



## Effect of Small Noise on the Speed of Reaction-Diffusion Equations

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# The Voter Model (Single Site)



Fix  $N \in \mathbb{N}$ . Let  $\boldsymbol{\eta}(t) = (\eta_1(t), \dots, \eta_N(t)) \in \{0, 1\}^N$  be a continuous-time Markov chain. Each voter  $i \in \{1, \dots, N\}$  has **two independent** Poisson clocks:

- **Neutral clock** (rate  $R$ ): when it rings, voter  $i$  samples a uniformly random voter  $j \neq i$  and sets  $\eta_i := \eta_j$ .
- **Biased clock** (rate  $s > 0$ ): when it rings, voter  $i$  samples a uniformly random voter  $j \neq i$ . If  $\eta_j = 1$ , voter  $i$  adopts opinion 1 (sets  $\eta_i := 1$ ); otherwise nothing changes.

The **empirical fraction of 1-voters** is

$$p_t^N := \frac{1}{N} \sum_{i=1}^N \eta_i(t).$$



## Theorem (Scaling Limit)

Fix  $s > 0$ . Let  $R = R_N$  satisfy  $R_N/N \rightarrow \gamma/2 \in [0, \infty)$  as  $N \rightarrow \infty$ . Then  $p_t^N \Rightarrow p_t$  in  $D([0, \infty), [0, 1])$  with the Skorokhod topology, where  $p_t$  solves

$$dp_t = s p_t(1 - p_t) dt + \sqrt{\gamma p_t(1 - p_t)} dB_t.$$

## Remark

- $\gamma > 0$ : **Wright–Fisher diffusion with selection.**
- The diffusion coefficient vanishes at  $p = 0$  and  $p = 1$ ; these are absorbing boundaries.
- Durrett–Fan (2015) extended this to a spatial setting, yielding a Wright–Fisher SPDE

$$\partial_t u = \Delta u + s u(1 - u) + \sqrt{\gamma u(1 - u)} \dot{W}.$$

# The Fisher–KPP Equation



Setting  $s = 1$  and  $\gamma = 0$  (pure deterministic limit):

$$\partial_t u(t, x) = \Delta u(t, x) + u(t, x)(1 - u(t, x)), \quad t \geq 0, x \in \mathbb{R}.$$

Three mechanisms:

- **Diffusion** ( $\Delta u$ ): spatial spreading
- **Exponential growth** ( $+u$ ): unchecked reproduction at low density
- **Saturation** ( $-u^2$ ): logistic competition at high density

This is the **Fisher–KPP equation**

(Fisher, 1937; Kolmogorov–Petrovsky–Piscounov, 1937).



## Definition (Traveling Wave)

A function  $u(t, x) = W_c(x - ct)$  is a traveling wave of speed  $c$  if  $W_c \in C^2(\mathbb{R})$ ,  $0 \leq W_c \leq 1$ , and

$$W_c(-\infty) = 1, \quad W_c(+\infty) = 0.$$

## Theorem (Fisher 1937; KPP 1937)

- 1 Traveling wave solutions to the FKPP equation exist for every speed  $c \geq 2$ .
- 2 For **Heaviside-type** initial data  $u(0, \cdot) = \mathbf{1}_{(-\infty, 0]}$ , the solution converges to a traveling wave solution with minimal speed  $c_{\min} = 2$ .



Let  $\{\text{BBM}\}$  be branching Brownian motion:

- Particles move independently as Brownian motions with generator  $\Delta$  (variance rate 2).
- Each particle splits into two at rate 1.

Let  $M_t := \max\{\text{position of all particles alive at time } t\}$ , and write  $\mathbb{P}_x$  for the law starting from a single particle at  $x$ .

## Theorem (McKean, 1975)

*The function*

$$u(t, x) := \mathbb{P}_x(M_t \leq 0)$$

*solves the KPP equation  $\partial_t u = \Delta u + u(1 - u)$  with Heaviside-type initial condition  $u(0, \cdot) = \mathbf{1}_{(-\infty, 0]}(\cdot)$ .*



Using McKean's duality and fine analysis of the BBM front:

## Theorem (Bramson, 1978, 1983)

Let  $M_t$  be the position of the rightmost particle in BBM. Then  $\xi_t$  converges in distribution as  $t \rightarrow \infty$  where

$$M_t = 2t - \frac{3}{2} \log t + \xi_t \quad \text{as } t \rightarrow \infty.$$

## Remark

The  $-\frac{3}{2} \log t$  correction is now known as the Bramson shift. The law of  $\xi$  depends on the initial condition.



The KPP minimal speed is  $c = 2$ . What happens when the population is finite? Brunet and Derrida (1997) studied this through **three distinct models with population constraint**:

- 1 A **cutoff PDE** (Artificial analytical model)
- 2 An  **$N$ -particle selection system** (particle model)
- 3 An **SPDE** (More realistic PDE model with random fluctuations)

Each model produces a front that propagates at a speed  $v_N < 2$ .

# Model 1: Cutoff FKPP Equation



## Definition (Cutoff FKPP)

For  $N \geq 2$ , let  $u_N(t, x)$  solve

$$\partial_t u_N = \partial_{xx} u_N + u_N(1 - u_N) \cdot \mathbf{1}_{\{u_N \geq 1/N\}}.$$

- The reaction term is **disabled** when  $u_N < 1/N$ .
- Physically: a population of size  $N$  cannot sustain densities below  $1/N$ , where  $N$  is called the **carrying capacity** (effective population scale).

## Remark

*This is mathematically convenient but physically artificial. A more faithful discrete model is needed.*

## Model 2: $N$ -BBM (Selection)



### Definition ( $N$ -BBM)

A system of **exactly**  $N$  **particles** on  $\mathbb{R}$ :

- Each particle moves as Brownian motion with generator  $\Delta$ .
- Each particle splits into two at rate 1.
- Immediately after each branching, the leftmost particle is killed, restoring the total to  $N$ .

### Remark

- $N$  is a **hard constraint** – the actual number of particles is exactly  $N$  at all times.
- This models genetic drift: only the “fittest” (rightmost) individuals survive.
- The selection mechanism pushes the front slower than the unconstrained BBM.



In the Durrett–Fan framework, Brunet–Derrida suggest taking  $R = 1$  (fixed) and  $N \rightarrow \infty$ :

## Definition (Stochastic FKPP)

For each  $N \geq 1$ , let  $w_N(t, x)$  solve

$$\partial_t w_N = \partial_{xx} w_N + w_N(1 - w_N) + \sqrt{\frac{w_N(1 - w_N)}{N}} \dot{W},$$

where  $\dot{W}$  is space–time white noise.

- Noise intensity  $\sim 1/\sqrt{N}$ : the finite-population fluctuation.
- $N \rightarrow \infty$  formally recovers the deterministic FKPP.
- **Well-posedness**: Shiga (1988) via moment duality with coalescing Brownian motion.



## Conjecture (Brunet–Derrida, 1997; BDMM, 2006)

For any of the three models above, the asymptotic front velocity  $v_N := \lim_{t \rightarrow \infty} \frac{1}{t} X_N(t)$  (where  $X_N(t)$  is the front position, or  $p$ -quantile position, whichever makes sense) admits the expansion:

$$v_N = 2 - \frac{\pi^2}{(\log N)^2} + \frac{6\pi^2 \log \log N}{(\log N)^3} + o\left(\frac{\log \log N}{(\log N)^3}\right) \quad \text{as } N \rightarrow \infty.$$

The first correction term was proposed by Brunet–Derrida (1997).

The second-order term was proposed by Brunet, Derrida, Mueller & Munier (2006).

# Status of the Conjecture (from what I know)



Term	Status
2 (FKPP minimal speed)	Trivial
$-\frac{\pi^2}{(\log N)^2}$	<b>Proved</b> Bérard–Gouéré 2010 ( $\asymp$ order only) MMQ 2011 (precise coefficient)
$+\frac{6\pi^2 \log \log N}{(\log N)^3}$	<b>Open</b> Consistent with Maillard 2016 (fluctuations)
$o\left(\frac{\log \log N}{(\log N)^3}\right)$	<b>Open</b>



## Theorem (Bérard–Gouéré, 2010)

For  $N$ -BRW (discrete-time branching random walk on  $\mathbb{R}$  with selection):

$$v_N \rightarrow 2 \quad \text{and} \quad 2 - v_N \asymp \frac{1}{(\log N)^2}.$$

- First proof of the  $(\log N)^{-2}$  **scaling**.
- Does **not** establish the precise coefficient  $\pi^2$ .
- Technique: coupling with a branching random walk killed upon exiting a moving interval.



## Theorem (MMQ, Inventiones Math. 2011)

For the Wright–Fisher SPDE model:

$$v_N = 2 - \frac{\pi^2}{(\log N)^2} + \mathcal{O}\left(\frac{\log \log N}{(\log N)^3}\right).$$

- First rigorous proof of the **precise coefficient**  $\pi^2$ .
- Key insight: the front is shaped by the behavior near “the tip” at distance  $O(\log N)$  from the leading edge.
- The error term  $\mathcal{O}\left(\frac{\log \log N}{(\log N)^3}\right)$  is of the *same order* as the conjectured second correction. Resolving whether the  $\mathcal{O}$  hides a  $6\pi^2 \log \log N$  term remains open.



## Theorem (Maillard, PTRF 2016)

For  $N$ -BBM, let  $\nu_t^N$  be the counting measure at time  $t$ :

$$\nu_t^N = \sum_{i=1}^N \delta_{X_i(t)}.$$

Define the  $\alpha$ -quantile:  $\text{med}_\alpha(\nu_t^N) := \inf\{x : \nu_t^N([x, \infty)) < \alpha N\}$ . Then

$$\frac{\text{med}_\alpha(\nu_{\log^3 N}^N)}{\log^3 N} = 2 - \frac{\pi^2}{(\log N)^2} + \frac{6\pi^2 \log \log N + \zeta_N}{(\log N)^3},$$

where  $\zeta_N \Rightarrow \zeta$  converges weakly as  $N \rightarrow \infty$ .



## Important Distinction

Maillard's result is about the front position at a fixed (albeit large) time  $T_N = \log^3 N$ . It does **not** establish the existence of an asymptotic **velocity**:

$$v_N := \lim_{t \rightarrow \infty} \frac{X_N(t)}{t}.$$

- The conjectured expansion is for  $v_N$ , the long-time speed.
- Maillard's expansion is for  $X_N(T_N)/T_N$  at a specific time scale.
- These are **related but not equivalent**: the random fluctuation  $\zeta_N$  could vanish or shift in the  $t \rightarrow \infty$  limit.
- Hence: the second-order term for the **velocity** remains open.



Consider the single-site voter model with **state-dependent** biased voting rate:

- **Neutral**: voter  $i$  samples a uniformly random  $j \neq i$  and sets  $\eta_i := \eta_j$  at rate  $R$ .
- **Biased (state-dependent)**: voter  $i$  samples a uniformly random  $j \neq i$ , and if  $\eta_j = 1$  adopts opinion 1 (sets  $\eta_i := 1$ ). The attempt rate is  $s(p_t^N) = (p_t^N)^{-1/2}$ .

If  $s(p) = 1/\sqrt{p}$ , then under the same scaling  $R/N \rightarrow \gamma/2$ ,  $p_t^N$  converges in distribution to  $p_t$  solving

$$dp_t = \sqrt{p_t(1-p_t)} dt + \sqrt{\gamma p_t(1-p_t)} dB_t.$$

The drift  $b(p) = \sqrt{p}(1-p)$  is **non-Lipschitz** at  $p = 0$ :  $b'(0+) = \infty$ . This stronger drift near  $p = 0$  reflects *intensified selection when the favored allele is rare*.



Generalizing the voter model to a spatial setting yields:

## Definition

For  $\alpha \in (0, 1)$  and  $N \geq 1$ , consider

$$\partial_t u = \partial_{xx} u + b_\alpha(u) + \sqrt{\frac{u(1-u)}{N}} \dot{W},$$

where  $b_\alpha(u) = u^\alpha(1-u)$ .

- $b_\alpha$  is **non-Lipschitz** at  $u = 0$  when  $\alpha < 1$ :  $b'_\alpha(0+) = \infty$ .
- This raises fundamental **well-posedness questions**.
- Mueller–Mytnik–Ryzhik (2021): weak uniqueness for  $\alpha \geq 1/2$ .



## Theorem (Barnes–Mytnik–Sun, 2024 – Wave Speed)

For  $\alpha \in [1/2, 1)$ , the asymptotic speed satisfies:

$$V_{\alpha, N} \asymp N^{\frac{1-\alpha}{1+\alpha}} \quad \text{as } N \rightarrow \infty.$$

## Theorem (Barnes–Mytnik–Sun, 2026)

For every  $\alpha \in [0, 1)$ , the SPDE

$$\partial_t u = \partial_{xx} u + b_\alpha(u) + \sqrt{\frac{u(1-u)}{N}} \dot{W}$$

with  $b_\alpha(u) = u^\alpha(1-u)$  for  $\alpha \in (0, 1)$  and  $b_0(u) = \mathbf{1}_{(0,1]}(u) - u$  admits a **unique-in-law** martingale solution.

# Conjecture: Power-Law Speed for All $\alpha \in [0, 1)$



## Conjecture (Barnes–Mytnik–Sun, 2024)

The scaling  $V_{\alpha, N} \asymp N^{(1-\alpha)/(1+\alpha)}$  persists for all  $\alpha \in [0, 1)$ .

Current status:

- $\alpha \in [1/2, 1)$ : proved (BMS 2024).
- $\alpha \in [0, 1/2)$ : **open** – non-Lipschitz singularity makes the analysis more delicate.



Consider  $N$ -BBM with heavy-tailed offspring:

$$\mathbb{P}(\text{offspring} = k) \sim C_\alpha k^{-1-\alpha}, \quad k \rightarrow \infty, \quad \alpha \in (0, 1).$$

What is the scaling of the front velocity  $V_{\alpha, N}^{\text{BBM}}$ ?

**Private Communication (Berestycki-Maillard-Pain-Sun, 2024), Simulation Evidence (Sun Jian, Master's Thesis 2025)**

$$V_{\alpha, N}^{\text{BBM}} \asymp N^{\frac{1-\alpha}{2}} \ll N^{\frac{1-\alpha}{1+\alpha}} \asymp V_{\alpha, N}^{\text{SPDE}}.$$



Consider one particle at random produces  $K_\alpha$  offspring with

$$\mathbb{P}(K_\alpha = k) \sim C_\alpha k^{-1-\alpha}, \quad \alpha \in (0, 1).$$

Immediately after branching, we have  $N + K_\alpha - 1$  particles. However, the selection kills the **leftmost** particles to maintain the total population to be  $N$ , so the effective number of offspring is no more than

$$\tilde{K}_\alpha = \min(K_\alpha, N).$$

Therefore, the model can be dominated by a BBM with offspring distribution  $\tilde{K}_\alpha$  whose velocity is

$$2\sqrt{\mathbb{E}[\tilde{K}_\alpha]} \asymp \sqrt{N^{1-\alpha}} = N^{\frac{1-\alpha}{2}}.$$



Thank You!