (8.4)$$

We will prove (8.4) with $m = 5$. We consider two cases:

Case 1: $z_1 \geq 1$. Define $A_1 \in \mathcal{F}^*$ as

$$A_1 \doteq \{-B_1(s) + 4gs - 1 \leq 0 \text{ for all } 0 \leq s \leq 5, \bar{Y}(5) < g\}.$$

It is easily seen that

$$\inf_{(v, \mathbf{z}) \in [0, 4g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_1) \doteq \kappa_1 > 0.$$

From (7.2) and (8.2) it follows that, for $(v, \mathbf{z}) \in [0, 4g] \times [1, \infty) \times \mathbb{R}_+^{N-1}$, under $\mathbb{P}_{(v, \mathbf{z})}^*$, on $A_1 \cap \{\hat{\tau}_{4g} \wedge \hat{\tau}_0 > 5\}$,

$$L_1(5) \leq N \sup_{s \leq 5} (-1 + 4gs - B_1(s))^+ + \bar{Y}(5) < g,$$

and consequently

$$V(5) \geq 5g - L_1(5) > 5g - g = 4g.$$

This says that $A_1 \cap \{\hat{\tau}_{4g} \wedge \hat{\tau}_0 > 5\}$ is $\mathbb{P}_{(v, \mathbf{z})}^*$ trivial and so

$$\inf_{(v, \mathbf{z}) \in [0, 4g] \times [1, \infty) \times \mathbb{R}_+^{N-1}} \mathbb{P}_{(v, \mathbf{z})}^*(\hat{\tau}_{4g} \wedge \hat{\tau}_0 \leq 5) \geq \inf_{(v, \mathbf{z}) \in [0, 4g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_1) = \kappa_1 > 0.$$

This proves (8.4) when $z_1 \geq 1$.

Case 2: $z_1 < 1$. Define $A_2 \in \mathcal{F}^*$ as

$$A_2 \doteq \{B_1(5) < -1 - 9g\}.$$

Clearly

$$\inf_{(v, \mathbf{z}) \in [0, 4g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_2) \doteq \kappa_2 > 0.$$

Also, for $(v, \mathbf{z}) \in [0, 4g] \times [0, 1) \times \mathbb{R}_+^{N-1}$, under $\mathbb{P}_{(v, \mathbf{z})}^*$, on $A_2 \cap \{\hat{\tau}_{4g} \wedge \hat{\tau}_0 > 5\}$,

$$L_1(5) = \sup_{0 \leq s \leq 5} (-z_1 + \frac{1}{2}L_2(s) + \int_0^s V(u)du - B_1(s))^+ \geq \sup_{0 \leq s \leq 5} (-1 - B_1(s))^+ > 9g$$

and consequently

$$V(5) \leq 4g + 5g - L_1(5) < 0.$$

This shows that $A_2 \cap \{\hat{\tau}_{4g} \wedge \hat{\tau}_0 > 5\}$ is $\mathbb{P}_{(v, \mathbf{z})}^*$ trivial and so

$$\inf_{(v, \mathbf{z}) \in [0, 4g] \times [0, 1) \times \mathbb{R}_+^{N-1}} \mathbb{P}_{(v, \mathbf{z})}^*(\hat{\tau}_{4g} \wedge \hat{\tau}_0 \leq 5) \geq \inf_{(v, \mathbf{z}) \in [0, 4g] \times \mathbb{R}_+^N} \mathbb{P}_{(v, \mathbf{z})}^*(A_2) = \kappa_2 > 0.$$

This completes the proof of (8.4) when $z_1 < 1$. The result follows. \square

8.2. Proofs of lemmas for Proposition 7.6. In this section we provide the proofs of Lemmas 7.7, 7.8, and 7.9 stated in Sect. 7.4 that were used in the proof of Proposition 7.6.

8.2.1. *Proof of Lemma 7.7* Fix $\hat{\mathbf{z}} \in \mathbb{R}^{N-1}$. All inequalities in this proof are $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$ -a.s. Observe that $4g = V(\sigma_1) = g\sigma_1 - L_1(\sigma_1) + 2g$, so that $L_1(\sigma_1) = g\sigma_1 - 2g$. Then

$$\begin{aligned} Z_1(\sigma_1) &= B_1(\sigma_1) - \frac{1}{2}L_2(\sigma_1) + L_1(\sigma_1) - \int_0^{\sigma_1} V(s)ds \\ &\leq \sup_{0 \leq s \leq \sigma_1} (B_1(s)) + g\sigma_1 - \frac{g\sigma_1^2}{2} - 2g\sigma_1 + \int_0^{\sigma_1} L_1(s)ds \\ &\leq \sup_{0 \leq s \leq \sigma_1} (B_1(s)) + L_1(\sigma_1)\sigma_1 \\ &\leq \sup_{0 \leq s \leq \sigma_1} (B_1(s)) + g\sigma_1^2. \end{aligned}$$

Thus, for $\beta > 0$,

$$\begin{aligned} e^{\beta(Z_1(\sigma_1))^{1/2}} &\leq e^{\beta(\sup_{0 \leq s \leq \sigma_1} (B_1(s)) + g\sigma_1^2)^{1/2}} \\ &\leq e^{\beta(\sup_{0 \leq s \leq \sigma_1} B_1(s))^{1/2} + \beta\sqrt{g}\sigma_1} \leq \frac{1}{2}e^{2\beta((\sup_{0 \leq s \leq \sigma_1} B_1(s))^{1/2})} + \frac{1}{2}e^{2\beta\sqrt{g}\sigma_1}, \end{aligned} \quad (8.5)$$

where in the final step we use Young's inequality. We now estimate each of the terms in (8.5). We begin by recalling that from Proposition 7.2, we can find $\beta_0 \in (0, 1/2)$ such that

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{\beta_0 \sigma_1} \doteq c(\beta_0) < \infty.$$

Hence, taking $\beta \in (0, \beta_0/(2\sqrt{g})]$, the second term in (8.5) is bounded as

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta\sqrt{g}\sigma_1} \leq c(\beta_0).$$

With $\beta \in (0, \beta_0]$ for the first term in (8.5), we have,

$$\begin{aligned} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta(\sup_{0 \leq s \leq \sigma_1} B_1(s))^{1/2}} &\leq e^{2\beta} + \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta \sup_{0 \leq s \leq \sigma_1} B_1(s)} \mathbf{1}_{\{\sup_{0 \leq s \leq \sigma_1} B_1(s) > 1\}} \\ &\leq e^{2\beta} + \sum_{k=0}^{\infty} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta \sup_{0 \leq s \leq \sigma_1} B_1(s)} \mathbf{1}_{\{k \leq \sigma_1 < k+1\}} \\ &\leq e^{2\beta} + \sum_{k=0}^{\infty} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta \sup_{0 \leq s \leq k+1} B_1(s)} \mathbf{1}_{\{k \leq \sigma_1 < k+1\}}. \end{aligned}$$

Using Cauchy-Schwarz inequality,

$$\begin{aligned}
& \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta(\sup_{0 \leq s \leq \sigma_1} B_1(s))^{1/2}} \\
& \leq e^{2\beta} + \sum_{k=0}^{\infty} (\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{4\beta \sup_{0 \leq s \leq k+1} B_1(s)})^{1/2} (\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(\sigma_1 \geq k))^{1/2} \\
& \leq e^{2\beta} + \varrho_1 \sum_{k=0}^{\infty} e^{8\beta^2 \varrho_2(k+1)} (\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(\sigma_1 \geq k))^{1/2} \\
& \leq e^{2\beta} + \varrho_1 c(\beta_0)^{1/2} e^{8\beta^2 \varrho_2} \sum_{k=0}^{\infty} e^{8\beta^2 \varrho_2 k - \frac{\beta k}{2}} \doteq c_1(\beta) < \infty, \tag{8.6}
\end{aligned}$$

where the finiteness follows on choosing $\beta \in (0, \beta_1]$ for sufficiently small $\beta_1 \in (0, \beta_0]$. The second line above follows from a standard concentration inequality (see (1.4)) and the last line from Markov's inequality. Thus for any $\beta \in (0, \beta_1]$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{2\beta(\sup_{0 \leq s \leq \sigma_1} B_1(s))^{1/2}} \doteq c_1(\beta) < \infty.$$

The result now follows on setting $\gamma_3 = \min\{\beta_0/(2\sqrt{g}), \beta_1\}$. \square

8.2.2. Proof of Lemma 7.8 Let $(z_1, \hat{\mathbf{z}}) \in (0, \infty) \times \mathbb{R}_+^{N-1}$. All inequalities of random quantities in this proof are $\mathbb{P}_{(4g,z_1,\hat{\mathbf{z}})}^*$ -almost sure. For $t \leq \tau_0^{Z_1}$, we have

$$\begin{aligned}
Z_1(t) &= z_1 + B_1(t) - \frac{1}{2}L_2(t) - \int_0^t V(s)ds \\
&= z_1 + B_1(t) - \frac{1}{2}L_2(t) - \int_0^t (gs + 4g)ds \\
&\leq z_1 + \sup_{0 \leq s \leq t} B_1(s) - \frac{gt^2}{2} \doteq H(t).
\end{aligned}$$

Consequently, $Z_1(t)$ must hit zero before $H(t)$, and so $\tau_0^H \doteq \inf\{t \geq 0 : H(t) = 0\} \geq \tau_0^{Z_1}$. Thus, for arbitrary $\gamma > 0$,

$$\begin{aligned}
\mathbb{E}_{(4g,z_1,\hat{\mathbf{z}})}^* e^{\gamma \tau_0^{Z_1}} &= 1 + \int_1^{\infty} \mathbb{P}_{(4g,z_1,\hat{\mathbf{z}})}^*(\tau_0^{Z_1} > \frac{\ln(s)}{\gamma})ds \\
&\leq 1 + \int_1^{\infty} \mathbb{P}_{(4g,z_1,\hat{\mathbf{z}})}^*(\tau_0^H > \frac{\ln(s)}{\gamma})ds \\
&\leq 1 + \int_1^{\infty} \mathbb{P}_{(4g,z_1,\hat{\mathbf{z}})}^*(H(\frac{\ln(s)}{\gamma}) > 0)ds.
\end{aligned}$$

Thus, using Markov's inequality, for $\theta > 0$,

$$\begin{aligned}
 \mathbb{E}_{(4g, z_1, \hat{\mathbf{z}})}^* e^{\gamma \tau_0^{z_1}} &\leq 1 + \int_1^\infty \mathbb{P}_{(4g, z_1, \hat{\mathbf{z}})}^* (z_1 + \sup_{0 \leq u \leq \frac{\ln(s)}{\gamma}} B_1(s) > \frac{g(\frac{\ln(s)}{\gamma})^2}{2}) ds \\
 &\leq 1 + \int_1^\infty \mathbb{P}_{(4g, z_1, \hat{\mathbf{z}})}^* (z_1^{1/2} + (\sup_{0 \leq u \leq \frac{\ln(s)}{\gamma}} B_1(s))^{1/2} > \sqrt{\frac{g}{2}} \frac{\ln(s)}{\gamma}) ds \\
 &\leq 1 + e^{\theta z_1^{1/2}} \int_1^\infty s^{-\sqrt{\frac{g}{2}} \frac{\theta}{\gamma}} \mathbb{E}_{(4g, z_1, \hat{\mathbf{z}})}^* e^{\theta (\sup_{0 \leq u \leq \frac{\ln(s)}{\gamma}} B_1(s))^{1/2}} ds \\
 &\leq 1 + \varrho_1 e^{\theta z_1^{1/2}} \int_1^\infty s^{-\sqrt{\frac{g}{2}} \frac{\theta}{\gamma}} s^{\frac{\varrho_2 \theta^2}{\gamma}} ds,
 \end{aligned}$$

where in the last line we have used a standard concentration inequality (see (1.4)). Now take $\gamma_4 \doteq g/(16\varrho_2)$ and for fixed $\gamma \in (0, \gamma_4]$, take $\theta = 4\sqrt{2}\gamma/\sqrt{g}$. Then it follows that

$$-\sqrt{\frac{g}{2}} \frac{\theta}{\gamma} + \frac{\varrho_2 \theta^2}{\gamma} \leq -2.$$

Thus

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(4g, z_1, \hat{\mathbf{z}})}^* e^{\gamma \tau_0^{z_1}} \leq 1 + \varrho_1 e^{4\sqrt{2}\gamma z_1^{1/2}/\sqrt{g}}.$$

The result follows. \square

8.2.3. Proof of Lemma 7.9 Let $v \in [2g, \infty)$, $\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}$, and $\gamma > 0$. All inequalities of random quantities in this proof are $\mathbb{P}_{(v, 0, \hat{\mathbf{z}})}^*$ -almost sure. For $t \leq \hat{\tau}_{2g}$, $V(t) \geq 2gt$, so

$$0 \leq Z_1(t) = B_1(t) + L_1(t) - \frac{1}{2} L_2(t) - \int_0^t V(s) ds \leq \sup_{0 \leq s \leq t} B_1(s) + L_1(t) - 2gt,$$

from which it follows that $-L_1(t) \leq \sup_{0 \leq s \leq t} B_1(s) - 2gt$. Hence,

$$V(t) = gt - L_1(t) + v \leq \sup_{0 \leq s \leq t} B_1(s) - gt + v \doteq Q(t).$$

From this inequality we see that $\tau_{2g}^Q \doteq \inf\{t \geq 0 : Q(t) = 2g\}$ satisfies $\tau_{2g}^Q \geq \hat{\tau}_{2g}$. Then, for any $\theta > 0$,

$$\begin{aligned}
 \mathbb{E}_{(v, 0, \hat{\mathbf{z}})}^* e^{\gamma \hat{\tau}_{2g}} &= \int_0^\infty \mathbb{P}_{(v, 0, \hat{\mathbf{z}})}^* (\hat{\tau}_{2g} \geq \frac{1}{\gamma} \ln(s)) ds \leq \int_0^\infty \mathbb{P}_{(v, 0, \hat{\mathbf{z}})}^* (\tau_{2g}^Q \geq \frac{1}{\gamma} \ln(s)) ds \\
 &\leq 1 + \int_1^\infty \mathbb{P}_{(v, 0, \hat{\mathbf{z}})}^* (Q(\frac{1}{\gamma} \ln(s)) > 2g) ds.
 \end{aligned}$$

Thus by Markov's inequality,

$$\begin{aligned}\mathbb{E}_{(v,0,\hat{\mathbf{z}})}^* e^{\gamma \hat{\tau}_{2g}} &\leq 1 + e^{-2g\theta} \int_1^\infty \mathbb{E}_{(v,0,\hat{\mathbf{z}})}^* e^{\theta Q(\frac{1}{\gamma} \ln(s))} ds \\ &= 1 + e^{\theta(v-2g)} \int_1^\infty e^{-\frac{g\theta}{\gamma} \ln(s)} \mathbb{E}_{(v,0,\hat{\mathbf{z}})}^* e^{\theta \sup_{0 \leq t \leq \frac{1}{\gamma} \ln(s)} B_1(t)} ds \\ &\leq 1 + \varrho_1 e^{\theta(v-2g)} \int_1^\infty s^{-\frac{g\theta}{\gamma} + \frac{\varrho_2 \theta^2}{\gamma}} ds,\end{aligned}$$

where we have once again used (1.4). Now let $\gamma_5 \doteq g^2/(8\varrho_2)$ and for fixed $\gamma \in (0, \gamma_5)$, take $\theta = 4\gamma/g$. Then, for any $\gamma \in (0, \gamma_5)$,

$$-\frac{g\theta}{\gamma} + \frac{\varrho_2 \theta^2}{\gamma} \leq -2.$$

It then follows, for $\gamma \in (0, \gamma_5)$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(v,0,\hat{\mathbf{z}})}^* e^{\gamma \hat{\tau}_{2g}} \leq 1 + \varrho_1 e^{4\gamma(v-2g)/g}.$$

The result follows. \square

8.3. Proofs of lemmas for Proposition 7.11. In this section we provide the proofs of Lemmas 7.12 and 7.13 stated in Sect. 7.5 that were used in the proof of Proposition 7.11.

8.3.1. Proof of Lemma 7.12 Fix $(2g, 0, \hat{\mathbf{z}}) \in \mathbb{R} \times \mathbb{R}^N$. Since $M(t) = \sum_{i=2}^N B_i(t) - (N-1)B_1(t)$, from Proposition 7.6 (which implies $\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* \sigma_{2l} < \infty$ for any $l \in \mathbb{N}$) and optional sampling theorem (cf. [28, Section 1.3.C]), we have from Lemma 7.10, for $l \in \mathbb{N}$,

$$\begin{aligned}\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* (\bar{Z}_2(\sigma_{2l}) - \bar{z}_2) &\leq \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* \left(M(\sigma_{2l}) + \frac{N}{k} Y_k^{(1)}(\sigma_{2l}) - \frac{(N-k)}{k} L_1(\sigma_{2l}) + \frac{N}{2k} L_{k+1}(\sigma_{2l}) \right) \\ &= \frac{N}{k} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* Y_k^{(1)}(\sigma_{2l}) - \frac{(N-k)}{k} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_1(\sigma_{2l}) \\ &\quad + \frac{N}{2k} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l}).\end{aligned}\tag{8.7}$$

Using standard martingale maximal inequalities we have

$$\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* B_i^*(\sigma_{2l}) \leq c_0 \sqrt{\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* \sigma_{2l}} = c_0 \left(\sum_{i=1}^l \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* (\sigma_{2i} - \sigma_{2(i-1)}) \right)^{1/2} \leq c'_0 \sqrt{l},$$

where $c_0, c'_0 \in (0, \infty)$ are independent of $\hat{\mathbf{z}}$ and l , and the last inequality once more uses Proposition 7.6. Thus, for some $c_1 \in (0, \infty)$, for all $k = 1, \dots, N, l \in \mathbb{N}$,

$$\sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* Y_k^{(1)}(\sigma_{2l}) \leq c_1 l^{1/2}.\tag{8.8}$$

Next note that

$$\begin{aligned}
\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_1(\sigma_{2l}) &= \sum_{i=1}^l \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* (L_1(\sigma_{2i}) - L_1(\sigma_{2i-2})) \geq l \inf_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* (L_1(\sigma_2)) \\
&\geq gl \inf_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (L_1(\sigma_2) > g) \\
&= gl \inf_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (g\sigma_2 - V(\sigma_2) + 2g > g) \\
&= gl \inf_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (\sigma_2 > 1) = gl,
\end{aligned} \tag{8.9}$$

where the last equality follows on observing that, under $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$, $\sigma_2 > \sigma_1 > 1$ a.s.

The result follows from (8.7), (8.8) and (8.9). \square

8.3.2. Proof of Lemma 7.13 Fix $\Delta > 0$ and $(2g, 0, \hat{\mathbf{z}}) \in \mathbb{R} \times \mathbb{R}_+^N$ such that $\hat{\mathbf{z}} \in \mathcal{S}_\Delta$. All inequalities will be a.s. under $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$. Let $k = k(\Delta)$ satisfy (7.16). Define

$$\theta_k = \inf\{t \geq 0 : Z_{k+1}(t) = 0\}.$$

Then for $t \leq \theta_k$,

$$\begin{aligned}
Z_{k+1}(t) &= z_{k+1} + B_{k+1}(t) - B_k(t) - \frac{1}{2}(L_k(t) + L_{k+2}(t)) \\
&\geq \frac{\Delta}{N^2} + B_{k+1}(t) - B_k(t) - \frac{1}{2}(L_k(t) + L_{k+2}(t)).
\end{aligned} \tag{8.10}$$

To bound $\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l})$, we will obtain an upper bound on the probability that $L_{k+1}(\sigma_{2l}) > 0$, or equivalently, the probability that $Z_{k+1}(\cdot)$ hits zero before time σ_{2l} , using (8.10). Next, we will estimate $\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* (L_{k+1}(\sigma_{2l})^2)$. These two will be combined using a Cauchy-Schwarz inequality to obtain an upper bound for $\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l})$.

We will first obtain an upper bound for $L_k(t)$ for $k < N$ and $t \leq \theta_k$. When $3 \leq k \leq N-1$, from (7.4), for $t \leq \theta_k$,

$$\begin{aligned}
L_k(t) &\leq B_k^*(t) + \frac{1}{2}L_{k-1}(t) \\
L_i(t) &\leq B_i^*(t) + \frac{1}{2}(L_{i-1}(t) + L_{i+1}(t)), \quad 3 \leq i \leq k-1 \text{ if } k \geq 4, \\
L_2(t) &\leq L_1(t) + B_2^*(t) + \frac{1}{2}L_3(t).
\end{aligned} \tag{8.11}$$

Thus,

$$\begin{aligned}
&\sum_{i=3}^{k-1} (i-1)(L_i(t) - \frac{1}{2}(L_{i+1}(t) + L_{i-1}(t))) \\
&\quad + (L_2(t) - L_1(t) - \frac{1}{2}L_3(t)) + (k-1)(L_k(t) - \frac{1}{2}L_{k-1}(t)) \\
&\leq \sum_{i=2}^k (i-1)B_i^*(t) \doteq Y_k^{(2)}(t),
\end{aligned}$$

where the first sum is taken to be zero if $k = 3$. The left side in the above inequality equals $\frac{k}{2}L_k(t) - L_1(t)$ and so we have, whenever $N > k \geq 3$, $t \leq \theta_k$,

$$L_k(t) \leq \frac{2}{k}L_1(t) + \frac{2}{k}Y_k^{(2)}(t). \quad (8.12)$$

Note that the above inequality holds trivially if $k = 1$, and by (8.11) if $k = 2$, and so in fact the above holds under $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$, with $\hat{\mathbf{z}} \in \mathcal{S}_\Delta$, for k satisfying (7.16) and $t \leq \theta_k$.

We now obtain a similar upper bound on $L_{k+2}(t)$ when $k < N$ and $t \leq \theta_k$. From (7.4), when $k < N - 1$, for $t \leq \theta_k$,

$$\begin{aligned} & (N - k - 1)(L_{k+2}(t) - \frac{1}{2}L_{k+3}(t)) + \sum_{i=k+2}^{N-1} (N - i)(L_{i+1}(t) - \frac{1}{2}(L_i(t) + L_{i+2}(t))) \\ & \leq \sum_{i=k+1}^{N-1} (N - i)B_{i+1}^*(t) \doteq Y_k^{(3)}(t). \end{aligned}$$

The left side equals $\frac{N-k}{2}L_{k+2}(t)$ and so we have, when $k < N - 1$,

$$L_{k+2}(t) \leq \frac{2}{N-k}Y_k^{(3)}(t), \quad \text{for all } t \leq \theta_k. \quad (8.13)$$

Note that when $k = N - 1$ the inequality is trivially true. Using (8.12) and (8.13) in (8.10), we have for $t \leq \theta_k$, under $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$

$$\begin{aligned} Z_{k+1}(t) & \geq \frac{\Delta}{N^2} - \sum_{i=1}^N B_i^*(t) - \frac{2}{k}L_1(t) - \frac{2}{k}Y_k^{(2)}(t) - \frac{2}{N-k}Y_k^{(3)}(t) \\ & \geq \frac{\Delta}{N^2} - Y_k^{(4)}(t) - \frac{2}{k}L_1(t), \end{aligned}$$

where $Y_k^{(4)}(t) = \sum_{i=1}^N B_i^*(t) + \frac{2}{k}Y_k^{(2)}(t) + \frac{2}{N-k}Y_k^{(3)}(t)$. Note that if $L_{k+1}(\sigma_{2l}) > 0$, then $\inf_{0 \leq s \leq \sigma_{2l}} Z_{k+1}(s) = 0$, which in turn implies that $\theta_k \in [0, \sigma_{2l}]$ and so from the above display

$$\frac{\Delta}{N^2} \leq Y_k^{(4)}(\theta_k) + \frac{2}{k}L_1(\theta_k) \leq Y_k^{(4)}(\sigma_{2l}) + \frac{2}{k}L_1(\sigma_{2l}).$$

As a consequence,

$$\begin{aligned} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(L_{k+1}(\sigma_{2l}) > 0) & \leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*\left(\frac{\Delta}{N^2} \leq Y_k^{(4)}(\sigma_{2l}) + \frac{2}{k}L_1(\sigma_{2l})\right) \\ & \leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*\left(\frac{\Delta}{2N^2} \leq Y_k^{(4)}(\sigma_{2l})\right) + \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*\left(\frac{\Delta}{2N^2} \leq \frac{2}{k}L_1(\sigma_{2l})\right). \end{aligned} \quad (8.14)$$

Consider now the first term on the right side. Then, for $T \geq 1$,

$$\begin{aligned} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*\left(\frac{\Delta}{2N^2} \leq Y_k^{(4)}(\sigma_{2l})\right) & \leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(\sigma_{2l} > T) + \mathbb{P}\left(\frac{\Delta}{2N^2} \leq Y_k^{(4)}(T)\right) \\ & \leq l \sup_{\hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(\sigma_2 > T/l) + c_1 e^{-c_2 \Delta^2/T} \\ & \leq c_3 l e^{-c_4 T/l} + c_1 e^{-c_2 \Delta^2/T}, \end{aligned} \quad (8.15)$$

where $c_i \in (0, \infty)$ are constants that do not depend on Δ or $\hat{\mathbf{z}} \in \mathcal{S}_\Delta$, the second inequality uses the strong Markov property and a standard concentration estimate (see (1.4)) and the last inequality is a consequence of Proposition 7.6. Now consider the second term on the right side of (8.14). For $T \geq 1$,

$$\begin{aligned} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* \left(\frac{\Delta}{2N^2} \leq \frac{2}{k} L_1(\sigma_{2l}) \right) &\leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (\sigma_{2l} > T) + \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* \left(\frac{\Delta}{2N^2} \leq \frac{2}{k} L_1(T) \right) \\ &\leq c_3 l e^{-c_4 T/l} + \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (L_1(T) \geq \frac{\Delta k}{4N^2}). \end{aligned} \quad (8.16)$$

Note that under $\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*$,

$$\sup_{0 \leq s \leq T} V(s) < 2g + gT.$$

Thus, using (7.2) and (7.3), for any $T \geq 1$,

$$\begin{aligned} L_1(T) &\leq \frac{1}{2} L_2(T) + (2g + gT)T + B_1^*(T) \\ &\leq \frac{(N-1)}{N} L_1(T) + \frac{1}{N} \bar{Y}(T) + (2g + gT)T + B_1^*(T) \end{aligned}$$

and thus, with $\tilde{Y}(T) = \bar{Y}(T) + NB_1^*(T)$ and $c_5 = 3gN$

$$L_1(T) \leq \bar{Y}(T) + (2g + gT)TN + NB_1^*(T) \leq \tilde{Y}(T) + c_5 T^2. \quad (8.17)$$

Take $T = T(\Delta) \doteq \frac{1}{2N} \sqrt{\frac{\Delta}{2c_5}}$ and observe that $c_5 T^2 \leq \Delta k / (8N^2)$ for all $k \in \{1, \dots, N\}$. Choose $\Delta_1 > 0$ such that $T(\Delta_1) \geq 1$. Then, it follows by a concentration estimate that, for $\Delta \geq \Delta_1$,

$$\begin{aligned} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (L_1(T) \geq \frac{\Delta k}{4N^2}) &\leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (\tilde{Y}(T) \geq \frac{\Delta k}{4N^2} - c_5 T^2) \\ &\leq \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (\tilde{Y}(T) \geq c_5 T^2) \leq c_6 e^{-c_7 T^3}. \end{aligned} \quad (8.18)$$

Then, using (8.18), (8.16) and (8.15) in (8.14), we obtain constants $c'_2, c'_7, c_8, c_9 > 0$ such that for all $\Delta \geq \Delta_1$,

$$\begin{aligned} \mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^* (L_{k+1}(\sigma_{2l}) > 0) &\leq 2c_3 l e^{-c_4 v \sqrt{\Delta}/l} + c_1 e^{-c'_2 \Delta^{3/2}} + c_6 e^{-c'_7 \Delta^{3/2}} \\ &\leq 2c_3 l e^{-c_4 v \sqrt{\Delta}/l} + c_8 e^{-c_9 \Delta^{3/2}}. \end{aligned} \quad (8.19)$$

We will now obtain an upper bound for $\mathbb{E}_{(2g,0,\hat{\mathbf{z}})} (L_{k+1}(\sigma_{2l})^2)$. From (7.3) we have that, for $m \geq 2$ and $t \geq 0$,

$$\begin{aligned} (L_N(t) - \frac{1}{2} L_{N-1}(t)) + \sum_{j=m}^{N-2} (N-j)(L_{j+1}(t) - \frac{1}{2}(L_{j+2}(t) + L_j(t))) \\ \leq \sum_{j=m}^{N-1} (N-j) B_{j+1}^*(t) \doteq Y_m^{(5)}(t). \end{aligned}$$

The left side above equals $\frac{N-m+1}{2}L_{m+1}(t) - \frac{N-m}{2}L_m(t)$ and so

$$\frac{N-m+1}{2}L_{m+1}(t) - \frac{N-m}{2}L_m(t) \leq Y_m^{(5)}(t).$$

Dividing by $(N-m)(N-m+1)/2$ throughout, we have

$$\frac{1}{N-m}L_{m+1}(t) - \frac{1}{N-m+1}L_m(t) \leq \frac{2}{(N-m)(N-m+1)}Y_m^{(5)}(t), \quad 2 \leq m \leq N-1.$$

Summing over m from 2 to k , the above yields

$$\frac{1}{N-k}L_{k+1}(t) - \frac{1}{N-1}L_2(t) \leq \sum_{m=2}^k \frac{2Y_m^{(5)}(t)}{(N-m)(N-m+1)} \doteq Y_k^{(6)}(t).$$

and thus

$$L_{k+1}(t) \leq \frac{N-k}{N-1}L_2(t) + (N-k)Y_k^{(6)}(t).$$

From (7.13) (recall it holds for any value of $k \geq 1$) we have

$$L_{k+1}(t) \leq \frac{N-k}{N-1} \left(\frac{2(k-1)}{k}L_1(t) + \frac{2}{k}Y_k^{(1)}(t) + \frac{1}{k}L_{k+1}(t) \right) + (N-k)Y_k^{(6)}(t).$$

Thus

$$\frac{N(k-1)}{k(N-1)}L_{k+1}(t) \leq \frac{2(k-1)(N-k)}{k(N-1)}L_1(t) + \frac{2(N-k)}{k(N-1)}Y_k^{(1)}(t) + (N-k)Y_k^{(6)}(t)$$

and consequently, when $k > 1$,

$$\begin{aligned} L_{k+1}(t) &\leq \frac{2(N-k)}{N}L_1(t) + \frac{2(N-k)}{N(k-1)}Y_k^{(1)}(t) + \frac{k(N-k)(N-1)}{N(k-1)}Y_k^{(6)}(t) \\ &= \frac{2(N-k)}{N}L_1(t) + Y_k^{(7)}(t), \end{aligned} \quad (8.20)$$

where $Y_k^{(7)}(t) = \frac{2(N-k)}{N(k-1)}Y_k^{(1)}(t) + \frac{k(N-k)(N-1)}{N(k-1)}Y_k^{(6)}(t)$. Recalling the inequality (7.2) we see that (8.20) also holds with $k = 1$ and $Y_1^{(7)}(t) \doteq \frac{2}{N}\tilde{Y}(t)$.

Using this, we obtain that

$$\begin{aligned} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*(L_{k+1}(\sigma_{2l})^2) &\leq 2\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*(Y_k^{(7)}(\sigma_{2l}))^2 + \frac{8(N-k)^2}{N^2}\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*L_1^2(\sigma_{2l}) \\ &\leq 2\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*(Y_k^{(7)}(\sigma_{2l}))^2 + \frac{16(N-k)^2}{N^2}\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*(\tilde{Y}(\sigma_{2l}))^2 \\ &\quad + \frac{16(N-k)^2c_5^2}{N^2}\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*(\sigma_{2l})^4. \end{aligned}$$

where the last line is from (8.17). Thus, using the above bound, the strong Markov property, and Proposition 7.6, there is a $b_1 \in (0, \infty)$ such that, for all $l \in \mathbb{N}$, $\hat{\mathbf{z}} \in \mathcal{S}_\Delta$, and k satisfying (7.16),

$$\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^*(L_{k+1}(\sigma_{2l})^2) \leq b_1 l^5.$$

Applying Cauchy-Schwarz inequality and using (8.19), we obtain positive constants D_1, D_2, D_3 such that for all $\Delta \geq \Delta_1$,

$$\begin{aligned} \mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l}) &\leq (\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* L_{k+1}(\sigma_{2l})^2)^{1/2} (\mathbb{P}_{(2g,0,\hat{\mathbf{z}})}^*(L_{k+1}(\sigma_{2l}) > 0))^{1/2} \\ &\leq b_1^{1/2} l^{5/2} (2c_3 l e^{-c_4 \sqrt{\Delta}/l} + c_8 e^{-c_9 \Delta^{3/2}})^{1/2} \\ &= D_1 l^{5/2} (\sqrt{l} e^{-D_2 \sqrt{\Delta}/l} + e^{-D_3 \Delta^{3/2}}), \end{aligned}$$

as desired. \square

8.4. Proof of Proposition 7.16. In this section we give the proof of Proposition 7.16. Proof relies on five preliminary lemmas which extend some estimates derived in Sects. 7.3 and 7.4 to more general starting configurations. The first four are required to verify part (1) of the proposition and the last one is used to check part (2). Proof of the proposition is at the end of the section. Recall the set C^* from (7.22) and stopping times σ_1, σ_2 from Sect. 7.4. Recall Γ from (7.18).

Lemma 8.1. *There exists a $\rho_0 > 0$ and $b_1, b_2 > 0$ such that, for all $\rho \in (0, \rho_0)$, there is a $b_3(\rho) \in (0, \infty)$ such that for any $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,*

$$\mathbb{E}_{(v,\mathbf{z})}^* e^{\rho \Gamma} \leq b_3(\rho) e^{b_1 \rho (|v| + z_1 + \bar{z}_2)} \mathbb{E}_{(v,\mathbf{z})}^* e^{b_2 \rho \sigma_1}.$$

Proof. Define the stopping time

$$\sigma^* \doteq \inf\{t \geq \sigma_2 : (V(t), \mathbf{Z}(t)) \in C^*\}.$$

From Proposition 7.15, there exist positive constants d_0, c' such that for any $\gamma \in (0, \gamma_7/2)$, where γ_7 is as in that lemma,

$$\mathbb{E}_{(2g,0,\hat{\mathbf{z}})}^* e^{\gamma \Gamma} \leq d_0 e^{c' \gamma \bar{z}_2}, \quad \hat{\mathbf{z}} \in \mathbb{R}_+^{N-1}.$$

Fix $\rho'_0 > 0$ such that $\rho'_0 < \min\{\gamma_7, \gamma_5, \gamma_4\}$ and $\rho'_0(1 + \kappa'_2 g) < \gamma_4$, where γ_5 and κ'_2 are as in Lemma 7.9 and γ_4 is as in Lemma 7.8. Then, for $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$ and $\rho \in (0, \rho'_0/2)$,

$$\begin{aligned} \mathbb{E}_{(v,\mathbf{z})}^* e^{\rho \Gamma} &\leq \mathbb{E}_{(v,\mathbf{z})}^* e^{\rho \sigma^*} = \mathbb{E}_{(v,\mathbf{z})}^* \left[\mathbb{E}_{(v,\mathbf{z})}^* \left[e^{\rho \sigma^*} \mid \mathcal{F}_{\sigma_2}^* \right] \right] \leq d_0 \mathbb{E}_{(v,\mathbf{z})}^* e^{\rho \sigma_2 + c' \rho \bar{z}_2(\sigma_2)} \\ &\leq d_0 \left(\mathbb{E}_{(v,\mathbf{z})}^* e^{2\rho \sigma_2} \right)^{1/2} \left(\mathbb{E}_{(v,\mathbf{z})}^* e^{2c' \rho \bar{z}_2(\sigma_2)} \right)^{1/2}. \end{aligned} \quad (8.21)$$

Recall the stopping time $\eta_1 \geq \sigma_1$ defined in (7.7). Proceeding as in the proof of Proposition 7.6 (see (7.8)), with $d_1 = 1 + \kappa'_1 e^{8\rho \kappa'_2 g}$ and $d_2 = (1 + \kappa'_2 g)$, using Lemmas 7.8 and 7.9,

$$\begin{aligned} \mathbb{E}_{(v,\mathbf{z})}^* e^{2\rho \sigma_2} &\leq \mathbb{E}_{(v,\mathbf{z})}^* \left[\mathbf{1}_{\eta_1 < \sigma_2} \mathbb{E}_{(v,\mathbf{z})}^* \left[e^{2\rho \sigma_2} \mid \mathcal{F}_{\eta_1}^* \right] \right] + \mathbb{E}_{(v,\mathbf{z})}^* e^{2\rho \eta_1} \\ &\leq \kappa'_1 \mathbb{E}_{(v,\mathbf{z})}^* e^{2\rho \kappa'_2 V(\eta_1) + 2\rho \eta_1} + \mathbb{E}_{(v,\mathbf{z})}^* e^{2\rho \eta_1} \\ &\leq d_1 \mathbb{E}_{(v,\mathbf{z})}^* e^{2\rho d_2 \eta_1} \\ &= d_1 \mathbb{E}_{(v,\mathbf{z})}^* \left[e^{2\rho d_2 \sigma_1} \mathbb{E}_{(v,\mathbf{z})}^* \left[e^{2\rho d_2 (\eta_1 - \sigma_1)} \mid \mathcal{F}_{\sigma_1}^* \right] \right] \\ &\leq \kappa_1 d_1 \mathbb{E}_{(v,\mathbf{z})}^* \left[e^{2\rho d_2 \sigma_1} e^{2\rho d_2 \kappa_2 Z_1(\sigma_1)^{1/2}} \right] \\ &\leq \kappa_1 d_1 \left(\mathbb{E}_{(v,\mathbf{z})}^* e^{4\rho d_2 \kappa_2 Z_1(\sigma_1)^{1/2}} \right)^{1/2} \left(\mathbb{E}_{(v,\mathbf{z})}^* e^{4\rho d_2 \sigma_1} \right)^{1/2}, \end{aligned} \quad (8.22)$$

where we used Lemma 7.9 for the second inequality, $V(\eta_1) \leq 4g + g\eta_1$ in the third inequality, and Lemma 7.8 in the penultimate inequality (and the observation that $\rho'_0 d_2 < \gamma_4$). Now we estimate exponential moments of $Z_1(\sigma_1)^{1/2}$. Note that, under $\mathbb{P}_{(v, \mathbf{z})}^*$, $L_1(\sigma_1) = g\sigma_1 + v - 4g$, from which it follows that

$$Z_1(\sigma_1) \leq \sup_{0 \leq s \leq \sigma_1} B_1(s) + z_1 + g\sigma_1^2 + g\sigma_1 + 2|v|\sigma_1 + |v|$$

and so, using $\sqrt{a+b} \leq \sqrt{a} + \sqrt{b}$ and $\sqrt{ab} \leq (a+b)/2$ for $a, b \geq 0$,

$$Z_1(\sigma_1)^{1/2} \leq \left(\sup_{0 \leq s \leq \sigma_1} B_1(s) \right)^{1/2} + (\sqrt{g} + 3/2)\sigma_1 + \frac{1}{2}(z_1 + 3|v| + g + 1).$$

Using this bound in (8.22) and Young's inequality, we obtain a finite positive constant d_3 not depending on v, \mathbf{z}, ρ such that

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{2\rho\sigma_2} &\leq d_3 e^{2\rho d_2 \kappa_2 (z_1 + 3|v|)} \\ &\left[\left(\mathbb{E}_{(v, \mathbf{z})}^* e^{8\rho d_2 \kappa_2 \left(\sup_{0 \leq s \leq \sigma_1} B_1(s) \right)^{1/2}} \right)^{1/2} + \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{8\rho d_2 \kappa_2 (\sqrt{g} + 3/2)\sigma_1} \right)^{1/2} \right] \\ &\left(\mathbb{E}_{(v, \mathbf{z})}^* e^{4\rho d_2 \sigma_1} \right)^{1/2}. \end{aligned} \quad (8.23)$$

The expectation involving $\sup_{0 \leq s \leq \sigma_1} B_1(s)$ above is bounded as in the proof of Lemma 7.7 (see (8.6)) to obtain $\rho''_0, d_4, d_5 \in (0, \infty)$ such that $d_5 \rho''_0 < d_2$, and for any $\rho \in (0, \rho''_0)$ and $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{8\rho d_2 \kappa_2 (\sup_{0 \leq s \leq \sigma_1} B_1(s))^{1/2}} &\leq e^{8\rho d_2 \kappa_2} + d_4 e^{d_5 \rho^2} \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{4\rho d_2 \sigma_1} \right)^{1/2} \sum_{k=0}^{\infty} e^{d_5 \rho^2 k - 2\rho d_2 k} \\ &\leq e^{8\rho d_2 \kappa_2} + d_6(\rho) \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{4\rho d_2 \sigma_1} \right)^{1/2}, \end{aligned} \quad (8.24)$$

where $d_6(\rho) \doteq d_4 e^{d_5 \rho^2} (1 - e^{-\rho d_2})^{-1}$. Using this bound in (8.23), we conclude that for every $0 < \rho < \min\{\rho'_0/2, \rho''_0\}$, there exists a finite positive constant $d_7(\rho)$ satisfying

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{2\rho\sigma_2} \leq e^{2\rho d_2 \kappa_2 (z_1 + 3|v|)} d_7(\rho) \mathbb{E}_{(v, \mathbf{z})}^* e^{d'_2 \rho \sigma_1}, \quad (8.25)$$

where $d'_2 = \max\{4d_2, 8d_2 \kappa_2 (\sqrt{g} + 3/2)\}$.

Now, we estimate $\mathbb{E}_{(v, \mathbf{z})}^* e^{2c'\rho \bar{Z}_2(\sigma_2)}$. From (7.11) and (7.2), $\bar{Z}_2(t) \leq \bar{z}_2 + M(t) + \bar{Y}(t)$, $t \geq 0$. Hence, writing $\tilde{Y}(t) \doteq M(t) + \bar{Y}(t)$,

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{2c'\rho \bar{Z}_2(\sigma_2)} \leq e^{2c'\rho \bar{z}_2} \mathbb{E}_{(v, \mathbf{z})}^* e^{2c'\rho \tilde{Y}(\sigma_2)}.$$

Proceeding exactly as in (8.24), we obtain $\rho'''_0 > 0$ such that for every $\rho \in (0, \rho'''_0)$, there exists a $d_8(\rho) \in (0, \infty)$ such that

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{2c'\rho \tilde{Y}(\sigma_2)} &\leq \sum_{k=0}^{\infty} \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{4c'\rho \sup_{0 \leq s \leq k+1} \tilde{Y}(s)} \right)^{1/2} (\mathbb{P}_{(v, \mathbf{z})}^*(\sigma_2 \geq k))^{1/2} \\ &\leq d_8(\rho) \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{2\rho\sigma_2} \right)^{1/2}, \end{aligned}$$

which, along with (8.25), gives

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{2c' \rho \bar{Z}_2(\sigma_2)} &\leq e^{2c' \rho \bar{z}_2} d_8(\rho) \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{2\rho \sigma_2} \right)^{1/2} \leq e^{2c' \rho \bar{z}_2} e^{\rho d_2 \kappa_2 (z_1 + 3|v|)} d_8(\rho) d_7(\rho)^{1/2} \\ &\quad \left(\mathbb{E}_{(v, \mathbf{z})}^* e^{d_2' \rho \sigma_1} \right)^{1/2}. \end{aligned} \quad (8.26)$$

The result, with $\rho_0 \doteq \min\{\rho_0'/2, \rho_0'', \rho_0'''\}$, now follows upon using (8.25) and (8.26) in (8.21). \square

Lemma 8.2. *Let $\vartheta \doteq \hat{\tau}_{4g} \wedge \hat{\tau}_0$. Then there is a $\beta_1 > 0$ and $D_1 > 0$ such that, for all $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,*

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_1 \vartheta} \leq D_1 e^{\beta_1 (|v| + z_1)}.$$

Proof. Fix $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$. We consider three cases.

Case 1: $v \in [0, 4g]$. In this case the result is immediate from Lemma 7.5.

Case 2: $v > 4g$. In this case, for all $t \leq \vartheta$, $V(t) > 4g$. Thus, for such t , we have

$$\begin{aligned} Z_1(t) &= z_1 + B_1(t) + L_1(t) - \frac{1}{2} L_2(t) - \int_0^t V(s) ds \\ &\leq z_1 + \sup_{0 \leq s \leq t} B_1(s) + L_1(t) - 4gt. \end{aligned}$$

Consequently, $-L_1(t) \leq z_1 + \sup_{0 \leq s \leq t} B_1(s) - 4gt$. Thus we have, for $t \leq \vartheta$,

$$V(t) = gt - L_1(t) + v \leq z_1 + \sup_{0 \leq s \leq t} B_1(s) - 3gt + v \doteq Q_1(t).$$

Letting $\tau_{4g}^{Q_1} \doteq \inf\{t \geq 0 : Q_1(t) = 4g\}$, we have $\tau_{4g}^{Q_1} \geq \vartheta$. Thus, for $\beta_1, \theta > 0$,

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_1 \vartheta} &\leq 1 + \int_1^\infty \mathbb{P}_{(v, \mathbf{z})}^* (Q_1(\frac{1}{\beta_1} \ln(s)) > 4g) ds \\ &\leq 1 + e^{-4g\theta\beta_1} \int_1^\infty \mathbb{E}_{(v, \mathbf{z})}^* e^{\theta\beta_1 Q_1(\frac{1}{\beta_1} \ln(s))} ds \\ &= 1 + e^{-4g\theta\beta_1 + \theta\beta_1(z_1 + v)} \int_1^\infty e^{-3\theta g \ln(s)} \mathbb{E}_{(v, \mathbf{z})}^* e^{\theta\beta_1 \sup_{0 \leq t \leq \frac{1}{\beta_1} \ln(s)} B_1(t)} ds \\ &\leq 1 + Q_1 e^{-4g\theta\beta_1 + \theta\beta_1(z_1 + v)} \int_1^\infty e^{-3\theta g \ln(s) + Q_2 \theta^2 \beta_1 \ln(s)} ds \end{aligned}$$

where the last line uses the estimate (1.4). Taking $\theta = g^{-1}$ and $\beta_1 = g^2/Q_2$, we now see that

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_1 \vartheta} \leq 1 + Q_1 e^{-4g\theta\beta_1 + \theta\beta_1(z_1 + v)}$$

which completes the proof for Case 2.

Case 3: $v < 0$. In this case, for $t \leq \vartheta$, we have $V(t) < 0$. Thus for such t , from (7.3) and (7.2),

$$\begin{aligned} L_1(t) &= \sup_{0 \leq s \leq t} (-z_1 - B_1(s) + \frac{1}{2} L_2(s) + \int_0^s V(u) du)^+ \leq B_1^*(t) + \frac{1}{2} L_2(t) \leq B_1^*(t) \\ &\quad + \frac{N-1}{N} L_1(t) + \frac{1}{N} \bar{Y}(t). \end{aligned}$$

Consequently,

$$L_1(t) \leq NB_1^*(t) + \bar{Y}(t)$$

and so

$$V(t) = gt - L_1(t) + v \geq gt - NB_1^*(t) - \bar{Y}(t) + v \doteq Q_2(t).$$

Letting, $\tau_0^{Q_2} \doteq \inf\{t \geq 0 : Q_2(t) = 0\}$, we then have, $\tau_0^{Q_2} \geq \vartheta$. Thus, for $\theta, \beta_1 > 0$,

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_1 \vartheta} &\leq 1 + \int_1^\infty \mathbb{P}_{(v, \mathbf{z})}^*(Q_2(\frac{\ln(s)}{\beta_1}) < 0) ds \\ &= 1 + \int_1^\infty \mathbb{P}_{(v, \mathbf{z})}^*(g \frac{\ln(s)}{\beta_1} + v < NB_1^*(\frac{\ln(s)}{\beta_1}) + \bar{Y}(\frac{\ln(s)}{\beta_1})) ds \\ &\leq 1 + e^{-\theta \beta_1 v} \int_1^\infty s^{-\theta g} \mathbb{E}_{(v, \mathbf{z})}^* e^{\theta \beta_1 (NB_1^*(\frac{\ln(s)}{\beta_1}) + \bar{Y}(\frac{\ln(s)}{\beta_1}))} ds \\ &\leq 1 + \varrho_1 e^{\theta \beta_1 |v|} \int_1^\infty s^{-\theta g} s^{\varrho_2 \beta_1 \theta^2} ds, \end{aligned}$$

where in the last line we have used the estimate (1.4). Take $\theta = 4g^{-1}$ and $\beta_1 = g^2/(8\varrho_2)$, then

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_1 \vartheta} \leq 1 + \varrho_1 e^{\theta \beta_1 |v|}.$$

This completes the proof for Case 3 and thus the result follows. \square

Lemma 8.3. *There is a $\beta_2 > 0$ and $\kappa_1, \kappa_2 > 0$ such that, for all $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,*

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_2 \sigma_1} \leq \kappa_1 e^{\kappa_2 (|v| + z_1)}.$$

Proof. From Proposition 7.2, with γ as in that proposition,

$$\sup_{\mathbf{z} \in \mathbb{R}_+^N} \mathbb{E}_{(\frac{g}{2N}, \mathbf{z})}^* e^{\gamma \hat{\tau}_{4g}} \doteq d_1 < \infty. \quad (8.27)$$

Also, from Lemma 7.3, with β as in that lemma,

$$\sup_{\mathbf{z} \in \mathbb{R}_+^N} \mathbb{E}_{(0, \mathbf{z})}^* e^{\beta \hat{\tau}_{g/(2N)}} \doteq d_2 < \infty. \quad (8.28)$$

With β_1 as in Lemma 8.2, let $\beta_2 \in (0, \min\{\gamma, \beta, \beta_1\})$. Recall the stopping time ϑ from Lemma 8.2. Define stopping times

$$\vartheta_1 \doteq \inf\{t \geq \vartheta : V(t) = g/(2N)\}, \quad \vartheta_2 \doteq \inf\{t \geq \vartheta_1 : V(t) = 4g\}.$$

Then, for $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_2 \sigma_1} \leq \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 = \vartheta\}} e^{\beta_2 \sigma_1}] + \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} e^{\beta_2 \sigma_1}].$$

From Lemma 8.2, with β_1 and D_1 as in the lemma,

$$\mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 = \vartheta\}} e^{\beta_2 \sigma_1}] \leq \mathbb{E}_{(v, \mathbf{z})}^* [e^{\beta_2 \vartheta}] \leq D_1 e^{\beta_1 (|v| + z_1)}.$$

Next, from (8.27),

$$\begin{aligned}\mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} e^{\beta_2 \sigma_1}] &\leq \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} e^{\beta_2 \vartheta_2}] = \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} \mathbb{E}_{(v, \mathbf{z})}^* [e^{\beta_2 \vartheta_2} \mid \mathcal{F}_{\vartheta_1}^*]] \\ &\leq d_1 \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} e^{\beta_2 \vartheta_1}].\end{aligned}$$

Also, from (8.28),

$$\begin{aligned}\mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} e^{\beta_2 \vartheta_1}] &= \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} \mathbb{E}_{(v, \mathbf{z})}^* [e^{\beta_2 \vartheta_1} \mid \mathcal{F}_{\vartheta}^*]] \\ &\leq d_2 \mathbb{E}_{(v, \mathbf{z})}^* [\mathbf{1}_{\{\sigma_1 > \vartheta\}} e^{\beta_2 \vartheta}] \leq d_2 D_1 e^{\beta_1(|v|+z_1)},\end{aligned}$$

where the last line is from Lemma 8.2. Combining the above estimates, for all $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\beta_2 \sigma_1} \leq D_1 e^{\beta_1(|v|+z_1)} + d_1 d_2 D_1 e^{\beta_1(|v|+z_1)}.$$

The result follows. \square

Lemma 8.4. *There is a $\kappa \in (0, \infty)$ such that for every $\alpha > 0$ there is a $s_\alpha > 0$ with*

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\alpha(|V(1)|+Z_1(1)+\bar{Z}_2(1))} \leq s_\alpha e^{\kappa \alpha(|v|+z_1+\bar{z}_2)}, \text{ for all } (v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N.$$

Proof. Since $V(t) \leq g + |v|$ for $t \leq 1$, we have from (7.1)

$$L_1(1) \leq (g + |v|) W_{11} + \sum_{i=1}^N W_{1,i} B_i^*(1) = N(g + |v|) + \sum_{i=1}^N W_{1,i} B_i^*(1).$$

Thus, under $\mathbb{P}_{(v, \mathbf{z})}^*$,

$$\begin{aligned}Z_1(1) + |V(1)| &\leq z_1 + B_1(1) + L_1(1) - \frac{1}{2} L_2(1) - \int_0^1 V(s) ds + g + L_1(1) + |v| \\ &\leq 2|v| + \frac{g}{2} + z_1 + \sup_{0 \leq s \leq 1} B_1(s) + 3L_1(1) \\ &\leq 2|v| + \frac{g}{2} + z_1 + \sup_{0 \leq s \leq 1} B_1(s) + 3(N(g + |v|) + \sum_{i=1}^N W_{1,i} B_i^*(1)).\end{aligned}$$

Moreover, from (7.11) and (7.2), $\bar{Z}_2(1) \leq \bar{z}_2 + M(1) + \bar{Y}(1)$, $t \geq 0$. The result is now immediate from the estimate in (1.4). \square

For $K \in \mathbb{N}$, with $K > 256/g$, define

$$R_K \doteq [\frac{1}{K}, \frac{g}{128} - \frac{1}{K}] \times [\frac{1}{K}, K] \times [0, K]^{N-1}.$$

Lemma 8.5. *There is a $K \in \mathbb{N}$, $K > 256/g$, such that*

$$\inf_{(v, \mathbf{z}) \in C^*} \mathbb{P}^1((v, \mathbf{z}), R_K) \doteq c_K > 0.$$

Proof. Suppose that, for every $K \in \mathbb{N}$, $K > 256/g$,

$$\inf_{(v, \mathbf{z}) \in C^*} \mathbb{P}^1((v, \mathbf{z}), R_K) = 0.$$

Then, we can find a sequence $\{(v_K, \mathbf{z}_K)\}_{K \in \mathbb{N}} \subset C^*$ such that

$$\mathbb{P}^1((v_K, \mathbf{z}_K), R_K) \leq \frac{1}{K}. \quad (8.29)$$

Since C^* is compact, we can find $(v^*, \mathbf{z}^*) \in C^*$ such that, along a subsequence (labeled again with K), $(v_K, \mathbf{z}_K) \rightarrow (v^*, \mathbf{z}^*)$. From the second statement in Lemma 5.1,

$$\mathbb{P}^1((v^*, \mathbf{z}^*), R) > 0.$$

Since R_K increase to R as $K \rightarrow \infty$, we can find a $K^* \in \mathbb{N}$, $K^* > 256/g$, such that

$$\mathbb{P}^1((v^*, \mathbf{z}^*), R_{K^*}) \doteq a_{K^*} > 0.$$

Choose a real, continuous function $f : \mathbb{R} \times \mathbb{R}_+^N$ such that $0 \leq f \leq 1$, $f = 1$ on R_{K^*} and $f = 0$ on $R_{2K^*}^c$. Then

$$\begin{aligned} \liminf_{K \rightarrow \infty} \mathbb{P}^1((v_K, \mathbf{z}_K), R_{2K^*}) &\geq \liminf_{K \rightarrow \infty} \int f(v, \mathbf{z}) \mathbb{P}^1((v_K, \mathbf{z}_K), (dv, d\mathbf{z})) \\ &= \int f(v, \mathbf{z}) \mathbb{P}^1((v^*, \mathbf{z}^*), (dv, d\mathbf{z})) \\ &\geq \mathbb{P}^1((v^*, \mathbf{z}^*), R_{K^*}) = a_{K^*} > 0, \end{aligned}$$

where the middle equality follows from the Feller property of the transition probability kernel \mathbb{P}^1 . The Feller property can be verified by analyzing two copies of the process (1.3) starting from different initial conditions but driven by the same Brownian motion. Using the Lipschitz property of the Skorohod map and Grönwall's lemma, the distance between the coupled processes in sup-norm on any given compact time interval can be made small (in a pathwise sense) if the initial conditions are close enough.

On the other hand, from (8.29)

$$\limsup_{K \rightarrow \infty} \mathbb{P}^1((v_K, \mathbf{z}_K), R_{2K^*}) \leq \limsup_{K \rightarrow \infty} \mathbb{P}^1((v_K, \mathbf{z}_K), R_K) = 0.$$

This is a contradiction which completes the proof of the lemma. \square

We can now complete the proof of Proposition 7.16.

Proof of Proposition 7.16 Fix $\eta > 0$ such that $\eta < \rho_0$ and $b_2\eta \leq \beta_2$, where ρ_0 and b_2 are as Lemma 8.1 and β_2 is as in Lemma 8.3. Combining Lemmas 8.1 and 8.3, for all $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\eta\Gamma} \leq b_3(\eta) e^{b_1\eta(|v|+z_1+\bar{z}_2)} \mathbb{E}_{(v, \mathbf{z})}^* e^{b_2\eta\sigma_1} \leq b_3(\eta) \kappa_1 e^{b_1\eta(|v|+z_1+\bar{z}_2)+\kappa_2(|v|+z_1)}.$$

Thus

$$\begin{aligned} \mathbb{E}_{(v, \mathbf{z})}^* e^{\eta\tau_{C^*}(1)} &= \mathbb{E}_{(v, \mathbf{z})}^* \left[\mathbb{E}_{(v, \mathbf{z})}^* \left[e^{\eta\tau_{C^*}(1)} \mid \mathcal{F}_1^* \right] \right] \\ &\leq b_3(\eta) \kappa_1 e^{\eta} \mathbb{E}_{(v, \mathbf{z})}^* e^{(b_1\eta+\kappa_2)(|V(1)|+Z_1(1)+\bar{Z}_2(1))}. \end{aligned}$$

Consequently, with $\alpha = b_1\eta + \kappa_2$, for all $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$,

$$\mathbb{E}_{(v, \mathbf{z})}^* e^{\eta \tau_{C^*}(1)} \leq b_3(\eta) \kappa_1 s_\alpha e^\eta e^{\kappa \alpha (|v| + z_1 + \bar{z}_2)},$$

where s_α and κ are as in Lemma 8.4. This immediately implies part (1) of the proposition.

We now consider part (2). Let K be as in Lemma 8.5. From Theorem 4.1, with $M_1 \doteq \inf_{(v, \mathbf{z}) \in R_K} K_{(v, \mathbf{z})}$ (which is positive)

$$\inf_{(v, \mathbf{z}) \in R_K} \mathbb{P}^\zeta((v, \mathbf{z}), B) \geq \lambda(B \cap D) \inf_{(v, \mathbf{z}) \in R_K} K_{(v, \mathbf{z})} = M_1 \lambda(B \cap D).$$

Also, from Lemma 8.5, with K as in the lemma, for $(v, \mathbf{z}) \in C^*$ and $B \in \mathcal{B}(\mathbb{R} \times \mathbb{R}_+^N)$,

$$\begin{aligned} \mathbb{P}^{1+\zeta}((v, \mathbf{z}), B) &\geq \int_{R_K} \mathbb{P}^1((v, \mathbf{z}), (d\tilde{v}, d\tilde{\mathbf{z}})) \mathbb{P}^\zeta((\tilde{v}, \tilde{\mathbf{z}}), B) \\ &\geq M_1 \lambda(B \cap D) \mathbb{P}^1((v, \mathbf{z}), R_K) \geq M_1 c_K \lambda(B \cap D). \end{aligned}$$

The result now follows on taking $\nu(\cdot) = M_1 c_K \lambda(\cdot \cap D)$ and $r_1 = 1 + \zeta$. \square

9. Law of Large Numbers

In this section, we prove Theorem 2.6. We begin with the following lemma.

Lemma 9.1. *For any $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, $\mathbb{P}_{(v, \mathbf{z})}^*$ -almost every ω , there exists a $m^*(\omega) \in (0, \infty)$ such that*

$$|V(t, \omega)| + \sum_{i=1}^N Z_i(t, \omega) \leq m^*(\omega) (\log t)^2, \text{ for all } t \geq 2.$$

Proof. Recall the set C^* from (7.22) and the stopping time $\tau_{C^*}(1)$ defined just after. Let $\Sigma(t) \doteq |V(t)| + \sum_{i=1}^N Z_i(t)$, $t \geq 0$. We will first show that there exist positive constants c_1, c_2 such that

$$\sup_{(v, \mathbf{z}) \in C^*} \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq \tau_{C^*}(1)} \Sigma(t) \geq x \right) \leq c_1 e^{-c_2 \sqrt{x}}, \quad x > 0. \quad (9.1)$$

Note that, from (1.3) and Lemma 7.1, for $(v, \mathbf{z}) \in C^*$ and $t \geq 0$,

$$\begin{aligned} \Sigma(t) &= |v| + \sum_{i=1}^N z_i + B_N(t) + gt + L_1(t) + \frac{1}{2} L_N(t) \\ &\leq 2g + \Delta + B_N(t) + gt + \sum_{i=1}^N W_{i,1} t \sup_{0 \leq s \leq t} (V(s))^+ + \sum_{i,j=1}^N W_{i,j} B_j^*(t) \\ &\leq 2g + \Delta + gt + 2gt \sum_{i=1}^N W_{i,1} + gt^2 \sum_{i=1}^N W_{i,1} + B_N(t) + \sum_{i,j=1}^N W_{i,j} B_j^*(t). \end{aligned}$$

Using this bound along with part (1) of Proposition 7.16 and (1.4), we obtain positive constants c_1, c_2, x_0 such that, for any $(v, \mathbf{z}) \in C^*$, $x \geq x_0$, and choosing $\delta > 0$ sufficiently small,

$$\begin{aligned} \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq \tau_{C^*}(1)} \Sigma(t) \geq x \right) &\leq \mathbb{P}_{(v, \mathbf{z})}^* (\tau_{C^*}(1) \geq \delta \sqrt{x}) + \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq \delta \sqrt{x}} \Sigma(t) \geq x \right) \\ &\leq \mathbb{P}_{(v, \mathbf{z})}^* (\tau_{C^*}(1) \geq \delta \sqrt{x}) + \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq \delta \sqrt{x}} \left(B_N(t) + \sum_{i,j=1}^N W_{i,j} B_j^*(t) \right) \geq \frac{x}{2} \right) \\ &\leq c_1 e^{-c_2 \sqrt{x}}. \end{aligned}$$

This proves (9.1). Define the following stopping times:

$$\bar{\tau}_0 = 0, \quad \bar{\tau}_{i+1} \doteq \inf \{t \geq \bar{\tau}_i + 1 : (V(t), \mathbf{Z}(t)) \in C^*\}, \quad i \geq 0.$$

Using (9.1), there exists a positive constant c_3 such that for any $(v, \mathbf{z}) \in C^*$, $n \geq 2$ and $m > 0$,

$$\begin{aligned} \mathbb{P}_{(v, \mathbf{z})}^* (\Sigma(t) \geq m(\log t)^2 \text{ for some } t \in [n, n+1]) \\ \leq \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq n+1} \Sigma(t) \geq c_3 m(\log(n+1))^2 \right) \leq \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq \bar{\tau}_{n+1}} \Sigma(t) \geq c_3 m(\log(n+1))^2 \right) \\ \leq (n+1) \sup_{(v, \mathbf{z}) \in C^*} \mathbb{P}_{(v, \mathbf{z})}^* \left(\sup_{t \leq \tau_{C^*}(1)} \Sigma(t) \geq c_3 m(\log(n+1))^2 \right) \leq c_1 (n+1) e^{-c_2 \sqrt{c_3 m} \log(n+1)}, \end{aligned}$$

where we used the strong Markov property to obtain the third inequality. Choosing m sufficiently large, we see from the Borel-Cantelli Lemma that, for any $(v, \mathbf{z}) \in C^*$,

$$\mathbb{P}_{(v, \mathbf{z})}^* \left(\limsup_{t \rightarrow \infty} \frac{\Sigma(t)}{(\log t)^2} < \infty \right) = 1.$$

Finally for an arbitrary $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$, and with Γ as in (7.18) and $A = \{\limsup_{t \rightarrow \infty} \frac{\Sigma(t)}{(\log t)^2} < \infty\}$, applying Lemma 8.1,

$$\mathbb{P}_{(v, \mathbf{z})}^*(A) = \mathbb{P}_{(v, \mathbf{z})}^*(A, \Gamma < \infty) = \mathbb{E}_{(v, \mathbf{z})}^* \left(\mathbb{P}_{(v, \mathbf{z})}^*(A \mid \mathcal{F}_\Gamma^*) 1_{\{\Gamma < \infty\}} \right) = \mathbb{P}_{(v, \mathbf{z})}^*(\Gamma < \infty) = 1.$$

The result follows. \square

Proof of Theorem 2.6. All limits in the proof hold $\mathbb{P}_{(v, \mathbf{z})}^*$ -almost surely for arbitrary $(v, \mathbf{z}) \in \mathbb{R} \times \mathbb{R}_+^N$. From (1.2),

$$\lim_{t \rightarrow \infty} \frac{\sum_{j=1}^N X_{(j)}(t)}{Nt} = \lim_{t \rightarrow \infty} \frac{V(t) + \sum_{j=1}^N X_{(j)}(t)}{Nt} = \frac{g}{N}. \quad (9.2)$$

where we used Lemma 9.1 in the first equality. Moreover, again using Lemma 9.1, for any $i \in \{0, 1, \dots, N\}$,

$$\frac{1}{t} \left| X_{(i)}(t) - \frac{1}{N} \sum_{j=1}^N X_{(j)}(t) \right| \leq \frac{1}{t} |X_{(N)}(t) - X_{(0)}(t)| \leq \frac{1}{t} \sum_{i=1}^N Z_i(t) \rightarrow 0 \quad (9.3)$$

as $t \rightarrow \infty$. The statement in (2.4) now follows from (9.2) and (9.3). Also, (2.5) follows from Lemma 9.1 on noting

$$0 = \lim_{t \rightarrow \infty} \frac{V(t)}{t} = g - \lim_{t \rightarrow \infty} \frac{L_1(t)}{t}.$$

To prove (2.6), note that from (1.2), (1.3) and Lemma 9.1,

$$0 = \lim_{t \rightarrow \infty} \frac{Z_1(t)}{t} = - \lim_{t \rightarrow \infty} \frac{X_{(0)}(t)}{t} - \frac{1}{2} \lim_{t \rightarrow \infty} \frac{L_2(t)}{t} + \lim_{t \rightarrow \infty} \frac{L_1(t)}{t},$$

which gives $\lim_{t \rightarrow \infty} \frac{L_2(t)}{t} = \frac{2(N-1)g}{N}$. Again using (1.3) and Lemma 9.1,

$$0 = \lim_{t \rightarrow \infty} \frac{Z_2(t)}{t} = - \frac{1}{2} \lim_{t \rightarrow \infty} \frac{L_3(t)}{t} + \lim_{t \rightarrow \infty} \frac{L_2(t)}{t} - \lim_{t \rightarrow \infty} \frac{L_1(t)}{t},$$

from which, we obtain $\lim_{t \rightarrow \infty} \frac{L_3(t)}{t} = \frac{2(N-2)g}{N}$. Suppose $N \geq 4$ and, for some $i \in \{3, \dots, N-1\}$, the limit $\lim_{t \rightarrow \infty} \frac{L_j(t)}{t}$ exists and equals $\frac{2(N-j+1)g}{N}$ for all $3 \leq j \leq i$. Then, using (1.3), $\lim_{t \rightarrow \infty} \frac{L_{i+1}(t)}{t}$ exists and

$$0 = \lim_{t \rightarrow \infty} \frac{Z_i(t)}{t} = - \frac{1}{2} \lim_{t \rightarrow \infty} \frac{L_{i+1}(t)}{t} + \lim_{t \rightarrow \infty} \frac{L_i(t)}{t} - \frac{1}{2} \lim_{t \rightarrow \infty} \frac{L_{i-1}(t)}{t}$$

which implies $\lim_{t \rightarrow \infty} \frac{L_{i+1}(t)}{t} = \frac{2(N-i)g}{N}$. The statement in (2.6) now follows by induction. \square

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The Inert Drift Atlas Model

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