

From 1 to 6 in Branching Brownian Motion

Lisa Hartung (joint with A. Bovier)

Courant Institute, NYU

Advances in Statistical Mechanics, CIRM, 27.08.2018



Outline

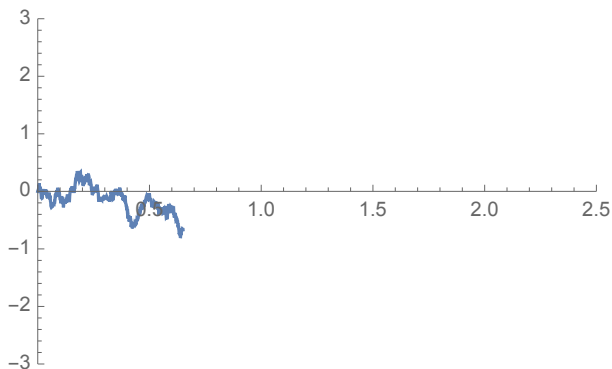
- 1 Definition variable speed BBM
- 2 Previous results
- 3 Closing in on the straight line
- 4 Main results
- 5 Ideas of proof

Definition BBM

1. Start a Brownian motion x in 0 .

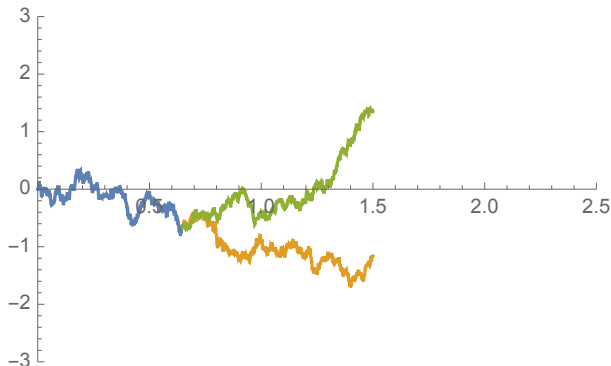
Definition BBM

1. Start a Brownian motion x in 0.



Definition BBM

2. After exponential holding time T particle splits into 2 offsprings.
3. Each performs independent Brownian motion starting at $x(T)$.

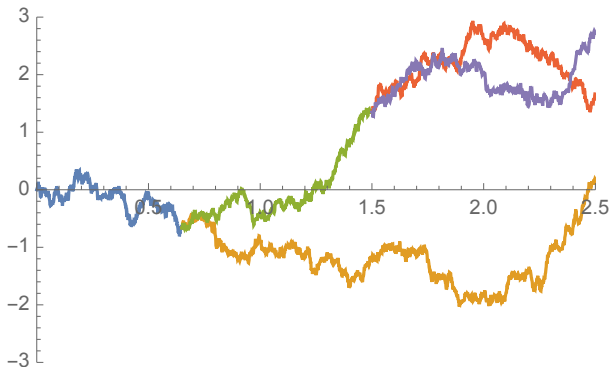


Definition BBM

4. The new particles are subject of the same splitting rule.

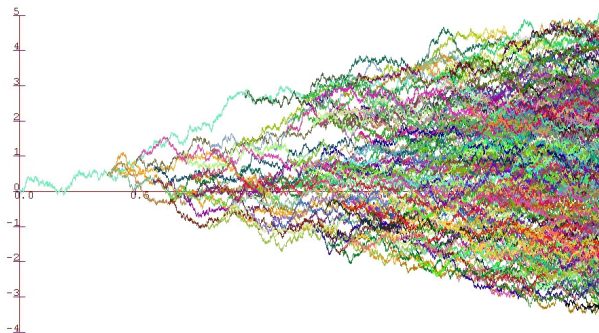
Definition BBM

- The new particles are subject of the same splitting rule.



Definition BBM

And after some time...



Picture by Matt Roberts, Bath

Why are we interested in this process?

- Prototype continuous spatial branching process
- Connection to F-KPP equation
- Extreme value theory for correlated Gaussian processes

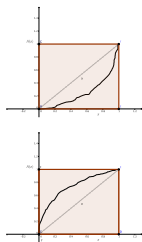
Variable speed BBM

Let $A : [0, 1] \rightarrow [0, 1]$ be increasing. Define

$$\Sigma^2(s) = tA(s/t).$$

Brownian motion with speed function Σ^2

$$B_s^\Sigma = B_{\Sigma^2(s)}.$$



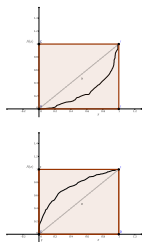
Variable speed BBM

Let $A : [0, 1] \rightarrow [0, 1]$ be increasing. Define

$$\Sigma^2(s) = tA(s/t).$$

Brownian motion with speed function Σ^2

$$B_s^\Sigma = B_{\Sigma^2(s)}.$$



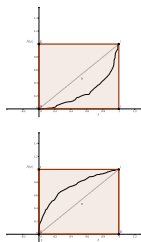
Variable speed BBM

Let $A : [0, 1] \rightarrow [0, 1]$ be increasing. Define

$$\Sigma^2(s) = tA(s/t).$$

Brownian motion with speed function Σ^2

$$B_s^\Sigma = B_{\Sigma^2(s)}.$$



Variable speed BBM:

same splitting rules, but if a particle splits at time $s < t$:

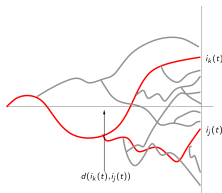
law of movement independent copies of $\{B_r^\Sigma - B_s^\Sigma\}_{t \geq r \geq s}$

Example for Gaussian process labelled by tree

- A time-homogeneous tree. Label individuals at time t as $\mathbf{i}_1(t), \dots, \mathbf{i}_{n(t)}(t)$.

Example for Gaussian process labelled by tree

- A time-homogeneous tree. Label individuals at time t as $\mathbf{i}_1(t), \dots, \mathbf{i}_{n(t)}(t)$.
- Canonical tree-distance: $d(\mathbf{i}_\ell(t), \mathbf{i}_k(t)) \equiv$ time of most recent common ancestor of $\mathbf{i}_\ell(t)$ and $\mathbf{i}_k(t)$

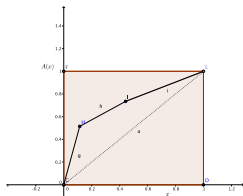
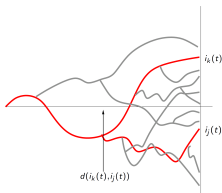


Example for Gaussian process labelled by tree

- A time-homogeneous tree. Label individuals at time t as $\mathbf{i}_1(t), \dots, \mathbf{i}_{n(t)}(t)$.
- Canonical tree-distance: $d(\mathbf{i}_\ell(t), \mathbf{i}_k(t)) \equiv$ time of most recent common ancestor of $\mathbf{i}_\ell(t)$ and $\mathbf{i}_k(t)$
- For fixed time horizon t , define **Gaussian process**, $(x_k^t(s), k \leq n(t), s \leq t)$, with covariance

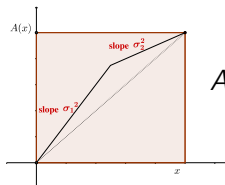
$$\mathbb{E}x_k^t(r)x_\ell^t(s) = tA(t^{-1}d(\mathbf{i}_k(r), \mathbf{i}_\ell(s)))$$

for $A : [0, 1] \rightarrow [0, 1]$, increasing.



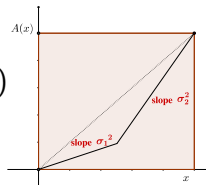
Previous results

In this talk we focus for clarity on the case of **two-speed BBM**, where



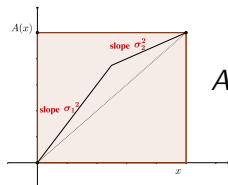
$$A(x) = \sigma_1^2 x + \mathbb{1}_{x \geq 1/2} (x - 1/2) (\sigma_2^2 - \sigma_1^2)$$

$$\sigma_1^2 + \sigma_2^2 = 2$$



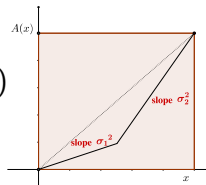
Previous results

In this talk we focus for clarity on the case of **two-speed BBM**, where



$$A(x) = \sigma_1^2 x + \mathbb{1}_{x \geq 1/2} (x - 1/2) (\sigma_2^2 - \sigma_1^2)$$

$$\sigma_1^2 + \sigma_2^2 = 2$$



Convergence of the maximum:

$$\lim_{t \uparrow \infty} \mathbb{P} \left(\max_{k \leq n(t)} x_k(t) \leq m(t) + y \right) = \mathbb{E} \left[e^{-CV e^{-\sqrt{2}y}} \right]$$

with

- $m(t) = m_{\sigma_1}(t)$ depending on σ_1
- $C = C(\sigma_1)$ a **constant** depending on σ_1
- $V = V(\sigma_1)$ a **random variable** depending on σ_1

Typical results

Moreover:

- $m_{\sigma_1}(t) = \sqrt{2}M(\sigma_1)t - \frac{c(\sigma_1)}{2\sqrt{2}} \ln t$, where

1

$$M(\sigma_1) = \begin{cases} 1, & \text{if } \sigma_1^2 \leq 1, \\ \frac{\sigma_1 + \sigma_2}{2}, & \text{if } \sigma_1^2 \geq 1, \end{cases}$$

is **continuous** in σ_1

2

$$c(\sigma_1) = \begin{cases} 1, & \text{if } \sigma_1^2 < 1, \\ 3, & \text{if } \sigma_1^2 = 1, \\ 3(\sigma_1 + \sigma_2), & \text{if } \sigma_1^2 > 1, \end{cases}$$

is **discontinuous** at $\sigma_1^2 = 1$.

Typical results

And the random variable:

- V is the limit of the **McKean martingale** if $\sigma_1^2 < 1$
- V is the limit of the **derivative martingale** if $\sigma_1^2 \geq 1$

and the constant:

- $C \equiv \lim_{r \uparrow \infty} \sqrt{\frac{2}{\pi}} e^{-a^2 r/2} \int_0^\infty u(r, y + \sqrt{2}r) e^{(\sqrt{2}+a)y} (1 - e^{-2ay}) dy$, with $a = \sqrt{2}(\sigma_2 - 1)$ if $\sigma_1^2 < 1$, u solution of $F - KPP$
- $C \equiv \lim_{r \uparrow \infty} \sqrt{\frac{2}{\pi}} \int_0^\infty u(r, y + \sqrt{2}r) e^{\sqrt{2}y} y dy$, if $\sigma_1^2 = 1$, u solution of $F - KPP$
- More complicated, if $\sigma_1^2 > 1$.

Closing in on the discontinuity

We have seen that various quantities behave **discontinuously** at the borderline case $A(x) = x$.

To analyse this singular behaviour more closely, we consider functions A that depend on t , specifically,

$$\sigma_1^2 = \sigma_1^2(t) \equiv 1 \pm t^{-\alpha}, \quad \alpha > 0$$

$$\sigma_2^2 = \sigma_2^2(t) \equiv 1 \mp t^{-\alpha}, \quad \alpha > 0$$

Closing in on the discontinuity

Theorem [B, H, '18] With the notation above, the following facts hold:

- 1 If $\alpha \geq 1/2$, then everything is the same as in BBM ($\sigma_1^2 = 1$)
- 2 If $\alpha \in (0, 1/2)$, $\sigma_1^2 = 1 - t^{-\alpha}$,



$$m(t) = \sqrt{2}t - \frac{1 + 4\alpha}{2\sqrt{2}} \ln t$$

- ▶ V is the derivative martingale
- ▶ C is the same as in *BBM*

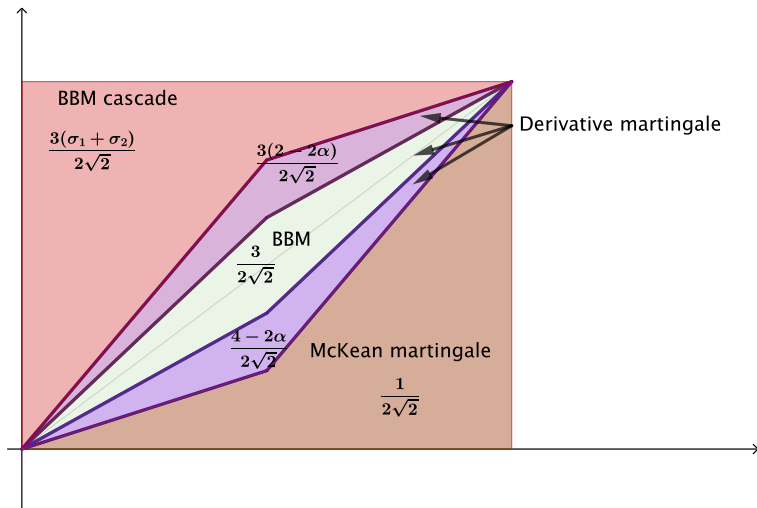
- 3 If $\alpha \in (0, 1/2)$, $\sigma_1^2 = 1 + t^{-\alpha}$,



$$m(t) = \sqrt{2}t \frac{\sigma_1 + \sigma_2}{2} - \frac{6(1 - \alpha)}{2\sqrt{2}} \ln t$$

- ▶ V is the derivative martingale
- ▶ C is $2/\sqrt{\pi}$ times the constant in *BBM*

The phase diagram



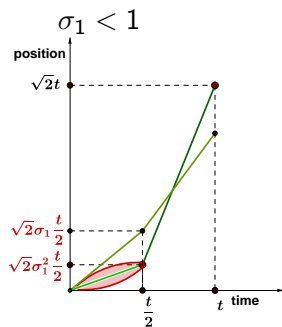
Ingredients of the proof

The proof is based on three basic ingredients:

- 1 Localisation of the ancestors of extremal particles at the time of the speed change
- 2 Tree recursions
- 3 Tail asymptotics

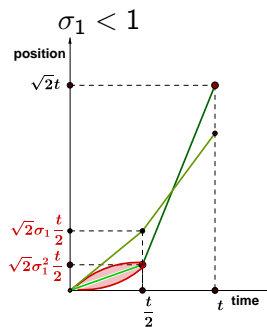
Localisation: the t -independent case

Two-speed BBM

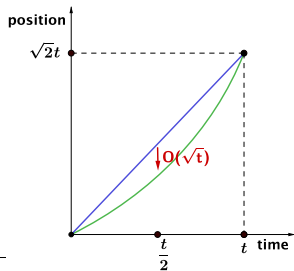


Localisation: the t -independent case

Two-speed BBM



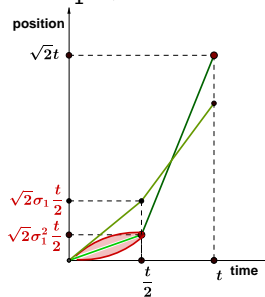
Standard BBM



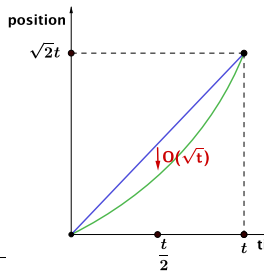
Localisation: the t -independent case

Two-speed BBM

$$\sigma_1 < 1$$

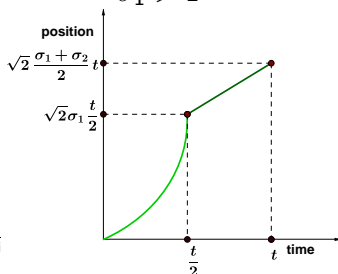


Standard BBM



Two-speed BBM

$$\sigma_1 > 1$$



A barrier for (standard) BBM

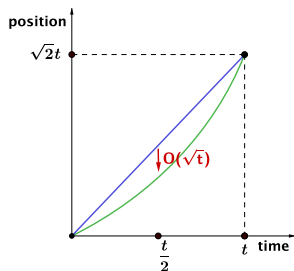
A priori information on all paths (Lalley-Sellke)

A barrier for (standard) BBM

A priori information on all paths (Lalley-Sellke)

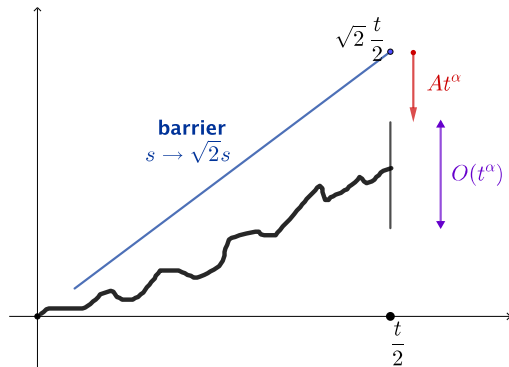
With high probability

$$\forall k \leq n(t) \forall r < s \leq t : x_k(s) < \sqrt{2}s$$



- All particles have to stay essentially below blue line
- Path of maximal particle \approx Brownian bridge $0 \rightarrow \sqrt{2}t$ in time t staying below straight line
 \Rightarrow Path is $\approx O(\sqrt{t})$ **below barrier**

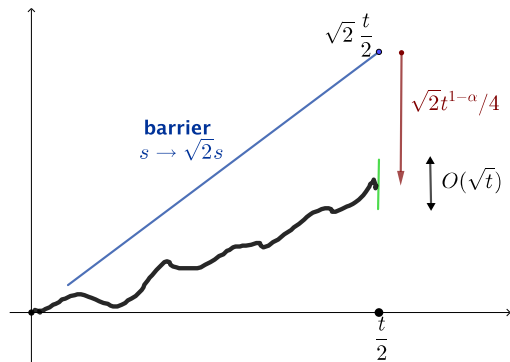
Localisation: $\sigma_1^2 = 1 + t^{-\alpha}$



To reach height $m(t)$ at time t

- Stay below barrier up to time $t/2$
- Be $O(t^\alpha)$ below $\sqrt{2} \frac{t}{2}$ at time $\frac{t}{2}$

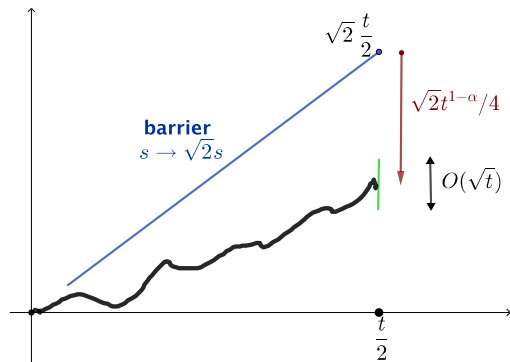
Localisation: $\sigma_1^2 = 1 - t^{-\alpha}$



To reach height $m(t)$ at time t

- Stay below barrier up to time $t/2$
- Be $\frac{t^{1-\alpha}}{2\sqrt{2}} \pm O(\sqrt{t})$ below $\sqrt{2} \frac{t}{2}$ at time $t/2$

Localisation: $\sigma_1^2 = 1 - t^{-\alpha}$

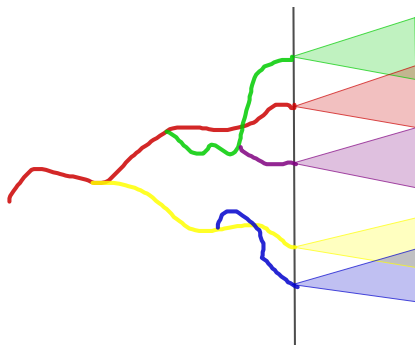


To reach height $m(t)$ at time t

- Stay below barrier up to time $t/2$
- Be $\frac{t^{1-\alpha}}{2\sqrt{2}} \pm O(\sqrt{t})$ below $\sqrt{2}\frac{t}{2}$ at time $t/2$

Localisation impose the size of the logarithmic corrections via probabilities of Brownian bridges to satisfy barrier conditions!

Branching property



$$\mathbb{P} \left[\max_{1 \leq k \leq n(t)} \tilde{x}_k(t) - m(t) \geq y \right]$$

$$= \mathbb{P} \left[\max_{\substack{k \leq n(t/2) \\ \ell \leq n^k(t/2)}} \sigma_1 x_k(t/2) + \sigma_2 x_\ell^k(t/2) - m(t) \geq y \right],$$

Asymptotics

Lemma (Bramson '83)

For $x = x(t)$ such that $\lim_{t \uparrow \infty} x(t)/t = 0$

$$\mathbb{P} \left(\max_{k \leq n(t)} x_k(t) > x + \sqrt{2t} \right) \sim Cx e^{-\sqrt{2}x} e^{-x^2/2t} t^{-3/2}, \quad \text{as } t \uparrow \infty$$

where C is a strictly positive constant.

Combining everything

$$\mathbb{P} \left[\max_{\substack{k \leq n(t/2) \\ \ell \leq n^k(t/2)}} \sigma_1 x_k(t/2) + \sigma_2 x_\ell^k(t/2) - m(t) \geq y \right],$$

- Use tail asymptotics.
- Use localization.
- Use the independence given through the branching property to put the pieces together.

Concluding remarks

- Analogous results for the extremal process: Same as in BBM

Concluding remarks

- Analogous results for the extremal process: **Same as in BBM**
- **Kistler and Schmidt** have shown a transition from 1 to 3 with a sequence of **step functions** with step length $t^{-1/(1-\alpha)}$ converging to the straight line. Here extremal process remains **Poisson** for all $\alpha < 1$.

Concluding remarks

- Analogous results for the extremal process: **Same as in BBM**
- **Kistler and Schmidt** have shown a transition from 1 to 3 with a sequence of **step functions** with step length $t^{-1/(1-\alpha)}$ converging to the straight line. Here extremal process remains **Poisson** for all $\alpha < 1$.
- One can consider a lot of different sequences of functions A_t that converge to $A(x) = x$ and obtain lots of different rescalings.
- Gaussian processes considered as functions of the function A have a very **complex discontinuity** at the identity function. BBM and other log-correlated processes are **natural borderlines**.

Thank you for your attention!