

Bálint Tóth:

Scaling limits for
self-interacting random
walks and diffusions, (3)

lecture notes for
Berlin-Zürich Summer
School on

Stochastic Models of
Complex processes

31 Aug-4 Sept 2009

Construction of the 1d scaling limit:
the "True Self-Repelling Motion" (TSRM)
[based on BST, W. Werner '98]

We have seen the limit theorem for

$$\frac{X([Nt])}{N^{2/3}} \quad \text{with } t \in \mathbb{R}_+ \text{ fixed.}$$

Next: $N^{-2/3} X([Nt]) \xrightarrow{?} X(t)$ as a process

In particular:

what could $t \mapsto X(t)$ be?

Basic properties of the scaling limit process $X(t)$

① $X(0) = 0,$

$t \mapsto X(t)$

a.s. continuous

a.s. recurrent

② Scale invariance

$$\left(X(t) \right)_{t \geq 0} \stackrel{\text{law}}{=} \left(\alpha^{2/3} X(t) \right)_{t \geq 0}$$

③ local counterpart:

local variation of power $3/2$:

$$\Theta_0^\varepsilon = 0; \quad \Theta_{n+1}^\varepsilon = \inf \left\{ t > \Theta_n^\varepsilon : |X(t) - X(\Theta_n^\varepsilon)| \geq \varepsilon \right\}$$

$$\varepsilon^{3/2} \sup \{ n : \Theta_n^\varepsilon < t \} \xrightarrow{P} \frac{2}{\sqrt{\pi}} \cdot t$$

④ existence and regularity of occupation time density

$$\mu_t(A) := \int_0^t \mathbb{1}(X(s) \in A) ds$$

$$L(t, x) := \frac{d\mu_t(dx)}{dx}$$

$(t, x) \mapsto L(t, x)$ a.s. continuous

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⑤ $t \mapsto (X(t), L(t, \cdot))$ (jointly)
Markov

More on this later...

⑥ **Dynamics:**
phenomenologically:

$$\begin{cases} dX(t) = -\partial_x L(t, X(t)) dt + \text{noise?} \\ \partial_x L(t, x) = \delta(x - X(t)) \end{cases} //$$

Comment...

Properly regularized:

$$X(T) + \int_0^T \frac{L(s, X(s) + \varepsilon) - L(s, X(s) - \varepsilon)}{2\varepsilon} ds$$

$$- \frac{1}{4} \left\{ \sup_{s < T} X(s) + \inf_{s < T} X(s) \right\}$$

$$\xrightarrow{P} 0$$

Construction: [inspired by the 1]

let $X(t)$ be a process with regular
(a.s. continuous) occupation time density $H(t, x)$

$$T(x, h) := \inf\{t \geq 0 : H(t, x) \geq h\} \quad \begin{array}{l} \text{inverse} \\ \text{local time} \end{array}$$

$$\Lambda_{xh}(y) := H(T(x, h), y)$$

Then $\{\Lambda_{xh}(y) : x \in \mathbb{R}, h \in \mathbb{R}_+, y \in \mathbb{R}\}$

contains all information about the process

X . can be reconstructed from $\Lambda_{\bullet, \bullet}(\bullet)$

Moreover: $\{\Lambda_{xh}(y) : x \in \mathbb{R}, h \in \mathbb{R}_+, \text{ }y \geq x\}$ is suff

Since, for $y \leq x$:

$$\Lambda_{xh}(y) = \sup\{h' : \Lambda_{yh'}(x) < h\}$$

Thus, we only need:

$$\{ \Lambda_{xh}(y) : x \in \mathbb{R}, h \geq 0, y \geq x \}$$

and go backwards

Look at the joint scaling of

$$\left(\frac{\Lambda_{[Nx_i], [N \circ h_i]}([N y_i])}{\delta \sqrt{N}} ; i=1, 2, \dots, p \right)$$

on pages 32-33!

$$\{ \Lambda_{xh}(y) : x \in \mathbb{R}, h \geq 0; y \geq x \}$$

must be the

Brownian Web

(with reflection/absorption at $h=0$)

Historical remarks, priorities ...

R. Arratia 1979 (unpublished manuscript)

BT, WW 1998 (full construction)

Ch Newman et al (200 (re)discover and coin term BW)

Theorem

$\exists!$ a process $\{\Lambda_{xh}(y) : x \in \mathbb{R}, h \geq 0, y \geq x\}$
with the following properties:

① for $(x_1, h_1), \dots, (x_p, h_p)$ fixed
 $(\Lambda_{x_1 h_1}(\cdot), \dots, \Lambda_{x_p h_p}(\cdot))$ are ICRA
(see page 32)

② a.s. $\forall (x, h) : \Lambda_{xh}(x) = h$

③ a.s. $\forall (x_1, h_1), (x_2, h_2), z \geq y \geq \max\{x_1, x_2\}$

$$\{\Lambda_{x_1 h_1}(y) < \Lambda_{x_2 h_2}(y)\} \Rightarrow \{\Lambda_{x_1 h_1}(z) < \Lambda_{x_2 h_2}(z)\}$$

no crossing of lines

④ regularity

a.s. $\forall x \leq y : h \mapsto \Lambda_{xh}(y)$ is left continuous

Remark: other regularity conditions are also possible
(see Arratia, Newman et al, ...)

Construction of the process $\Lambda_{\cdot, \cdot}(\cdot)$:

let $(\tilde{x}_i, \tilde{h}_i)_{i \in \mathbb{N}}$ be dense in $\mathbb{R} \times \mathbb{R}_+$

Step 1: $\left\{ \Lambda_{\tilde{x}_i, \tilde{h}_i}(\cdot) : i=1, 2, 3, \dots \right\}$ easily constructed sequentially

Step 2: the law of $\left\{ \Lambda_{\tilde{x}_i, \tilde{h}_i}(\cdot) : i \in \mathbb{N} \right\}$ will not depend on the order of the sequential construction

Step 3: extend by left continuity in h

$$\Lambda_{x, h}(y) := \sup \left\{ \Lambda_{\tilde{x}_i, \tilde{h}_i}(y) : \tilde{x}_i \leq x, \Lambda_{\tilde{x}_i, \tilde{h}_i}(x) \leq h \right\}$$

Step 4: properties (1)(2)(3)(4) drop out

Step 5: check uniqueness / relies on five estimates on BM.

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First properties of the forward lines:

[to be completed, ...]

see on the blackboard...

The backward lines:

$$y \leq x: \Lambda_{xh}^*(y) := \sup \{ h' : \Lambda_{yh'}^*(x) < h \}$$

reverted:

$$y \geq x: \Lambda_{xh}(y) = \sup \{ h' : \Lambda_{yh'}^*(x) < h \}$$

$$\text{Duality: } \Lambda_{-x,h}^*(-y) \stackrel{\text{law}}{=} \Lambda_{xh}(y)$$

Topology:

$$I(x,h) := \lim_{y \nearrow x} \sup \# \{ p \in \mathbb{N} : \exists (x_1, h_1), \dots, (x_p, h_p) : x_i \leq x \}$$

$$\forall y \in (y, x) : \Lambda_{x_1, h_1}(y) < \dots < \Lambda_{x_p, h_p}(y),$$

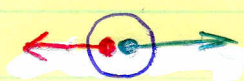
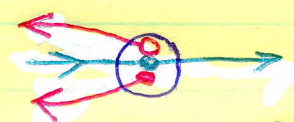
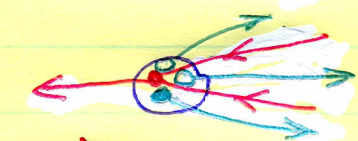

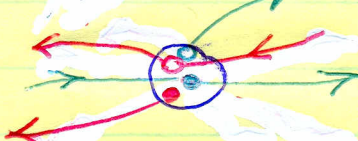
and $\Lambda_{x_i, h_i}(x) = h_i$ - the number of lines coalescing exactly at (x, h_i)

$$O(x,h) := \lim_{y \downarrow x} \lim_{\varepsilon > 0} \# \{ \Lambda_{x', h'}(y) : \text{dist}((x', h'), (x, h)) < \varepsilon \}$$

- number of lines emerging from (x, h) .

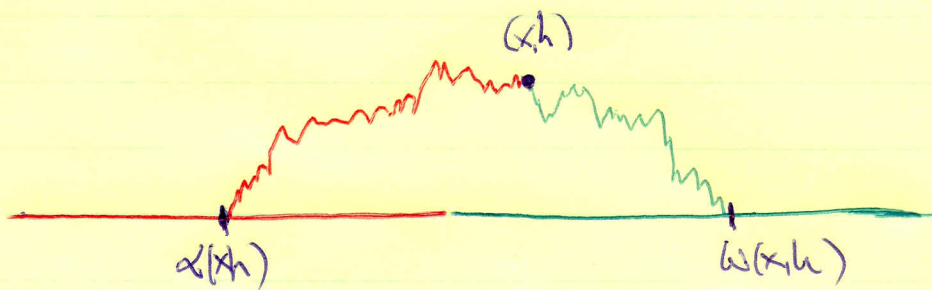
$I^*(x,h), O^*(x,h)$ defined similarly for the backward lines

Topological types of points $(x,h) \in \mathbb{R} \times \mathbb{R}_+$

$[I, 0]$	$[I^*, 0^*]$	picture	<u>Hausdorff dimension</u>
$[0, 1]$	$[0, 1]$		2
$[1, 1]$	$[0, 2]$		3/2
$[0, 2]$	$[1, 1]$		3/2
$[0, 3]$	$[2, 1]$		countably many densely
$[2, 1]$	$[0, 3]$		— " —
$[1, 2]$	$[1, 2]$		1
$0 - I^* = 1 = O^* - I$			

Construction of the process:

$D(x, h) :=$



Almost surely

If $(x_1, h_1) \neq (x_2, h_2)$

either $D(x_1, h_1) \not\subseteq D(x_2, h_2)$ and $\text{int}(D(x_1, h_1) \setminus D(x_2, h_2)) \neq \emptyset$
or the other way around.

$T_{x, h} := |D(x, h)| = \int_{\alpha(x, h)}^{\omega(x, h)} \lambda_{x, h}(y) dy$

$T: \mathbb{R} \times \mathbb{R}_+ \rightarrow [0, \infty)$ almost surely injective

This defines almost surely a linear ordering of $\mathbb{R} \times \mathbb{R}_+$!

And fact

A.s. properties of $(x, h) \mapsto T(x, h)$:

- injective
- for $x \in \mathbb{R}$ fixed:
 $h \mapsto T(x, h)$ strictly increasing
left continuous
- $(x, h) \mapsto T(x, h)$ lower semicontinuous
(\Rightarrow Borel)
- preserves Lebesgue measure
from $\mathbb{R} \times \mathbb{R}_+$ to $(0, \infty)$
- $\text{Ran}(T)$ is of full Lebesgue measure
in $(0, \infty)$

Though: T is not injective

for $I(x, h)$ not of topological type $(0, 1)$

$\lim_{\varepsilon \downarrow 0} T(x, h + \varepsilon) \notin \text{Ran}(T)$!

$$\bigcap_{\varepsilon > 0} \{ (x, h) : T(x, h) \in (t - \varepsilon, t + \varepsilon) \} =: (X(t), H(t))$$

it is a singleton!

First properties of $t \mapsto (X(t), H(t))$

- a.s. $t \mapsto (X(t), H(t))$ is continuous

- a.s. $t \mapsto X(t)$ is recurrent

- $(X(\alpha t), H(\alpha t)) \stackrel{\text{law}}{=} (\alpha^{2/3} X(t), \alpha^{1/3} H(t))$

- a.s. $t \mapsto (X(t), H(t))$ is plane filling - Peano points of type $(0,1)/(0,1)$ are visited once

- $\left\{ \begin{array}{l} (1,1)/(0,2) \\ (0,2)/(1,1) \end{array} \right\}$ - twice

- $\left\{ \begin{array}{l} (0,3)/(2,1) \\ (2,1)/(0,3) \\ (1,2)/(1,2) \end{array} \right\}$ - three times

Local time & Dynamics:

$$\begin{aligned} L(t, y) &:= \sup \{ h > 0 : T(y, h) < t \} \\ &= \inf \{ h > 0 : T(y, h) \geq t \} \end{aligned}$$

since $h \mapsto T(y, h)$ is strictly increasing

$t \mapsto L(t, y)$ is continuous!

reverted:

$$\begin{aligned} T(x, h) &= \sup \{ t \geq 0 : L(t, x) < h \} \\ &= \inf \{ t \geq 0 : L(t, x) \geq h \} \end{aligned}$$

Theorem $L(t, y)$ is (indeed!) the occupation time density of $X(t)$:

$$\text{a.s. } \int_0^t g(s, X(s)) ds = \int_{-\infty}^{\infty} \left\{ \int_0^t g(s, x) d_s L(s, x) \right\} dx$$

relies on the fact that $T: \mathbb{R} \times \mathbb{R}_+ \rightarrow (0, \infty)$ preserves Lebesgue measure.

"Ray-Knight Theorem": $L(T(x,h), y) = \Lambda_{xh}(y)$

for free, by construction; $H(t) = L(t, X(t))$

Dynamics: apply the local time identity

to $g(s, y) = L(s, y \pm \varepsilon)$

$$\frac{1}{2\varepsilon} \int_0^{T_{xh}} (L(s, X(s) + \varepsilon) - L(s, X(s) - \varepsilon)) ds =$$

$$\frac{1}{2\varepsilon} \int_{-\infty}^{+\infty} \left\{ \int_0^{T_{xh}} (L(s, y + \varepsilon) - L(s, y - \varepsilon)) d_s L(s, y) \right\} dy =$$

$$\frac{1}{2\varepsilon} \int_{\alpha(x,h)}^x \left\{ \int_0^{\Lambda_{xh}(y)} (\Lambda_{y,u}(y + \varepsilon) - \Lambda_{y,u}(y - \varepsilon)) du \right\} dy +$$

$$\frac{1}{2\varepsilon} \int_x^{\omega(x,h)} \left\{ \int_0^{\Lambda_{xh}(y)} (\Lambda_{y,u}(y + \varepsilon) - \Lambda_{y,u}(y - \varepsilon)) du \right\} dy$$

An identity:

$$(x, h) \in \mathbb{R} \times \mathbb{R}_+, \quad x \leq y \leq z$$

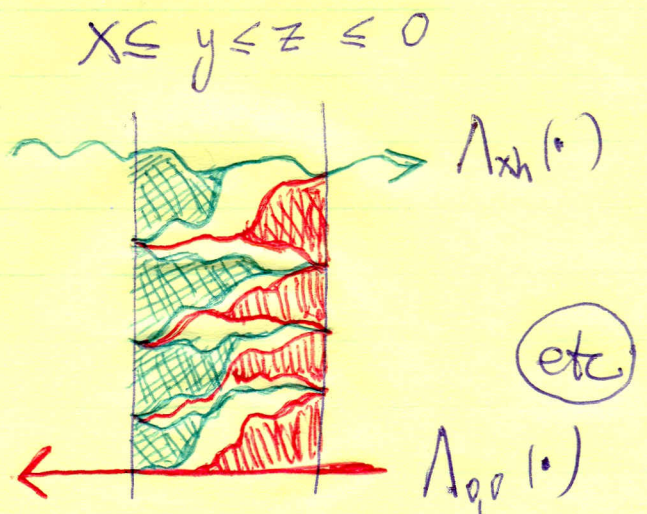
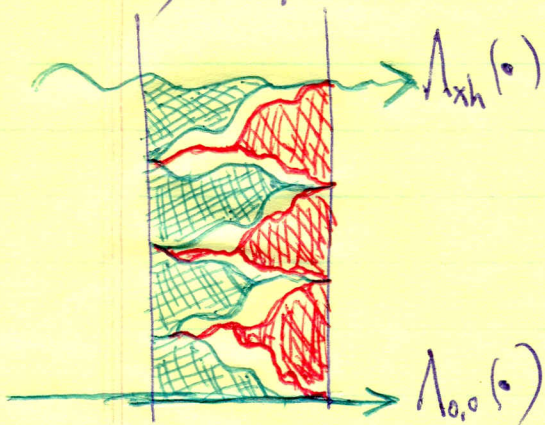
$$(*) \int_0^{\Lambda_{xh}(y)} \Lambda_{yu}(z) du + \int_0^{\Lambda_{xh}(z)} \Lambda_{zu}(y) du = \Lambda_{xh}(y) \cdot \Lambda_{xh}(z)$$

or, equivalently:

$$(**) \int_0^{\Lambda_{xh}(y)} \Lambda_{yu}(z) du - \int_0^{\Lambda_{xh}(z)} \Lambda_{xu}(y) du =$$

$$\frac{1}{2} (\Lambda_{xh}(z) - \Lambda_{xh}(y))^2 - \frac{1}{2} (\Lambda_{xh}^2(z) - \Lambda_{xh}^2(y)) + 2 \int_0^{\Lambda_{xh}(y)} (\Lambda_{yu}(z) - u) du$$

Proof of (*):
 $x, 0 \leq y \leq z$



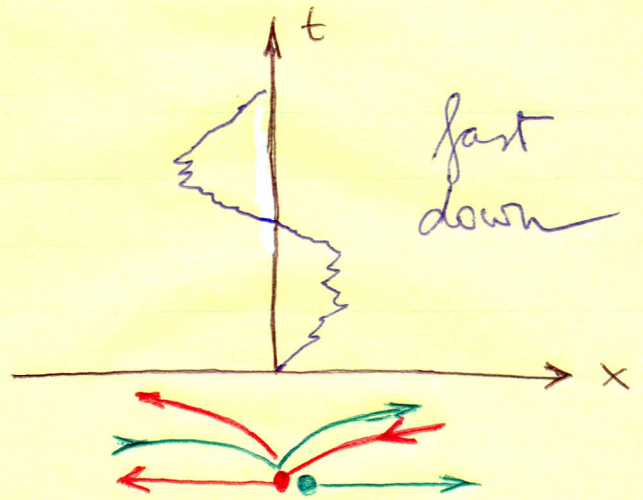
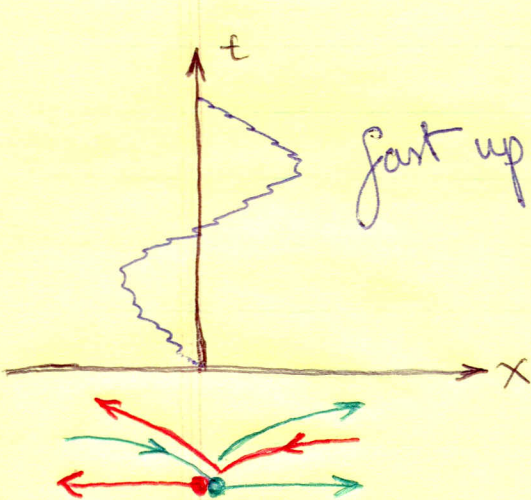
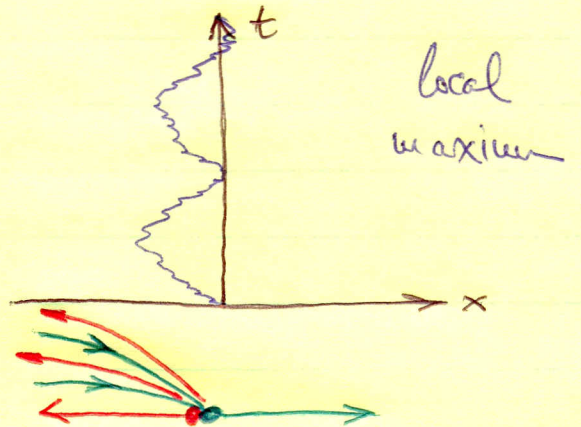
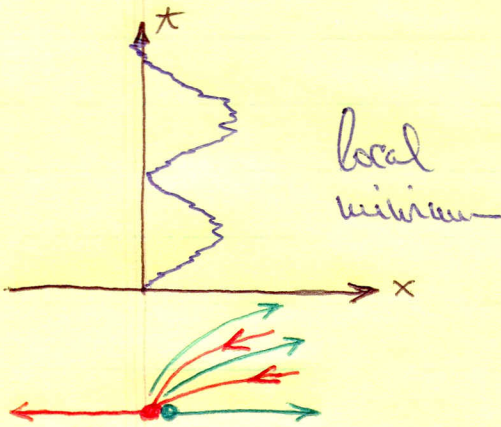
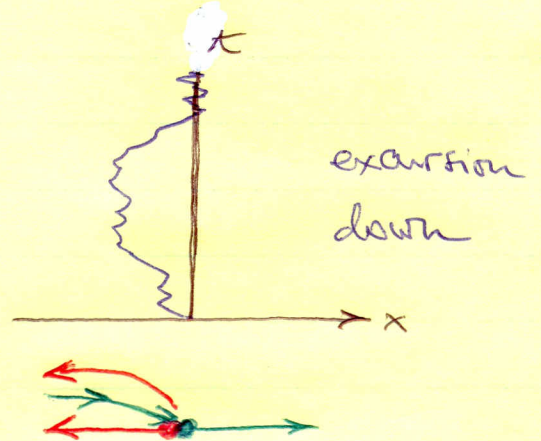
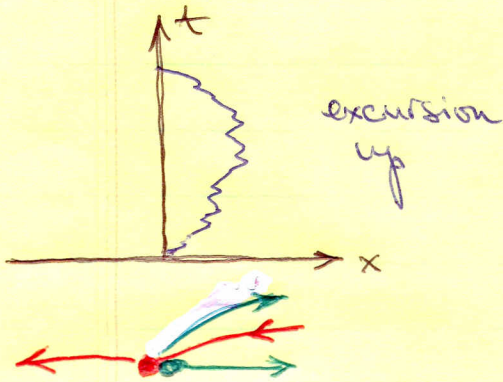
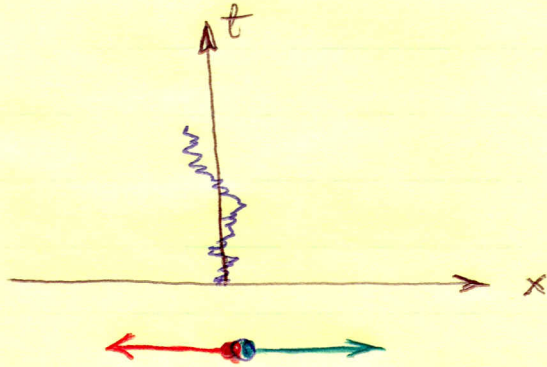
Contributions of the three terms on r.h.s. of ~~(*)~~
(applied with $z = x + \epsilon$) ...

$$\left\{ \frac{1}{4} (w(x, h) - x) + 0 - \frac{1}{2} x \cdot \mathbb{I}(x < 0) \right\} +$$

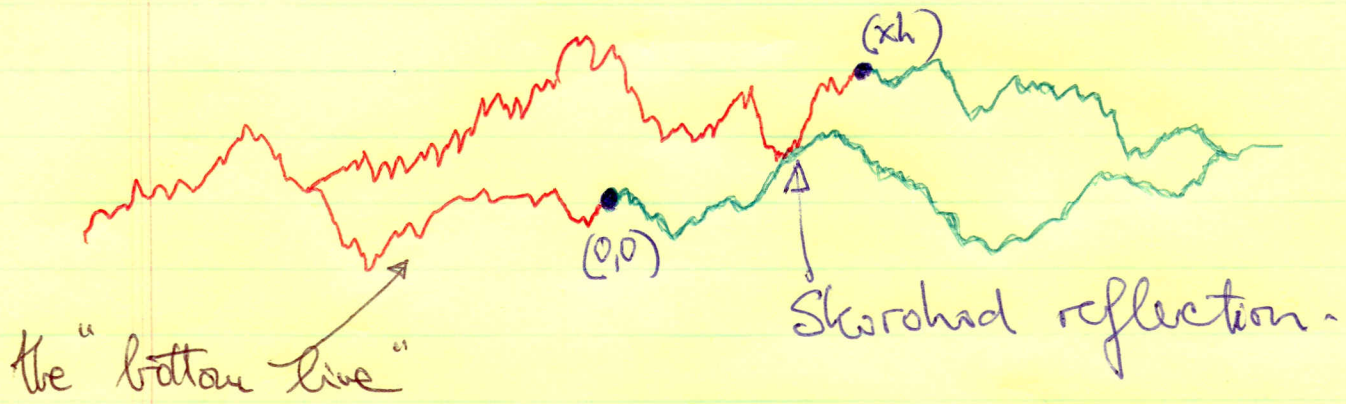
$$\left\{ -\frac{1}{4} (x - d(x, h)) + 0 - \frac{1}{2} x \cdot \mathbb{I}(x > 0) \right\} =$$

$$\frac{1}{4} (d(x, h) + w(x, h)) - x \quad \text{q.e.d.}$$

Topological types revisited:



The stationary process: start with $\Lambda_{0,0}(\gamma)$ - Brownian (rather than straight line)



$$t \mapsto \Lambda_{X(t), H(t)}(X(t) + \cdot) - H(t)$$

or

Stationary Markov process

$$t \mapsto \eta(t, \cdot) := \Lambda_{X(t), H(t)}(X(t) + \cdot)$$

$$\in \mathcal{S}'(\mathbb{R})$$

Stationary Markov process

Stationary distribution: white noise

Major open problem:

understand better $\gamma(t, \cdot)$

e.g. What is its infinitesimal generator acting on

$$L^2(\mathcal{J}, \pi) ?$$

white noise

The construction of the process
 $(X(t), H(t))$

$(1-x)P - (x)P$

