

# Symbiotic Branching

## Compact Interface Property

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## Part I

(3-Step Motivation of Symbiotic Branching)

## Part II

(Compact Support Property)

# Part I -

## 3-Step Motivating of Symbiotic Branching

## Step 1

A stochastic evolution equation (SPDE):

### Heat Equation with Feller Noise

$$\frac{\partial u}{\partial t}(t, x) = \frac{1}{2} \Delta u(t, x) + \sqrt{u(t, x)} \dot{W}_t$$

*with initial condition  $u(0, x) = u_0(x)$  and a white noise  $\dot{W}$ .*

$u_0$  is non-negative with subexponential growth. (Always:  $t \geq 0$ ,  $x \in \mathbb{R}$ )

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- the 'branching' term
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## White Noise/Weak Solutions

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*White noise on a  $\sigma$ -finite measure space  $(E, \mathcal{E}, \mu)$  is a mean zero Gaussian process indexed by measurable, finite measure subsets of  $E$  with covariance function*

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→ weak solution satisfy for  $\phi \in C_c^2$  the integral equation

$$\begin{aligned} \int u(t, x) \phi(x) dx &= \int u_0(x) \phi(x) dx + \frac{1}{2} \int_0^t \int u(s, x) \Delta \phi(x) dx ds \\ &+ \int_0^t \int \sqrt{u(s, x)} \phi(x) \dot{W}(ds, dx), \quad a.s. \end{aligned}$$

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## From Galton-Watson to Feller's Diffusion

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There are  $Z_0 = x$  males at time 0 and

$$Z_{n+1} = \sum_{k=1}^{Z_n} X_n^k$$

at time  $n + 1$ . The  $X_n^k$  are iid with mean 1 (critical) and variance  $\sigma^2 < \infty$ .  $X_n^k$  denotes the number of sons of male  $k$  at time  $n$ .

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rescaled process  $Z_t^N$  converges to  $X_t$ , solution of Feller's Diffusion

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→ connection of branching and  $\sqrt{\dots}$  (note: no space yet)

## From SuperBrownianMotion to Stochastic PDEs

- SuperBrownianMotion (SBM)  $U_t$ : spatial generalization of Feller's Diffusion
- recall:  $U_t(\omega)$  is a measure
- $U_t$  has a density only in spatial dimension 1, i.e. there is  $u(t, x)$  such that

$$U_t(A) = \int_A u(t, x) dx$$

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→ spatial movement yields additional  $\frac{1}{2}\Delta$

## Step 2

A stochastic evolution system:

### Mutually Catalytic Branching Model

$$\begin{aligned}\frac{\partial u_1}{\partial t}(t, x) &= \frac{1}{2} \Delta u_1(t, x) + \sqrt{u_1(t, x) u_2(t, x)} \dot{W}_t^1 \\ \frac{\partial u_2}{\partial t}(t, x) &= \frac{1}{2} \Delta u_2(t, x) + \sqrt{u_2(t, x) u_1(t, x)} \dot{W}_t^2\end{aligned}$$

with initial condition  $u_1(0, x) = u_1(x)$ ,  $u_2(0, x) = u_2(x)$ .

$\dot{W}^1, \dot{W}^2$  are independent white noises,  $u_1, u_2$  are non-negative with subexponential growth.

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- the 'interaction' terms

## From Catalytic SBM to Mutually Catalytic Branching

Consider a SuperBrownianMotion  $U_1$  where

- branching rate depends on spatial position ( $\rightarrow$  catalytic)
- branching rate depends on a second species  $U_2$ : rate is proportional to the amount of the second species at this site
- assume the second species  $U_2$  is also given by a SBM

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- one-way interaction

$\rightarrow$  two-way interaction yields the terms

$$\sqrt{u_1(t, x)u_2(t, x)}, \sqrt{u_2(t, x)u_1(t, x)}$$

## Step 3

A stochastic evolution system:

### Symbiotic Branching Model (Etheridge/Fleischmann (2004))

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with initial condition  $u_1(0, x) = u_1(x)$ ,  $u_2(0, x) = u_2(x)$ .

$u_1, u_2$  are non-negative with subexponential growth and the white noises are now **correlated**:

$$E[\dot{W}^1(t, x)\dot{W}^2(t', x')] = \varrho\delta(t, t')\delta(x, x')$$

for  $\varrho \in [-1, 1]$ .

We have seen so far that

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- comes from the mutually catalytic behaviour
- additional correlation

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goals: any properties that change their behaviour with the correlation parameter  $\rho$



# Part II -

## Compact Interface Property for Symbiotic Branching

From now on we restrict to  $u_1 = \mathbb{1}_{\mathbb{R}^+}$ ,  $u_2 = \mathbb{1}_{\mathbb{R}^-}$ . For a realization of  $(u_1, u_2)$  the interface is defined as:

### Interface

$$Int_t(\omega) = cl\{x \mid u_1(t, x)(\omega)u_2(t, x)(\omega) \neq 0\}$$

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## Question

*Does the compact interface property hold?*

*i.e. is*

$$\bigcup_{t \leq T} Int_t$$

*a.s. compact?*

## Answer (Etheridge/Fleischmann (2004))

- For each  $\varrho \in [-1, 1]$  there is a set  $\Omega_1$  of measure 1 such that for each  $\omega \in \Omega_1$  and  $T > 0$

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is compact.

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is compact.

- More precisely, there is a positive constant  $c$  and a random time  $T_0$  such that

$$\bigcup_{t \leq T} \text{Int}_t(\omega) \subset [-cT, cT]$$

for each  $T \geq T_0(\omega)$ .

## Conjecture

There is a constant  $c$ , a random time  $T_0$  and a set  $\Omega_1$  of measure 1 such that for  $\varrho < 0$  each  $\omega \in \Omega_1$

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(for  $\varrho = -1$  and  $u_1(t, x) + u_2(t, x) = 1$  (heat-equation with 'Wright-Fisher noise') Tribe (1995) proved propagation with  $\sqrt{T}$ .)



## How to prove the Conjecture

The proof of Etheridge/Fleischmann is based on estimates of

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- if we can manage to give much better bounds, the rest of their proof implies the conjecture

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techniques to estimate the moments:

- perturbed duality
- moment equations

## Using Generators to Prove Duality

Given markov processes  $X_t, N_t$  with statespaces  $X, N$ , generators  $\Omega^X, \Omega^N$  and a duality function  $H : X \times N \rightarrow \mathbb{R}$  bounded.

### Proposition (duality)

$$\begin{aligned}\Omega^X H(\cdot, n)(x) &= \Omega^N H(x, \cdot)(n) \\ \Rightarrow \mathbb{E}^{X_0}[H(X_t, N_0)] &= \mathbb{E}^{N_0}[H(X_0, N_t)]\end{aligned}$$

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Further, for bounded  $f$

### Proposition (perturbed duality)

$$\begin{aligned}\Omega^X H(\cdot, n)(x) &= \Omega^N H(x, \cdot)(n) + f(n)H(x, \cdot)(n) \\ \Rightarrow \mathbb{E}^{X_0}[H(X_t, N_0)] &= \mathbb{E}^{N_0} \left[ H(X_0, N_t) \exp\left(\int_0^t f(N_s) ds\right) \right]\end{aligned}$$

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→ nice generators are nice to prove duality

# A Perturbated Duality in Symbiotic Branching

Etheridge/Fleischmann:

- used a dual process to estimate the moments
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$$\begin{aligned} & \mathbb{E}[u_1(t, x)u_2(t, x)]^q \\ = & \mathbb{E}^{N_0} \left[ (\mathbb{1}_{\mathbb{R}^+}, \mathbb{1}_{\mathbb{R}^-})^{N_t} \exp(L_{[N_t^{red}, N_t^{red}]} + L_{[N_t^{blue}, N_t^{blue}]} + \varrho L_{[N^{red}, N^{blue}]}) \right] \end{aligned}$$

where

$$(\mathbb{1}_{\mathbb{R}^+}, \mathbb{1}_{\mathbb{R}^-})^{N_t} = \begin{cases} 1, & \text{all } B_t^{red} \geq 0, \text{ all } B_t^{blue} \leq 0 \\ 0, & \text{otherwise} \end{cases}$$

goal: use the dual process to prove better moment bounds



Thank You For Listening

# Existence and Uniqueness

## Existence

*There is a weak solution with paths a.s. in  $C(\mathbb{R}^+, C_{tem})$ .*

## Uniqueness in Law

*Uniqueness in law is known.*

## Pathwise Uniqueness

*Pathwise uniqueness is unknown. But: known for certain coloured noises*