# Some mathematical models from population genetics 1: Some classical models 

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## Kingman's Coalescent

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Neutral (haploid) population of constant size $N$
Wright-Fisher model: new generation determined by multinomial sampling with equal weights

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

## Some simple extensions

## Variable population size $N \rho_{t}$.

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Genetic structure:

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## Genetic structure:

e.g. 2 populations of sizes $N \rho_{1}, N \rho_{2}$ with migration between. Add mutation step to Wright-Fisher: after reproduction a (small) fixed proportion $\bar{\mu}_{i}$ of individuals migrates from population $i$ to population $j$. $\bar{\mu}_{1} \rho_{1}=\bar{\mu}_{2} \rho_{2}$.

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The structured coalescent: within populations coalescence at rate $\frac{1}{\rho_{i}}\binom{n_{i}}{2}$. Each lineage migrates $1 \mapsto 2$ at rate $\mu_{2} \frac{\rho_{2}}{\rho_{1}}$ and $2 \mapsto 1$ at rate $\mu_{1} \frac{\rho_{1}}{\rho_{2}}$.

## The Moran model

## The neutral Wright-Fisher model:

A population of $N$ genes evolves in discrete generations. Generation
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- In the Moran model each individual has zero or two offspring.
- The Moran model is already in ‘diffusion’ timescale.


## Graphical representation



For each pair of indices $(i, j)$ Poiss(1) process of arrows pointing left or right with equal probability.

## Graphical representation



## Graphical representation



Genealogy given by Kingman's coalescent (independent of $N$ ).

## The infinite population limit

Suppose that the gene in question has two alleles, $a, A$.
Write $p_{t}$ for the proportion of $a$-alleles at time $t$.

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For a 'nice' function $f$ on $[0,1]$, the infinitesimal generator of $p$ is

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\begin{aligned}
\mathcal{L} f(p) \equiv & \left.\frac{d}{d t} \mathbb{E}\left[f\left(p_{t}\right) \mid p_{0}=p\right]\right|_{t=0} \\
= & \binom{N}{2} p(1-p)\left(f\left(p+\frac{1}{N}\right)-f(p)\right) \\
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To see what happens as $N \rightarrow \infty$, perform a Taylor expansion ...

## . . a Taylor expansion

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& +\binom{N}{2} p(1-p)\left(f\left(p-\frac{1}{N}\right)-f(p)\right) \\
= & \binom{N}{2} p(1-p)\left(f(p)+\frac{1}{N} f^{\prime}(p)+\frac{1}{2 N^{2}} f^{\prime \prime}(p)+\mathcal{O}\left(\frac{1}{N^{3}}\right)-f(p)\right. \\
& \left.+f(p)-\frac{1}{N} f^{\prime}(p)+\frac{1}{2 N^{2}} f^{\prime \prime}(p)+\mathcal{O}\left(\frac{1}{N^{3}}\right)-f(p)\right) \\
= & \frac{1}{2} p(1-p) f^{\prime \prime}(p)+\mathcal{O}\left(\frac{1}{N}\right) .
\end{aligned}
$$

## The diffusion limit

It is reasonable to guess then that for the infinite population limit,

$$
\left.\frac{d}{d t} \mathbb{E}\left[f\left(p_{t}\right) \mid p_{0}=p\right]\right|_{t=0}=\frac{1}{2} p(1-p) f^{\prime \prime}(p) .
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d p_{t}=\sqrt{p_{t}\left(1-p_{t}\right)} d W_{t},
\end{gathered}
$$

where $W_{t}$ is Brownian motion.

## Differential reproductive success

## Wright-Fisher model:

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## Diffusion limit:

For $N \sigma \rightarrow s$, let $N \rightarrow \infty$ (and in WF model measure time in units of size $N$ ) to obtain

$$
d p_{t}=s p_{t}\left(1-p_{t}\right) d t+\sqrt{p_{t}\left(1-p_{t}\right)} d W_{t} .
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\mathbb{P}\left[\left(Z_{1}, \ldots Z_{N}\right)=\left(m_{1}, \ldots, m_{N}\right) \mid \sum_{i=1}^{N} Z_{i}=N\right]=\frac{N!}{m_{1}!\cdots m_{N}!} \frac{1}{N^{N}}
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... just as for the Wright-Fisher model.

## Feller's diffusion approximation

Offspring $\sim \operatorname{Poiss}\left(1+\frac{a}{N}\right)$, generation times $\frac{k}{N}$ and $X_{t}^{(N)}=\frac{1}{N} Z_{t}^{(N)}$.

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\begin{aligned}
\left.\frac{d}{d t} \mathbb{E}_{x}\left[s^{X_{t}}\right]\right|_{t=0} & \approx N\left\{\mathbb{E}_{x}\left[s^{X_{1 / N}^{(N)}}\right]-s^{x}\right\} \\
& =N\left\{e^{N x\left(1+\frac{a}{N}\right)\left(s^{1 / N}-1\right)}-s^{x}\right\} \\
& \approx \frac{1}{2} x(\log s)^{2} s^{x}+a x(\log s) s^{x} \\
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& =\frac{1}{2} x \frac{d^{2} f}{d x^{2}}+a x \frac{d f}{d x} \\
d X_{t} & =a X_{t} d t+\sqrt{X_{t}} d B_{t}
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## Spatially structured populations

Population (with two alleles) subdivided into demes, labelled by $i \in I$.

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d X_{i}(t)=a_{1} X_{i}(t) d t+\sqrt{\sigma X_{i}(t)} d B_{i}(t)+\sum_{j} m_{i j}\left(X_{j}(t)-X_{i}(t)\right) d t
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d Y_{i}(t)=a_{2} Y_{i}(t) d t+\sqrt{\sigma Y_{i}(t)} d \tilde{B}_{i}(t)+\sum_{j} m_{i j}\left(Y_{j}(t)-Y_{i}(t)\right) d t
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& N_{i}(t)=X_{i}(t)+Y_{i}(t)
\end{aligned}
$$

$$
p_{i}(t)=\frac{X_{i}(t)}{X_{i}(t)+Y_{i}(t)}
$$

## The proportion of type $a$

$$
\begin{aligned}
d p_{i}(t)=\left(a_{1}-a_{2}\right) p_{i}(t)\left(1-p_{i}(t)\right) d t & +\sum_{j} \frac{N_{j}}{N_{i}} m_{i j}\left(p_{j}(t)-p_{i}(t)\right) d t \\
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Condition on $N_{i} \equiv$ constant, to arrive at
The stepping stone model

$$
\begin{aligned}
d p_{i}(t) & =s p_{i}(t)\left(1-p_{i}(t)\right) d t+\sum_{j} m_{i j}\left(p_{j}(t)-p_{i}(t)\right) d t \\
s & =\left(a_{1}-a_{2}\right), \quad \gamma=\frac{\sigma}{N} \quad+\sqrt{\gamma p_{i}(t)\left(1-p_{i}(t)\right)} d W_{i}(t)
\end{aligned}
$$

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\begin{equation*}
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$n_{i} \in \mathbb{Z}_{+}, \underline{n}=\left(n_{i}\right)_{i \in I}, \underline{e}_{j}=\left(\delta_{i j}\right)_{i \in I}, \underline{p^{n}}=\prod_{i} p_{i}^{n_{i}}$.

## Duality

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\end{gathered}
$$

Strategy: Calculate $d\left(\underline{p}^{\underline{n}}\right)$ for $\underline{n}$ fixed. Choose the process $\underline{n}$ in such a way that equation (*) is satisfied.

Itô's formula gives

$$
\begin{aligned}
& d\left(\underline{p}^{\underline{n}}\right)=\sum_{i} n_{i} \underline{\underline{p}}^{\underline{n}-\underline{e}_{i}}\left[s p_{i}\left(1-p_{i}\right)+\sum_{j} m_{i j}\left(p_{j}-p_{i}\right)\right] d t \\
&+\sum_{i} \gamma \frac{1}{2} n_{i}\left(n_{i}-1\right) \underline{p}^{\underline{n}-2 \underline{e}_{i}} p_{i}\left(1-p_{i}\right) d t+\sum_{i}(\ldots) d B_{i}
\end{aligned}
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## Rearranging,

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\begin{aligned}
d(\underline{p} \underline{n})=\sum_{i} n_{i} s & \left(\underline{p}^{\underline{n}}-\underline{p} \underline{\underline{n}}+\underline{e}_{i}\right) d t \\
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\underline{n} \mapsto \underline{n}-\underline{e}_{i} \text { at rate } \frac{\gamma}{2} n_{i}\left(n_{i}-1\right)
\end{aligned}
$$

$$
+\sum_{i}(\ldots) d B_{i} .
$$

## The 'coalescent' dual

The dual process $\underline{n}$ evolves as follows:

- $n_{i} \mapsto n_{i}+1$ at rate $-s n_{i}$
- $\left\{\begin{array}{l}n_{i} \mapsto n_{i}-1 \\ n_{j} \mapsto n_{j}+1\end{array}\right.$ at rate $n_{i} m_{i j}$
- $n_{i} \mapsto n_{i}-1$ at rate $\frac{1}{2} \gamma n_{i}\left(n_{i}-1\right)$

$$
\mathbb{E}\left[\underline{p}_{t}^{\underline{n}_{0}}\right]=\mathbb{E}\left[\underline{p}_{0}^{\underline{n}_{t}}\right] .
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## Long-time behaviour

Suppose that $I=\mathbb{Z}^{d}$ and $m_{i j}$ simple random walk.

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## Long-time behaviour

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