Populations in statistical genetics

What are they, and how can we infer them from whole genome data?

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Work with:



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Overview

- Some views of 'population'
- A statistical definition
- Generative approaches
- ▶ Inference from sparse weakly linked data (STRUCTURE)
- ► Inference from dense linked data (fineSTRUCTURE)
- Selected results

What is a population in general?

Like species, populations are elusive objects.

- ► Definitions are a function of knowledge and application
- Some definitions Need not be biologically meaningful
- Example: Individuals found on same island
- Example: Sample location due to clustered sampling procedure
- Example: Disease status for case/control studies
- ▶ In statistics: we generalize from samples to a population
- ► Share: individuals are equivalent under some measure

But do populations really exist? Does it matter if they are useful?

Diversion: some fuzzy definitions of species

Are definitions of species analogous to definitions of populations?

- ► **Typological species:** Organisms that share the same set of phenotypes
- ► **Ecological species:** Organisms that compete for the same environment
- Phylogenetic species: Organisms with a single common ancestor
- ▶ **Biological species:** Organisms that can share DNA via sexual reproduction



What is a population? Motivation from genetics

We are interested in non-subjective definitions of populations.

- Individuals have equivalent genetic ancestry
- ▶ This depends on the data available ...
- ... and the model used
- Knowledge driven definition
- Requires defining equivalent!

The birds and the bees (vs the plants and bacteria)

- Sex dramatically affects genetic transmission
- Different population concept required
- Sexual species
 - Generalise the Biological Species concept
 - Random mating within a population
 - Try to detect deviations from random mating to identify populations

Asexual species

- Generalise the Ecological Species concept
- Neutral competition within a population
- Try to detect deviations from neutrality to identify populations
- Not examined further in this talk

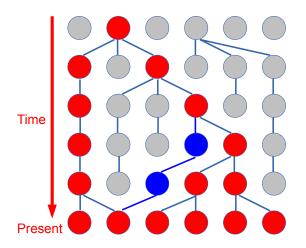
Top down approaches

Start from a theoretical generative model of reproduction. For example:

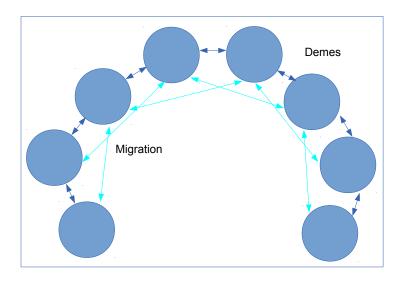
- Individuals move between populations via migration
- Discrete generations (for simplicity)
- Each generation randomly chooses parent(s) within a population

Population structure: individuals migrate between populations but mate randomly within each

Ancestry Process



Populations as Demes



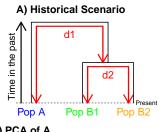
Ammerman and Cavalli-Sforza, 1984 The Neolithic Transition and the Genetics of Populations in Europe.

Diversion: Principal Components Analysis

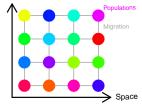
- ▶ PCA is widely used in genetics, and helpful for visualisation
- ▶ It does not make any explicit modelling assumptions...
- BUT to interpret the output, you do!
 - 'Just' rotating the data
 - Similar individuals tend to be close
 - Differences shared by many individuals tend to appear first
 - ► Each component describes a different direction of variation
 - ▶ If we are lucky, these correspond to real shared drift events
 - ► There is no unique interpretation of PCA (see McVean 2009)

See Lawson & Falush 2012 for details.

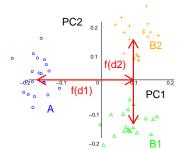
Example



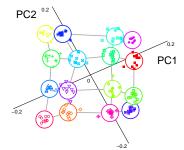
B) Spatial Scenario



C) PCA of A



D) PCA of B

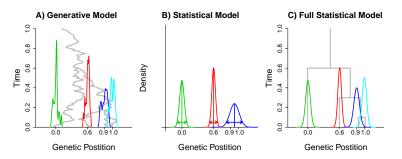


Bottom up approaches

Start from a definition of population as 'equivalent' individuals

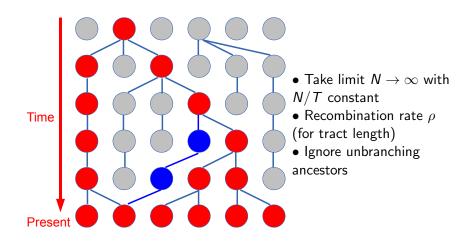
- Within a population, individuals are randomly mating
- Small samples of large populations: individuals are approximately independent
- Smaller populations: relationships must be accounted for
- What does random mating mean for the data?
- How are populations related?

Top down vs Bottom up



- ► A) Generative approaches are best in theory, if we can make the model match reality
- But, hard to use in practice how to do inference?
- ▶ B) Bottom up approaches are approximate might lose power
- C) But can be refined until they are close to the generating process

Ancestry Process - Ancestral Recombination Graph



Structure model

Pritchard, Stephens & Donnelly 2000.

- Populations are large and well mixed
- ► SNPs D_{il} are unlinked*
- loci have some ancestral frequency

$$p_{0I} \sim P(\cdot)$$

▶ Population k has frequency p_{kl} drifted from ancestral p_{0l} **

$$p_{kl} \sim \text{Dirichlet}(p_{0l})$$

▶ Individual *i* is in population *k* if $Q_{ik} = 1$, assigned by

$$\prod_{l} P(D_{il}|p_{Q_{ik},l})$$

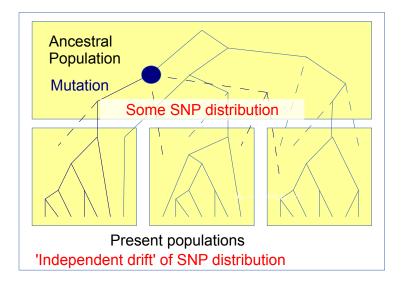
▶ Individuals in the same population are *exchangeable* with respect to the SNP frequencies



^{*} Solutions to this have been explored (computationally inconvenient)

^{**} Valid approximation, originally derived by Wright

Single SNP with populations



Scaling STRUCTURE

STRUCTURE approach has a parameter for every SNP. But:

Assuming that drift is weak, $p_{0l} = E(D_{\cdot l})$ and:

$$p(p_{kl}) \sim N(p_{0l}, p_{0l}(1-p_{0l}))$$

Probability SNP / is shared not by chance:

$$X_{ijl} = D_{il}D_{jl}/p_{0l} + (1 - D_{il})(1 - D_{jl})/(1 - p_{0l})$$

Invoke the Central limit theorem:

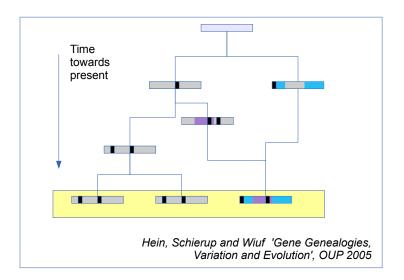
$$X_{ij} = \sum_{l=1}^{L} X_{ijl} \sim N(\mu_{ij}, \sigma_{ij}^2)$$

- ► This is the Coancestry Matrix
- ▶ It is a sufficient statistic for p(D|Q)
- \blacktriangleright μ and σ known and same for all individuals in a population
- Exchangability again!

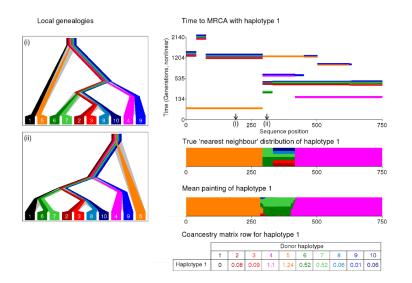
Now lets think about linkage...



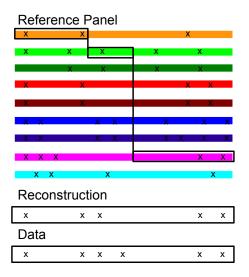
Ancestral Recombination Graph



ChromoPainter



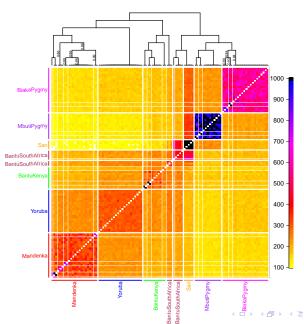
ChromoPainter Hidden Markov Model



FineSTRUCTURE model

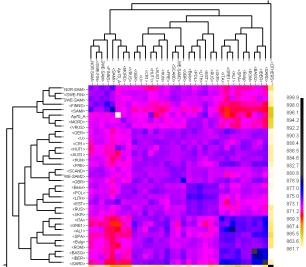
- ▶ Each population has a characteristic rate P_{ab} of sharing 'chunks' with each other population
- Individuals are again exchangeable within populations
- ▶ Each recombination event has probability $\hat{P}_{ab} = P_{ab}/\hat{n}_b$ when coming from an individual in population b into population a
- Dirichlet Process prior on the parameters P_a.
- ▶ We integrate out *P*, leaving no population level parameters
- ▶ We can put a meaningful prior on the variation of *P* between populations, and the number of populations
- In practice these details don't matter much

Example: Africa HGDP



Ancient Genomes: fennoscandia.blogspot.co.uk

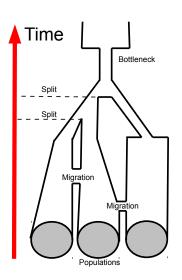
Ajv70 Gotland hunter gatherer



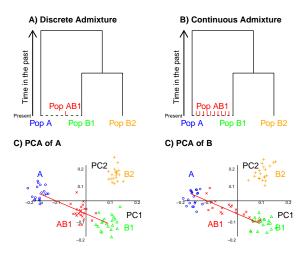
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History of populations

- We have assumed that we observe individuals from real populations
- Populations differ by genetic drift
- This works, even if there is historical migration, provided that the mixture fractions are equal (exchangeability)
- Real individuals are related by a combination of drift and admixture
- 'Ancestral population graph'

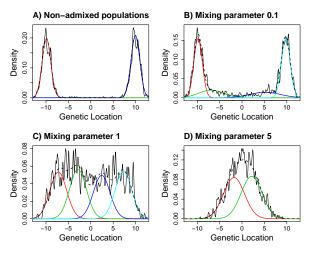


Admixture



Admixture describes mixture without drift.

Admixture



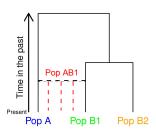
Continuous admixture is a problem.

Admixture models

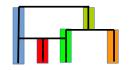
- STRUCTURE can infer 'pure' populations from admixed populations
- ▶ Population assignment Q_{ik} sums to 1 without requiring a single element
- Interpretation: Observed individuals are mixtures of pure populations, without drift
- How can we tell apart:
 - Large drift, mixed by admixture?
 - small drift without admixture?
- ▶ Solution: SNPs have fixed in some populations $p_{ik} = 0$ (or 1)
- FineSTRUCTURE cannot use this description, as we've integrated out these details

Drift model

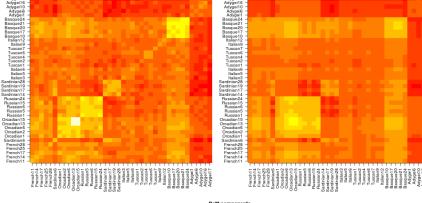
- ► See each drift event as independent
- Population assignment Q_{ik} now takes any value
- 'Amount' of each drift event retained by individual i
- Reconstructs the coancestry matrix
- Requires a strong prior to obtain a unique solution

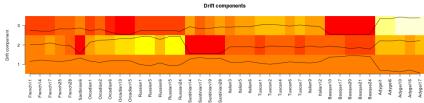


Drift Components









Projects

- Siberian cold adaptation (Alexia & Toomas, in review)
- GlobeTrotter (Hellenthal et al: very accurate admixture dating using ChromoPainter, to appear in Science)
- Peopling of the British Isles (in review)
- UK10K, ALSPAC (4K whole genomes, use in genome-wide association studies)
- Highly recombining asexuals (fungus, bacteria)
- Model improvements:
 - Relatedness
 - Admixture/history of populations
 - Complete recoding for usability
 - Efficient computation (fastFineSTRUCTURE)

See www.paintmychromosomes.com

- ▶ Lawson, Hellenthal, Myers & Falush, 'Inference of population structure using dense haplotype data', 2012. PLoS Genetics.
- ▶ Lawson & Falush 'Similarity matrices and clustering algorithms for population identification using genetic data', 2012. ARHG.
- ▶ Lawson 2014 'Populations in statistical genetics modelling and inference', in 'Populations in the Human Sciences', Eds. Kreager, Capelli, Ulijaszek & Winney.



Garrett Hellenthal UCL



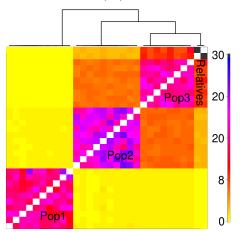
Simon Myers Oxford



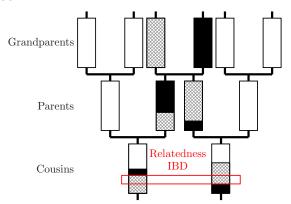
Daniel Falush Max Planck, Leipzig

Relatedness

What if individuals within a population are related?

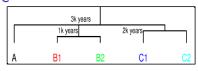


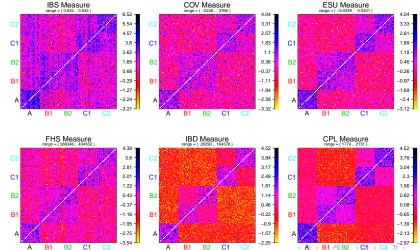
Relatedness



- Samples are cousins
- ▶ This is very easy to tell from their tract length distribution
- Excluding these tracts, we sample from the population distribution of chunks
- Multiple ordering model in progress

Choice of measure





Choice of measure

