

Results for ARW on \mathbb{Z}

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Today: methods for ARW on \mathbb{Z}

Input for ARW:

- $\zeta \in [0, \infty)$ the 'mass' parameter
- $\lambda \in (0, \infty]$ the 'sleep rate'

The rules:

- The particles can be in two states: *active* and *sleepy*
- Each site starts with an active particle with probability ζ (iid)
- Active particles perform independent simple random walks at rate 1
- Sleepy particles do not move
- Active particles turn into sleepy particles at rate λ
- When two or more particles occupy the same site, all particles at that site become active

ARW **fixates** if each vertex is visited finitely many times.

Monotonicity + 0-1 law \implies critical density ζ_c

Existence of the critical density

For each $\lambda \in [0, \infty]$, there is a critical density $\zeta_c(\lambda) \in [0, 1]$ satisfying

$$\mathbb{P}(\text{ARW}(\zeta, \lambda) \text{ fixates}) = \begin{cases} 1, & \zeta < \zeta_c \\ 0, & \zeta > \zeta_c \end{cases}$$

Driven-dissipative version of ARW:

- 1 Start with any(!) initial configuration of particles on $[-N, N]^d \subset \mathbb{Z}^d$
- 2 Run ARW dynamics, killing particles that hit the boundary of the box
- 3 When all particles are asleep, add a particle at a random site
- 4 Return to step 2

Different from ARW on a graph: no mass conservation!

Questions

- Critical density ζ_c ?
- How quickly does the density converge? Fluctuations?
- Distribution of particles at criticality?

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Concrete problems:

- Time to fixate?
- Behavior of the odometer function?
- Existence of stationary distribution?

Results

Consider ARW on \mathbb{Z} .

Rolla, Sidoravicius '12

$$\zeta_c \in \left[\frac{\lambda}{1+\lambda}, 1\right].$$

Basu, Ganguly, Hoffman '15

$$\zeta_c \rightarrow 0 \text{ as } \lambda \rightarrow 0.$$

Hoffman, R., Rolla '20

$$\text{For any } \lambda > 0, \zeta_c < 1.$$

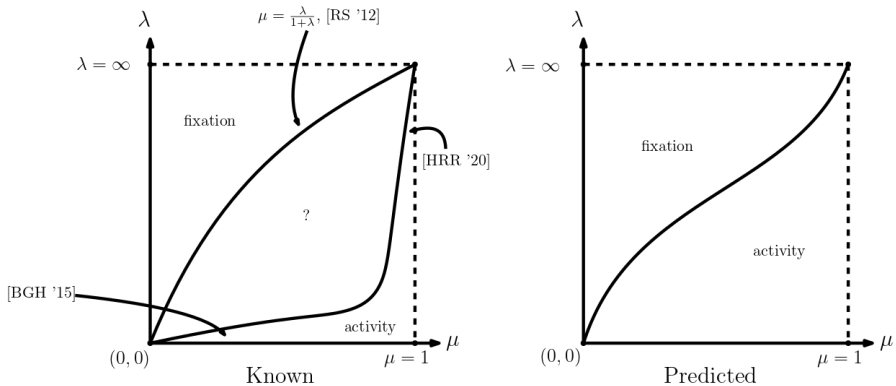


Figure: Phase diagram for ARW on \mathbb{Z} .

Let T denote the time to fixation on $G = \mathbb{Z}/n\mathbb{Z}$.

Basu, Ganguly, Hoffman, R. '17

- For any $\lambda \in (0, \infty]$ and $\zeta < \frac{\lambda}{1+\lambda}$,

$$T < Cn \log(n)^2$$

with high probability as $n \rightarrow \infty$ for some $C > 0$.

- For any $\zeta \in (0, 1)$, there exists $\lambda_0 > 0$ such that for $\lambda < \lambda_0$,

$$T > e^{cn}$$

with high probability as $n \rightarrow \infty$ for some $c > 0$.

Site-wise representation:

Define iid instructions $\{\xi_{v,j}\}$ for $v \in G$ and $j \geq 1$ by

$$\xi_{v,j} = \begin{cases} \text{move the particle at } v \text{ to a uniform neighbor} \\ \text{of } v \text{ with probability } \frac{1}{\deg(v)(1+\lambda)} \\ \text{put the particle at site } v \text{ to sleep with probability } \frac{\lambda}{1+\lambda} \end{cases}$$

At each site v , the j th time we topple a particle at v , the state of the system changes according to $\xi_{v,j}$.

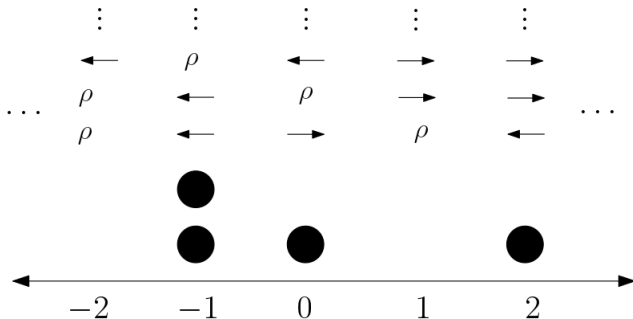


Figure: The stacks of instructions, visualized as lying above each vertex of the graph $G = \mathbb{Z}$. ρ represents a 'sleep' instruction.

Issue: what if the evolution of the system depends on the order of the instructions?

The order doesn't matter!

Abelian property

Consider any initial configuration η , and any (legal) sequence of stack instructions $\bar{\xi} = (\xi_{v^1, j^1}, \xi_{v^2, j^2}, \dots, \xi_{v^N, j^N})$. If $\bar{\xi}'$ is any (legal) re-ordering of the instructions in $\bar{\xi}$, then $\bar{\xi}\eta = \bar{\xi}'\eta$.

Key idea: we can choose clever toppling sequences.

Rolla, Sidoravicius '12

ARW fixates almost surely on $G = \mathbb{Z}$ for $\zeta < \frac{\lambda}{\lambda+1}$.

Proof sketch: find 'traps' for the particles to fall asleep in.

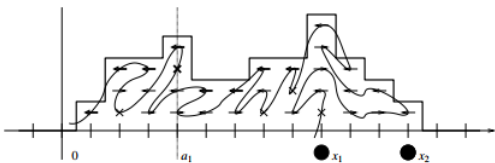


Figure: A diagram from [RS '12], showing the first trap a_1 for the particle x_1 .

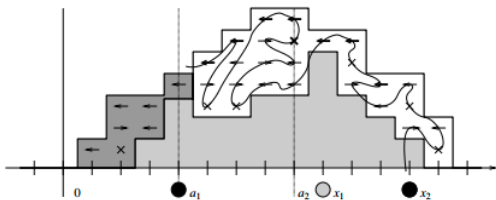


Figure: The trap a_2 for the particle x_2 , obtained recursively by exploring the stack instructions.

Let $x_k =$ position of k th particle to the right of 0, $k = 1, 2, \dots$

Define the traps a_k recursively:

- $a_0 = 0$.
- For $k > 0$: send a ghost particle out from x_k , ignoring sleep instructions, until it hits a_{k-1} .
- $a_k =$ leftmost site to the right of a_{k-1} where the second to last instruction seen by the ghost was a sleep instruction.

Particles follow the paths of their ghosts, except that they fall asleep in the trap.

Proof sketch: Trap setting succeeds if $a_{k-1} < x_k$ for all $k \in \mathbb{N}$.

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Thus if $\zeta < \frac{\lambda}{1+\lambda}$, $x_k > a_k$ for all k large a.s.

So $\mathbb{P}(\text{fixation}) > 0$. By the 0-1 law, $\mathbb{P}(\text{fixation}) = 1$. \square

Basu, Ganguly, Hoffman, R. '17

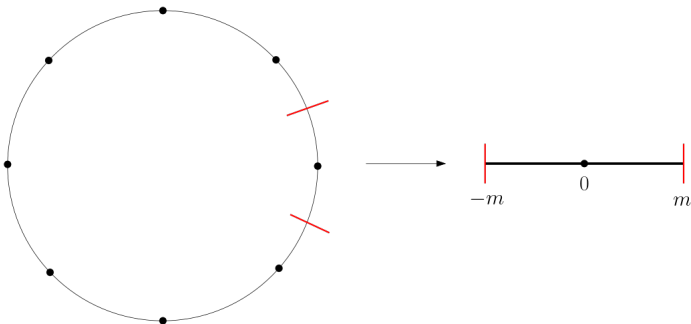
Consider ARW on $\mathbb{Z}/n\mathbb{Z}$ started from iid $\text{Ber}(\zeta)$ particles per site. For any $\lambda \in (0, \infty]$ and $\zeta < \frac{\lambda}{1+\lambda}$,

$$\# \text{ stack instructions to fixate} < Cn \log(n)^2$$

with high probability as $n \rightarrow \infty$ for some $C > 0$.

The fixation speed depends on the initial condition.

First step: gather $O(\log n)$ particles at each of $O\left(\frac{n}{\log n}\right)$ sites.



Focus on a single sub-interval.

How to adapt the traps for an interval?

Two-sided traps: ghosts start at 0, traps are set recursively at the boundary. Procedure fails if the traps reach 0.

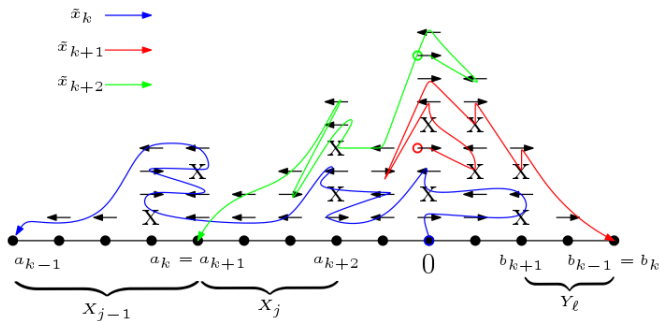


Figure: Setting traps 'in both directions' on an interval.

Internal erosion on an interval:

- 1 Start with the interval $X_0 = [-m, m] \cap \mathbb{Z}$.
- 2 Start a simple random walker from 0, stopped when she hits a boundary point $B \in \partial X_t$.
- 3 Remove the point B from X_0 , to obtain $X_{t+1} = X_t \setminus \{B\}$.
- 4 Return to step 2

Idea: replace each segment $[j - 1, j]$ and $[-j, -j + 1]$ by an independent $\text{Exponential}(j)$ length of rope, connect them all together, and initialize by lighting both ends on fire.

Properties of exponentials give a coupling between this process and the erosion process. Key computation: for $a, b > 0$,

$$\mathbb{P}^0(\text{hit } b \text{ before } -a) = \frac{a}{a+b} = \mathbb{P}(\text{Exp}(b) < \text{Exp}(a)).$$

(+ memoryless-ness)

Levine, Peres, '07

Let $R(m)$ be the number of sites remaining when the origin is eroded. As $m \rightarrow \infty$,

$$\frac{R(m)}{m^{3/4}} \rightarrow_d \left(\frac{8}{3}\right)^{1/4} \sqrt{|Z|},$$

where $Z \sim N(0, 1)$.

Note: the number of remaining sites is $O(m^{3/4}) = o(m)$.

Issue: at each stage, one of the traps moves a random distance – distributed as $\text{Geo}\left(\frac{1+\lambda}{\lambda}\right)$ – not distance 1.

We are still able to couple with the rope process, but the exponentials have random means. Many concentration estimates necessary.

Conclusion: the left and right side traps still shrink to 0 at the same rate (up to lower order stuff). Two-sided trap setting succeeds for $\zeta < \frac{\lambda}{1+\lambda}$.

Hoffman, R., Rolla '20

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Fix λ , choose ζ very close to 1, show that the odometer at 0 is infinite.

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Count how many particles exit a large interval $[0, N]$ after stabilizing.

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Initial condition is not iid (use RSZ '19)

Initial condition & dynamics:

- - free particle
- - carpet particle
- - transit regions
- - hole

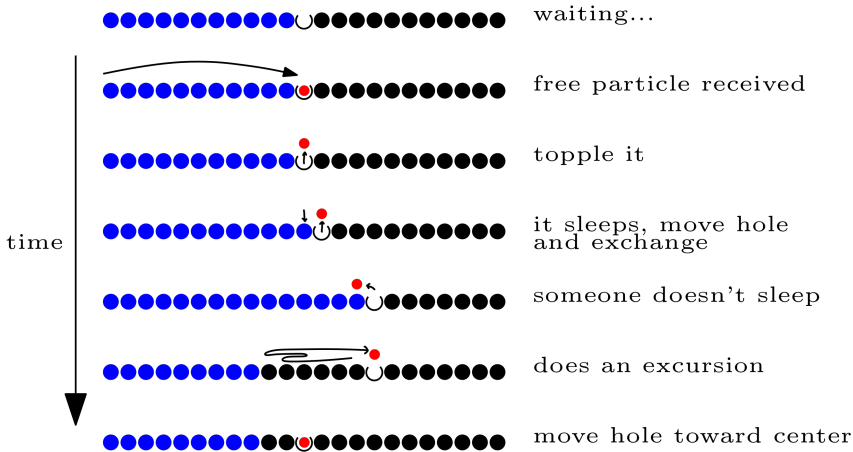
Topple left-most block with a free particle
Carpet sometimes gets turned into free
Transit regions are all active, but never move
Holes can move inside their blocks



Typical situation during the dynamics:

- - free particle
- - sleepy carpet particle
- - active carpet particle





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Frozen configuration (bad event)



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Solution: the hole has a strong bias towards 0. Expected maximum distance reached by a SRW excursion = ∞

Blocks / transit regions (i.e. ζ) must be sufficiently large.

Main lemma: the probability that the *last* free particle to interact with a given block causes a 'freeze' is small.

Formulating this is tricky: we condition on previous blocks, and bound uniformly over all numbers of free particle inputs.

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Successively condition on blocks, use single-block-estimate on each to upper bound total number of frozen blocks remaining at the end.