

# Root's and Rost's solution of the SEP

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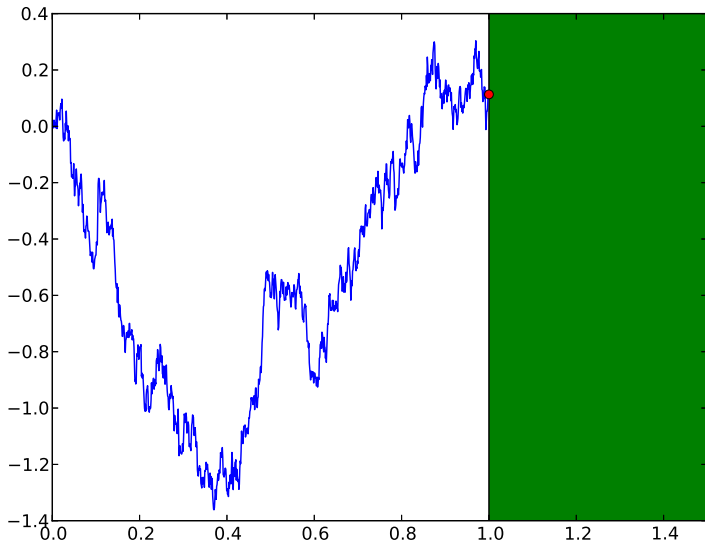
TU Berlin

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- I. An example: generating Brownian increments by Skorokhod stopping times of minimal variance
- II. Calculating Root barriers with integral equations
- III. Root/Rost barriers, viscosity solutions of obstacle problems and FBSDEs

# I. Generating Brownian increments by Skorokhod stopping times

# Simulating Brownian motion $B = (B_t)_{t \geq 0}$



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## Algorithm 1. (Euler)

▶  $(\tau_0, X_0) = (0, 0)$

▶ Draw

$$\begin{cases} X_{k+1} &= X_k + N_k \text{ with } N_k \sim \mathcal{N}(0, 1) \\ \tau_{k+1} &= \tau_k + 1 \end{cases}$$

▶ Gives  $(\tau_k, X_k)_{k \geq 0} \stackrel{\mathcal{L}}{=} (\tau_k, B_{\tau_k})_{k \geq 0}$

...can be seen as Skorokhod embedding

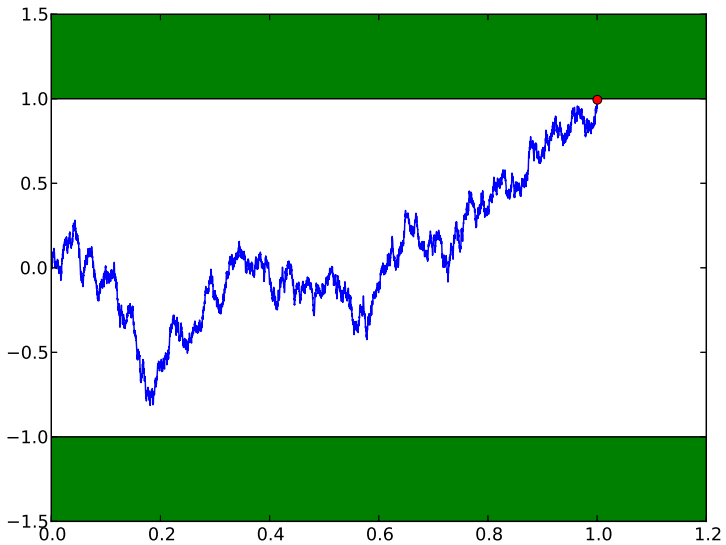
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$\tau_1 \equiv 1$  solves SEP  $B_\tau \sim \mathcal{N}(0, 1)$



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### Algorithm 2 (Bichteler–Karandhikar)

- ▶  $(\tau_0, X_0) = (0, 0)$
- ▶ Draw

$$\begin{cases} X_{k+1} = X_k + N_k \text{ with } \mathbb{P}(N_k = 1) = \mathbb{P}(N_k = -1) = \frac{1}{2} \\ \tau_{k+1} = \tau_k + D_k \text{ with } D_k \text{ s.t. } \mathbb{E}[\exp \lambda D_k] = \frac{1}{\cosh(\sqrt{2\lambda})} \end{cases}$$

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$\tau_1$  solves SEP  $B_\tau \sim \frac{1}{2}\delta_{-1} + \frac{1}{2}\delta_1$

## Root's barrier

- ▶  $\tau_1$  solution of SEP  $B_{\tau_1} \sim \mu$ 
  - ▶ in algorithm 1  $\mu = \mathcal{N}(0, 1)$
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- ▶  $\tau_1$  hitting time of time-space process  $t \mapsto (t, B_t)$

$$\tau_1 = \inf \{t > 0 : (t, B_t) \in R\}$$

- ▶ in algorithm 1  $R = \{(s, x) : s \geq 1, x \in \mathbb{R}\}$
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### Theorem (Root's barrier, 1968)

*Let  $\mu$  be centered and have finite second moment. Then there exists a closed set*

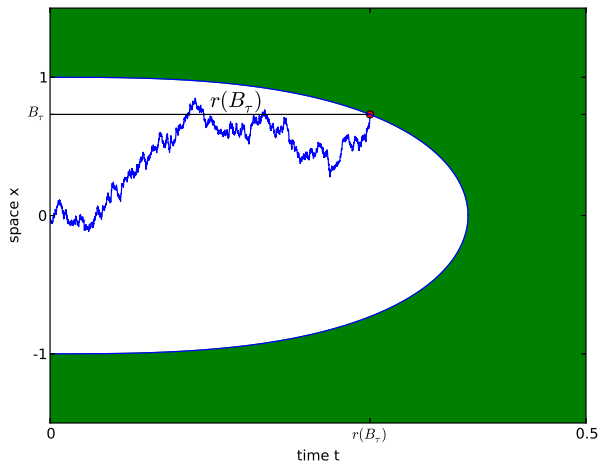
$$R \subset [0, \infty] \times [-\infty, \infty]$$

*such that  $\tau_R := \inf \{t \geq 0 : (t, B_t) \in R\}$  solves the Skorokhod embedding  $B_{\tau} \sim \mu$ ,  $B^{\tau} = (B_{t \wedge \tau})_{t \geq 0}$  u.i.*

Better: solve SEP with  $\mu = \mathcal{U}[-1, 1]$

### Corollary

$\exists r \in C(\mathbb{R}, [0, \infty))$  s.t.  $R = \{(t, x) : t \geq r(x)\}$  is the Root barrier for the SEP  $B_\tau \sim \mathcal{U}[-1, 1]$ .



### Algorithm 3

- ▶  $(\tau_0, X_0) = (0, 0)$
- ▶ Draw

$$\begin{cases} X_{k+1} &= X_k + U_k \text{ with } U_k \sim \mathcal{U}[-1, 1] \\ \tau_{k+1} &= \tau_k + r(U_k) \end{cases}$$

- ▶ Then  $(\tau_k, X_k)_{k \geq 0} \stackrel{\mathcal{L}}{=} (\tau_k, B_{\tau_k})_{k \geq 0}$ .
  - ▶ Trivial to simulate (once you know  $r$ )
  - ▶ Increments bounded in space AND time (scaled Monte-Carlo; example knock-out options)
  - ▶ Similar schemes without SEP (Milstein–Tretyakov, Lejay, Deaconu–Hermann, etc.)
  - ▶ For more applications see Gassiat&Mijatovic&O13

**Problem:** Find Root barrier  $R$  for any given distribution  $\mu$

**Unfortunately:**

- ▶ Root's existence proof of barrier  $R$  not constructive

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**Rest of talk:**

- ▶ 1-1 correspondence of Root barrier and solutions of nonlinear integral equation
- ▶ 1-1 correspondence between Root barrier and viscosity solution of parabolic obstacle problem (Dupire, Cox–Wang)
- ▶ 1-1 correspondence of  $R$  with solution of reflected FBSDE
- ▶ Barles–Souganidis for numerical schemes parabolic obstacle problem



## II. Root barrier and integral equations

Set

$$g(t, x) = \mathbb{E}L_t^x = \sqrt{\frac{2}{\pi}} \sqrt{t} e^{-\frac{x^2}{2t}} - x \operatorname{erfc}\left(\frac{x}{\sqrt{2}\sqrt{t}}\right)$$
$$K(r, \bar{r}, x, y) = \frac{1}{2} (g(r - \bar{r}, x - y) + g(r - \bar{r}, x + y))$$

Theorem (Gassiat&Mijatovic&O13)

$\exists! r \in C_b([-1, 1], \mathbb{R}_{\geq 0})$  which solves the integral equation

$$\frac{x^2 + 1}{2} - x = g(r(x), x) - \int_x^1 K(r(x), r(y), x, y) dy$$

Moreover, if we extend  $r$  to  $\mathbb{R}$  by  $r(x) = 0$  for  $x \in \mathbb{R} \setminus [-1, 1]$  then

$$R = \{(t, x) : t \geq r(x)\}$$

is the Root barrier for the SEP  $B_\tau \sim \mathcal{U}[-1, 1]$ .

## Potential functions

$B$  one-dimensional Brownian motion; denote semigroup (of transformations on measures)  $(P_t^B)$ . Define operator  $U^B$

$$\mu \mapsto U^B \mu := \int_0^\infty P_t^B \mu dt$$

If  $\mu$  is a signed measure with  $\mu(\mathbb{R}) = 0$  then

$$\frac{dU^B \mu}{dx} = - \int_{\mathbb{R}} |x - y| \mu(dy) =: u_\mu(x)$$

(rhs well-defined also for positive measures with finite moment).

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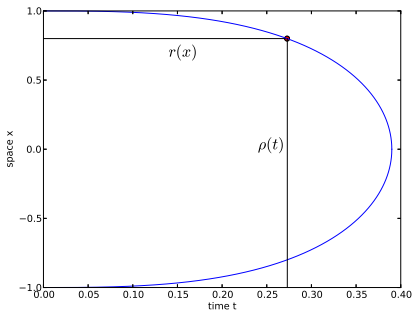
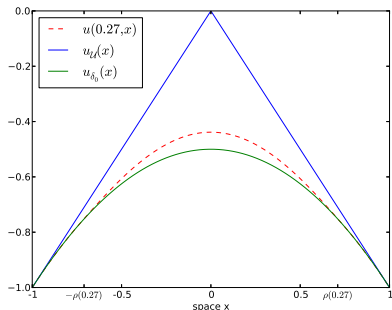
### Definition

For probability measure  $\mu$  on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  with finite first moment, associate with it a function  $u_\mu \in C(\mathbb{R}, \mathbb{R}_{\leq 0})$

$$u_\mu(x) = - \int_{\mathbb{R}} |x - y| \mu(dy)$$

We call  $u_\mu$  the **potential function** of  $\mu$ .

$$u(t, x) = -\mathbb{E} [|B_{t \wedge \tau} - x|]$$



- ▶  $u(0, x) = u_{\delta_0}(x) = -x, u(\infty, x) = u_{\mathcal{U}}(x) = \frac{x^2+1}{2}$
- ▶  $R = \{(t, x) : u(t, x) = u_{\mathcal{U}}(x)\}$

- ▶ Let  $R$  be the Root barrier for  $B_\tau \sim \mathcal{U}[-1, 1]$ ,  $r(x)$   
s.t.  $R = \{(t, x) : t \geq r(x)\}$  and  $\rho(t) = r^{-1}(t)$  (positive)
- ▶ Tanaka:  $\forall (t, x)$

$$\begin{aligned} u(t, x) &= u_\delta(x) - \mathbb{E}[L_{t \wedge \tau}^x] = -|x| - \mathbb{E}[L_t^x + \mathbf{1}_{t > \tau}(L_\tau^x - L_t^x)] \\ &= -|x| - g(t, x) - \mathbb{E}[\mathbf{1}_{t > \tau}(L_\tau^x - L_t^x)] \end{aligned}$$

- ▶ At  $x = \rho_t$ ,  $u(t, \rho_t) = u_{\mathcal{U}}(x)$  above becomes

$$u_{\mathcal{U}}(\rho_t) = -|\rho_t| - g(t, \rho_t) - \mathbb{E}[\mathbf{1}_{t > \tau}(L_\tau^{\rho_t} - L_t^{\rho_t})]$$

- ▶ Finished if we can write as explicit  $\mathbb{E}[\mathbf{1}_{t > \tau}(L_\tau^{\rho_t} - L_t^{\rho_t})]$   
functional of  $\rho$ .

- ▶ Note  $\mathbb{P}(\tau < t) = \mathbb{P}(\mathcal{U} \notin [-\rho_t, \rho_t]) = 1 - \rho_t$  hence

$$\mathbb{P}(\tau \in dt) = -d\rho_t$$

- ▶ Using Markovianity and symmetry

$$\begin{aligned} \mathbb{E}[(L_t^x - L_\tau^x) 1_{t > \tau}] &= \int_0^t \mathbb{E}[(L_t^x - L_s^x) | \tau = s] \mathbb{P}(\tau \in ds) \\ &= - \int_0^t \mathbb{E}[(L_t^x - L_s^x) | \tau = s] d\rho_s \\ &= - \int_0^t \frac{1}{2} \left( \mathbb{E}[L_{t-s}^{x-\rho_s}] + \mathbb{E}[L_{t-s}^{x+\rho_s}] \right) d\rho_s. \end{aligned}$$

- ▶ Putting this into above

$$u_{\mathcal{U}}(\rho_t) = u_{\delta}(\rho_t) - g(t, \rho_t) + \frac{1}{2} \int_0^t \overbrace{\mathbb{E}[L_{t-s}^{\rho_t - \rho_s}] + \mathbb{E}[L_{t-s}^{\rho_t + \rho_s}]}^{=g(t-s, \rho_t - \rho_s) + g(t-s, \rho_t + \rho_s)} d\rho_s$$

- ▶ Finish by change of variable  $dy = d\rho(s)$ :

$$u_{\mathcal{U}}(x) = u_{\delta,}(x) - g(r(x), x) + \frac{1}{2} \int_1^x K(r(x), r(y), x, y) dy$$

- ▶ Derivation purely probabilistic...no PDE techniques
- ▶ Extends to other target distributions (but there are limits)
- ▶ Uniqueness of solutions is hard (without using PDE uniqueness)! see Gassiat&Mijatovich&O13
- ▶ Useful? Solving this integral equation is numerically MUCH MUCH better than solving for free boundary via PDE
- ▶ The integral term in

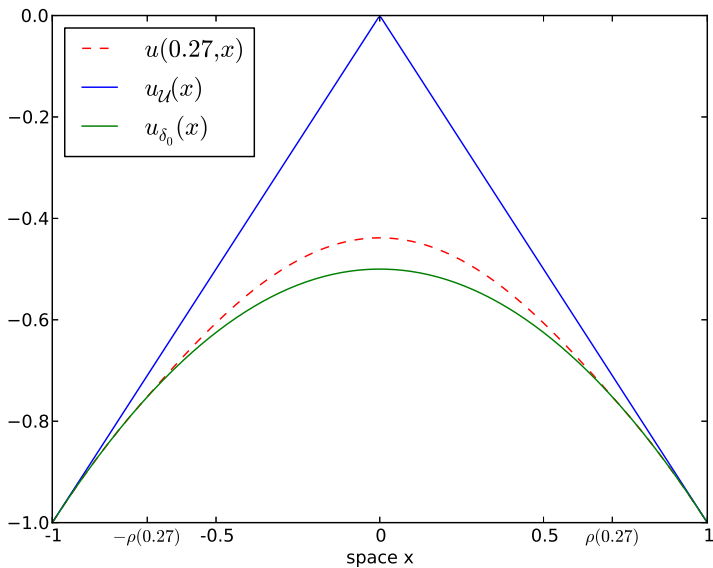
$$\frac{x^2 + 1}{2} - x = g(r(x), x) - \int_x^1 K(r(x), r(y), x, y) dy$$

is very small. Hence applying  $\frac{d}{dx}$  to both both sides of  $\frac{x^2+1}{2} - x = g(r(x), x)$  gives ODE for  $r$  which is a very good approximation.



Root/Rost barriers, viscosity solutions of obstacle problems and  
FBSDEs

$$u(t, x) = -\mathbb{E} [|B_{t \wedge \tau} - x|]$$



## Recall viscosity theory

### Definition

$\mathcal{O}$  a locally compact subset of  $\mathbb{R}$ ,  $\mathcal{O}_T = (0, T) \times \mathcal{O}$  for  $T \in (0, \infty]$ .  
Let  $u : \mathcal{O}_T \rightarrow \mathbb{R}$  and define for  $(s, z) \in \mathcal{O}_T$  the parabolic superjet  $\mathcal{P}_{\mathcal{O}}^{2,+} u(s, z)$  as the set of triples  $(a, p, m) \in \mathbb{R} \times \mathbb{R} \times \mathbb{R}$  which fulfill

$$u(t, x) \leq u(s, z) + a(t - s) + p(x - z) + m \frac{(x - z)^2}{2} + o(|t - s| + |x - z|^2)$$

as  $\mathcal{O}_T \ni (t, x) \rightarrow (s, z)$ .

Similarly we define the parabolic subjet  $\mathcal{P}_{\mathcal{O}}^{2,-} u(s, z)$  such that  $\mathcal{P}_{\mathcal{O}}^{2,-} u = -\mathcal{P}_{\mathcal{O}}^{2,+}(-u)$ .

## Definition

A function  $F : \mathcal{O}_T \times \mathbb{R} \times \mathbb{R} \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$  is proper if

$$\forall (t, x, a, p) \in \mathcal{O}_T \times \mathbb{R} \times \mathbb{R}$$

$$F(t, x, r, a, p, m) \leq F(t, x, s, a, p, m') \quad \forall m \geq m', s \geq r.$$

Denote the real-valued, upper semicontinuous functions on  $\mathcal{O}_T$  with  $USC(\mathcal{O}_T)$ . A subsolution of

$$\begin{cases} F(t, x, u, \partial_t u, Du, D^2 u) = 0 \\ u(0, \cdot) = u_0(\cdot) \end{cases} \quad (1)$$

is a function  $u \in USC(\mathcal{O}_T)$  such that

$$\begin{aligned} F(t, x, a, p, m) &\leq 0 \text{ for } (t, x) \in \mathcal{O}_T \text{ and } (a, p, m) \in \mathcal{P}_O^{2,+} u(t, x) \\ u(0, \cdot) &\leq u_0(\cdot) \text{ on } \mathcal{O} \end{aligned}$$

The definition of a supersolution follows by replacing upper by lower semicontinuous,  $\mathcal{P}_O^{2,+}$  by  $\mathcal{P}_O^{2,-}$  and  $\leq$  by  $\geq$ .

- ▶ If  $u$  is a classic  $C^{1,2}((0, T) \times \mathbb{R}, \mathbb{R})$  solution of

$$\begin{cases} F(t, x, u, \partial_t u, Du, D^2 u) = 0 \\ u(0, \cdot) = u_0(\cdot) \end{cases}$$

then  $v$  is also a viscosity solution

- ▶ Comparison Theorem (Maximum Principle):  $u(0, \cdot) \leq v(0, \cdot)$ ,  
 $u$  sub- and  $v$  supersolution implies  $u \leq v$

# Barles-Perthame's semi-relaxed limits

## Proposition

Let  $(u^n)_n \subset USC(\mathcal{O}_T)$ ,  $\mathcal{O}$  a locally compact subset of  $\mathbb{R}$ ,  $(F_n)$  a sequence of maps

$$F_n : \mathcal{O}_T \times \mathbb{R} \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$$

each  $u^n$  a subsolution of  $F_n(t, x, v, \partial_t v, D^2 v) = 0$ . Assume  $(u^n)_n$  and  $(F_n)_n$  are locally uniformly bounded. Then

$\underline{u}(t, x) = \liminf_{(s, y) \rightarrow (t, x), n \rightarrow \infty} u^n(s, y)$  is a subsolution of

$$\underline{F}(t, x, \underline{u}, \partial_t \underline{u}, D^2 \underline{u}) = 0 \text{ on } \mathcal{O}$$

The analogous statement holds for a sequence of LSC  $(\mathcal{O}_T)$  functions which are supersolutions. Further, if  $\bar{u} = \underline{u}$  then the convergence of  $(u_n)$  to  $\bar{u} = \underline{u}$  is locally uniform.

## Theorem (O&Reis13)

Let  $\mu, \nu \in \mathcal{M}^2$ ,  $\mu \leq_{\text{cx}} \nu$  and denote with  $R$  the Root barrier for the SEP

$$dX_t = \sigma(t, X_t) dB_t, \quad X_0 \sim \mu, \quad X_{\tau_R} \sim \nu$$

Define

$$u(t, x) := -\mathbb{E}[|X_{t \wedge \tau_R} - x|].$$

Then  $u$  is the unique viscosity solution of the obstacle problem

$$\begin{cases} \min\left(u - u_\nu, \partial_t u - \frac{\sigma^2}{2} \Delta u\right) = 0, \\ u(0, \cdot) = u_\mu(\cdot). \end{cases} \quad (2)$$

Moreover,

1.  $t \mapsto u(t, x)$  is non-increasing and  $x \mapsto u(t, x)$  is Lipschitz (uniformly in time)
2.  $u_\nu(x) \leq u(t, x) \leq u_\mu(x)$ ,
3.  $\lim_{t \rightarrow \infty} u(t, x) = u_\nu(x)$ .

## Proof (Sketch).

► Show

$$\left(\partial_t - \frac{\sigma^2}{2}\Delta\right) u \geq 0 \text{ on } [0, \infty) \times \mathbb{R},$$

$$u - u_\nu \geq 0 \text{ on } [0, \infty) \times \mathbb{R},$$

$$\left(\partial_t - \frac{\sigma^2}{2}\Delta\right) u = 0 \text{ on } R^c,$$

$$u - u_\nu = 0 \text{ on } R,$$

► **Step 1.**  $u - U_\nu \geq 0$ . By Jensen

$$\begin{aligned} u(t, x) &= -\mathbb{E}[|X_t^{TR} - x|] \geq \mathbb{E}[\mathbb{E}[-|X_{TR} - x| | \mathcal{F}_{t \wedge TR}]] \\ &= -\mathbb{E}[|X_{TR} - x|] = u_\nu(x) \end{aligned}$$



- ▶ Take  $(\psi_n), \psi_n \in C^2(\mathbb{R}, \mathbb{R})$ ,  $\psi_n \rightarrow |\cdot|$  uniformly,  $\Delta\psi_n(\cdot)$  is continuous,  $\Delta\psi_n \geq 0$  and  $\text{supp}(\Delta\psi_n) \subset [-\frac{1}{n}, \frac{1}{n}]$ ;
- ▶ Define

$$\begin{aligned} u^n(t, x) &= -\mathbb{E}[\psi^n(X_t^\tau - x)] \rightarrow_{n \rightarrow \infty} u(t, x) \\ u_\mu^n &= -\mathbb{E}[\psi^n(X_0 - x)] \rightarrow_{n \rightarrow \infty} u_\mu(x) \\ u_\nu^n &= -\mathbb{E}[\psi^n(X_\tau - x)] \rightarrow_{n \rightarrow \infty} u_\nu(x) \end{aligned}$$

- ▶ Apply Ito to  $\psi^n(X_{\cdot}^{\tau R} - x)$

$$u^n(t, x) = u_\mu^n(x) - \int_0^t \mathbb{E} \left[ \frac{\sigma^2(r, X_r)}{2} \Delta\psi^n(X_r - x) \mathbf{1}_{r < \tau R} \right] dr$$

► **Step 2.**  $u - U_\mu = 0$  on  $R$

- By Ito, applied to  $\psi^n(X_r - x)$

$$u^n(t, x) = u_\mu^n(x) - \int_0^t \mathbb{E} \left[ \frac{\sigma^2(r, X_r)}{2} \Delta \psi^n(X_r - x) 1_{r < \tau_R} \right] dr$$

Take  $\lim_{t \rightarrow \infty}$

$$u_\nu^n(x) = u_\mu^n(x) - \int_0^\infty \mathbb{E} \left[ \frac{\sigma^2(r, X_r)}{2} \Delta \psi^n(X_r - x) 1_{r < \tau_R} \right] dr$$

- Hence

$$u^n(t, x) - u_\nu^n(x) = \int_t^\infty \mathbb{E} \left[ \frac{\sigma^2(r, X_r)}{2} \Delta \psi^n(X_r - x) 1_{r < \tau_R} \right] dr$$

- Fix  $(t, x) \in R^o$ , then  $(t+r, x) \in R^o$  for  $r \geq 0$  hence

$$\begin{aligned} \lim_{n \rightarrow \infty} (u^n - u_\nu^n)(t, x) &= \int_t^\infty \mathbb{E} \left[ \frac{\sigma^2(r, X_r)}{2} \lim_{n \rightarrow \infty} \Delta \psi^n(X_r - x) 1_{r < \tau_R} \right] dr \\ &= 0 \end{aligned}$$

**Step 3.**  $(\partial_t - \frac{\sigma^2}{2} \Delta) u \geq 0$  on  $[0, \infty) \times \mathbb{R}$ .

► First show that  $u^n$  solves

$$\begin{cases} (\partial_t - \frac{\sigma^2}{2} \Delta) u^n - \frac{1}{2} l_n = 0 & \text{on } (0, \infty) \times \mathbb{R} \\ u^n(0, \cdot) = u_{\mu}^n(\cdot). \end{cases}$$

- $u^n(t, x)$  has a right- and left derivative  $\forall (t, x) \in [0, \infty) \times \mathbb{R}$  and continuous derivative  $\Delta u^n(t, x)$
- $\lim_{n \rightarrow 0} l^n = 0$  loc. uniformly

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- $u^n(t, x)$  has a right- and left derivative  $\forall (t, x) \in [0, \infty) \times \mathbb{R}$  and continuous derivative  $\Delta u^n(t, x)$
  - $\lim_{n \rightarrow 0} l^n = 0$  loc. uniformly
- Apply the method of semi-relaxed limits:  $u^n \rightarrow u$  and  $u_\mu^n(\cdot) \rightarrow u_\mu(\cdot)$  uniformly hence  $u$  is viscosity supersolution of

$$\begin{cases} (\partial_t - \frac{\sigma^2}{2} \Delta) u = 0 & \text{on } [0, \infty) \times \mathbb{R} \\ u(0, \cdot) = u_\mu(\cdot) \end{cases}$$

**Step 4.**  $(\partial_t - \frac{\sigma^2}{2} \Delta) u = 0$  on  $R^c$ :

- ▶ need to show that  $u$  is a subsolution (supersolution follows from above)
- ▶  $R$  is a Root barrier, hence

$$(\tau_R + r, X_{\tau_R}) \in R \quad \forall r \geq 0,$$

hence if  $(t, x) \in R^c$  and  $t \geq \tau_R$  then  $X_{\tau_R} \neq x$ . Therefore

$$\lim_{n \rightarrow \infty} \sup_{(t,x) \in K} \Delta \psi^n (X_{\tau_R} - x) \mathbf{1}_{t \geq \tau_R} = 0 \quad \text{for every compact } K \subset R^c$$

c) From PDE to barrier

## Proposition

Under (TC) there exists a  $v_\infty \in C([0, \infty) \times \mathbb{R}, \mathbb{R})$  which is a viscosity solution of

$$\begin{cases} \min \left( v - h, \partial_t v - \frac{\sigma^2}{2} \Delta v \right) = 0, & (t, x) \in (0, \infty) \times \mathbb{R} \\ v(0, x) = u_0(x), & x \in \mathbb{R}. \end{cases}$$

Moreover, for  $T < \infty$

1.  $\forall (t, x) \in [0, T] \times \mathbb{R} \exists!$   $(X_s^{t,x}, Y_s^{t,x}, Z_s^{t,x}, K_s^{t,x})_{s \in [t, T]}$  of  $\{\mathcal{F}_s^t\}$ -progressively measurable processes, solution of the rFBSDE

$$\begin{cases} X_s^{t,x} &= x + \int_t^s \sigma(T-r, X_r^{t,x}) dW_r, \\ Y_s^{t,x} &= u_0(X_T^{t,x}) + K_T^{t,x} - K_s^{t,x} - \int_s^T Z_r^{t,x} dW_r, \\ Y_s^{t,x} &\geq h(X_s^{t,x}), t < s \leq T, \int_t^T (Y_s^{t,x} - h(X_s^{t,x})) dK_s^{t,x} = 0, \end{cases}$$

2.  $(K_s^{t,x})_{s \in [t, T]}$  increasing, continuous, and  $K_t^{t,x} = 0$ .
3.  $v_\infty|_{[0, T] \times \mathbb{R}}(t, x) \equiv Y_{T-t}^{T-t, x}$ .

## Theorem (O&Reis13)

Let  $\mu, \nu$  have a second moment,  $\mu \leq_{cx} \nu$  and  $\sigma$  Lip+LG. Then the free boundary  $R$

$$R = \{(t, x) : u(t, x) = u_\nu(x)\}$$

of the obstacle problem

$$\begin{cases} \min \left( u - u_\nu, \partial_t u - \frac{\sigma^2}{2} \Delta u \right) = 0, \\ u(0, \cdot) = u_\mu(\cdot). \end{cases} \quad (3)$$

solves the SEP

$$dX_t = \sigma(t, X_t) dB_t, \quad X_0 \sim \mu, \quad X_{TR} \sim \nu.$$



Proof.

(sketch)

- ▶ Under above assumptions the PDE has exactly one solution

$$\mathcal{O}(u_\mu, u_\nu, \sigma) = \{u\}$$

- ▶  $\mathcal{R}(\mu, \nu, \sigma) = \{R\}$

- ▶ From previous Theorem  $u(t, x) = -\mathbb{E}[|X_{t \wedge \tau_R} - x|]$

- ▶ By proof of previous theorem, hence

$$R \subset \{(t, x) : u(t, x) = u_\nu(x)\}$$

- ▶ To see  $R \supset \{(t, x) : u(t, x) = u_\nu(x)\}$  use the representation

$$u(t+r, x) - u(t, x) = \mathbb{E}[L_t^x - L_{t+r}^x]$$

□

(d) Numerics

Assume  $\mu, \nu$  support in interval  $[a, b] \subset \mathbb{R}$

$$S^h [u^h](t, x) := \begin{cases} u_\mu(x) & \text{in } [0, \Delta t) \times (a, b) \\ u^h(t, x) + \frac{\Delta t \sigma^h(t, x)}{2(\Delta x)^2} (u^h(t, x + \Delta x) - 2u^h(t, x) + u^h(t, x - \Delta x)) & \text{in } [0, T] \times (a, b) \\ u_\mu(x) = u_\nu(x) & \text{in } [0, T] \times \{a, b\} \end{cases}$$

### Proposition (O&Reis13)

Let  $T < \infty$  and let  $\mu, \nu$  have second moments, compact support,  $\mu \leq_{cx} \nu$ . If  $\Delta t |\sigma|_{\infty; [a, b] \times [0, T]} < (\Delta x)^2$  Then  $u^h \in \mathcal{B}([0, T] \times \mathbb{R}, \mathbb{R})$  and

$$\left| u^h - u \right|_{\infty; [0, T] \times \mathbb{R}} \rightarrow 0 \text{ as } h \rightarrow (0, 0)$$

on  $[0, T]$ ,  $\{u\} \in \mathcal{O}(u_\mu, u_\nu, \sigma)$ .

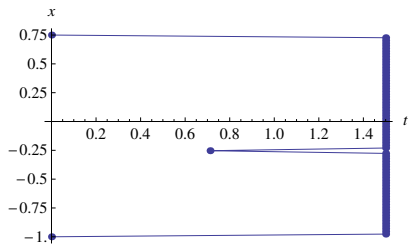
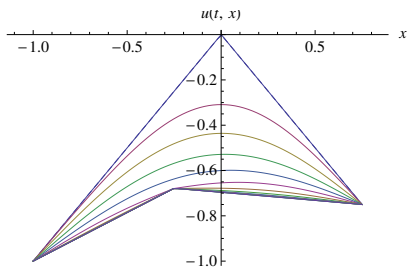


Figure:  $\sigma = 1, \mu = \delta_0$  and  $\nu = \frac{2}{7}\delta_{-1} + \frac{1}{4}\delta_{-\frac{1}{4}} + \frac{13}{28}\delta_{\frac{3}{4}}$ .

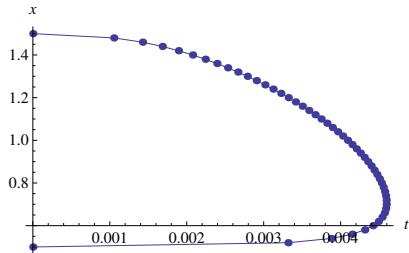
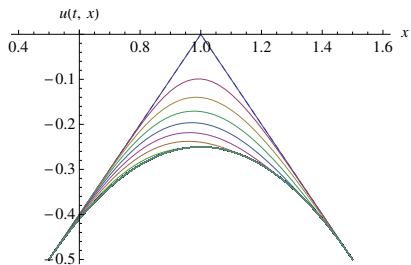


Figure:  $\sigma(x) = x$ ,  $\mu = \delta_1$  and  $\nu = \mathcal{U}\left(\left[\frac{1}{2}, \frac{3}{2}\right]\right)$

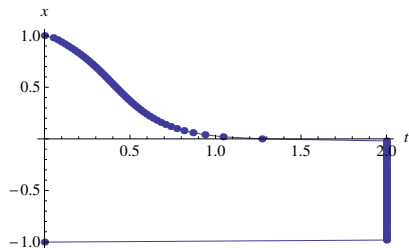
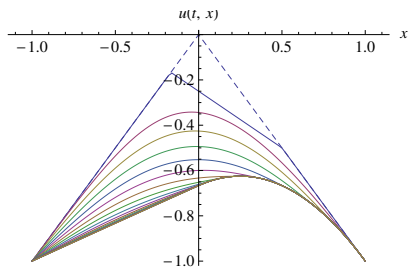


Figure:  $\sigma(x) = 1, \mu = \frac{3}{4}\delta_{-\frac{1}{6}} + \frac{1}{4}\delta_{0.5}$ ,  $\nu = \frac{1}{3}\delta_{-1} + \frac{2}{3}\mathcal{U}([0, 1])$ .

[1, 2, 3, 4, 5, 6, 7]



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THANK YOU FOR YOUR TIME!