



# Evolution in a spatial continuum

*Drift, draft and structure*

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Joint work with Nick Barton (Edinburgh) and Tom Kurtz (Wisconsin)

# Kingman's Coalescent

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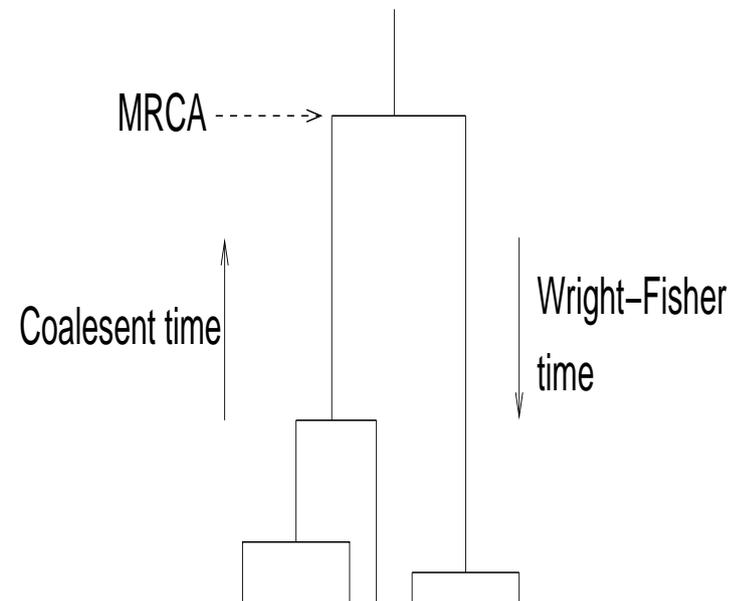
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Coalescence rate  $\binom{k}{2}$



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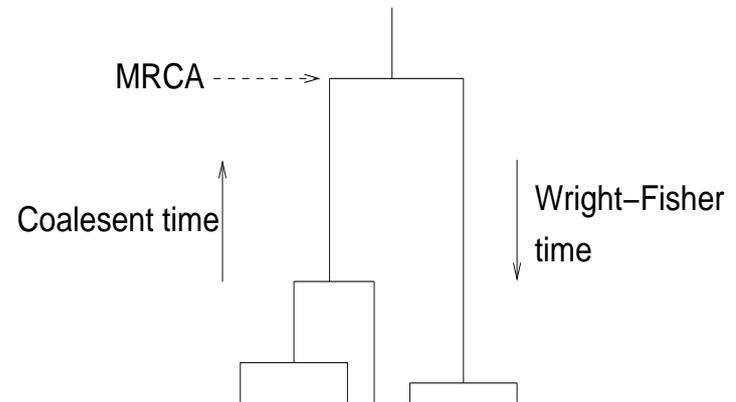
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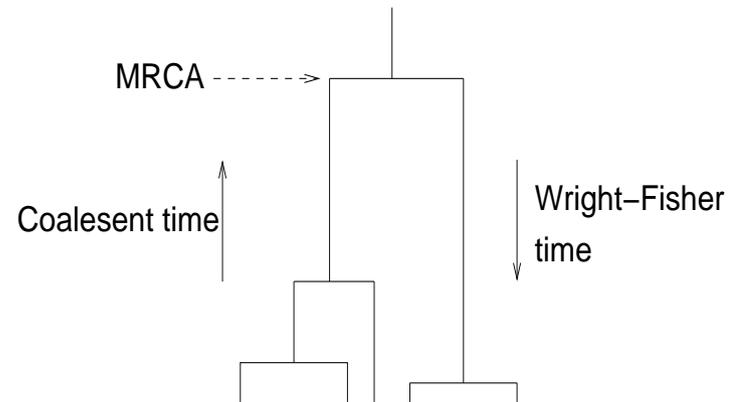
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$$dp_\tau = \sqrt{\frac{1}{N}p_\tau(1 - p_\tau)}dW_\tau,$$

Backwards in time



Coalescence rate  $\frac{1}{N} \binom{k}{2}$

# Basic observation

Genetic diversity is orders of magnitude lower than expected from census numbers and genetic drift.

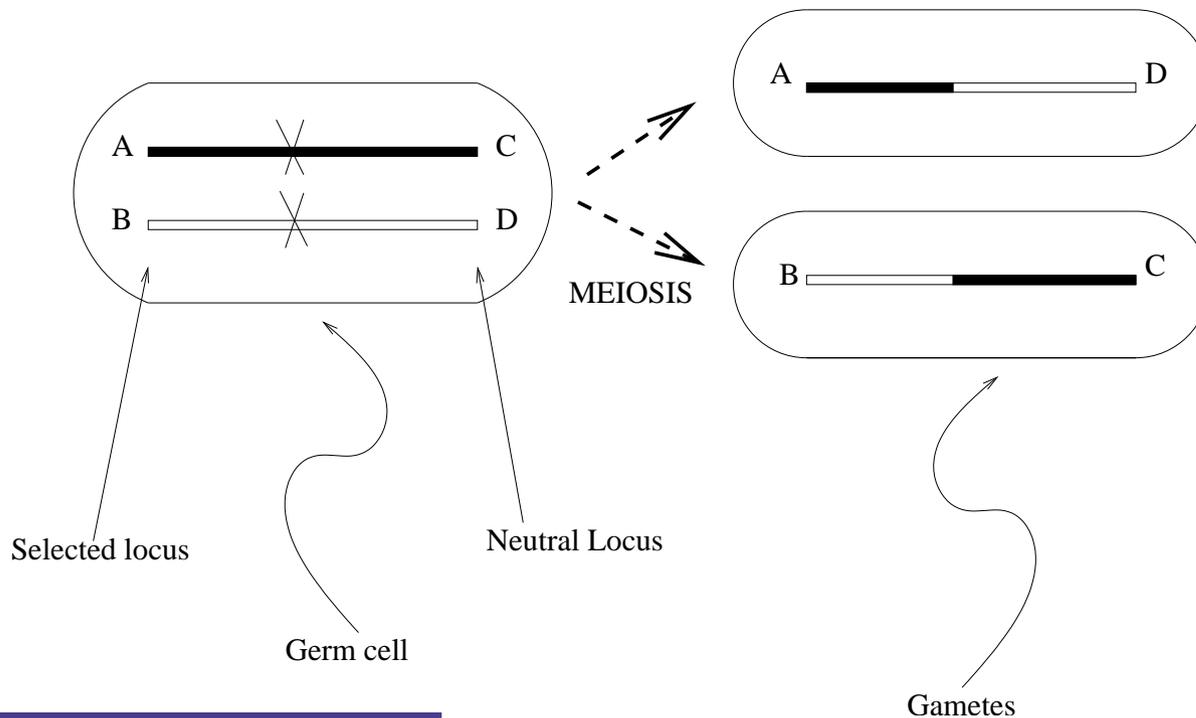
Something else is going on...

# Genetic hitchhiking

When a selectively advantageous allele arises, it is either lost or it *sweeps* to fixation  
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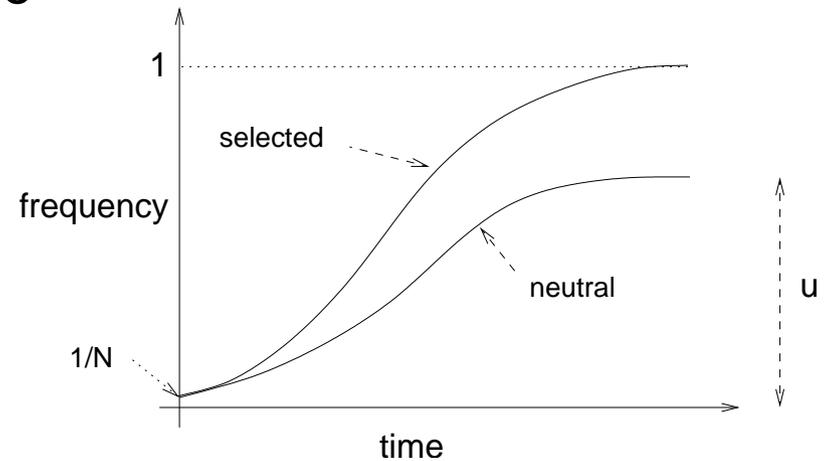
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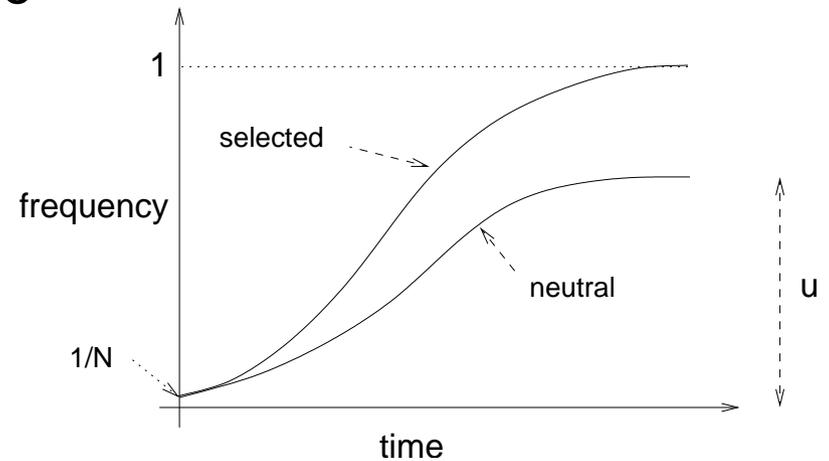


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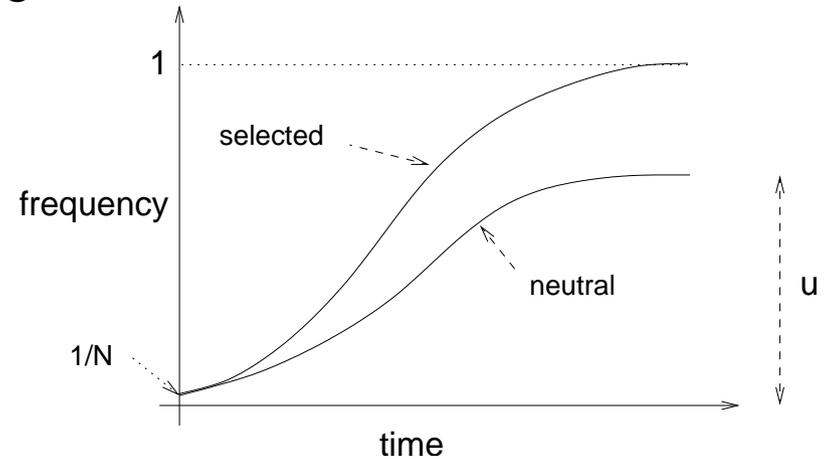


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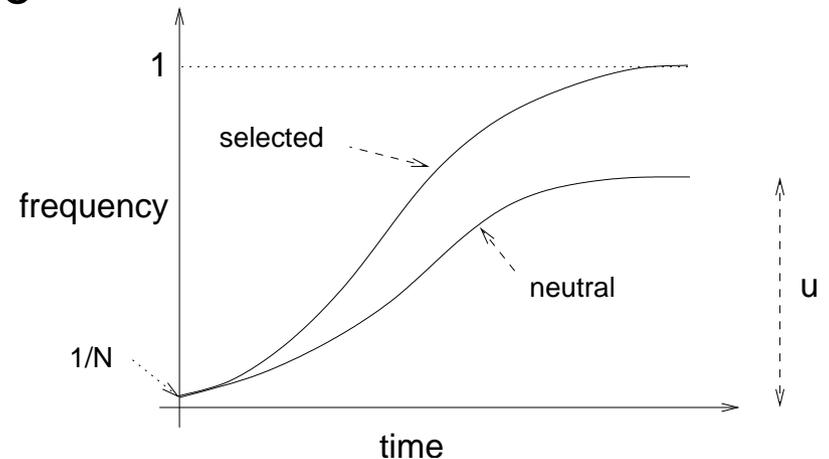
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$$\mathbb{E}[\Delta p] = 0, \quad \mathbb{E}[(\Delta p)^2] = p(1 - p)\mathbb{E}[u^2]$$

$$\mathbb{E}[(\Delta p)^3] = O(1) \implies \text{multiple coalescences}$$

# $\Lambda$ -coalescents

Pitman (1999), Sagitov (1999)

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

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

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

# Forwards in time

Bertoin & Le Gall (2003)

The  $\Lambda$ -coalescent describes the genealogy of a sample from a population evolving according to a  $\Lambda$ -Fleming-Viot process.

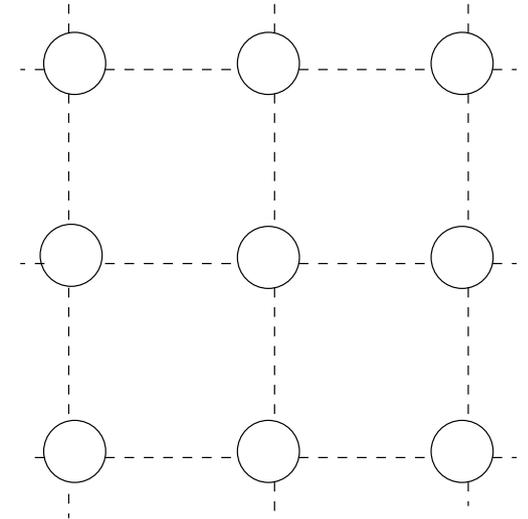
- Poisson point process intensity  $dt \otimes u^{-2} \Lambda(du)$
- individual sampled at random from population
- proportion  $u$  of population replaced by offspring of chosen individual

# Spatial structure

Kimura's stepping stone model

$$dp_i = \sum_j m_{ij} (p_j - p_i) dt + \sqrt{\frac{1}{N_e} p_i (1 - p_i)} dW_i$$

System of interacting Wright-Fisher dif-  
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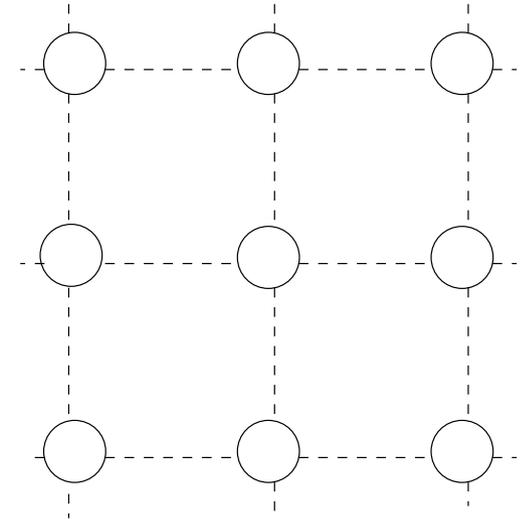
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Genealogy described by system of coalescing  
random walks



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- In two dimensions the equation has *no* solution
- Diffusive rescaling leads to the heat equation
- But anyway local populations are *finite*

# Another basic observation

Real populations experience large scale fluctuations in which the movement and reproductive success of many individuals are correlated

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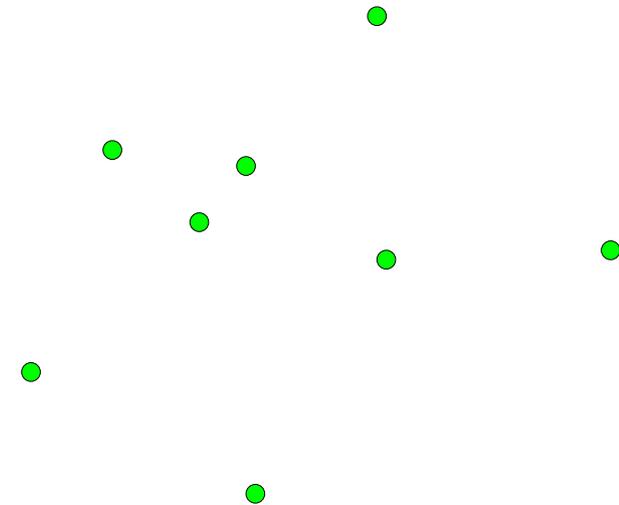
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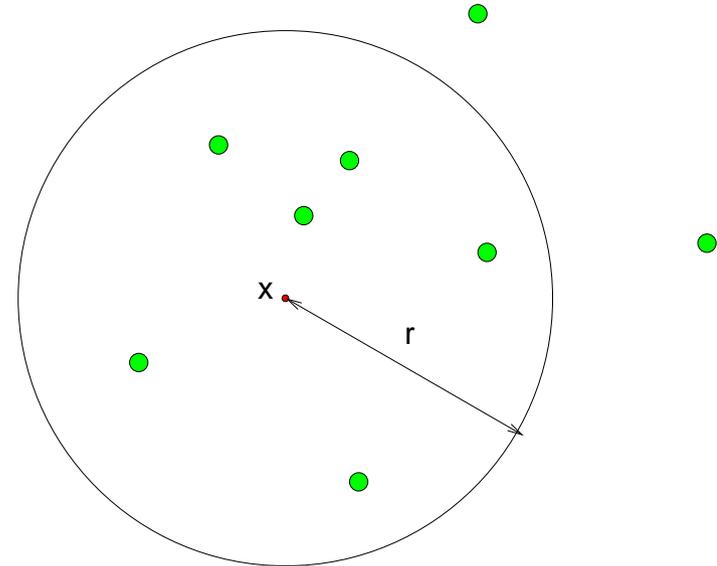
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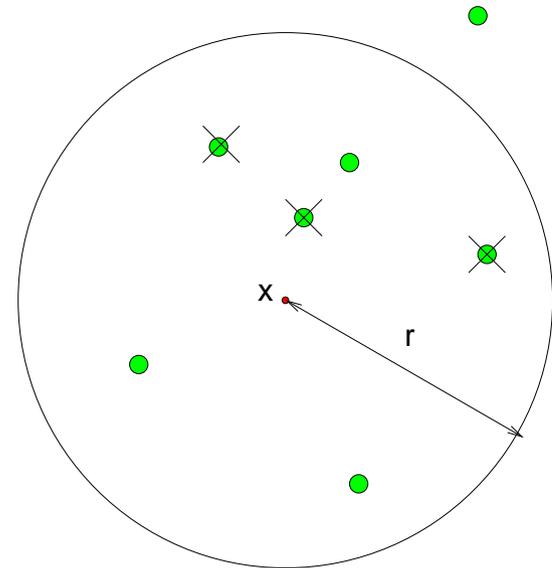
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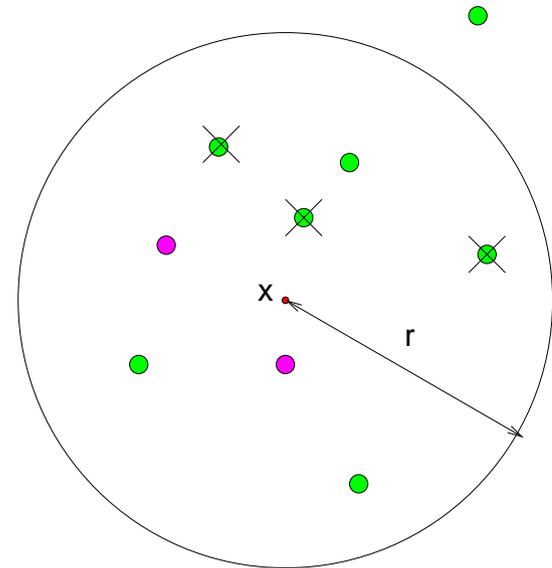
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- Rescale space and time to investigate large scale effects

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

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

# Conditions (1)

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

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

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

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Two lineages currently at separation  $y \in \mathbb{R}^2$  coalesce at instantaneous rate

$$\int_{(|y|/2, \infty)} L_r(y) \left( \int_{[0,1]} u^2 \nu_r(du) \right) \mu(dr).$$

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