Quetzal: a library for integrating distributional, demographic and coalescent models

Computational Medicine and Bioinformatics Tools & Technology Seminar Series

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Introduction

Environmental demogenetics

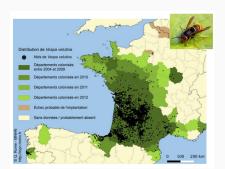
Environmental demogenetics aims to analyze geographic patterns of genetic diversity to inform the influence of **environment** on **demography**.

- 1. what data?
- 2. how to formalize these processes (what model)?
- 3. how to extract information from data (what inference method)?
- 4. how to implement the method (what tools)?

Example of a spatial genetic dataset

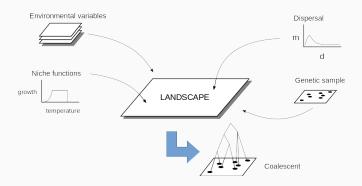
Vespa velutina nigritorax (yellow-legged hornet) invasion in Europe.

- first encountered in South-West France in 2004
- fast expansion
- economical/ecological impact (honey bee predator)
- 84 females genotyped for 22 SSR loci in 2008



lat	lon	List2003		VMA8	
43,50	2,21	163	163	262	264
43,50	2,21	163	169	262	264
43,56	1,46	163	169	260	264
43,97	3,68	177	177	262	264
44,14	4,17	177	177	264	267
44,22	0,55	177	177	264	264
44,22	0,55	163	177	262	264
44,22	0,55	169	177	262	264
44,29	0,69	177	177	264	264
44,29	0,69	163	163	246	264
44,37	3,41	165	169	0	0
44,57	0,23	163	169	264	264
44,57	0,23	163	177	249	264
44,57	0,23	163	163	249	264
44,57	0,23	153	163	249	264
44,57	0,23	169	177	249	264
44,57	0,19	163	163	264	264
44,57	0,19	163	169	264	264
44,57	0,19	163	169	249	264
44,57	0,19	163	177	249	264
44,67	2,16	169	177	264	264

Model summary



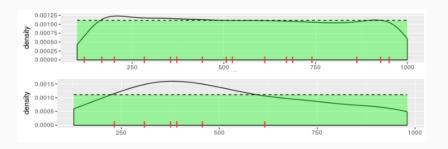
The main challenges are:

- chose submodels (model selection)
- estimate parameters from genetic data (statistical inference)

Inferential method: Approximate Bayesian Computation

Simulation/reject algorithm:

- ullet sample parameters in a prior distribution: $heta' \sim p(heta)$
- simulate data from the model: $y' \sim p(y|\theta')$
- weigh (θ', y') as a function of the distance $||y' y_{obs}||$



What tools for spatially explicit simulations of gene flow within landscapes?

Simulation tools, see Yannic et al. (2020)

	Program	Simulator	Level	Lg.	Reference
ĺ	Splatche3	Backward	Population	C++	Currat et al. (2019)
	${\sf PhyloGeoSim}$	Backward	Population	Java	Dellicour (2013)
	IBDsim	Backward	Population	С	Leblois et al. (2009)
	SLIM3	Forward	Individual	C++	Haller and Messer (2019)

There is an abundance of simulation **programs**, but no **library**:

C++ Library

egglib-cpp is the underlying C++ library of EggLib. It was designed with the aim of improving performance at the expense of safety and intuitive design. Therefore it might be difficult to use directly. The complete contents are listed below:

egglib._eggwrapper

 $C++\ library-direct\ use\ is\ strongly\ discouraged!$

C Library

The low-level code for msprime is written in C, and is structured as a standalone library. This code is all contained in the lib directory. Although the code is structured as a library, it is not intended to be used outside of the msprime project! The interfaces at the C level change considerably over time, and are deliberately undocumented.

The no-library-available game

- 1. If your biological model fits the assumptions of an existing simulation program x, then use x and win.
- 2. If your biological model presents important deviations from *x*, then used *x* anyway and win.
- 3. If your biological model present critical deviations from x, then a new progam y is required. Since developing y is too time consuming, go back to 2 or forfeit.

The library-available game

- 1. If your biological model fits the assumptions of an existing simulation program x, then use x and win.
- 2. If not, use the library to assemble simulation components as you see fit into a program and win.

Creating a new coalescence-based library

MOLECULAR ECOLOGY RESOURCES

The Quetzal Coalescence template library: A C++ programmers resource for integrating distributional, demographic and coalescent models

Arnaud Becheler M, Camille Coron, Stéphane Dupas

Programming choices for QuetzalCoalTL:

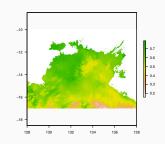
- C++ for a good compromise between design and performances
- Give as much information as possible to the compiler to make the code faster/safer
- Large use of generic programming (templates, metaprogramming)
- Focus on modularity and user experience

Using Quetzal for an Australian

lizard: demonstration

Heterogeneous landscape

- Landscape is discretized in n demes (grid cells).
- Let be L=1 environmental variables (suitability derived from a niche model).



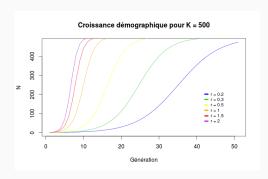
```
string path = "suitability.tiff"; using landscape_t = DiscreteLandscape<string,int>; landscape_t land( \{\{"suitability", path\}\},\{0\}\}); // auto s = land["suitability"]; // -> callable s(x,t) // For temporal heterogeneity: // land(\{\{"s1", path1\},\{"s2",path2\}\},\{0,-1000\});
```

Demography initialization

- The ancestral population is a Wright-Fisher of N_A haploid individuals
- Very ancient demography assumed non-spatial.
- At time t_0 , N_0 individuals are introduced in deme x_0 .
- The following history is then spatially explicit.

```
\label{eq:using_core_t} \begin{array}{rl} using & core_{-}t &= SpatiallyExplicit < coord_{-}t \;, \; time_{-}t \;, \\ & & demographic_{-}policy \;, \\ & & coalescence_{-}policy >; \\ core_{-}t & core \big(x_{-}0 \;, \; t_{-}0 \;, \; N_{-}0 \,\big); \\ core_{-} ancestral_Wright_Fisher_{-}N \, \big(N_{-}0 \,\big); \end{array}
```

- r: growth rate, constant
- k: carrying capacity, function of suitability



Carrying capacity:

- scaled by the suitability value on continental cells
- null in ocean cells "most of the time", but draft dispersal possible

```
auto s = land["suitability"]; // \rightarrow callable <math>s(x,t)
auto K = [\&gen, s]
                   // \rightarrow callable K(x,t)
(coord_type const& x, time_type)
  // if ocean cell:
            return 0 with proba 0.9, or 1
  // if continental cell:
            return 100*s(x,t)
};
```

Carrying capacity:

- scaled by the suitability value on continental cells
- null in ocean cells "most of the time", but draft dispersal possible

```
if( s(x,0) <= 0.001){ //ocean cell
  std::bernoulli_distribution dist(0.1);
  return dist(gen) ? 1 : 0;
}else{ // continental cell
  return 100*s(x,0);</pre>
```

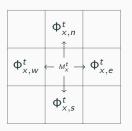
Number of children: $\tilde{N}_{x}^{t} \sim Poisson(g(x, t))$

```
auto g = N*(Iit(1)+r)/(Iit(1)+((r*N)/K));
auto children = [g](auto\& gen, auto x, auto t)\{
std::poisson_distribution < unsigned int > dist(g(x,t));
return dist(gen);
};
```

Dispersal

- ullet number of effective emigrants going out of deme x: $M_{\scriptscriptstyle X}^t=e imes ilde{N}_{\scriptscriptstyle X}^t$
- set of x neighbours (North, South, East, West): \mathbb{V}_{x}
- number of individuals going from x to $y \in \mathbb{V}_x$ at time t: $\Phi^t_{x,y}$
- ullet sampling emigrants destination in a multinomial law defines $\Phi^t_{x,y}$:

$$(\Phi_{x,y}^t)_{y\in\mathbb{V}_x}\sim \mathcal{M}(\tilde{N}_x^t,(p_{xy})_y)$$
.



Quetzal code:

using demographic_policy = strategy::mass_based; auto neighbors = make_neighboring_cells_functor(land);15

Friction

The term $(p_{xy})_y$ denotes the parameters of the multinomial law, giving for an emigrant in x its probability to go to $y \in \mathbb{V}_x$:

$$\begin{array}{cccc} \rho & : & \mathbb{X} \times \mathbb{V}_{\times} & \mapsto & [0,1] \\ & & (x,y) & \mapsto & \frac{1}{h(y) \cdot \sum_{i \in \mathbb{V}_{\times}} \frac{1}{h(i)}} \end{array}.$$

where h is a function of the suitability.

```
auto h = [s](auto x)\{ if (s(x,0) \le 0.5) {return 0.99;} // ocean or deserts else return 1.0 - s(x, 0);}; kernel = make_light_neighboring_migration(emigrant_rate, h, neighbors);
```

Demographic process

Flow of migrants
$$\Phi:$$
 $(\Phi^t_{x,y})_{y\in\mathbb{V}_x}\sim \mathcal{M}(\tilde{N}^t_x,(p_{xy})_y)$.

Population size
$$N: N_j^{t+1} = \sum_{i \in \mathbb{X}} \Phi_{i,j}^t$$
:

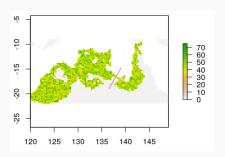


Figure 1: Click.

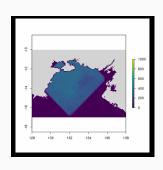


Figure 2: Click.

Quetzal code:

 $core.expand_demography(2021, children, kernel, gen)); 17$

Genetic process

Let be a set S de n gene copies sampled at time t_s . Coalescent trees are then simulated backward in time, from t_s to t_0 .

Knowing that a child node c is in deme $j \in \mathbb{X}$, the probability for its parents p to be in $i \in \mathbb{X}$ is a function of the migration flow Φ :

$$P(p \in i \mid e \in j) = \frac{\Phi_{i,j}^t}{\sum_k \Phi_{k,j}^t}$$

Knowing that the parent nodes p_1 (p_2) of the nodes c_1 (c_2) are in i at time t, the probability for the children to coalesce in the same parent is:

$$P(p_1 = p_2 \mid p_1 \in i, p_2 \in i) = 1/N_i^t$$

```
using coal_policy=distance_to_parent_leaf_name <...>;
core.coalesce_to_mrca <>(
sample, 2021, get_position, get_name, gen);
```

Conclusion

- Very flexible resource for demographic processes
- Easy to couple to coalescence simulators
- Next big step is to couple it to Tskit library (Kelleher et al., 2014) for efficient generation of correlated trees.
- Open source on github

Thank you for your attention!

DECRYPT

Simulation pipeline

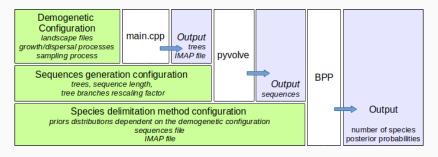


Figure 3: Pipeline to simulate coalescence trees conditioned on a complex spatially explicit demographic history and sampling schemes using Quetzal (Becheler et al., 2019), then simulating sequences along the trees using Pyvolve (Spielman and Wilke, 2015) and delimiting species using BPP (Flouri et al., 2018)

Results

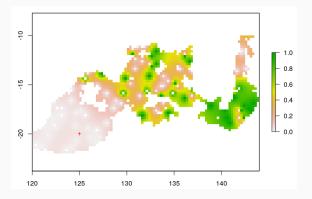


Figure 4: Spatial interpolation of p_x the probability to detect 2 species in a population expanding in an heterogeneous landscape under the MSC when the sequences sample is constructed at time t_s by two 2D gaussian sampling processes centered on (i) the population origin x_0 (red cross), and (ii) on a random coordinate x (with $N(x, t_s) > 30$ to avoid inconsistent sampling in very low density areas).

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