

**Maximal displacement of
branching symmetric stable processes**

available at [arXiv:2106.15215](https://arxiv.org/abs/2106.15215)

Yuichi Shiozawa

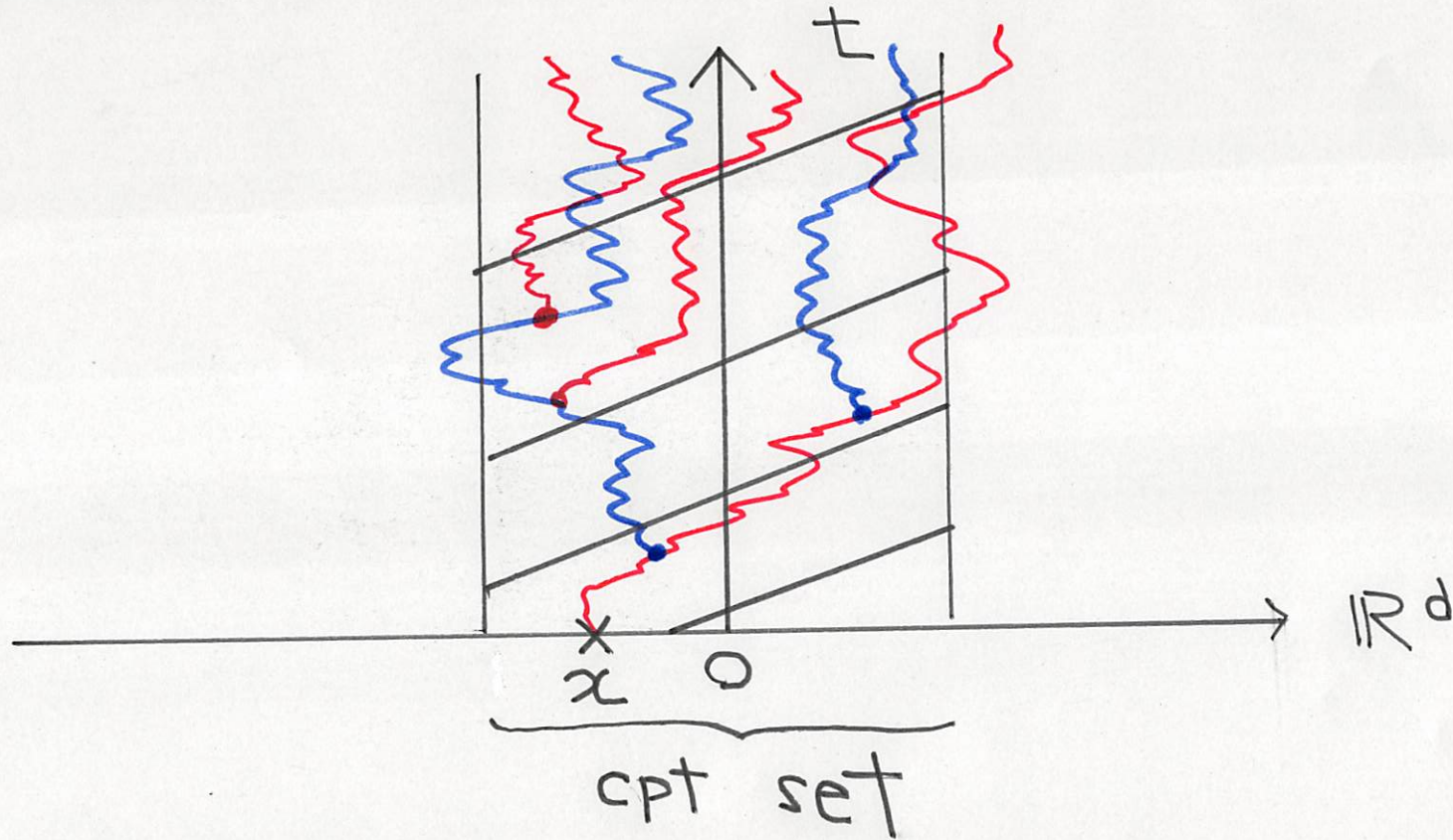
(Osaka University, Japan)

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1. Introduction



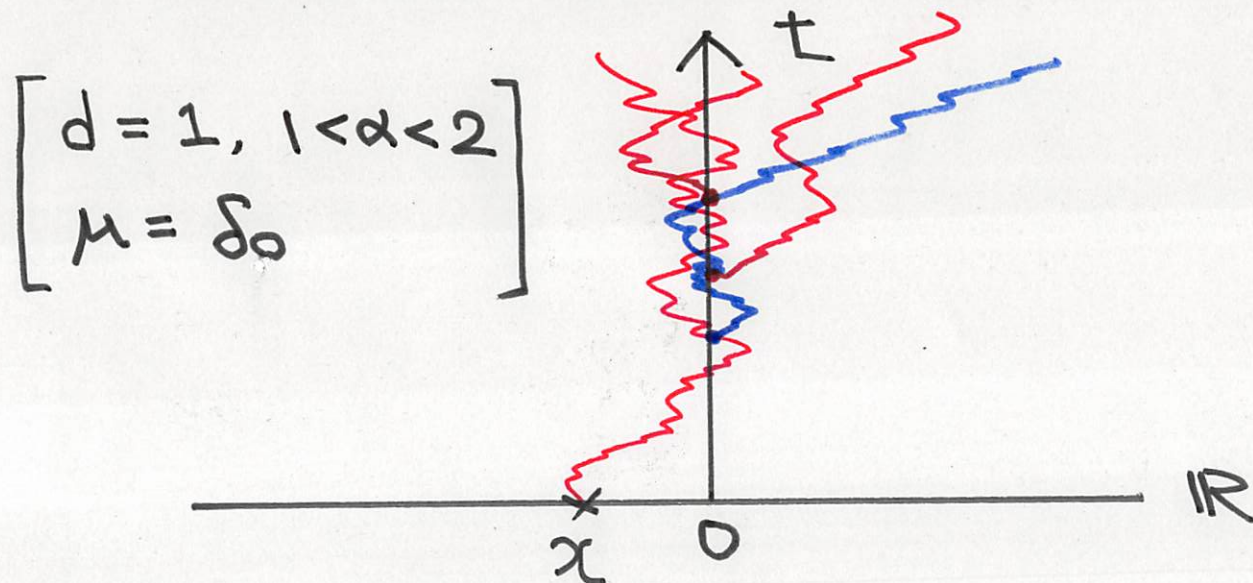
- Reproduction only on a **compact set** in \mathbb{R}^d
- ▷ L_t : **forefront** of the particle range at time t

▷ $(\{X_t\}_{t \geq 0}, \{P_x\}_{x \in \mathbb{R}^d})$: **symm. α -stable proc. on \mathbb{R}^d**
generated by $-\frac{1}{2}(-\Delta)^{\alpha/2}$ ($\alpha \in (0, 2)$)

$$\Rightarrow P_x(|X_t| > R) \sim \frac{c_0 t}{R^\alpha} \quad (R \rightarrow \infty) \quad (\text{heavy tail})$$

- **Branching RW on \mathbb{Z} with spatially homogeneous branching**
[Durrett(83), Bhattacharya-Hazra-Roy(17)]
- **(conti. time) Catalytic branching RW on \mathbb{Z}** [Bulinskaya(21)]
(reproduction only on **finite points**)

2. Model and results



- (1) μ -killed symm. stable proc.
- (2) binary branching at the lifetime
- (3) indep. reproduction

▷ μ : positive Radon meas. on \mathbb{R}^d with **compact support**

▷ $G_\beta(x, y)$: β -resolvent of the symm. α -stable proc. on \mathbb{R}^d

$$\lim_{\beta \rightarrow \infty} \sup_{x \in \mathbb{R}^d} \int_{\mathbb{R}^d} G_\beta(x, y) \mu(dy) = 0 \quad (\text{Kato class})$$

▷ $\lambda := \inf \text{Spec} \left(\frac{1}{2}(-\Delta)^{\alpha/2} - \mu \right)$: intensity of branching

In what follows, we assume $\lambda < 0$

⇒ the ground state $h \in C_b^+(\mathbb{R}^d)$ exists and

$$h(x) \sim \frac{C_0}{|x|^{d+\alpha}} \int_{\mathbb{R}^d} h(y) \mu(dy) \quad (|x| \rightarrow \infty)$$

▷ $Z_t :=$ population at time t

▷ X_t^k : position of the k th particle at time t ($1 \leq k \leq Z_t$)

▷ $L_t := \max_{1 \leq k \leq Z_t} |X_t^k|$:

maximal norm of particles alive at time t (forefront)

▷ $M_t := e^{\lambda t} \sum_{k=1}^{Z_t} h(X_t^k)$: nonneg. square integrable martingale

Theorem. $\exists c_* > 0$ (explicit), $\forall \kappa > 0$,

$$\lim_{t \rightarrow \infty} \mathbb{P}_x \left(e^{\lambda t / \alpha} L_t \leq \kappa \right) = \mathbb{E}_x \left[\exp \left(-\kappa^{-\alpha} c_* M_\infty \right) \right]$$

RHS: average over the **Fréchet distribution** with parameter α
scaled by $c_* M_\infty$ [Bovier(17), Thm 1.12]

Remark. [(Non)degeneracy of M_∞]

- $d = 1, \alpha \in (1, 2) \Rightarrow \lambda < 0$ and $\mathbb{P}_x(M_\infty > 0) = 1$
- $d > \alpha \Rightarrow \mathbb{P}_x(M_\infty = 0) \in (0, 1)$

3. Comment on the proof of Theorem

$$\triangleright R^\kappa(t) = \kappa e^{-\lambda t/\alpha} \quad (\kappa > 0: \text{fixed})$$

By the Markov and branching properties at time $T(\leq t)$,

$$\mathbb{P}_x(L_t \leq R^\kappa(t)) = \mathbb{E}_x \left[\prod_{k=1}^{Z_T} \mathbb{P}_{X_T^k}(L_{t-T} \leq R^\kappa(t)) \right] \cdots \text{(A)}$$

By the **second moment method** [Nishimori-S(21+)],

$$\mathbb{P}_{X_T^k}(L_{t-T} \leq R^\kappa(t)) \asymp \exp \left(-\frac{c_*}{\kappa^\alpha} e^{\lambda T} h(X_T^k) \right) \cdots \text{(B)}$$

By (A) and (B), we have as $t \rightarrow \infty$ and $T \rightarrow \infty$,

$$\mathbb{P}_x(L_t \leq R^\kappa(t)) \asymp \mathbb{E}_x \left[\exp \left(-\frac{c_*}{\kappa^\alpha} M_\infty \right) \right] \cdots \text{(C)}$$

4. Tail probability and examples

▷ $a(t)$: positive m'ble funct. s.t. $a(t) \rightarrow \infty$ ($t \rightarrow \infty$)

Theorem B. $\exists c_* > 0$ (as in Thm), loc. uniformly in $x \in \mathbb{R}^d$,

$$\mathbb{P}_x \left(e^{\lambda t/\alpha} L_t > a(t) \right) \sim \frac{c_*}{a(t)^\alpha} h(x) \quad (t \rightarrow \infty)$$

Example. [Catalytic branching] ▷ $d = 1$, $\alpha \in (1, 2)$

▷ $\mu = c\delta_0$ ($c > 0$, δ_0 : Dirac meas. at the origin)

⇒ reproduction only at the origin

Theorems hold, λ and $h(x)$ can be written explicitly [S(08)].