

Super-Brownian Motion, Critical Spatial  
Stochastic Systems  
and Measure-Valued Diffusions

Ed Perkins  
University of British Columbia

Fields Institute  
November 5, 2003.

## 1. Brownian Motion:

Start at 0 in  $\mathbb{Z}^d$ , take ind't random steps  $(W_i)$  to neighbours with equal probability.

$S_n = \sum_{i=1}^n W_i =$  position at time  $n$ .

Rescale time by  $N$ , space by  $1/\sqrt{N}$ :  
 $B^{(N)}(t) = S_{[Nt]}/\sqrt{N}$ ,  $t \geq 0$ .

**Theorem 1.1** (CLT, Demoivre 1718,...,  
Lindeberg 1922)

$$\lim_{N \rightarrow \infty} P(B^{(N)}(1) \in I) = \int_I e^{-d|x|^2/2} (2\pi/d)^{-d/2} dx.$$

**Theorem 1.2** (FCLT, Donsker 1951)

$$\lim_{N \rightarrow \infty} P(B^{(N)} \in A) = P(d^{-1/2} B \in A)$$

"for any" set  $A$  of paths from  $\mathbb{R}_+$  to  $\mathbb{R}^d$ .

$B$  is standard Brownian motion (Wiener 1923).

## Universality of $B$

- $\{W_i\} \in \mathbb{R}^d$  any repeated independent random quantities with mean 0 and covariance  $\sigma^2 I_{d \times d}$  leads to  $\sigma B$  in limit.  $S_n$  could be a fair game.
- Instead of  $t = i/N$  may take steps "at rate  $N$ ", i.e. at random times  $T_i$ ,  $P(T_i - T_{i-1} > t) = e^{-Nt}$  (mean  $1/N$  exponential inter-step times).
- $B$  is a stochastic model for particle suspended in liquid (Brown 1828), stock market fluctuations (Bachelier 1900), as building block for more realistic stochastic models (Itô 1946)...

Why Normal? If  $Z_1$ ,  $Z_2$  and  $Z$  are independent copies of limit in CLT, then

$$\begin{aligned} \sqrt{t_1} \frac{S_{Nt_1}}{\sqrt{Nt_1}} + \sqrt{t_2} \frac{S_{N(t_1+t_2)} - S_{Nt_1}}{\sqrt{Nt_2}} \\ = \sqrt{t_1 + t_2} \frac{S_{N(t_1+t_2)}}{\sqrt{N(t_1 + t_2)}} \end{aligned}$$

$$\sqrt{t_1} Z_1 + \sqrt{t_2} Z_2 \equiv \sqrt{t_1 + t_2} Z \quad (\text{same distribution})$$

Normal distributions (mean 0) are only such distributions.

## 2. Super-Brownian Motion (SBM)

$b > 0$  (branching rate),  $\sigma^2 > 0$  (diffusion rate),  
 $g \in \mathbb{R}$  (growth rate).

$W \in \mathbb{R}^d$  random displacement with mean 0,  
covariance  $\frac{\sigma^2}{b} I_{d \times d}$ .

Start  $O(N)$  particles in  $\mathbb{R}^d$ .

Particles behave independently.

With rate  $bN$  particle at  $x$  dies.

With rate  $bN + g$  particle at  $x$  gives birth to  
particle at  $x + W/\sqrt{N}$ .

Define a finite measure on  $\mathbb{R}^d$ ,  $X_t^N$ , by

$$X_t^N(A) = \frac{1}{N} \#(\text{particles in } A \text{ at time } t).$$

**Theorem 2.1** (S. Watanabe 68). If  $X_0^N \rightarrow X_0$ ,  
 $\lim_{N \rightarrow \infty} P(X^N \in A) = P_{X_0}(X \in A)$  "for any"  
set  $A$  of measure-valued paths.

$X$  is a continuous measure-valued Markov process whose law  $P_{X_0}$  depends only on  $(X_0, b, \sigma^2, g)$ .

Call  $X$  super-Brownian motion.

Set  $g = 0$ . If  $X_t^{N,i}$  describes descendants of  $x_i$  at  $t = 0$ , then  $X_t^N$  decomposes into a sum of  $NX_0^N(\mathbb{R}^d)$  ind't clusters:

$$(*) \quad X_t^N = \sum_{x_i} X_t^{N,i}.$$

Sad fact: Critical branching processes die out.  
 $P(X_t^{N,i} \neq 0) \sim (Ntb)^{-1}$  as  $N \rightarrow \infty$ .

(\*) has  $NX_0^N(\mathbb{R}^d)$  summands each non-zero with prob.  $\sim (Ntb)^{-1}$ . Let  $N \rightarrow \infty$  in (\*):

$$X_t = \sum_{j=1}^{M(t)} X_t^j,$$

where  $(X_t^j)$  are independent clusters descending from a single ancestor at  $t = 0$ , and  $M(t)$  has mean  $\frac{X_0(\mathbb{R}^d)}{tb}$  (Poisson distribution).

**Additive Property:** Run 2 independent SBM's  $X^1, X^2$ , then clearly  $X^1 + X^2$  is a SBM starting at  $X_0^1 + X_0^2$ .

### 3A Properties of Brownian Motion

1. PDE.  $u(t, x) = E_x(\phi(B_t))$  solves  $\frac{\partial u}{\partial t} = \frac{\Delta}{2}u$ ,  $u_0 = \phi$ .

2. Longterm Behaviour.

$d \leq 2$ .  $B_t \in G$  infinitely often as  $t \rightarrow \infty$  for any open set  $G$  a.s. (neighbourhood recurrence).

$d \geq 3$ .  $\lim_{t \rightarrow \infty} \|B(t)\| = \infty$  a.s. (transience).

3. Local Behaviour (Ciesielski-Taylor (1962))

Let

$$\phi_d(r) = \begin{cases} r^2 \log(1/r) \log \log \log(1/r) & d = 2 \\ r^2 \log \log 1/r & d \geq 3. \end{cases}$$

$\phi_d - m(\{B_s : s \leq t\}) = c_d t$  for all  $t \geq 0$  a.s. So the range of Brownian motion is a random set of Hausdorff dimension 2.

### 3B. Properties of SBM

1. PDE.  $E_{X_0}(e^{-\langle X_t, \phi \rangle}) = e^{-\langle X_0, v_t \rangle}$ , where  
 $\frac{\partial v}{\partial t} = \frac{\sigma^2 \Delta v}{2} - bv^2 + gv$ ,  $v_0 = \phi \geq 0$ .  
(Dynkin, Le Gall, Mselati; Iscoe,...)

2. Longterm Behaviour.

$X_0$  finite.  $X$  becomes extinct in finite time almost surely iff  $g \leq 0$  (eg. use PDE).

$g = 0$ ,  $X_0(dx) = m dx$ , Dawson (1977) showed:

(a) For  $d \leq 2$ , for any  $R > 0$ ,  $X_t(|x| \leq R)$  becomes negligible as  $t \rightarrow \infty$ .

(b) For  $d \geq 3$ ,  $\lim_{t \rightarrow \infty} P(X_t \in A) = P(X_\infty \in A)$  where  $E(X_\infty(A)) = m \text{Leb}(A)$  and  $X_\infty$  are (the only extremal) equilibrium distributions for  $X$ .

3. Local Behaviour.

$S(X_t)$  = closed support of  $X_t$ . If  $d \geq 2$

$X_t(A) = C_d \frac{b}{\sigma^2} \phi_d - m(A \cap S(X_t)) \quad \forall A \text{ a.s. } t > 0$ .

(Dawson-Hochberg 79, P. 90, Le Gall-P. 95)

$S(X_t)$  is a Leb. null set of Hausdorff dimension 2 for all  $t > 0$  a.s.

If  $d = 1$ ,  $X_t(dx) = X(t, x)dx$ , where  $X(t, x)$  is the unique solution of a stochastic pde.

## 4. Voter Model

Each site in  $\mathbb{Z}^d$  has type 0 or 1.  $\xi_t(x) = 0$  or 1. With rate 1, the type at  $x$  chooses a neighbour at random and imposes its type on it.

Rescale space and time:

$\xi_t^N(x) = \xi_{tN}(x\sqrt{N})$ ,  $x \in \mathbb{Z}^d/\sqrt{N}$ .  $V_t^N(A) = m_N^{-1} \sum_{x \in A} \xi_t^N(x)$  is empirical distribution of 1's.

### Notation

$$m_N = \begin{cases} N, & d \geq 3 \\ N/\log N & d = 2 \end{cases}$$

$(S_n)$  is a nearest neighbour rw.

$$p_{\text{esc}} = \begin{cases} P_0(S_n \neq 0 \ \forall n \geq 1) & d \geq 3 \\ 2\pi\sigma^2 = \pi & d = 2 \end{cases}$$

**Theorem 4.1** (Cox, Durrett, P. 00) Assume  $d \geq 2$ ,  $V_0^N \rightarrow X_0$ . Then

$\lim_{N \rightarrow \infty} P(V^N \in A) = P_{X_0}(X \in A)$  " $\forall$ "  $A$ , where  $X$  is SBM with  $g = 0$ ,  $\sigma^2 = 1/d$  and  $b = p_{\text{esc}}$ .



**Remark** Changing local dynamics (eg. new definition of "neighbour") leads to a new  $S_n$  and hence a new  $p_{\text{esc}}$ —local dynamics affect limit only through value of  $p_{\text{esc}}$ .

**Proof.**

Reinterpret dynamics:  $\xi_t(x) = 1 \iff$  particle at  $x$   
 $\xi_t(x) = 0 \iff$  no particle at  $x$ .

$$f_0^N(t, x) = \{\text{no. of neighbouring 0's to } x\} / 2d$$

Particle at  $x$  dies with rate  $Nf_0^N(t, x)$ , and with rate  $Nf_0^N(t, x)$  produces child at  $y$  chosen at random from the neighbouring 0 sites.

Similar to branching random walk with  $g = 0$ ,  $\sigma^2 = 1/d$  and a random  $b = f_0^N(t, x)$ .

Proof shows that if  $\xi_t(x) = 1$ ,  $f_0^N(t, x) \sim p_{\text{esc}}$  on average by using a "dual coalescing random walk" to find conditional third moments. Dual RW traces back your history in time.

## An Application

Let  $\xi_t$  be voter model starting from a single 1 at  $x = 0$ .  $S(\xi_t) = \{x : \xi_t(x) = 1\}$ . Take  $d \geq 2$ .

Q. (Bramson, Griffeath 81).

Conditional on  $\xi_t \neq 0$ , what is asymptotic shape of  $S(\xi_t)$  as  $t \rightarrow \infty$ ?

Set  $t = N$ .  $S(\xi_N)/\sqrt{N} = S(V_1^N)$  and so expect  
 $P(S(\xi_N)/\sqrt{N} \in \cdot | \xi_N \neq 0)$   
 $\sim P_{\delta_0/N}(S(X_1) \in \cdot | X_1 \neq 0)$

LHS is the law of a single cluster of SBM.

**Theorem 4.2.** (Bramson, Cox, Le Gall 01)

For  $d \geq 2$  and "for all" sets of sets  $A$

$$\lim_{t \rightarrow \infty} P(S(\xi_t)/\sqrt{t} \in A | \xi_t \neq 0)$$

$= N_0(S(X_1) \in A | X_1 \neq 0)$  (law of a single cluster conditioned on non-extinction).

## 5. Other Limit Theorems

(Slade, Notices AMS '02)

(a) Interacting Particle systems.

(i) Rescaled Contact Process (stochastic model for spread of disease) at, or near criticality, will converge to super-Brownian motion with non-trivial parameters.

Durrett-P. (99), long range,  $d > 1$ .

van der Hofstad-Sakai (04), medium range,  $d > 4$ .

(ii) Near critical rescaled stochastic Lotka Volterra models of Neuhauser and Pacala converge to super-Brownian motion with non-trivial drift and branching rates.  $d > 2$ . (Cox-P. 03).

(b) Lattice Trees. A connected set of neighbouring (range  $L$ ) bonds in  $\mathbb{Z}^d$  containing 0 with no cycles. Pick a lattice tree with  $N^2$  vertices at random.  $I^N$  assigns mass  $N^{-2}$  to each vertex scaled by  $1/\sqrt{N}$ .

(Derbez-Slade 98) (conj. Aldous 93). For  $d > 8$  and  $L > L_0$   $P(I^N \in \cdot)$  approaches  $N_0(\int_0^\infty X_s ds \in \cdot | \int_0^\infty X_s(\mathbb{R}^d) ds = 1)$ , i.e., the law of an integrated cluster conditioned on total mass 1, Integrated Super Excursion (ISE). Here  $X$  is SBM  $(b(L, d), \sigma^2(L, d))$

$\Lambda = \sum_i N X_{i/N}^N(\mathbb{R}^d) = \text{total progeny of br. process}$   
 $ISE = \lim_N P(N^{-2} \sum_i \sum_{x \in S(X_{i/N}^N)} \delta_x | \Lambda = N^2).$

Proof uses lace expansion (Brydges-Spencer, Madras-Slade).  $d > 8$  is needed since  $S(\int_0^\infty X_s ds)$  has dimension 4 and so will be a tree iff  $d \geq 8$ . Get strong non-local interactions for  $d < 8$  and logarithmic corrections if  $d = 8$ .

(c) Critical Percolation in  $\mathbb{Z}^d$ .  $d > 6$ ,  $p = p_c$ . Assign mass  $N^{-2}$  to each vertex in cluster containing 0 conditioned to have size  $N^2$  and rescaled by  $N^{-1/2}$ .

Conj. (Hara, Slade 98):  $X^N$  converges to ISE.

Theorem (Hara, Slade 00):  $L > L_0$ . First and second moments converge.

(d) Fleming-Viot Processes

Modified Moran particle system.

$\mathbb{R}^d$  is space of allele types. Migration is now mutation of type of offspring from that of parent.  $N$  is *fixed* population size. Branching becomes resampling from gene pool: at  $t = i/N$  each particle (j) at  $x$  is replaced by  $k_j$  offspring of at  $x + W_m/\sqrt{N}$ ,  $m = 1, \dots, k_j$ , where  $(k_1, \dots, k_N)$  is multinomial  $(N; 1/N, \dots, 1/N)$ .

Then the empirical measure of types  $F_t^N$  is a random probability on types which converges to  $F_t$ , the Fleming-Viot process.

**Theorem 6.1** (Etheridge-March 91)

$$P(F \in \cdot) = \lim_n P_{F_0}(X \in \cdot \mid \sup_{t \leq n} |X_t(\mathbb{R}^d) - 1| \leq 1/n)$$

Here  $X$  is SBM ( $g = 0$ ,  $\sigma^2$ ,  $b = 1$ ).

**"Proof".** Consider discrete time branching random walk in which at  $t = 1/N$  a particle is replaced by a Poisson (1) number of offspring displaced from its parent by  $W_m/\sqrt{N}$ . If  $X_t^N$  assigns mass  $1/N$  to each particle at time  $t$ , then  $X^N \rightarrow X$ , SBM as before ( $g = 0$ ,  $b = 1$ ,  $\sigma^2$ ).  
 Easy Fact:  $Z_1, \dots, Z_N$  independent Poisson (1),

$$P((Z_1, \dots, Z_N) \in \cdot \mid \sum_i Z_i = N) \\
\sim \text{multinomial}(N : 1/N, \dots, 1/N).$$

This implies

$$P(X^N|_{[0,T]} \in \cdot \mid X_t^N(\mathbb{R}^d) = 1 \forall t \leq T) \\
= P(F^N|_{[0,T]} \in \cdot).$$

Let  $N \rightarrow \infty$  and hope interchange of limits on left side is OK. □

## 7. A Family of Stochastic pde's

SBM  $X$ , " $X_t(dx) = X(t, x)dx$ " where  $X(t, x)$  is the unique solution of the stochastic pde  
(SPDE)  $\frac{\partial X}{\partial t} = \frac{\sigma^2 \Delta X}{2} + gX + \sqrt{2bX}\dot{W}$ ,

Here  $\dot{W}(s, x)dsdx$  are iid Normal with mean 0 and variance  $dsdx$  (space time white noise).

$d = 1$  (Reimers, Konno-Shiga 1988)  
 $X(t, x)$  exists, is continuous and unique.

$d > 1$   $X_t(dx) \perp dx$ . Solutions to (SPDE) exist and are unique when interpreted in a generalized sense.

Why  $\sqrt{\cdot}$ ? By Additive Property we need  
 $\sqrt{2bX^1}\dot{W}_1 + \sqrt{2bX^2}\dot{W}_2 \equiv \sqrt{2(b(X^1 + X^2))}\dot{W}$   
which is true since  $\sqrt{c_1}Z_1 + \sqrt{c_2}Z_2 \equiv \sqrt{c_1 + c_2}Z$   
for independent normals.

In previous limit theorems, local interactions and LLN led to constant (non-obvious) parameters in limit. For many models truly interactive models arise:

- attraction/repulsion of particles
- competing species/predator prey models
- symbiotic/diploid branching

$$(\text{Itô '46}) \quad dY_t = d(Y_t)dt + \sigma(Y_t)dB_t$$

( Dawson '87) Use of SBM as a building block for interactive population models.

(ISPDE)

$$\begin{aligned} \frac{\partial X}{\partial t} &= (A_{X_t}^* + g(x, X_t))X_t(x) + \sqrt{2b(x, X_t)X_t(x)}\dot{W} \\ A_{X_t}\phi(x) &= \frac{1}{2} \sum_{1 \leq i, j \leq d} a_{ij}(x, X_t)\phi_{ij} + \sum_{i=1}^d d_i(x, X_t)\phi_i \\ a_{ij} &= (\sigma\sigma^*)_{ij}. \end{aligned}$$

Question: Does (ISPDE) characterize a unique measure-valued process? Existence via weak limits of natural branching particle systems. Uniqueness open—infinite dimensional, non-Lipschitz degenerate coefficients, generalized solutions (with nonlinearity).



Dawson '78: Solutions with  $g$  are absolutely continuous wrt solutions with  $g = 0$ . Set  $g = 0$ .

P. '95, Donnelly Kurtz '99: Strong equation, genealogical information, exchangeable particles systems and historical processes to construct and characterize solutions with  $b$  constant and  $\sigma, d$  Lipschitz. Gave tools for further study: stochastic calc., lookdown system.

$$\begin{aligned} \text{(SE)} \quad Y_t(\omega, y) &= y_0 + \int_0^t \sigma(Y_s, X_s) dy_s + \int_0^t d(Y_s, X_s) ds, \\ X_t(\cdot) &= \int 1(Y_t(\omega, y) \in \cdot) H_t(dy). \end{aligned}$$

$\omega$  selects random tree;  $y$  selects branch on tree.

**Theorem 7.1.**  $\sigma, b$  Lipschitz,  $b$  constant,  $g = 0$ . (a)  $\exists$  a unique solution to (SE) which is a function of  $H$  and is strong Markov.

(b)  $X$  satisfies (ISPDE) iff  $X$  satisfies (SE) and so (ISPDE) has a unique solution.

(ISPDE)  $\frac{\partial X}{\partial t} = A_{X_t}^* X_t(x) + \sqrt{2b(x, X_t)} X_t(x) \dot{W}$   
 Uniqueness for  $b(x, X_t)$  is still open.

Finite dimensional problem: Replace  $x \in \mathbb{R}^d$  by  $i \in \{1, \dots, k\}$  and  $X_t \in M_f(\mathbb{R}^d)$  by  $X_t \in \mathbb{R}_+^k$ .

$A_{X_t} \phi(i) = \sum_{j=1}^k q_{ij}(X_t) \phi(j)$ ;  $i \rightarrow j$  at rate  $q_{ij}(X_t)$ .

(ISPDE) becomes:

(SDE)  $dX_t^i = \sum_{j=1}^k X_t^j q_{ji}(X_t) dt + \sqrt{b_i(X_t)} X_t^i dB_t^i$ ,  
 $B^i$  independent Brownian motions,  $i = 1, \dots, k$ .

More generally consider:

(GSDE)  $dX_t^i = d_i(X_t) dt + \sqrt{b_i(X_t)} X_t^i dB_t^i$ .

Assume: (i)  $b_i > 0$ , continuous.

(ii)  $d_i \geq 0$  on  $\{x_i = 0\}$ , continuous.

Problem: Degenerate diffusion so Stroock-Varadhan (1969) won't apply. Non-Lipschitz diffusion so Itô (1951) won't apply.

**Counter-example:** Uniqueness may fail in (GSDE).

$$X_t = \int_0^t \sqrt{2X_s} dB_s + \int_0^t C(1 + \log^+(\frac{1}{X_s}))^{-1} ds$$

If  $C > 1$ , 0 is a regular boundary point and so there are solutions s.t.  $\int_0^\infty 1(X_s = 0) ds = 0$  and  $X \equiv 0$ .

**Thm. 7.2.(a)** (Athreya, Barlow, Bass, P. 02)  
 Assume (\*)  $d_i(x) > 0$  on  $\{x_i = 0\}$ .  
 There is a unique (in law) solution to (GSDE).

(b) (Bass, P. 03) If  $d_i, b_i$  are locally Hölder continuous then there is a unique solution to (GSDE).

**Remark.** (\*) will fail for many examples of interest such as nearest neighbour random walk migration,  $q_{ij} = 1(|i - j| = 1)$  for  $i \neq j$ . Hence interest in (b).

Proof. Stroock-Varadhan perturbation.  
 $\mathcal{L}^0 \phi = \sum_i b_i^0 x_i \phi_{ii} + d_i^0 \phi_i$ ,  $R_\lambda^0 f = (\lambda - \mathcal{L}^0)^{-1} f$ .  
 Must show  $\frac{\partial^2 R_\lambda^0 f}{\partial x_i \partial x_j}$  is bounded on an Banach space  $B$ . For (a)  $B = L^2(\mu)$ . For (b)  $B$  is weighted Hölder space.

**Remark.** Constants in bound for (b) do not depend on  $k$  and so results extend to infinite dimensional sde's. Some hope for (ISPDE)?