

Genealogy and spatial distribution
of the N -particle branching random walk
with polynomial tails

Sarah Penington

University of Bath

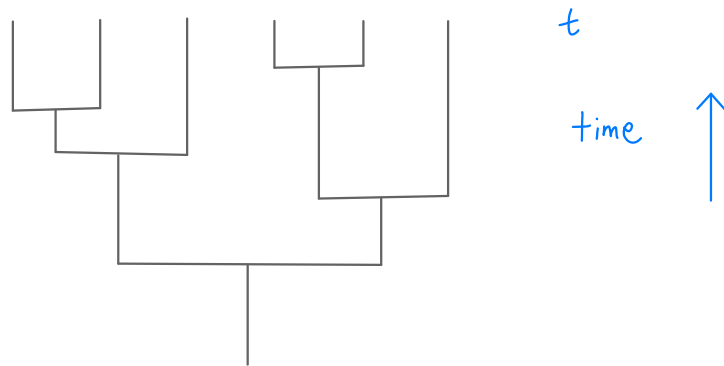
Joint work with Matt Roberts and Zsófia Talyigás

Branching-selection systems

- Particle systems: particles branch (produce offspring) and move in space
killing rule keeps total number of particles constant.
- Toy models for a population under selection.
Location of a particle (= individual) represents its evolutionary fitness.
- Introduced by Brunet and Derrida in 1990s.
Recent results and open conjectures about long-term behaviour.

Genealogy:

Coalescent process



N-particle branching random walk (N-BRW)

Discrete-time branching-selection system.

N particles with locations in \mathbb{R} at each timestep.

Let X be a real-valued random variable (jump distribution).

At each time $n \in \mathbb{N}_0$, each particle has two offspring.

Each of the $2N$ offspring particles makes an independent jump from its parent's location, with the same law as X .

The N rightmost particles (of the $2N$ offspring particles) form the population at time $n+1$.



Notation: $X_1^{(N)}(n) \leq X_2^{(N)}(n) \leq \dots \leq X_N^{(N)}(n)$ ordered particle positions at time n .

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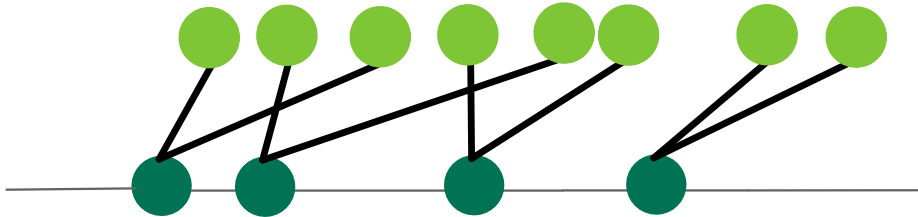
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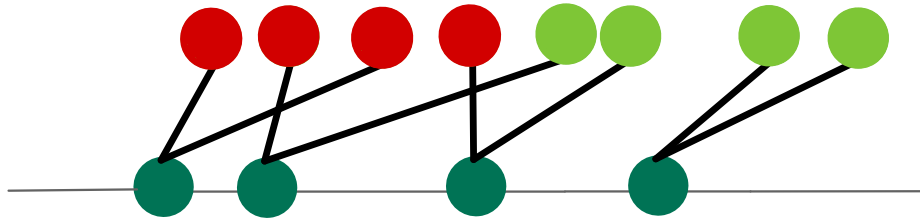
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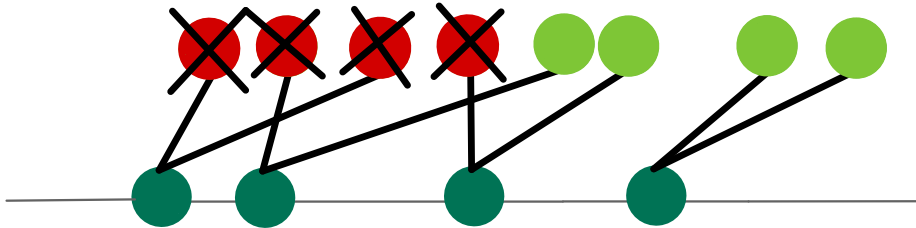
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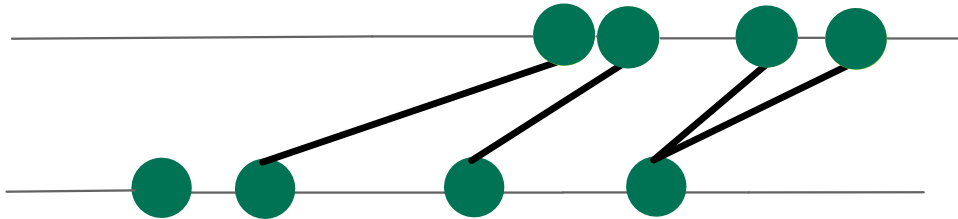
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Light-tailed jump distribution

Asymptotic speed

If $\mathbb{E}[X] < \infty$ then $\exists v_N \in (0, \infty)$ s.t. $\lim_{n \rightarrow \infty} \frac{X_N^{(N)}(n)}{n} = v_N = \lim_{n \rightarrow \infty} \frac{X_1^{(N)}(n)}{n}$ a.s. and in L^1 .

Theorem (Bérard and Guéré 2010) If $\mathbb{E}[e^{\lambda X}] < \infty$ for some $\lambda > 0$ (+technical assumptions) then $\lim_{N \rightarrow \infty} v_N = v_\infty$ exists and $v_\infty - v_N \sim c(\log N)^{-2}$ as $N \rightarrow \infty$.

Conjectured by Brunet + Derrida 1997. Related result for Fisher-KPP equation with noise (Mueller, Mytnik, Quastel 2009)

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Genealogy

Sample k particles from the N particles and trace their ancestry backwards in time \rightarrow coalescent process.

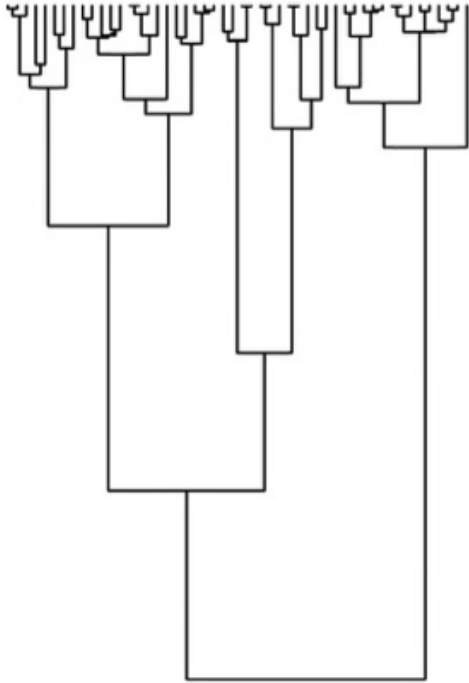
Conjecture (Brunet, Derrida, Mueller, Munier)

If X is light-tailed then the genealogy of a sample on a $(\log N)^3$ timescale converges to a Bolthausen-Sznitman coalescent as $N \rightarrow \infty$.

Coalescent processes

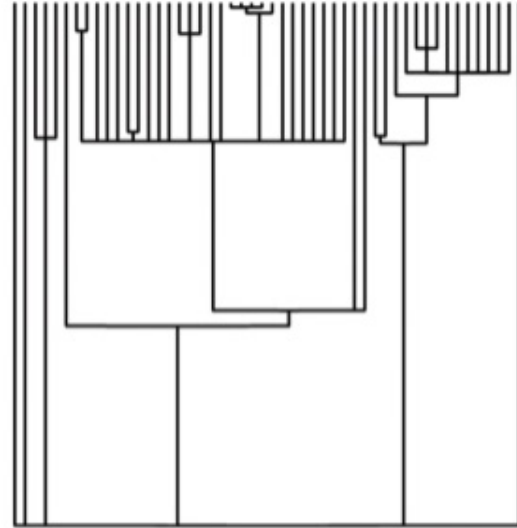
Kingman's coalescent

Neutral population: choose particles to kill uniformly at random in each generation.



Bolthausen-Sznitman coalescent

Population under selection.



Thanks to Götz Kersting
and Anton Wakolbinger

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If X is light-tailed then the genealogy of a sample on a $(\log N)^3$ timescale converges to a Bolthausen-Sznitman coalescent as $N \rightarrow \infty$.

N-BRW with heavy-tailed jump distribution

Suppose $P(X > x) \sim x^{-\alpha}$ as $x \rightarrow \infty$, for some $\alpha > 0$.

Asymptotic speed

Theorem (Bérard and Maillard 2014)

If $E[X] < \infty$, $\lim_{n \rightarrow \infty} \frac{X_N^{(N)}(n)}{n} = v_N$ where $v_N \sim c_\alpha N^{1/\alpha} (\log N)^{1/\alpha - 1}$ as $N \rightarrow \infty$.

If $E[X] = \infty$, cloud of particles accelerates.

Genealogy

Conjecture (Bérard and Maillard)

The genealogy on a $\log N$ timescale is approximately given by a star-shaped coalescent when N is large.

Time and space scales

$$\text{Let } \mathbb{P}(X > x) = \frac{1}{h(x)} \text{ for } x \geq 0.$$

Assume h is regularly varying with index $\alpha > 0$

$$\text{i.e. for any } y > 0, \quad \frac{h(xy)}{h(x)} \longrightarrow y^\alpha \text{ as } x \rightarrow \infty.$$

and $\mathbb{P}(X \geq 0) = 1$ (no negative jumps).

$$\text{Let } \ell_N = \lceil \log_2 N \rceil \quad \text{time scale}$$

$$\text{Let } a_N = h^{-1}(2N\ell_N), \quad \text{where } h^{-1}(x) := \inf \{y \geq 0 : h(y) > x\}. \quad \text{space scale}$$

$$\begin{aligned} \mathbb{E} \left[\begin{array}{l} \# \text{ jumps of size } > c_1 a_N \text{ in} \\ \text{a time interval of length } c_2 \ell_N \end{array} \right] &= 2N \cdot c_2 \ell_N \mathbb{P}(X > c_1 a_N) \\ &= \frac{2N c_2 \ell_N}{h(c_1 a_N)} \sim \frac{2N c_2 \ell_N}{c_1^\alpha 2N \ell_N} = \frac{c_2}{c_1^\alpha} \\ &\quad \text{as } N \rightarrow \infty. \end{aligned}$$

Main result

w.h.p. = with probability $\rightarrow 1$ as $N \rightarrow \infty$.

Theorem (P., Roberts, Talyigás 2021)

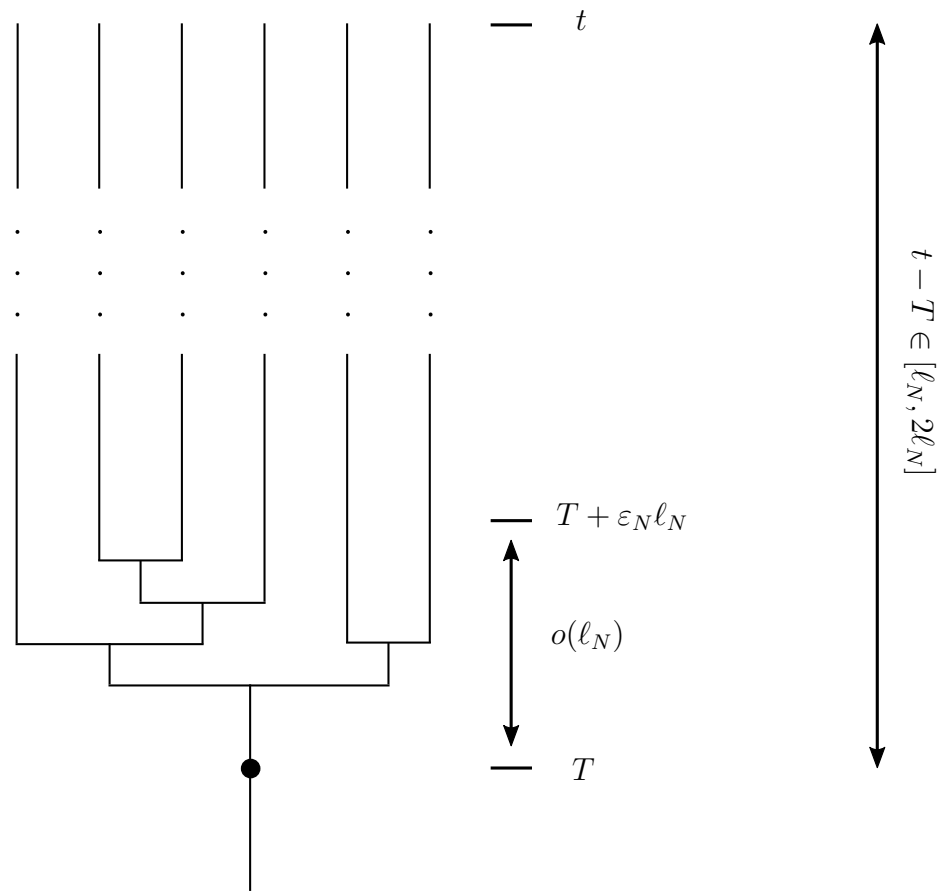
For $\eta > 0$, $k \in \mathbb{N}$ and $t > 4\ell_N$, the following occurs w.h.p.:

- **Spatial distribution:** At time t , there are $N - o(N)$ particles in

$$[X_i^{(N)}(t), X_i^{(N)}(t) + \eta a_N].$$

- **Genealogy:** The genealogy on an ℓ_N -timescale is asymptotically given by a star-shaped coalescent.

i.e. $\exists T \in [t - 2\ell_N, t - \ell_N]$ s.t. w.h.p., for a uniform sample of k particles at time t , every particle is descended from the rightmost particle at time T and no pair of particles in the sample has a common ancestor after time $T + \varepsilon_N \ell_N$, for any $(\varepsilon_N)_N$ with $\varepsilon_N \rightarrow 0$ and $\varepsilon_N \ell_N \rightarrow \infty$ as $N \rightarrow \infty$.



$\exists T \in [t - 2l_N, t - l_N]$ s.t. w.h.p., for a uniform sample of k particles at time t , every particle is descended from the rightmost particle at time T and no pair of particles in the sample has a common ancestor after time $T + \varepsilon_N l_N$, for any $(\varepsilon_N)_N$ with $\varepsilon_N \rightarrow 0$ and $\varepsilon_N l_N \rightarrow \infty$ as $N \rightarrow \infty$.

Spatial distribution

At time t , there are $N - o(N)$ particles in $[X_i^{(N)}(t), X_i^{(N)}(t) + \eta a_N]$ w.h.p.

Proposition (PRT 2021) There exist $0 < p_r \leq q_r < 1$ s.t. $q_r \rightarrow 0$ as $r \rightarrow \infty$ and $p_r \rightarrow 1$ as $r \rightarrow 0$ s.t. for $r > 0$, for N sufficiently large and $t > 3\ell_N$,

$$\mathbb{P}(X_N^{(N)}(t) - X_1^{(N)}(t) \geq r a_N) \in [p_r, q_r].$$

Genealogy

w.h.p. $\exists T \in [t - 2\ell_N, t - \ell_N]$ s.t. w.h.p., for a uniform sample of k particles at time t , every particle is descended from the rightmost particle at time T and no pair of particles has a common ancestor after time $T + o(\ell_N)$.

Proposition (PRT 2021) For $0 \leq s_1 < s_2 \leq 1$, $\exists p > 0$ s.t. for N sufficiently large and $t > 4\ell_N$,

$$\mathbb{P}(T \in [t - 2\ell_N + s_1 \ell_N, t - 2\ell_N + s_2 \ell_N]) > p.$$

N-BRW genealogy

Jump distribution X .

Light-tailed $\mathbb{P}(X > x) \leq e^{-cx}, c > 0$

Heavy-tailed $\mathbb{P}(X > x) \sim x^{-\alpha}, \alpha > 0$

Time to coalesce

$(\log N)^3$

Coalescent

Bolthausen-Sznitman

$\log N$

Star-shaped

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Work in progress with Z. Talyigás.

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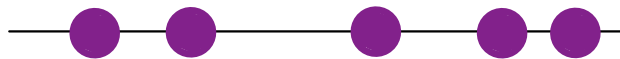
Simulation by Z. Talyigás.

N-particle branching Brownian motion (N-BBM)

- N particles move in \mathbb{R} according to independent Brownian motions.
- Each particle, independently, branches into two particles after an $\text{Exp}(1)$ time.
- Each time a particle branches, the leftmost particle in the system is killed.

N particles in the system at all times.

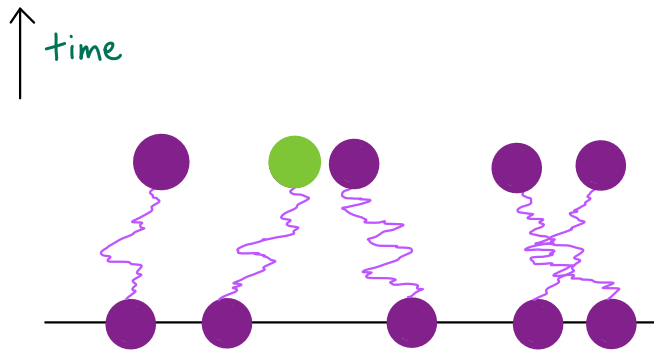
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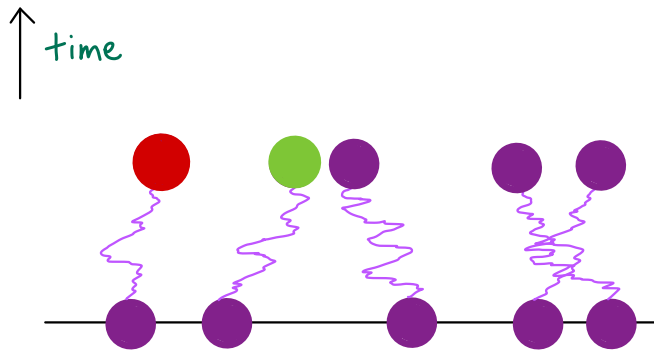
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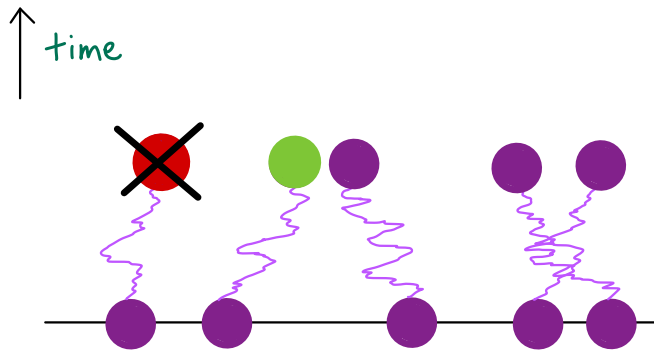
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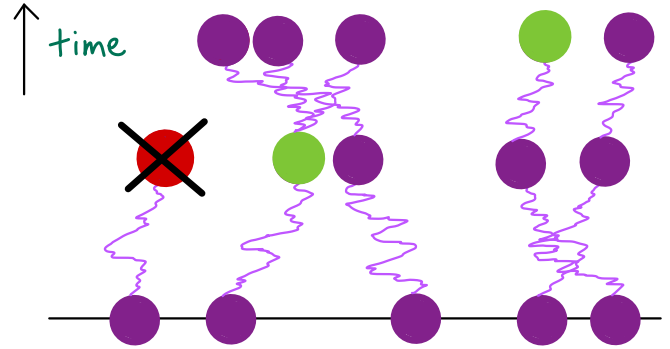
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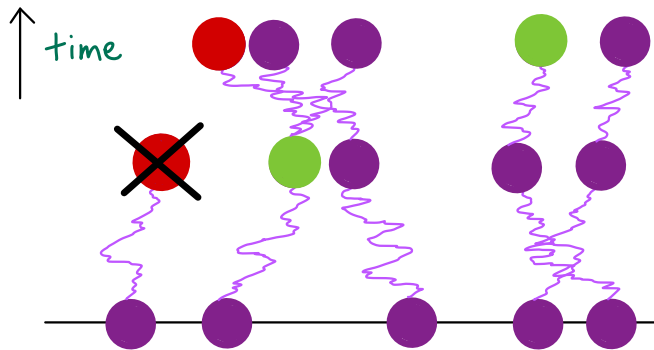
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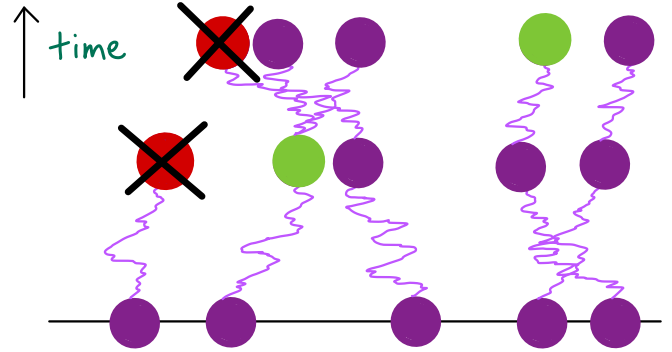
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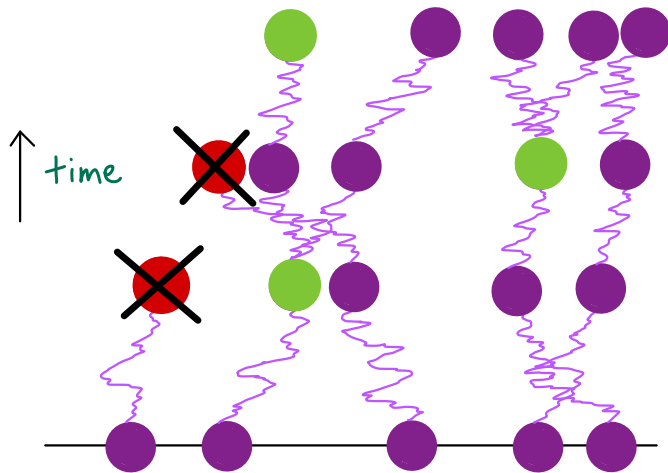
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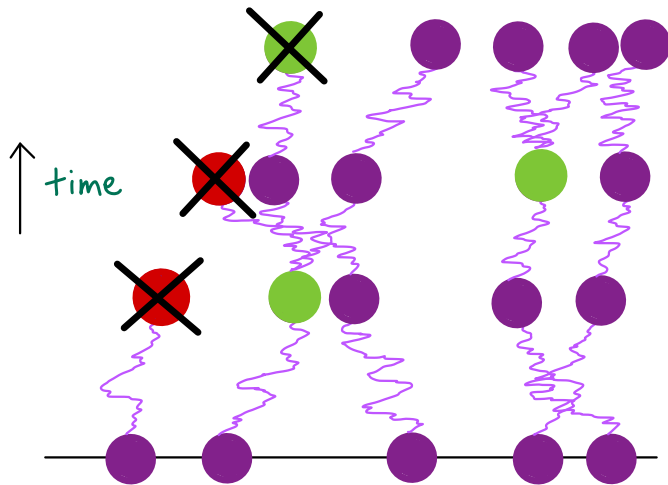
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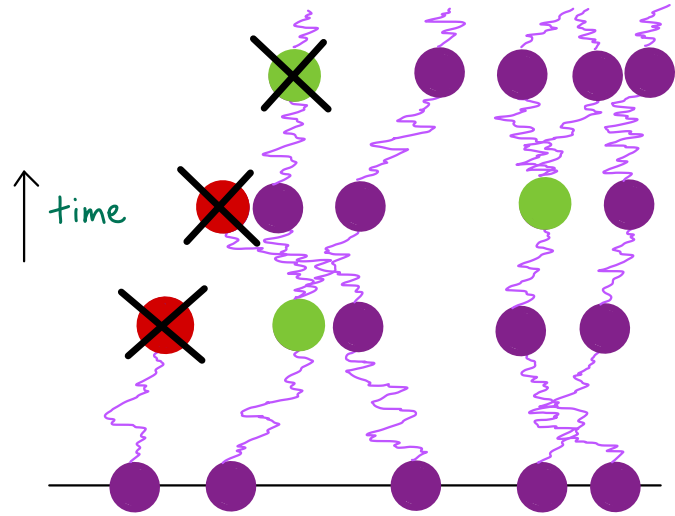
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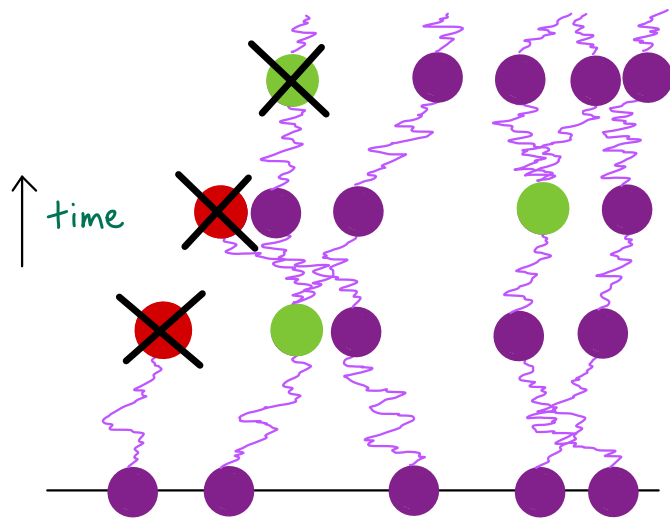
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Introduced by Maillard (2012).

Conjecture (Brunet/Derrida, Maillard): Genealogy of a sample on a $(\log N)^3$ timescale converges to a Bolthausen-Sznitman coalescent as $N \rightarrow \infty$.

One tool: over a fixed timescale, as $N \rightarrow \infty$, density converges to solution of a free boundary problem. (Hydrodynamic limit: De Masi/Ferrari/Presutti/Soprano-Loto '17)

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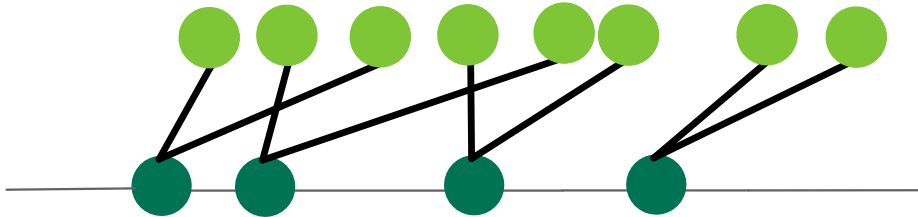
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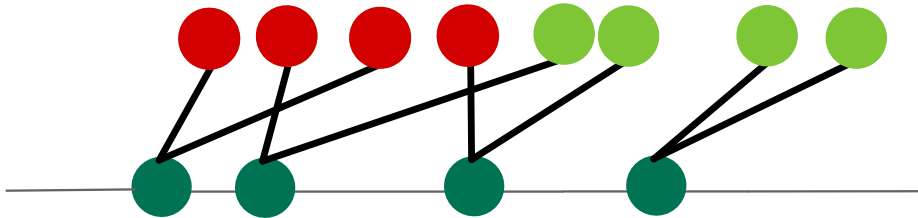
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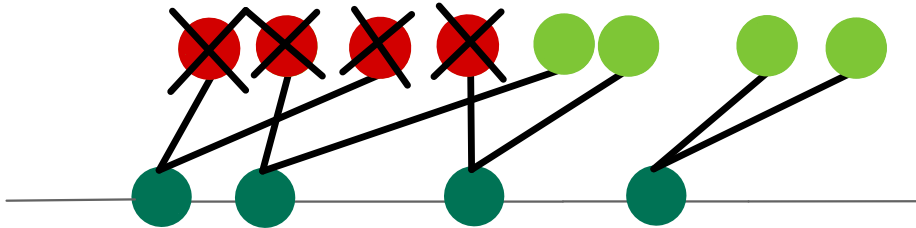
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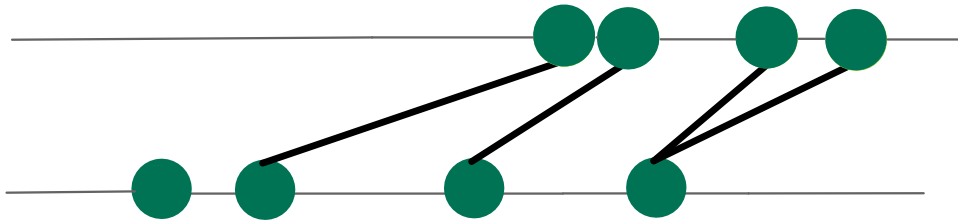
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and $\mathbb{P}(X \geq 0) = 1$ (no negative jumps).

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$$\text{Let } a_N = h^{-1}(2N \ell_N), \quad \text{where } h^{-1}(x) := \inf \{y \geq 0 : h(y) > x\}. \quad \text{space scale}$$

Main result

w.h.p. = with probability $\rightarrow 1$ as $N \rightarrow \infty$.

Theorem (P., Roberts, Talyigás 2021)

For $\eta > 0$, $k \in \mathbb{N}$ and $t > 4\ell_N$, the following occurs w.h.p.:

- **Spatial distribution:** At time t , there are $N - o(N)$ particles in

$$[X_i^{(N)}(t), X_i^{(N)}(t) + \eta a_N].$$

- **Genealogy:** The genealogy on an ℓ_N -timescale is asymptotically given by a star-shaped coalescent.

i.e. $\exists T \in [t - 2\ell_N, t - \ell_N]$ s.t. w.h.p., for a uniform sample of k particles at time t , every particle is descended from the rightmost particle at time T and no pair of particles in the sample has a common ancestor after time $T + \varepsilon_N \ell_N$, for any $(\varepsilon_N)_N$ with $\varepsilon_N \rightarrow 0$ and $\varepsilon_N \ell_N \rightarrow \infty$ as $N \rightarrow \infty$.

Warm up lemma

Recall $l_N = \lceil \log_2 N \rceil$.

Recall $X_1^{(N)}(t) \leq X_2^{(N)}(t) \leq \dots \leq X_N^{(N)}(t)$ ordered particle positions at time t .

Lemma For $s \in \mathbb{N}_0$, $X_1^{(N)}(s + l_N) \geq X_N^{(N)}(s)$.

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Lemma For $s \in \mathbb{N}_0$, $X_1^{(N)}(s+l_N) \geq X_N^{(N)}(s)$.

Proof: Suppose (for a contradiction) that $X_1^{(N)}(t) < X_N^{(N)}(s) \quad \forall t \in [s, s+l_N] \cap \mathbb{N}_0$.

Then since all jumps are non-negative, the rightmost particle at time s has 2^{l_N} descendants at time $s+l_N$.

Since $2^{l_N} \geq N$, this implies $X_1^{(N)}(s+l_N) \geq X_N^{(N)}(s)$. ~~///~~

So $\exists s^* \in [s, s+l_N] \cap \mathbb{N}_0$ s.t. $X_1^{(N)}(s^*) \geq X_N^{(N)}(s)$.

All jumps are ≥ 0 , so $X_1^{(N)}(s+l_N) \geq X_1^{(N)}(s^*)$. \square .

Construction of N-BRW from BRWs.

BRW: Initial particle at $x \in \mathbb{R}$ at time 0.

At each time $n \in \mathbb{N}_0$, each time- n particle has two offspring, each of which makes an independent jump from its parent's location with the same distribution as X .

The time- $(n+1)$ particles are these offspring particles.

Number of time- t particles is 2^t .

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Construction of N-BRW with initial particle locations x_1, \dots, x_N :

Take N independent BRWs with initial particles at x_1, \dots, x_N .

Colour BRW particles blue or red. All time-0 particles are blue.

For $n \in \mathbb{N}_0$, the N rightmost offspring particles of time- n blue particles are coloured blue. All other time- $(n+1)$ particles are coloured red.

Blue particles form an N-BRW.

Path of jumps from ancestor to descendant in N-BRW = path in one of the BRWs.

Random walk with heavy tailed jump distribution

$\mathbb{P}(X > x) = \frac{1}{h(x)}$, h regularly varying with index $\alpha > 0$.

X_1, X_2, X_3, \dots i.i.d. with $X_1 \stackrel{d}{=} X$.

Fix $c > 0$ small. For x v. large, unlikely that $\sum_{k=1}^n X_k \geq x$ and $X_k \leq cx \quad \forall k \leq n$.

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Lemma (Durrett '83, Gantert '00)

For $m \in \mathbb{N}$, $q > 0$, $\lambda > 0$ and $0 < r < 1 \wedge \frac{\lambda(1 \wedge \alpha)}{8q}$, for N sufficiently large, if

$x_N > N^\lambda$ then

$$\mathbb{P}\left(\sum_{j=1}^{m \wedge N} X_j \mathbb{1}_{X_j \leq r x_N} \geq x_N\right) \leq N^{-q}.$$

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Proof: Take $2q \log 2 < c < \frac{\lambda(1 \wedge \alpha) \log 2}{r}$.

By Markov's inequality,

$$\begin{aligned} \mathbb{P}\left(\sum_{j=1}^{ml_N} X_j \mathbb{1}_{X_j \leq rx_N} \geq x_N\right) &= \mathbb{P}\left(e^{cl_N x_N^{-1}} \sum_{j=1}^{ml_N} X_j \mathbb{1}_{X_j \leq rx_N} \geq e^{cl_N}\right) \\ &\leq N^{-\frac{c}{\log 2}} \mathbb{E}\left[e^{cl_N x_N^{-1}} X \mathbb{1}_{X \leq rx_N}\right]^{ml_N}. \end{aligned}$$

Use identity

$$\mathbb{E}\left[e^{vY} \mathbb{1}_{Y \leq K_2} \mathbb{1}_{Y \geq K_1}\right] = \int_{K_1}^{K_2} v e^{vu} \mathbb{P}(Y > u) du + e^{vK_1} \mathbb{P}(Y \geq K_1) - (e^{vK_2} - 1) \mathbb{P}(Y > K_2)$$

to show $\mathbb{E}\left[e^{cl_N x_N^{-1}} X \mathbb{1}_{X \leq rx_N}\right] = 1 + \mathcal{O}(N^{-\varepsilon})$ for some $\varepsilon > 0$. \square

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$$\mathbb{P}\left(\sum_{j=1}^{m\ell_N} X_j \mathbb{1}_{X_j \leq r x_N} \geq x_N\right) \leq N^{-q}.$$

Use with $x_N = \text{const} \cdot a_N$.

Recall $a_N = h^{-1}(2N\ell_N)$, so $a_N \sim (2N\ell_N)^{1/\alpha}$.

Fix $\epsilon \in (0, 1)$ small. A jump larger than ϵa_N is a "big jump".

For $c \gg \epsilon$, it is very unlikely that there is a time- t particle $> ca_N$ away from its time- $(t - \mathcal{O}(\ell_N))$ ancestor unless an ancestor made a big jump.

A jump is big w.p. $h(\epsilon a_N)^{-1} \sim e^{-\alpha} (2N\ell_N)^{-1}$,

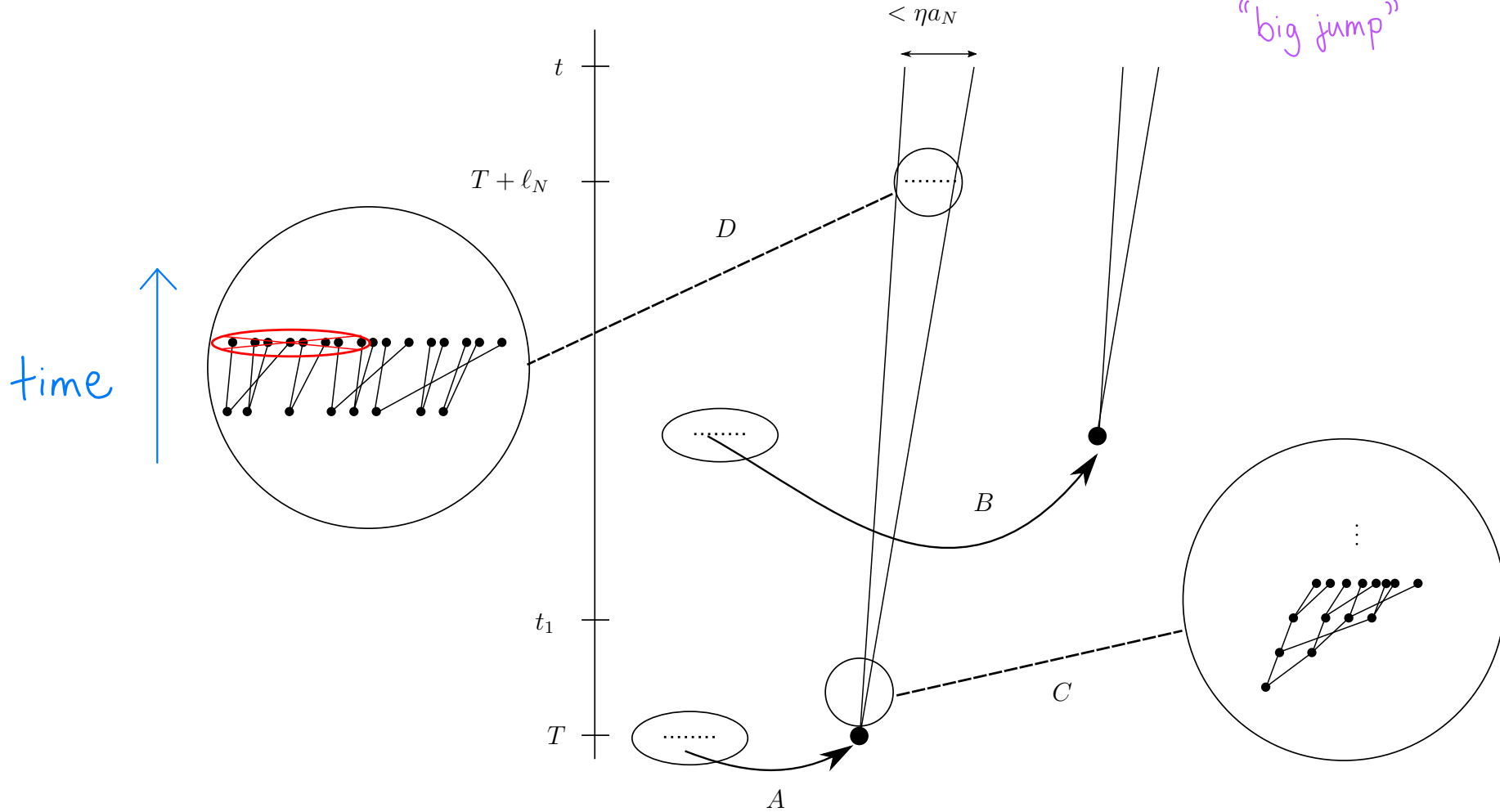
so we see big jumps at rate $\Theta(\ell_N^{-1})$.

Proof heuristics

$$t_{\perp} := t - l_N.$$

Let T = last time before time t_{\perp} when a particle makes a jump $\geq \rho a_N$ and takes the lead.

"big jump"

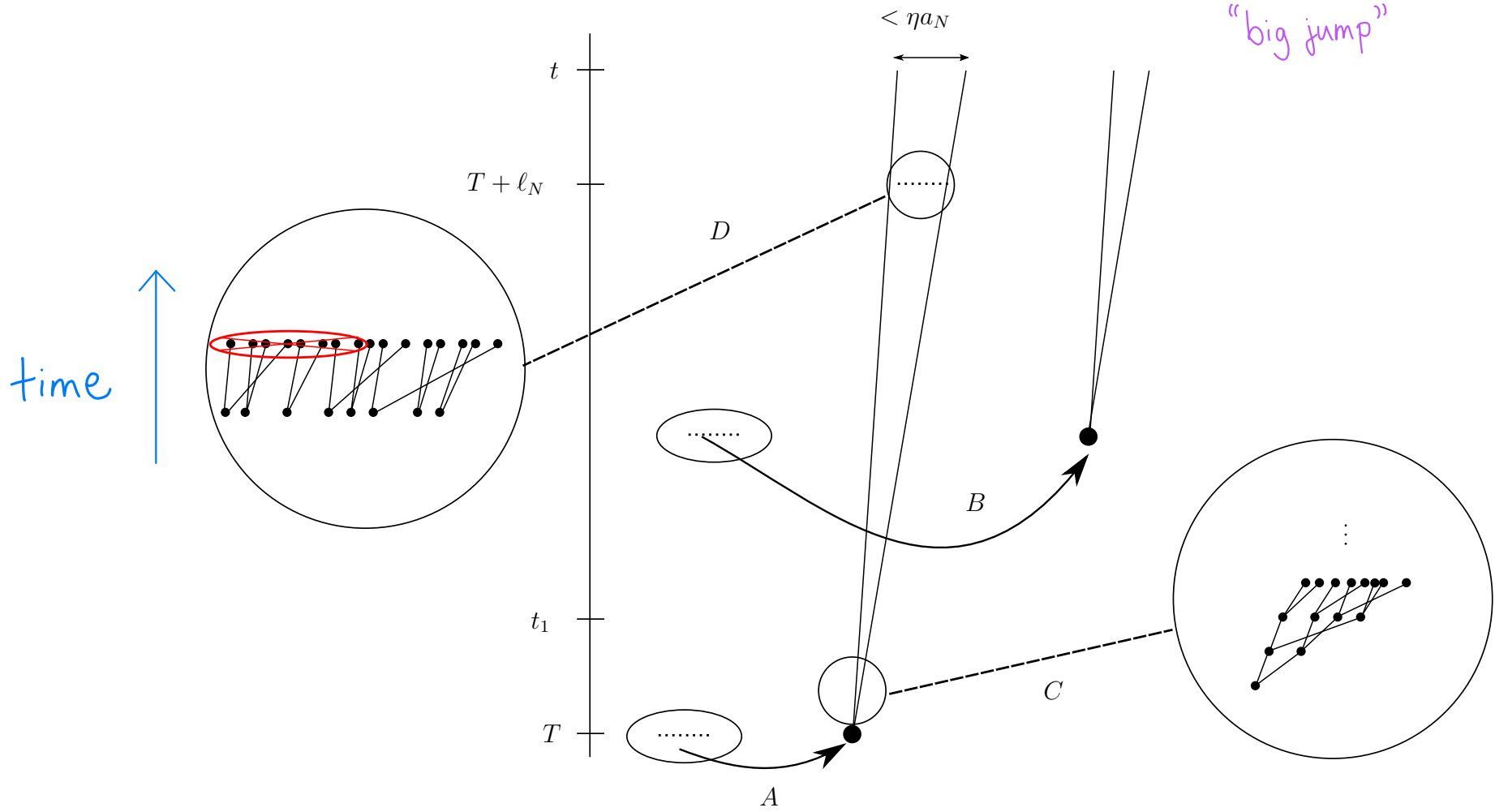


Proof heuristics

$$t_1 := t - \ell_N.$$

Let $T =$ last time before time t_1 when a particle makes a jump $\geq \rho a_N$ and takes the lead.

"big jump"



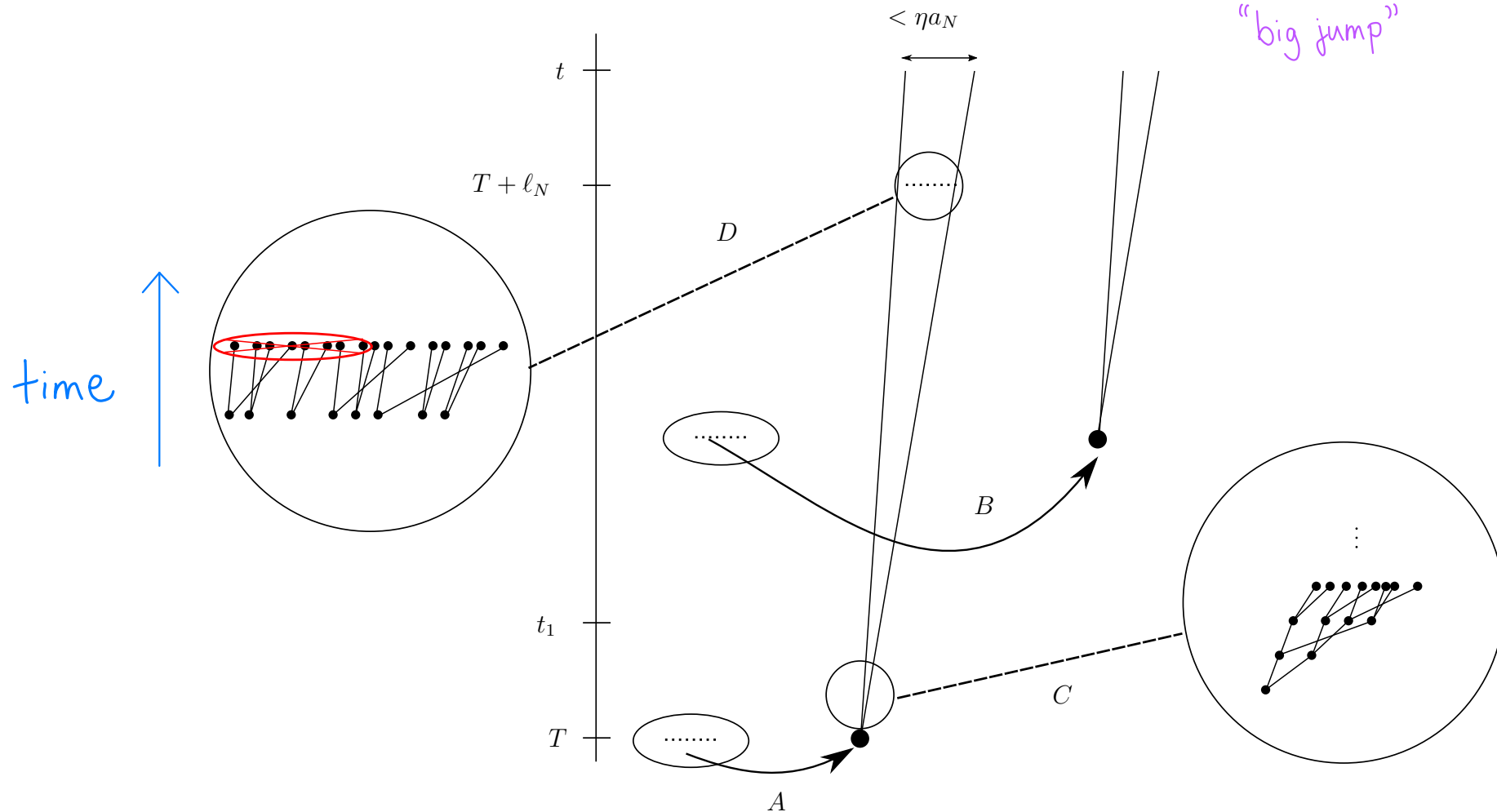
A: A particle makes a big jump at time T and takes the lead (by $\Theta(a_N)$). Its descendants stay in the lead until time t_1 (other particles can't take the lead with a big jump, and can't move far without a big jump).

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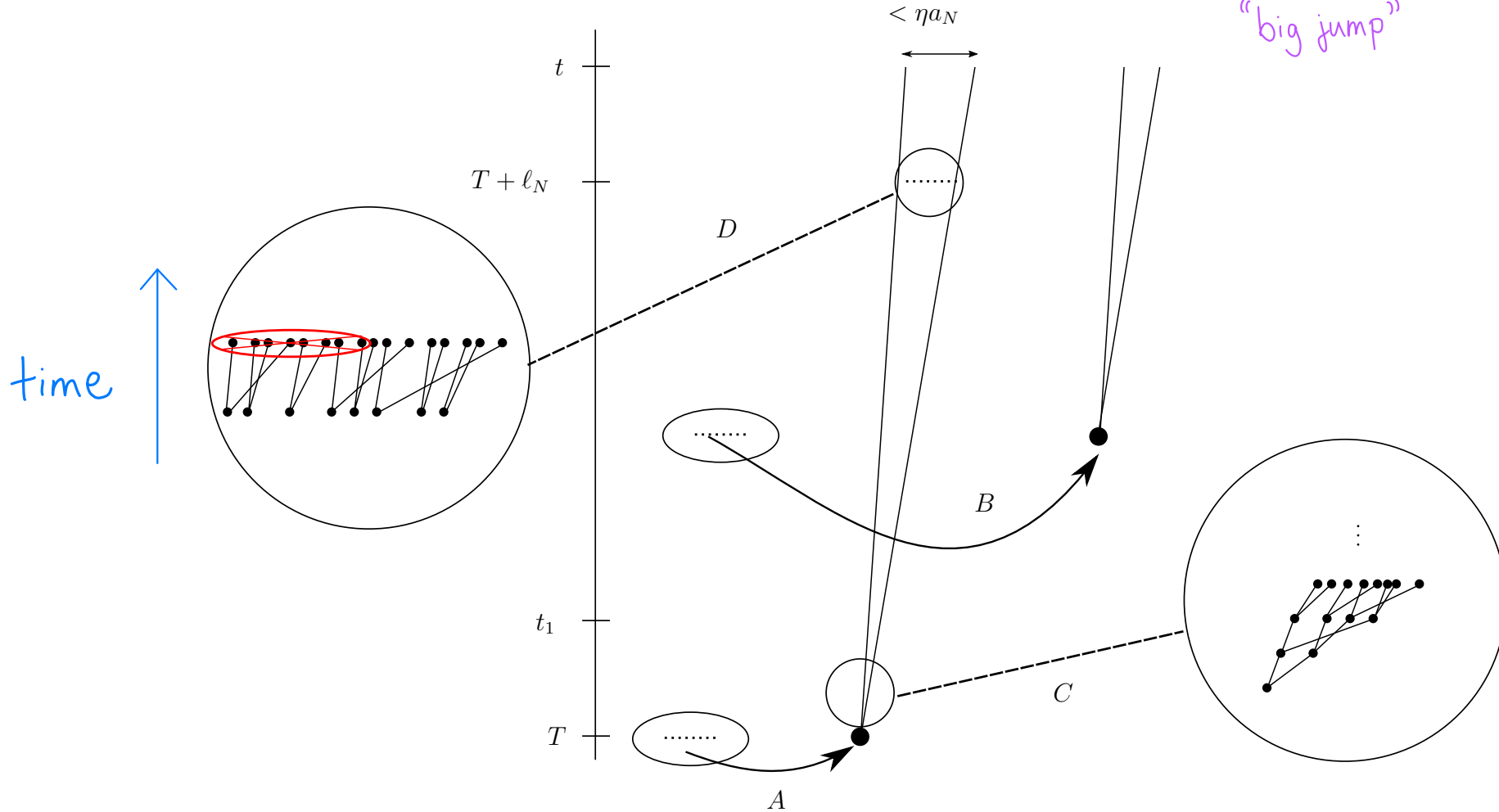
B: There are $O(1)$ big jumps in time interval $[t_1, t]$, each with $o(N)$ descendants at time t .

Proof heuristics

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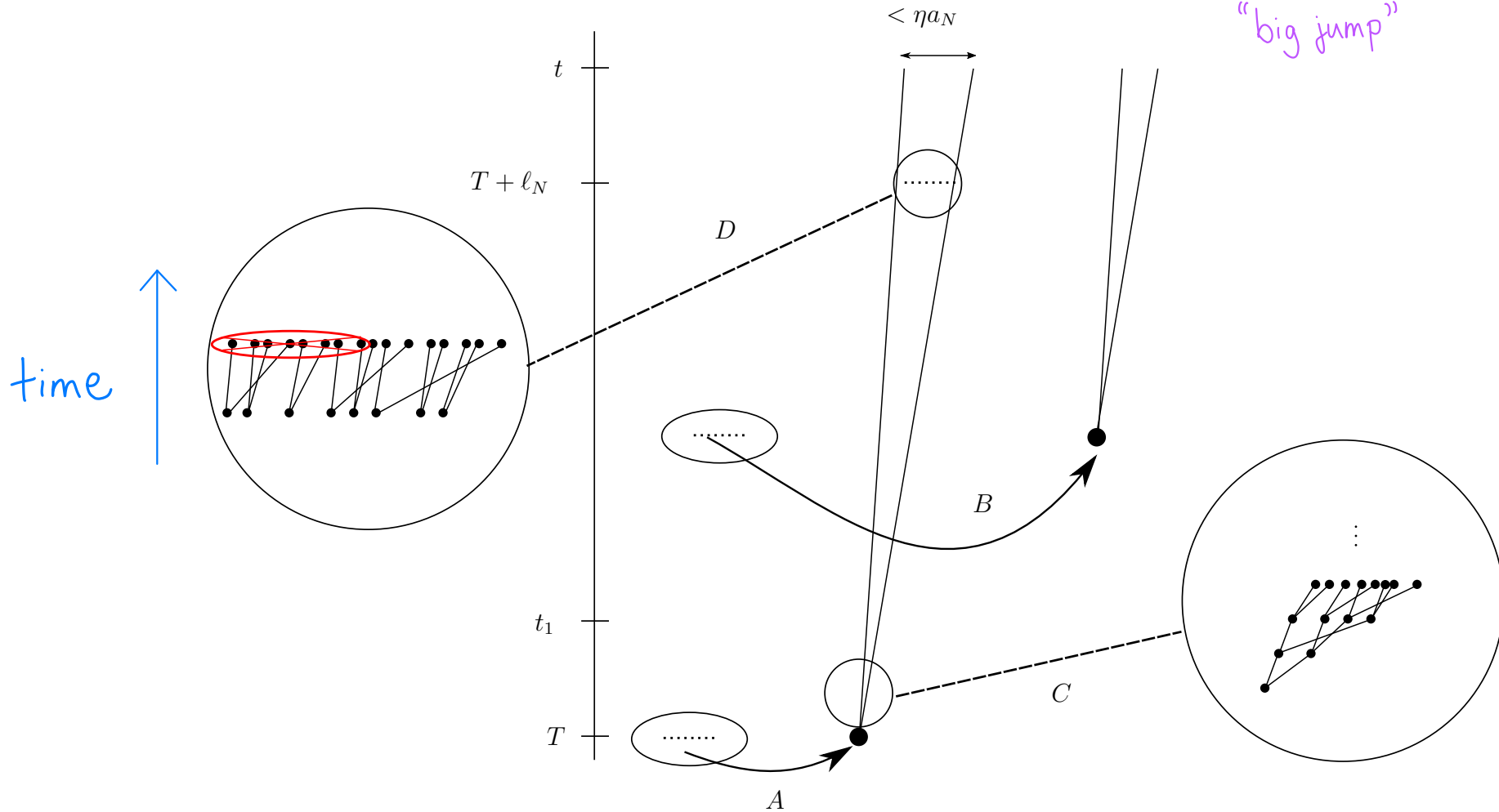
C : The tribe descended from the time- T leader doubles in size at each timestep until almost time $T + \ell_N$.

Proof heuristics

$$t_1 := t - \ell_N.$$

Let $T =$ last time before time t_1 when a particle makes a jump $\geq \rho a_N$ and takes the lead.

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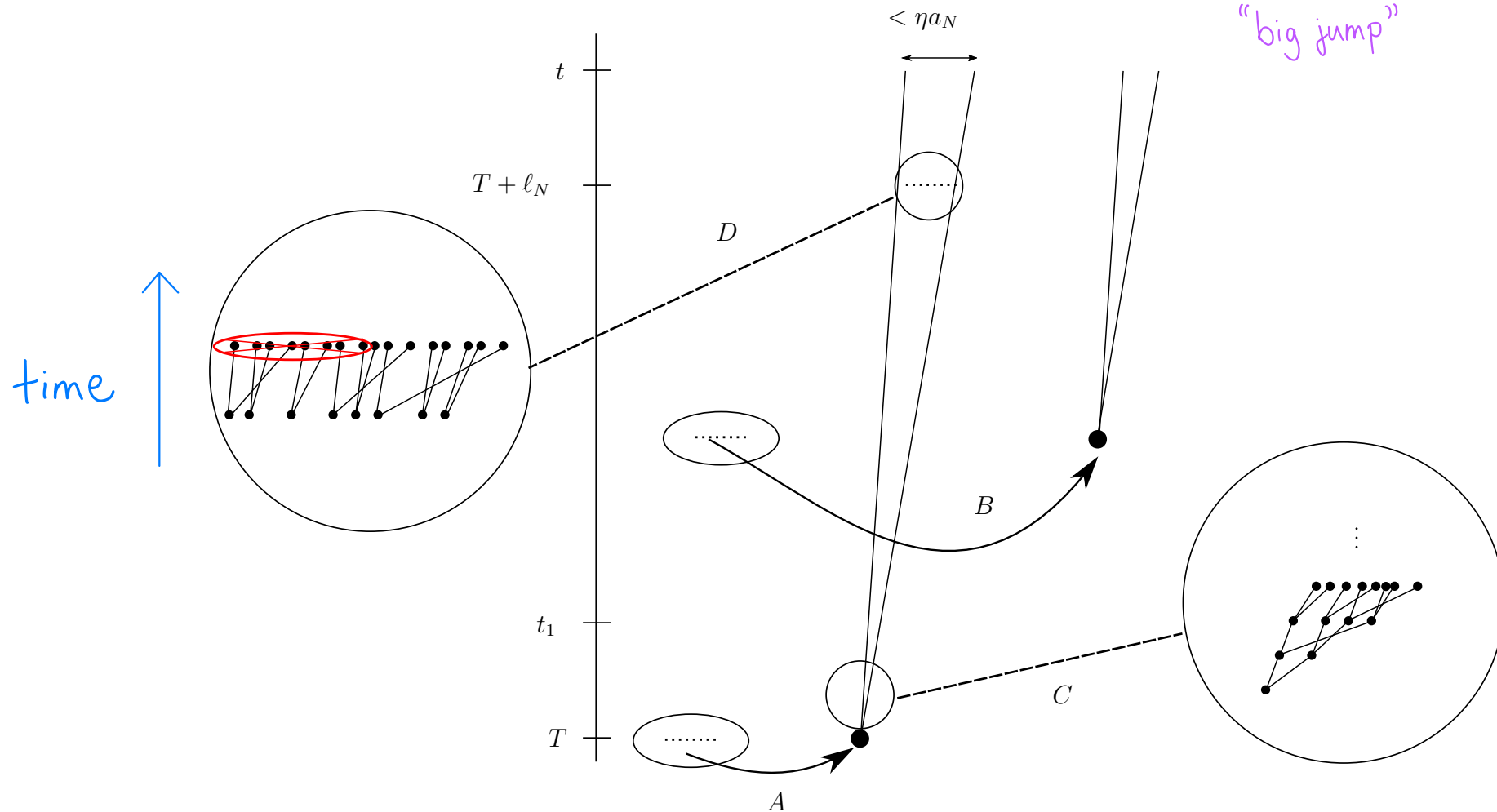
D: On the time interval $[T + \ell_N, t]$, the time- T leader's tribe has size $N - o(N)$.

Proof heuristics

$$t_1 := t - \ell_N.$$

Let $T =$ last time before time t_1 when a particle makes a jump $\geq \rho a_N$ and takes the lead.

"big jump"



$T \in [t_1 - \ell_N, t_1]$ w.h.p.:

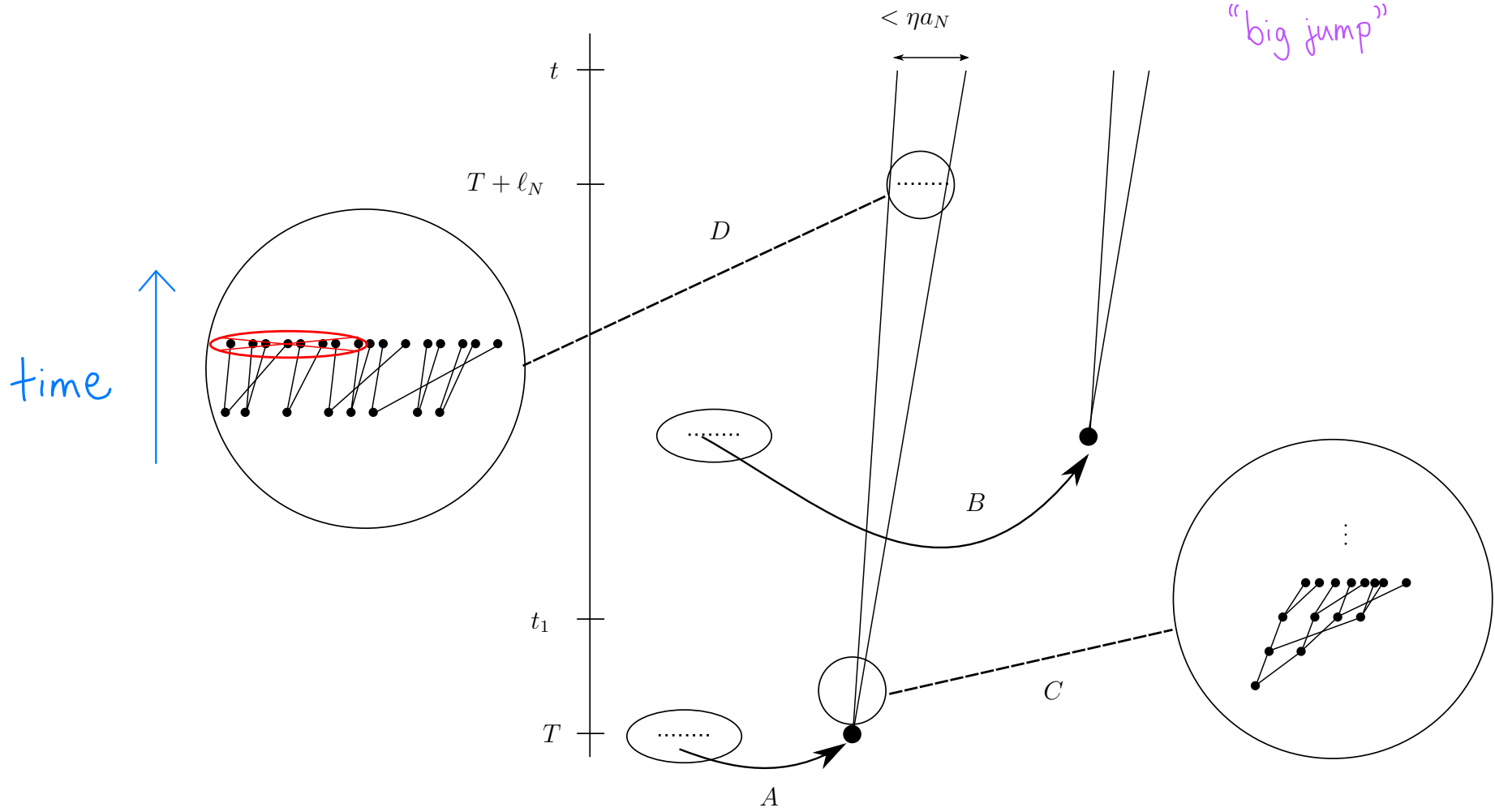
If no particle takes the lead with a big jump during $[s, s + \ell_N]$, then $X_1^{(N)}(s + \ell_N) \geq X_N^{(N)}(s)$ and so $\frac{1}{a_N} (X_N^{(N)}(s + \ell_N) - X_1^{(N)}(s + \ell_N))$ is small w.h.p.

Proof heuristics

$$t_1 := t - \ell_N.$$

Let $T =$ last time before time t_1 when a particle makes a jump $\geq \rho a_N$ and takes the lead.

"big jump"



$N - o(N)$ particles are close to leftmost at time t (on a_N space scale)

No big jumps in the time- T leader's tribe up to time $T + \ell_N > t_1$.

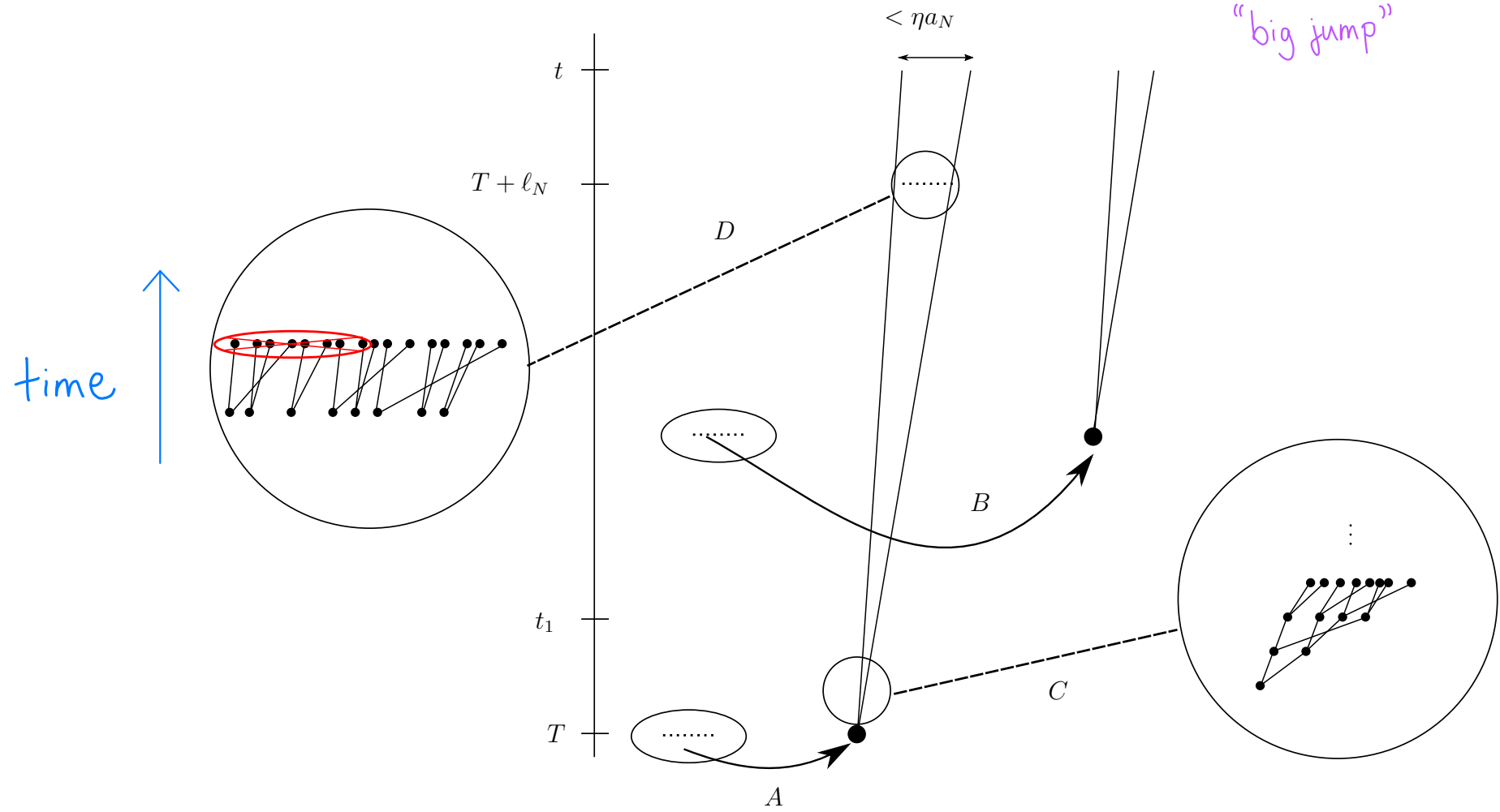
$O(1)$ big jumps in the tribe during $[T + \ell_N, t]$, each with $o(N)$ descendants.

Proof heuristics

$$t_{\perp} := t - \ell_N.$$

Let T = last time before time t_{\perp} when a particle makes a jump $\geq \rho a_N$ and takes the lead.

"big jump"



Star-shaped genealogy

No time- $(T + \varepsilon_N \ell_N)$ particles have $\Theta(N)$ time- t descendants.

None of the particles in the time- T leader's tribe have moved far by time $T + \varepsilon_N \ell_N$, so each has $\Theta(N 2^{-\varepsilon_N \ell_N}) = o(N)$ descendants at time t .