

F-statistics and admixture

Fernando Racimo

Adelaide, January 2018

Today

- F2 statistics
- Outgroup F3 statistics
- Admixture F3 statistics
- F4 and D-statistics
- qpWave / qpAdm

An excellent resource!

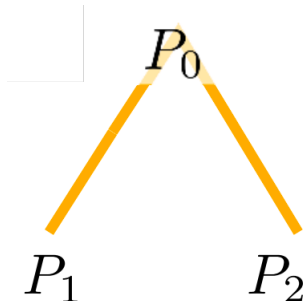
Admixture, Population Structure and F -statistics

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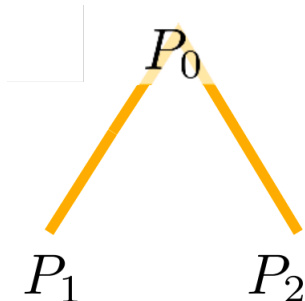
F_2 statistics

- Let's imagine we have two populations: P_1 and P_2
- At a particular site, the allele frequency of a (randomly chosen) allele is denoted as p
- $F_2(P_1, P_2) = E[(p_1 - p_2)^2]$



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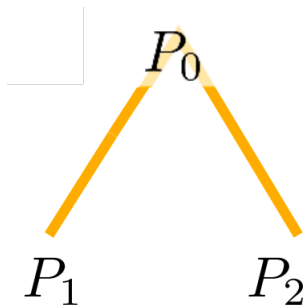


- $F_2(P_1, P_2) = E[(p_1 - p_2)^2]$
- $E[]$ denotes an expectation.
- This expectation is over multiple **independent runs of the evolutionary process** of an allele. In practice, we don't have multiple runs.
- However, we can look at **multiple sites** across the genome
- Sites are not exactly independent (due to linkage), but we'll later see ways to account for this problem

- $F_2(P_1, P_2)$ can also be interpreted as a variance
- $\text{Var}[p_1 - p_2] = E[(p_1 - p_2)^2] - (E[p_1 - p_2])^2$
- But $E[p_1 - p_2] = E[p_1] - E[p_2] = E[p_0] - E[p_0] = 0$
- So $\text{Var}[p_1 - p_2] = E[(p_1 - p_2)^2]$, and therefore:
- **$F_2(\mathbf{P1}, \mathbf{P2}) = \text{Var}[\mathbf{p1} - \mathbf{p2}]$**

F_2 additivity

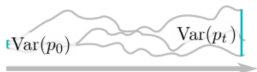
- If we consider a common ancestral population P_0 , then:
- $F_2(P_1, P_2) = F_2(P_1, P_0) + F_2(P_2, P_0)$



F_2 as a measure of genetic drift

- If we compute an F_2 statistic between an ancestral and a descendant population, we can consider an F_2 statistic to be:
 - A measure of the increase in allele frequency variance over time
 - A measure of the decrease in heterozygosity over time
 - A measure of the probability that two gene copies in the descendant population originate from a single copy in the ancestral population
 - In essence, a measure of **genetic drift** (time scaled by population size) or “population inbreeding”

A $F_2 = \text{Var}(p_t) - \text{Var}(p_0)$



B $F_2 = \frac{\mathbb{E}H_0 - \mathbb{E}H_t}{2}$



C $F_2 = \frac{1}{2}f\mathbb{E}H_0$

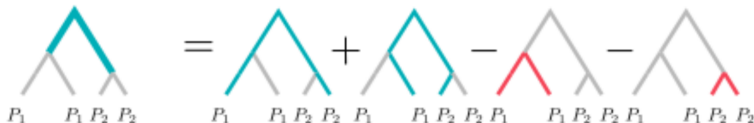


A coalescent interpretation

A. Equation

$$2F_2(P_1, P_2) = \theta \left(\mathbb{E}T_{12} + \mathbb{E}T_{12} - \mathbb{E}T_{11} - \mathbb{E}T_{22} \right)$$

B. Concordant genealogy

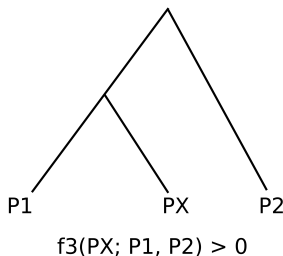


A coalescent interpretation

- The estimator for F_2 can be written as a function of Tajima's estimator for θ (π)
- $F_2 = \theta(E[T_{12}] - \frac{E[T_{11}] + E[T_{22}]}{2})$
- We know (from the first class) that θT is the expected number of differences between two sequences separated by time T , under the infinite sites model
- We also know (from the first class) that, an estimator for the expected number of differences for two sequences is π
- $\theta \hat{T}_{12} = \pi_{12}$
- $\theta \hat{T}_{11} = \pi_{11}$
- $\theta \hat{T}_{22} = \pi_{22}$
- $\hat{F}_2 = \pi_{12} - \frac{\pi_{11} + \pi_{22}}{2}$

F_3 statistics

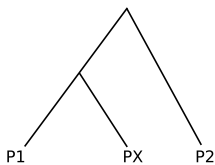
- F_3 statistics can be used to determine if a population X is admixed¹
- $F_3(P_X; P_1, P_2) = E[(p_X - p_1)(p_X - p_2)]$
- They can also be expressed in terms of F_2 statistics
- $F_3(P_X; P_1, P_2) = \frac{1}{2}(F_2(p_X, p_1) + F_2(p_X, p_2) - F_2(p_1, p_2))$
- Note that if the populations can be described in terms of a tree, then $F_2(p_1, p_2) \leq F_2(p_X, p_1) + F_2(p_X, p_2)$



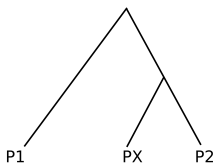
¹Reich et al. (2009)

Admixture F_3 statistics

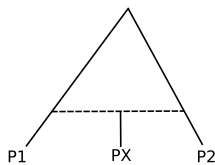
- One application of F_3 is to detect violations in “treeness” (admixture or populations structure)
- If $F_2(p1, p2) > F_2(pX, p1) + F_2(pX, p2)$, then a tree is not a good descriptor of the populations, and $F_3(P_X; P_1, P_2) < 0$
- Run F_3 statistics a Test population in the first position
- If the demographic history (with respect to 2 other populations) can be described as a tree, then $F_3 > 0$
- Violations in treeness result in $F_3 < 0$



$$f_3(PX; P1, P2) > 0$$



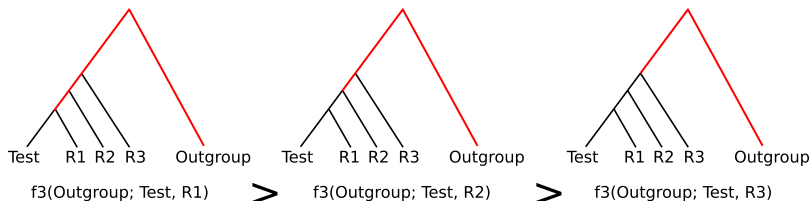
$$f_3(PX; P1, P2) > 0$$



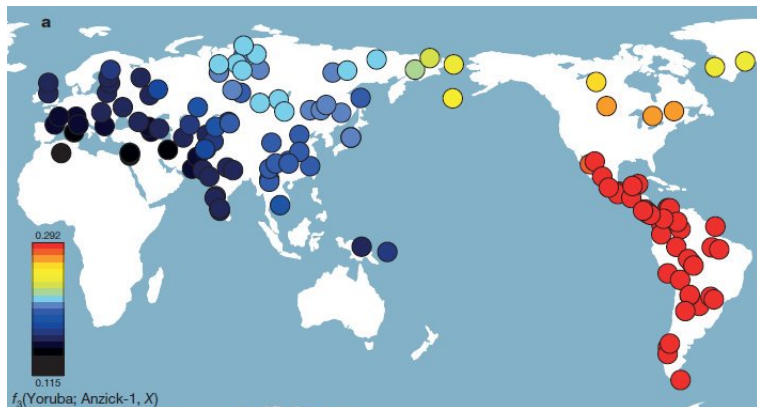
$$f_3(PX; P1, P2) < 0$$

Outgroup F_3 statistics

- Another application of F_3 is to determine which populations are closer (have more of a shared history) to a Test population
- Run F_3 statistics with an Outgroup in the first position, followed by a Test population and several candidate Reference populations
- F_3 can be interpreted as the shared drift-path between a Test + Reference X and Test + Outgroup
- The more shared history between Test and Reference X, the larger the F_3 statistic



Outgroup F_3 statistics



- F_4 statistics can be used to detect admixture and estimate admixture parameters.
- $F_4(P_1, P_2; P_3, P_4) = E[(p_1 - p_2)(p_3 - p_4)]$
- They can also be expressed in terms of F_2 statistics:
- $F_4(P_1, P_2; P_3, P_4) = \frac{1}{2}(F_2(p_1, p_4) + F_2(p_2, p_3) - F_2(p_1, p_3) - F_2(p_2, p_4))$

F_4 statistics can be used to detect admixture

- A scaled version of the F_4 statistic (D) has been widely used to determine if admixture occurred in a population tree
- We'll describe D in detail in a few slides...

F-statistics vs. F_{ST}

- An F-statistic can be thought of as a covariance (or a linear combination of covariances) between population allele frequencies
 - It ranges between $-\infty$ and ∞
 - Easier to work with mathematically
 - It is additive: $F_2(P_1, P_2) = F_2(P_1, P_0) + F_2(P_2, P_0)$
 - Value **depends** on heterozygosity in the population
 - Highly used in models involving well-defined splits and admixture events
- F_{ST} can be thought of as an “absolute correlation” between population allele frequencies
 - It ranges between 0 (panmixia) and 1 (complete divergence).
 - It is not additive: $F_{ST}(1, 2) \neq F_{ST}(1, 0) + F_{ST}(2, 0)$
 - Value does not depend on heterozygosity in the population
 - Highly used in stepping stone / migration models

Different models, different interpretations

- F-statistics will have different interpretations depending on underlying model
- Admixture graphs may not necessarily be the best descriptor of a biological system!








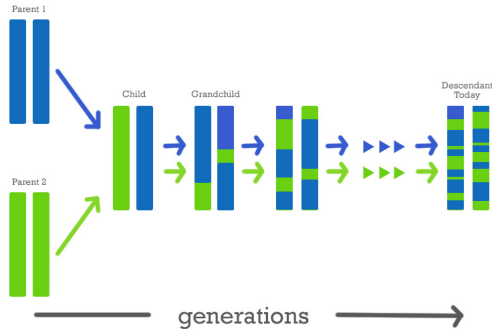
Model		$F_3(P_X; P_1, P_2)$	$F_4(P_1; P_X; P_2, P_3)$	Parameters
Panmictic		0	0	
Admixture Graph		$t_1 - 2\alpha(1-\alpha) \times (1-c_a)t_r$	$(1-\alpha)(t_2 - t_1)$	α : admixture ratio; t_1 : admixture time; t_2 : merging time of P_2 and P_3 ; t_r : global ancestor
Island Model		$\frac{1}{M}$	0	M: Migration rate
Hierarchical Island Model		$\frac{n(d-1)}{M}$	0	M: Migration rate n: # of island d: # of demes per island
Stepping stone		$\frac{2}{7M}$	$-\frac{8}{7M}$	M: Migration rate between adjacent demes
Hierarchical stepping stone		$-\frac{0.06}{M}$	$\frac{14}{55M}$	M: Migration rate between adjacent demes
Serial founder model		t_x	0	t_x : time when P_X is first colonized

Figure 6. Expectations for F_3 and F_4 under select models.

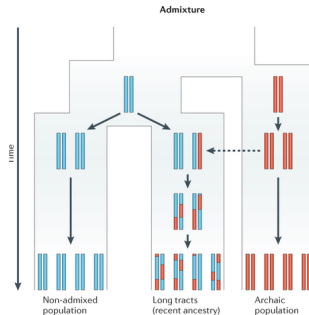
Admixture

- Admixture is the process by which two previously isolated populations interbreed.
- It results in the introduction of genetic material from a foreign source into a population.



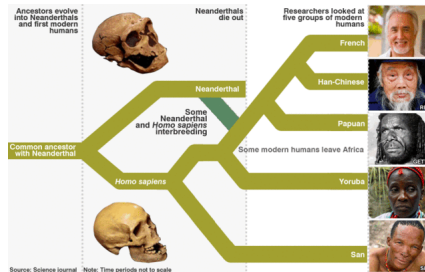
Admixture

- The signatures of admixture can be detected in the genomes of the descendants of the admixed individuals.



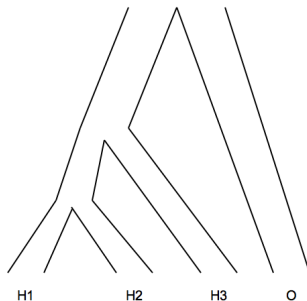
ABBA-BABA

- The ABBA-BABA test (or D-statistic) was developed to test for ancient gene flow between populations (Green et al. 2010, Durand et al. 2011, Patterson et al. 2012).
- Originally used as evidence for Neanderthal introgression into non-African modern humans (Green et al. 2010, Prufer et al. 2014).



ABBA-BABA: assumptions

- We need to have sequence data from 3 populations (H1, H2 and H3) and an outgroup (O).
- The population tree should be known.
- There has been no recurrent mutations (short time-scales).
- Null hypothesis: no gene flow between H3 and H1 or between H3 and H2 after their respective splits.



ABBA-BABA: test using individual genomes

- Look at all diallelic loci where:
 - O and H3 have different alleles (called A and B)
 - H1 and H2 have different alleles
 - In other words, we look for sites where:
 - $(H1, H2, H3, O) = (A, B, B, A)$
 - $(H1, H2, H3, O) = (B, A, B, A)$
 - For example, (C,T,T,C) or (A,T,A,T).

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- Calculate $D = \frac{\#ABBA - \#BABA}{\#ABBA + \#BABA}$

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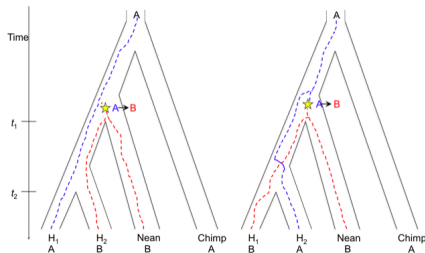
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- Calculate $D = \frac{\#ABBA - \#BABA}{\#ABBA + \#BABA}$
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- Test if D is significantly different from 0 (more on this in a second).
- If so, reject the null hypothesis of no gene flow.

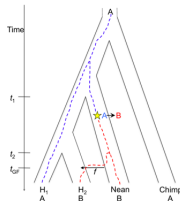
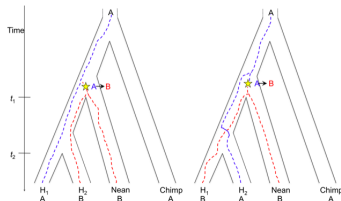
ABBA-BABA: rationale

- If there was no admixture, the only way to generate coalescent trees consistent with ABBA or BABA is by incomplete lineage sorting (ILS).
- In that case, we expect the same number of ABBA trees as of BABA trees.



ABBA-BABA: rationale

- However, if there was gene flow from H3 to H2, we expect an excess of ABBA trees.
- Therefore, $\#ABBA > \#BABA$ and $D > 0$.



ABBA-BABA: testing for significance

- Perform block jackknife to get an estimate, \hat{s} , of the standard deviation of D.
- Assume that under the null hypothesis, $D \sim \text{Normal}(0, \hat{s}^2)$
- Use this distribution to calculate a Z-score
- Reject null hypothesis if $|Z| > 3$

ABBA-BABA: calculated from low-coverage data

- Look at 1 individual from each of the H1, H2, H3 and O populations

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- Randomly sample 1 read from each individual in each site

ABBA-BABA: calculated from low-coverage data

- Look at 1 individual from each of the H1, H2, H3 and O populations
- Randomly sample 1 read from each individual in each site
- Practical problems:
 - Not using all the information we could theoretically use
 - Bias can occur if H1 and H2 were sequenced using different platforms.
 - Bias can occur if H1 and H2 have different error rates.
 - SNP chip data is improperly used (without accounting for ascertainment bias).
 - With ancient genomes, increased error rates at specific positions (e.g. C-to-T) can also generate problems.



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Powerful Inference with the D-statistic on Low-Coverage Whole-Genome Data

 Samuele Soraggi,  Carsten Wiuf,  Anders Albrechtsen

doi: <https://doi.org/10.1101/127852>

This article is a preprint and has not been peer-reviewed [what does this mean?].

Abstract

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ABBA-BABA: example (from Green et al. 2010)

- We're testing for admixture from Neanderthals into French, using San Africans as the non-admixed sister population.

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- $D(\text{San}, \text{French}, \text{Neanderthal}, \text{Chimpanzee}) = 0.047$
- After performing a block jackknife, $Z = 7.6$
- Conclusion: reject null hypothesis of no admixture.

ABBA-BABA: alternative formulation

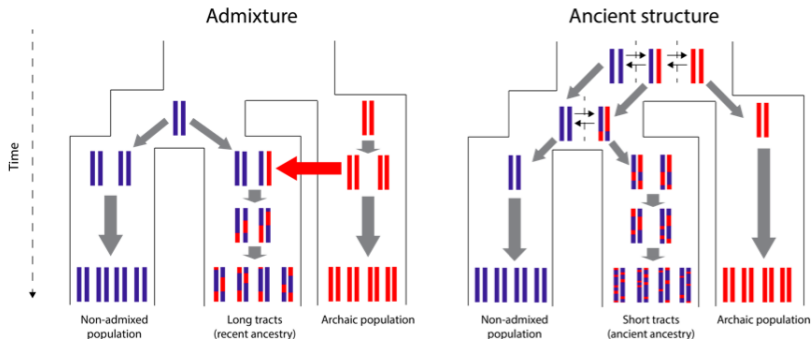
- Using sample allele frequencies (Durand et al. 2011)
- $$D = \frac{\sum_{i=1}^n [(1-\hat{p}_{i1})\hat{p}_{i2}\hat{p}_{i3}(1-\hat{p}_{i4}) - \hat{p}_{i1}(1-\hat{p}_{i2})\hat{p}_{i3}(1-\hat{p}_{i4})]}{\sum_{i=1}^n [(1-\hat{p}_{i1})\hat{p}_{i2}\hat{p}_{i3}(1-\hat{p}_{i4}) + \hat{p}_{i1}(1-\hat{p}_{i2})\hat{p}_{i3}(1-\hat{p}_{i4})]}$$
- \hat{p}_{i1} is the sample allele frequency in H1 at SNP i.
- \hat{p}_{i2} is the sample allele frequency in H2 at SNP i.
- \hat{p}_{i3} is the sample allele frequency in H3 at SNP i.
- \hat{p}_{i4} is the sample allele frequency in O at SNP i.

ABBA-BABA: caveats

- The value of D is not the same as the admixture rate!
- D depends on both the admixture rate AND the split times between the populations.
- Should not be deployed locally: ILS can generate local regions with $D \neq 0$.
- A genome-wide value of D significantly different from 0 could also be caused by ancestral population structure.

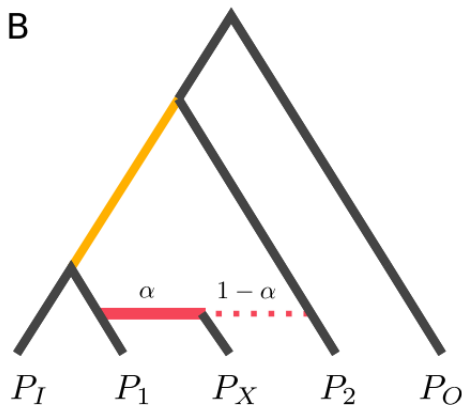
ABBA-BABA: caveats

- Important to find admixture tracts with lengths consistent with introgression.
- Hard problem: requires probabilistic models like HMMs.

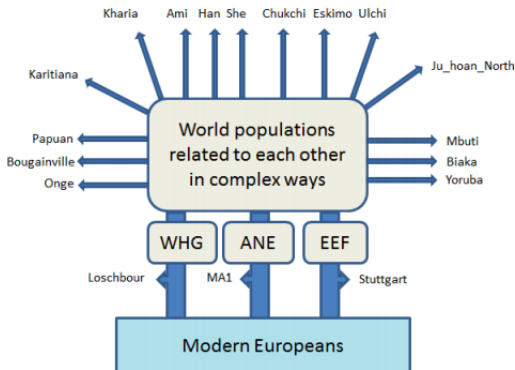


F_4 statistics can be used to estimate admixture proportions

- Assuming admixture occurred, F_4 statistics can be used to estimate the **amount** of admixture
- $\alpha = \frac{F_4(P_O, P_I; P_X, P_1)}{F_4(P_O, P_I; P_2, P_1)}$

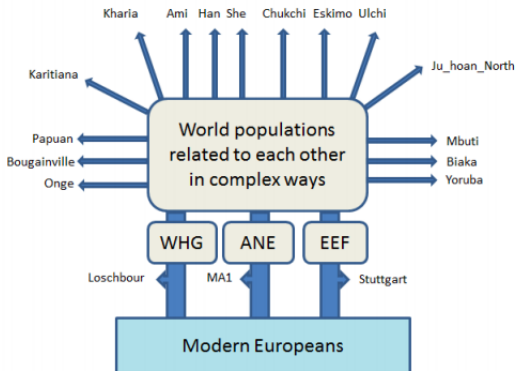


- The qpWave / qpAdm methodology² is a way to model admixture without detailed phylogenetic modeling
- This was originally used to argue for at least 3 highly-differentiated streams of ancestry contributing to present-day European genomes

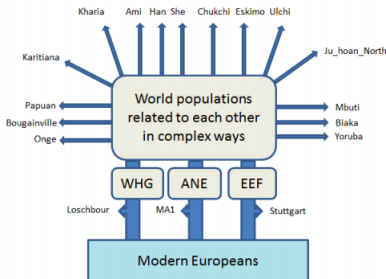


²Lazaridis et al. (2014, 2016)

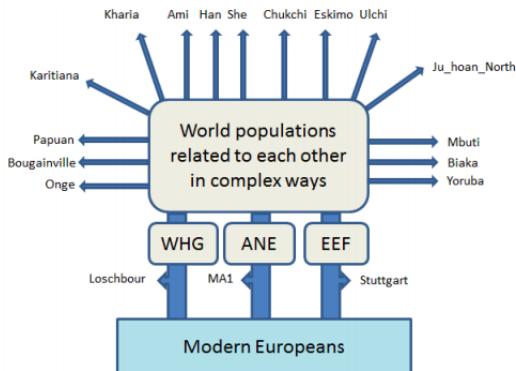
- We need:
 - A) A Test population
 - B) A set of Outgroup populations
 - C) A set of Reference populations that are clades (with respect to the Outgroups) of populations potentially contributing ancestry to the Test



- We can write F4 statistics for the Test as a weighted sum of N F4 statistics for the Reference populations
- $f_4(\text{Test}, O_1; O_2, O_3) = \sum_{i=1}^N \alpha_i f_4(\text{Ref}_i, O_1; O_2, O_3)$
- Given m Outgroups, there are $m \binom{m}{2}$ equations of the above form
- We can use regression to fit the mixture coefficients with the mixture coefficients (α_i) by regression



- qpAdm is a program used to find the best-fitting admixture coefficients under this framework



- qpWave is a program used to find whether the Reference + Test populations can be modeled as being descended from as few as X source populations (that are differentially related to the Outgroups)
- Typically one runs qpWave before qpAdm

