

Asymptotic Theory of the Planar Stepping Stone Model

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We focus on

- spatially explicit stepping stone model on the torus in \mathbb{Z}^2
- coalescing random walks on the torus in \mathbb{Z}^2
- effect of spatial structure on some measures of kinship

Based on work with Rick Durrett and Iljana Zähle:

*The stepping stone model: new formulas expose old myths (with R. Durrett). **Ann. Appl. Probab.**, 12 (2002), 1348-1377.*

*The stepping stone model II: Genealogies and the infinite sites model (with R. Durrett and I. Zähle). To appear, **Ann. Appl. Probab.***

1 Discrete Time Stepping Stone Model

Model for the evolution of gene frequencies in a population.

Colonies of individuals subject to migration and mutation.

- **Time** discrete, $n = 0, 1, \dots$
- **Colonies** located at sites $x \in \mathcal{S}$
- **Colony size** N (diploid), $2N$ (haploid case)
- **Types** mutations produce new types
- **Mutation probability** $\mu > 0$
- **Migration probability** $\nu \in (0, 1]$
- **Migration kernel** $q(x, y), x, y \in \mathcal{S}, q(x, x) = 0$

Formation of generation $n + 1$ at colony C_x

Each individual, independently:

- prob. μ : assumes a new type
- prob. $1 - \mu$: assumes type of individual chosen at random from colony C_y with probability

$$p(x, y) = (1 - \nu)I(x, y) + \nu q(x, y)$$

Note

- Wright-Fisher type model
- keep kernel $q(x, y)$ fixed, allow other parameters to vary

Duality

We can trace **lineages** of individuals backward in time, following paths of **coalescing random walks**.

Two lineages

W_n^1, W_n^2 : independent walks on \mathcal{S} , kernel $p(x, y)$.

U_n^1, U_n^2 : independent, uniformly distributed on $\{1, 2, \dots, 2N\}$.

Note. In k steps of W_n^1 , about νk are steps to different colonies.

Hitting Times

Starting with walks at 0 and x ,

T_{0x} = time required for two lineages to **reside in the same colony**

$$= \inf\{n : W_n^1 = W_n^2\}$$

t_{0x} = time required for two lineages to **coalesce**

$$= \inf\{n : (W_n^1, U_n^1) = (W_n^2, U_n^2)\}$$

Given the distribution of t_{0x} , we can compute

$$\begin{aligned} h &= \text{probability of } \textit{identity by descent} \text{ for two} \\ &\quad \text{(different) individuals picked at random} \\ &\quad \text{from entire population} \\ &= E(1 - \mu)^{2t_0} \end{aligned}$$

$$\begin{aligned} \phi(x) &= \text{probability of } \textit{identity by descent} \text{ for two} \\ &\quad \text{(different) individuals picked at random} \\ &\quad \text{from colonies } C_0 \text{ and } C_x \\ &= E(1 - \mu)^{2t_{0x}} \end{aligned}$$

$$\begin{aligned} F_{ST} &= \text{Wright's statistic} \\ &= \frac{\phi(0) - h}{1 - h} \quad (\text{following Nei (1975)}) \end{aligned}$$

2 The Single Colony Case

One colony, migration parameter $\nu = 0$

t_0 is **geometric**,

$$P(t_0 = k) = \left(\frac{1}{2N}\right)\left(1 - \frac{1}{2N}\right)^{k-1}, \quad k = 1, 2, \dots$$

For large N , $\frac{t_0}{2N} \stackrel{d}{\approx} \mathcal{E}(1)$

Calculation of h

$$\begin{aligned} h &= E(1 - 2\mu)^{2t_0} \\ &= E((1 - 2\mu)^{2N})^{t_0/2N} \\ &\approx E(e^{-4N\mu\mathcal{E}(1)}) \\ &= \frac{1}{1 + 4N\mu} \end{aligned}$$

$s > 2$ lineages

Consider s distinct lineages (coalescing rw's).

ζ_n^s = number of remaining lineages at time n

For large N , can show

$$\zeta_{[2Nt]}^s \stackrel{d}{\approx} D_t^s, \quad t > 0$$

where $D_t^s, t \geq 0$ on $\{1, 2, \dots, s\}$ is the pure death process, with

$$k \rightarrow 1 \text{ at rate } \binom{k}{2}$$

Can also keep track (on time scale $2N$) of which of the s random walks have coalesced, leading to **Kingman's coalescent**.

Consistent with previous $s = 2$ case?

$$\zeta_{[2Nt]} = 2 \iff t_0 > [2Nt]$$

and

$$P(D_t^2 = 2) = P(\mathcal{E}(1) > t) = e^{-t}$$

3 $\mathcal{S} = \text{the torus } \Lambda(L) \text{ in } \mathbb{Z}^d$

$$\Lambda(L) = (-L/2, L]^2 \cap \mathbb{Z}^2 \text{ (wrap around)}$$

Dynamics: each individual, independently:

- prob. μ : assumes a new type
- prob. $1 - \mu$: assumes type of individual chosen at random from colony C_y with prob.

$$p^L(x, y) = (1 - \nu)I(x, y) + \nu q^L(x, y)$$

where

$$q^L(x, y) = \sum_{z \in \mathbb{Z}^2} q(x, y + Lz)$$

Assumptions: $q(x, y) = q(y, x) = q(0, y - x)$, finite range, covariance matrix $\sigma^2 I$

Other models: island model, circular, ...

Seek distributions of

$$t_{0x}^L \quad \text{and} \quad \zeta_n^s$$

as $L \rightarrow \infty$, and possibly $N \rightarrow \infty, \nu \rightarrow 0$

4 Migration rate vs. system size

We consider the following “regimes”

Migration rate vs. system size

$$N\nu \gg \log L$$

$$N\nu \approx \log L$$

$$N\nu \ll \log L$$

Why? Let \tilde{t}_{00}^L be coalescent time for two walks starting in same colony. Then

$$t_{0x}^L \stackrel{d}{=} T_{0x}^L + \tilde{t}_{00}^L$$

and we will see that

$$\tilde{t}_{00}^L \approx 2NL^2$$

and if $|x| \approx L$ then

$$T_{0x}^L \approx \frac{L^2 \log L}{\nu}$$

Previous work: $2N = 1, \nu = 1$, nearest-neighbor kernel

Migration rate vs. system size

$$T_{0x}^L \ll \tilde{t}_{00}^L \quad \frac{N\nu}{\log L} \rightarrow \infty$$

$$T_{0x}^L \asymp \tilde{t}_{00}^L \quad \frac{4\pi\sigma^2 N\nu}{\log L} \rightarrow \alpha \in (0, \infty)$$

$$T_{0x}^L \gg \tilde{t}_{00}^L \quad \frac{N\nu}{\log L} \rightarrow 0$$

Behavior of t_{00}^L : high migration rate

Theorem (C., Durrett) Assume $\frac{N\nu}{\log L} \rightarrow \infty$ as $L \rightarrow \infty$.

Then for any $t > 0$,

$$\sup_{x \in \Lambda_L} \left| P \left(\frac{t_{0,x}^L}{2NL^2} > t \right) - e^{-t} \right| \rightarrow 0.$$

That is, uniformly in $x \in \Lambda(L)$, for large L ,

$$\frac{t_{0,x}^L}{2NL^2} \stackrel{d}{\approx} \mathcal{E}(1)$$

We use

Theorem (Strobeck (1987))

$$E(t_{00}^L) = 2NL^2 (= \text{population size})$$

Spatial scales

$$x_1, x_2, \dots, x_s \in \Lambda(L)$$

For $i \neq j$,

$$|x_i - x_j| \approx L^\beta, \quad 0 \leq \beta \leq 1$$

$\beta = 1 \iff$ uniformly spread out on torus

Behavior of T_{0x}^L

Theorem (C., Durrett) Suppose $x = x_L$ satisfies

$$\frac{\log^+ |x|}{\log L} \rightarrow \beta \in [0, 1] \quad \text{as } L \rightarrow \infty.$$

For all $t > 0$, uniformly for $\nu \in (0, 1]$,

$$P \left(\frac{T_{0x}^L}{\left(\frac{L^2 \log L}{2\pi\sigma^2\nu} \right)} > t \right) \rightarrow \beta e^{-t}$$

That is, if $|x| \approx L^\beta$ and $\tau_L = \frac{L^2 \log L}{2\pi\sigma^2\nu}$ then

$$\frac{t_{0,x}^L}{\tau_L} \stackrel{d}{\approx} \begin{cases} 0 & \text{prob. } 1 - \beta \\ \mathcal{E}(1) & \text{prob. } \beta \end{cases}$$

Low migration rate: $N\nu = O(\log L)$

Theorem (C., Durrett) Assume $\frac{4\pi\sigma^2 N\nu}{\log L} \rightarrow \alpha \in [0, \infty)$
and $x = x_L$ satisfies $\frac{\log^+ |x|}{\log L} \rightarrow \beta \in [0, 1]$ as $L \rightarrow \infty$.

Then for any $t > 0$,

$$P\left(\frac{t_{0,x}^L}{\tau_L} > (1 + \alpha)t\right) \rightarrow \left(\beta + (1 - \beta)\frac{\alpha}{1 + \alpha}\right) e^{-t}.$$

That is, if $\frac{4\pi\sigma^2 N\nu}{\log L} \approx \alpha$ and $|x| \approx L^\beta$ then

$$\frac{t_{0,x}^L}{(1 + \alpha)\tau_L} \stackrel{\text{d}}{\approx} \begin{cases} 0 & \text{prob. } \frac{1 - \beta}{1 + \alpha} \\ \mathcal{E}(1) & \text{prob. } \frac{\beta + \alpha}{1 + \alpha} \end{cases}$$

$$\tau_L = \frac{L^2 \log L}{2\pi\sigma^2\nu}$$

Example. Approximation of $\phi(x)$ using

$$\frac{t_{0,x}^L}{(1+\alpha)\tau_L} \stackrel{d}{\approx} \begin{cases} 0 & \text{prob. } \frac{1-\beta}{1+\alpha} \\ \mathcal{E}(1) & \text{prob. } \frac{\beta+\alpha}{1+\alpha} \end{cases}$$

where $\frac{4\pi\sigma^2 N\nu}{\log L} \approx \alpha$ and $|x| \approx L^\beta$

$$\begin{aligned} \phi(x) &= E((1-\mu)^{2t_{0,x}^L}) \\ &= E(((1-\mu)^{2(1+\alpha)\tau_L})^{t_{0,x}^L/(1+\alpha)\tau_L}) \\ &\approx \frac{1-\beta}{1+\alpha} + \frac{\beta+\alpha}{1+\alpha} E(((1-\mu)^{2(1+\alpha)\tau_L})^{\mathcal{E}(1)}) \\ &= \frac{1-\beta}{1+\alpha} + \frac{\beta+\alpha}{1+\alpha} \frac{1}{2(1+\alpha)\mu\tau_L} \end{aligned}$$

Some special cases.

1. $N\nu \ll \log L$ and $|x| \approx L$ ($\alpha = 0$ and $\beta = 1$)

$$h = \phi(x) \approx \frac{1}{1 + 2\mu\tau_L} = \frac{1}{1 + 4N_e\mu}$$

where

$$N_e = \frac{\tau_L}{2} = \frac{L^2 \log L}{4\pi\sigma^2\nu}$$

For $N = 20$, $L = 50$, $\nu = 0.1$ ($N\nu = 2$), $q(x, y)$ uniform on $[-2, 2]^2 \setminus (0, 0)$, $\sigma^2 = 50/24$,
 $N_e = 3736 \ll NL^2 = 50,000$.

Compare with **the island model**: there are k colonies,

$$p(x, y) = \begin{cases} 1 - \nu & x = y \\ \nu/(k - 1) & x \neq y \end{cases}$$

$$N_e = Nk \left(1 + \frac{(k - 1)^2}{4N\nu k^2} \right) > \text{actual population size}$$

2. Correlation with distance $N\nu \ll \log L$ ($\alpha = 0$) and $|x| \approx L^\beta$ or $\beta \approx \log |x| / \log L$,

$$\begin{aligned}\phi(x) &= E(1 - \mu)^{2t_{0,x}^L} \\ &\approx (1 - \beta) + \beta \frac{1}{2\mu\tau_L}\end{aligned}$$

$$\begin{aligned}\phi(0) - \phi(x) &\approx 1 - \left((1 - \beta) + \frac{\beta}{2\mu\tau_L} \right) \\ &= \beta \left(1 - \frac{\beta}{2\mu\tau_L} \right) \\ &\approx \beta 2\mu\tau_L \\ &\approx \log |x| \frac{L^2}{\pi\sigma^2\nu}\end{aligned}$$

3. Wright's statistic F_{ST}

$$\frac{4\pi\sigma^2 N\nu}{\log L} \rightarrow \alpha > 0$$

$$F_{ST} = \frac{\phi(0) - h}{1 - h}$$

Approximate h ($\beta = 1$) and $\phi(0)$ ($\beta = 0$) as before, get

$$F_{ST} \approx \frac{1}{1 + \alpha} \approx \frac{1}{8\pi\sigma^2 N\mu} \log L^2$$

Crow and Aoki (1984), numerical studies, F_{ST} proportional to the log of the number of colonies.

$$5 \quad \text{Times } t \ll \tau_L = \frac{L^2 \log L}{2\pi\sigma^2\nu}$$

If $\frac{4\pi\sigma^2 N\nu}{\log L} \approx \alpha$ and $|x| \approx L^\beta$ then

$$\frac{t_{0,x}^L}{(1+\alpha)\tau_L} \stackrel{\text{d}}{\approx} \begin{cases} 0 & \text{prob. } \frac{1-\beta}{1+\alpha} \\ \mathcal{E}(1) & \text{prob. } \frac{\beta+\alpha}{1+\alpha} \end{cases}$$

Change to continuous time version of model (Moran type).

For $0 < \delta \leq 1$ and $c > 0$ let

$$\Gamma(L, c, \delta) = (L^\delta / \log L, c\delta L^\delta \log L)$$

Theorem (C. Durrett, Zähle) *Assume*

$2N\nu\pi\sigma^2 / \log L \rightarrow \alpha \in [0, \infty)$ as $L \rightarrow \infty$. For any fixed $\beta_0 > 0$, uniformly in $\beta_0 \leq \beta \leq \gamma \leq 1$ and $|x| \in \Gamma(L, c, \beta)$,

$$P \left(t_{0x}^L > \frac{L^{2\gamma}}{2\nu} \right) \rightarrow \frac{\beta + \alpha}{\gamma + \alpha}$$

as $L \rightarrow \infty$.

6 $s > 2$ lineages

$$h_L = (1 + \alpha)\tau_L = (1 + \alpha)L^2 \log L / (2\pi\sigma^2\nu)$$

$$\mathcal{G}(L, n, 1) = \left\{ A = \{x_1, \dots, x_s\} : \right.$$

$$\left. \forall i, x_i \in \Lambda(L), \forall i \neq j, |x_i - x_j| \geq \frac{L}{\log L} \right\}.$$

Theorem (C. Durrett, Zähle) *Assume*

$2N\nu\pi\sigma^2 / \log L \rightarrow \alpha \in [0, \infty)$ as $L \rightarrow \infty$. *Uniformly in*
 $t \geq 0, A \in \mathcal{G}(L, s, 1)$,

$$|P(|\zeta_{h_L t}^A| = k) - P(D_t^s = k)| \rightarrow 0.$$

$$\mathcal{G}(L, n, c, \delta) = \left\{ A = \{x_1, \dots, x_n\} : \right. \\ \left. \forall i, x_i \in \Lambda(L), \forall i \neq j, |x_i - x_j| \in \Gamma(L, c, \delta) \right\}.$$

Theorem (C. Durrett, Zähle) *Assume*

$2N\nu\pi\sigma^2 / \log L \rightarrow \alpha \in [0, \infty)$ as $L \rightarrow \infty$. *Uniformly in*
 $\beta_0 \leq \beta \leq \gamma \leq 1$ and $A \in \mathcal{G}(L, s, c, \beta)$,

$$P \left(\left| \zeta_{\frac{L^{2\gamma}}{2\nu}}^A \right| = k \right) - P \left(D_{\log \frac{\gamma+\alpha}{\beta+\alpha}}^s = k \right) \rightarrow 0$$

Uniformly for $t \geq 0$ and $A \in \mathcal{G}(L, s, c, \beta)$,

$$\Pr \left(\left| \zeta_{\frac{L^2}{2\nu} + h_L t}^A \right| = k \right) - P \left(D_{\log \frac{1+\alpha}{\beta+\alpha} + t}^s = k \right) \rightarrow 0.$$