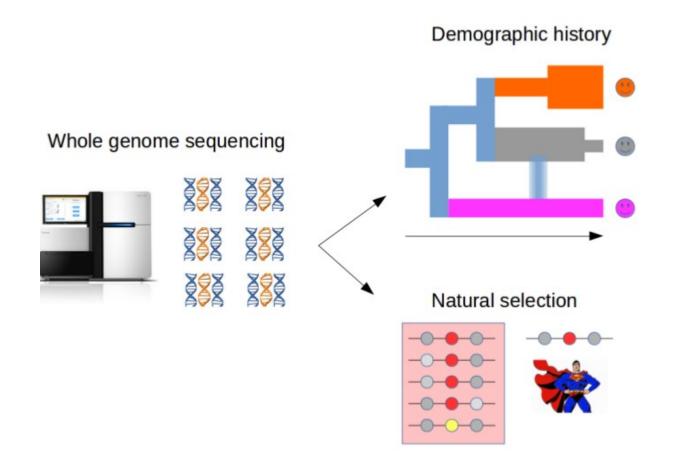
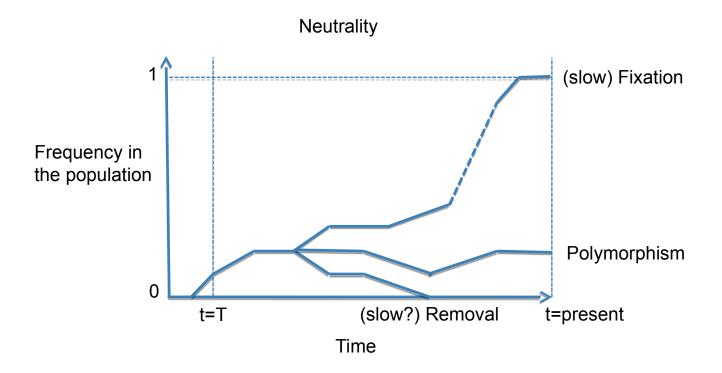
Detecting natural selection from genomic data (part 2)

Matteo Fumagalli

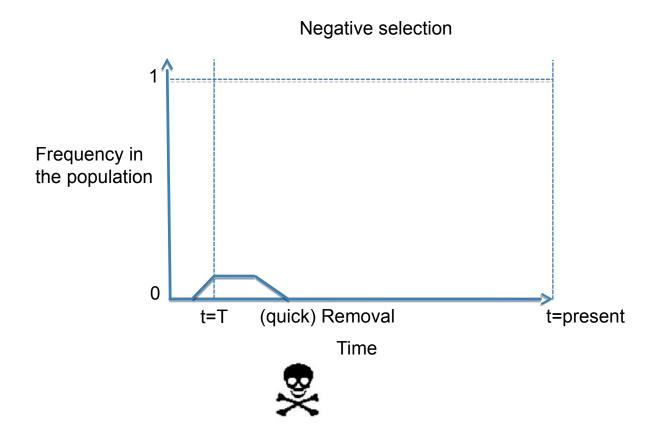
Motivation



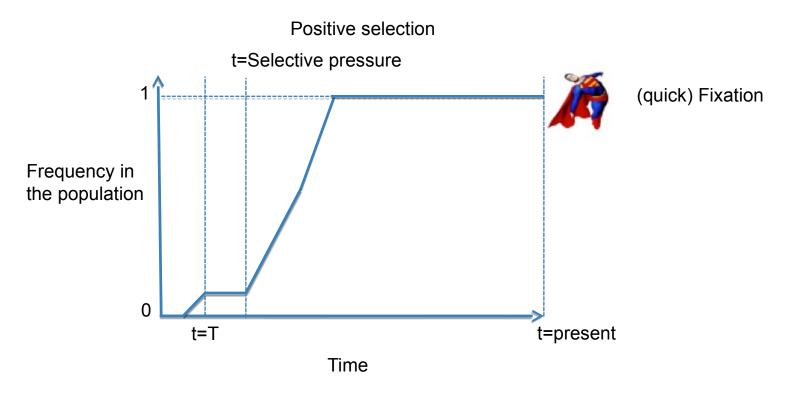
Heritable traits that increase the fitness of the become more common.



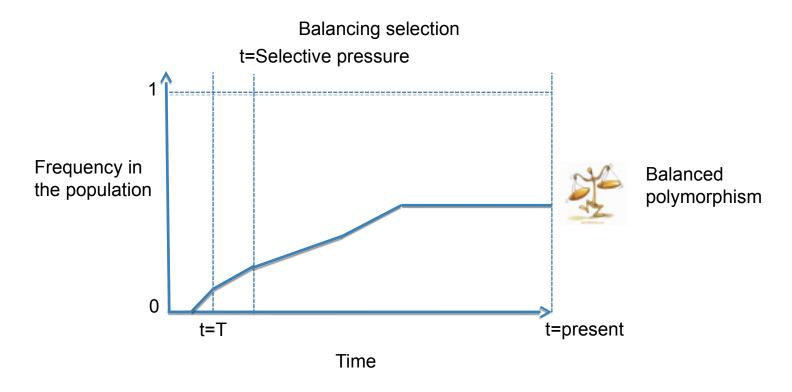
Heritable traits that increase the fitness of the become more common.



Heritable traits that increase the fitness of the become more common.

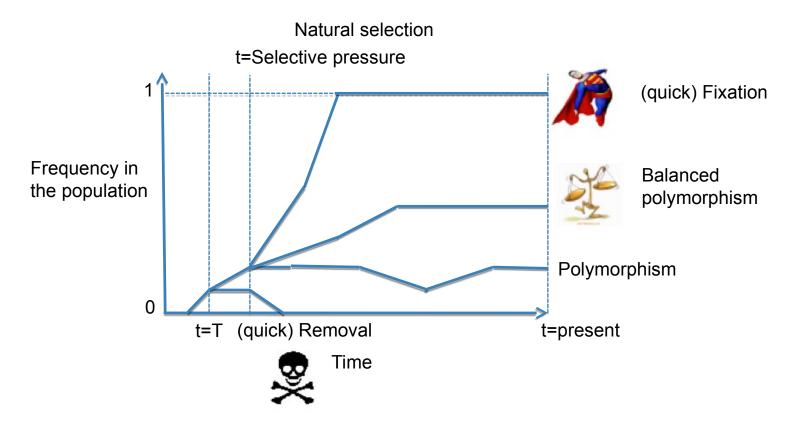


Heritable traits that increase the fitness of the become more common.



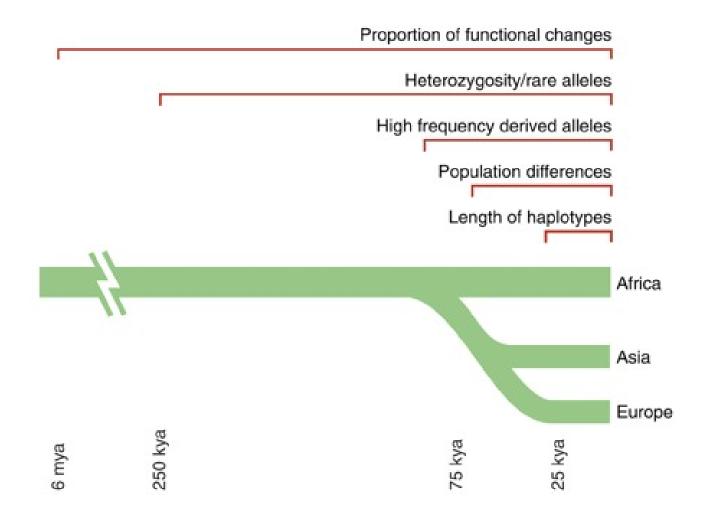
Heritable traits that increase the fitness of the become more common.

1) Mutations arise randomly and evolve according to their effect on the fitness of the carrier

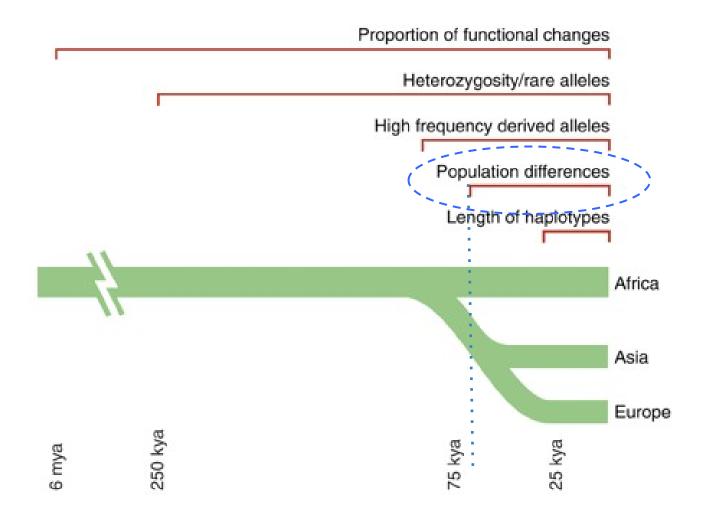


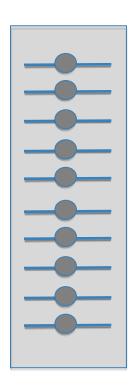
2) Sites targeted by natural selection are likely to harbour functionality

Methods to infer recent selection

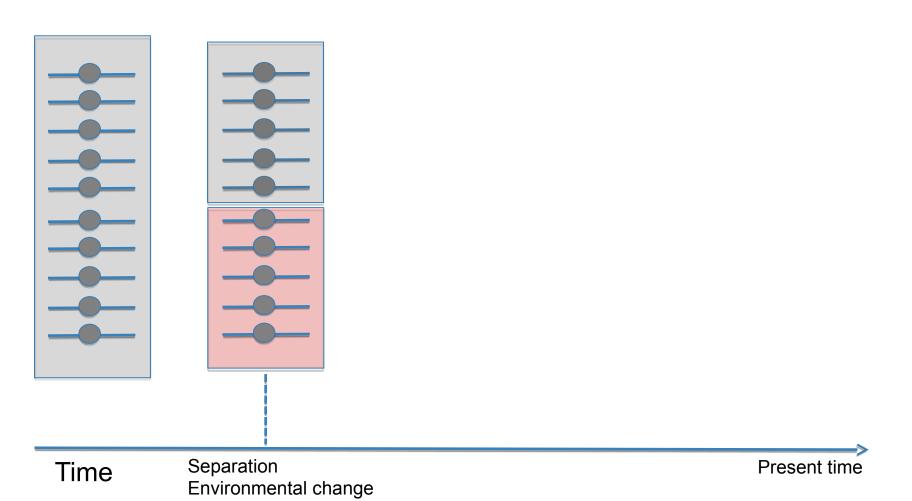


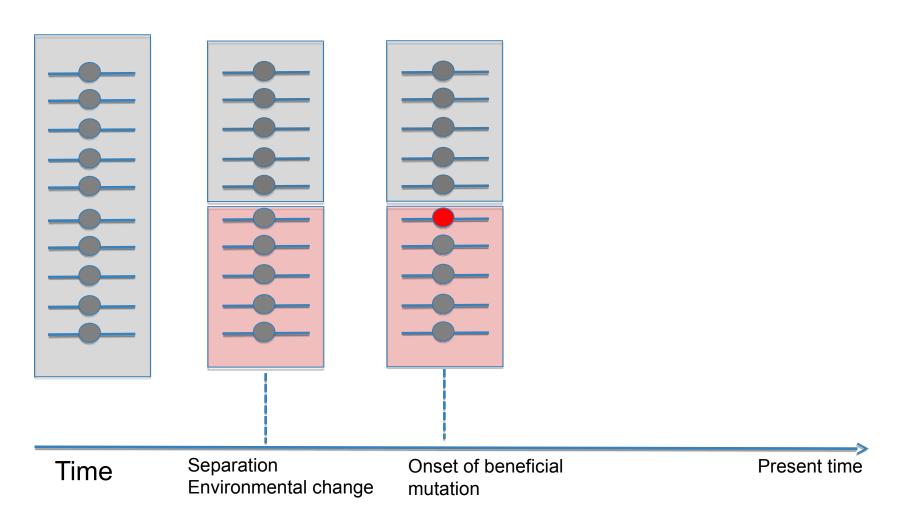
Methods to infer recent selection

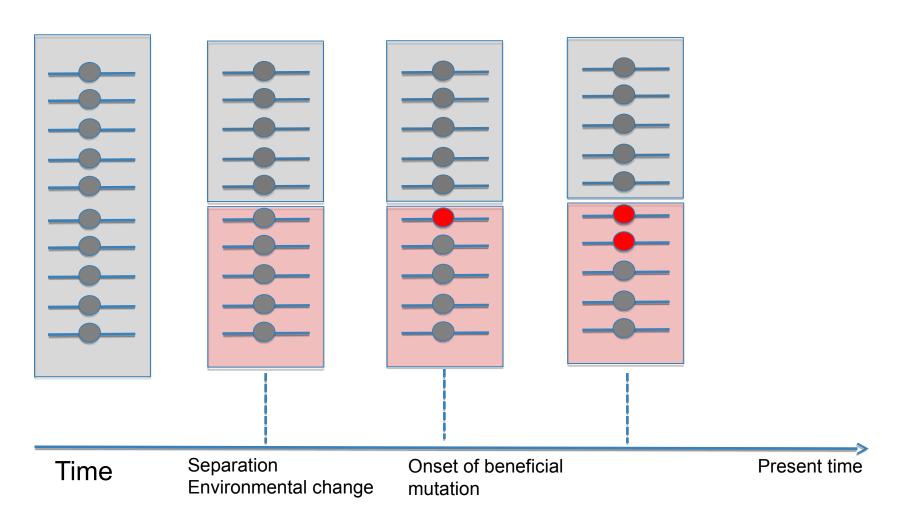


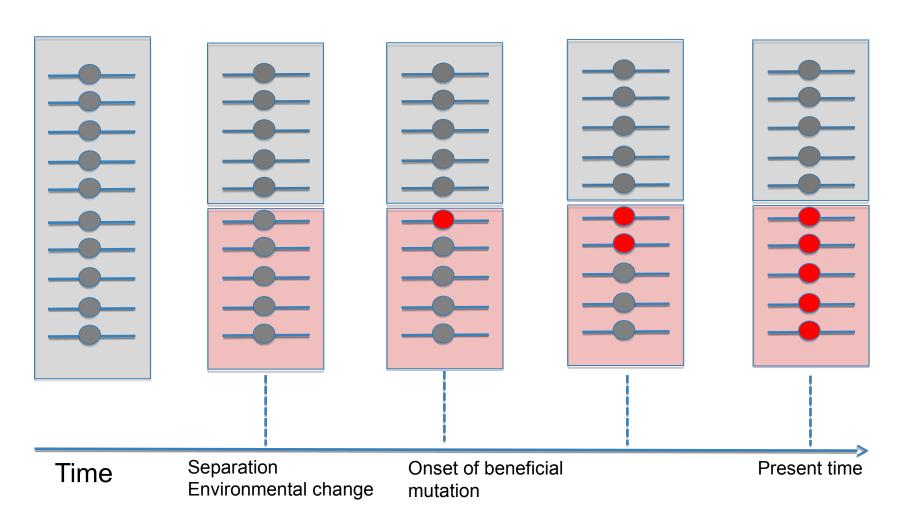


Time Present time









F_{ST}

Common measure for quantifying population subdivision.

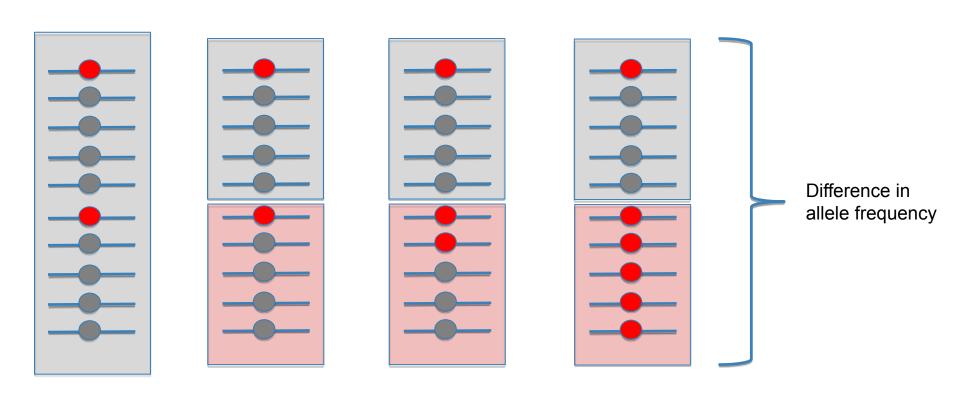
$$F_{ST} = H_B / (H_W + H_B)$$

H_B: between populations

H_w: average within populations

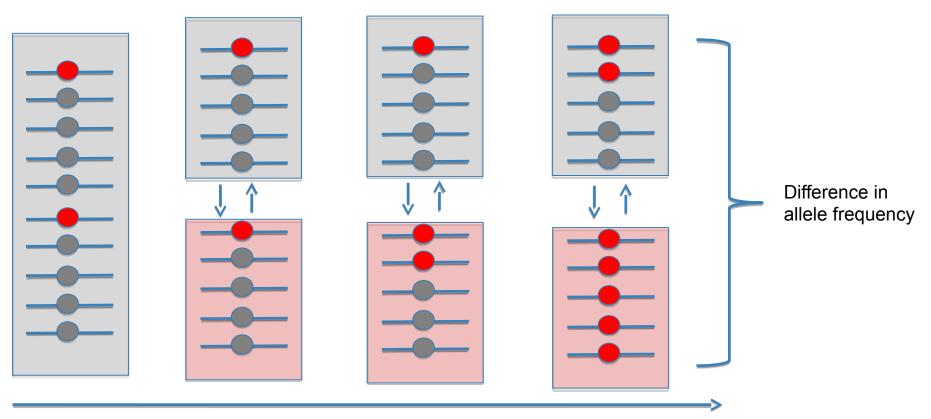
- \rightarrow if $H_W << H_B 0$ then $F_{ST} \sim 1$
- \rightarrow if H_B=0 then F_{ST}=0

From standing variation



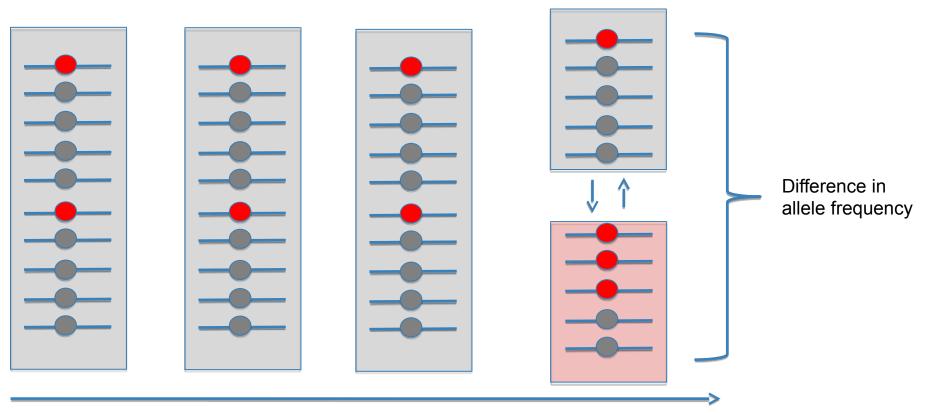
Time

With migration



Time

With recent divergence

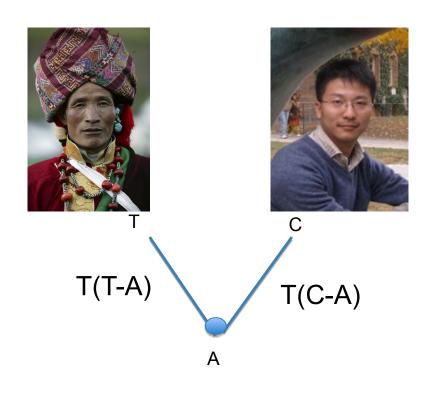


Time



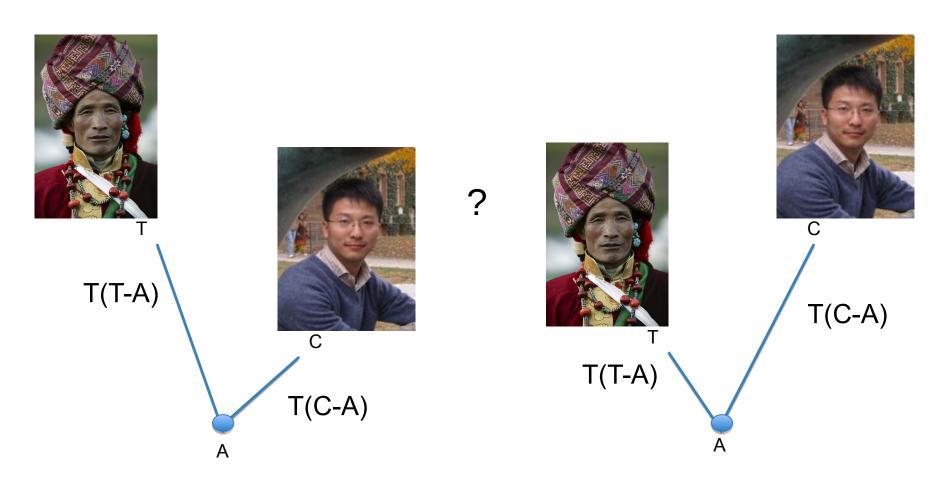


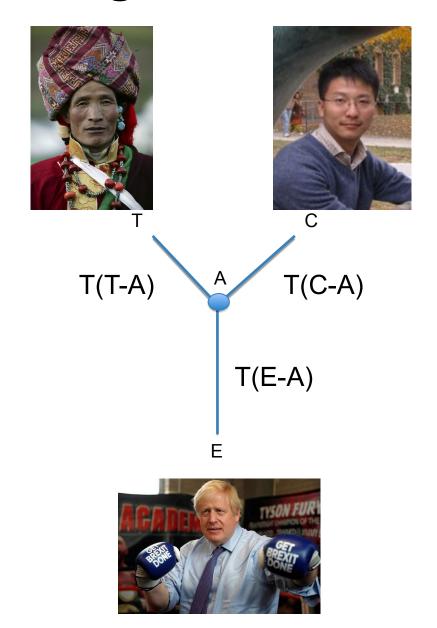
 $F_{ST}(T-C)$

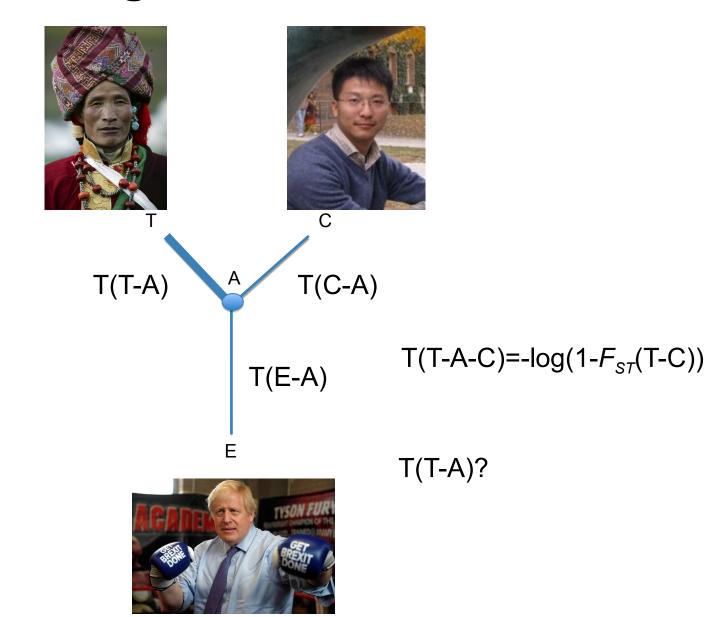


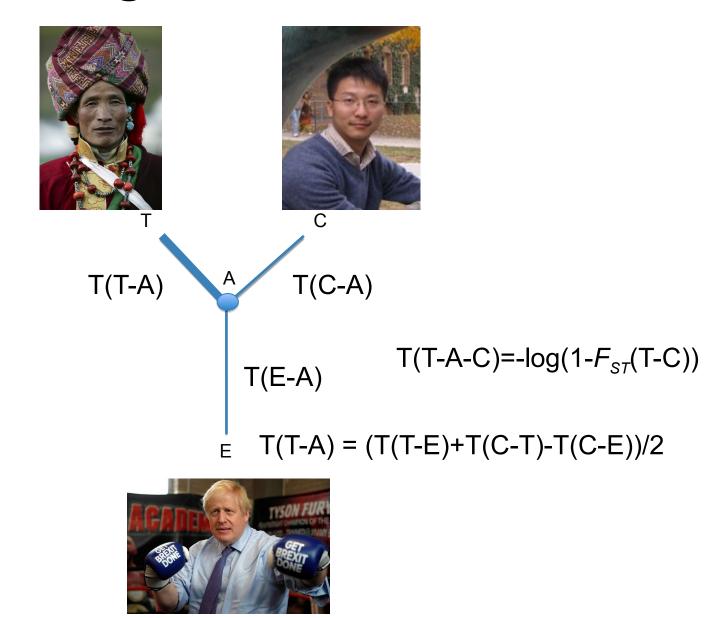
$$F_{ST}(T-C)\sim T(T-A-C)$$

 $F_{ST}(T-C)\sim T(T-A-C)$

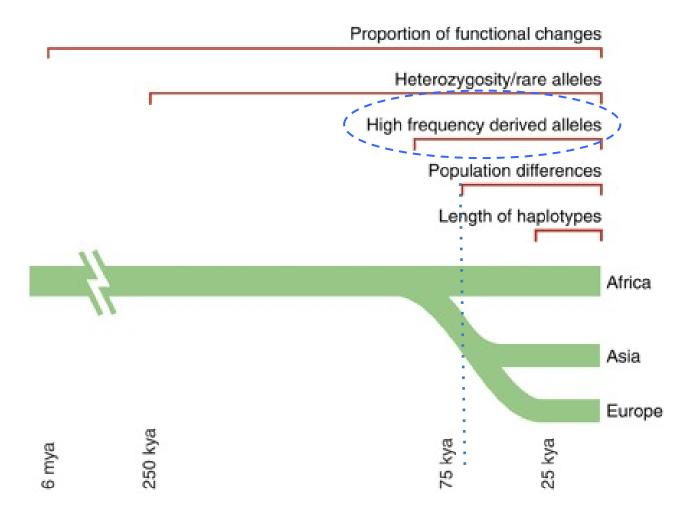


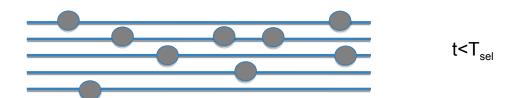


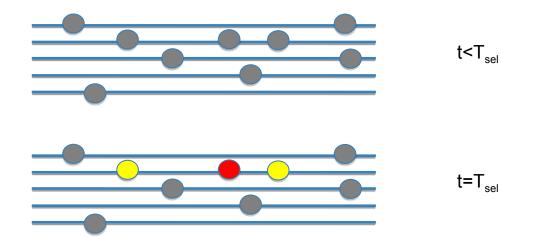


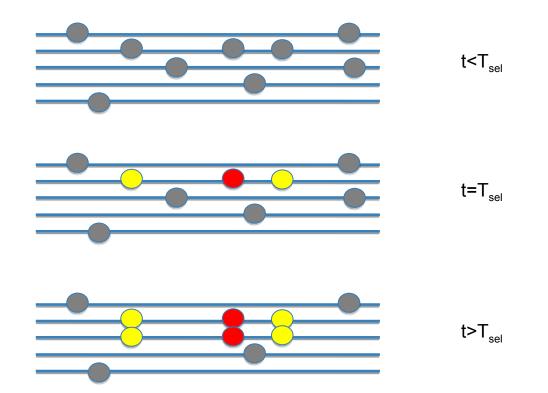


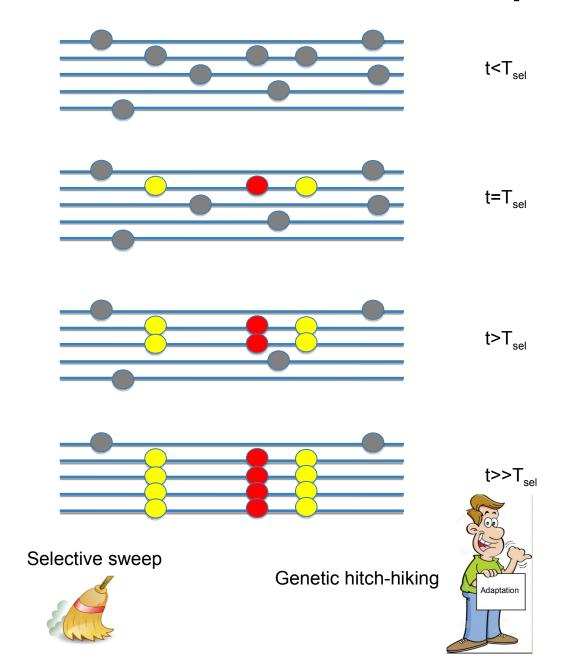
Methods to infer selection

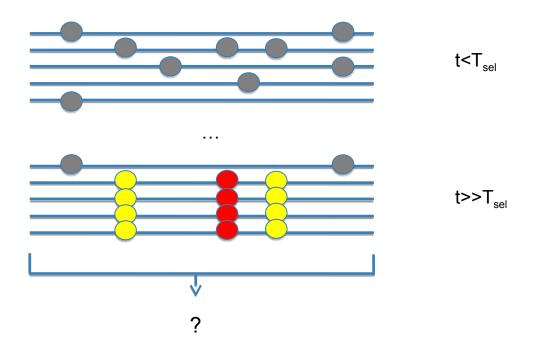


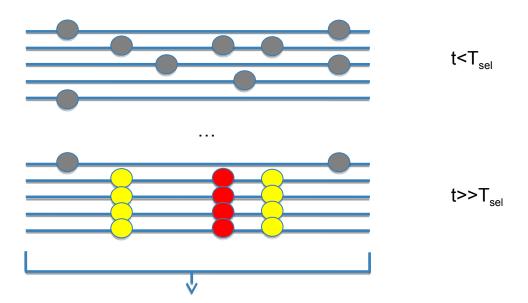




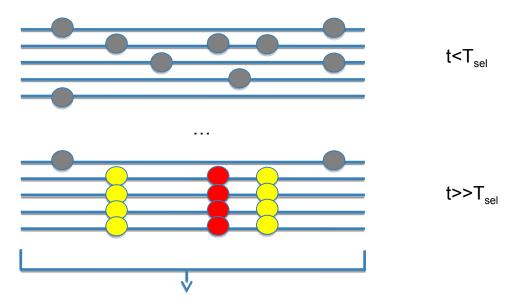








- Reduction of polymorphisms levels (e.g. from 7 to 5 SNPs)

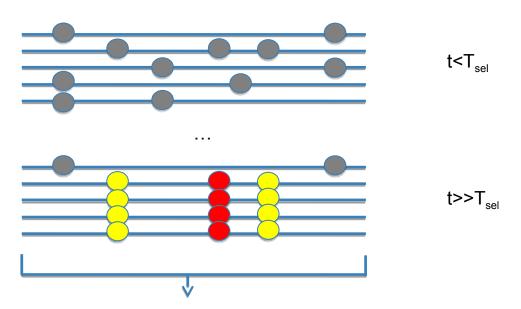


- Reduction of polymorphisms levels (e.g. from 7 to 5 SNPs)

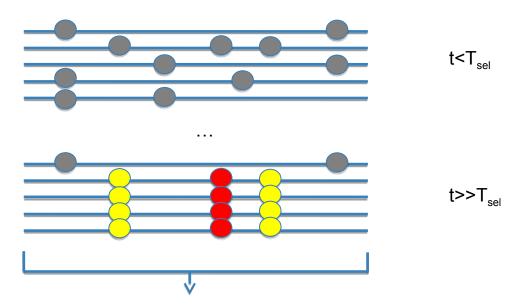
Nucleotide diversity index: Watterson's Theta with K SNPs and n chromosomes

$$\theta_W = \frac{K}{a_n}$$

$$a_n = \sum_{i=1}^{n-1} \frac{1}{i}$$



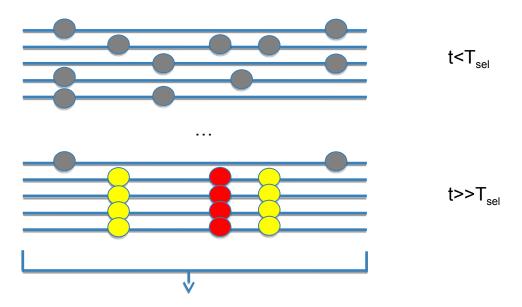
- Reduction of polymorphisms levels (Theta)
- ?



- Reduction of polymorphisms levels (Theta)
- Excess of low-frequency variants

Nucleotide diversity index: average pairwise nucleotide differences (Pi) with $k_{i,j}$ equal to the number of nucleotide differences between sequences i and j

$$\pi = \frac{\sum_{i=1}^{m-1} \sum_{j=+1}^{m} k_{i,j}}{\binom{n}{2}}$$



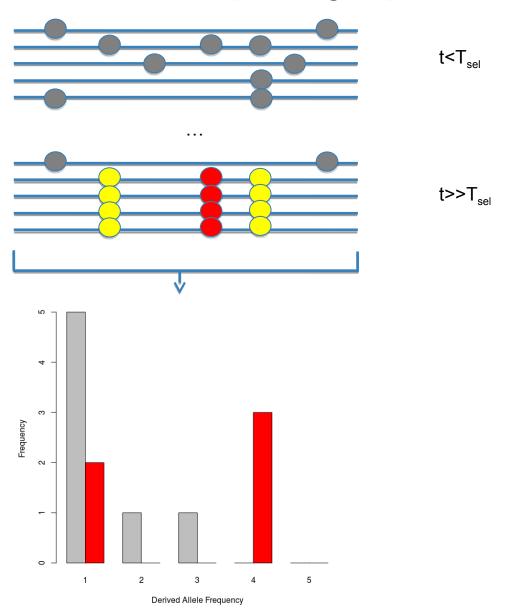
- Reduction of polymorphisms levels (Theta)
- Excess of low-frequency variants (Pi)

Under neutrality, Theta and Pi are expected to be the same. Tajima's D measures their difference.

$$D = \frac{\pi - \theta_W}{\sqrt{\hat{V}(\pi - \theta_W)}}$$

D<0 is suggestive of an excess of low-frequency variants

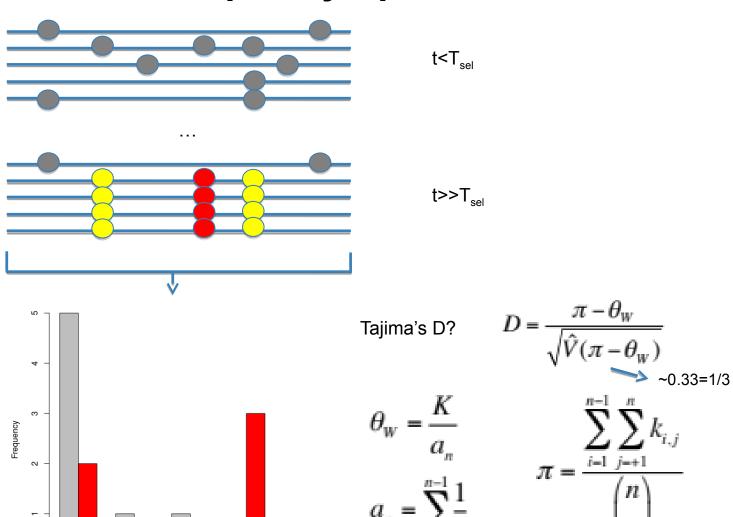
The Site Frequency Spectrum



The Site Frequency Spectrum

2

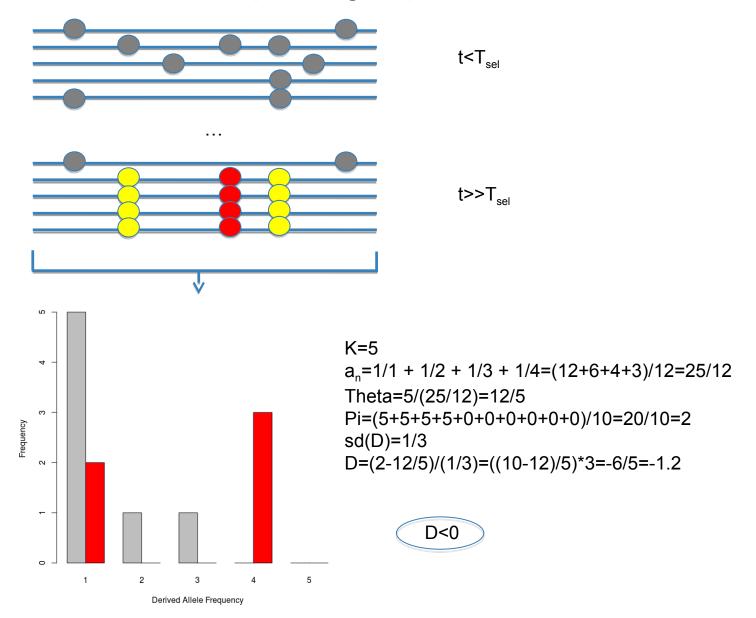
Derived Allele Frequency



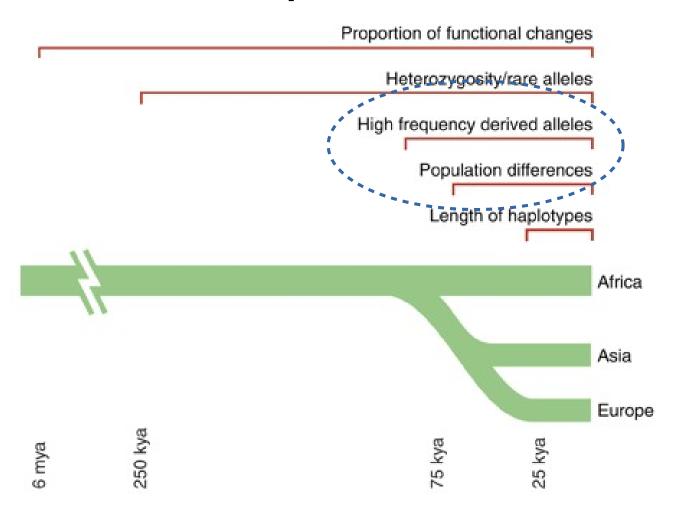
5

= 10, the number of comparisons you need to make

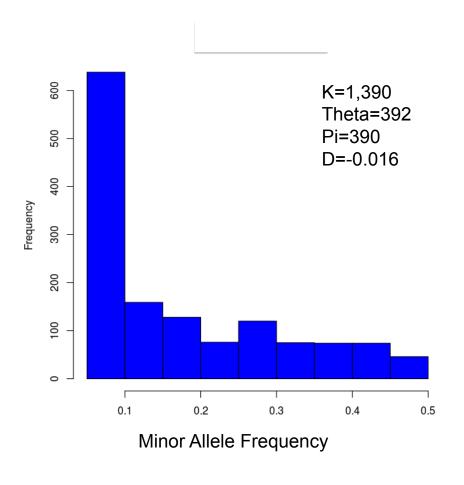
The importance of being... The Site Frequency Spectrum



Inference of positive selection



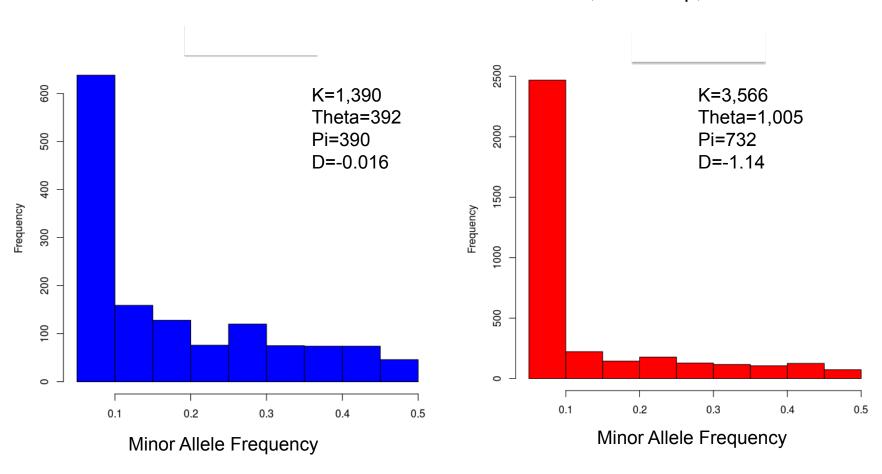
Confounding factor



Confounding factor

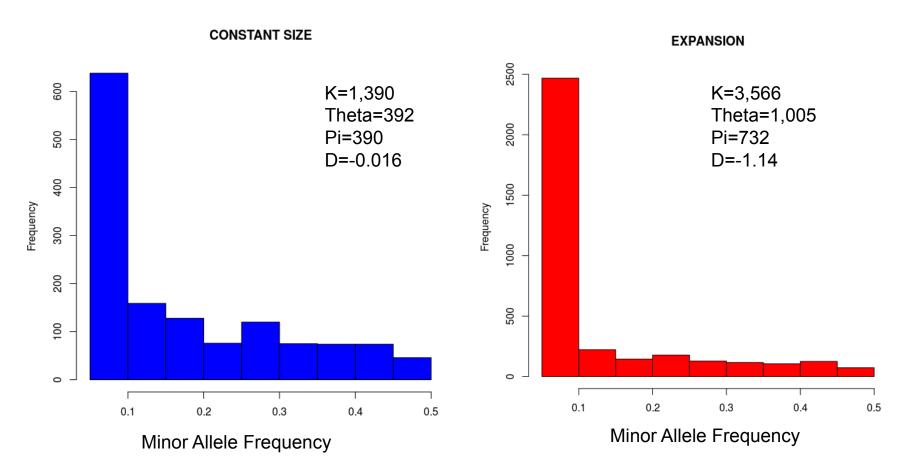


n=20; L=500kbp; no selection



Demography matters!

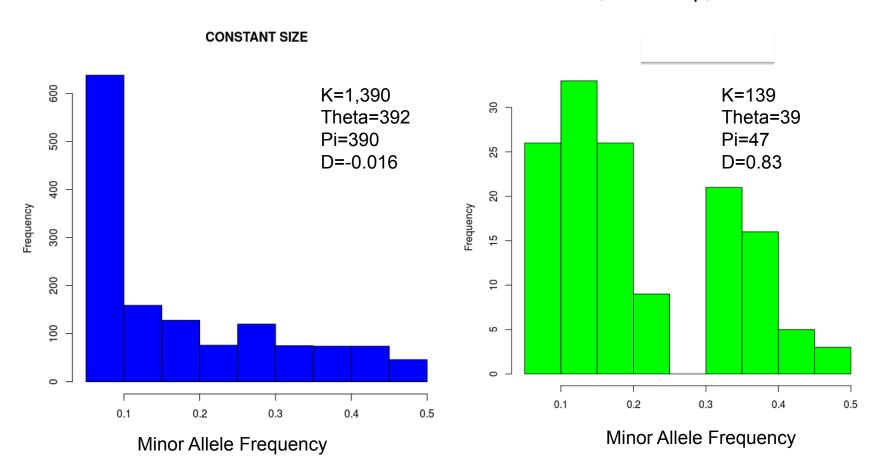
n=20; L=500kbp; no selection



- Excess of segregating sites
- Excess of low-frequency variants
- SFS-derived summary statistics may fail to distinguish between the effects of demography and selection

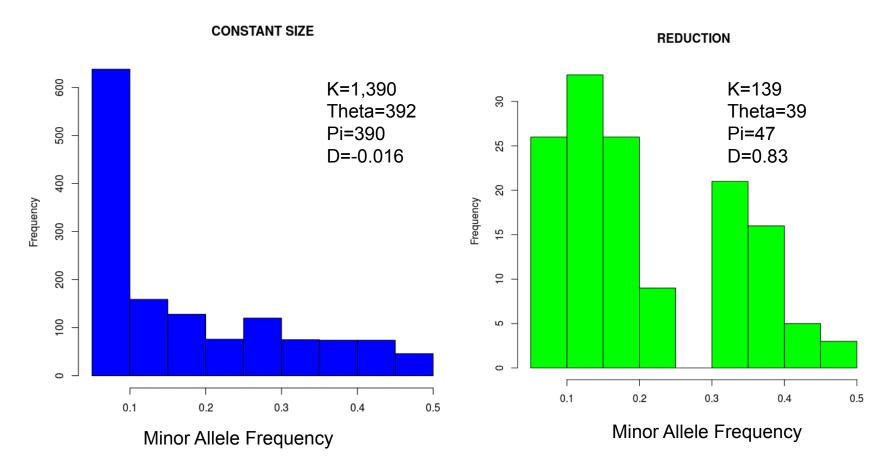
Demography matters?

n=20; L=500kbp; no selection



Demography matters!

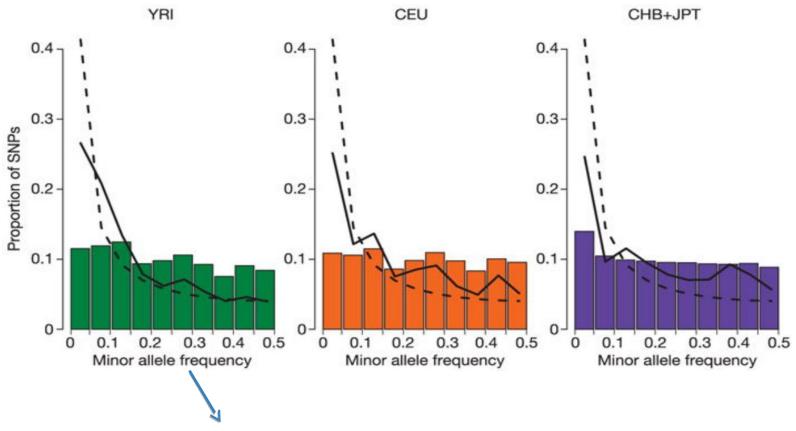
n=20; L=500kbp; no selection



- Depletion of segregating sites
- Excess of intermediate-frequency variants
- SFS-derived summary statistics may fail to distinguish between the effects of demography and selection

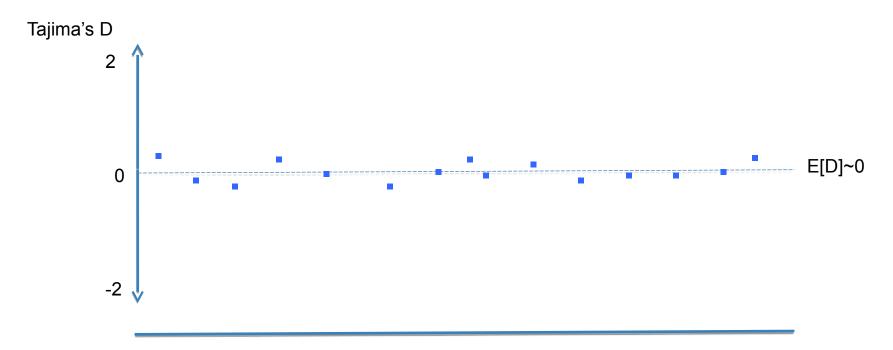
Experimental design matters?

The effect of ascertainment bias



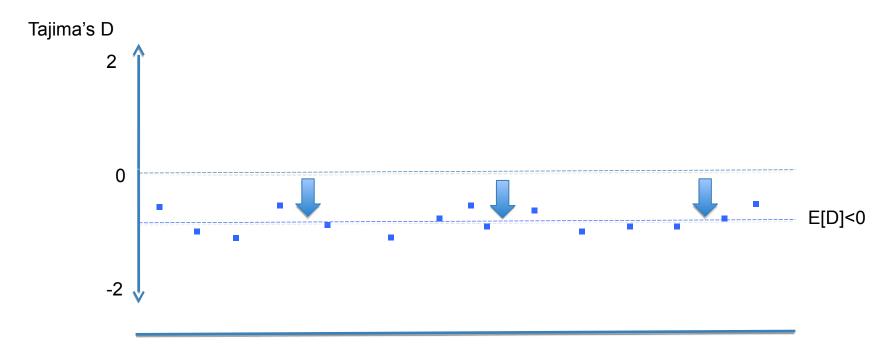
Deficiency of low-frequency variants

Under constant population size:



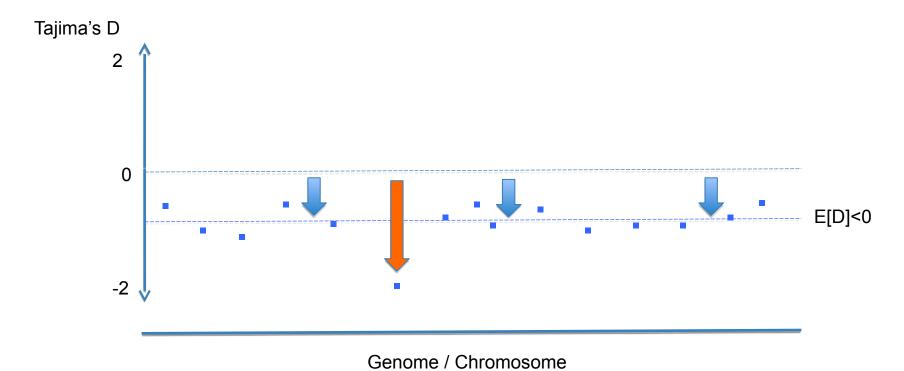
Genome / Chromosome

Under expanding population size:



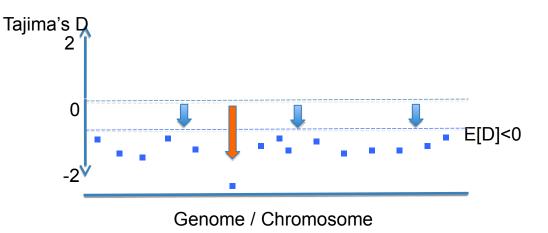
Genome / Chromosome

Under expanding population size and positive selection:

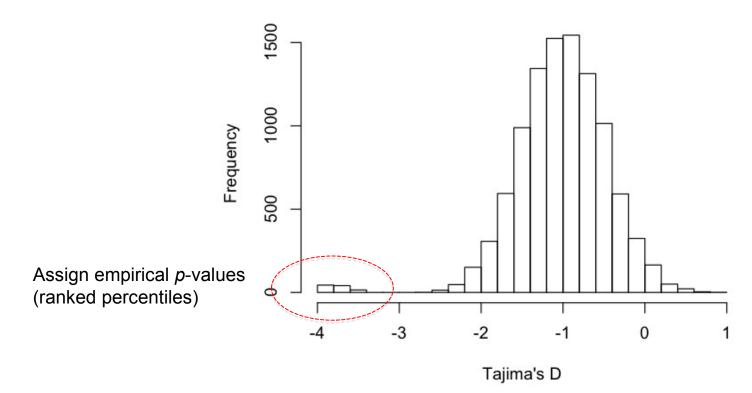


Demography affects all loci equally, while selection changes local patterns

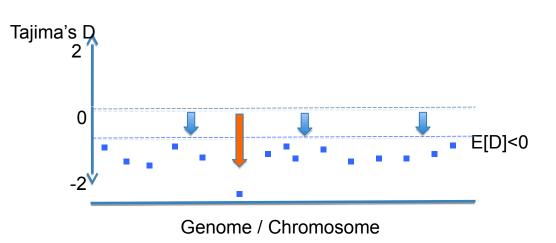
Outlier approach



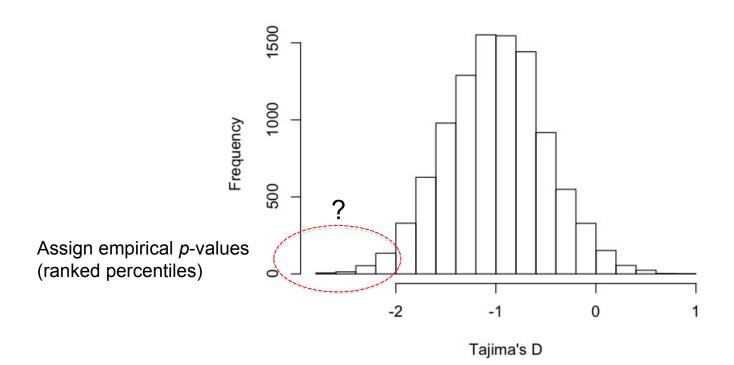
Empirical distribution



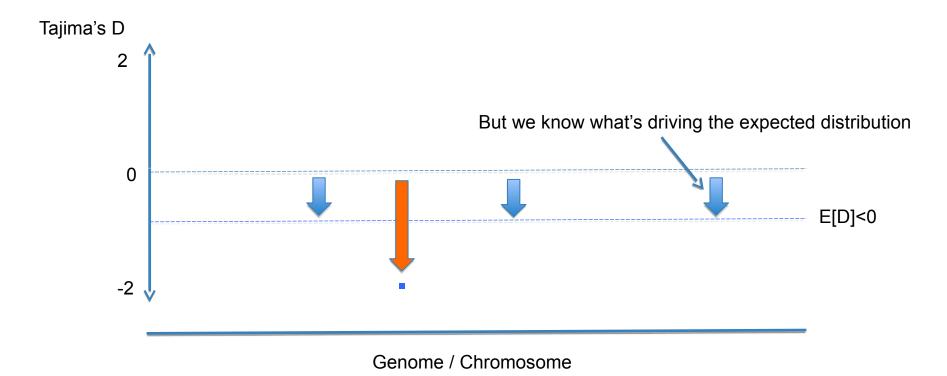
Outlier approach



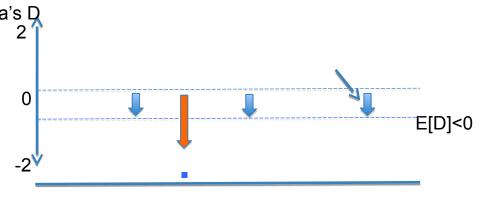
Empirical distribution



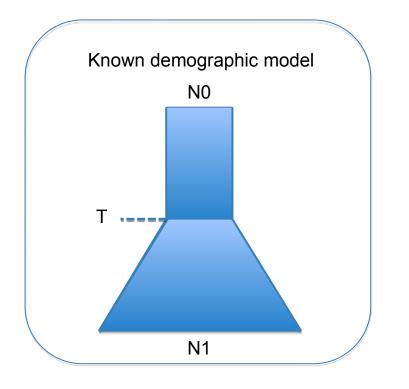
Under expanding population size and positive selection:



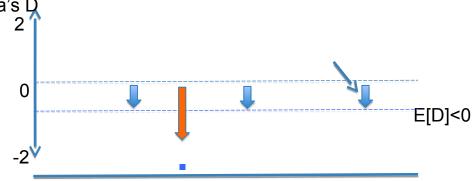
Demography affects all loci equally, while selection changes local patterns What should we do if we don't have genome-wide data? Simulations-based 2 approach 0



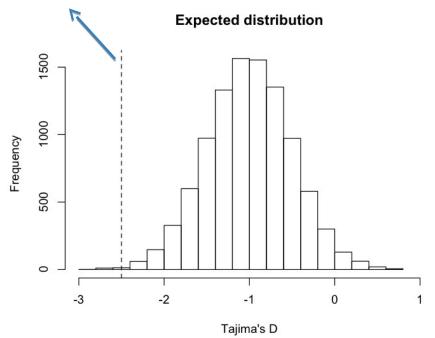
Genome / Chromosome

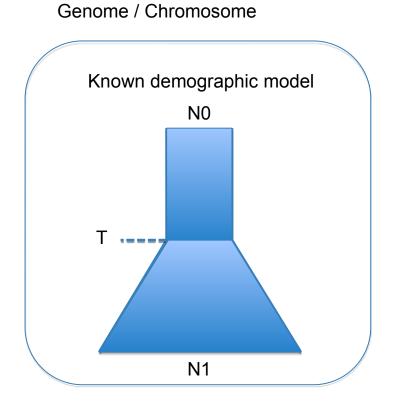


Simulations-based 2 approach 0

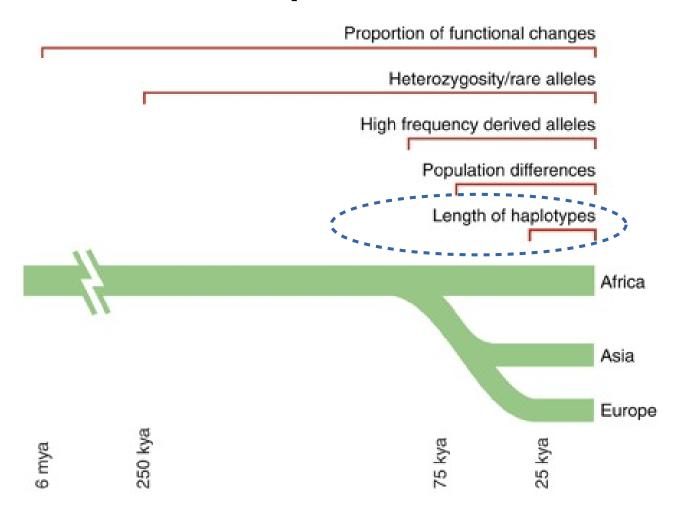


Assign *p*-values (based on ranked percentile of observed value)

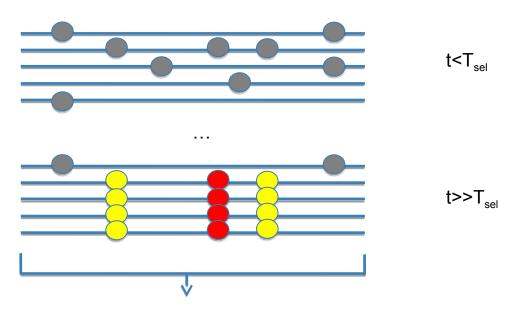




Inference of positive selection

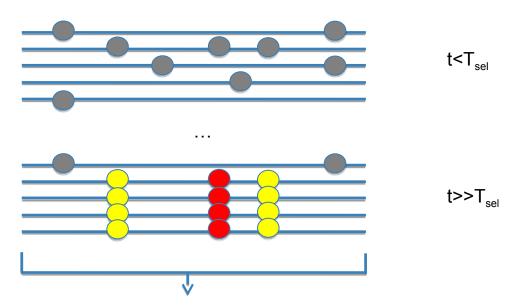


Positive selection

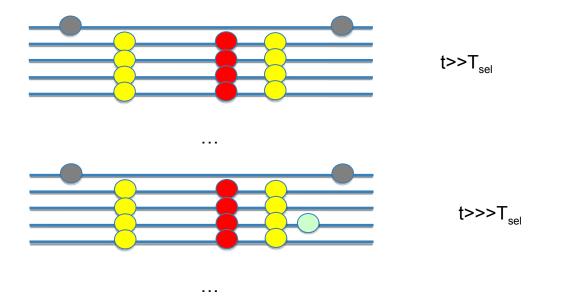


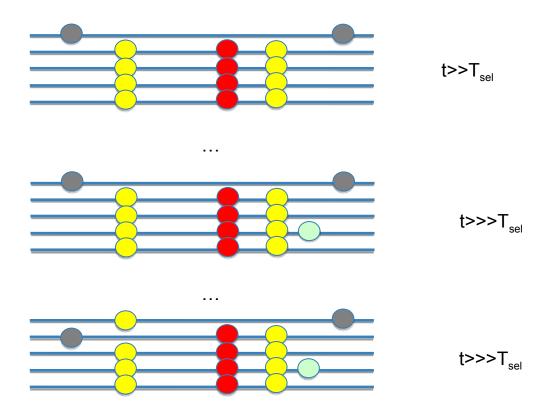
- Reduction of polymorphisms levels (Theta)
- Excess of low-frequency variants (Pi, Tajima's D, SFS)
- 7

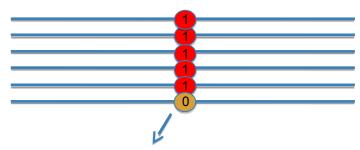
Positive selection



- Reduction of polymorphisms levels (Theta)
- Excess of low-frequency variants (Pi, Tajima's D, SFS)
- Extended haplotype homozygosity / Extended LD

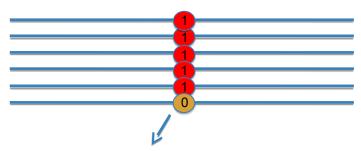






Core haplotype is 1 (Biallelic: 0 is ancestral, 1 is derived allele)

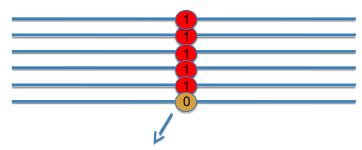
$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$
Core SNP



Core haplotype is 1 (Biallelic: 0 is ancestral, 1 is derived allele)

$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$

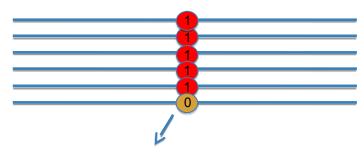
Until marker x_i (starting from x_o)



Core haplotype is 1 (Biallelic: 0 is ancestral, 1 is derived allele)

$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$

Sum across all unique haplotypes carrying the core SNP



Core haplotype is 1

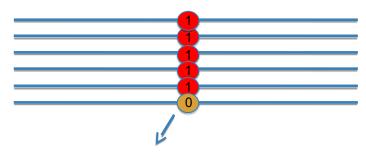
(Biallelic: 0 is ancestral, 1 is derived allele)

$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$

Sum across all unique haplotypes carrying the core SNP

 n_h is haplotype frequency of h

 n_h is haplotype frequency of the core SNP



Core haplotype is 1

(Biallelic: 0 is ancestral, 1 is derived allele)

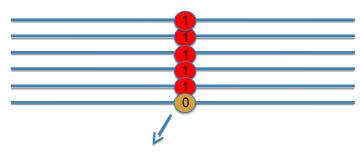
$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$

Sum across all unique haplotypes carrying the core SNP

 n_h is haplotype frequency of h

 n_h is haplotype frequency of the core SNP

$$EHH_c(x_i = 0) = ?$$



Core haplotype is 1

(Biallelic: 0 is ancestral, 1 is derived allele)

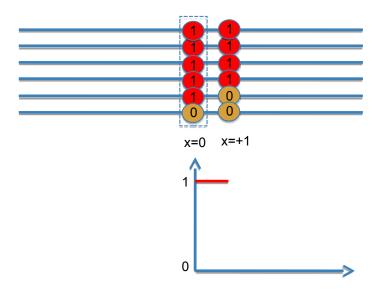
$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$

$$n_{h} \text{ is haplotype frequency of } h$$

$$n_{c} \text{ is haplotype frequency of the core SNP}$$

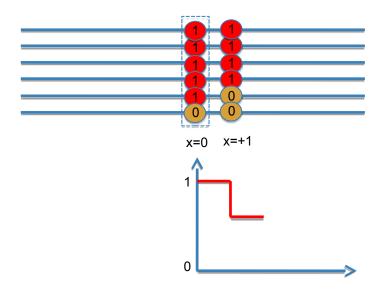
Sum across all unique haplotypes carrying the core SNP

$$EHH_{c}(x_{j}=0) = \frac{\binom{5}{2}}{\binom{5}{2}} = 1$$

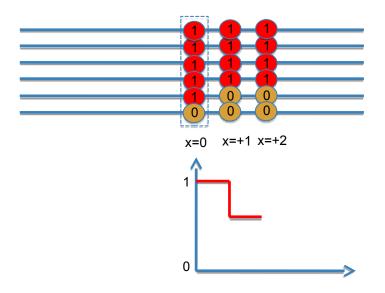


$$EHH_{c}(x_{i} = +1) = ?$$

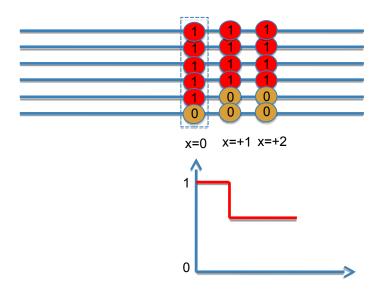
How many unique haplotypes carrying the core SNP? What is their frequency?



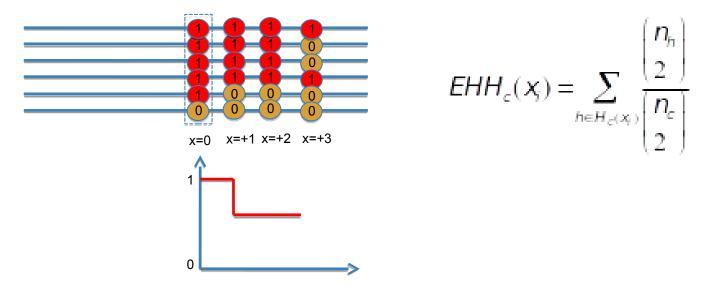
$$EHH_{c}(x_{5} = +1) = \frac{\binom{4}{2} + \binom{1}{2}}{\binom{5}{2}} = \frac{6+0}{10} = 0.60$$



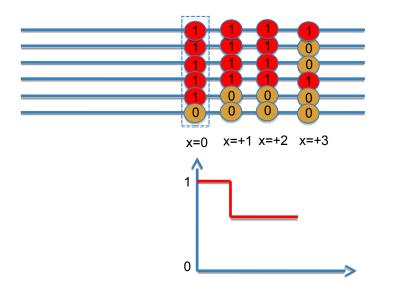
$$EHH_{c}(x_{i} = +2) = ?$$



$$EHH_c(x_i = +2) = EHH_c(x_i = +1) = 0.60$$



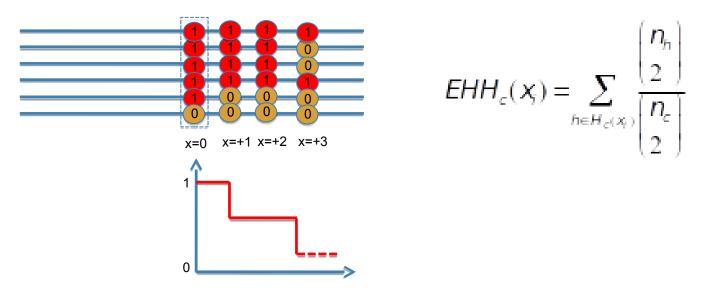
How many unique haplotypes carrying the core SNP? What is their frequency?



$$EHH_{c}(x_{j}) = \sum_{h \in H_{c}(x_{j})} \frac{\binom{n_{h}}{2}}{\binom{n_{c}}{2}}$$

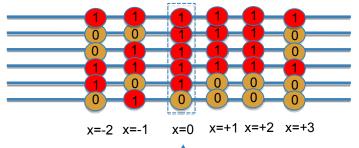
How many unique haplotypes carrying the core SNP? What is their frequency?
1111 with freq=2
1100 with freq=1

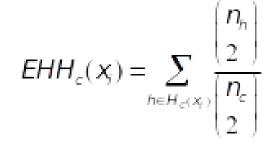
$$EHH_{c}(x_{i} = +3) = ?$$



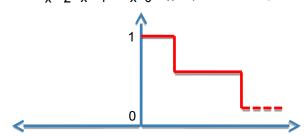
How many unique haplotypes carrying the core SNP? What is their frequency?
1111 with freq=2
11000 with freq=1

$$EHH_{c}(x_{i} = +3) = \frac{\binom{2}{2} + \binom{2}{2} + \binom{1}{2}}{\binom{5}{2}} = \frac{1+1+0}{10} = 0.20$$





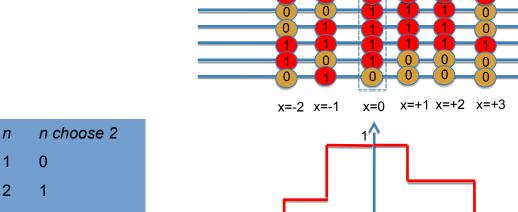
n	n choose 2
1	0
2	1
3	3
4	6
5	10
6	15



$$EHH_c(x_i =-1) = ?$$

 $EHH_c(x_i =-2) = ?$

Comment on differences (if any) between EHH(x=+2) and EHH(x=-2).

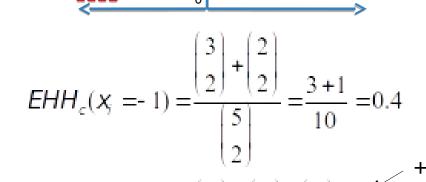


10

15

6

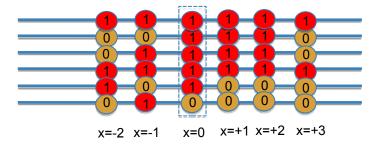
$$EHH_{c}(x_{i}) = \sum_{h \in H_{c}(x_{i})} \frac{ \binom{n_{h}}{2}}{ \binom{n_{c}}{2}}$$



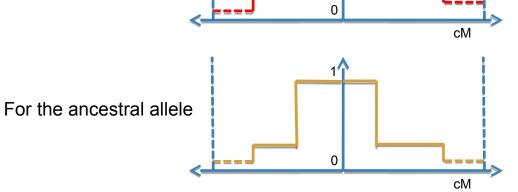
$$EHH_{c}(x_{5} = -2) = \frac{\binom{2}{2} + \binom{1}{2} + \binom{1}{2}}{\binom{5}{2}} = \frac{1+0+0}{10} = 0.1$$

Comment on differences (if any) between EHH(x=+2) and EHH(x=-2)?

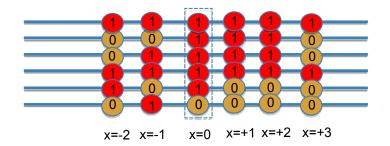
Integrated Haplotype Score

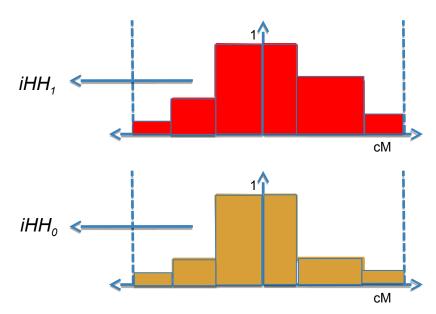






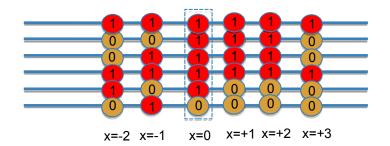
Integrated Haplotype Score

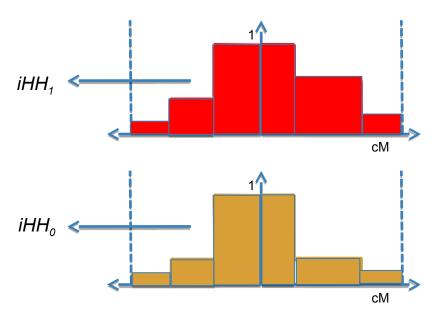




Integrated haplotype homozygosity (*iHH*)

Integrated Haplotype Score



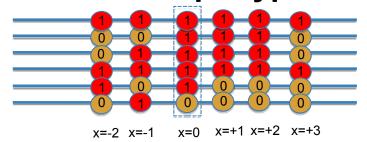


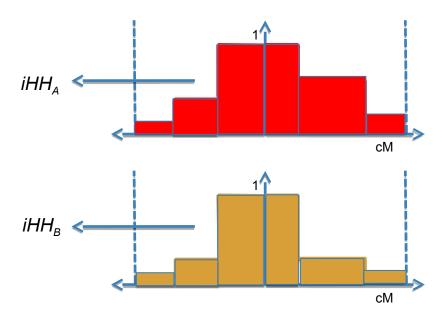
Integrated haplotype homozygosity (*iHH*)

Integrated haplotype score: iHs = In(iHH₁/iHH₀)

Genome-wide normalization in frequency bins (to mean=0 and sd=1)

Cross-population Extended Haplotype Homozygosity



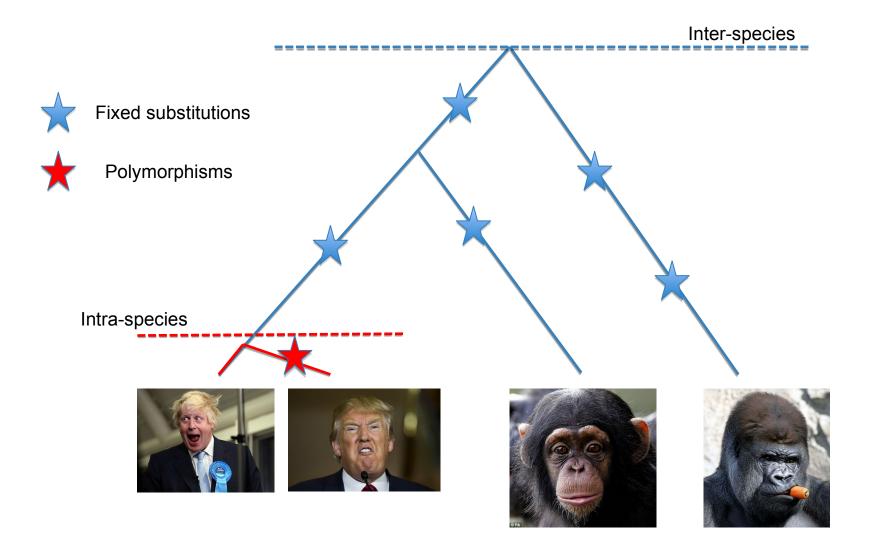


Integrated haplotype homozygosity (*iHH*) for **populations** A and B

Integrated haplotype score: $XP\text{-}EHH = In(iHH_A/iHH_B)$

Genome-wide normalization in frequency bins (to mean=0 and sd=1)

Inferring inter-species selection



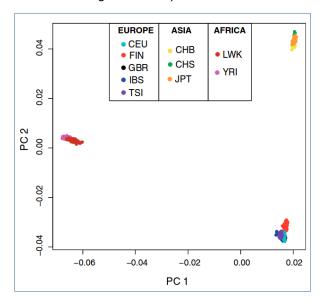
State-of-the-art methods to detect natural selection

1. Composite scores (Grossman et al. 2013)

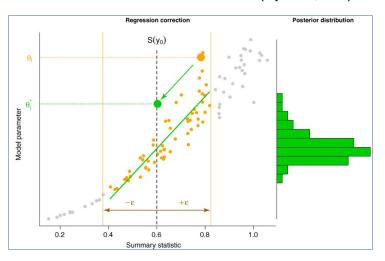
$BF_t = \frac{P(v_t \in bin_{t,k} \big| selected)}{P(v_t \in bin_{t,k} \big| unselected)}$ and defined the composite score as the product of the Bayes factor of each test: $CMS_{GW} = \prod_{t \in tests} BF_t$

3. Unsupervised machine learning

(PCA, Duforet-Frebourg et al. 2016)

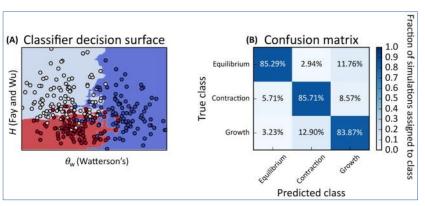


2. Simulations-based (rejection, ABC)



4. Supervised machine learning

(SVM, Schrider & Kern 2018)



How do we infer signals of natural selection from genomic data using machine learning and deep learning?

(let's move to part 3)

What are the main limitations of currently employed methods to detect selection?