

HOW DOES SPATIAL STRUCTURE INTERACT WITH NATURAL SELECTION?

Alison Etheridge

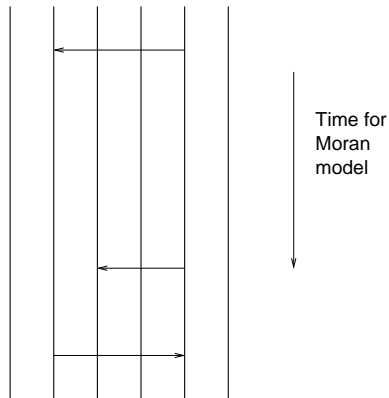
Nick Barton, IST; Nic Freeman, Sheffield; Jerome Kelleher, Oxford; Sarah Penington, Oxford; Daniel Straulino, Oxford; Amandine Véber, École Polytechnique; Feng Yu, Bristol

University of Oxford

The logo for the Engineering and Physical Sciences Research Council (EPSRC). It features the acronym "EPSRC" in a bold, dark red serif font. The letters are framed by two horizontal teal lines, one above and one below.

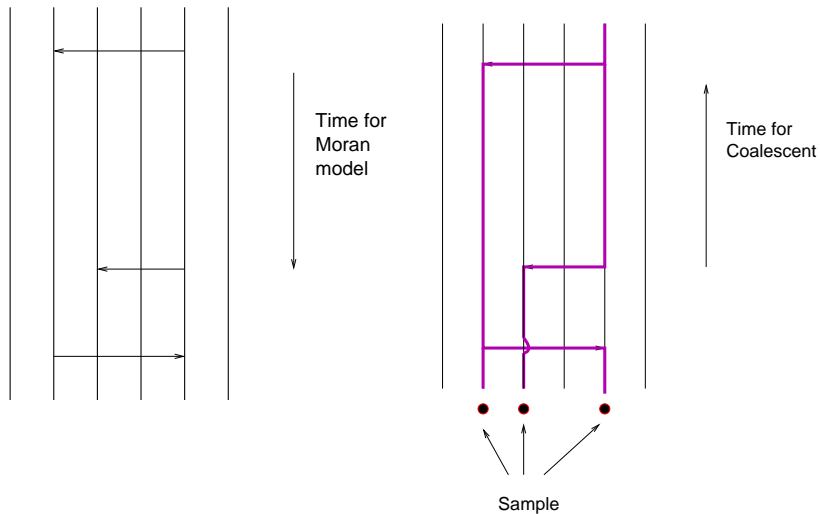
Engineering and Physical Sciences
Research Council

Genetic drift

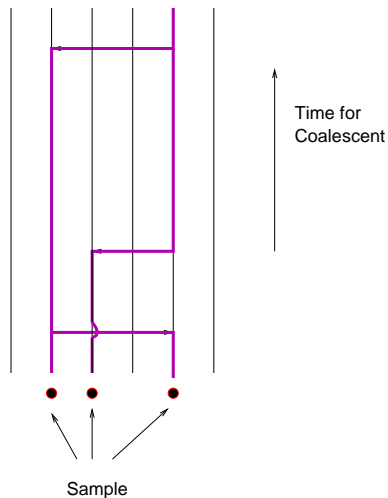


- ▶ Events determined by Poisson Process intensity $\binom{N}{2} dt$;
- ▶ Parent chosen at random;
- ▶ Individual chosen at random to die

Genetic drift



Genetic drift

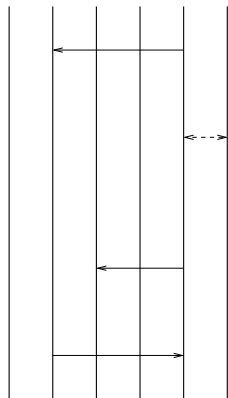


Two types, a , A .

$$dp_t = \sqrt{p_t(1-p_t)}dW_t,$$

$$\mathbb{E}[p(t)^{n(0)}] = \mathbb{E}[p(0)^{n(t)}].$$

Adding selection



Time for
Moran
model

Relative fitnesses:

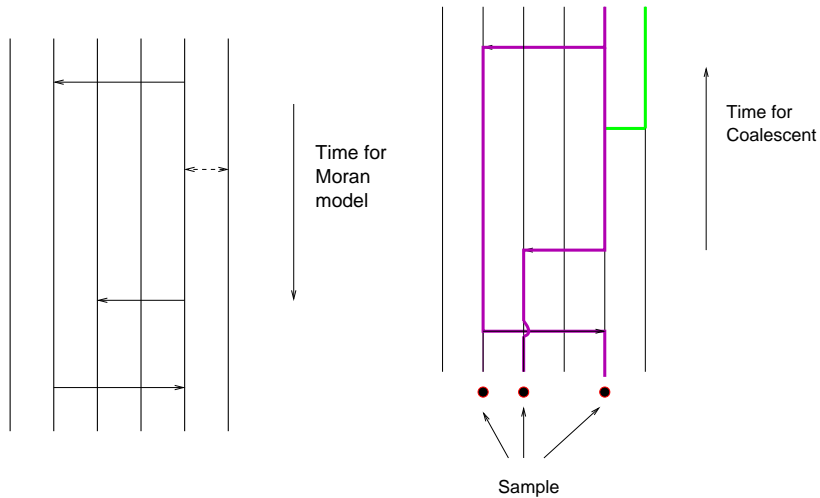
$$\frac{a}{1 - s_N} \mid \frac{A}{1}$$

- ▶ Events fall at rate $\binom{N}{2}$;
- ▶ Parent chosen in *weighted* way;
- ▶ Individual chosen at random to die

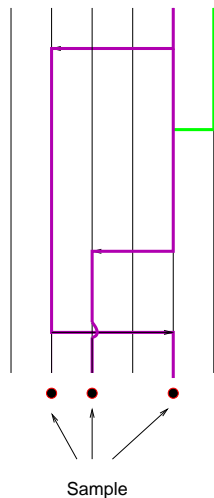
Probability parent of type a :

$$\frac{(1 - s_N)p}{1 - s_N p} = (1 - s_N)p + s_N p^2 + \mathcal{O}(s_N^2)$$

Adding selection



Adding selection



Time for
Coalescent

Neutral events at rate $(1 - s_N) \binom{N}{2}$;

Selective events at rate $s_N \binom{N}{2}$:
if aA chosen, A reproduces.

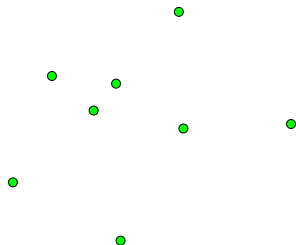
$$Ns_N \rightarrow \sigma,$$

$$dp_t = -\sigma p_t(1 - p_t)dt + \sqrt{p_t(1 - p_t)}dW_t$$

$$\mathbb{E}[p(t)^{n(0)}] = \mathbb{E}[p(0)^{n(t)}].$$

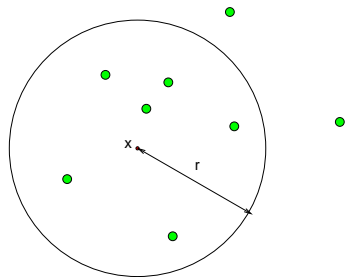
Introducing Space: An individual based model

- ▶ Start with Poisson point process intensity λdx ;



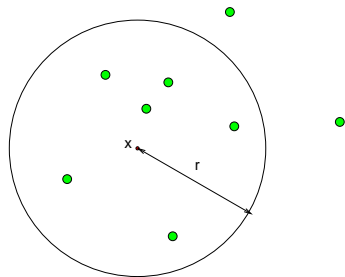
Introducing Space: An individual based model

- ▶ Start with Poisson point process intensity λdx ;
- ▶ At rate $\mu(dr) \otimes dx \otimes dt$ throw down ball centre x , radius r ;



Introducing Space: An individual based model

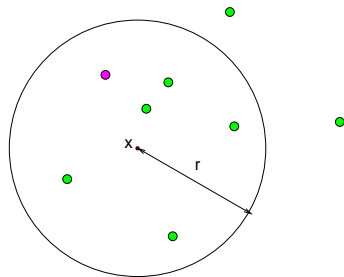
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If ball empty, do nothing

Introducing Space: An individual based model

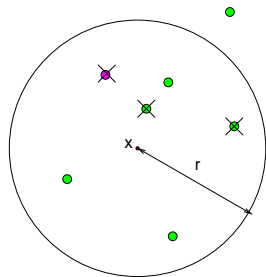
- ▶ Start with Poisson point process intensity λdx ;
- ▶ At rate $\mu(dr) \otimes dx \otimes dt$ throw down ball centre x , radius r ;
- ▶ Choose parent uniformly at random from individuals in $B(x, r)$;



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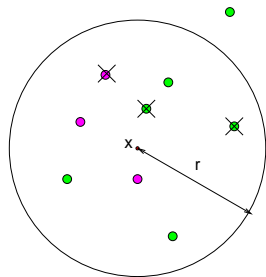
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- ▶ Choose parent uniformly at random from individuals in $B(x, r)$;
- ▶ Each individual in region dies with probability $u \sim \nu_r(du)$;



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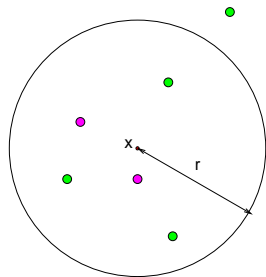
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- ▶ New individuals born according to a Poisson $\lambda u \mathbf{1}_{B(x,r)} dy$.



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If ball empty, do nothing

$\lambda \rightarrow \infty$ limit (no space)

Distribution of types.

State $\{\rho(t, \cdot) \in \mathcal{M}_1(K), t \geq 0\}$.

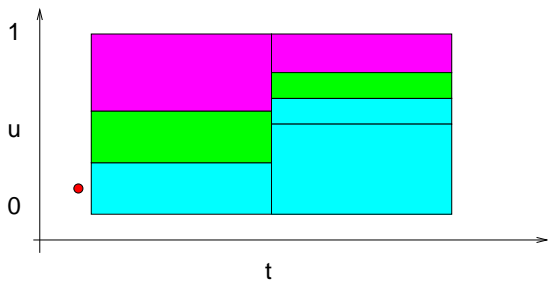
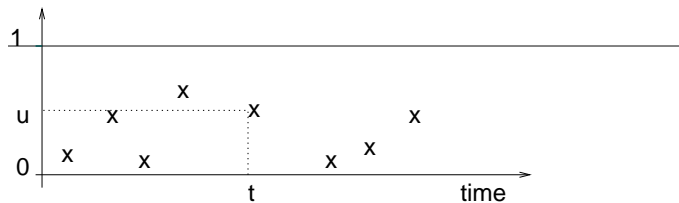
- ▶ Poisson point process intensity $dt \otimes F(du)$
- ▶ individual sampled at random from population
- ▶ proportion u of population replaced by offspring of chosen individual

$$\rho(t, \cdot) = (1 - u)\rho(t-, \cdot) + u\delta_k.$$

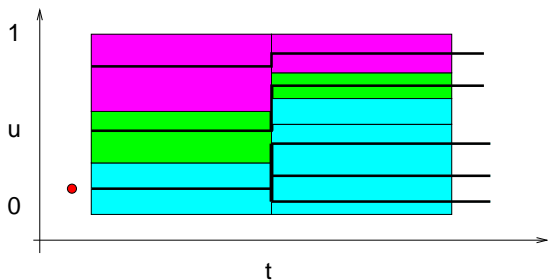
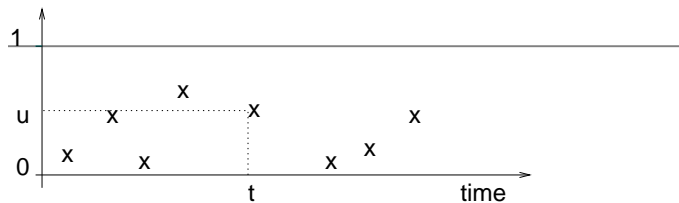
Generalised Fleming-Viot process Bertoin & Le Gall (2003)

$F(du) = \frac{\Lambda(du)}{u^2}$, Λ finite measure on $[0, 1]$.

The Λ -Fleming-Viot process



The Λ -Fleming-Viot process



Λ -coalescent dual

Donnelly & Kurtz (1999), Pitman (1999), Sagitov (1999)

If there are currently n ancestral lineages, each transition involving j of them merging happens at rate

$$\beta_{n,j} = \int_0^1 u^j (1-u)^{n-j} \frac{\Lambda(du)}{u^2}$$

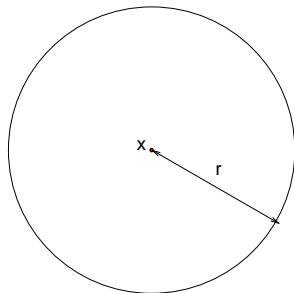
- ▶ Λ a finite measure on $[0, 1]$
- ▶ Kingman's coalescent, $\Lambda = \delta_0$

The spatial Λ -Fleming-Viot process

State $\{\rho(t, x, \cdot) \in \mathcal{M}_1(K), x \in \mathbb{R}^d, t \geq 0\}$.

Π Poisson point process rate $dt \otimes dx \otimes \xi(dr, du)$ on $[0, \infty) \times \mathbb{R}^d \times [0, \infty) \times [0, 1]$.

Dynamics: for each $(t, x, r, u) \in \Pi$,



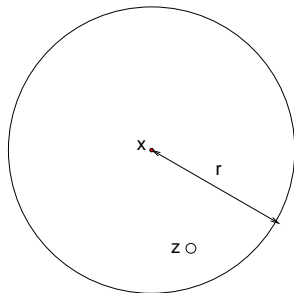
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- ▶ $z \sim U(B_r(x))$



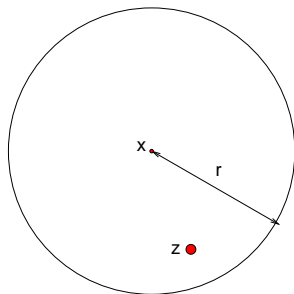
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- ▶ $z \sim U(B_r(x))$
- ▶ $k \sim \rho(t-, z, \cdot)$.



The spatial Λ -Fleming-Viot process

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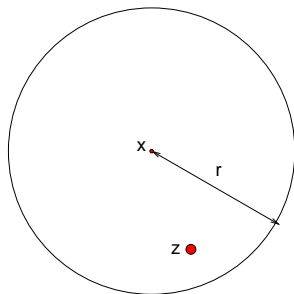
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- ▶ $z \sim U(B_r(x))$
- ▶ $k \sim \rho(t-, z, \cdot)$.

For all $y \in B_r(x)$,

$$\rho(t, y, \cdot) = (1 - u)\rho(t-, y, \cdot) + u\delta_k.$$

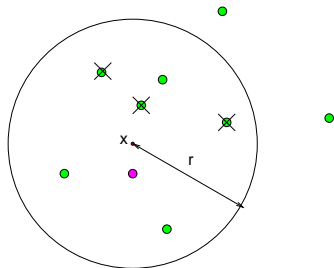


Backwards in time

- ▶ A *single* ancestral lineage evolves in series of jumps with intensity

$$dt \otimes \int_{(|x|/2, \infty)} \int_{[0,1]} \frac{L_r(x)}{\pi r^2} u \xi(dr, du) dx$$

on $\mathbb{R}_+ \times \mathbb{R}^2$ where $L_r(x) = |B_r(0) \cap B_r(x)|$.



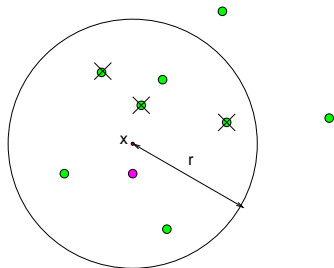
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- ▶ Lineages can coalesce when hit by same 'event'.



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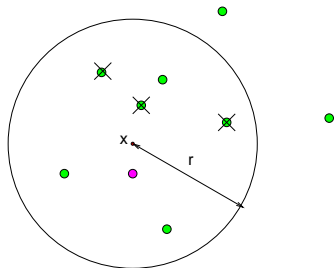
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- ▶ Lineages can coalesce when hit by same 'event'.

Note: If $\xi(dr, du) = \mu(dr) \otimes \delta_u$, rate of jumps $\propto u$.



Introducing selection to the SLFV

Warning: There are **lots** of different ways to do this.

- ▶ Easy to cook up examples where allele frequencies the same, but genealogies different

... **but that's another talk**

Introducing selection to the SLFV

Here we mimic what we did for the Moran model

Introducing selection to the SLFV

$K = \{a, A\}$, $w(t, x) = \rho(t, x, a)$ proportion of type a

- ▶ (i) Two types, a, A . Weight type a by $(1 - s)$. If a reproduction event affects a region $B(x, r)$ in which current proportion of a -alleles is \bar{w} , then probability offspring are type a is

$$\frac{(1 - s)\bar{w}}{1 - s\bar{w}}$$

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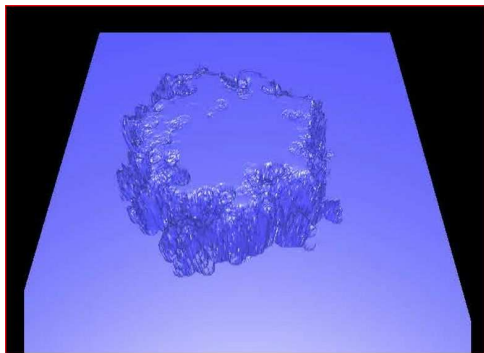
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$$\frac{(1 - s)\bar{w}}{1 - s\bar{w}} = \bar{w}(1 - s) + s\bar{w}^2 + \mathcal{O}(s^2).$$

- ▶ (ii) Neutral events rate $\propto (1 - s)$, selective events rate $\propto s$. At selective reproduction events, sample **two** potential parents. If types aa , then an a reproduces, otherwise an A does.

Spread of a favoured allele

- ▶ Two types, a , A . If a reproduction event affects a region $B(x, r)$ in which current proportion of a -alleles is \bar{w} , then probability offspring are type a is $\frac{(1-s)\bar{w}}{1-s\bar{w}}$.



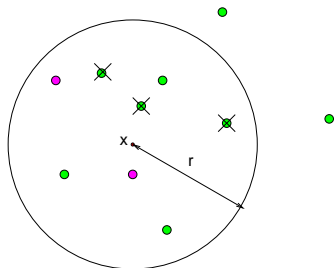
(Spatial) Ancestral selection graph

Evolution of ancestry due to neutral events as before:

- ▶ lineages evolve in a series of jumps;
- ▶ they can coalesce when covered by same event.

At *selective* events

- ▶ *Two 'potential' parents must be traced;*
- ▶ Lineages can coalesce when hit by same 'event'.



A sampled individual is type a iff all lineages in the corresponding ASG are type a at any previous time.

Why rescale?

Neutral mutation rate, μ , sets timescale

- ▶ Mutation rates are low;
- ▶ Scaling limits are 'robust'.

Natural question:

Over what spatial scales can we expect to observe the signature of natural selection?

Scaling limits I: Small neighbourhood size:

Fix $u \in (0, 1)$.

Consider only 'small' events.

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Heuristics:

- ▶ At a 'branching' event in ASG, two lineages born at separation $\mathcal{O}(1/\sqrt{n})$.
- ▶ Probability they separate to $\mathcal{O}(1)$ before coalescing is
 - ▶ $d = 1$: $\mathcal{O}(1/\sqrt{n})$,
 - ▶ $d = 2$: $\mathcal{O}(1/\log n)$,
 - ▶ $d \geq 3$: $\mathcal{O}(1)$.
- ▶ Selection will only be visible if expect to see at least one pair 'separate' by time 1.

Scaling limits I: Small neighbourhood size:

Fix $u \in (0, 1)$.

Consider only 'small' events.

Set $n = 1/\mu$ and rescale: $w(nt, \sqrt{nx})$.

Ability to detect selection depends on dimension:

- ▶ $d = 1$, selection only visible if $s = \mathcal{O}(1/\sqrt{n})$,
limiting ASG embedded in Brownian net;
- ▶ $d = 2$, selection only visible if $s = \mathcal{O}(\log n/n)$,
limiting ASG 'Branching BM';
- ▶ $d \geq 3$, selection only visible if $s = \mathcal{O}(1/n)$,
limiting ASG Branching BM.

Technical challenges because $ns_n \rightarrow \infty$.

Straulino (2015), E., Freeman, Straulino (2015)

Scaling limits II: High neighbourhood size (small events)

Set $u_n = u/n^\gamma$, $s_n = s/n^\delta$, $w^{(n)}(t, x) = w(nt, n^\beta x)$,

Jump rate nu_n , jump size $1/n^\beta$. Diffusive scaling: $2\beta = 1 - \gamma$

- ▶ At 'branching' event, two lineages at separation $\mathcal{O}(1/n^\beta)$.
- ▶ Probability separate to $\mathcal{O}(1)$ is $\mathcal{O}(1/n^\beta)$, ($d = 1$);
 $\mathcal{O}(1/\log n)$, ($d = 2$); $\mathcal{O}(1)$. ($d \geq 3$).
- ▶ If two lineages hit by same event, given one jumps, they coalesce with probability $\mathcal{O}(1/n^\gamma)$.

$d \geq 2$: Probability 'long' excursion before coalesce $\mathcal{O}(1)$;

$d = 1$: Number attempts to reach separation $\mathcal{O}(1)$

\sim number of attempts to coalesce: $\beta = \gamma$;

Selection events rate $nu_n s_n \mathcal{O}(1)$: $1 - \gamma - \delta = 0$.

$\rightsquigarrow \beta = \gamma = 1/3, \delta = 2/3$.

Scaling limits II: High neighbourhood size (small events)

- ▶ Choose $\xi(dr, du) = \mu(dr) \otimes \delta_u$ with $\mu(dr) = \delta_r$
- ▶ Set $u_n = u/n^{1/3}$, $s_n = s/n^{2/3}$, $w^{(n)}(t, x) = w(nt, n^{1/3}x)$,

$$dw = \frac{1}{2}\Delta w dt + sw(1-w)dt + \mathbf{1}_{d=1}\epsilon\sqrt{w(1-w)}W(dt, dx)$$

E. Véber, Yu.

Hybrid zones

Hybrid zones develop when have selection against heterozygosity.

Two spatial dimensions;

only small events

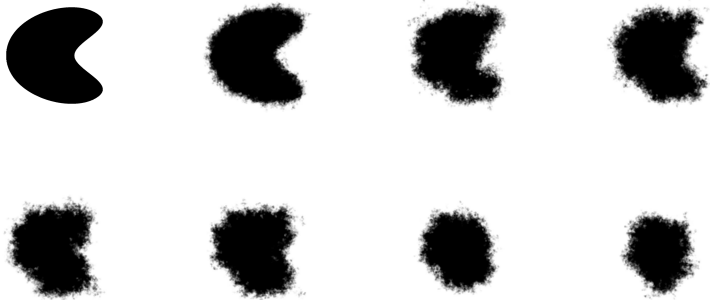
During selective events:

- ▶ sample *three* individuals;
- ▶ the type of the parent is chosen by 'majority voting'.

Scaling limits III



Scaling limits III



Under suitable rescaling, interface becomes sharp and evolves (deterministically) according to motion by mean curvature.

(E. Freeman, Penington)