

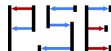
Extremal Processes of Gaussian Processes Indexed by Trees

Anton Bovier
with Louis-Pierre Arguin, Nicola Kistler
and Lisa Hartung

Institute for Applied Mathematics Bonn

EURANDOM, Eindhoven, 2014

hausdorff center for mathematics



PROBABILISTIC STRUCTURES
IN EVOLUTION
DFG SPP 1590

Plan

Plan

- 1 Gaussian processes on trees
- 2 Branching Brownian motion
- 3 The extremal process of BBM
- 4 Variable speed BBM
- 5 Universality

Motivation

Motivation

- **Spin glasses:** What is the structure of ground states for (mean field) spin glasses?

Motivation

- **Spin glasses:** What is the structure of ground states for (mean field) spin glasses?
- **Extreme value theory:** What are the extreme values and the extremal process of dependent random processes?

Motivation

- **Spin glasses:** What is the structure of ground states for (mean field) spin glasses?
- **Extreme value theory:** What are the extreme values and the extremal process of dependent random processes?
- **Spatial branching processes:** Describe the cloud of spatial branching processes, in particular near their propagation front!

Motivation

- **Spin glasses:** What is the structure of **ground states** for (mean field) spin glasses?
- **Extreme value theory:** What are the extreme values and the extremal process of dependent random processes?
- **Spatial branching processes:** Describe the cloud of spatial branching processes, in particular near their **propagation front!**
- **Reaction diffusion equations:** Characterise convergence to travelling wave solutions in certain non-linear pdes!

Motivation

- **Spin glasses:** What is the structure of **ground states** for (mean field) spin glasses?
- **Extreme value theory:** What are the extreme values and the extremal process of dependent random processes?
- **Spatial branching processes:** Describe the cloud of spatial branching processes, in particular near their **propagation front!**
- **Reaction diffusion equations:** Characterise convergence to travelling wave solutions in certain non-linear pdes!

This is too hard in general, but we will look at a setting where these questions have a chance to be answered. **Branching Brownian motion** is at the heart of this setting.

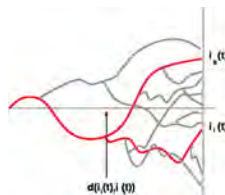
Gaussian processes labelled by trees

Gaussian processes labelled by trees

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

Gaussian processes labelled by trees

- 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 $\mathbf{i}_k(t)$

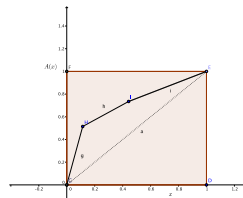
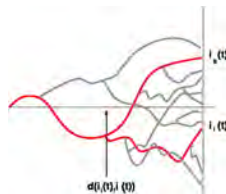


Gaussian processes labelled by trees

- 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.

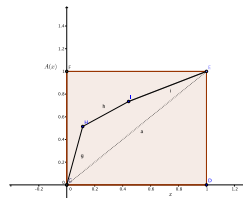
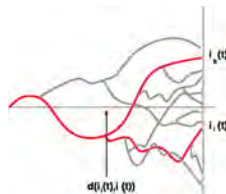


Gaussian processes labelled by trees

- 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.

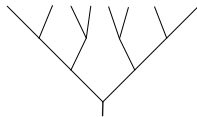


Can be constructed as time change of **branching Brownian motion**

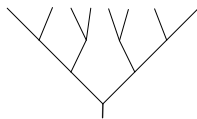
Examples

Examples

Binary tree, branching at integer times



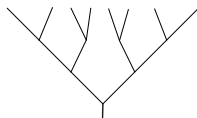
Examples



Binary tree, branching at integer times

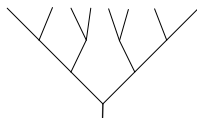
- $A(x) = x$: Branching random walk [Harris '63]

Examples



Binary tree, branching at integer times

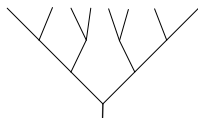
- $A(x) = x$: Branching random walk [Harris '63]
- A step function: Generalised Random Energy models (GREM) [Gardner-Derrida '82]



Examples

Binary tree, branching at integer times

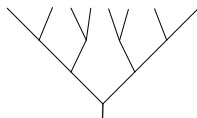
- $A(x) = x$: Branching random walk [Harris '63]
- A step function: Generalised Random Energy models (GREM) [Gardner-Derrida '82]
- Special case $A(x) = 0, x < 1, A(1) = 1$: Random energy model (REM), i.e. $n(t)$ iid $\mathcal{N}(0, t)$ r.v.s



Examples

Binary tree, branching at integer times

- $A(x) = x$: Branching random walk [Harris '63]
- A step function: Generalised Random Energy models (GREM) [Gardner-Derrida '82]
- Special case $A(x) = 0, x < 1, A(1) = 1$: Random energy model (REM), i.e. $n(t)$ iid $\mathcal{N}(0, t)$ r.v.s
- Continuous A : CREM [Gardner-Derrida '82, B-Kurkova '04]

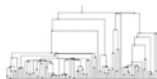


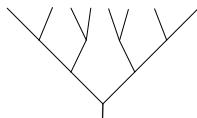
Examples

Binary tree, branching at integer times

- $A(x) = x$: Branching random walk [Harris '63]
- A step function: Generalised Random Energy models (GREM) [Gardner-Derrida '82]
- Special case $A(x) = 0, x < 1, A(1) = 1$: Random energy model (REM), i.e. $n(t)$ iid $\mathcal{N}(0, t)$ r.v.s
- Continuous A : CREM [Gardner-Derrida '82, B-Kurkova '04]

Supercritical Galton-Watson tree





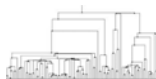
Examples

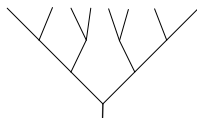
Binary tree, branching at integer times

- $A(x) = x$: Branching random walk [Harris '63]
- A step function: Generalised Random Energy models (GREM) [Gardner-Derrida '82]
- Special case $A(x) = 0, x < 1, A(1) = 1$: Random energy model (REM), i.e. $n(t)$ iid $\mathcal{N}(0, t)$ r.v.s
- Continuous A : CREM [Gardner-Derrida '82, B-Kurkova '04]

Supercritical Galton-Watson tree

- $A(x) = x$: Branching Brownian motion (BBM) [Moyal '62]

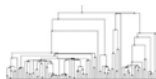




Examples

Binary tree, branching at integer times

- $A(x) = x$: Branching random walk [Harris '63]
- A step function: Generalised Random Energy models (GREM) [Gardner-Derrida '82]
- Special case $A(x) = 0, x < 1, A(1) = 1$: Random energy model (REM), i.e. $n(t)$ iid $\mathcal{N}(0, t)$ r.v.s
- Continuous A : CREM [Gardner-Derrida '82, B-Kurkova '04]



Supercritical Galton-Watson tree

- $A(x) = x$: Branching Brownian motion (BBM) [Moyal '62]
- General A : variable speed BBM [Derrida-Spohn '88, Fang-Zeitouni '12]

Extreme value theory

Extreme value theory

In the class of models we have described, we are interested in three main questions:

Extreme value theory

In the class of models we have described, we are interested in three main questions:

- How big is $M(t)/t \equiv \max_{k \leq n(t)} x_k(t)/t$, as $t \uparrow \infty$?

Extreme value theory

In the class of models we have described, we are interested in three main questions:

- How big is $M(t)/t \equiv \max_{k \leq n(t)} x_k(t)/t$, as $t \uparrow \infty$?
- Is there a rescaling $u_t(x)$, such that

$$\mathbb{P}(M(t) \leq u_t(x)) \rightarrow F(x)?$$

Extreme value theory

In the class of models we have described, we are interested in three main questions:

- How big is $M(t)/t \equiv \max_{k \leq n(t)} x_k(t)/t$, as $t \uparrow \infty$?
- Is there a rescaling $u_t(x)$, such that

$$\mathbb{P}(M(t) \leq u_t(x)) \rightarrow F(x)?$$

- Is there a limiting **extremal process**, \mathcal{P} , such that

$$\sum_{k \leq n(t)} \delta_{u_t^{-1}(x_k(t))} \rightarrow \mathcal{P}?$$

Reference: The REMs

Reference: The REMs

If $x_k(t)$ are just $n(t)$ iid Gaussian rv's with variance t :

Reference: The REMs

If $x_k(t)$ are just $n(t)$ iid Gaussian rv's with variance t :

- $M(t)/t \rightarrow \sqrt{2 \lim_{t \uparrow \infty} t^{-1} \ln n(t)} \equiv \sqrt{2r}$

Reference: The REMs

If $x_k(t)$ are just $n(t)$ iid Gaussian rv's with variance t :

- $M(t)/t \rightarrow \sqrt{2 \lim_{t \uparrow \infty} t^{-1} \ln n(t)} \equiv \sqrt{2r}$

With $u_t(x) = t\sqrt{2r} - \frac{\ln(rt)}{2\sqrt{2r}} + \frac{x}{\sqrt{r}} + \frac{\ln(n(t)/\mathbb{E}n(t))}{\sqrt{2r}}$, where $n(t)/\mathbb{E}n(t) \rightarrow RV$, a.s.

- $$\mathbb{P}(M(t) \leq u_t(x)) \rightarrow \exp\left(-\frac{1}{4\pi} e^{-\sqrt{2}x}\right)$$

Reference: The REMs

If $x_k(t)$ are just $n(t)$ iid Gaussian rv's with variance t :

- $M(t)/t \rightarrow \sqrt{2 \lim_{t \uparrow \infty} t^{-1} \ln n(t)} \equiv \sqrt{2r}$

With $u_t(x) = t\sqrt{2r} - \frac{\ln(rt)}{2\sqrt{2r}} + \frac{x}{\sqrt{r}} + \frac{\ln(n(t)/\mathbb{E}n(t))}{\sqrt{2r}}$, where $n(t)/\mathbb{E}n(t) \rightarrow RV$, a.s.

- $$\mathbb{P}(M(t) \leq u_t(x)) \rightarrow \exp\left(-\frac{1}{4\pi} e^{-\sqrt{2}x}\right)$$

- $$\sum_{k \leq n(t)} \delta_{u_t^{-1}(x_k(t))} \rightarrow \text{PPP}\left(\frac{1}{4\pi} e^{-\sqrt{2}x} dx\right)$$

where $\text{PPP}(\mu)$ denotes the **Poisson Point Process** with intensity μ .

Universality 1: the order of the maximum

Universality 1: the order of the maximum

In all models, it has been shown (or can be shown easily) that the order of the maximum is only a function of the **concave hull** of the function A (and on the growth rate of $n(t)$):

Universality 1: the order of the maximum

In all models, it has been shown (or can be shown easily) that the order of the maximum is only a function of the **concave hull** of the function A (and on the growth rate of $n(t)$):

If \bar{A} denotes the concave hull of A , then :

$$\lim_{t \rightarrow \infty} t^{-1} M(t) = \sqrt{2 \lim_{t \rightarrow \infty} t^{-1} \ln n(t)} \int_0^1 \sqrt{\frac{d}{ds} \bar{A}(s)} ds$$

[B-Kurkova 01, for binary tree, Fang-Zeitouni 11, GW tree]

Universality 1: the order of the maximum

In all models, it has been shown (or can be shown easily) that the order of the maximum is only a function of the **concave hull** of the function A (and on the growth rate of $n(t)$):

If \bar{A} denotes the concave hull of A , then :

$$\lim_{t \rightarrow \infty} t^{-1} M(t) = \sqrt{2 \lim_{t \rightarrow \infty} t^{-1} \ln n(t)} \int_0^1 \sqrt{\frac{d}{ds} \bar{A}(s)} ds$$

[B-Kurkova 01, for binary tree, Fang-Zeitouni 11, GW tree]

Note in particular that as long as $A(s) \leq s$, for all $s \leq 1$, then $\bar{A}(s) = s$, and the order of the maximum is the same as in the REM.

The GREM

The GREM

The full picture is known (or easy to get) if A is a **step function**. In that case:

The GREM

The full picture is known (or easy to get) if A is a **step function**. In that case:

- If $A(s) < s$, for all $s \in (0, 1)$, then **all** results are the same as in the corresponding REM!

The GREM

The full picture is known (or easy to get) if A is a **step function**. In that case:

- If $A(s) < s$, for all $s \in (0, 1)$, then **all** results are the same as in the corresponding REM!
- If $A(s) \leq s$, with equality in a finite number of points, the REM picture holds, but a prefactor appears in front of the e^{-x} 's.

The GREM

The full picture is known (or easy to get) if A is a **step function**. In that case:

- If $A(s) < s$, for all $s \in (0, 1)$, then **all** results are the same as in the corresponding REM!
- If $A(s) \leq s$, with equality in a finite number of points, the REM picture holds, but a prefactor appears in front of the e^{-x} 's.
- If $\bar{A}(s) \neq s$, then the leading order and the logarithmic correction are changed and depend on \bar{A} ; the extremal process is a **Poisson cascade process**.

The GREM

The full picture is known (or easy to get) if A is a **step function**. In that case:

- If $A(s) < s$, for all $s \in (0, 1)$, then **all** results are the same as in the corresponding REM!
- If $A(s) \leq s$, with equality in a finite number of points, the REM picture holds, but a prefactor appears in front of the e^{-x} 's.
- If $\bar{A}(s) \neq s$, then the leading order and the logarithmic correction are changed and depend on \bar{A} ; the extremal process is a **Poisson cascade process**.

This is all proven for the binary tree, but extension to general trees are straightforward.

The GREM

The full picture is known (or easy to get) if A is a **step function**. In that case:

- If $A(s) < s$, for all $s \in (0, 1)$, then **all** results are the same as in the corresponding REM!
- If $A(s) \leq s$, with equality in a finite number of points, the REM picture holds, but a prefactor appears in front of the e^{-x} 's.
- If $\bar{A}(s) \neq s$, then the leading order and the logarithmic correction are changed and depend on \bar{A} ; the extremal process is a **Poisson cascade process**.

This is all proven for the binary tree, but extension to general trees are straightforward.

Note the special role of the linear function $A(s) = s$



Branching Brownian motion

(BBM) is a classical object in probability, combining the standard models of **random motion** and **random genealogies** into one: Each particle of the Galton-Watson process performs Brownian motion independently of any other. This produces an immersion of the Galton-Watson process in space.

Branching Brownian motion



(BBM) is a classical object in probability, combining the standard models of **random motion** and **random genealogies** into one: Each particle of the Galton-Watson process performs Brownian motion independently of any other. This produces an immersion of the Galton-Watson process in space.



Picture by **Matt Roberts**, Bath

BBM is the canonical model of a spatial branching process.

The F-KPP



The F-KPP



One of the simplest **reaction-diffusion equations** is the Fisher-Kolmogorov-Petrovsky-Piscounov (F-KPP) equation:

$$\partial_t v(x, t) = \frac{1}{2} \partial_x^2 v(x, t) + v - v^2$$

The F-KPP



One of the simplest **reaction-diffusion equations** is the Fisher-Kolmogorov-Petrovsky-Piscounov (F-KPP) equation:

$$\partial_t v(x, t) = \frac{1}{2} \partial_x^2 v(x, t) + v - v^2$$

Fischer used this equation to model the evolution of biological populations. It accounts for:

- **birth:** v ,
- **death:** $-v^2$,
- **diffusive migration:** $\partial_x^2 v$.



F-KPP equation and BBM



F-KPP equation and BBM

Lemma (McKeane '75, Ikeda, Nagasawa, Watanabe '69)

Let $f : \mathbb{R} \rightarrow [0, 1]$ and $\{x_k(t) : k \leq n(t)\}$ BBM.

$$u(t, x) = \mathbb{E} \left[\prod_{k=1}^{n(t)} f(x - x_k(t)) \right]$$

Then $v \equiv 1 - u$ is the solution of the F-KPP equation with initial condition $v(0, x) = 1 - f(x)$.

Travelling waves





Travelling waves

Theorem (Bramson '78)

The equation

$$\frac{1}{2}\omega'' + \sqrt{2}\omega' - \omega^2 + \omega = 0.$$

has a unique solution satisfying $0 < \omega(x) < 1$, $\omega(x) \rightarrow 0$, as $x \rightarrow +\infty$, and $\omega(x) \rightarrow 1$, as $x \rightarrow -\infty$, up to translation, i.e. if ω, ω' are two solutions, then there exists $a \in \mathbb{R}$ s.t. $\omega'(x) = \omega(x + a)$.



Travelling waves

Theorem (Bramson '78)

The equation

$$\frac{1}{2}\omega'' + \sqrt{2}\omega' - \omega^2 + \omega = 0.$$

has a unique solution satisfying $0 < \omega(x) < 1$, $\omega(x) \rightarrow 0$, as $x \rightarrow +\infty$, and $\omega(x) \rightarrow 1$, as $x \rightarrow -\infty$, up to translation, i.e. if ω, ω' are two solutions, then there exists $a \in \mathbb{R}$ s.t. $\omega'(x) = \omega(x + a)$.

For suitable initial conditions,

$$u(t, x + m(t)) \rightarrow \omega(x),$$

where $m(t) = \sqrt{2}t - \frac{3}{2\sqrt{2}} \ln t$, where ω is one of the stationary solutions.

Examples

Examples

Choosing suitable initial conditions, this theorem applies to

Examples

Choosing suitable initial conditions, this theorem applies to

- $u(t, x) = \mathbb{P}(\max_{k \leq n(t)} x_k(t) \leq x),$

Examples

Choosing suitable initial conditions, this theorem applies to

- $u(t, x) = \mathbb{P}(\max_{k \leq n(t)} x_k(t) \leq x)$, and
- the Laplace functional $u(t, x) = \mathbb{E} \exp(-\sum_{k \leq n(t)} \phi(x_k(t)))$
Needs a bit extra work...

Examples

Choosing suitable initial conditions, this theorem applies to

- $u(t, x) = \mathbb{P}(\max_{k \leq n(t)} x_k(t) \leq x)$, and
- the Laplace functional $u(t, x) = \mathbb{E} \exp(-\sum_{k \leq n(t)} \phi(x_k(t)))$
Needs a bit extra work...

In particular, it gives Bramson's celebrated result

$$\lim_{t \rightarrow \infty} \mathbb{P}(\max_{k \leq n(t)} x_k(t) - m(t) \leq x) = \omega(x)$$

The derivative martingale





The derivative martingale

Lalley-Sellke, 1987: $\omega(x)$ is random shift of Gumbel-distribution

$$\omega(x) = \mathbb{E} \left[e^{-CZ e^{-\sqrt{2}x}} \right], \quad (*)$$



The derivative martingale

Lalley-Sellke, 1987: $\omega(x)$ is random shift of Gumbel-distribution

$$\omega(x) = \mathbb{E} \left[e^{-CZ e^{-\sqrt{2}x}} \right], \quad (*)$$

$Z \stackrel{(d)}{=} \lim_{t \rightarrow \infty} Z(t)$, where $Z(t)$ is the **derivative martingale**,

$$Z(t) = \sum_{k \leq n(t)} \{\sqrt{2}t - x_k(t)\} e^{-\sqrt{2}\{\sqrt{2}t - x_k(t)\}}$$



The derivative martingale

Lalley-Sellke, 1987: $\omega(x)$ is random shift of Gumbel-distribution

$$\omega(x) = \mathbb{E} \left[e^{-cZ e^{-\sqrt{2}x}} \right], \quad (*)$$

$Z \stackrel{(d)}{=} \lim_{t \rightarrow \infty} Z(t)$, where $Z(t)$ is the **derivative martingale**,

$$Z(t) = \sum_{k \leq n(t)} \{ \sqrt{2}t - x_k(t) \} e^{-\sqrt{2}\{ \sqrt{2}t - x_k(t) \}}$$

The form (*) seems universal, but Z is particular.

For the REM on the GW tree (*) holds with Z a standard exponential.

Description of the extremal process

Description of the extremal process

Poisson Point Process: $\mathcal{P}_Z = \sum_{i \in \mathbb{N}} \delta_{\rho_i} \equiv \text{PPP} \left(CZe^{-\sqrt{2}x} dx \right)$

Description of the extremal process

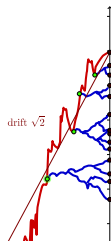
Poisson Point Process: $\mathcal{P}_Z = \sum_{i \in \mathbb{N}} \delta_{p_i} \equiv \text{PPP} \left(CZe^{-\sqrt{2}x} dx \right)$

Cluster process:

$$\Delta(t) \equiv \sum_k \delta_{x_k(t) - \max_{j \leq n(t)} x_j(t)}.$$

conditioned on the event $\{ \max_{j \leq n(t)} x_j(t) > \sqrt{2}t \}$
converges in law to point process, Δ .

[Chauvin, Rouault '90]



$$\mathcal{E} \equiv \sum_{i,j \in \mathbb{N}} \delta_{p_i + \Delta_j^{(i)}}, \quad \Delta^{(i)} \text{ iid copies of } \Delta$$

The extremal process

The extremal process

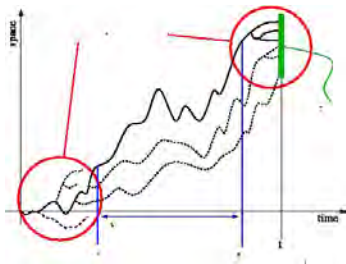
Theorem (Arguin-B-Kistler '11, Aidékon, Brunet, Berestycki, Shi '11)

The point process $\mathcal{E}_t \equiv \sum_{i=1}^{n(t)} \delta_{x_i(t)-m(t)} \rightarrow \mathcal{E}$.

The extremal process

Theorem (Arguin-B-Kistler '11, Aidékon, Brunet, Berestycki, Shi '11)

The point process $\mathcal{E}_t \equiv \sum_{i=1}^{n(t)} \delta_{x_i(t)-m(t)} \rightarrow \mathcal{E}$.



Interpretation:

p_i : positions of maxima of clusters with recent common ancestors.

$\Delta^{(i)}$: positions of members of clusters seen from their maximal one

The extremal process

The extremal process

Technically, proven by showing convergence of Laplace functionals:

The extremal process

Technically, proven by showing convergence of Laplace functionals:

$$\mathbb{E} \left[\exp \left(- \int \phi(y) \mathcal{E}_t(dy) \right) \right] \rightarrow \mathbb{E} [\exp(-C(\phi)Z)]$$

for any $\phi \in \mathcal{C}_c(\mathbb{R})$ non-negative, where

$$C(\phi) = \lim_{t \rightarrow \infty} \sqrt{\frac{2}{\pi}} \int_0^\infty (1 - u(t, y + \sqrt{2t})) y e^{\sqrt{2}y} dy$$

$u(t, y)$: solution of F-KPP with initial condition $u(0, y) = e^{-\phi(y)}$.

The extremal process

Technically, proven by showing convergence of Laplace functionals:

$$\mathbb{E} \left[\exp \left(- \int \phi(y) \mathcal{E}_t(dy) \right) \right] \rightarrow \mathbb{E} [\exp(-C(\phi)Z)]$$

for any $\phi \in \mathcal{C}_c(\mathbb{R})$ non-negative, where

$$C(\phi) = \lim_{t \rightarrow \infty} \sqrt{\frac{2}{\pi}} \int_0^\infty (1 - u(t, y + \sqrt{2t})) y e^{\sqrt{2}y} dy$$

$u(t, y)$: solution of F-KPP with initial condition $u(0, y) = e^{-\phi(y)}$.

Then show that the limit is the Laplace functional of the process \mathcal{E} described above.

Adding an extra dimension...

The following is inspired by an analogous result conjectured for the Gaussian free field by Biskup and Luidor.

Adding an extra dimension...

The following is inspired by an analogous result conjectured for the Gaussian free field by Biskup and Louidor.

Chose an embedding $\gamma : \{1, \dots, n(t)\} \rightarrow \mathbb{R}_+$, such that

$$|(\gamma(i_k(t)), -\gamma(i_j(t)))| \sim e^{-d(i_k(t), i_j(t))}$$

Define for $x \in \mathbb{R}_+$,

$$Z(r, t, u) \equiv \sum_{k: \gamma(i_k(r)) \leq u} \{\sqrt{2}t - x_k(t)\} e^{-\sqrt{2}\{\sqrt{2}t - x_k(t)\}}$$

$$\lim_{r \uparrow \infty} \lim_{t \uparrow \infty} Z(r, t, u) \rightarrow Z(u)$$

Adding an extra dimension...

The following is inspired by an analogous result conjectured for the Gaussian free field by Biskup and Louidor.

Chose an embedding $\gamma : \{1, \dots, n(t)\} \rightarrow \mathbb{R}_+$, such that

$$|(\gamma(i_k(t)), -\gamma(i_j(t)))| \sim e^{-d(i_k(t), i_j(t))}$$

Define for $x \in \mathbb{R}_+$,

$$Z(r, t, u) \equiv \sum_{k: \gamma(i_k(r)) \leq u} \{\sqrt{2}t - x_k(t)\} e^{-\sqrt{2}\{\sqrt{2}t - x_k(t)\}}$$

$$\lim_{r \uparrow \infty} \lim_{t \uparrow \infty} Z(r, t, u) \rightarrow Z(u)$$

Adding an extra dimension

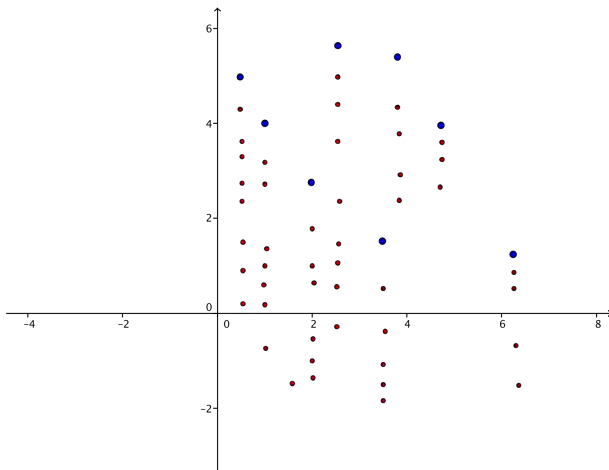
Theorem (B, Hartung '14)

The point process $\mathcal{E}_t \equiv \sum_{k=1}^{n(t)} \delta_{(\gamma(i_k(t)), x_i(t) - m(t))} \rightarrow \tilde{\mathcal{E}}$ on $\mathbb{R}_+ \times \mathbb{R}$, where

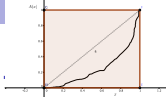
$$\tilde{\mathcal{E}} \equiv \sum_{i,j} \delta_{(q_i, p_i) + (0, \Delta_j^{(i)})},$$

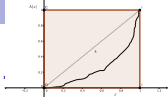
with (q_i, p_i) atoms of a Cox process on $\mathbb{R}_+ \times \mathbb{R}$ with intensity measure $Z(du) \times Ce^{-\sqrt{2}x} dx$, and $\Delta_j^{(i)}$ as before.

Adding another dimension



Variable speed BBM.....below the straight line...

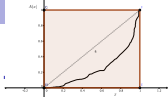




Variable speed BBM.....below the straight line...

Theorem (B-Hartung '13,'14)

Assume that $A(x) < x, \forall x \in (0, 1), A'(0) = a^2 < 1, A'(1) = b^2 > 1$.

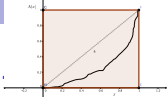


Variable speed BBM.....below the straight line...

Theorem (B-Hartung '13,'14)

Assume that $A(x) < x, \forall x \in (0, 1), A'(0) = a^2 < 1, A'(1) = b^2 > 1$.

Then $\exists C(b)$ and a r.v. Y_a such that



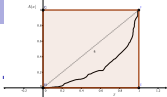
Variable speed BBM.....below the straight line...

Theorem (B-Hartung '13,'14)

Assume that $A(x) < x, \forall x \in (0, 1), A'(0) = a^2 < 1, A'(1) = b^2 > 1$.

Then $\exists C(b)$ and a r.v. Y_a such that

- $\mathbb{P}(M(t) - \tilde{m}(t) \leq x) \rightarrow \mathbb{E}e^{-C(b)Y_a e^{-\sqrt{2}x}}$



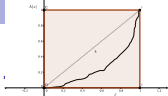
Variable speed BBM.....below the straight line...

Theorem (B-Hartung '13,'14)

Assume that $A(x) < x, \forall x \in (0, 1), A'(0) = a^2 < 1, A'(1) = b^2 > 1$.

Then $\exists C(b)$ and a r.v. Y_a such that

- $\mathbb{P}(M(t) - \tilde{m}(t) \leq x) \rightarrow \mathbb{E}e^{-C(b)Y_a e^{-\sqrt{2}x}}$
- $\sum_{k \leq n(t)} \delta_{x_k(t) - \tilde{m}(t)} \rightarrow \mathcal{E}_{a,b} = \sum_{i,j} \delta_{\rho_i + b\Delta_j^{(i)}}$



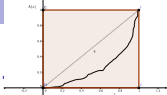
Variable speed BBM.....below the straight line...

Theorem (B-Hartung '13,'14)

Assume that $A(x) < x, \forall x \in (0, 1), A'(0) = a^2 < 1, A'(1) = b^2 > 1$.

Then $\exists C(b)$ and a r.v. Y_a such that

- $\mathbb{P}(M(t) - \tilde{m}(t) \leq x) \rightarrow \mathbb{E}e^{-C(b)Y_a e^{-\sqrt{2}x}}$
- $\sum_{k \leq n(t)} \delta_{x_k(t) - \tilde{m}(t)} \rightarrow \mathcal{E}_{a,b} = \sum_{i,j} \delta_{p_i + b\Delta_j^{(i)}}$
- $\tilde{m}(t) \equiv \sqrt{2}t - \frac{1}{2\sqrt{2}} \ln t$.
- p_i : e the atoms of a PPP($C(b)Y_a e^{-\sqrt{2}x} dx$),
- $Y_s \equiv \sum_{i=1}^{n(s)} e^{-s(1+\sigma_1^2) + \sqrt{2}x_i(s)}$



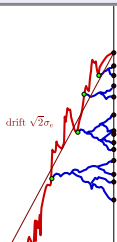
Variable speed BBM.....below the straight line...

Theorem (B-Hartung '13,'14)

Assume that $A(x) < x, \forall x \in (0, 1), A'(0) = a^2 < 1, A'(1) = b^2 > 1$.

Then $\exists C(b)$ and a r.v. Y_a such that

- $\mathbb{P}(M(t) - \tilde{m}(t) \leq x) \rightarrow \mathbb{E}e^{-C(b)Y_a e^{-\sqrt{2}x}}$
- $\sum_{k \leq n(t)} \delta_{x_k(t) - \tilde{m}(t)} \rightarrow \mathcal{E}_{a,b} = \sum_{i,j} \delta_{p_i + b\Delta_j^{(i)}}$
- $\tilde{m}(t) \equiv \sqrt{2}t - \frac{1}{2\sqrt{2}} \ln t$.
- p_i : e the atoms of a PPP($C(b)Y_a e^{-\sqrt{2}x} dx$),
- $Y_s \equiv \sum_{i=1}^{n(s)} e^{-s(1+\sigma_1^2) + \sqrt{2}x_i(s)}$
- Δ : are as in BBM but with the conditioning on the event $\{\max_k x_k(t) \geq \sqrt{2}bt\}$.



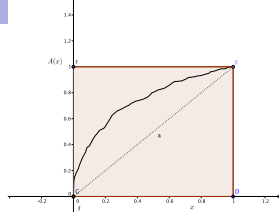
Elements of the proof:

Elements of the proof:

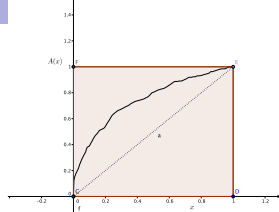
- 1) Explicit construction for the case of two speeds:
- 2) Gaussian comparison for general A .

For details, go to Lisa's talk (Friday, 9h)!!

Above the straight line

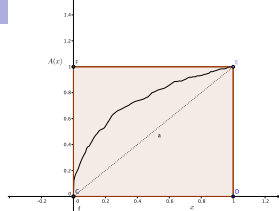


Above the straight line



When the concave hull of A is above the straight line, everything changes.

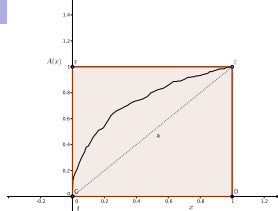
Above the straight line



When the concave hull of A is above the straight line, everything changes.

- If A is **piecewise linear**, it is quite easy to get the full picture:
Cascade of BBM processes.

Above the straight line



When the concave hull of A is above the straight line, everything changes.

- If A is **piecewise linear**, it is quite easy to get the full picture:
Cascade of BBM processes.
- If A is strictly concave, Fang and Zeitouni '12 and Maillard and Zeitouni '13 have shown that the correct rescaling is

$$m(t) = C_\sigma t - D_\sigma t^{1/3} - \sigma^2(1) \ln t$$

(with explicit constants C_σ and D_σ) but there are no explicit limit laws or limit processes available.

Universality

Universality

The **new extremal processes** should not be limited to BBM:

Universality

The **new extremal processes** should not be limited to BBM:

- **Branching random walk** [Bramson '78, Addario-Berry, Aïdékon '13 (law of max), Madaule '13 (full extremal process),...]
- **Gaussian free field in $d = 2$** [Bolthausen, Deuschel, Giacomin '01, Bramson-Ding-Zeitouni '13, Biskup-Loudon '13 [Poisson cluster extremes]]

Universality

The **new extremal processes** should not be limited to BBM:

- **Branching random walk** [Bramson '78, Addario-Berry, Aïdékon '13 (law of max), Madaule '13 (full extremal process),...]
- **Gaussian free field in $d = 2$** [Bolthausen, Deuschel, Giacomin '01, Bramson-Ding-Zeitouni '13, Biskup-Loudon '13 [Poisson cluster extremes]]
- **Cover times of random walks** [Lawler '93, Dembo-Peres-Rosen-Zeitouni '06, Belius-Kistler '14]

Universality

The **new extremal processes** should not be limited to BBM:

- **Branching random walk** [Bramson '78, Addario-Berry, Aïdékon '13 (law of max), Madaule '13 (full extremal process),...]
- **Gaussian free field in $d = 2$** [Bolthausen, Deuschel, Giacomin '01, Bramson-Ding-Zeitouni '13, Biskup-Loudon '13 [Poisson cluster extremes]]
- **Cover times** of random walks [Lawler '9,3 Dembo-Peres-Rosen-Zeitouni '06, Belius-Kistler '14]
- **Spin glasses** with log-correlated potentials [Fyodorov, Bouchaud '08, Arguin, Zindy '12..]

Universality

The **new extremal processes** should not be limited to BBM:

- **Branching random walk** [Bramson '78, Addario-Berry, Aïdékon '13 (law of max), Madaule '13 (full extremal process),...]
- **Gaussian free field in $d = 2$** [Bolthausen, Deuschel, Giacomin '01, Bramson-Ding-Zeitouni '13, Biskup-Loudon '13 [Poisson cluster extremes]]
- **Cover times** of random walks [Lawler '9,3 Dembo-Peres-Rosen-Zeitouni '06, Belius-Kistler '14]
- **Spin glasses** with log-correlated potentials [Fyodorov, Bouchaud '08, Arguin, Zindy '12..]
- **Statistics of zeros of Riemann zeta-function** [Fyodorov, Keating '12]

references

- L.-P. Arguin, A. Bovier, and N. Kistler, *The genealogy of extremal particles of branching Brownian motion*, Commun. Pure Appl. Math. **64**, 1647–1676 (2011).
- L.-P. Arguin, A. Bovier, and N. Kistler, *Poissonian statistics in the extremal process of branching Brownian motion*, Ann. Appl. Probab. **22**, 1693–1711 (2012).
- L.-P. Arguin, A. Bovier, and N. Kistler, *The extremal process of branching Brownian motion*, Probab. Theor. Rel. Fields **157**, 535–574 (2013).
- E. Aïdékon, J. Berestycki, É. Brunet, and Z. Shi, *Branching Brownian motion seen from its tip*, Probab. Theor. Rel. Fields **157**, 405–451 (2013).
- A. Bovier and L. Hartung, *The extremal process of two-speed branching Brownian motion*, EJP **19**, No 18 (2014)
- A. Bovier and L. Hartung, *Variable speed branching Brownian motion 1. Extremal processes in the weak correlation regime*, arxiv 1403.6332

Thank you for your attention!



hausdorff center for mathematics



universität bonn



PSE PROBLEMS OF STRUCTURE IN STOCHASTIC EVOLUTION