# Instantaneous everywhere-blowup of parabolic SPDEs

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#### Abstract

We consider the following stochastic heat equation

$$\partial_t u(t,x) = \frac{1}{2} \partial_x^2 u(t,x) + b(u(t,x)) + \sigma(u(t,x)) \dot{W}(t,x),$$

defined for  $(t, x) \in (0, \infty) \times \mathbb{R}$ , where  $\dot{W}$  denotes space-time white noise. The function  $\sigma$  is assumed to be positive, bounded, globally Lipschitz, and bounded uniformly away from the origin, and the function b is assumed to be positive, locally Lipschitz and nondecreasing. We prove that the Osgood condition

$$\int_1^\infty \frac{\mathrm{d}y}{b(y)} < \infty$$

implies that the solution almost surely blows up everywhere and instantaneously, In other words, the Osgood condition ensures that  $P\{u(t, x) = \infty \text{ for all } t > 0 \text{ and } x \in \mathbb{R}\} = 1$ . The main ingredients of the proof involve a hitting-time bound for a class of differential inequalities (Remark 4.3), and the study of the spatial growth of stochastic convolutions using techniques from the Malliavin calculus and the Poincaré inequalities that were developed in Chen et al [3,4].

*Keywords:* SPDEs, ergodicity, the Malliavin calculus, Poincaré inequalities. *AMS 2010 subject classification:* 60H15; 60H07, 60F05.

### 1 Introduction

We consider the following stochastic heat equation

$$\partial_t u(t,x) = \frac{1}{2} \partial_x^2 u(t,x) + b(u(t,x)) + \sigma(u(t,x)) \dot{W}(t,x) \quad \text{for } (t,x) \in (0,\infty) \times \mathbb{R},$$
  
subject to  $u(0,x) = u_0(x) \qquad \qquad \text{for all } x \in \mathbb{R}.$  (1.1)

The initial condition  $u_0$  is assumed to be a non-random bounded function, and the noise term is space-time white noise; that is,  $\dot{W}$  is a centered, generalized Gaussian random field with

$$\operatorname{Cov}[W(t,x), W(s,y)] = \delta_0(t-s)\delta_0(x-y) \quad \text{for all } t, s \ge 0 \text{ and } x, y \in \mathbb{R}.$$

Throughout, we assume that  $\sigma$  and b satisfy the following hypotheses:

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**Assumption 1.1.**  $\sigma : \mathbb{R} \to (0, \infty)$  is Lipschitz continuous, and satisfies  $0 < \inf_{\mathbb{R}} \sigma \le \sup_{\mathbb{R}} \sigma < \infty$ . **Assumption 1.2.**  $b : \mathbb{R} \to (0, \infty)$  is locally Lipschitz continuous, as well as nondecreasing.

We recall that a random field solution to (1.1) is a predictable random field  $u = \{u(t, x)\}_{t \ge 0, x \in \mathbb{R}}$  that satisfies the following integral equation:

$$u(t,x) = (p_t * u_0)(x) + \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x)b(u(s,y)) \,\mathrm{d}s \,\mathrm{d}y + \mathcal{I}(t,x), \tag{1.2}$$

where

$$\mathcal{I}(t,x) = \int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)\sigma(u(s,y)) W(\mathrm{d} s \,\mathrm{d} y),$$

the symbol \* denotes convolution, and

$$p_r(z) = \frac{\exp\{-z^2/(2r)\}}{\sqrt{2\pi r}} \quad \text{for all } r > 0 \text{ and } z \in \mathbb{R}.$$

When b and  $\sigma$  are Lipschitz continuous, general theory ensures that the SPDE (1.2) is well posed; see Dalang [5] and Walsh [19]. However, general theory fails to be applicable when b and/or  $\sigma$  are assumed to be only locally Lipschitz continuous. Here, we can exploit the fact that b is nondecreasing in order to ensure the existence of a "minimal solution" u under Assumptions 1.1 and 1.2; see the beginning of the proof of Theorem 1.5 in Section 5 for more details. With that under way, we turn to the main objective of this paper and prove that, under Assumptions 1.1 and 1.2, the classical Osgood condition (1.3) of ODEs ensures that the minimal solution, and hence every solution, to (1.1) blows up everywhere and instantaneously.

There is a large and distinguished literature in PDEs that focuses on these types of questions; see for example Cabré and Martel [2], Peral and Vázquez [16], and Vázquez [18]. To the best of our knowledge, the present paper contains the first instantaneous blowup result for SPDEs of the type given by (1.1). For PDEs, various different definitions for instantaneous blowup are used but all these notions basically mean that the solution blows up for every t > 0. We provide a different definition that is particularly well suited for our purposes.

**Definition 1.3.** Let  $u = \{u(t, x)\}_{t \ge 0, x \in \mathbb{R}}$  denote a space-time random field with values in  $[-\infty, \infty]$ . We say that u blows up everywhere and instantaneously when

$$P \{u(t, x) = \infty \text{ for every } t > 0 \text{ and } x \in \mathbb{R} \} = 1.$$

Our notion of instantaneous, everywhere blowup is sometimes referred to as *instantaneous and* complete blowup.

We are not aware of any prior known results on instantaneous nor everywhere blowup in the SPDE literature. However, broader questions of blowup for SPDEs have received recent attention. Recent examples include Ref.s [7,9–11], where criteria for the blowup in finite time with positive probability or almost surely are studied. And De Bouard and Debussche [8] investigate blowup for the stochastic nonlinear Schrödinger equation, valid in arbitrarily small time, and with positive probability; see also the references in [8].

In order to state our result precisely, we need the well-known Osgood condition from the classical theory of ODEs.

**Condition 1.4.** A function  $b : \mathbb{R} \mapsto (0, \infty)$  is said to satisfy the Osgood condition if

$$\int_{1}^{\infty} \frac{\mathrm{d}y}{b(y)} < \infty, \tag{1.3}$$

where  $1/0 = \infty$ .

It was proved in Foondun and Nualart [10] that, when  $\sigma$  is a positive constant, the Osgood condition implies that the solution to (1.1) blows up almost surely. Earlier, this fact was previously proved by Bonder and Groisman [9] for SPDEs on a finite interval. In the converse direction, and for the same equations on finite intervals, Foondun and Nualart [10] have shown that if  $\sigma$  is locally Lipschitz continuous and bounded, then the Osgood condition is necessary for the solution to blow up somewhere with positive probability.

Recall Assumptions 1.1 and 1.2. The aim of the present paper is to show that the Osgood condition in fact implies that, almost surely, the solution to equation (1.1) blows up everywhere and instantaneously.

**Theorem 1.5.** If b satisfies the Osgood Condition 1.4, then the minimal solution to (1.1) blows up everywhere and instantaneously almost surely.

A few years ago, Professor Alison Etheridge asked one of us a number of questions about the time to blow up and the nature of the blowup for stochastic reaction-diffusion equations of the general type studied here. This paper provides the answer to Professor Etheridge's questions in the case that  $\sigma$  satisfies Assumption 1.1. We do not have sharp blowup results when Assumption 1.1 fails. Perhaps a noteworthy example is  $\sigma(u) = u$ , which lies well outside the present theory.

We now describe the main strategy behind the the proof of Theorem 1.5. We may recast (1.2) as

u = Term A + Term B + Term C,

notation being clear. Term A is deterministic, involves the initial condition, and plays no role in the blowup phenomenon because the initial condition is a nice function. In the PDE literature, there are many results about blowup that hold because the initial condition is assumed to be singular. Here, the initial data is a very nice function with no singularities. In our setting, blowup occurs for very different reasons, and is caused by the interplay between the stochastic Term B, which is the highly non-linear term, and the other stochastic Term C, which is regarded as a Walsh stochastic integral. Next, we will say a few words about this interplay.

As part of our analysis, we prove that, when b is in fact a Lipschitz continuous function that satisfies the Osgood condition (1.3), the process  $x \mapsto u(t, x)$  is almost surely unbounded for every t > 0. The proof of this fact makes use of ideas from the Malliavin calculus and Poincaré inequalities developed in a recent paper by Chen et al [4]. The limiting procedure used to define the solution then allows us to use the growth property of b to show blowup of the solution and thus complete the proof of the main result.

We end this introduction with a plan of the paper. In §2 we study ergodicity and growth properties for a family of stochastic convolutions. In §3 we use some of these results to show that, when b is Lipschitz and the initial condition is a constant, the solution to (1.1) is spatially stationary and ergodic. In §4 we develop a hitting-time estimate for a family of differential inequalities and subsequently use that estimate in order to obtain a lower bound for u. The remaining details of the proof of Theorem 1.5 are gathered in §5, using the earlier results of the paper.

Throughout this paper, we write

 $||X||_p = \{ E(|X|^p) \}^{1/p}$  for all  $p \ge 1$  and  $X \in L^p(\Omega)$ .

For every function  $f : \mathbb{R} \to \mathbb{R}$ , Lip(f) denotes the optimal Lipschitz constant of f; that is,

$$\operatorname{Lip}(f) = \sup_{-\infty < a < b < \infty} \frac{|f(b) - f(a)|}{b - a}.$$

In particular, f is Lipschitz continuous iff  $\operatorname{Lip}(f) < \infty$ .

## 2 Spatial growth of stochastic convolutions

### 2.1 Spatial ergodicity via the Malliavin calculus

We introduce following Nualart [15] some elements of the Malliavin calculus that we will need. Let  $\mathcal{H} = L^2(\mathbb{R}_+ \times \mathbb{R})$ . For every Malliavin-differentiable random variable F, we let DF denote the Malliavin derivative of F, and observe that  $DF = \{D_{r,z}F\}_{r>0,z\in\mathbb{R}}$  is a random field indexed by  $(r, z) \in \mathbb{R}_+ \times \mathbb{R}$ .

For every  $p \ge 2$ , let  $\mathbb{D}^{1,p}$  denote the usual Gaussian Sobolev space endowed with the semi-norm

$$||F||_{1,p}^p := \mathrm{E}(|F|^p) + \mathrm{E}(||DF||_{\mathcal{H}}^p).$$

We will need the following version of the Poincaré inequality due to Chen et al [4, (2.1)]:

$$|\operatorname{Cov}(F,G)| \le \int_0^\infty \mathrm{d}r \int_{-\infty}^\infty \mathrm{d}z \ ||D_{r,z}F||_2 ||D_{r,z}G||_2 \quad \text{for every } F, G \text{ in } \mathbb{D}^{1,2}.$$
 (2.1)

Next, let us recall some notions from the ergodic theory of multiparameter processes (see for example Chen et al [3]): We say that a predictable random field  $Z = \{Z(t,x)\}_{(t,x)\in(0,\infty)\times\mathbb{R}}$  is spatially mixing when the random field  $x \to Z(t,x)$  is weakly mixing in the usual sense for every t > 0. This property can be stated as follows: For all  $k \in \mathbb{N}$ , t > 0,  $\xi^1, ..., \xi^k \in \mathbb{R}$ , and Lipschitz-continuous functions  $g_1, ..., g_k : \mathbb{R} \to \mathbb{R}$  that satisfy  $g_j(0) = 0$  and  $\operatorname{Lip}(g_j) = 1$  for every j = 1, ..., k,

$$\lim_{|x| \to \infty} \operatorname{Cov}[\mathcal{G}(x), \mathcal{G}(0)] = 0, \qquad (2.2)$$

where

$$\mathcal{G}(x) = \prod_{j=1}^{k} g_j(Z(t, x + \xi^j)), \quad x \in \mathbb{R}.$$
(2.3)

Whenever the process  $x \to Z(t, x)$  is stationary and weakly mixing for all t > 0, it is ergodic.

Finally, we will require two elementary identities for products of the heat kernel. Namely, that

$$p_{t-s}(x-y)p_s(y-z) = p_t(x-z)p_{s(t-s)/t}\left(y-z-\frac{s}{t}(x-z)\right),$$
(2.4)

and

$$\int_{-\infty}^{\infty} \left[ p_{t-s}(x-y) \right]^2 \left[ p_{s-r}(y-z) \right]^2 \, \mathrm{d}y = \sqrt{\frac{t-r}{4\pi(t-s)(s-r)}} \left[ p_{t-r}(x-z) \right]^2.$$
(2.5)

See Chen et al [3, below (6.10)] for (2.4) and Chen et al [4, below (2.7)] for (2.5).

#### 2.2 Ergodicity of stochastic convolutions

Let  $Z = \{Z(t, x)\}_{(t,x) \in (0,\infty) \times \mathbb{R}}$  be a predictable random field that satisfies

$$c_1 \le \inf_{(t,x)\in(0,\infty)\times\mathbb{R}} Z(t,x) \le \sup_{(t,x)\in(0,\infty)\times\mathbb{R}} Z(t,x) \le c_2,$$
(2.6)

for two positive and finite constants  $c_1$  and  $c_2$  that are fixed throughout. Set  $I_Z(0, x) = 0$ , and consider the associated stochastic convolution

$$I_Z(t,x) = \int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)Z(s,y)W(\mathrm{d} s\,\mathrm{d} y) \qquad \text{for every } t>0 \text{ and } x\in\mathbb{R}.$$
 (2.7)

The main aim of this section is to study the growth properties of the random field  $x \to I_Z(t, x)$ . Next we develop natural conditions under which the random field  $x \to I_Z(t, x)$  is stationary and ergodic at all times t > 0. **Proposition 2.1.** Assume that  $x \to Z(t, x)$  is stationary for all t > 0. Assume also that  $Z(t, x) \in \mathbb{D}^{1,p}$  for all  $p \ge 2$ , t > 0 and  $x \in \mathbb{R}$ , and that its Malliavin derivative DZ(t, x) has the following property: For every T > 0 and  $p \ge 2$  there exists a number  $C_{T,p} > 0$  such that

$$||D_{r,z}Z(t,x)||_p \le C_{T,p} \, p_{t-r}(x-z)p_r(z), \tag{2.8}$$

for every  $t \in (0,T)$  and  $x \in \mathbb{R}$  and for almost every  $(r,z) \in (0,t) \times \mathbb{R}$ . Then the process  $x \to Z(t,x)$  is ergodic for every t > 0, and  $x \to I_Z(t,x)$  is stationary and ergodic for every t > 0.

*Proof.* Thanks to the Poincaré inequality (2.1), the proof of ergodicity follows the same pattern as [3, Proof of Theorem 1.3]. Therefore, we describe the argument quickly mainly where adjustments are needed.

We start with the process Z and use a similar argument as Chen et al [3, Proof of Corollary 9.1]; see also Chen et al [4, Theorem 1.1]. Define  $\mathcal{G}(x)$  as was done in (2.3). It then follows from (2.8) and (2.4) that there exists a constant  $c_{T,k} > 0$  such that

$$\begin{split} \|D_{r,z}\mathcal{G}(x)\|_{2} &\leq \sum_{j_{0}=1}^{k} \left(\prod_{j=1, j\neq j_{0}}^{k} \|g_{j}(Z(t, x+\xi^{j}))\|_{2k}\right) \|D_{r,z}Z(t, x+\xi^{j_{0}})\|_{2k} \\ &\leq c_{T,k} \sum_{j=1}^{k} p_{t-r}(x+\xi^{j}-z)p_{r}(z) \\ &\leq c_{T,k} \sum_{j=1}^{k} p_{t}(x+\xi^{j})p_{r(t-r)/t}\left(z-\frac{r}{t}(x+\xi^{j})\right), \end{split}$$

valid uniformly for all  $0 < r < t \leq T$  and  $x, z \in \mathbb{R}^{1}$ .

We can combine the Poincaré inequality (2.1), the heat-kernel identity (2.4), and the semigroup property of the heat kernel to find that

$$|\operatorname{Cov}[\mathcal{G}(x), \mathcal{G}(0)]| \le c_{T,k} \sum_{j,\ell=1}^{k} p_t(x+\xi^j) p_t(x+\xi^\ell) \int_0^t p_{2r(t-r)/t} \left(\frac{r}{t}(x+\xi^j-\xi^\ell)\right) \,\mathrm{d}r.$$

Therefore, the dominated convergence implies (2.2), whence follows the ergodicity of  $x \to Z(t, x)$  for every t > 0.

Next, we show that the process  $x \to I_Z(t, x)$  is stationary for all t > 0. The proof of this fact follows the proof of Lemma 7.1 in [3] closely. First, let us choose and fix some  $y \in \mathbb{R}$  and apply (7.2) in [3] as follows:

$$(I_Z \circ \theta_y)(t, x) = I_Z(t, x+y) = \int_{(0,t)\times\mathbb{R}} p_{t-s}(x+y-z)Z(s, z-y+y) W(\mathrm{d}s \,\mathrm{d}z)$$
$$= \int_{(0,t)\times\mathbb{R}} p_{t-s}(x-z)Z(s, z+y) W_y(\mathrm{d}s \,\mathrm{d}z)$$
$$= \int_{(0,t)\times\mathbb{R}} p_{t-s}(x-z)(Z \circ \theta_y)(s, z) W_y(\mathrm{d}s \,\mathrm{d}z),$$

where  $\theta_y$  denotes the shift operator (see Chen et al [3]), and  $W_y$  is the associated shifted Gaussian noise [3, (7.1)]. The spatial stationarity of  $I_Z$  follows from the facts that W and  $W_y$  have the same

<sup>&</sup>lt;sup>1</sup>The notation  $c_{t,k}$  may refer to a constant that changes from line to line but in any case depends only on (t,k).

law and the random field  $Z \circ \theta_y$  has the same finite-dimensional distributions as Z because Z is assumed to be spatially stationary.

We now turn to the spatial ergodicity of the process  $I_Z$ . By the properties of the divergence operator [15, Proposition 1.3.8],  $I_Z(t,x) \in \mathbb{D}^{1,k}$  for all  $k \geq 2, t > 0$ , and  $x \in \mathbb{R}$ . Moreover, the Malliavin derivative  $DI_Z(t,x)$  a.s. satisfies

$$D_{r,z}I_Z(t,x) = p_{t-r}(x-z)Z(r,z) + \int_{(r,t)\times\mathbb{R}} p_{t-s}(y-x)D_{r,z}Z(s,y)W(\mathrm{d} s\,\mathrm{d} y).$$

In principle, the above is valid for a.e. (r, z) but in fact the right-hand side can be used to define the Malliavin derivative everywhere a.s. And that is what we do here. In particular, for any integer  $k \ge 2$ , the Burkholder-Davis-Gundy inequality and the estimate (2.8) together imply that

$$\begin{split} \|D_{r,z}I_{Z}(t,x)\|_{2k} &\leq cp_{t-r}(x-z) + c_{k} \left(\int_{r}^{t} \mathrm{d}s \int_{\mathbb{R}} \mathrm{d}y \left[p_{t-s}(x-y)\right]^{2} \|D_{r,z}Z(s,y)\|_{2k}^{2}\right)^{1/2} \\ &\leq cp_{t-r}(x-z) + c_{T,k} \left(\int_{r}^{t} \mathrm{d}s \int_{\mathbb{R}} \mathrm{d}y \left[p_{t-s}(x-y)\right]^{2} \left[p_{s-r}(y-z)\right]^{2} \left[p_{r}(z)\right]^{2}\right)^{1/2}. \end{split}$$

Thanks to (2.5), this yields

$$\|D_{r,z}I_Z(t,x)\|_{2k} \le cp_{t-r}(x-z) + c_{T,k}p_r(z)p_{t-r}(x-z) \left(\int_r^t \sqrt{\frac{t-r}{4\pi(t-s)(s-r)}} \,\mathrm{d}s\right)^{1/2} \le c_{T,k}p_{t-r}(x-z)(1+p_r(z)(t-r)^{1/4}).$$
(2.9)

Define

$$\mathcal{J}(x) = \prod_{j=1}^{k} g_j(I_Z(t, x + \xi^j)) \quad \text{for } x \in \mathbb{R},$$

using the same  $g^1, \ldots, g^k$  and  $\xi^1, \ldots, \xi^k$  that were introduced earlier. In this way we can conclude from (2.9) and elementary properties of the Malliavin derivative that

$$\begin{split} \|D_{r,z}\mathcal{J}(x)\|_{2} &\leq \sum_{j_{0}=1}^{k} \left(\prod_{j=1, j\neq j_{0}}^{k} \|g_{j}(I_{Z}(t, x+\xi^{j}))\|_{2k}\right) \|D_{r,z}I_{Z}(t, x+\xi^{j_{0}})\|_{2k} \\ &\leq c_{T,k} \sum_{j=1}^{k} p_{t-r}(x+\xi^{j}-z)(1+p_{r}(z)(t-r)^{1/4}) \\ &= c_{T,k} \sum_{j=1}^{k} \left[p_{t}(x+\xi^{j}-z) + p_{r(t-r)/t}\left(z-\frac{r}{t}(x+\xi^{j})\right)p_{t}(x+\xi^{j})(t-r)^{1/4}\right] \end{split}$$

valid uniformly for all  $0 < r < t \leq T$  and  $x, z \in \mathbb{R}$ .

Now we apply (2.1) together with the semigroup property of the heat kernel to see that

$$\begin{aligned} |\operatorname{Cov}[\mathcal{J}(x),\mathcal{J}(0)]| &\leq c_{T,k} \sum_{j,\ell=1}^{k} \left[ t p_{2t}(x+\xi^{j}-\xi^{\ell}) + \int_{0}^{t} p_{t+\frac{r(t-r)}{t}} \left(x+\xi^{j}-\frac{r}{t}\xi^{\ell}\right) p_{t}(\xi^{\ell})(t-r)^{1/4} \, \mathrm{d}r + \int_{0}^{t} p_{t+\frac{r(t-r)}{t}} \left(\frac{r}{t}(x+\xi^{j})-\xi^{\ell}\right) p_{t}(x+\xi^{j})(t-r)^{1/4} \, \mathrm{d}r + p_{t}(x+\xi^{j})p_{t}(x+\xi^{\ell}) \int_{0}^{t} p_{2r(t-r)/t} \left(\frac{r}{t}(x+\xi^{j}-\xi^{\ell})\right) (t-r)^{1/4} \, \mathrm{d}r \right]. \end{aligned}$$

The dominated convergence implies that  $\lim_{|x|\to\infty} \operatorname{Cov}[\mathcal{J}(x), \mathcal{J}(0)] = 0$ , and hence follows the ergodicity of  $x \to I_Z(t, x)$  for every t > 0. This concludes the proof.

#### 2.3 Spatial growth of stochastic convolutions

We are ready to state the main result of this section.

**Theorem 2.2.** Choose and fix  $c_2 > c_1 > 0$ . Then, there exists  $\eta = \eta(c_1, c_2) > 0$  such that

$$\mathbf{P}\left\{\limsup_{c\to\infty}\inf_{t\in(a,a+(\eta a)^2)}\inf_{x\in(0,\eta a)}I_Z(t,c+x)=\infty\right\}=1,$$

valid for every non-random number a > 0 and every predictable random field Z that satisfies the boundedness condition (2.6) and for which  $x \mapsto I_Z(t, x)$  is stationary and ergodic for all t > 0.

**Remark 2.3.** Note, in particular, that the constant  $\eta$  does not depend on the choice of Z. This is the crucial part of the message of Theorem 2.2.

The proof of Theorem 2.2 requires a few prefatory steps that we present as a series of lemmas. Once those lemmas are under way, we are able to prove Theorem 2.2 promptly.

**Lemma 2.4.** For every  $c_2 > c_1 > 0$  there exist  $C_2, C_1 > 0$  such that

$$\frac{C_1}{1+\lambda} \exp\left(-\frac{\lambda^2}{2c_1^2}\right) \le \mathbf{P}\left\{I_Z(t,x) \ge (t/\pi)^{1/4}\lambda\right\} \le \frac{C_2}{1+\lambda} \exp\left(-\frac{\lambda^2}{2c_2^2}\right),$$

uniformly for all  $t, \lambda \ge 0$  and  $x \in \mathbb{R}$ , and for every predictable random field Z that satisfies (2.6).

*Proof.* Choose and fix t > 0 and consider

$$M_0 = 0$$
 and  $M_r = \int_{(0,r) \times \mathbb{R}} p_{t-s}(y-x)Z(s,y) W(\mathrm{d} s \,\mathrm{d} y)$  for  $0 < r \le t$ .

Because Z is uniformly bounded, the above is a continuous,  $L^2$ -martingale with quadratic variation

$$\langle M \rangle_r = \int_0^r \mathrm{d}s \int_{-\infty}^\infty \mathrm{d}y \ [p_{t-s}(y-x)]^2 |Z(s,y)|^2 \quad \text{for } 0 \le r \le t.$$

Because

$$\int_0^r \mathrm{d}s \int_{-\infty}^\infty \mathrm{d}y \ [p_{t-s}(y-x)]^2 = \int_0^r \frac{\mathrm{d}s}{\sqrt{4\pi(t-s)}} = \sqrt{\frac{t}{\pi}} - \sqrt{\frac{t-r}{\pi}} \qquad \text{for } 0 \le r \le t,$$

the inequalities (2.6) yield

$$\frac{c_1^2}{\sqrt{\pi}} \left[ \sqrt{t} - \sqrt{t-r} \right] \le \langle M \rangle_r \le \frac{c_2^2}{\sqrt{\pi}} \left[ \sqrt{t} - \sqrt{t-r} \right] \quad \text{for } 0 \le r \le t.$$
(2.10)

The Dubins, Dambis-Schwartz theorem, see [17], ensures that  $M_r = B(\langle M \rangle_r)$  for a standard, linear Brownian motion *B*. Since  $I_Z(t, x) = M_t$  is the terminal point of our martingale *M*, and because (2.10) implies that  $\langle M \rangle_t \leq c_2^2 \sqrt{t/\pi}$ , we learn from the reflection principle and the scaling property that

$$P\left\{I_Z(t,x) \ge c_2(t/\pi)^{1/4}\lambda\right\} \le P\left\{\sup_{0 \le r \le c_2^2\sqrt{t/\pi}} B(r) \ge c_2(t/\pi)^{1/4}\lambda\right\} = \sqrt{2/\pi} \int_{\lambda}^{\infty} e^{-z^2/2} dz.$$

A standard estimate yields the upper bound. For the lower bound we observe in like manner to the preceding that

$$P\left\{ I_{Z}(t,x) \ge c_{1}(t/\pi)^{1/4} \lambda \right\}$$

$$\ge P\left\{ B\left(c_{1}^{2}\sqrt{t/\pi}\right) \ge 2c_{1}(t/\pi)^{1/4} \lambda \right\} P\left\{ \sup_{\nu \in [c_{1}^{2},c_{2}^{2}]} \left| B\left(\nu\sqrt{t/\pi}\right) - B\left(c_{1}^{2}\sqrt{t/\pi}\right) \right| \le c_{1}(t/\pi)^{1/4} \right\}$$

$$= \frac{\varpi}{\sqrt{2\pi}} \int_{2\lambda}^{\infty} e^{-z^{2}/2} dz,$$

where  $\varpi = P\{\sup_{\nu \in [1, (c_2/c_1)^2]} | B(\nu) - B(1)| \le 1\} \in (0, 1)$ . This proves that

$$\mathsf{P}\left\{I_Z(t,x) \ge c_1(t/\pi)^{1/4}\lambda\right\} \gtrsim \lambda^{-1}\exp(-\lambda^2/2) \quad \text{for all } \lambda \ge 1,$$

where the implied constant depends only on  $c_1$  and  $c_2$ . When  $\lambda \in (0, 1)$ , it suffices to lower bound the integral by a constant.

**Lemma 2.5.** Choose and fix a non-random number  $c_0 > 0$ . Then,

$$\sup_{t \ge 0} \sup_{-\infty < x \ne z < \infty} \mathbf{E}\left(\left|\frac{I_Z(t, x) - I_Z(t, z)}{|x - z|^{1/2}}\right|^k\right) \le (2c_0^2 k)^{k/2},$$

for every  $k \in [2,\infty)$  and for all predictable random fields Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \leq c_0$ .

**Remark 2.6.** We emphasize that Lemma 2.5 assumes that Z is bounded. This is a much weaker condition than (2.6), as the latter implies also that, among other things,  $\inf_{p \in \mathbb{R}_+ \times \mathbb{R}} Z(p)$  is a.s. bounded from below by a strictly positive, deterministic number. The next lemmas also in fact require only this weaker boundedness condition.

*Proof.* Choose and fix  $t \ge 0$  and  $x \ne z \in \mathbb{R}$ , and let Z be as described. By the Burkholder-Davis-Gundy inequality in the form [6], for every real number  $k \ge 2$ ,

$$\begin{split} \|I_{Z}(t,x) - I_{Z}(t,z)\|_{k}^{2} &\leq 4k \int_{0}^{t} \mathrm{d}s \int_{-\infty}^{\infty} \mathrm{d}y \; [p_{t-s}(y-x) - p_{t-s}(y-z)]^{2} \|Z(s,y)\|_{k}^{2} \\ &\leq 4c_{0}^{2}k \int_{0}^{\infty} \mathrm{d}s \int_{-\infty}^{\infty} \mathrm{d}y \; [p_{s}(y-x+z) - p_{s}(y)]^{2} \\ &= \frac{2c_{0}^{2}k}{\pi} \int_{0}^{\infty} \mathrm{d}s \int_{-\infty}^{\infty} \mathrm{d}\xi \; \mathrm{e}^{-s\xi^{2}} \left|1 - \mathrm{e}^{-i\xi(x-z)/2}\right|^{2} \quad \text{[Plancherel's theorem]} \\ &= \frac{8c_{0}^{2}k}{\pi} \int_{0}^{\infty} \frac{1 - \cos(|x-z|\xi/2)}{\xi^{2}} \; \mathrm{d}\xi = 2c_{0}^{2}k|x-z|. \end{split}$$

This proves the lemma.

**Lemma 2.7.** Choose and fix a non-random number  $c_0 > 0$ . Then,

$$\sup_{t,h>0} \sup_{x\in\mathbb{R}} \operatorname{E}\left(\left|\frac{I_Z(t+h,x) - I_Z(t,x)}{h^{1/4}}\right|^k\right) \le (5c_0^2k)^{k/2},$$

for every  $k \in [2,\infty)$  and for all predictable random fields Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \le c_0$ .

*Proof.* Choose and fix t, h > 0 and  $x \in \mathbb{R}$ , and a predictable random field Z as above, and then write

$$||I_Z(t+h, x) - I_Z(t, x)||_k \le T_1 + T_2,$$

where

$$T_{1} = \left\| \int_{(0,t)\times\mathbb{R}} \left[ p_{t+h-s}(y-x) - p_{t-s}(y-x) \right] Z(s,y) W(\mathrm{d}s \,\mathrm{d}y) \right\|_{k},$$
  
$$T_{2} = \left\| \int_{(t,t+h)\times\mathbb{R}} p_{t+h-s}(y-x) Z(s,y) W(\mathrm{d}s \,\mathrm{d}y) \right\|_{k}.$$

By the Burkholder-Davis-Gundy inequality in the form [6], for every real number  $k \ge 2$ ,

$$T_{1}^{2} \leq 4k \int_{0}^{t} ds \int_{-\infty}^{\infty} dy \left[ p_{t+h-s}(y-x) - p_{t-s}(y-x) \right]^{2} \|Z(s,y)\|_{k}^{2}$$

$$\leq 4c_{0}^{2}k \int_{0}^{\infty} ds \int_{-\infty}^{\infty} dy \left[ p_{s+h}(y) - p_{s}(y) \right]^{2}$$

$$= \frac{2c_{0}^{2}k}{\pi} \int_{0}^{\infty} ds \int_{-\infty}^{\infty} d\xi \, e^{-s\xi^{2}} \left| 1 - e^{-h\xi^{2}/2} \right|^{2} \qquad \text{[Plancherel's theorem]}$$

$$= \frac{2\sqrt{2}c_{0}^{2}k}{\pi} \int_{0}^{\infty} \frac{|1 - \exp(-y^{2})|^{2}}{y^{2}} \, dy \sqrt{h} \leq \frac{2\sqrt{2}c_{0}^{2}k}{\pi} \left( \frac{1}{3} + \int_{1}^{\infty} \frac{dy}{y^{2}} \right) \sqrt{h} = \frac{8\sqrt{2}c_{0}^{2}k}{3\pi} \sqrt{h},$$

where we have used the bound  $1 - \exp(-y^2) \le y^2 \land 1$  in order to obtain the last concrete numerical estimate. Similarly, we obtain

$$T_2^2 \le 4k \int_t^{t+h} \mathrm{d}s \int_{-\infty}^{\infty} \mathrm{d}y \ [p_{t+h-s}(y-x)]^2 ||Z(s,y)||_k^2$$
$$\le 4c_0^2 k \int_0^h \mathrm{d}s \int_{-\infty}^{\infty} \mathrm{d}y \ [p_{s+h}(y)]^2 = \frac{2c_0^2 k}{\pi} \int_h^{2h} \mathrm{d}s \int_{-\infty}^{\infty} \mathrm{d}\xi \ \mathrm{e}^{-s\xi^2}$$
$$= \frac{2c_0^2 k}{\sqrt{\pi}} \int_h^{2h} \frac{\mathrm{d}s}{\sqrt{s}} = \frac{4(\sqrt{2}-1)c_0^2 k}{\sqrt{\pi}} \sqrt{h}.$$

We finally get

$$\|I_Z(t+h,x) - I_Z(t,x)\|_k \le c_0 \sqrt{k} \left[ \sqrt{\frac{8\sqrt{2}}{3\pi}} + \sqrt{\frac{4(\sqrt{2}-1)}{\sqrt{\pi}}} \right] h^{1/4} \le 2.1 c_0 \sqrt{k} h^{1/4},$$

and complete the proof, with room to spare for the constants of the inequality.

Define

$$\varrho(p) = |p_1|^{1/4} + |p_2|^{1/2} \quad \text{for all } p = (p_1, p_2) \in \mathbb{R}^2,$$

and for convenience, we use the following notation,  $I_Z(p) := I_Z(p_1, p_2)$ .

**Lemma 2.8.** For every non-random numbers  $c_0, m > 0$  and  $\delta \in (0, 1)$ ,

$$\sup_{Z,\mathbb{I}} \operatorname{E} \exp\left( \alpha \sup_{\substack{p,q \in [0,1] \times \mathbb{I} \\ 0 < \varrho(p-q) \le 1}} \left| \frac{I_Z(p) - I_Z(q)}{[\varrho(p-q)]^{1-\delta}} \right|^2 \right) < \infty$$

where  $\sup_{Z,\mathbb{I}}$  denotes the supremum over all predictable random fields Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \le c_0$  and over all intervals  $\mathbb{I} \subset \mathbb{R}$  that have length  $\le m$ , and  $\alpha$  is any positive number that satisfies

$$\alpha < \frac{(1 - 2^{-\delta/2})^2}{2^{25} e c_0^2}.$$

*Proof.* Since  $(a + b)^k \leq 2^k(a^k + b^k)$  for all  $k \geq 1$  and  $a, b \geq 0$ , Lemmas 2.5 and 2.7 together and Jensen's inequality imply that

$$\mathbb{E}\left(\left|\frac{I_{Z}(p) - I_{Z}(q)}{\varrho(p-q)}\right|^{k}\right) \leq \left\{ \mathbb{E}\left(\left|\frac{I_{Z}(p) - I_{Z}(q)}{\varrho(p-q)}\right|^{2k}\right) \right\}^{1/2} \\ \leq c_{0}^{k} 2^{k} (4^{k/2} + 10^{k/2}) k^{k/2} \leq (13c_{0})^{k} k^{k/2},$$
(2.11)

valid uniformly for all real numbers  $k \geq 1$ , distinct  $p, q \in \mathbb{R}_+ \times \mathbb{R}$ , and predictable Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \leq c_0$ .

We are going to use a suitable form of Garsia's lemma [13, Appendix C], and will begin by verifying the conditions that can be found in that reference. Note that  $\rho(0) = 0$  and  $\rho$  is subadditive:  $\rho(p+q) \leq \rho(p) + \rho(q)$  for all  $p, q \in \mathbb{R}^d$ . We use the notation of [13, Appendix C] and let

$$\mathcal{B}_{\varrho}(s) = \left\{ y \in \mathbb{R}^2 : \, \varrho(y) \le s \right\} \qquad \text{for all } s \ge 0,$$

and for all real numbers  $k \geq 1$ ,

$$\mathcal{I}_{k} = \int_{[0,1]\times\mathbb{I}} \mathrm{d}p \int_{[0,1]\times\mathbb{I}} \mathrm{d}q \, \left| \frac{I_{Z}(p) - I_{Z}(q)}{\varrho(p-q)} \right|^{k}$$

We know that  $\mathcal{I}_k < \infty$  a.s. for every  $k \ge 1$ . In fact, (2.11) ensures that

$$E(\mathcal{I}_k) \le m^2 (13c_0)^k k^{k/2},$$
(2.12)

uniformly for all real numbers  $k \geq 1$ , distinct  $p, q \in \mathbb{R}_+ \times \mathbb{R}$ , and predictable Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \leq c_0$ . If  $(s, y) \in \mathbb{R}_+ \times \mathbb{R}^2$  satisfies  $|y_1| \leq (s/2)^4$  and  $|y_2| \leq (s/2)^2$  then certainly  $y \in B_{\varrho}(s)$ . Similarly, if  $y \in B_{\varrho}(s)$ , then certainly  $|y_1| \leq s^4$  and  $|y_2| \leq s^2$ . This argument shows that  $(s/2)^6 \leq |B_{\varrho}(s)| \leq 2s^6$  for all  $s \geq 0$ , where  $|\cdots|$  denotes the Lebesgue measure on  $\mathbb{R}^2$ . Consequently,  $\int_0^{r_0} |B_{\varrho}(s)|^{-2/k} ds < \infty$  for one, hence all,  $r_0 > 0$ , if and only if k > 12 and

$$\int_{0}^{r_{0}} \frac{\mathrm{d}s}{|B_{\varrho}(s)|^{2/k}} \leq 2^{12/k} \int_{0}^{r_{0}} s^{-12/k} \,\mathrm{d}s \leq \frac{2kr_{0}^{(k-12)/k}}{k-12} \qquad \text{for every } r_{0} > 0 \text{ and } k > 12$$
$$\leq 4r_{0}^{(k-12)/k} \qquad \text{for every } r_{0} > 0 \text{ and } k \geq 24.$$

Apply Theorem C.4 of [13] with  $\mu(z) = z$  - so that  $C_{\mu} = 2$  there – in order to see that

$$\sup_{\substack{p,q \in [0,r] \times \mathbb{I}\\ \varrho(p-q) \le r_0}} |I_Z(p) - I_Z(q)| \le 32\mathcal{I}_k^{1/k} \int_0^{r_0} \frac{\mathrm{d}s}{|B_\varrho(s)|^{2/k}} \le 128\mathcal{I}_k^{1/k} r_0^{(k-12)/k} \qquad \text{a.s.},$$

for every nonrandom  $k \ge 24$  and  $r_0 > 0$ . In particular, we learn from (2.12) that

$$\mathbf{E}\left(\sup_{\substack{p,q\in[0,1]\times\mathbb{I}\\\varrho(p-q)\leq r_0}}|I_Z(p)-I_Z(q)|^k\right)\leq 128^k r_0^{k-12}\mathbf{E}(\mathcal{I}_k)\leq m^2(1664c_0)^k r_0^{k-12}k^{k/2},$$

for every  $k \geq 24$  and  $r_0 > 0$ , as well as all r > 0, all intervals  $\mathbb{I}$  of length m, and all predictable fields Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \leq c_0$ . We freeze all variables and define for every  $\delta \in (0, 1)$ and  $n \in \mathbb{Z}_+$ ,

$$S_{n,\delta} = \left\{ E\left( \sup_{\substack{p,q \in [0,1] \times \mathbb{I} \\ 2^{-n-1} < \varrho(p-q) \le 2^{-n}}} \left| \frac{I_Z(p) - I_Z(q)}{[\varrho(p-q)]^{1-\delta}} \right|^k \right) \right\}^{1/k}$$

It follows that as long as  $k \ge 24$ ,

$$S_{n,\delta} \le 2^{(1-\delta)(n+1)} \left\{ E \left( \sup_{\substack{p,q \in [0,1] \times \mathbb{I}\\ \varrho(p-q) \le 2^{-n}}} |I_Z(p) - I_Z(q)|^k \right) \right\}^{1/k} \le 2^{12-\delta} c_0 m^{2/k} 2^{-n[\delta - (12/k)]} \sqrt{k}.$$

Sum the preceding over all  $n \in \mathbb{Z}_+$  to see that, as long as  $k \ge (24/\delta) > (12/\delta) \lor 24$ ,

$$\left\{ \mathbf{E} \left( \sup_{\substack{p,q \in [0,1] \times \mathbb{I} \\ \varrho(p-q) \le 1}} \left| \frac{I_Z(p) - I_Z(q)}{[\varrho(p-q)]^{1-\delta}} \right|^k \right) \right\}^{1/k} \le \frac{2^{12-\delta} c_0 m^{2/k} \sqrt{k}}{1 - 2^{-[\delta - (12/k)]}} \le \frac{2^{12}}{1 - 2^{-\delta/2}} c_0 m^{2/k} \sqrt{k}.$$

Replace k by 2k and restrict attention to integral choices of k in order to see that

$$\mathbb{E}\left(\sup_{\substack{p,q\in[0,1]\times\mathbb{I}\\\varrho(p-q)\leq 1}} \left|\frac{I_Z(p) - I_Z(q)}{[\varrho(p-q)]^{1-\delta}}\right|^{2k}\right) \leq m^2 \left(\frac{2^{25/2}\sqrt{e}\,c_0}{1-2^{-\delta/2}}\right)^{2k} k! =: m^2 Q^k k!,$$

for every integer  $k \ge 12/\delta$ , as well as all r > 0, all intervals  $\mathbb{I}$  of length m, and all predictable fields Z that satisfy  $\sup_{p \in \mathbb{R}_+ \times \mathbb{R}} |Z(p)| \le c_0$ , where where we have used the inequality  $k^k \le e^k k!$  valid for all positive integers k. An appeal to the Taylor series expansion of the exponential function  $v \mapsto \exp(\alpha v^2)$  yields

$$\operatorname{E}\exp\left(\alpha\sup_{\substack{p,q\in[0,1]\times\mathbb{I}\\\varrho(p-q)\leq 1}}\left|\frac{I_Z(p)-I_Z(q)}{[\varrho(p-q)]^{1-\delta}}\right|^2\right)\leq\frac{m^2}{1-\alpha Q}<\infty,$$

for every  $\alpha$  that satisfies  $\alpha < Q^{-1}$ . This proves the lemma.

We are ready to conclude this section.

Proof of Theorem 2.2. Lemma 2.4 ensures that

$$P\left\{I_Z(a,c) > M\left(\frac{a}{\pi}\right)^{1/4}\right\} \ge \frac{C_1 e^{-M^2/(2c_1^2)}}{1+M},$$

uniformly for all  $a > 0, c \in \mathbb{R}$ , and  $M \ge 1$ . In particular,

$$\mathbb{P}\left\{\inf_{t\in(a,a+\varepsilon^4)}\inf_{x\in(c,c+\varepsilon^2)}I_Z(t,x)\leq M\left(\frac{a}{\pi}\right)^{1/4}\right\} \leq 1-\frac{C_1\mathrm{e}^{-(2M)^2/(2c_1^2)}}{1+2M}+\mathbb{P}\left\{\sup_{t\in(a,a+\varepsilon^4)}\sup_{x\in(c,c+\varepsilon^2)}\left|I_Z(t,x)-I_Z(a,c)\right|\geq M\left(\frac{a}{\pi}\right)^{1/4}\right\}.$$

Chebyshev's inequality yields the following:

$$\begin{aligned} & \operatorname{P}\left\{\sup_{t\in(a,a+\varepsilon^{4})}\sup_{x\in(c,c+\varepsilon^{2})}\left|I_{Z}(t,x)-I_{Z}(a,c)\right| \geq M\left(\frac{a}{\pi}\right)^{1/4}\right\} \\ & \leq \operatorname{P}\left\{\sup_{t\in(a,a+\varepsilon^{4})}\sup_{x\in(c,c+\varepsilon^{2})}\left|\frac{I_{Z}(t,x)-I_{Z}(a,c)}{\sqrt{\varrho\left((t,x)-(a,c)\right)}}\right| \geq \frac{M(a/\pi)^{1/4}}{\sqrt{2\varepsilon}}\right\} \\ & \leq \operatorname{E}\exp\left(\alpha\sup_{t\in(a,a+\varepsilon^{4})}\sup_{x\in(c,c+\varepsilon^{2})}\left|\frac{I_{Z}(t,x)-I_{Z}(a,c)}{\sqrt{\varrho((t,x)-(a,c))}}\right|^{2}\right) \times \exp\left(-\frac{\alpha M^{2}\sqrt{a/\pi}}{2\varepsilon}\right),
\end{aligned}$$

uniformly for all  $M \geq 1$  and  $a, c, \varepsilon, \alpha > 0$ . Choose and fix

$$\alpha = \frac{(1 - 2^{-1/4})^2}{2^{26} e(c_1 \vee c_2)^2} \quad \text{and} \quad \varepsilon = \frac{c_1^2 \alpha}{8} \sqrt{\frac{a}{\pi}}.$$
(2.13)

and apply Lemma 2.8 [with  $\delta = \frac{1}{2}$  and  $c_0 = c_1 \lor c_2$ ] in order to see that there exists  $K = K(c_1, c_2) > 1$  such that

$$\mathbb{P}\left\{ \inf_{t \in (a,a+\varepsilon^4)} \inf_{x \in (c,c+\varepsilon^2)} I_Z(t,x) \le M\left(\frac{a}{\pi}\right)^{1/4} \right\} \le 1 - \frac{C_1 e^{-(2M)^2/(2c_1^2)}}{1+2M} + K e^{-(2M)^2/c_1^2} \\ \le 1 - e^{-(2M)^2/(2c_1^2)} \left[\frac{C_1}{3M} - K e^{-(2M)^2/(2c_1^2)}\right],$$

uniformly for all  $M \ge 1$  and a > 0. In particular, there exists  $M_0 = M_0(c_1, c_2) > 1$  such that for all  $M \ge 1$  and a > 0,

$$\sup_{a,c>0} \mathbb{P}\left\{\inf_{t\in(a,a+\varepsilon^4)} \inf_{x\in(c,c+\varepsilon^2)} I_Z(t,x) \le M\left(\frac{a}{\pi}\right)^{1/4}\right\} \le 1 - \frac{C_1 \mathrm{e}^{-(2M)^2/(2c_1^2)}}{6M}$$

uniformly for all  $M \ge M_0$ . To be sure, we remind also that  $\varepsilon = \varepsilon(a, c_1, c_2)$  is defined in (2.13). In any case, this readily yields

$$\inf_{a>0} \mathbb{P}\left\{\limsup_{c\to\infty} \inf_{t\in(a,a+\varepsilon^4)} \inf_{x\in(c,c+\varepsilon^2)} I_Z(t\,,x) > M\left(\frac{a}{\pi}\right)^{1/4}\right\} \ge \frac{C_1 \mathrm{e}^{-(2M)^2/(2c_1^2)}}{6M} > 0,\tag{2.14}$$

uniformly for all  $M \ge M_0$ . Since we are assuming that the infinite-dimensional process  $x \mapsto I_Z(\cdot, x)$  is ergodic, we can improve (2.14) to the following without need for additional work:

$$P\left\{\limsup_{c \to \infty} \inf_{t \in (a, a + \varepsilon^4)} \inf_{x \in (c, c + \varepsilon^2)} I_Z(t, x) > M\left(\frac{a}{\pi}\right)^{1/4}\right\} = 1,$$

uniformly for all  $M \ge M_0$  and a > 0. We now can send  $M \to \infty$  to deduce the theorem from the particular form of  $\varepsilon$  that is given in (2.13).

## 3 Ergodicity of the solution

In this section, we consider equation (1.1) with constant initial condition  $\rho \in \mathbb{R}$ . That is,

$$u(t,x) = \rho + \int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)b(u(s,y)) \,\mathrm{d}s \,\mathrm{d}y + \mathcal{I}(t,x), \tag{3.1}$$

where

$$\mathcal{I}(t,x) = \int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)\sigma(u(s,y)) W(\mathrm{d} s \,\mathrm{d} y).$$

The aim of this section is to show that when  $\sigma$  and b are Lipschitz continuous the solution to (3.1) is spatially ergodic. This follows from an application of Theorem 2.2. Note that because we are discussing Lipschitz continuous b, there is no need to describe what we mean by solution; that is done already in Walsh [19].

According to Bally and Pardoux [1] (see also Nualart [15, Proposition 1.2.4]), under these conditions  $u(t,x) \in \mathbb{D}^{1,P}$  for all  $p \geq 2, t > 0$ , and  $x \in \mathbb{R}$ , and the Malliavin derivative Du(t,x)satisfies

$$D_{r,z}u(t,x) = p_{t-r}(x-z)\sigma(u(r,z)) + \int_{(r,t)\times\mathbb{R}} p_{t-s}(y-x)B_{s,y}D_{r,z}u(s,y)\,\mathrm{d}s\,\mathrm{d}y + \int_{(r,t)\times\mathbb{R}} p_{t-s}(y-x)\Sigma_{s,y}D_{r,z}u(s,y)\,W(\mathrm{d}s\,\mathrm{d}y) \qquad \text{a.s},$$

for a.e.  $(r, z) \in (0, t) \times \mathbb{R}$  where B and  $\Sigma$  are a.s. bounded random fields. We have the following estimate on the Malliavin derivative.

**Lemma 3.1.** If  $\sigma$  and b are Lipschitz continuous, then for every T > 0 and  $p \ge 2$  there exists  $C_{T,p} > 0$  such that

$$||D_{r,z}u(t,x)||_p \le C_{T,p} p_{t-r}(x-z)p_r(z).$$

uniformly for  $t \in (0,T)$  and  $x \in \mathbb{R}$ , and for almost every  $(r,z) \in (0,t) \times \mathbb{R}$ .

*Proof.* The proof follows closely the proof of Lemma 2.1 in Chen et al [4] but we must account for a few of the changes that are caused by the drift b: By Minkowski's inequality,

$$\left\| \int_{(r,t)\times\mathbb{R}} p_{t-s}(y-x) B_{s,y} D_{r,z} u(s,y) \,\mathrm{d}s \,\mathrm{d}y \right\|_p^2 \le c \int_r^t \mathrm{d}s \int_{-\infty}^\infty \mathrm{d}y \left[ p_{t-s}(x-y) \right]^2 \| D_{r,z} u(s,y) \|_p^2.$$

This is the same expression that appears in the right-hand side of (2.6) in [4]. Therefore, the rest of the proof follows the analogous argument in [4, Proof of Lemma 2.1].

We are now ready to state the main result of this section.

**Corollary 3.2.** If  $\sigma$  and b are Lipschitz continuous, then the random fields  $x \to u(t, x)$  and  $x \to \mathcal{I}(t, x)$  are stationary and ergodic for every t > 0.

*Proof.* Stationarity follows from Chen et al [3, Lemma 7.1], and ergodicity is a direct consequence of Lemma 3.1 and Theorem 2.7.

## 4 A lower bound via differential inequalities

In this section, we continue to assume that b is Lipschitz continuous. Our aim is to prove the following key result.

**Theorem 4.1.** If  $b : \mathbb{R} \to (0, \infty)$  is Lipschitz continuous, then for every non-random number a > 0, there exists a non-random number  $\varepsilon = \varepsilon(a) > 0$  – not depending on the choice of b – that satisfies the following for every  $M > ||u_0||_{L^{\infty}(\mathbb{R})}$ : There exists an a.s.-finite random variable c = c(a, M) > 0such that

$$\inf_{t \in [a+\varepsilon,a+2\varepsilon]} \inf_{x \in (c,c+\sqrt{\varepsilon})} u(t,x) \ge \sup\left\{N > M : \int_{M+\rho}^{N+\rho} \frac{\mathrm{d}y}{b(y)} < \varepsilon\right\} \qquad a.s. \quad [\sup \emptyset = 0],$$

where  $\rho := \inf_{x \in \mathbb{R}} u_0(x)$ .

The following result will be useful for the proof of the above theorem.

**Lemma 4.2.** Fix two numbers N > A > 0 and suppose  $B : \mathbb{R}_+ \to (0, \infty)$  is Lipschitz continuous. Let  $T = \int_A^N ds/B(s)$ , and suppose that  $F : \mathbb{R}_+ \to \mathbb{R}_+$  solves

$$F(t) \ge A + \int_0^t B(F(s)) \,\mathrm{d}s \qquad \text{for all } t \in [0, 2T].$$

Then  $\inf_{t \in [T,2T]} F(t) \ge N$ .

**Remark 4.3.** Lemma 4.2 can recast in slightly weaker terms as a statement about the differential inequality,

$$\begin{bmatrix} F' \ge B \circ F & \text{on } \mathbb{R}_+, \\ \text{subject to } F(0) \ge A. \end{bmatrix}$$

In this case,  $F(t) \ge N$  some time t between  $T = \int_A^N ds/B(s)$  and time 2T.

Proof. Choose and fix an A > 0. The ordinary differential equation  $G(t) = A + \int_0^t B(G(s)) ds$  has a unique continuous solution that is strictly increasing (hence also has an inverse) up to time  $T = \sup\{t > 0 : G(t) \le N\}$  for every N > A, and  $G(T) = \lim_{s \uparrow T} G(s) = N$ . We also have that  $G(t) \ge N$  for all  $t \in [T, 2T]$ . A comparison theorem yields  $F \ge G$  on [0, 2T], and completes the proof.

Proof of Theorem 4.1. We first assume that the initial data is equal to a constant  $\rho \in \mathbb{R}$ . Choose and fix a > 0. According to Corollary 3.2 and Theorem 2.2, we can associate to a a non-random number  $\varepsilon > 0$  such that

$$\limsup_{c \to \infty} \inf_{t \in (a+\varepsilon, a+2\varepsilon)} \inf_{x \in (0,\sqrt{\varepsilon})} \mathcal{I}(t, c+x) = \infty, \quad \text{a.s.}$$

Also choose and fix a number M > 0. According to Theorem 2.2, we can find a random number c > 0 such that

$$\inf_{t \in (a+\varepsilon, a+2\varepsilon)} \inf_{x \in (0,\sqrt{\varepsilon})} \mathcal{I}(t, c+x) > M \quad \text{a.s.}$$

$$(4.1)$$

Because  $b \ge 0$  and b is nondecreasing,

$$\begin{aligned} u(a+t,c+x) &\geq \rho + \int_{(0,t)\times\mathbb{R}} p_{a+t-s}(y-x-c)b(u(s,y)) \,\mathrm{d}s \,\mathrm{d}y + \mathcal{I}(a+t,c+x) \\ &\geq \rho + \int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)b(u(a+s,c+y)) \,\mathrm{d}s \,\mathrm{d}y + \mathcal{I}(a+t,c+x) \\ &\geq \rho + \int_0^t b\left(\inf_{z\in(0,\sqrt{\varepsilon})} u(a+s,c+z)\right) \,\mathrm{d}s \int_0^{\sqrt{\varepsilon}} \mathrm{d}y \ p_{t-s}(y-x) + \mathcal{I}(a+t,c+x), \end{aligned}$$

a.s., for every t, c > 0 and  $x \in \mathbb{R}$ . If in addition  $x \in (0, \sqrt{\varepsilon})$  and  $t \in (0, 2\varepsilon)$ , then

$$\int_{0}^{\sqrt{\varepsilon}} p_{t-s}(y-x) \, \mathrm{d}y = \int_{-x}^{-x+\sqrt{\varepsilon}} p_{t-s}(y) \, \mathrm{d}y \ge \int_{-\sqrt{\varepsilon}}^{0} p_{t-s}(y) \, \mathrm{d}y \ge \int_{-1/2}^{0} p_1(y) \, \mathrm{d}y = \ell > 0,$$

uniformly for all  $s \in (0, t)$ . Therefore, (4.1) tells us that, for all  $x \in (0, \sqrt{\varepsilon})$  and  $t \in (0, 2\varepsilon)$ ,

$$u(a+t,c+x) \ge \ell \int_0^t b\left(\inf_{z \in (0,\sqrt{\varepsilon})} u(a+s,c+z)\right) \mathrm{d}s + M + \rho.$$

In other words, we have shown that the function

$$f(t) = \inf_{x \in (0,\sqrt{\varepsilon})} u(a+t,c+x) \qquad [t > 0]$$

satisfies

$$f(t) \ge M + \rho + \ell \int_0^t b(f(s)) \, \mathrm{d}s$$
 uniformly for all  $t \in (0, 2\varepsilon)$ .

Since  $\int_{M+\rho}^{N+\rho} [b(y)]^{-1} dy < \varepsilon$ , Lemma 4.2 assures us that  $\inf_{t \in [\varepsilon, 2\varepsilon]} f(t) \ge N$ . Hence,

$$\inf_{s\in [a+\epsilon,a+2\epsilon]} \inf_{y\in (c,c+\sqrt{\varepsilon})} u(s\,,y) \ge N \quad \text{a.s.},$$

which yields the theorem in the case that the initial data is constant. For the general case that the initial condition is bounded, using a standard comparison theorem we can deduce the proof of the theorem.  $\hfill \Box$ 

### 5 Minimal solutions, and proof of Theorem 1.5

We begin by revisiting the well posedness of (1.1) under Assumptions 1.1 and 1.2. After that, we prove Theorem 1.5 and conclude the paper.

#### 5.1 Minimal solutions

Let  $\mathscr{L}_{loc}$  denote the collection of all functions  $f : \mathbb{R} \to (0, \infty)$  that are nondecreasing and locally Lipschitz continuous. In particular, Assumption 1.2 is shortened to the assertion that  $b \in \mathscr{L}_{loc}$ . We also define  $\mathscr{L}$  to be the collection of all elements of  $\mathscr{L}_{loc}$  that are Lipschitz continuous.

Throughout this subsection, we write the solution to (1.1) as  $u_b$  provided that (1.1) well posed for a given  $b \in \mathscr{L}_{loc}$ . As a consequence of the theory of Walsh [19], (1.1) is well posed for example when  $b \in \mathscr{L}$ ; see also Dalang [5]. Moreover,  $u_b$  is defined uniquely provided additionally that  $\sup_{t \in (0,T)} \sup_{x \in \mathbb{R}} ||u(t,x)||_2 < \infty$  for all T > 0. Finally,

$$P\{u_b \le u_c\} = 1 \quad \text{for all } b, c \in \mathscr{L} \text{ that satisfy } b \le c;$$

see Mueller [14] and [12].

Now suppose that  $b \in \mathscr{L}_{loc}$ , as is the case in the Introduction. Let  $b^{(n)} = b \wedge n$  for every  $n \in \mathbb{N}$ . The monotonicity of b implies that every  $b^{(n)} \in \mathscr{L}$  for every  $n \in \mathbb{N}$ , and  $b^{(n)} \leq b^{(m)}$  when  $n \leq m$ . Since  $u_{b^{(n)}} \leq u_{b^{(m)}}$  whenever  $n \leq m$ , it follows that the random field

$$u = \lim_{n \to \infty} u_{b^{(n)}}$$

exists and has lower-semicontinuous sample functions. Note also that if  $c \in \mathscr{L}$  satisfies  $c \leq b$ , then  $u_c \leq u$ . This proves immediately that

$$u = \sup_{c \in \mathscr{L}} u_c.$$

Therefore, we refer to u as the *minimal solution* to (1.1) when b satisfies Assumption 1.2.

Next we describe why u can justifiably be called the minimal "solution" to (1.1). Minimality is clear from context. However, "solution" deserves some words.

If b is in addition Lipschitz continuous, then u is the solution to (1.1) that the Walsh theory yields and there is nothing to discuss. Now suppose  $b \in \mathscr{L}_{loc}$  and recall  $b^{(n)} \in \mathscr{L}$ . We may observe that

$$b^{(n)}(u_{b^{(n)}}(t,x)) \le b^{(m)}(u_{b^{(m)}}(t,x))$$
 whenever  $n \le m_{t}$ 

off a single null set that does not depend on (b, n, m). Since

$$b^{(n)}(x) = \frac{b(x) + n - |b(x) - n|}{2}$$

it follows that

$$\lim_{n \to \infty} b^{(n)} \left( u_{b^{(n)}}(t, x) \right) = b(u(t, x)) \quad \text{for all } t > 0 \text{ and } x \in \mathbb{R}, \tag{5.1}$$

again off a single null set [these are real-variable, sure, assertions]. Therefore, the monotone convergence theorem yields

$$\lim_{n \to \infty} \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x) b^{(n)}(u^{(n)}(s,y)) \,\mathrm{d}s \,\mathrm{d}y = \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x) b(u(s,y)) \,\mathrm{d}s \,\mathrm{d}y,$$

where  $b(\infty) = \sup b$ .

Next, let us consider the  $[0, \infty]$ -valued random variable

$$\tau = \inf \left\{ t > 0 : u(s, y) = \infty \quad \text{for all } s \le t \text{ and } y \in \mathbb{R} \right\},\$$

where  $\inf \emptyset = 0$ . Because u is lower semicontinuous, one can show that  $\tau$  is a stopping time with respect to the filtration of the noise, which we assume satisfies the usual conditions of martingale theory, without loss of generality. Of course,  $\tau$  is the first blowup time of u. Since  $\sigma$  is a bounded and continuous function,

$$\lim_{n \to \infty} \left\| \int_{(0,t\wedge\tau)\times\mathbb{R}} p_{t-s}(y-x) [\sigma(u^{(n)}(s,y)) - \sigma(u(s,y))] W(\mathrm{d}s\,\mathrm{d}y) \right\|_2^2$$
$$= \mathrm{E}\left( \int_{(0,t\wedge\tau)\times\mathbb{R}} \left[ p_{(t\wedge\tau)-s}(y-x) \right]^2 \lim_{n \to \infty} [\sigma(u^{(n)}(s,y)) - \sigma(u(s,y))]^2 \mathrm{d}s\,\mathrm{d}y \right) = 0,$$

where  $\int_{\emptyset} (\cdots) = 0$ . Taken together, these comments prove that if  $\tau > 0$  – that is if the solution to (1.1) does not instantly blow up – then u satisfies (1.2) for all  $x \in \mathbb{R}$  and all times  $t < \tau$ .<sup>2</sup> In this sense, our extension of the solution theory of Walsh [19] indeed produces solutions for  $b \in \mathscr{L}_{loc}$  if there is chance for non-instantaneous blowup, and the smallest such solution is u.

Theorem 1.5 says that if  $b \in \mathscr{L}_{loc}$  satisfies the Osgood condition (1.3), then the minimal solution satisfies  $u(t) \equiv \infty$  for all t > 0.

Now suppose the Osgood condition holds, and consider any solution theory that extends the Walsh theory and has a comparison theorem. The preceding comments prove that if that solution theory produces a solution v, then that solution satisfies  $u \leq v$  and hence  $v(t) \equiv \infty$  for all t > 0 by Theorem 1.5. This is the precise conditional sense in which Theorem 1.5 says that "the solution" to (1.1) blows up instantaneously and everywhere.

We can now conclude the paper with the following.

### 5.2 Proof of Theorem 1.5

We now return to the notation of the Introduction and write u in place of  $u_b$ , and prove the everywhere and instantaneous blow up of u under (1.3), where the symbol u denotes the minimal solution to (1.1) as was described in the previous subsection.

Choose and fix a > 0, and in light of (1.3) we may choose and fix  $M > ||u_0||_{L^{\infty}(\mathbb{R})}$  such that

$$\int_{M+\rho}^{\infty} \frac{\mathrm{d}y}{b(y)} < \varepsilon.$$

Then, the construction of u and Theorem 4.1 together yield  $\varepsilon$  – independently of the choice of b and M – such that the following holds for every  $n \in \mathbb{N}$ :

$$\inf_{t\in[a+\varepsilon,a+2\varepsilon]} \inf_{x\in(c,c+\sqrt{\varepsilon})} u(t\,,x) \ge \inf_{t\in(a+\varepsilon,a+2\varepsilon)} \inf_{x\in(c,c+\sqrt{\varepsilon})} u^{(n)}(t\,,x)$$
$$\ge \sup\left\{N > M: \int_{M+\rho}^{N+\rho} \frac{\mathrm{d}y}{b^{(n)}(y)} < \varepsilon\right\} \qquad \text{a.s.}$$

Let  $n \uparrow \infty$  to see from the monotone convergence theorem that

$$\inf_{t \in [a+\varepsilon,a+2\varepsilon]} \inf_{x \in (c,c+\sqrt{\varepsilon})} u(t,x) \ge \sup \left\{ N > M : \int_{M+\rho}^{N+\rho} \frac{\mathrm{d}y}{b(y)} < \varepsilon \right\} = \infty \qquad \text{a.s.}$$

This proves that the blowup time is a.s.  $\leq a + 2\varepsilon(a)$  and that the solution blows up everywhere in a random interval of the type  $(c, c + \sqrt{\varepsilon})$ . Consequently, for every non-random t > 0 there a.s. is a random closed interval  $I(t) \subset (0, \infty)$  and and a non-random closed interval  $\tilde{I}(t) = [a + \varepsilon, a + 2\varepsilon] \subset$ (0, t) such that

$$\inf_{(s,x)\in \tilde{I}(t)\times I(t)} u(s,x) = \infty \qquad \text{a.s.}$$
(5.2)

$$u(t,x) = (p_t * u_0)(x) + \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x)b(u(s,y)) \, \mathrm{d}s \, \mathrm{d}y + a \text{ finite term},$$

where  $b(\infty) = \sup b$ . Theorem 1.5 implies that both sides of the above identity are infinite when (1.3) holds.

<sup>&</sup>lt;sup>2</sup>In fact, one can show that the limit of the stochastic integrals in the mild formulation of  $u^{(n)}$  is finite a.s. See the end of the proof of Theorem 1.5. This implies the stronger statement that, for all t > 0 and  $x \in \mathbb{R}$ ,

We now consider the process  $u^{(n)} = u_{b^{(n)}}$ , as defined in the previous subsection. For every  $n \in \mathbb{N}$ , the random field  $u^{(n)}$  solves

$$u^{(n)}(t,x) = (p_t * u_0)(x) + \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x)b^{(n)}(u^{(n)}(s,y)) \,\mathrm{d}s \,\mathrm{d}y + \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x)\sigma(u^{(n)}(s,y)) \,W(\mathrm{d}s \,\mathrm{d}y).$$

By the monotone convergence theorem,

$$\int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)b^{(n)}(u^{(n)}(s,y))\,\mathrm{d}s\,\mathrm{d}y \ge \int_{\tilde{I}(t)\times I(t)} p_{t-s}(y-x)b^{(n)}(u^{(n)}(s,y))\,\mathrm{d}s\,\mathrm{d}y\uparrow\infty,$$

as  $n \to \infty$ ; see (5.1) and (5.2). At the same time, standard estimates such as those in §2 show that

$$\sup_{n\in\mathbb{N}} \mathbb{E}\left(\sup_{(t,x)\in K} \left| \int_{(0,t)\times\mathbb{R}} p_{t-s}(y-x)\sigma(u^{(n)}(s,y)) W(\mathrm{d} s \,\mathrm{d} y) \right|^2 \right) < \infty.$$

for every compact set  $K \subset \mathbb{R}_+ \times \mathbb{R}$ . Therefore, Fatou's lemma ensures that a.s.,

$$\liminf_{n \to \infty} \sup_{(t,x) \in K} \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x) \sigma(u^{(n)}(s,y)) W(\mathrm{d} s \, \mathrm{d} y) < \infty.$$

It follows that  $\inf_{K} u = \infty$  a.s. for all compact sets  $K \subset \mathbb{R}_{+} \times \mathbb{R}$ . This concludes the proof.

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