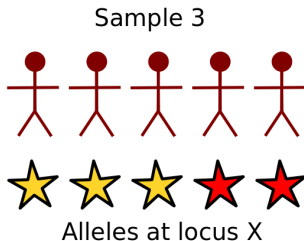
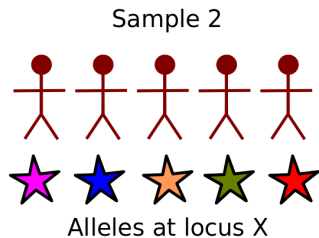
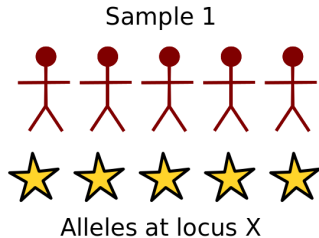


Introduction to the Coalescent

Mark Reppell

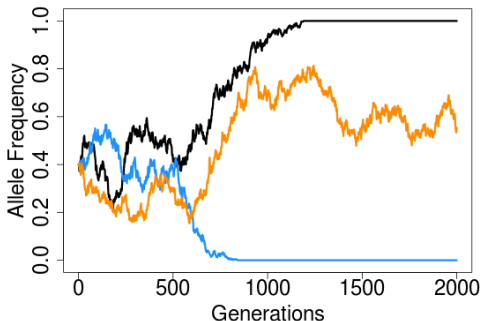
July 21, 2015

Why do we need models for how genes behave in populations?



- ▶ Genetic drift and the Wright-Fisher model
- ▶ What is the coalescent?
- ▶ Math behind the model
- ▶ Features of a genealogy
- ▶ Mutations and the infinite sites model

Genetic drift: the luck of the draw



Over time alleles may:

become fixed

remain segregating

become lost

Genetic drift refers to changes in allele frequencies over time as alleles **by chance** produce more/less offspring each generation

This contrasts with *natural selection* where alleles **systematically** produce more/less offspring each generation

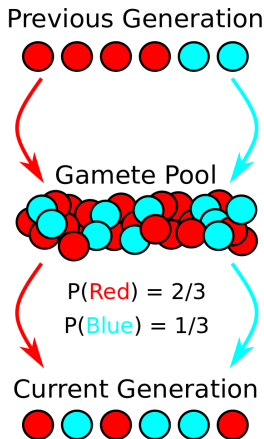
The Wright-Fisher model of genetic drift

In a population with size $2N$, two alleles are segregating at a genetic locus

Each new generation, every chromosome inherits one of the allele types, with probability p equal to $\frac{x}{2N}$ where x is the number of alleles of that type in the proceeding generation

At the population level the number of alleles of type i in a generation follows a binomial distribution, with p the allele frequency in the proceeding generation:

$$\text{Prob}(X_i = x) = \binom{2N}{x} p^x (1 - p)^{2N-x}$$

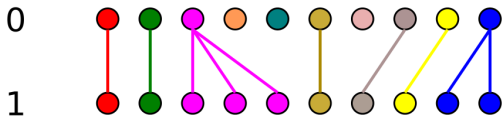


Assumptions of the Wright-Fisher model

- ▶ Discrete and non-overlapping generations
- ▶ Haploid chromosomes
- ▶ Randomly mating population with constant size
- ▶ No recombination or selection

A sample evolving under Wright-Fisher

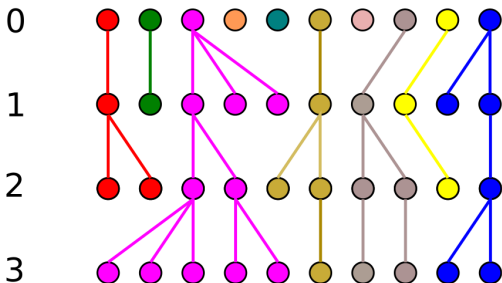
Generation



Every generation
each sample has
an ancestor in the
preceding
generation

A sample evolving under Wright-Fisher

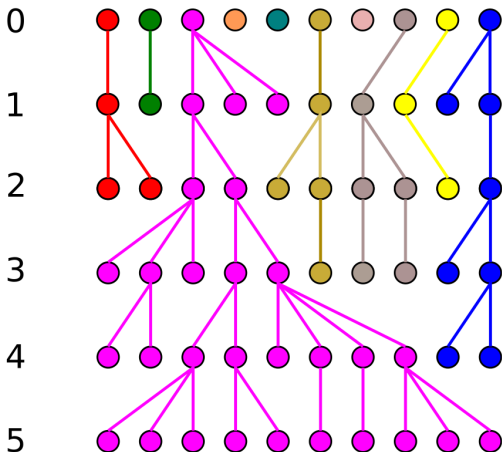
Generation



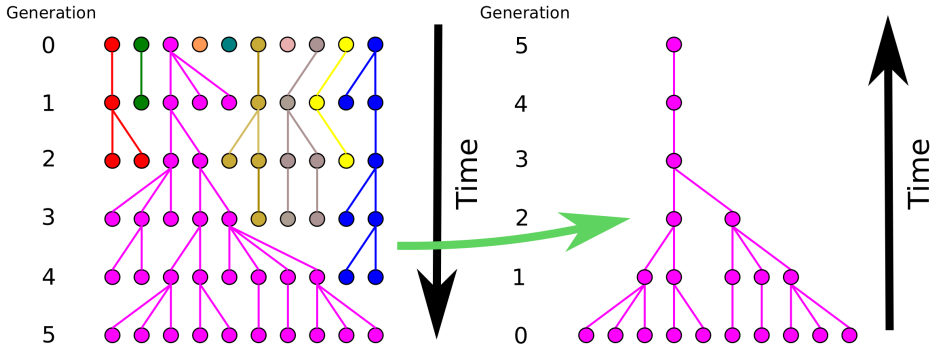
After several generations, all samples are descendants of a limited number of original individuals

A sample evolving under Wright-Fisher

Generation



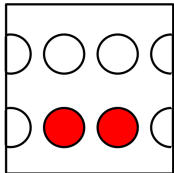
Reversing our perspective



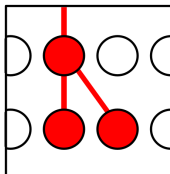
In the present, we cannot observe lineages that have been lost from the population

The discrete time Kingman coalescent

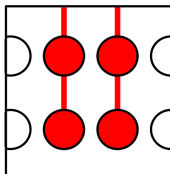
Begin with a haploid sample in the current generation



Each has an ancestor in the previous generation



If it is the same ancestor, this is a **coalescence**, and all preceding ancestors are common to both samples



If the ancestors are different, separate lineages continue into the past

How many generations does it take to find a common ancestor?

In a population with size $2N$

$$\underline{n = 2}$$

$$P(\text{coalesce in next gen}) = \frac{1}{2N}$$

$$P(\text{don't coalesce in 1 gen}) = \left(1 - \frac{1}{2N}\right)$$

$$P(\text{don't coalesce in } j \text{ gen}) = \left(1 - \frac{1}{2N}\right)^j$$

$$P(j \text{ gens until coalescence}) =$$

$$\left(1 - \frac{1}{2N}\right)^{j-1} \frac{1}{2N}$$

$$\underline{n \geq k > 2}$$

$$P(\text{don't coalesce in 1 gen}) =$$

$$\frac{(2N-1)}{2N} \frac{(2N-2)}{2N} \dots \frac{(2N-k+1)}{2N}$$

$$= \prod_{i=1}^{k-1} \left(1 - \frac{i}{2N}\right) = 1 - \sum_{i=1}^{k-1} \frac{i}{2N} + O\left(\frac{1}{N^2}\right)$$

$$= 1 - \binom{k}{2} \frac{1}{2N} + O\left(\frac{1}{N^2}\right)$$

Time until the first coalescence

For $n \geq k > 2$, from last slide:

$$P(\text{don't coalesce in 1 gen}) \approx 1 - \binom{k}{2} \frac{1}{2N}$$

so, with assumption of at most 1 coalescence in a generation:

$$P(\text{coalesce in 1 gen}) = \binom{k}{2} \frac{1}{2N}$$

and, for T_k^* time until first coalescent event:

$$P(T_k^* = j \text{ generations}) = \left\{ 1 - \binom{k}{2} \frac{1}{2N} \right\}^{j-1} \binom{k}{2} \frac{1}{2N}$$

The continuous time coalescent

While the Wright-Fisher works with discrete generations, it is computationally beneficial to work with continuous time

$$P(T_k^* > j \text{ gens}) = \left(1 - \frac{\binom{k}{2}}{2N}\right)^j$$

Scale by the average time for two lineages to find a common ancestor, $t = \frac{j}{2N}$:

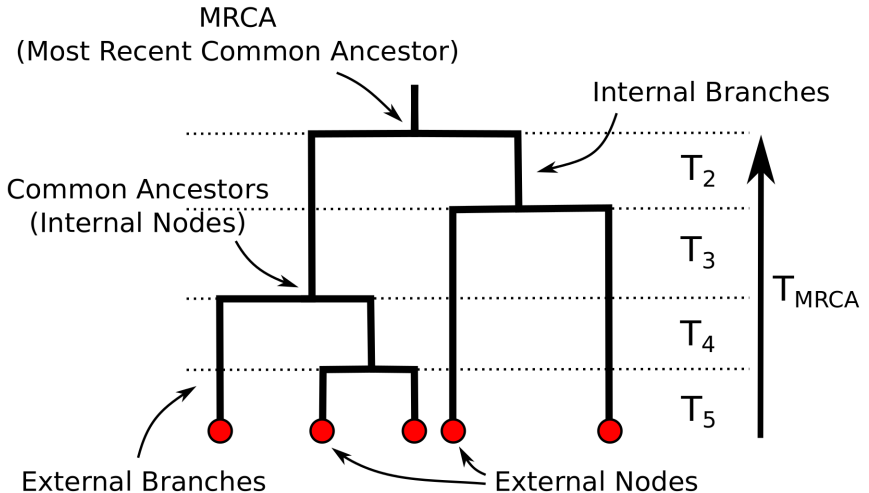
$$P(T_k > t) = \left(1 - \frac{\binom{k}{2}}{2N}\right)^{2Nt} \rightarrow \lim_{2N \rightarrow \infty} \left(1 - \frac{\binom{k}{2}}{2N}\right)^{2Nt} = e^{-\binom{k}{2}t}$$

T_k is an exponential random variable, with rate $\binom{k}{2}$

The basic coalescent algorithm

- 1) Start with $k = n$
- 2) Simulate waiting time T_k to next event, $T_k \sim \text{Exp}\left(\binom{k}{2}\right)$
- 3) Choose pair of lineages (i, j) uniformly among $\binom{k}{2}$ possible pairs
- 4) Merge i and j into single lineage, and decrease sample size by one, $k \rightarrow k - 1$
- 5) If $k \geq 2$ go to **2)**, otherwise stop

Anatomy of a coalescent genealogy



Each coalescent time is independent of all other times

Majority of $E(T_{MRCA})$ is $E(T_2)$

Variance in times is small for big k and big for small k

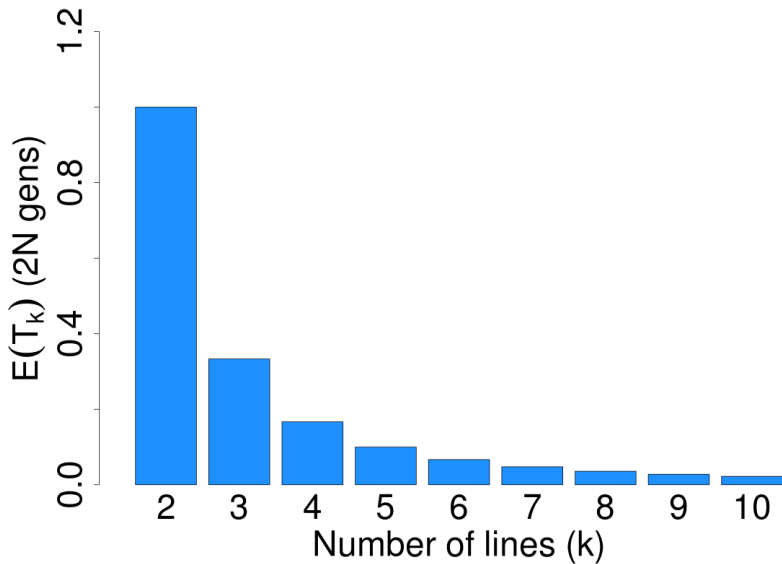
$$E(T_k) = \binom{k}{2} = \frac{2}{k(k-1)}$$

$$Var(T_k) = \left(\frac{2}{k(k-1)} \right)^2$$

$$E(T_2) = 1 = 2N \text{ gens}$$

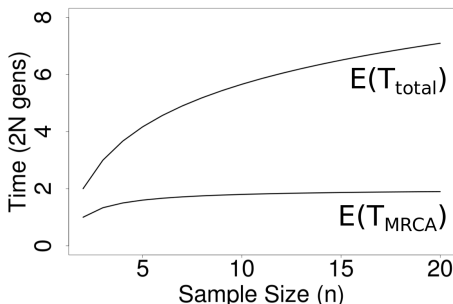
$$Var(T_2) = 1$$

Expected coalescent times



Total tree length is the summed length of all the branches on the tree

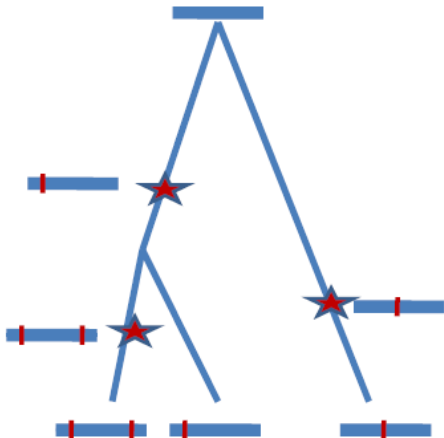
$$T_{total} = \sum_{k=2}^n kT_k$$
$$E(T_{total}) = \sum_{k=2}^n kE(T_k) = \sum_{k=2}^n k \frac{2}{k(k-1)} = \sum_{k=1}^{n-1} \frac{2}{k} \approx 2\ln(n-1)$$



The infinite sites mutation model

Infinite sites is a mutation model where every polymorphism is the result of a single mutation event. Each mutation creates a new polymorphic site

Infinite sites is a reasonable assumption for long genetic sequences with low mutation rates



The parameter θ is the **scaled mutation rate** (also called the population mutation rate)

Usually defined as $\theta = 4N\mu b$, where $2N$ is the population size, μ is the per base per gen mutation rate, and b is the length in bases of the locus

θ can be interpreted as the expected number of mutations separating a sample of 2 sequences
(2 branches $\times 2N$ gen $\times b$ bases $\times \mu$ mutations/base*gen)

Mutations are modeled as a Poisson process

Along a coalescent genealogy, mutations are modeled as Poisson distributed with rate $\frac{\theta}{2}$ per **coalescent time unit** ($2N$ generations)

The number of mutations, x during time t :

$$P(X = x|t) = \frac{\left(\frac{\theta t}{2}\right)^x}{x!} e^{-\frac{\theta t}{2}} \quad E(X|t) = \text{Var}(X|t) = \frac{\theta t}{2}$$

With this definition the time between mutation events is exponentially distributed with rate $\frac{\theta}{2}$

Adding mutations to the coalescent algorithm

Previous algorithm

- 1) Start with $k = n$
- 2) Simulate waiting time T_k to next event, $T_k \sim \text{Exp}(\binom{k}{2})$
- 3) Choose pair of lineages (i, j) uniformly among $\binom{k}{2}$ possible pairs
- 4) Merge i and j into single lineage, and decrease sample size by one $k \rightarrow k - 1$
- 5) If $k \geq 2$ go to 2), otherwise stop

6) For each branch along genealogy, with length ℓ

a) draw x mutations, $x \sim \text{Pois}(\frac{\theta\ell}{2})$

b) select location of each mutation along sequence uniformly

Generally, sequence is treated as having length 1, so mutation locations follow $\text{Uniform}(0,1)$

- ▶ Review working with the Linux command line
- ▶ Install the program *ms*
- ▶ Run a simple coalescent simulation and explore the output