# The $\Lambda$ -Fleming-Viot process and a connection with Wright-Fisher diffusion

Bob Griffiths University of Oxford A *d*-dimensional  $\Lambda$ -Fleming-Viot process  $\{X(t)\}_{t\geq 0}$  representing frequencies of *d* types of individuals in a population has a generator described by

$$\mathcal{L}g(\boldsymbol{x}) = \int_0^1 \sum_{i=1}^d x_i \Big( g(\boldsymbol{x}(1-y) + y\boldsymbol{e}_i) - g(\boldsymbol{x}) \Big) \frac{F(dy)}{y^2}.$$

The population is partitioned at events of change by choosing type  $i \in [d]$  to reproduce with probability  $x_i$ , then rescaling the population with additional offspring y of type i to be  $\mathbf{x}(1-y)+y\mathbf{e}_i$  at rate  $y^{-2}F(dy)$ .

### Examples

Eldon and Wakeley (2006). A model where F has a single point of increase in (0, 1] with a possible atom at 0.

A natural class that arises from discrete models is when F has a Beta $(\alpha, \beta)$  density, particularly a Beta $(2 - \alpha, \alpha)$  density coming from a discrete model where the offspring distribution tails are asymptotic to a power law of index  $\alpha$ . Birkner, Blath, Capaldo, Etheridge, Möhle, Schweinsberg, Wakolbinger (2005) give a connection to stable processes.

Birkner and Blath (2009) describe the  $\Lambda$ -Fleming-Viot process and discrete models whose limit gives rise to it. If F has a single atom at 0, then  $\{X_t\}_{t\geq 0}$  is the d-dimensional Wright-Fisher diffusion process with generator

$$\mathcal{L} = \frac{1}{2} \sum_{i,j=1}^{d} x_i (\delta_{ij} - x_j) \frac{\partial^2}{\partial x_i \partial x_j}$$

 $X_1(t)$  is a one-dimensional Wright-Fisher diffusion process with generator

$$\mathcal{L} = \frac{1}{2}x(1-x)\frac{\partial^2}{\partial x^2}$$

The  $\Lambda$ -coalescent process is a random tree process back in time which has multiple merger rates for a specific k lineages coalescing while n edges in the tree of

$$\lambda_{nk} = \int_0^1 x^k (1-x)^{n-k} \frac{F(dx)}{x^2}, \ k \ge 2$$

After coalescence there are n - k + 1 edges in the tree.

This process was introduced by Pitman (1999), Sagitov (1999) and has been extensively studied. Berestycki (2009); recent results in the  $\Lambda$ -coalescent.

There is a connection between continuous state branching processes and the  $\Lambda\text{-}coalescent.$  The connection is through the Laplace exponent

$$\psi(q) = \int_0^1 \left( e^{-qy} - 1 + qy \right) y^{-2} F(dy)$$

Bertoin and Le Gall (2006) showed that the  $\Lambda$ -coalescent comes down from infinity under the same condition that the continuous state branching process becomes extinct in finite time, that is when

$$\int_1^\infty \frac{dq}{\psi(q)} < \infty$$

Some papers on the  $\Lambda$ -coalescent

Berestycki (2009) Recent progress in coalescent theory.

Berestycki, Berestycki, and Limic (2012) Asymptotic sampling formulae for  $\Lambda$ -coalescents.

Berestycki, Berestycki, and Limic (2012) A small-time coupling between  $\Lambda$ -coalescents and branching processes.

Bertoin and Le Gall (2003) Stochastic flows associated to coalescent processes.

Bertoin and Le Gall (2006) J. Bertoin and J.-F. Le Gall (2006). Stochastic flows associated to coalescent processes III: Limit theorems.

Birkner, Blath, Capaldo, Etheridge, Möhle, Schweinsberg, Wakolbinger, (2005) Alpha-stable branching and Beta-coalescents.

Birkner and Blath (2009) Measure-Valued diffusions, general coalescents and population genetic inference.

## A Wright-Fisher generator connection

#### Theorem

Let  $\mathcal{L}$  be the  $\Lambda$ -Fleming-Viot generator, V be a uniform random variable on [0,1], U a random variable on [0,1] with density 2u, 0 < u < 1 and W = YU, where Y has distribution F and V, U, Y are independent. Denote the second derivatives of a function  $g(\mathbf{x})$  by  $g_{ij}(\mathbf{x})$ .

Then

$$\mathcal{L}g(\boldsymbol{x}) = \frac{1}{2} \sum_{i,j=1}^{d} x_i (\delta_{ij} - x_j) \mathbb{E} \Big[ g_{ij} \big( \boldsymbol{x}(1-W) + WV\boldsymbol{e}_i \big) \Big]$$

where expectation  $\mathbb E$  is taken over V, W.

Wright-Fisher generator

$$\mathcal{L}g(\boldsymbol{x}) = \frac{1}{2} \sum_{i,j=1}^{d} x_i (\delta_{ij} - x_j) g_{ij}(\boldsymbol{x})$$

 $\Lambda\textsc{-}\mathsf{Fleming}\textsc{-}\mathsf{Viot}$  generator

$$\mathcal{L}g(\boldsymbol{x}) = \frac{1}{2} \sum_{i,j=1}^{d} x_i (\delta_{ij} - x_j) \mathbb{E} \Big[ g_{ij} \Big( \boldsymbol{x}(1 - W) + WV \boldsymbol{e}_i \Big) \Big]$$

Method of proof

$$\mathcal{L}g(\boldsymbol{x}) = \int_0^1 \sum_{i=1}^d x_i \Big( g(\boldsymbol{x}(1-y) + y\boldsymbol{e}_i) - g(\boldsymbol{x}) \Big) \frac{F(dy)}{y^2}$$

$$\mathcal{L}g(\boldsymbol{x}) = \frac{1}{2} \sum_{i,j=1}^{d} x_i (\delta_{ij} - x_j) \mathbb{E} \Big[ g_{ij} \Big( \boldsymbol{x}(1-W) + WV\boldsymbol{e}_i \Big) \Big]$$

Show that the generators have the same answer acting on

$$g(\boldsymbol{x}) = \exp\Big\{\sum_{i=1}^{d} \eta_i x_i\Big\}, \ \boldsymbol{\eta} \in \mathbb{R}^d$$

## 1-dimensional generator

Wright-Fisher diffusion generator

$$\mathcal{L}g(x) = \frac{1}{2}x(1-x)g''(x)$$

# $\Lambda\textsc{-}\mathsf{Fleming}\textsc{-}\mathsf{Viot}$ process generator

$$\mathcal{L}g(x) = \frac{1}{2}x(1-x)\mathbb{E}\left[g''\left(x(1-W) + WV\right)\right]$$

or

$$\mathcal{L}g(x) = \frac{1}{2}x(1-x)\mathbb{E}\left[\frac{g'\left(x(1-W)+W\right) - g'\left(x(1-W)\right)}{W}\right]$$

The Laplace transform of  $\boldsymbol{W}$  is related to the Laplace exponent by

$$\mathbb{E}\left[e^{-\eta W}\right] = 2\int_0^1 \frac{e^{-\eta y} - 1 + \eta y}{(y\eta)^2} F(dy)$$

W = UY is continuous in (0, 1) with a possible atom at 0.

$$P(W=0) = P(Y=0)$$

## Adding mutation

The generator has an additional term added of

$$\frac{\theta}{2} \sum_{i=1}^{d} \left( \sum_{j=1}^{d} p_{ji} x_j - x_i \right) \frac{\partial}{\partial x_i}$$

If mutation is parent independent  $\theta p_{ij} = \theta_j$ , not depending on i, and the additional term is

$$\frac{1}{2}\sum_{i=1}^{d} \left(\sum_{j=1}^{d} \theta_{j} x_{j} - \theta x_{i}\right) \frac{\partial}{\partial x_{i}}$$

Eigenstructure of the  $\Lambda\text{-}\mathsf{Fleming}\text{-}\mathsf{Viot}$  process

#### Theorem

Let  $\{\lambda_n\}$ ,  $\{P_n(\boldsymbol{x})\}$  be the eigenvalues and eigenvectors of  $\mathcal{L}$ , the generator which includes mutation, satisfying

$$\mathcal{L}P_{\boldsymbol{n}}(\boldsymbol{x}) = -\lambda_{\boldsymbol{n}}P_{\boldsymbol{n}}(\boldsymbol{x})$$

Denote the d-1 eigenvalues of the mutation matrix P which have modulus less than 1 by  $\{\phi_k\}_{k=1}^{d-1}$ .

The eigenvalues of  $\mathcal L$  are

$$\lambda_{\mathbf{n}} = \frac{1}{2}n(n-1)\mathbb{E}\left[(1-W)^{n-2}\right] + \frac{\theta}{2}\sum_{k=1}^{d-1}(1-\phi_k)n_k$$

## Polynomial Eigenvectors

Denote the d-1 eigenvalues of P which have modulus less than 1 by  $\{\phi_k\}_{k=1}^{d-1}$  corresponding to eigenvectors which are rows of a  $d-1\times d$  matrix R satisfying

$$\sum_{i=1}^{d} r_{ki} p_{ji} = \phi_k r_{kj}, \ k = 1, \dots, d-1.$$

Define a d-1 dimensional vector  $\boldsymbol{\xi} = R\boldsymbol{x}$ .

The polynomials  $P_n(\mathbf{x})$  are polynomials in the d-1 terms in  $\boldsymbol{\xi} = (\xi_1, \dots, \xi_{d-1})$  whose only leading term of degree n is

$$\prod_{j=1}^{d-1} \xi_j^{n_j}$$

In the parent independent model of mutation

$$\lambda_{\boldsymbol{n}} = \lambda_n = \frac{1}{2}n\left\{(n-1)\mathbb{E}\left[(1-W)^{n-2}\right] + \theta\right\}$$

repeated  $\binom{n+d-2}{n}$  times with non-unique polynomial eigenfunctions within the same degree n.

The non-unit eigenvalues of the mutation matrix with identical rows are zero.

Wright-Fisher diffusion, general mutation structure.

$$\lambda_{\mathbf{n}} = \frac{1}{2}n(n-1) + \frac{\theta}{2}\sum_{k=1}^{d-1}(1-\phi_k)n_k$$

 $\Lambda$ -coalescent eigenvalues and rates

$$\frac{1}{2}n(n-1)\mathbb{E}\left[(1-W)^{n-2}\right] = \sum_{k=2}^{\infty} \binom{n}{k} \lambda_{nk}$$

which is the total jump rate away from n individuals.

These are the eigenvalues in the  $\Lambda$ -coalescent tree.

The individual rates can be expressed as

$$\binom{n}{k}\lambda_{nk} = \binom{n}{k}\int_0^1 y^k (1-y)^{n-k} \frac{F(dy)}{y^2}$$
$$= \frac{n}{2}\mathbb{E}\left[\frac{P_k(n,W) - P_{k-1}(n,W)}{W^2}\right],$$

where

$$P_k(n,w) = \binom{n-1}{k-1} (1-w)^{n-k} w^k$$

is a negative binomial probability of a waiting time of n trials to obtain k successes, where the success probability is w.

## Two types

The generator is specified by

$$\mathcal{L}g(x) = \frac{1}{2}x(1-x)\mathbb{E}\left[g''(x(1-W)+WV)\right] + \frac{1}{2}(\theta_1 - \theta_x)g'(x)$$

The eigenvalues are

$$\lambda_n = \frac{1}{2}n\left\{(n-1)\mathbb{E}\left[(1-W)^{n-2}\right] + \theta\right\}$$

and the eigenvectors are polynomials satisfying

$$\mathcal{L}P_n(x) = -\lambda_n P_n(x), \ n \ge 1.$$

## Polynomial eigenvectors

$$\frac{1}{2}x(1-x)\mathbb{E}\left[\frac{P'_n(x(1-W)+W) - P'_n(x(1-W))}{W} + \frac{1}{2}(\theta_1 - \theta_x)P'_n(x) \\ = \frac{1}{2}n\left[(n-1)\mathbb{E}\left[(1-W)^{n-2}\right] + \theta\right]P_n(x)$$

The monic polynomial  $P_n(x)$  is uniquely defined by recursion of its coefficients.

Stationary distribution  $\psi(x)$ 

$$\int_0^1 \mathcal{L}g(x)\psi(x)dx = 0$$
  
$$\sigma^2(x) = x(1-x), \ \mu(x) = \theta_1 - \theta x$$
  
$$k(x) = \mathbb{E}\Big[(1-W)^{-2}g\Big(x(1-W) + VW\Big)\Big]$$

An equation for the stationary distribution

$$0 = \int_0^1 \left[ k(x) \frac{1}{2} \frac{d^2}{dx^2} \left[ \sigma^2(x) \psi(x) \right] - g(x) \frac{d}{dx} \left[ \mu(x) \psi(x) \right] \right] dx + k(x) \frac{d}{dx} \left[ \frac{1}{2} \sigma^2(x) \psi(x) \right] \Big|_0^1 + g(x) \mu(x) \psi(x) \Big|_0^1$$

In a diffusion process k(x) = g(x) and the boundary terms vanish. Then there is a solution found by solving

$$\frac{1}{2}\frac{d^2}{dx^2} \Big[\sigma^2(x)\psi(x)\Big] - \frac{d}{dx} \Big[\mu(x)\psi(x)\Big] = 0$$

however  $k(x) \neq g(x)$  so we do not have an equation like this.

Green's function,  $\gamma(x)$ 

Solve, for a given function g(x)

$$\mathcal{L}\gamma(x) = -g(x), \ \gamma(0) = \gamma(1) = 0.$$

Then

$$\gamma(x) = \int_0^1 G(x,\xi)g(\xi)d\xi$$

A non-linear equation, equivalent to

$$\frac{1}{2}x(1-x)\mathbb{E}\Big[\gamma''(x(1-W)+VW)\Big] = -g(x)$$

# Green's function solution

Define

$$k(x) = \mathbb{E}\left[(1-W)^{-2}\gamma\left(x(1-W)+VW\right)\right]$$

then

$$k''(x) = -2\frac{g(x)}{x(1-x)}$$

with a solution

$$k(x) = k(0)(1-x) + k(1)x + (1-x) \int_0^x \frac{2g(\eta)}{(1-\eta)} d\eta + x \int_x^1 \frac{2g(\eta)}{\eta} d\eta$$

Mean time to absorption

If g(x) = 1,  $x \in (0, 1)$  then  $\gamma(x)$  is the mean time to absorption at 0 or 1 when X(0) = x.

$$k(x) = k(0)(1-x) + k(1)x + (1-x) \int_0^x \frac{2}{(1-\eta)} d\eta + x \int_x^1 \frac{2}{\eta} d\eta$$

There is a non-linear equation to solve of

$$k(x) = k(0)(1-x) + k(1)x - 2(1-x)\log(1-x) - 2x\log x$$
 where

$$k(x) = \mathbb{E}\left[(1-W)^{-2}\gamma\left(x(1-W)+VW\right)\right]$$

Stationary distribution,  $\Lambda$ -Fleming process with mutation

Let Z be a random variable with the size-biassed distribution of X,  $Z_*$  a size-biassed Z random variable and  $Z^*$  a size-biassed random variable with respect to 1 - Z.

Let B be Bernoulli random variable, independent of the other random variables in the following equation such that

$$P(B=1) = \frac{\theta - \theta_1}{\theta(\theta_1 + 1)}$$

An interesting distributional identity

$$VZ_* = \mathcal{D}\left(1-B\right)VZ + B\left(Z^*(1-W) + WV\right)$$

The frequency spectrum in the infinitely-many-alleles model

Take a limit from a *d*-allele model with  $\theta_i = \theta/d$ ,  $i \in [d]$ . The limit is a point process  $\{X_i\}_{i=1}^{\infty}$ . The 1-dimensional frequency spectrum h(x) is a non-negative measure such that for suitable functions k on [0,1] in the stationary distribution

$$\mathbb{E}\left[\sum_{i=1}^{\infty} k(X_i)\right] = \int_0^1 k(x)h(x)dx.$$

Symmetry in the d-allele model shows that

$$\int_0^1 k(x)h(x)dx = \lim_{d \to \infty} d\mathbb{E}\Big[k(X_1)\Big].$$

The classical Wright-Fisher diffusion gives rise to the Poisson-Dirichlet process with a frequency spectrum of

$$h(x) = \theta x^{-1} (1-x)^{\theta-1}, \ 0 < x < 1.$$

 $\Lambda$  Fleming-Viot frequency spectrum

Let  ${\boldsymbol Z}$  have a density

$$f(z) = zh(z), \ 0 < z < 1$$

Interesting identity

$$VZ_* =^{\mathcal{D}} Z^*(1-W) + WV$$

where  $Z_*$  is size-biassed with respect to Z, and  $Z^*$  is size-biassing with respect to 1 - Z.

The constant  $\theta$  appears in the identity through scaling in the size-biassed distributions.

Limit distribution of excess life in a renewal process

$$P(VZ_* > \eta) = \int_{\eta}^{1} \frac{P(Z > z)}{\mathbb{E}[Z]} dz$$

## Typed dual $\Lambda$ -coalescent

The  $\Lambda$ -Fleming-Viot process is dual to the system of coalescing lineages  $\{L(t)\}_{t\geq 0}$  which takes values in  $\mathbb{Z}^d_+$  and for which the transition rates are, for  $i, j \in [d]$  and  $l \geq 2$ ,

$$\begin{split} q_{\Lambda}(\xi,\xi-e_{i}(l-1)) &= \int_{[0,1]} {\binom{|\xi|}{l}} y^{l}(1-y)^{|\xi|-l} \frac{F(dy)}{y^{2}} \\ &\times \frac{\xi_{i}+1-l}{|\xi|+1-l} \frac{\mathcal{M}(\xi-e_{i}(l-1))}{\mathcal{M}(\xi)} \\ q_{\Lambda}(\xi,\xi+e_{i}-e_{j}) &= \mu_{ij}(\xi_{i}+1-\delta_{ij}) \frac{\mathcal{M}(\xi+e_{i}-e_{j})}{\mathcal{M}(\xi)} \end{split}$$

The process is constructed as a moment dual from the generator.

## A different dual process

Define a sequence of monic polynomials  $\{g_n(x)\}$  by the generator equation

$$\frac{1}{2}x(1-x)\mathbb{E}g_n''(x(1-W)+VW) + \frac{1}{2}(\theta_1 - \theta_x)g_n'(x)$$
  
=  $\binom{n}{2}\mathbb{E}(1-W)^{n-2}[g_{n-1}(x) - g_n(x)] + n\frac{1}{2}[\theta_1g_{n-1}(x) - \theta_n(x)]$ 

The defining equation mimics the Wright-Fisher diffusion acting on test functions  $g_n(x) = x^n$ 

$$\frac{1}{2}x(1-x)\frac{d^2}{dx^2}x^n + \frac{1}{2}(\theta_1 - \theta_x)\frac{d}{dx}x^n$$
$$= \binom{n}{2}(x^{n-1} - x^n) + \frac{1}{2}n(\theta_1 x^{n-1} - \theta_x^n)$$

Jacobi polynomial analogues

The eigenfunctions are polynomials  $\{P_n(x)\}$  satisfying

$$\mathcal{L}P_n(x) = \lambda_n P_n(x)$$
$$P_n(x) = g_n(x) + \sum_{r=0}^{n-1} c_{nr} g_r(x)$$

The coefficients are

$$c_{nr} = \frac{\lambda_{r+1}^{\circ} \cdots \lambda_{n}^{\circ}}{(\lambda_{r} - \lambda_{n}) \cdots (\lambda_{n-1} - \lambda_{n})}$$

where 
$$\lambda_n = \frac{n}{2} \Big[ (n-1)\mathbb{E}(1-W)^{n-2} + \theta \Big]$$
  
 $\lambda_n^{\circ} = \frac{n}{2} \Big[ (n-1)\mathbb{E}(1-W)^{n-2} + \theta_1 \Big]$ 

In the stationary distribution

$$\mathbb{E}\Big[g_n(X)\Big] = \omega_n$$

where  $\omega_n$  is a Beta moment analog

$$\omega_n = \frac{\prod_{j=1}^n \left( (j-1)\mathbb{E}(1-W)^{j-2} + \theta_1 \right)}{\prod_{j=1}^n \left( (j-1)\mathbb{E}(1-W)^{j-2} + \theta \right)}$$

Let

$$h_n = \frac{g_n}{\omega_n}$$

SO

$$\mathbb{E}\Big[h_n(X)\Big] = 1$$

**Dual Generator** 

$$\mathcal{L}h_n = \lambda_n \Big[ h_{n-1} - h_n \Big]$$

Dual equation

$$\mathbb{E}_{X(0)=x}\left[h_n(X(t))\right] = \mathbb{E}_{N(0)=n}\left[h_{N(t)}(x)\right]$$

where  $\{N(t),t\geq 0\}$  is a death process with rates

$$\lambda_n = \frac{1}{2}n\left((n-1)\mathbb{E}\left[(1-W)^{n-2}\right] + \theta\right)$$

The process  $\{N(t),t\geq 0\}$  comes down from infinity if and only if

$$\sum_{n=2}^{\infty} \lambda_n^{-1} < \infty$$

which implies the  $\Lambda\text{-}\mathrm{coalescent}$  comes down from infinity.

The transition functions are then

$$P\left(N(t) = j \mid N(0) = i\right)$$
$$= \sum_{k=j}^{i} e^{-\lambda_k t} (-1)^{k-j} \frac{\prod_{\{l:j \le l \le k+1\}} \lambda_l}{\prod_{\{l:j \le l \le k+1, \ l \ne k\}} (\lambda_l - \lambda_k)}$$

 $P \Big( N(t) = j \mid N(0) = \infty \Big)$  is well defined if N(t) comes down from infinity.

In the Kingman coalescent the death rates are  $n(n - 1 + \theta)/2$ and N(t) describes the number of non-mutant lineages at time t back in the population. Wright-Fisher diffusion. Fixation probability with selection. Two types.

Let P(x) be the probability that the 1st type fixes, starting from an initial frequency of x. P(x) is the solution of

$$\mathcal{L}^{\sigma}P(x) = \frac{1}{2}x(1-x)P''(x) - \sigma x(1-x)P'(x) = 0$$
 with  $P(0) = 0$ ,  $P(1) = 1$ . The solution of this differential equation is

$$P(x) = \frac{e^{2\sigma x} - 1}{e^{2\sigma} - 1}$$

 $\Lambda$ -Fleming Viot process. Fixation probability with selection.

Let P(x) be the probability that the 1st type fixes, starting from an initial frequency of x. P(x) is the solution of

$$\mathcal{L}^{\sigma}P(x) = \frac{1}{2}x(1-x)\mathbb{E}\Big[P''(x(1-W)+WV)\Big] - \sigma x(1-x)P'(x) = 0$$
  
with  $P(0) = 0$ ,  $P(1) = 1$ .

Der, Epstein and Plotkin (2011,2012). For some measures Fand  $\sigma$  it is possible that P(x) = 1 or 0 for all  $x \in (0,1)$ . If  $\sigma > 0$ , fixation is certain if

$$\sigma > -\int_0^1 \frac{\log(1-y)}{y^2} F(dy)$$

A computational solution for P(x) when fixation or loss is not certain from  $x \in (0, 1)$ .

Define a sequence of polynomials  $\{h_n(x)\}_{n=0}^{\infty}$  for a choice of pre-specified constants  $\{h_n(0)\}$  as solutions of

$$\mathbb{E}\left[\frac{h_n(x(1-W)+W) - h_n(x(1-W))}{W}\right] = nh_{n-1}(x)$$

where the leading coefficient in  $h_n(x)$  is

$$\frac{1}{\prod_{j=1}^{n-1} \mathbb{E}\left[(1-W)^j\right]}$$

Polynomial solution for P(x)

$$P(x) = \left(e^{2\sigma} - 1\right)^{-1} \sum_{n=1}^{\infty} \frac{(2\sigma)^n}{n!} H_n(x),$$

where  $\{H_n(x)\}$  are polynomials derived from

$$H_n(x) = \int_0^x nh_{n-1}(\xi)d\xi$$

and the constants  $\{h_n(0)\}$  are chosen so that

$$\int_0^1 nh_{n-1}(\xi)d\xi = 1$$

The coefficients of  $H_n(x)$  are well defined by a recurrence relationship with the coefficients of  $H_{n-1}(x)$ .