

# Some properties of the true self-repelling motion

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# What is a **(real-valued)** self-repelling process?

If  $(X_t)_{t \geq 0}$  is a one-dimensional continuous process, define its occupation time measure:

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→  $L_t : \mathbb{R} \rightarrow \mathbb{R}_+$  is called **“local time”**.

# What is a **(real-valued)** self-repelling process?

Heuristical (vague) definition:  $(X_t)_t$  is called **self-repulsive** when:

- ▶  $(X_t, L_t(\cdot))_t$  is a Markov process.
- ▶  $(X_t)_t$  prefers to go to the less visited places  
→ pushed away from  $x \in \mathbb{R}$  where  $L_t(x)$  is large.

## Process with its local time

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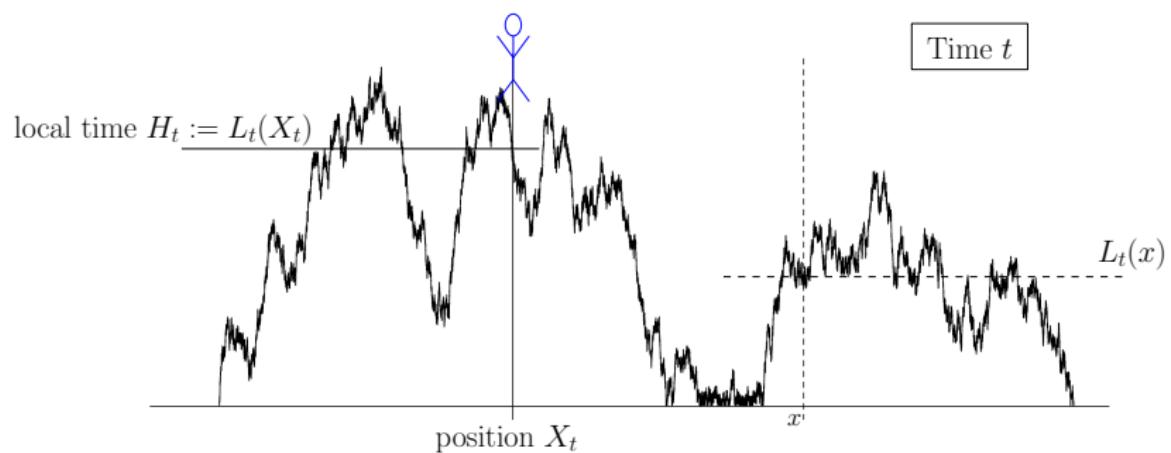


Figure : Picture at time  $t$ :  $X_t$  and its local time  $L_t(\cdot)$

## A first example: Brownian polymers

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$$dX_t = -\nabla(\textcolor{blue}{b} * L_t(X_t + \cdot))|_0 dt + dB_t$$

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What happens when  $\varepsilon \rightarrow 0$ ? Should converge to the true self-repelling motion (introduced after).

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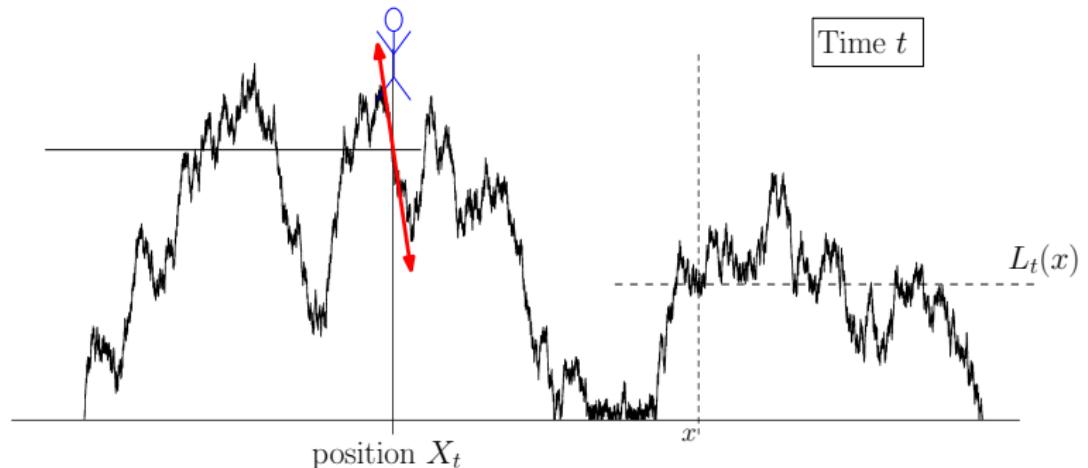


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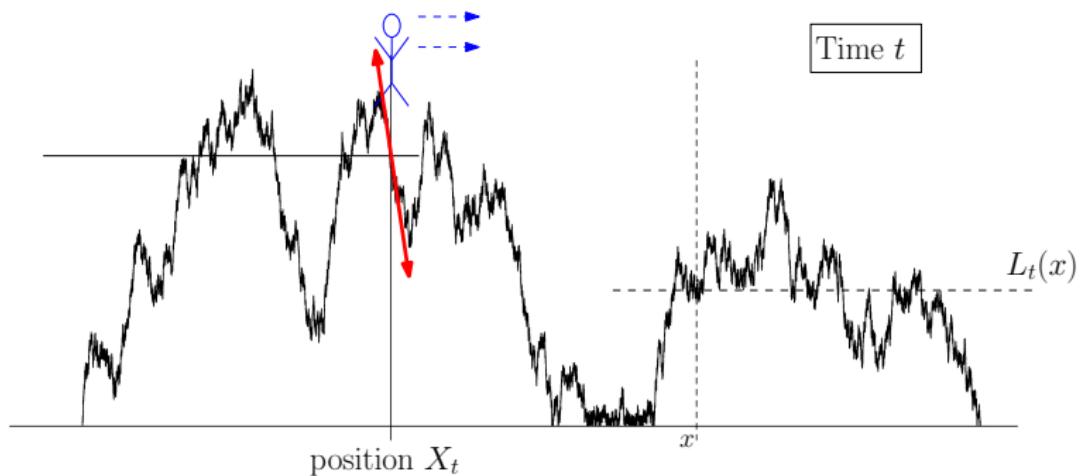


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## TSRM definition and first properties

Results of the papers of B. Toth - W. Werner 1998 and F. Soucaliuc - B. Toth - W. Werner 2000

# Short introduction to the true self-repelling motion (TSRM)

B. Tóth and W. Werner constructed a one-dimensional continuous self-repelling process  $(X_t, t \geq 0)$  called the **true self-repelling motion (TSRM)**.

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B. Tóth and W. Werner constructed a one-dimensional continuous self-repelling process ( $X_t, t \geq 0$ ) called the **true self-repelling motion (TSRM)**.

For the construction, they used a family of coalescing reflected Brownian motions in the upper half plane now called **Brownian Web**. The TSRM is defined as the trace of the contour of the tree of these coalescing Brownian motions.

## TSRM construction

Take the **Brownian Web**  $(\Lambda_{x,h}, (x, h) \in \mathbb{R} \times \mathbb{R}_+)$ .

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**Parametrization:** by the area it has swept. For every  $(x, h)$  in the upper half plane, the process  $(X_t, H_t)$  visits the point  $(x, h)$  at the random time  $t = T_{x,h} := \int_{\mathbb{R}} \Lambda_{x,h}(y) dy$

TSRM = first coordinate  $(X_t)$ .

## Some first properties

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First properties (established by Bálint Tóth and Wendelin Werner):

- ▶ Continuity and recurrence.
- ▶ Scaling and local variation: For all  $a > 0$ ,  $(X_{at}, t \geq 0)$  and  $(a^{2/3}X_t, t \geq 0)$  have the same distribution and the TSRM is of finite variation of order  $3/2$ .

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- ▶ Local time: The TSRM admits a local time  $L_t(\cdot)$  and a.s., for every  $(x, h) \in \mathbb{R} \times \mathbb{R}^+$ , the Brownian Web curves corresponds to the local time at times  $T_{x,h}$  (strong Ray Knight theorem). It implies  $H_t = L_t(X_t)$ .

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- ▶ **Markov property:**  $(X_t, L_t(\cdot))_t$  is a Markov process.
- ▶ **Localization:** Interaction is local: the law of  $X$  just after  $t$  depends only on  $L_t$  around the point  $X_t$ .  
Moreover, we have a **dynamical equation**:

$$dX_t = \lim_{\varepsilon \rightarrow 0} \frac{1}{2\varepsilon} (L_t(X_t + \varepsilon) - L_t(X_t - \varepsilon)) dt.$$

Limit holds in the probability sense.

# My contributions

# Large deviations of the TSRM

TSRM is an unusual process and it gives motivation to study some of its finest properties to discover the features it shares/does not share with the other processes.

## Proposition (L.D.)

- ▶ When  $x \rightarrow \infty$ ,  $P(X_1 > x) = \exp(-\kappa x^3 + O(\ln(x)))$  for some explicit  $\kappa$  (in terms of zeros of Airy function).
- ▶ When  $h \rightarrow \infty$ ,  $P(H_1 > h) = \exp(-8h^3/9 + O(\ln(h)))$ .

# Law of the iterated logarithm

Pushing forward those results permits to derive a LIL for the TSRM when both  $t$  is large and  $t$  is small:

## Proposition (L.D.)

$$\text{a.s., } \limsup_{t \rightarrow 0} t^{-2/3} (\ln(\ln(1/t)))^{-1/3} X_t = 1/\kappa^{1/3}.$$

## Marginal distributions

In a joint work with Bálint Tóth, we computed the marginal distributions of this process.

### Proposition (L.D., B. Tóth)

- ▶ The density of  $X_1$  denoted by  $\nu_1(x)$  is equal to:

$$\nu_1(x) = \sum_{k=1}^{\infty} \frac{3^{2/3}}{2^{7/3}} \left( \frac{\Gamma(2/3)}{\Gamma(1/3)} \right)^2 |a'_k|^{-3} f_{2/3}(2^{1/3} |a'_k| |x|)$$

where the scaling factors  $a'_k$  are the zeros of the derivative of the Airy function and  $f_{2/3}$  is the Mittag-Leffler's function.

- ▶ The density of  $H_1$  denoted by  $\nu_2(h)$  is equal to:

$$\nu_2(h) = \frac{2 \cdot 6^{1/3} \sqrt{\pi}}{\Gamma(1/3)^2} \exp(-(8h^3)/9) U(1/6, 2/3; (8h^3)/9)$$

where  $U$  is the hypergeometric function.

## Marginal distributions

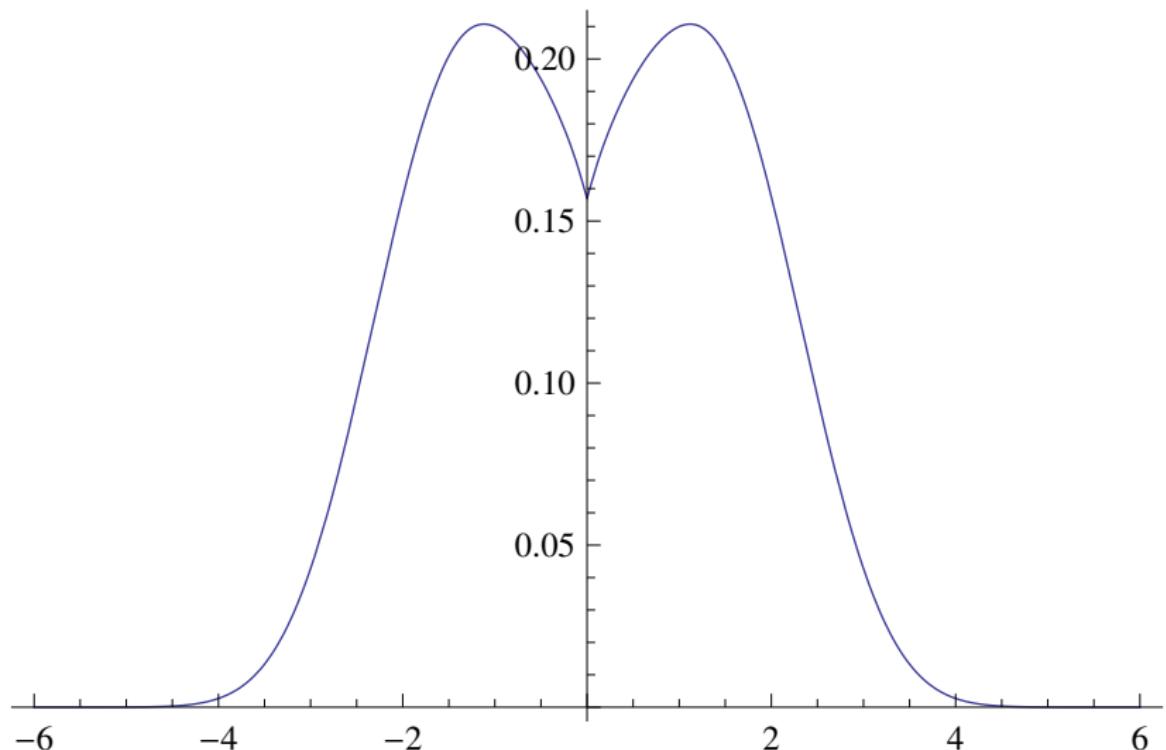


Figure : Density of  $X_1$  (displacement at time 1)

# A clever (self-repelling) burglar

What is the conditional law of the position  $X_1$  knowing  $L_1(\cdot)$ ?

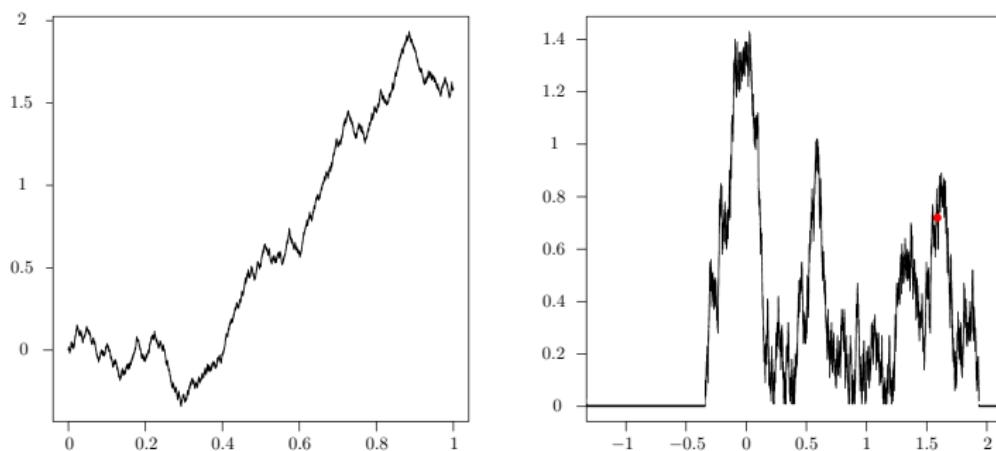


Figure : On the left ( $X_t$ ,  $t \in [0, 1]$ ), and on the right the local time  $L_1(\cdot)$

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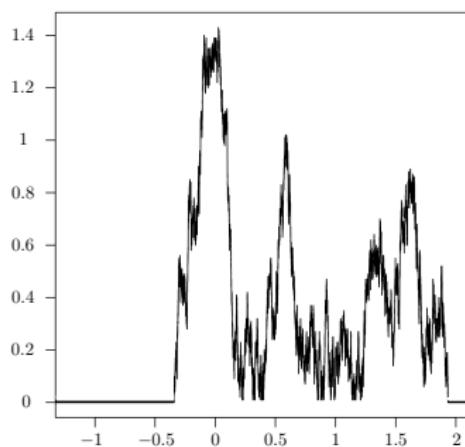
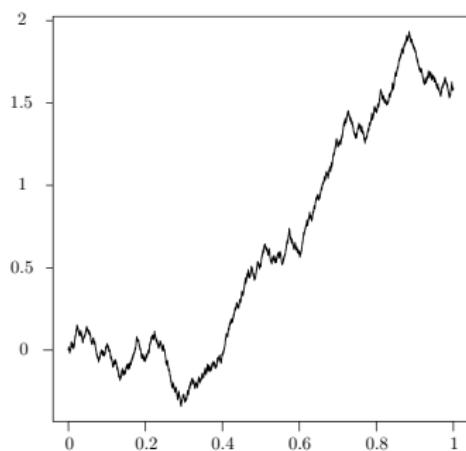
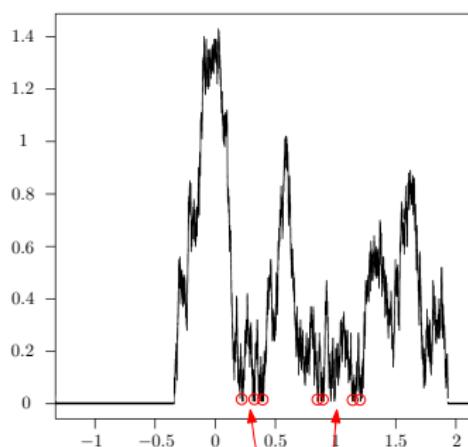
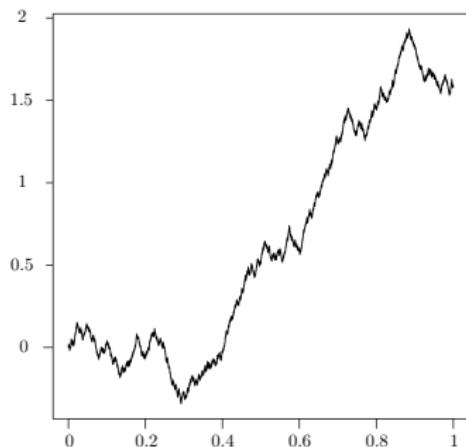


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Fast points

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# Result

## Proposition (L.D.)

*The conditional law of  $X_1$  knowing its local time at time 1,  $L_1(\cdot)$ , is uniform on the interval  $I$  defined by:*

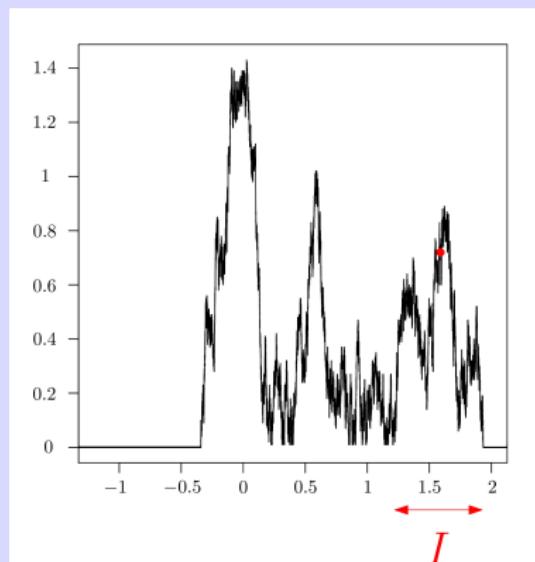


Figure : Definition of  $I$

# Thank you!