

A random string near a wall

Lorenzo Zambotti
(LPMA, Univ. Paris 6)

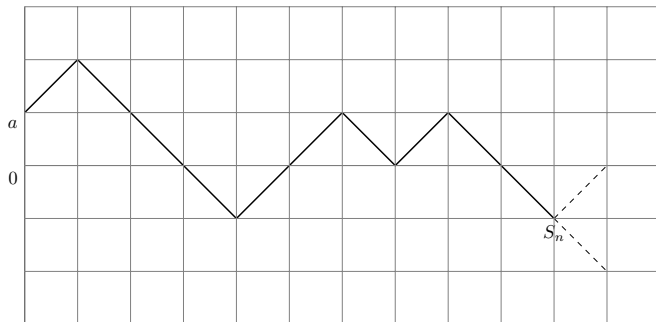
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Symmetric simple random walk

$(Y_i)_{i \geq 1}$ i.i.d. sequence with $\mathbb{P}(Y_i = 1) = \mathbb{P}(Y_i = -1) = 1/2$

$$S_0 := a \in \mathbb{Z}, \quad S_{n+1} = S_n + Y_{n+1},$$

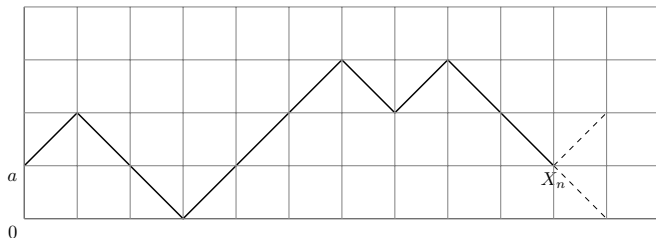
$(S_n)_{n \geq 0}$ defines a Markov chain with values in \mathbb{Z} .



Symmetric simple random walk with reflection

$$X_{n+1} := \begin{cases} X_n + Y_{n+1} & \text{if } X_n > 0 \\ 1 & \text{if } X_n = 0. \end{cases}$$

This defines a Markov chain with values in $\mathbb{Z}_+ = \{0, 1, \dots\}$.



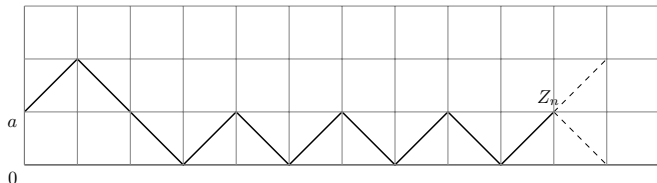
Absolute value of the SSRW

Let $Z_n := |S_n|$, with $S_0 = a \geq 0$. Although this is not obvious at first sight, $(Z_n)_{n \geq 0}$ is also a Markov chain:

$$Z_{n+1} = \begin{cases} Z_n + W_{n+1} & \text{if } Z_n > 0 \\ 1 & \text{if } Z_n = 0, \end{cases}$$

where $W_{n+1} := Y_{n+1}(\mathbb{1}_{S_n \geq 0} - \mathbb{1}_{S_n < 0})$.

It is easy to see that $(W_i)_{i \geq 1}$ is a copy of $(Y_i)_{i \geq 1}$ and then $(Z_n)_{n \geq 0}$ and $(X_n)_{n \geq 0}$ (for the same fixed a) **have the same law**.



Discrete interfaces

We define now a Markov chain with values in discrete paths. We fix $N \in \mathbb{N}$ and we define the state space

$$E_N := \{w \in \mathbb{Z}^{2N} : w(0) = w(2N) = 0, \\ |w(i) - w(i-1)| = 1, \forall i = 1, \dots, 2N\}.$$

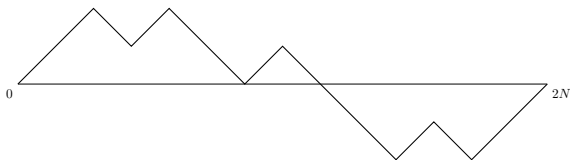
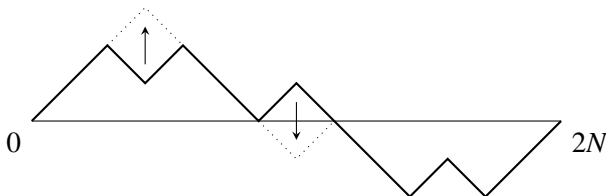


Figure: A typical path in E_N

The free evolution

Then we define a Markov chain with values in E_N as follows: we define a map $F : E_N \times \{1, \dots, 2N - 1\} \rightarrow E_N$, $F(w, j) = \hat{w} \in E_N$, where

$$\hat{w}_i = \begin{cases} w_i + 2 & \text{if } i = j \text{ and } w_{i-1} = w_{i+1} > w_i \\ w_i - 2 & \text{if } i = j \text{ and } w_{i-1} = w_{i+1} < w_i \\ w_i & \text{otherwise.} \end{cases}$$



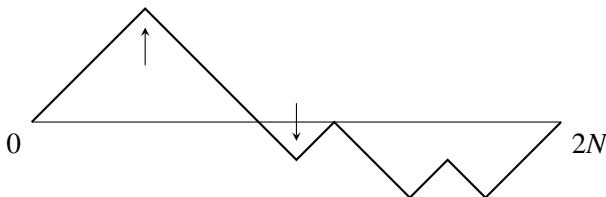
We let $(U_n)_{n \geq 1}$ be an i.i.d. sequence of uniform random variables on $\{1, \dots, 2N - 1\}$; then

$$e_{n+1} := F(e_n, U_{n+1}), \quad e_0 \in E_N.$$

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$$e_{n+1} := F(e_n, U_{n+1}), \quad e_0 \in E_N.$$

The free evolution

We denote the transition matrix of $(e_n)_{n \geq 0}$ by

$$P(x, y) = \mathbb{P}(F(x, U_1) = y).$$

It is easy to see that $P(x, y) = P(y, x)$ and therefore the uniform measure on E_N is invariant and reversible for P .

This is in fact the unique probability invariant measure of $(e_n)_{n \geq 0}$ by the following

Lemma

The Markov chain $(e_n)_{n \geq 0}$ is aperiodic and irreducible.

The reflected interface

Let us now add reflection to our discrete interface. We set

$$E_N^+ := \{w \in E_N : w(i) \geq 0, \forall i = 0, \dots, 2N\}.$$

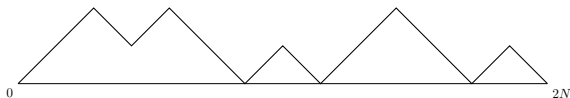
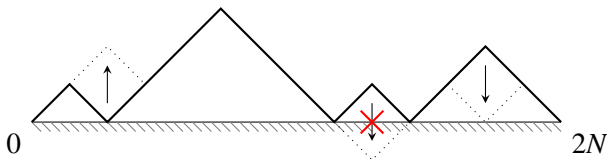


Figure: A typical path in E_N^+

Reflection means now suppression of transitions which would let e_n^+ take negative values:



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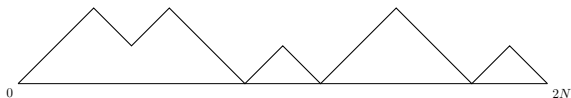
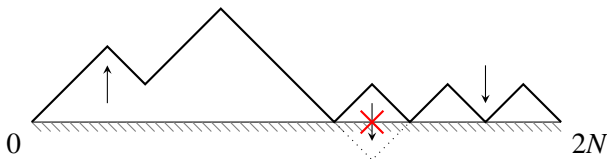


Figure: A typical path in E_N^+

Reflection means now suppression of transitions which would let e_n^+ take negative values:



The reflected interface

The Markov evolution in E_N^+ is defined as follows:

$$F^+ : E_N^+ \times \{1, \dots, 2N - 1\} \rightarrow E_N^+$$

$$F^+(w, j) = \begin{cases} F(w, j) & \text{if } F(w, j) \in E_N^+ \\ w & \text{otherwise.} \end{cases}$$

Then our E_N^+ -valued Markov chain $(e_n^+)_{n \geq 0}$ is defined by

$$e_{n+1}^+ := F^+(e_n^+, U_{n+1}), \quad e_0^+ \in E_N^+.$$

Lemma

The Markov chain $(e_n^+)_{n \geq 0}$ has a unique invariant probability measure, the uniform probability measure on E_N^+ , which is furthermore reversible for $(e_n^+)_{n \geq 0}$.

We see that the **reflection** for the dynamics is equivalent to a **conditioning** for the invariant measure.

An important remark

Let $|e_n|$ be the absolute value of the free interface $(e_n)_{n \geq 0}$. If $(e_n)_{n \geq 0}$ is stationary then the distribution of $|e_0|$ is a probability measure on E_N^+

$$\mathbb{P}(|e_0| = w) \propto \#\{w' \in E_N : |w'| = w\} = 2^{L(w)},$$

$$L(w) := \sum_{i=1}^{2N} \mathbb{1}_{(w_i=0)}, \quad w \in E_N^+.$$

In other words $L(w)$ is the number of excursions of w .

On the other hand, if $(e_n^+)_{n \geq 0}$ is stationary then the law of e_0^+ is uniform on E_N^+ , so that e_0^+ and $|e_0|$ have different laws.

Moreover $(|e_n|)_{n \geq 0}$ is **not Markovian**.

Scaling limits

Let $(S_n)_{n \geq 0}$ be the SSRW with $S_0 := 0$ and

$$B_t^N := \frac{1}{\sqrt{N}} S_{\lfloor Nt \rfloor}, \quad t \geq 0.$$

By Donsker's theorem $(B_t^N)_{t \geq 0} \Longrightarrow (B_t)_{t \geq 0}$ as $N \rightarrow +\infty$.

Under the same scaling the reflecting SSRW $(X_n)_{n \geq 0}$ converges to the reflecting Brownian motion $(\rho_t)_{t \geq 0}$.

This process is given by a stochastic differential equation

$$d\rho_t = dB_t + dl_t, \quad \rho_t \geq 0, \quad dl_t \geq 0, \quad \int_0^\infty \rho_t dl_t = 0,$$

- ▶ $t \mapsto (\rho_t, l_t)$ is continuous, ρ_t is non-negative
- ▶ $l_0 = 0, l_s \leq l_t$ for $s \leq t$
- ▶ $\text{supp}(dl_t) \subset \{t \geq 0 : \rho_t = 0\}$.

The measure dl_t is the **reflection term**.

Scaling of the free interface

Let us first consider the stationary version of $(e_n)_{n \geq 0}$ and define

$$v_N(t, x) = \frac{1}{\sqrt{2N}} e_{\lfloor 4N^2 t \rfloor}(\lfloor 2Nx \rfloor), \quad t \geq 0, x \in [0, 1].$$

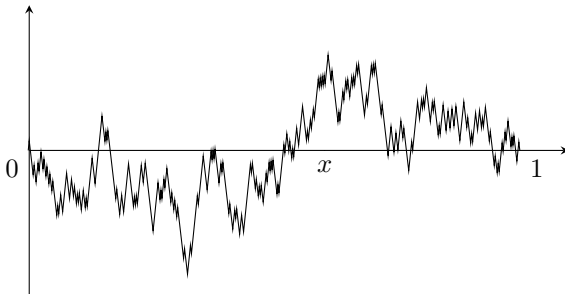


Figure: A typical path of $v_N(t, \cdot)$ for any $t \geq 0$ when N is large

The stochastic heat equation

As $N \rightarrow +\infty$, $v_N \Longrightarrow (v(t, x), t \geq 0, x \in [0, 1])$, stationary solution to a **stochastic partial differential equation (SPDE)**

$$\left\{ \begin{array}{l} \frac{\partial v}{\partial t} = \frac{1}{2} \frac{\partial^2 v}{\partial x^2} + W, \\ v(t, 0) = v(t, 1) = 0, \quad t \geq 0, \\ v(0, x) = v_0(x), \quad x \in [0, 1]. \end{array} \right.$$

Here W is a **space-time white noise**.

Scaling of the reflected interface

Let us now consider the stationary version of $(e_n^+)_{n \geq 0}$ and define

$$u_N(t, x) = \frac{1}{\sqrt{2N}} e_{\lfloor 4N^2 t \rfloor}^+(\lfloor 2Nx \rfloor), \quad t \geq 0, x \in [0, 1].$$

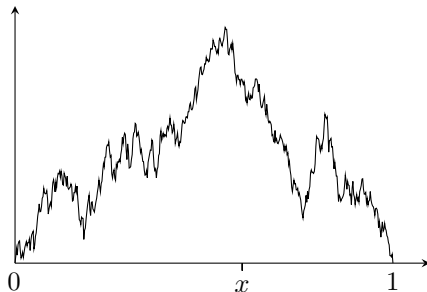


Figure: A typical path of $u_N(t, \cdot)$ for any $t \geq 0$ when N is large

A SPDE with reflection

As $N \rightarrow +\infty$, $u_N \Longrightarrow (u(t, x), t \geq 0, x \in [0, 1])$, that we are going to study in detail from chapter 5 on and is a stationary solution to a **SPDE with reflection** (Funaki-Olla, Z., Etheridge-Labbé)

$$\left\{ \begin{array}{l} \frac{\partial u}{\partial t} = \frac{1}{2} \frac{\partial^2 u}{\partial x^2} + W + \eta \\ u(0, x) = u_0(x), \quad u(t, 0) = u(t, 1) = 0 \\ u \geq 0, \quad d\eta \geq 0, \quad \int u \, d\eta = 0. \end{array} \right.$$

Here (u, η) is a random pair that consists of

- ▶ a continuous non-negative functions $u(t, x) \geq 0$
- ▶ a Radon measure η on $]0, +\infty[\times]0, 1[$,

such that the support of η is contained in $\{(t, x) : u(t, x) = 0\}$.

The contact set

For all $t \geq 0$ the typical profile of $u(t, \cdot)$ is positive on $]0, 1[$. Where does the reflection act?

The contact set

For all $t \geq 0$ the typical profile of $u(t, \cdot)$ is positive on $]0, 1[$. Where does the reflection act?

This apparent paradox is solved if we formulate the sentence more precisely: the correct result is that for all $t \geq 0$, a.s. $u(t, \cdot) > 0$ on $]0, 1[$:

$$\forall t > 0, \quad \mathbb{P}(\exists x \in]0, 1[: u(t, x) = 0) = 0.$$

However this does not exclude the existence, with positive probability, of **exceptional** times $t \geq 0$ and $x \in]0, 1[$ such that $u(t, x) = 0$:

$$\mathbb{P}(\exists t > 0, x \in]0, 1[: u(t, x) = 0) > 0.$$

The contact set

The next question is: what can be said about the **contact set**

$$\mathcal{L} := \{(t, x) : t > 0, x \in]0, 1[, u(t, x) = 0\}.$$

After proving that with positive probability u visits 0, one can ask:

- ▶ what is the **typical** behavior at exceptional times $t \geq 0$?
- ▶ That is, if $t > 0$ is such that there exists $x \in]0, 1[$ so that $u(t, x) = 0$, then **how many** such points x exist?

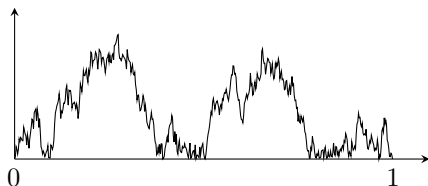


Figure: How many x such that $u(t, x) = 0$: infinitely many?

The contact set

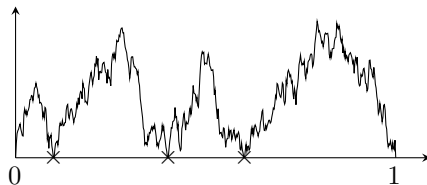


Figure: How many x such that $u(t, x) = 0$: finitely many?

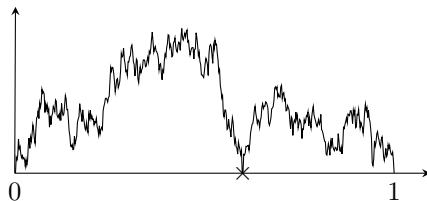


Figure: How many x such that $u(t, x) = 0$: just one? or two? or three?

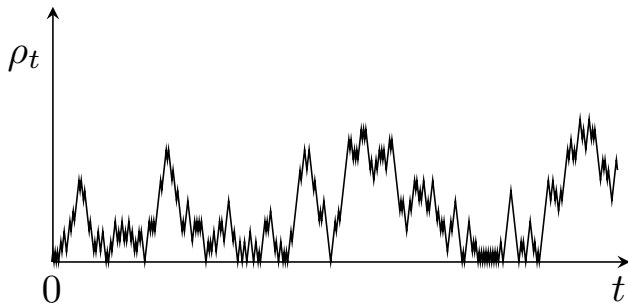
Proposition

Let $(B_t)_{t \geq 0}$ be a standard BM and $x \geq 0$. Then there exists a unique couple $(\rho_t, \ell_t)_{t \geq 0}$ of continuous real processes such that

$$\left\{ \begin{array}{l} \rho_t = x + B_t + \int_0^t f(\rho_s) ds + \ell_t, \quad t \geq 0 \\ \ell_0 = 0, \\ \rho_t \geq 0, \quad d\ell_t \geq 0, \quad \int_0^\infty \rho_t d\ell_t = 0. \end{array} \right. \quad (1)$$

If $f \equiv 0$ we call ρ the **reflecting BM**.

The reflecting BM



Penalisation

Let $n \geq 1$, $x \geq 0$ and

$$\rho_t^n = x + B_t + n \int_0^t (\rho_s^n)^- ds + \int_0^t f(\rho_s^n) ds, \quad t \in [0, T],$$

$$\text{where} \quad r^- = (r)^- := \max\{-r, 0\}, \quad r \in \mathbb{R}.$$

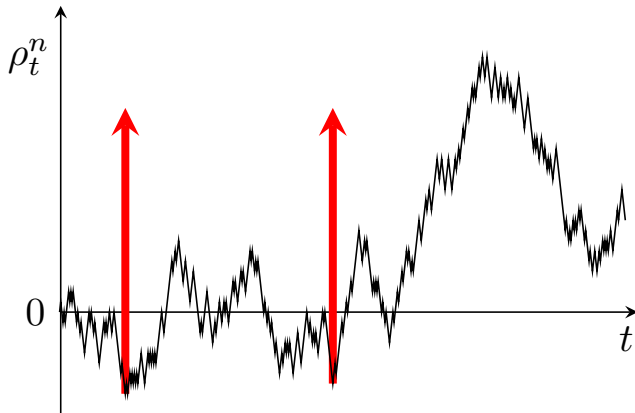
Additive noise and Lipschitz drift, so clearly pathwise uniqueness and existence of solutions by a Picard iteration.

Proposition

1. if $n \leq m$ then $\rho_t^n \leq \rho_t^m$ for all $t \in [0, T]$.
2. $\rho^n \uparrow \rho$ uniformly on $[0, T]$ as $n \uparrow +\infty$, where $(\rho_t, \ell_t)_{t \geq 0}$ is the unique solution to the equation with reflection (1). Moreover

$$\lim_{n \uparrow +\infty} n \int_0^t (\rho_s^n)^- ds = \ell_t, \quad t \in [0, T].$$

Penalisation



The penalised SDE

For $x \in \mathbb{R}$ and B a standard BM we set

$$\rho_t^n(x) = x + B_t + n \int_0^t (\rho_s^n(x))^- ds + \int_0^t f(\rho_s^n(x)) ds, \quad t \geq 0,$$

The infinitesimal generator of ρ^n is for $\varphi \in C_c^2(\mathbb{R})$

$$L^n \varphi(x) := \frac{1}{2} \varphi''(x) + (nx^- + f(x)) \varphi'(x), \quad x \in \mathbb{R}.$$

Moreover ρ^n admits as reversible invariant measure

$$\mu_n(dx) = e^{-n(x^-)^2 + 2F(x)} dx$$

where $F : \mathbb{R} \mapsto \mathbb{R}$ is any function such that

$$F'(x) = f(x), \quad x \in \mathbb{R}.$$

Note that $\mu_n([-\infty, 0]) < +\infty$ for n large, but $\mu_n([0, +\infty]) \leq +\infty$ in general.

The penalised SDE

Lemma

The measure μ is invariant and reversible for $(\rho_t)_{t \geq 0}$ and

$$\lim_{n \rightarrow +\infty} \int_{\mathbb{R}} \varphi \, d\mu_n = \int_{\mathbb{R}} \varphi \, d\mu, \quad \forall \varphi \in C_c(\mathbb{R}),$$

where

$$\mu(dx) := \mathbb{1}_{(x \geq 0)} e^{2F(x)} dx.$$

Here is an important message, that we have already noticed for discrete interfaces:

Remark

A *reflection* for the dynamics means a *conditioning* for the invariant measure.

δ -Bessel processes are solutions $(\rho_t)_{t \geq 0}$ to the SDE with $\delta > 1$

$$\rho_t = x + \frac{\delta - 1}{2} \int_0^t \rho_s^{-1} ds + B_t, \quad \rho_t \geq 0, \quad t \geq 0, \quad (2)$$

where $(B_t)_{t \geq 0}$ is a BM. If $\delta \downarrow 1$ the equation becomes

$$\rho_t = x + \ell_t + B_t, \quad \ell_0 = 0, \quad d\ell \geq 0, \quad \int_0^t \rho_s d\ell_s = 0, \quad (3)$$

i.e. the reflecting Brownian motion.

Bessel processes have the same **scaling invariance** of BM.

It is well known that a δ -Bessel process visits 0 with positive probability iff $\delta < 2$.

White noise

$W : L^2(\mathbb{R}_+^2) \rightarrow L^2(\Omega)$ isometry

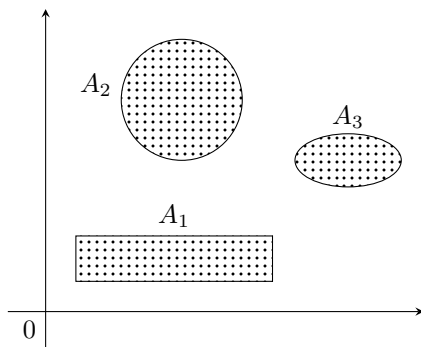


Figure: The family $(W(A_i))_i$ is Gaussian and independent since the sets A_i are pairwise disjoint. $W(A_i) \sim \mathcal{N}(0, m(A_i))$ and $W(\cup_i A_i) \sim \mathcal{N}(0, \sum_i m(A_i))$.

For all measurable set A , $W(A)$ is the amount of noise contained in A .

Lemma

Let $(e_i)_i$ be a complete orthonormal system in $L^2([0, +\infty[)$. Then

1. Let $w_t^i := W(\mathbb{1}_{[0,t]} \otimes e_i)$, $t \geq 0$, $i \in \mathbb{N}$. Then $(w_t^i)_i$ is an iid sequence of standard Brownian motions.
2. For all $h \in L^2([0, +\infty[)$ and $t \geq 0$

$$W(\mathbb{1}_{[0,t]} \otimes h) = \sum_i w_t^i \langle e_i, h \rangle$$

where the equality is in $L^2(\Omega)$.

A **cylindrical Brownian motion** in a separable Hilbert space H is

$$\langle W_t, h \rangle := \sum_i B_t^i \langle e_i, h \rangle, \quad t \geq 0,$$

where $(e_i)_i$ is a complete orthonormal system in H and $(B^i)_i$ is an iid sequence of Brownian motions.

The stochastic heat equation

We want to study the stochastic PDE

$$\left\{ \begin{array}{l} \frac{\partial v}{\partial t} = \frac{1}{2} \frac{\partial^2 v}{\partial x^2} + W, \\ v(t, 0) = v(t, 1) = 0, \quad t \geq 0, \\ v(0, x) = v_0(x), \quad x \in [0, 1] \end{array} \right. \quad (4)$$

where $W(t, x)$ is a space-time white-noise over $[0, +\infty[\times [0, 1]$.

This SPDE is interpreted in the PDE-weak sense: for all $h \in C_c^2(0, 1)$ and $t \geq 0$

$$\langle v_t, h \rangle = \langle v_0, h \rangle + \frac{1}{2} \int_0^t \langle v_s, h'' \rangle ds + \int_0^t \int_0^1 h(x) W(ds, dx).$$

Fourier decomposition

We set for all $k \geq 1$:

$$e_k(x) := \sqrt{2} \sin(k\pi x), \quad x \in [0, 1]. \quad (5)$$

Note that $\{e_k\}_{k \geq 1}$ is a complete basis of eigenvectors of the second derivative with homogeneous Dirichlet boundary conditions:

$$\frac{d^2}{dx^2} e_k = -(\pi k)^2 e_k, \quad e_k(0) = e_k(1) = 0, \quad k \geq 1.$$

Setting $v_t^k := \langle v(t, \cdot), e_k \rangle$ we obtain

$$dv_t^k = -\frac{(k\pi)^2}{2} v_t^k dt + dB_t^k, \quad v_0^k = \langle v_0, e_k \rangle$$

$$B_t^k := \int_{[0,t] \times [0,1]} e_k(x) W(ds, dx) = W(\mathbb{1}_{[0,t]} \otimes e_k).$$

Fourier decomposition

We proved in Lemma 7 that $(B_t^k, t \geq 0)_{k \geq 1}$ is an independent sequence of Brownian motions. Then $(v_t^k)_{k \geq 1}$ is an independent family of O-U processes of respective parameter $\frac{(\pi k)^2}{2} > 0$, and

$$v(t, x) = \sum_k \left(e^{-\frac{(\pi k)^2}{2}t} v_0^k + \int_0^t e^{-\frac{(\pi k)^2}{2}(t-s)} dB_s^k \right) e_k(x).$$

An important remark is the following:

$$\sum_k \frac{2}{(\pi k)^2} < +\infty \quad (d = 1).$$

Proposition

There exists a continuous modification of v s.t.

$$\sup_{x, y \in [0, 1], t, s \in [0, T]} \frac{|v(t, x) - v(s, y)|}{|t - s|^{\frac{1-\varepsilon}{4}} + |x - y|^{\frac{1-\varepsilon}{2}}} < +\infty.$$

The invariant measure

If we let $t \rightarrow +\infty$ in

$$v(t, \cdot) = \sum_k \left(e^{-\frac{(\pi k)^2}{2}t} v_0^k + \int_0^t e^{-\frac{(\pi k)^2}{2}(t-s)} dB_s^k \right) e_k$$

we obtain that the invariant measure of v is the law of

$$\beta := \sum_{k=1}^{+\infty} \frac{1}{\pi k} Z_k e_k,$$

where $(Z_k)_{k \geq 1}$ is an i.i.d. sequence of $\mathcal{N}(0, 1)$.

Proposition

$\beta = (\beta(x), x \in [0, 1])$ is a Brownian bridge.

Obstacle problems

Let $a \geq 0$. We fix a space-time white noise W on $[0, +\infty[\times [0, 1]$. We study the following SPDE with reflection:

$$\left\{ \begin{array}{l} \frac{\partial u}{\partial t} = \frac{1}{2} \frac{\partial^2 u}{\partial x^2} + f(u) + W + \eta \\ u(0, x) = u_0(x), \quad u(t, 0) = u(t, 1) = a \\ u \geq 0, \quad d\eta \geq 0, \quad \int u \, d\eta = 0 \end{array} \right. \quad (6)$$

where we assume that:

1. $u_0 : [0, 1] \mapsto \mathbb{R}$ is continuous and $u_0 \geq 0$.
2. $f : \mathbb{R} \mapsto \mathbb{R}$ is Lipschitz and bounded.

Reduction to a PDE with random obstacle

Let $a \geq 0$ and v be the unique solution to

$$\begin{cases} \frac{\partial v}{\partial t} = \frac{1}{2} \frac{\partial^2 v}{\partial x^2} + W, \\ v(t, 0) = v(t, 1) = a, \quad v(0, x) = u_0(x). \end{cases} \quad (7)$$

Then the function $z := u - v$ solves

$$\begin{cases} \frac{\partial z}{\partial t} = \frac{1}{2} \frac{\partial^2 z}{\partial x^2} + f(z + v) + \eta \\ z(0, x) = 0, \quad z(t, 0) = z(t, 1) = 0 \\ z \geq -v, \quad d\eta \geq 0, \quad \int (z + v) d\eta = 0. \end{cases} \quad (8)$$

The important remark here is that equation (8) is a PDE (rather than a SPDE) with **random obstacle** $-v$.

The Nualart-Pardoux equation

Theorem (Nualart-Pardoux, 1992)

Let $w \in C([0, T] \times [0, 1])$ with $w(0, \cdot) \geq 0$, $w(\cdot, 0) \geq 0$, $w(\cdot, 1) \geq 0$.
Then there exists a unique pair (z, η) such that

- ▶ $z \in C([0, T] \times [0, 1])$, $z(0, \cdot) = 0$, $z(\cdot, 0) = z(\cdot, 1) = 0$
- ▶ $\eta(dt, dx)$ is a measure on $]0, T[\times]0, 1[$ such that $\eta(]0, T[\times [\delta, 1 - \delta]) < +\infty$ for all $\delta > 0$
- ▶ For all $t \in [0, T]$ and $h \in C_c^\infty(0, 1)$

$$\begin{aligned} \langle z_t, h \rangle &= \frac{1}{2} \int_0^t \langle z_s, h'' \rangle ds + \int_0^t \langle f(z_s + w_s), h \rangle ds \\ &\quad + \int_0^t \int_0^1 h(x) \eta(ds, dx) \end{aligned} \tag{9}$$

- ▶ $z \geq -w$, $\int (z + w) d\eta = 0$.

We introduce the following approximating problem:

$$\left\{ \begin{array}{l} \frac{\partial u^\varepsilon}{\partial t} = \frac{1}{2} \frac{\partial^2 u^\varepsilon}{\partial x^2} + f(u^\varepsilon) + \frac{(u^\varepsilon)^-}{\varepsilon} + W \\ u^\varepsilon(0, \cdot) = u_0, \quad u^\varepsilon(t, 0) = u^\varepsilon(t, 1) = a. \end{array} \right.$$

Proposition

The pair (u, η) is the limit of the pair $(u^\varepsilon, \eta^\varepsilon)$ where

$$\eta^\varepsilon(dt, dx) := \frac{(u^\varepsilon(t, x))^-}{\varepsilon} dt dx.$$

The invariant measure

Let us consider a Brownian motion $(B_t^{(d)})_{t \geq 0}$ in \mathbb{R}^d where $B^{(d)} = (B^1, \dots, B^d)$ and the B^i 's are iid standard BMs.

Let $V : \mathbb{R}^d \rightarrow \mathbb{R}$ be a smooth function and let

$$dX_t = \nabla V(X_t) dt + dB_t^{(d)}, \quad X_0 = x \in \mathbb{R}^d.$$

It is a classical fact that an invariant measure for $(X_t)_{t \geq 0}$ is given by

$$\exp(2V(x)) dx.$$

If this measure is finite on \mathbb{R}^d , we obtain an invariant probability measure.

The penalised invariant measure

We consider now the penalised SPDE

$$\left\{ \begin{array}{l} \frac{\partial u^\varepsilon}{\partial t} = \frac{1}{2} \frac{\partial^2 u^\varepsilon}{\partial x^2} + f(u^\varepsilon) + \frac{(u^\varepsilon)^-}{\varepsilon} + W \\ u^\varepsilon(0, \cdot) = u_0, \quad u^\varepsilon(t, 0) = u^\varepsilon(t, 1) = a. \end{array} \right.$$

The invariant measure is

$$\nu_\varepsilon^a(d\zeta) := \frac{1}{Z_\varepsilon^a} \exp(2\langle F_\varepsilon(\zeta), 1 \rangle) \mathbf{W}_{a,a}(d\zeta),$$

where $\mathbf{W}_{a,a}$ is the law of $a + \beta$ and F_ε satisfies

$$F_\varepsilon(0) = 0, \quad F'_\varepsilon(y) := f(y) + \frac{y^-}{\varepsilon} = f_\varepsilon(y).$$

The penalised invariant measure

Note that

$$\frac{d}{dr} (r^-)^2 = -2r^-.$$

Then for $a > 0$

$$\nu_\varepsilon^a(d\zeta) = \frac{1}{Z_\varepsilon^a} \exp\left(2\langle F(\zeta), 1 \rangle - \frac{1}{\varepsilon} \langle (\zeta^-)^2, 1 \rangle\right) \mathbf{W}_{a,a}(d\zeta),$$

converges as $\varepsilon \downarrow 0$ to

$$\nu^a(d\zeta) := \frac{1}{Z^a} \exp(2\langle F(\zeta), 1 \rangle) \mathbb{1}_K(\zeta) \mathbf{W}_{a,a}(d\zeta),$$

where $K := \{u_0 : [0, 1] \rightarrow \mathbb{R} : u_0 \in L^2(0, 1), u_0 \geq 0\}$.

The Brownian excursion, or the 3-Bessel bridge

It turns out that

$$\mathbf{P}_{a,a}^3 = \mathbf{W}_{a,a}(\cdot | K).$$

Then

$$\nu^a(d\zeta) := \frac{1}{\hat{Z}^a} \exp(2\langle F(\zeta), 1 \rangle) \mathbf{P}_{a,a}^3(d\zeta),$$

and as $a \downarrow 0$

$$\nu^a(d\zeta) \implies \nu^0(d\zeta) := \frac{1}{\hat{Z}^0} \exp(2\langle F(\zeta), 1 \rangle) \mathbf{P}_{0,0}^3(d\zeta).$$

In particular if $f \equiv 0$ then the invariant measure of the SPDE with reflection is simply $\mathbf{P}_{a,a}^3$.

Integration by parts

Consider a regular bounded open set $O \subset \mathbb{R}^d$. Then the classical Gauss-Green formula states that for all $h \in \mathbb{R}^d$

$$\int_O (\partial_h \varphi) \rho \, dx = - \int_O \varphi \frac{\partial_h \rho}{\rho} \rho \, dx - \int_{\partial O} \varphi \langle \hat{n}, h \rangle \rho \, d\sigma$$

- ▶ $\varphi, \rho \in C_b^1(O)$ with $\lambda \leq \rho \leq \lambda^{-1}$, $\lambda \in]0, 1]$ is a constant,
- ▶ \hat{n} is the inward-pointing normal vector to the boundary ∂O
- ▶ σ is the surface measure on ∂O
- ▶ $\partial_h \varphi$ is the directional derivative of φ along h
- ▶ $\partial_h \log \rho = (\partial_h \rho) / \rho$.

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- ▶ $\partial_h \log \rho = (\partial_h \rho) / \rho$.

For us, $\mathbf{W}_{a,a} = \rho \, dx$, $K = O$.

What is the analog of $\rho \, d\sigma$? and of \hat{n} ? and of ∂O ?

The boundary measure

$$\begin{aligned} \mathbf{P}_{a,a}^3 [\partial_h \varphi] &= -\mathbf{P}_{a,a}^3 [\varphi(X) \langle X, h'' \rangle] \\ &\quad - \int_0^1 dr h(r) \gamma(r, a) \mathbf{P}_{a,a}^3 [\varphi(X) | X_r = 0]. \end{aligned}$$

where $\gamma(r, a) \geq 0$ is an explicit function of $r \in]0, 1[$, $a \geq 0$.

$$\int_O (\partial_h \varphi) \rho \, dx = - \int_O \varphi \frac{\partial_h \rho}{\rho} \rho \, dx - \int_{\partial O} \varphi \langle \hat{n}, h \rangle \rho \, d\sigma.$$

The boundary measure

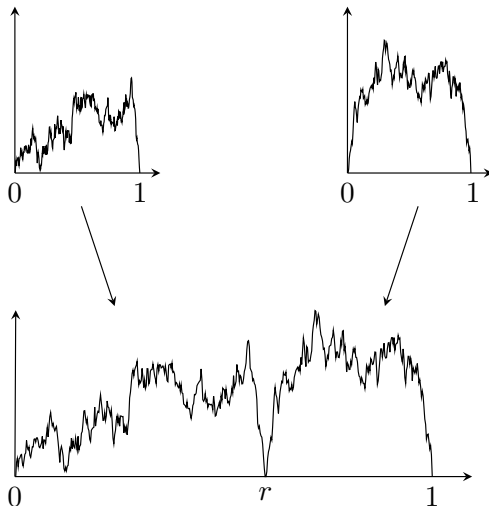


Figure: A typical path under the boundary measure.

Theorem

For all bounded Borel $\varphi : H \mapsto \mathbb{R}$ and $h \in C_c(0, 1)$

$$\begin{aligned} \int \nu^a(\mathrm{d}u_0) \mathbb{E} \left[\int_0^t \int_0^1 h(x) \varphi(u_s) \eta(\mathrm{d}s, \mathrm{d}x) \right] &= \\ &= \frac{t}{2Z^a} \int_0^1 \mathrm{d}r h(r) \gamma(r, a) \int \varphi(\zeta) e^{2F(\zeta)} \Sigma_a(r, \mathrm{d}\zeta). \end{aligned}$$

where $\Sigma_a(r, \cdot) := \mathbf{P}_{a,a}^3[\cdot | X_r = 0]$.

The contact set

We denote by $\pi : [0, +\infty[\times [0, 1] \mapsto [0, +\infty[$ the projection $(t, x) \mapsto t$, and for a set $S \subset [0, +\infty[\times [0, 1]$ we write

$$S_t := \{x \in [0, 1] : (t, x) \in S\}, \quad t \geq 0.$$

Theorem

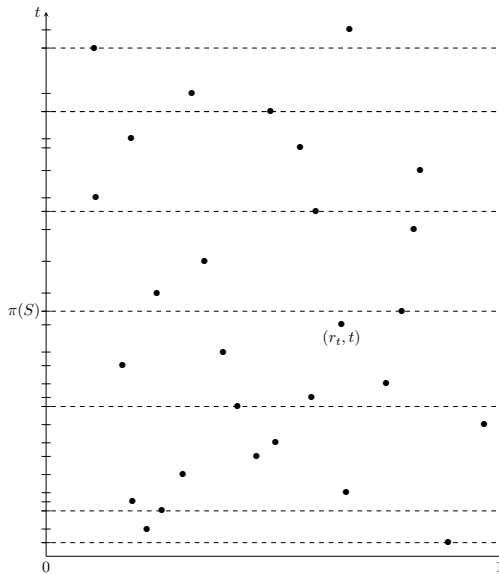
Let (u, η) be the stationary solution to equation (6). Let us denote by

$$\mathcal{C} := \{(t, x) : u(t, x) = 0, t > 0, x \in]0, 1[\}$$

the contact set and let us recall that the support of η is contained in \mathcal{C} . Then a.s. the set $\pi(\mathcal{C})$ has zero Lebesgue measure and there exists a measurable set $S \subset \mathcal{C}$ such that

1. $\eta(\mathcal{C} \setminus S) = 0$
2. for all $t > 0$, either $S_t = \emptyset$ or $S_t = \{r_t\}$, with $r_t \in]0, 1[$.
3. if $S_t = \{r_t\}$, then $u(t, x) > 0$ for all $x \in]0, 1[\setminus \{r_t\}$ and $u(t, r_t) = 0$.

The contact set



SPDEs with repulsion from 0

We study now the SPDE

$$\left\{ \begin{array}{l} \frac{\partial u}{\partial t} = \frac{1}{2} \frac{\partial^2 u}{\partial x^2} + \frac{c}{u^3} + W \\ u(t, 0) = u(t, 1) = a, \quad t \geq 0 \\ u(0, x) = u_0(x), \quad x \in [0, 1] \end{array} \right.$$

where $a \geq 0$ and $c > 0$ are fixed and we search for solutions $u \geq 0$.

This SPDE has the same **invariance scaling** as the linear and the reflected SPDE.

This SPDE is an analogue of Bessel processes for $\delta > 1$ (see the slide no. 28).

Theorem

Let $a \geq 0$, $c > 0$ and $u_0 \in C([0, 1]) \cap K$. Then there exists a unique continuous $u : [0, +\infty[\times [0, 1] \mapsto [0, +\infty[$ such that

1. $u^{-3} \in L^1_{loc}([0, +\infty[\times]0, 1])$
2. A.s. for all $t \geq 0$ and $h \in C_c^\infty(0, 1)$

$$\begin{aligned} \langle u_t, h \rangle &= \langle u_0, h \rangle + \frac{1}{2} \int_0^t \langle u_s, h'' \rangle ds + \int_0^t \int_0^1 h(x) W(ds, dx) \\ &\quad + c \int_0^t \int_0^1 h(x) u^{-3}(s, x) ds dx. \end{aligned} \tag{10}$$

If $\delta > 3$ is such that $c = \frac{(\delta-3)(\delta-1)}{8}$, then the only invariant probability measure of (10) is $\mathbf{P}_{a,c}^\delta$, law of the δ -Bessel bridge.

We have now functions $u = u^\delta$ for $\delta \geq 3$, stationary solutions to equations with reflection ($\delta = 3$) or repulsion from 0 ($\delta > 3$).

One of the main results of this course is the following

Theorem (Dalang, Mueller, Z. 2006)

Let $\delta \geq 3$. If $k \in \mathbb{N}$ satisfies

$$k > \frac{4}{\delta - 2},$$

the probability that there exist $t > 0$ and $x_1, \dots, x_k \in [0, 1]$ such that $0 < x_1 < \dots < x_k < 1$ and $u(t, x_i) = 0$ for all $i = 1, \dots, k$, is zero.

Hitting of zero

In particular, setting for $\delta \geq 3$

$$\zeta(\delta) := \sup\{k : \exists (t, x_1, \dots, x_k) \in]0, 1[\times]0, 1[, u(t, x_i) = 0\}$$

then

- ▶ for $\delta = 3$, a.s. $\zeta(\delta) \leq 4$
- ▶ for $\delta \in]3, 3 + 1/3]$, a.s. $\zeta(\delta) \leq 3$
- ▶ for $\delta \in]3 + 1/3, 4]$, a.s. $\zeta(\delta) \leq 2$
- ▶ for $\delta \in]4, 6]$, a.s. $\zeta(\delta) \leq 1$
- ▶ for $\delta > 6$, a.s. $\zeta(\delta) = 0$.

In any case $\zeta(\delta) \leq 4$ a.s. for all $\delta \geq 3$. The behavior at the transition points $\delta \in \{3, 3 + 1/3, 4, 6\}$ might be non-optimal. Indeed, we conjecture that a.s.

$$\zeta(3) \leq 3, \quad \zeta(3 + 1/3) \leq 2, \quad \zeta(4) \leq 1, \quad \zeta(6) = 0.$$

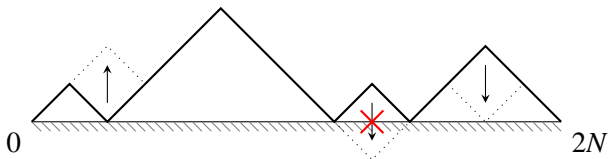
Theorem (Dalang, Mueller, Z.)

- (a) For all $\delta \in [3, 5]$, with positive probability, there exist $t > 0$ and $x \in]0, 1[$ such that $u_t(x) = 0$.
- (b) For $\delta = 3$, with positive probability there exist $t > 0$ and $\{x_1, x_2, x_3\} \subset]0, 1[$, $x_1 < x_2 < x_3$, such that $u_t(x_i) = 0$, $i = 1, 2, 3$.

We conjecture that for all $\delta \geq 3$ a.s.

$$\zeta(\delta) = \left\lceil \frac{4}{\delta - 2} \right\rceil - 1.$$

Back to the discrete interface



- ▶ What can be said on the **contact set** of the dynamical discrete interface as $N \rightarrow +\infty$?

Other open problems

- ▶ Construct SPDEs whose invariant measure is the δ -Bessel bridge for $\delta < 3$ (log-concavity is lost).
- ▶ One can conjecture that a.s.

$$\zeta(\delta) = \left\lceil \frac{4}{\delta - 2} \right\rceil - 1, \quad \delta \geq 2.$$

- ▶ For $\delta < 2$ the situation is even more complicated since 0 is hit by the stationary profile.
- ▶ The case $\delta = 1$ (reflecting BM) is the most intriguing since it is the limit of **homogeneous pinning models**.
- ▶ There is an IbPF for $\delta = 1$ but the form of the dynamics is hard even to conjecture.
- ▶ Dynamics of random trees (Aldous' CRT)