

What can we learn from experimental evolution in sexual populations?



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experimental evolution allows us to test many questions...

- how repeatable is evolution?
- how well can we localize/identify QTL?
- what are the origins and fates of adaptive alleles?

Elucidating the molecular architecture of adaptation via evolve and resequence experiments

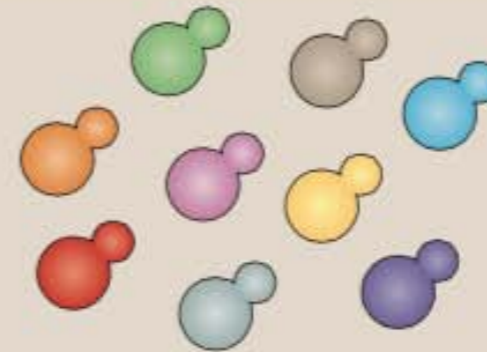
Anthony Long¹, Gianni Liti², Andrej Luptak³ and Olivier Tenaillon⁴



In vitro



Microbial isogenic



Microbial outbred



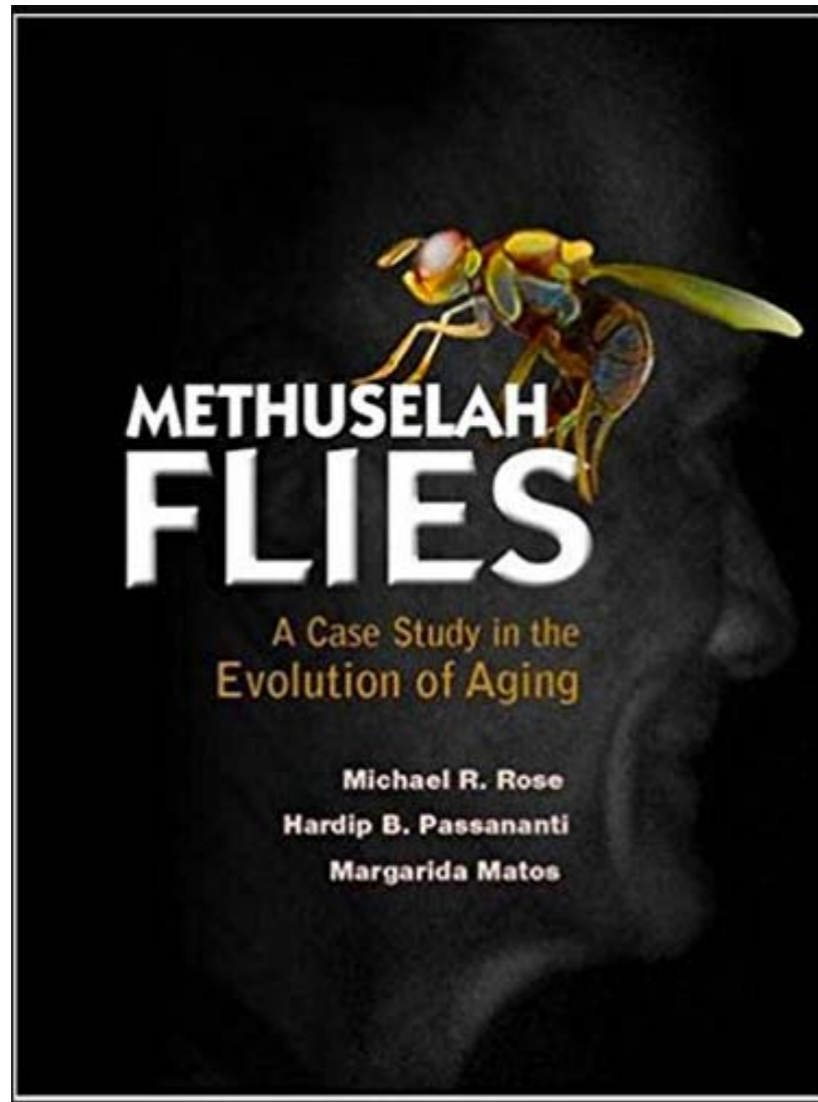
Obligate sexual higher eukaryotes

Example models	<i>In vitro</i>	Microbial isogenic	Microbial outbred	Obligate sexual higher eukaryotes
	Synthetic DNA or RNA molecules	Bacteria, haploid yeast or diploid yeast	Diploid yeast	<i>Drosophila melanogaster</i>

“The time is right for practitioners in the different systems to learn from one another.”



Drosophila melanogaster



selected treatment:

ACO



only 9-day
old adults
reproduce

600 generations by 2009

control treatment:

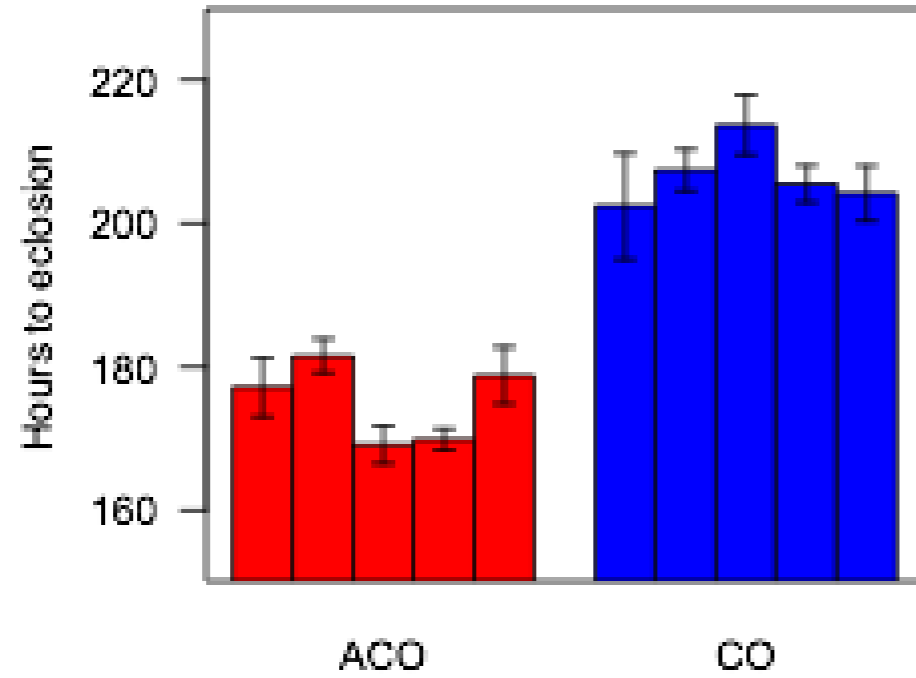
CO



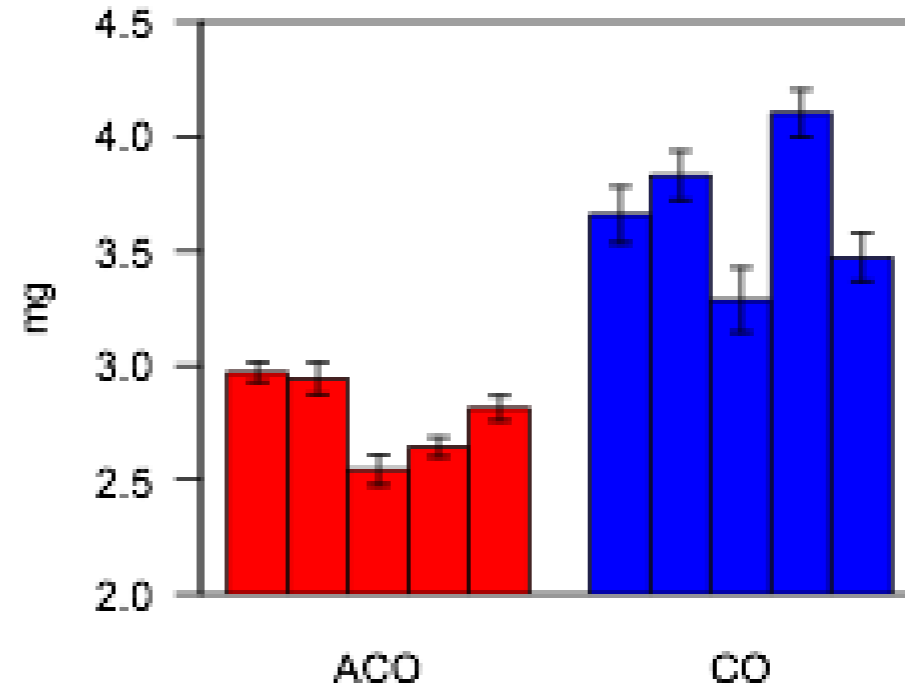
only 28-
day old
adults
reproduce

250 generations by 2009

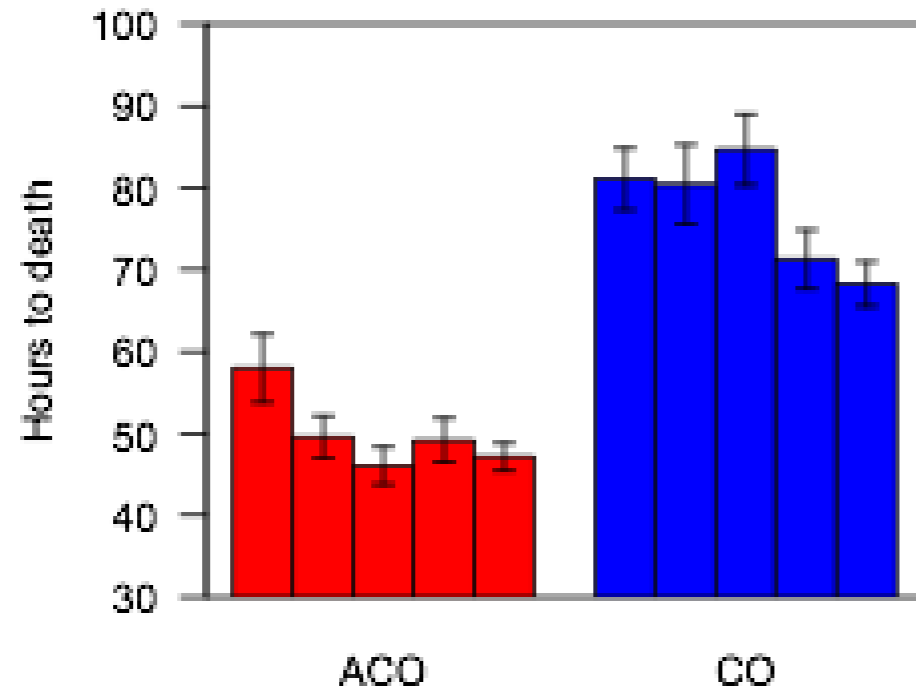
Mean development time



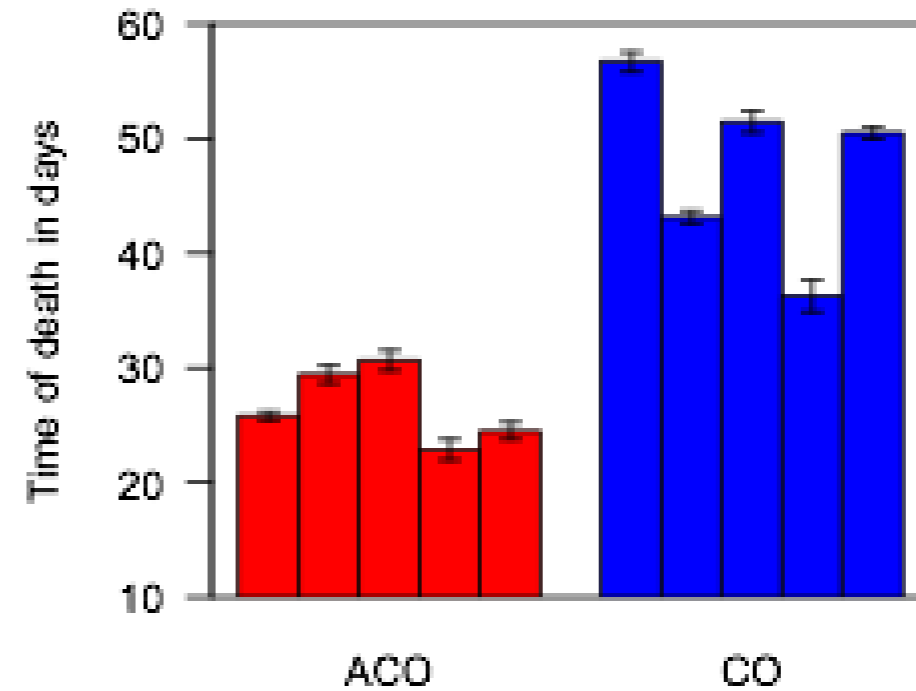
Dry weight at eclosion



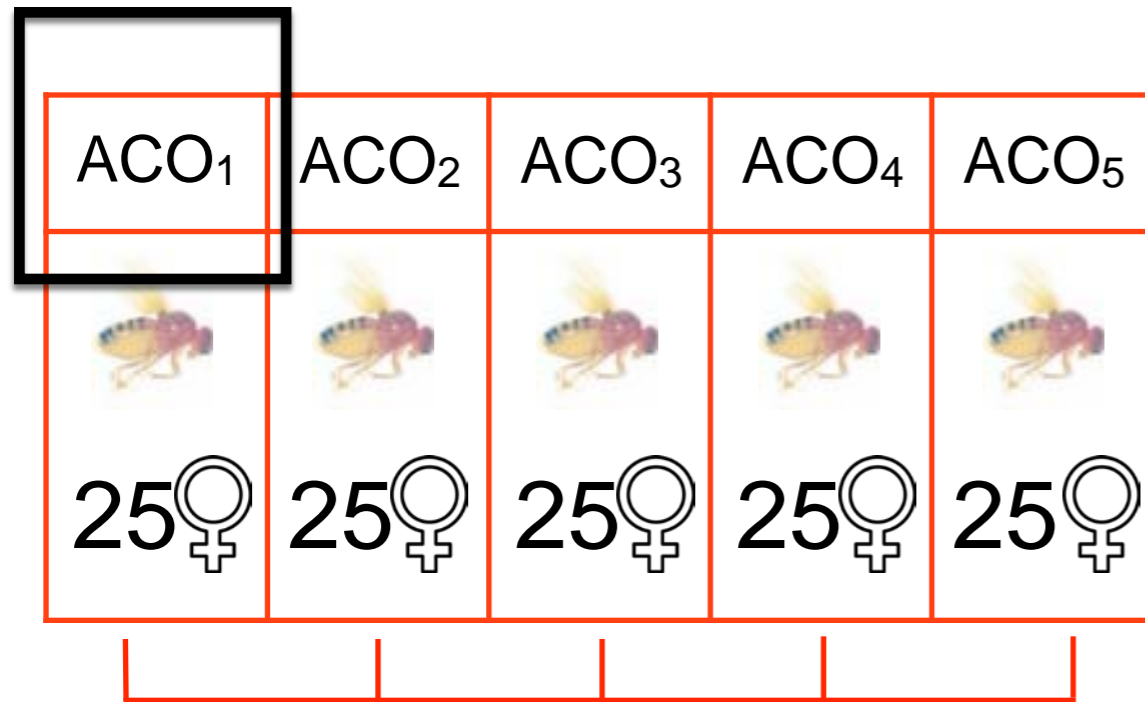
Mean starvation time



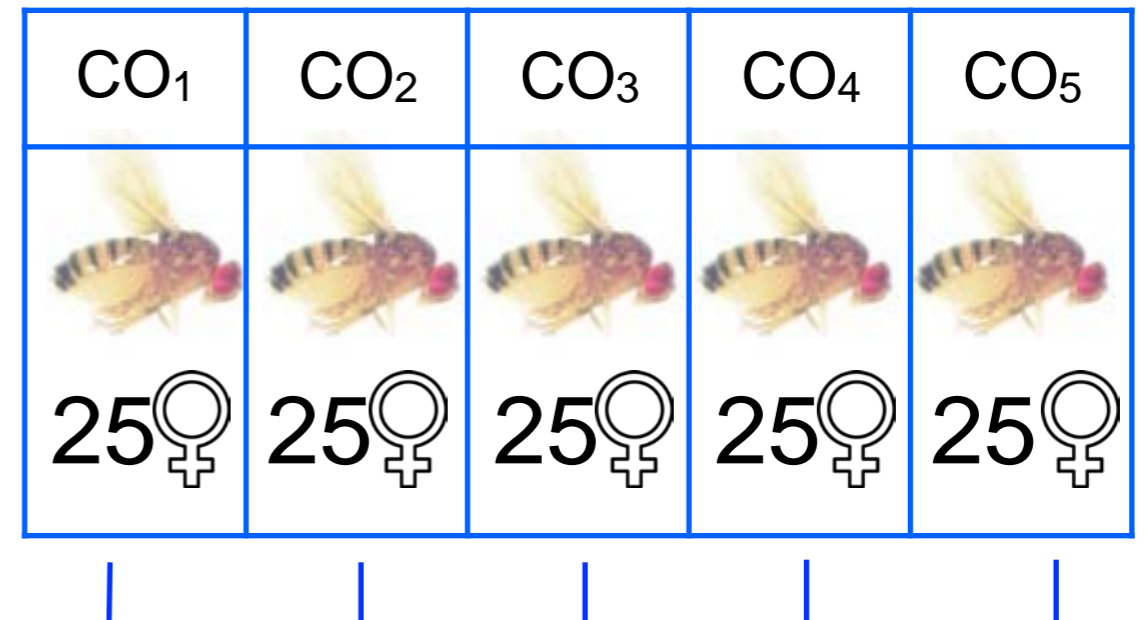
Mean longevity



Pooled genome sequencing



ACO library



CO library

third library sequenced: ACO₁

Pool- SEQ

Drosophila reference genome

————— G —————

ACO sequences

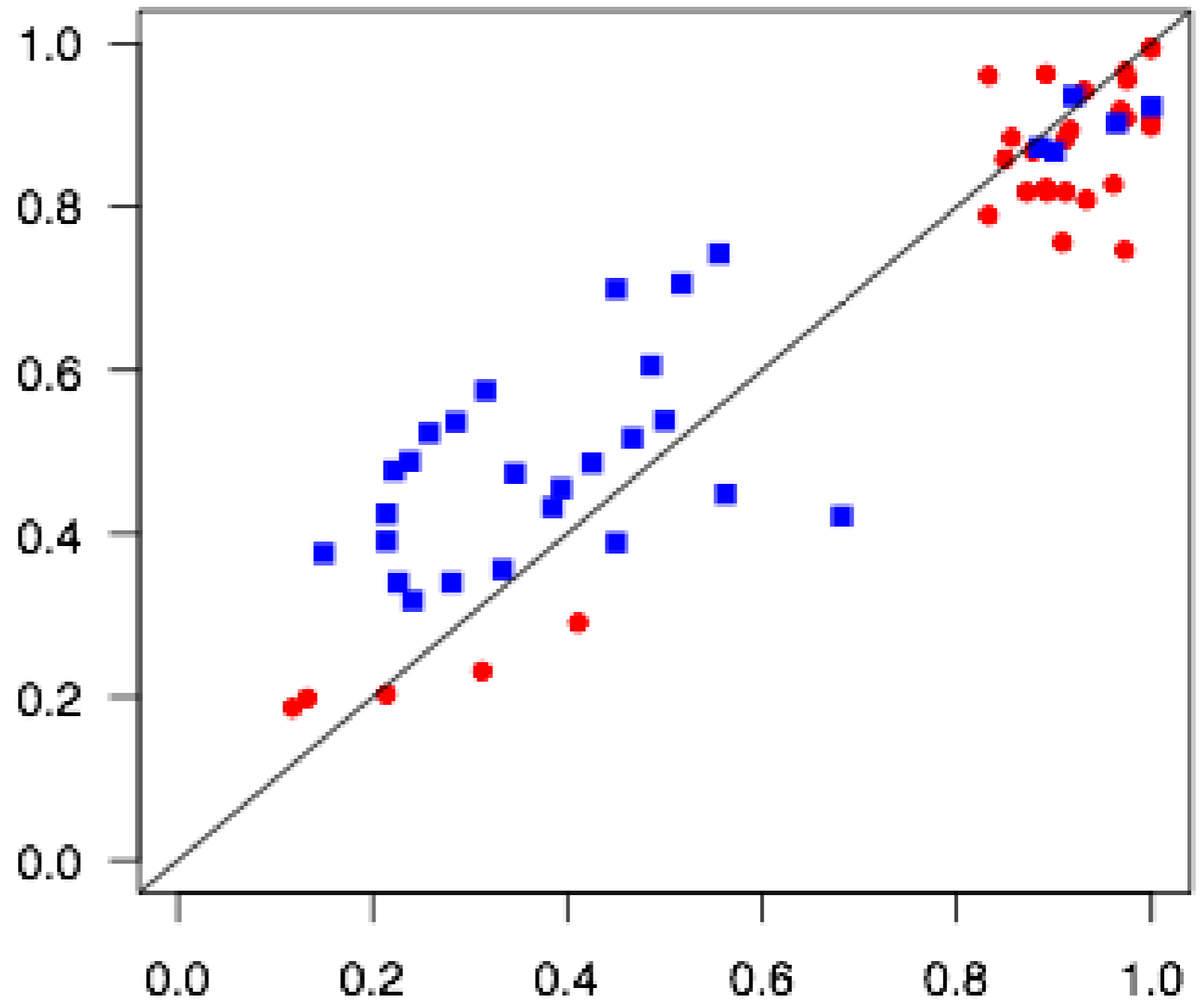
————— G —————
————— G —————
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————— G —————
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————— T —————
————— T —————
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————— T —————

CO sequences

————— T —————
————— T —————
————— T —————
————— T —————
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————— G —————
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————— G —————
————— G —————

utility of pooled samples for Pool-SEQ

allele frequency
from individual
genotypes



allele frequency from Pool-SEQ

 APPLICATIONS OF NEXT-GENERATION SEQUENCING

Sequencing pools of individuals — mining genome-wide polymorphism data without big funding

Christian Schlötterer¹, Raymond Tobler^{1,2}, Robert Kofler¹ and Viola Nolte¹

Abstract | The analysis of polymorphism data is becoming increasingly important as a complementary tool to classical genetic analyses. Nevertheless, despite plunging sequencing costs, genomic sequencing of individuals at the population scale is still restricted to a few model species. Whole-genome sequencing of pools of individuals (Pool-seq) provides a cost-effective alternative to sequencing individuals separately. With the availability of custom-tailored software tools, Pool-seq is being increasingly used for population genomic research on both model and non-model organisms. In this Review, we not only demonstrate the breadth of questions that are being addressed by Pool-seq but also discuss its limitations and provide guidelines for users.

About a decade ago, a fully sequenced genome was big news. But now, owing to rapid advances in next-generation sequencing (NGS) technology and computer algorithms for assembling short reads, we are enjoying the availability of an ever increasing number of genomes from a broad spectrum of non-model organisms¹. In parallel with the growing catalogue of reference genomes, a variety of approaches have emerged that seek to characterize the genome-wide polymorphism patterns. Arguably, the most comprehensive polymorphism data so far have been generated by single-nucleotide polymorphism (SNP) microarrays in humans^{2,3}. More recently, the field has begun moving towards the characterization of full genome sequences, with 1000 genome projects completed for humans⁴, *Arabidopsis thaliana*⁵ and cattle⁶. In *Drosophila melanogaster* too, hundreds of genomes have already been sequenced, and other species, such as pigs and dogs, are catching up. Does this imply that we have now captured all of the relevant variation and that, despite some minor bits and pieces of data remaining to be filled, in essence we are close to what we need to understand variation in these species?

Probably the best demonstration that this is not the case comes again from human genetics. The analysis of human diseases and other complex traits indicated that even the analysis of several thousand individuals frequently turned out to be insufficient to determine the underlying genetic architecture⁷. Given this scale, it is clear that many research questions cannot be addressed

by whole-genome sequencing of individuals, even though the sequencing costs of a human genome have now decreased below the 'magic line' of US\$1,000 (REF. 8).

In this Review, we discuss whole-genome sequencing of pools of individuals (Pool-seq) — an approach that provides genome-wide polymorphism data at considerably lower costs than sequencing of individuals. We explain why Pool-seq is more cost-effective, compare it to other approaches, review dedicated software tools, and discuss limitations and further directions. On the basis of various intraspecific whole-genome Pool-seq studies, we demonstrate its versatility and efficacy in facilitating a broad range of genome-wide analyses. However, we do not cover the metagenomic analysis of pools consisting of multiple species, as this has been reviewed elsewhere⁹.

The cost-effectiveness of Pool-seq

Key to population genetic surveys is information about polymorphic positions in the genome and the frequencies of variant alleles in various populations. The power of many genetic analyses increases with the accuracy to which allele frequencies can be determined from population samples. Pool-seq provides more accurate allele frequency estimation at a lower cost than sequencing of individuals^{10,11}. To understand the basis of this difference, it is important to remember that allele frequencies are typically estimated from samples drawn from a larger population. Smaller sample sizes

Next-generation sequencing (NGS; also known as second-generation sequencing). An umbrella term for different sequencing platforms delivering millions of short DNA sequence reads

Reads

DNA sequences that are generated by next-generation sequencing.

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doi:10.1038/nrg3803
Published online
23 September 2014

best practices:

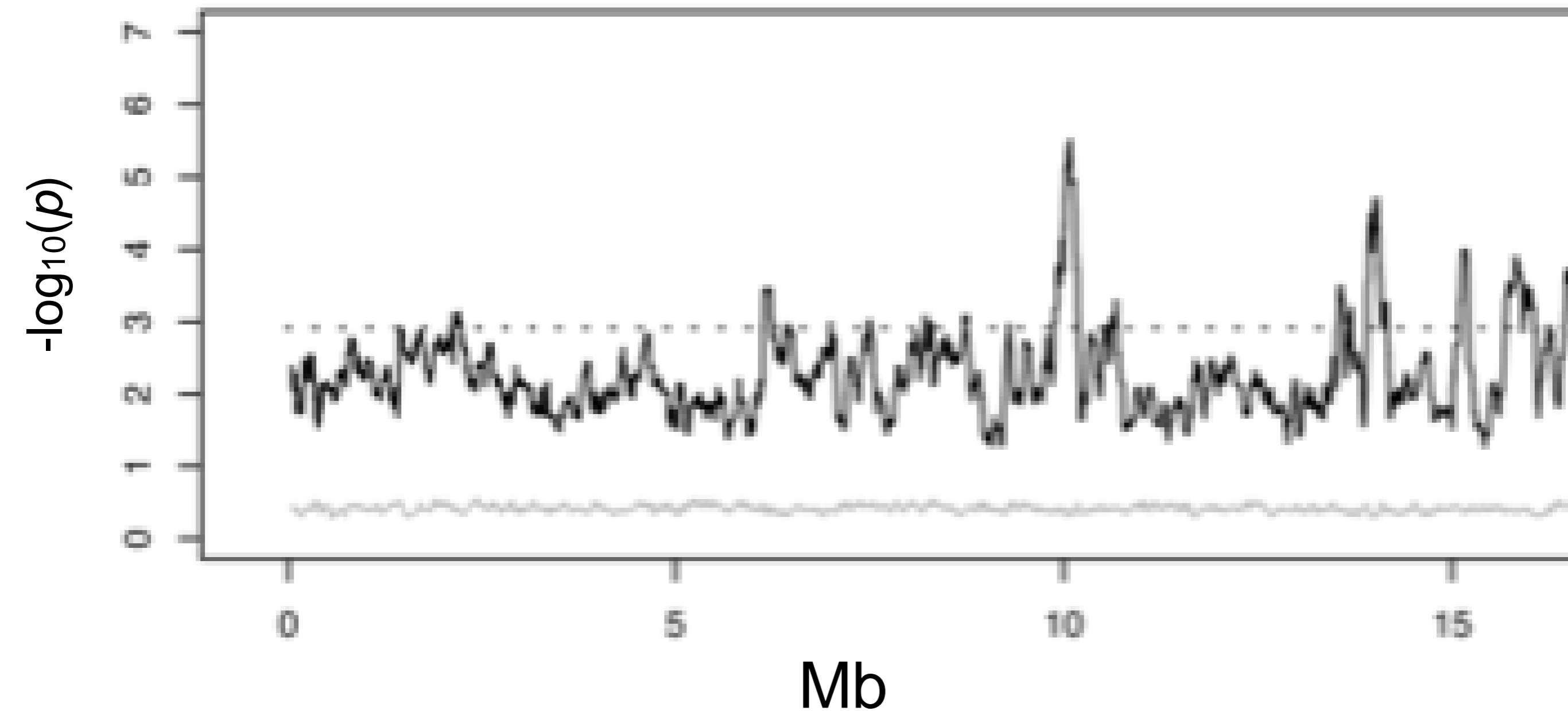
pool > 40 individuals

coverage > 50X

read lengths > 75bp

black line = frequency differences between
ACO and CO treatments

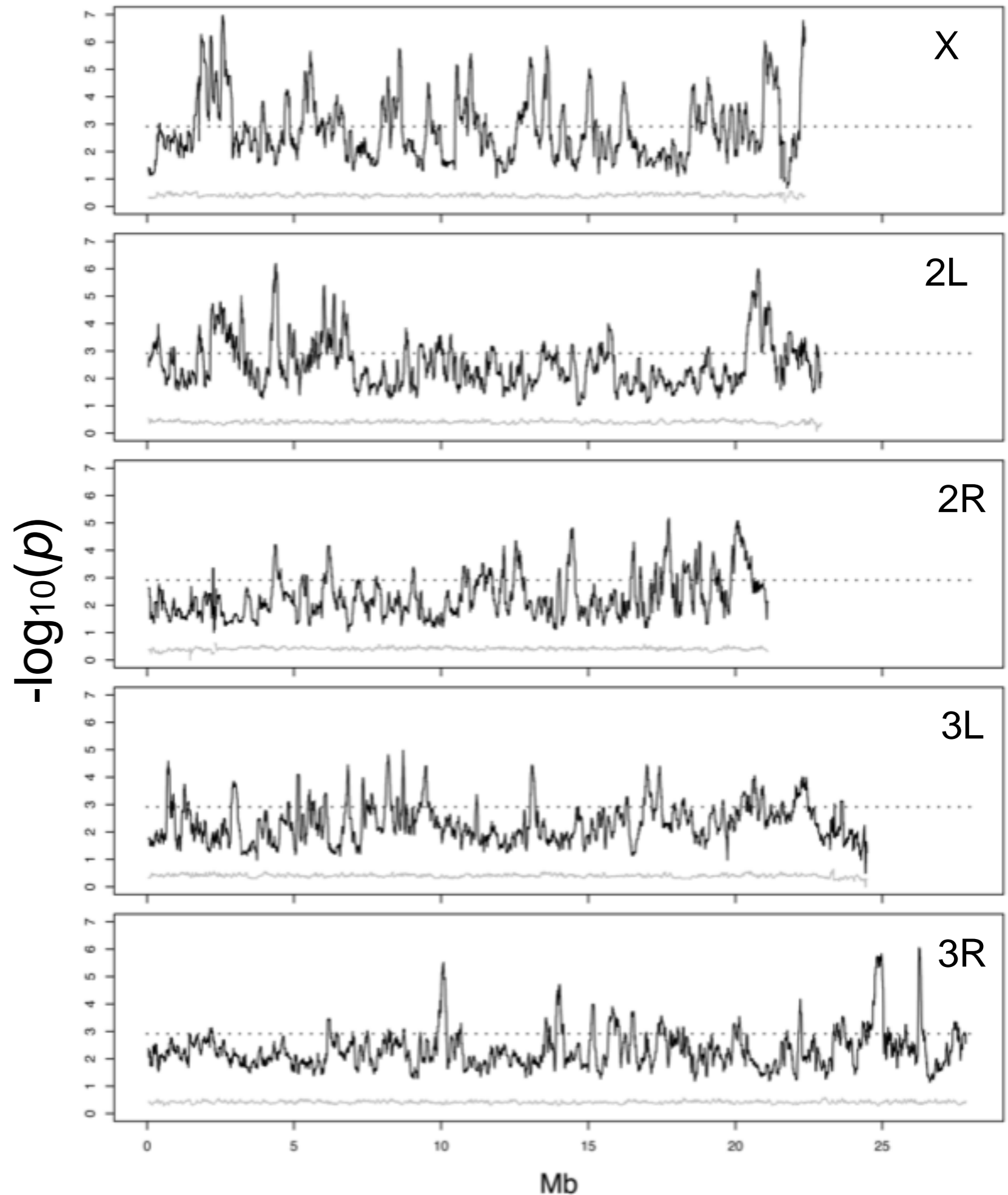
gray line = frequency differences between
ACO₁ and the entire ACO pool



lots of differentiation
between the **ACO**
and **CO** pools

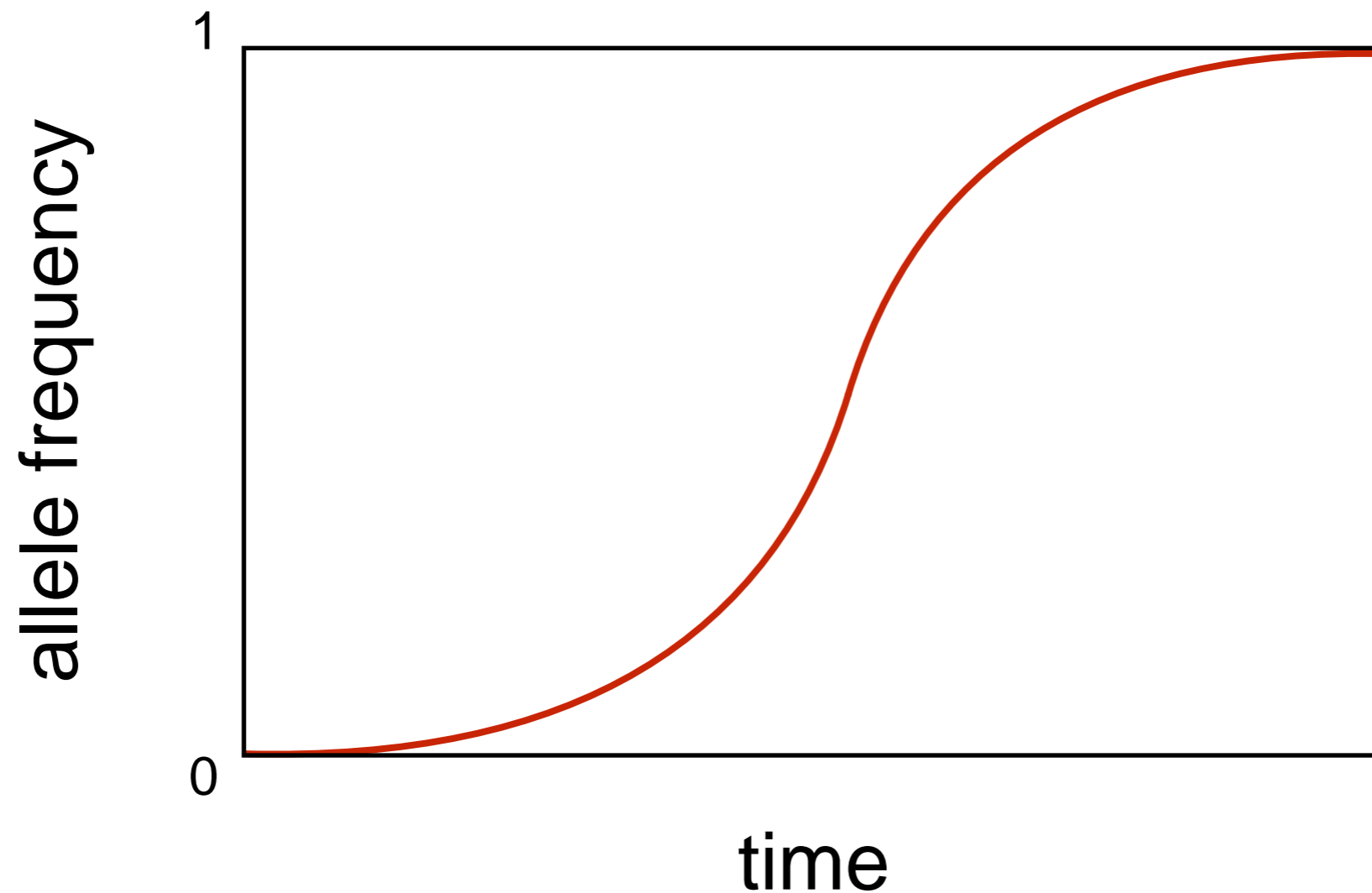
no differentiation
between **ACO₁** and
the **ACO** pool

~500 genes under
peaks



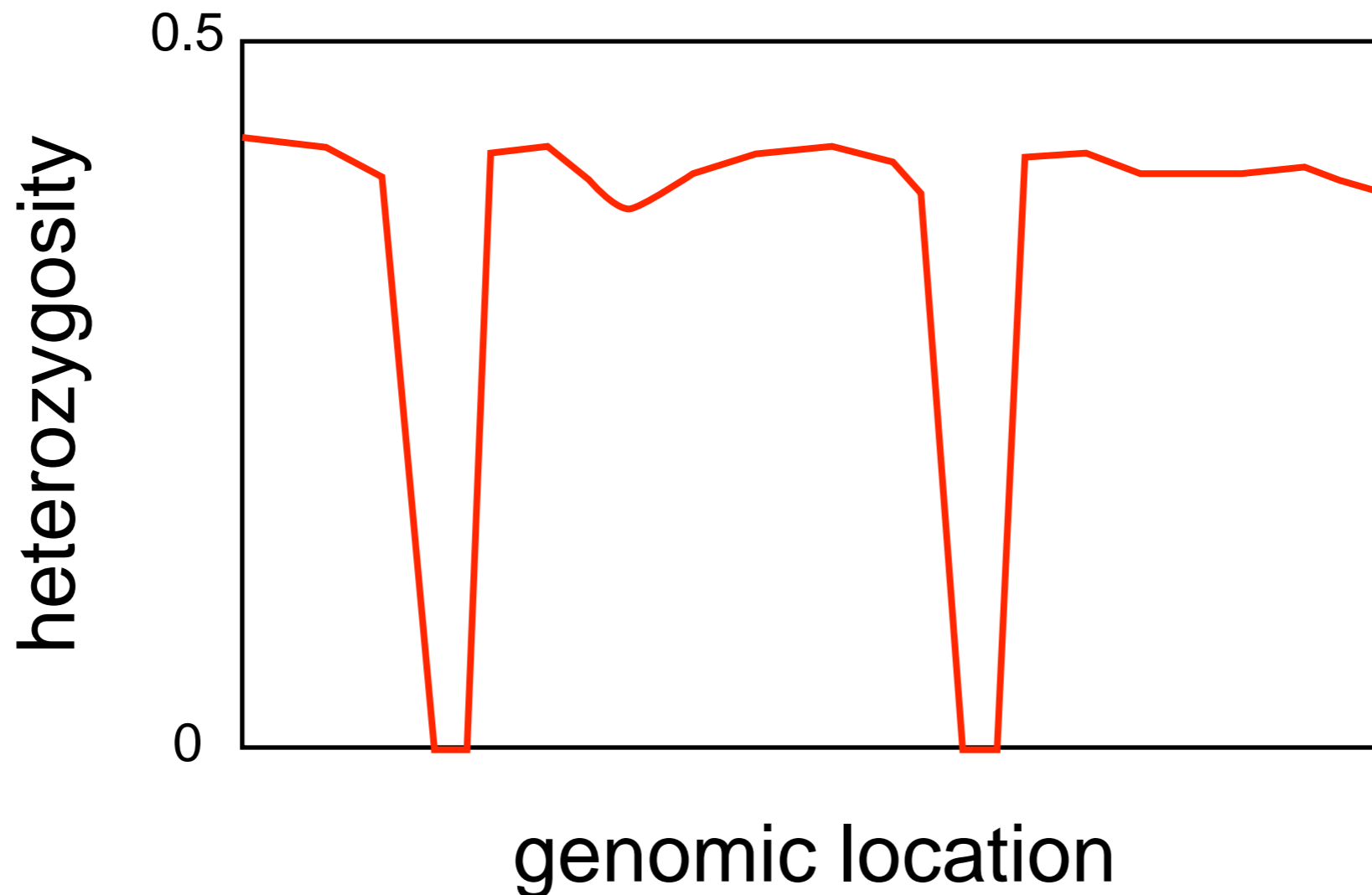
classic selective sweep

a beneficial mutation arises, natural selection increases the frequency of this allele until fixation



classic selective sweep

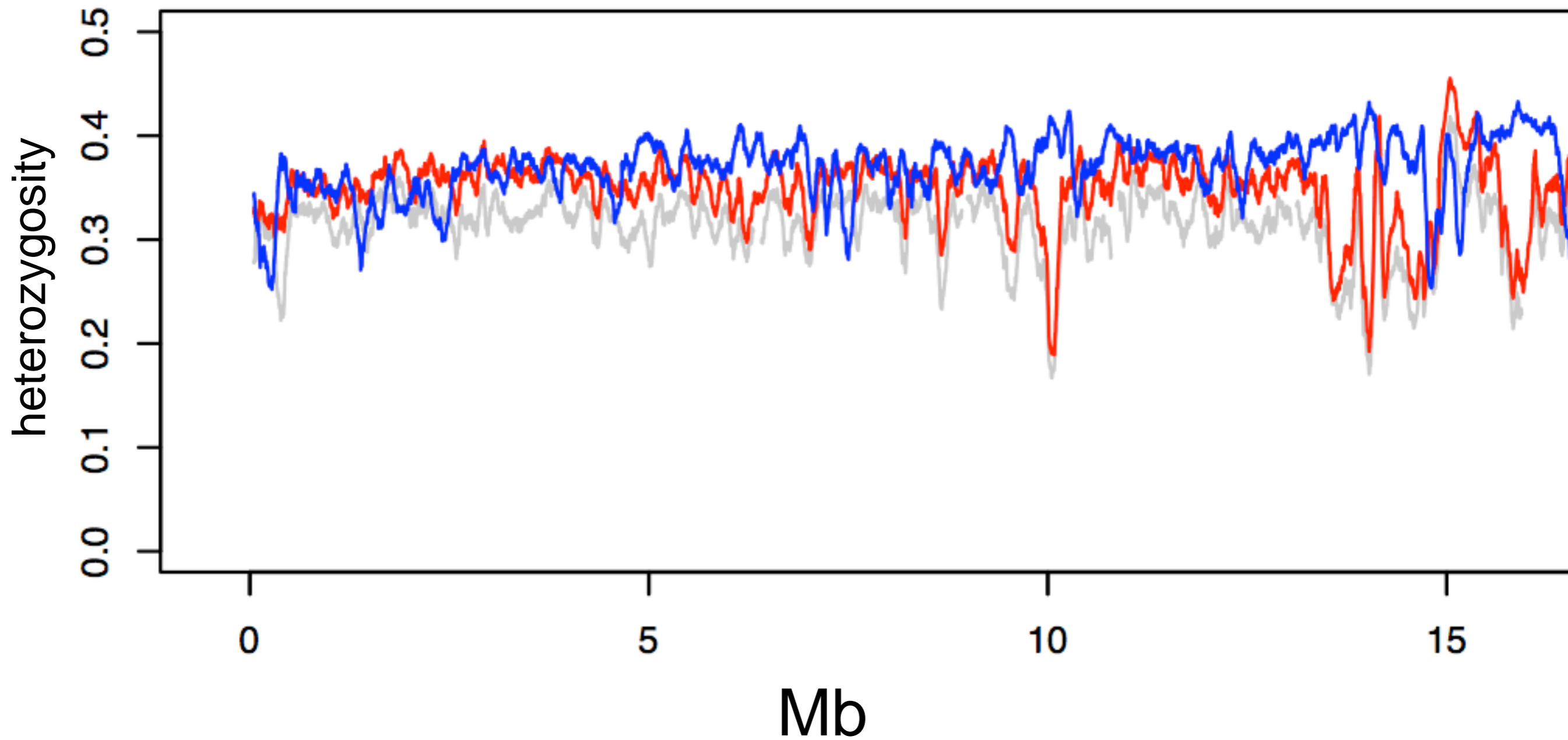
at sweep locations, heterozygosity losses should occur at selected and linked sites



blue = CO pool

red = ACO pool

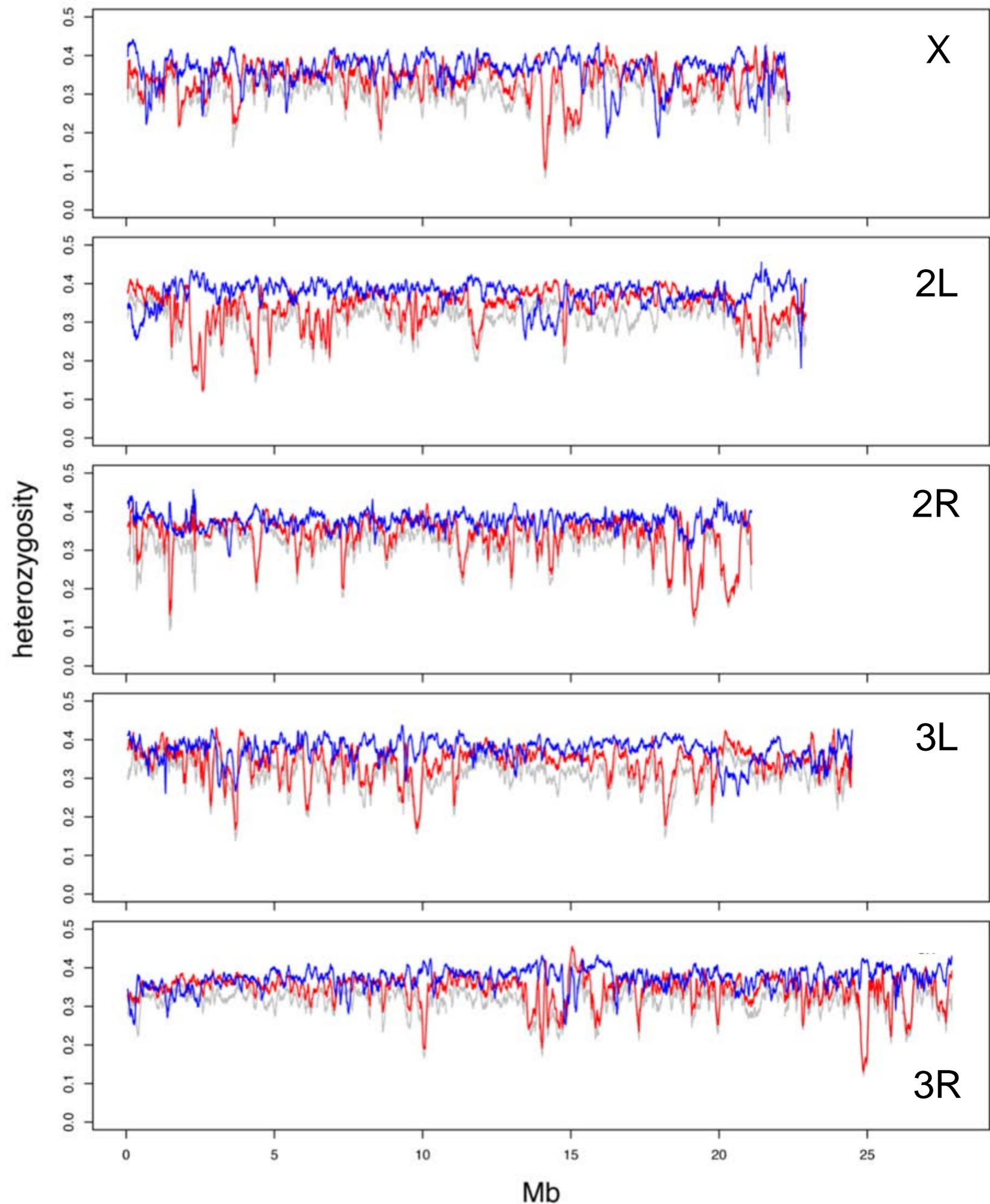
gray = ACO₁



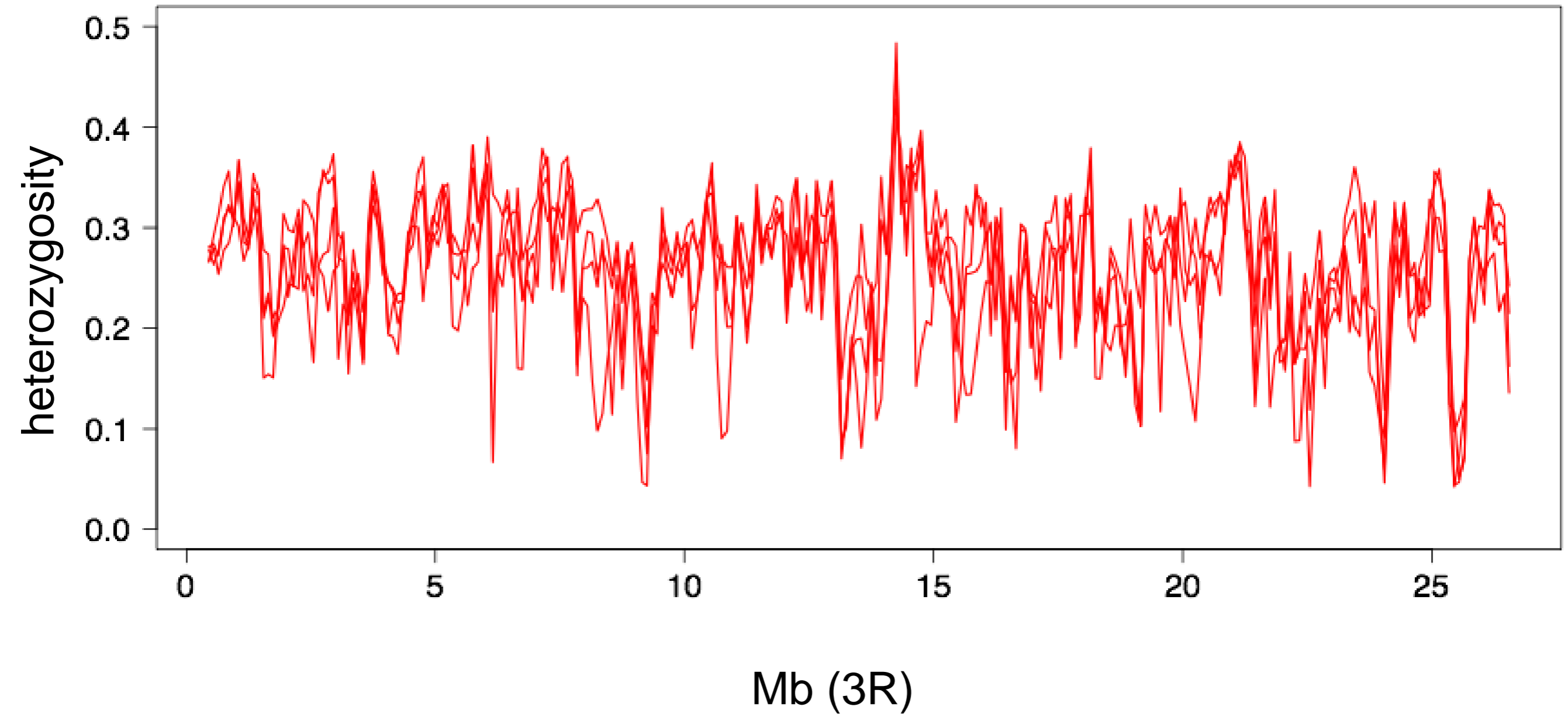
more losses of heterozygosity in **ACO**

local losses correspond to differentiated regions

ACO₁ heterozygosity resembles **ACO** pool

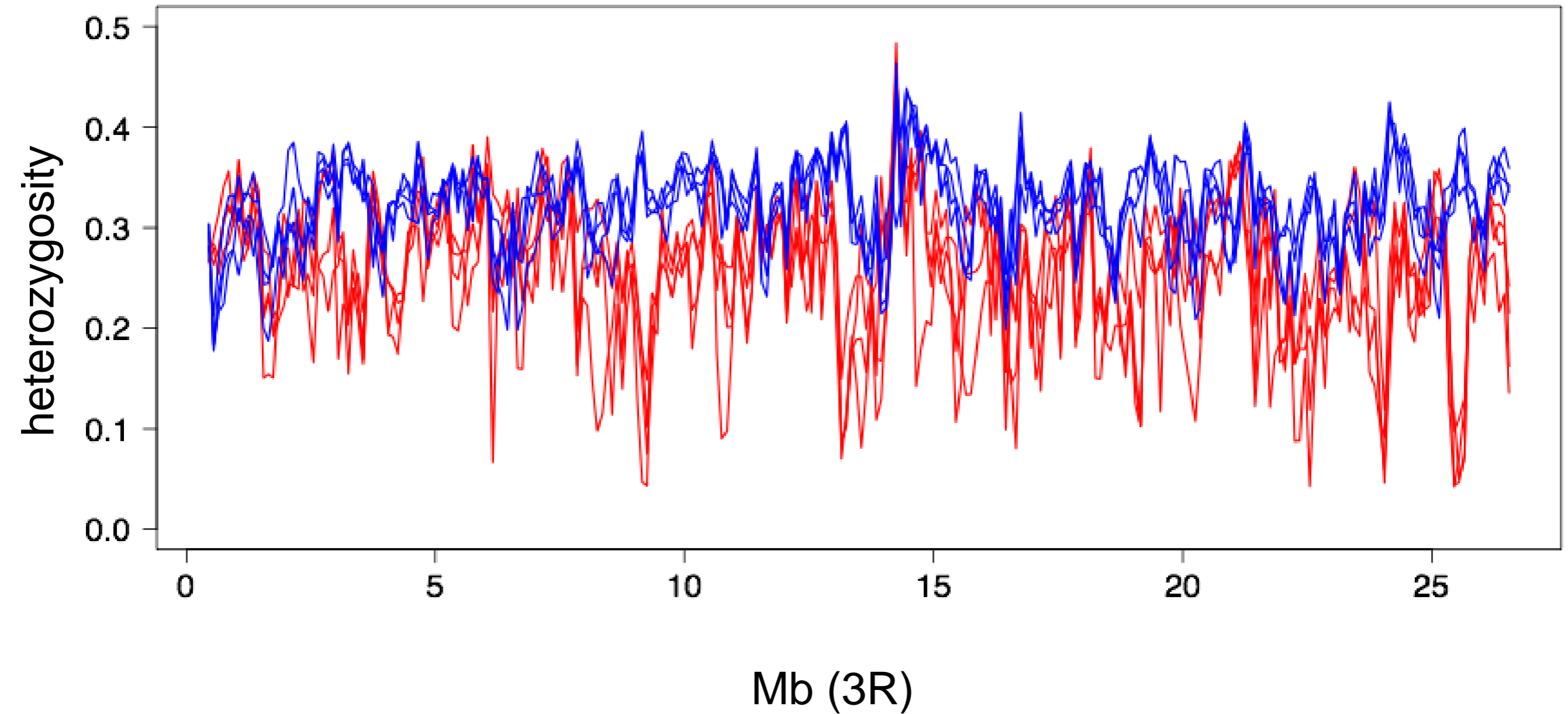


~50X coverage per replicate



data from Graves *et al.* 2016 **MBE**

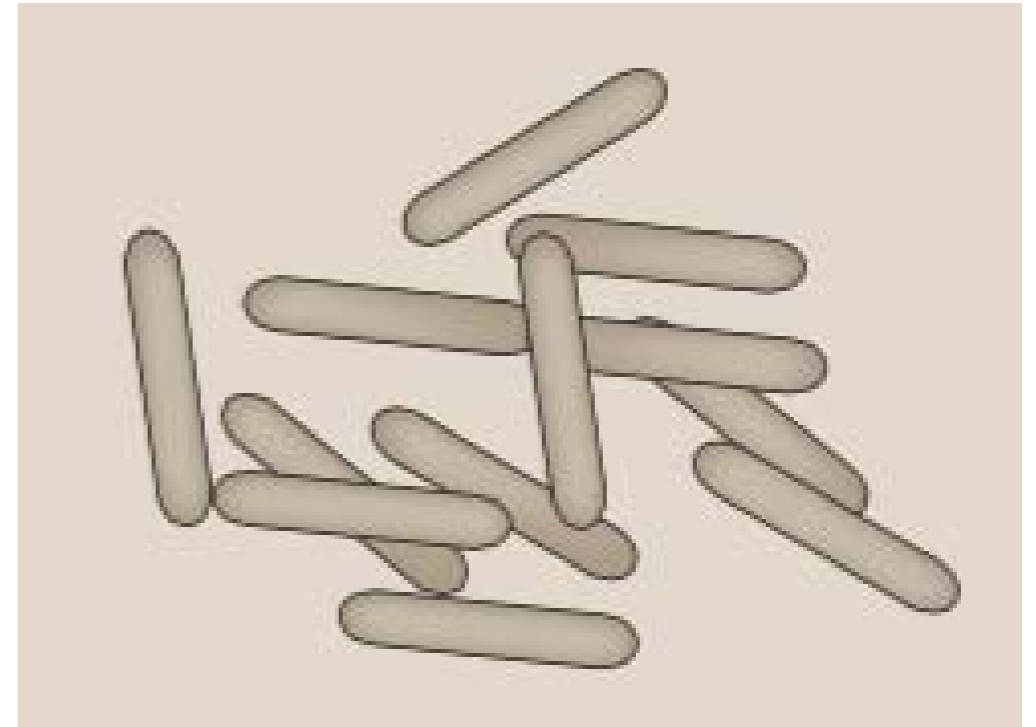
~50X coverage per replicate



data from Graves *et al.* 2016 **MBE**

conclusions

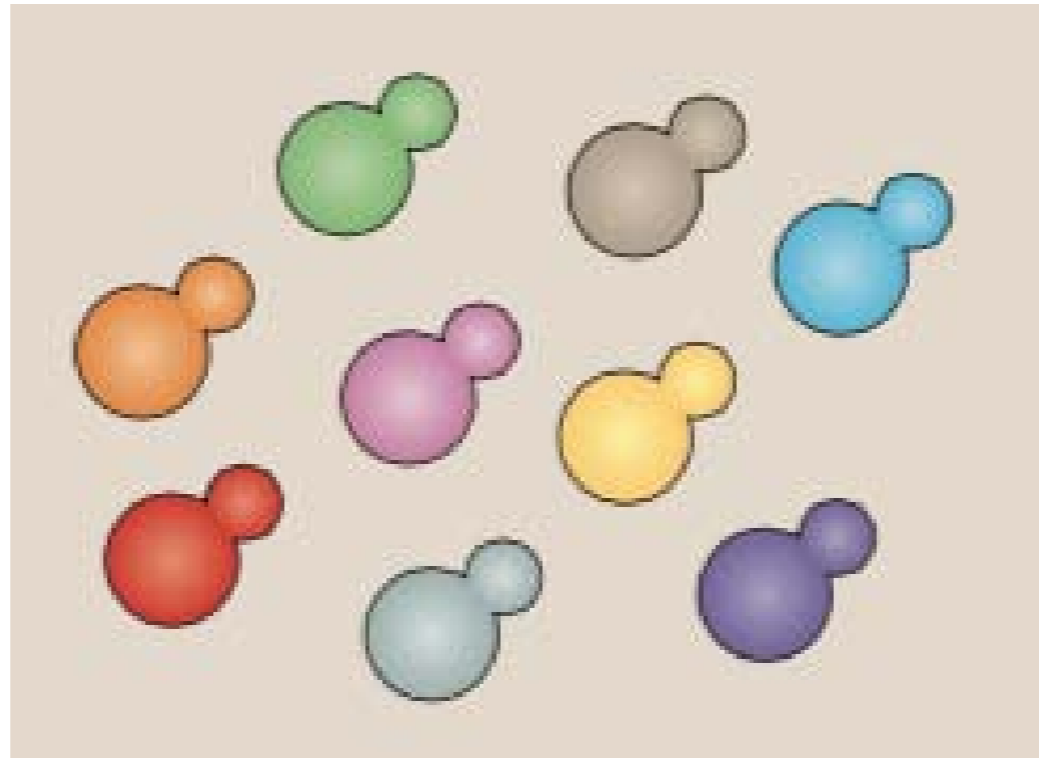
- how repeatable is evolution?
- how well can we localize/identify QTL?
- what are the origins and fates of adaptive alleles?



- eukaryote
- sexual recombination

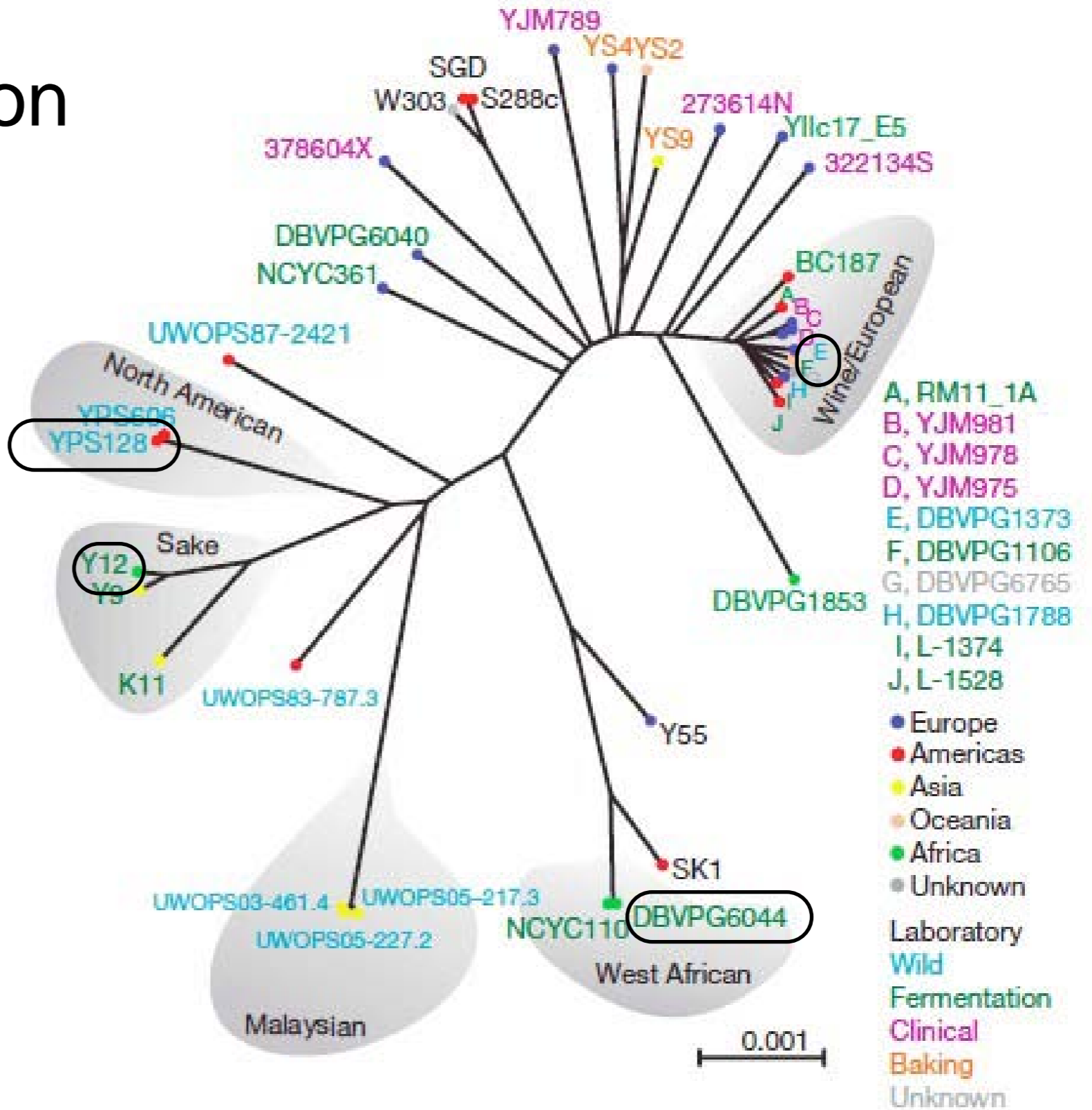
- short generations
- archivable

Saccharomyces cerevisiae



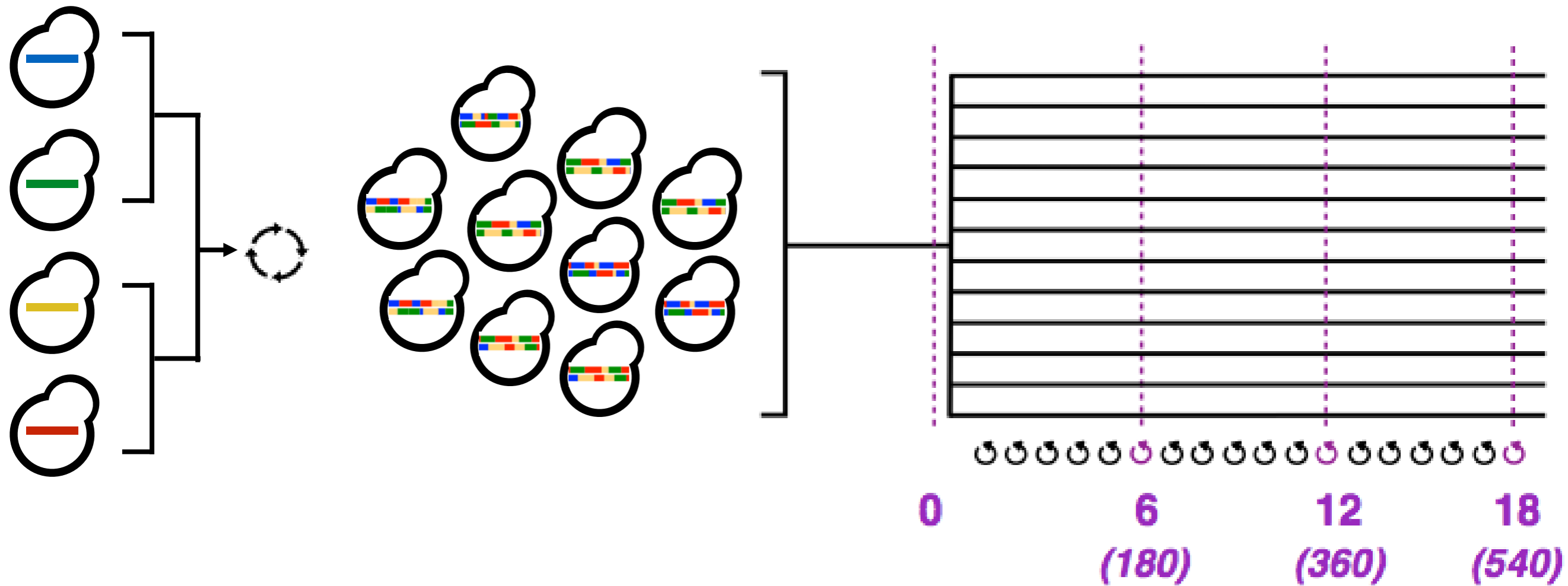
- eukaryote
- sexual recombination
- short generations
- archivable

recombinant population
with 4 founders



Liti *et al.* 2009 **Nature**

Cubillos *et al.* 2013 **Genetics**



selection treatment = regular outcrossing

15 candidate de novo mutations

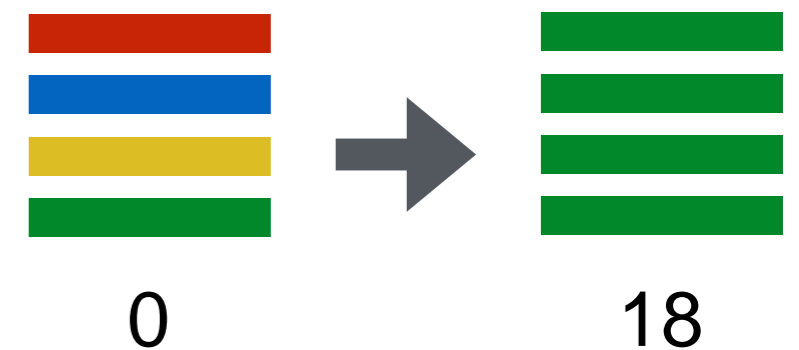
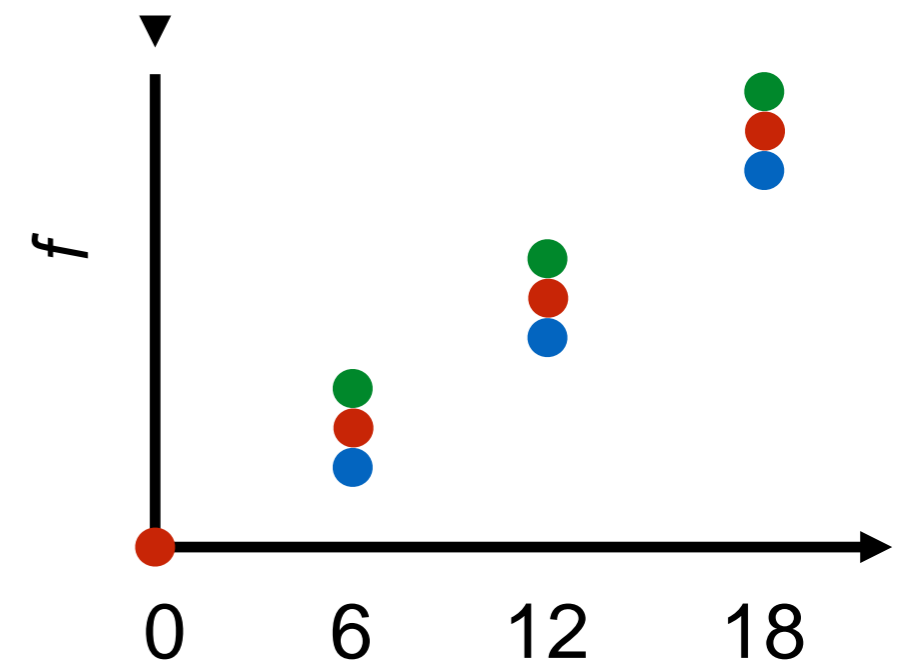
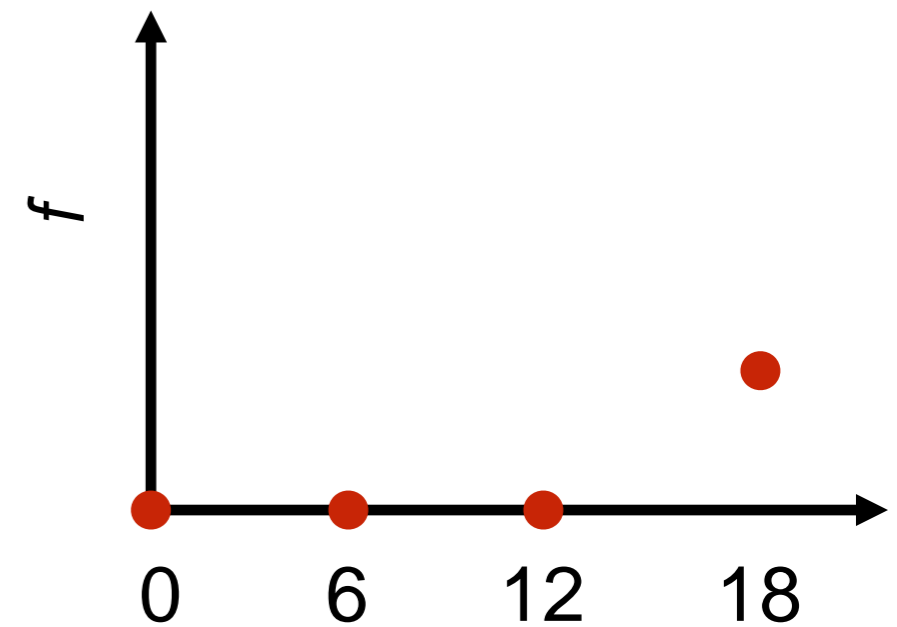
- none > 0.2 in any replicate
- none private to any replicate

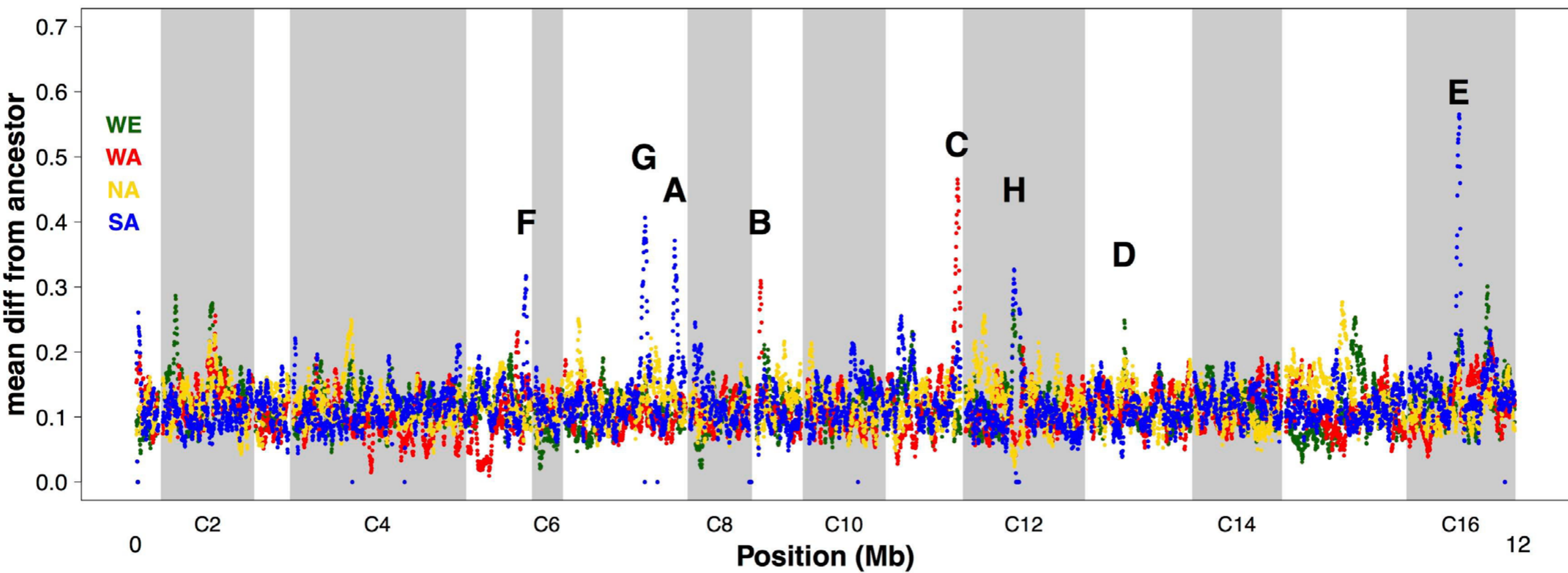
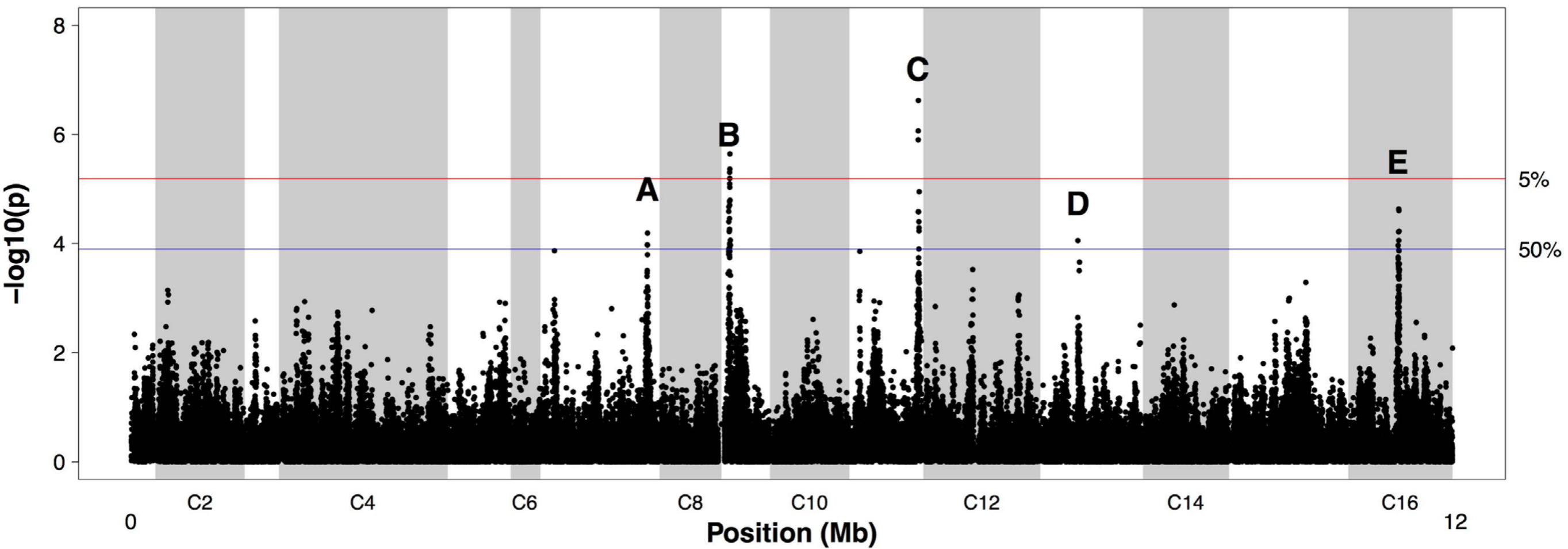
~75K standing variants

- fit linear models
- permutation tests

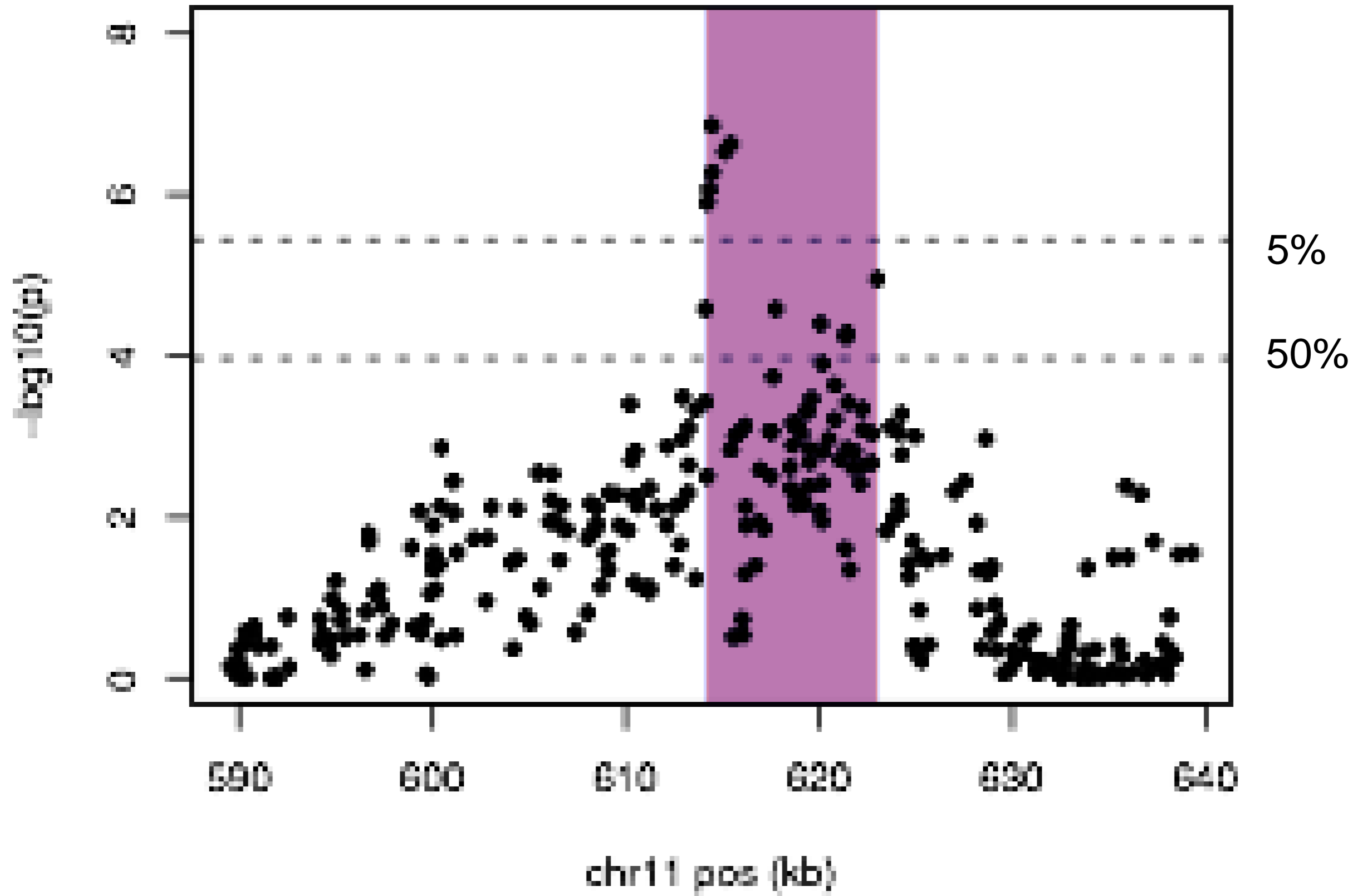
haplotype frequencies

- founder alleles known
- enrichment in evolved lines

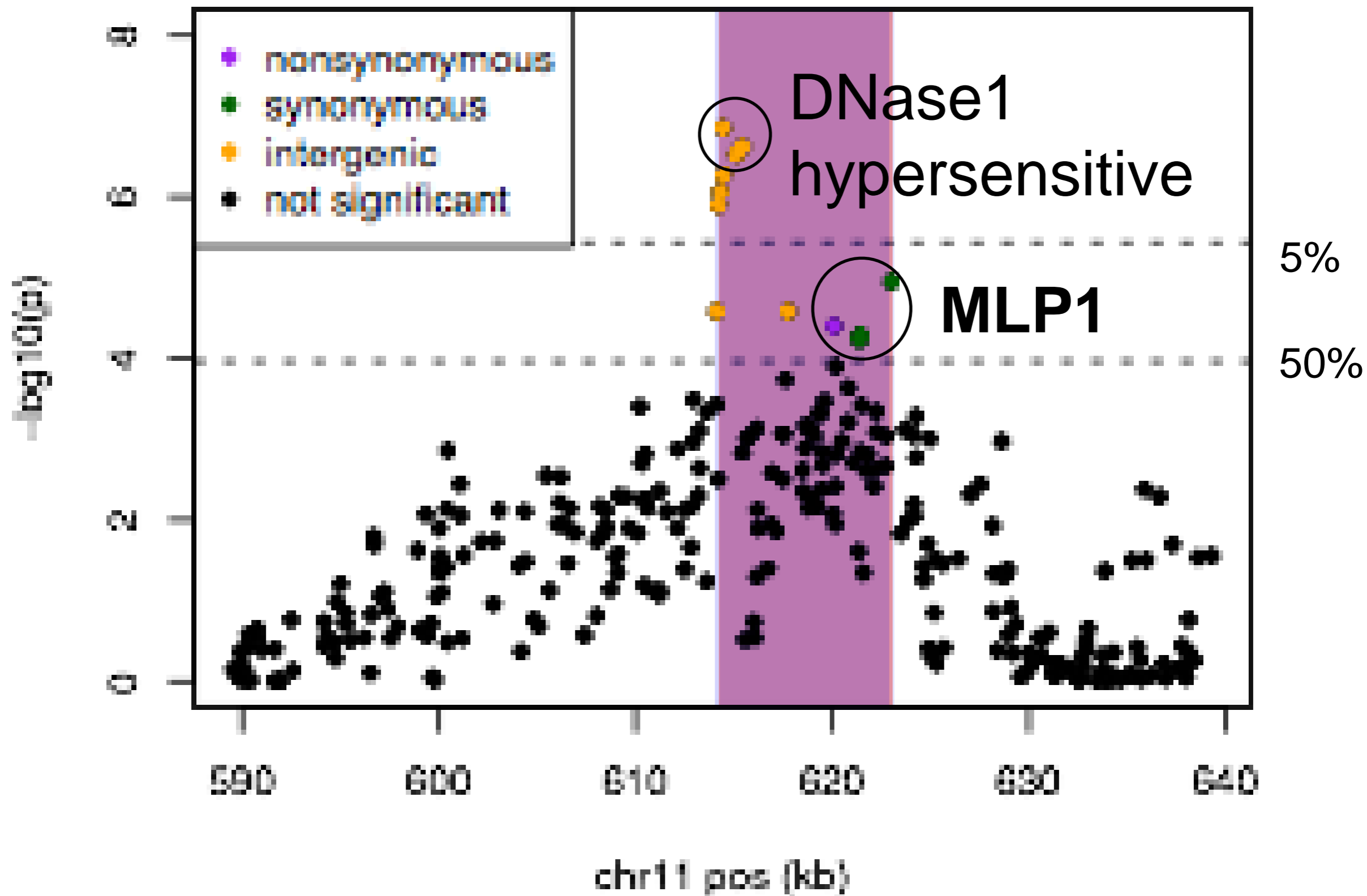




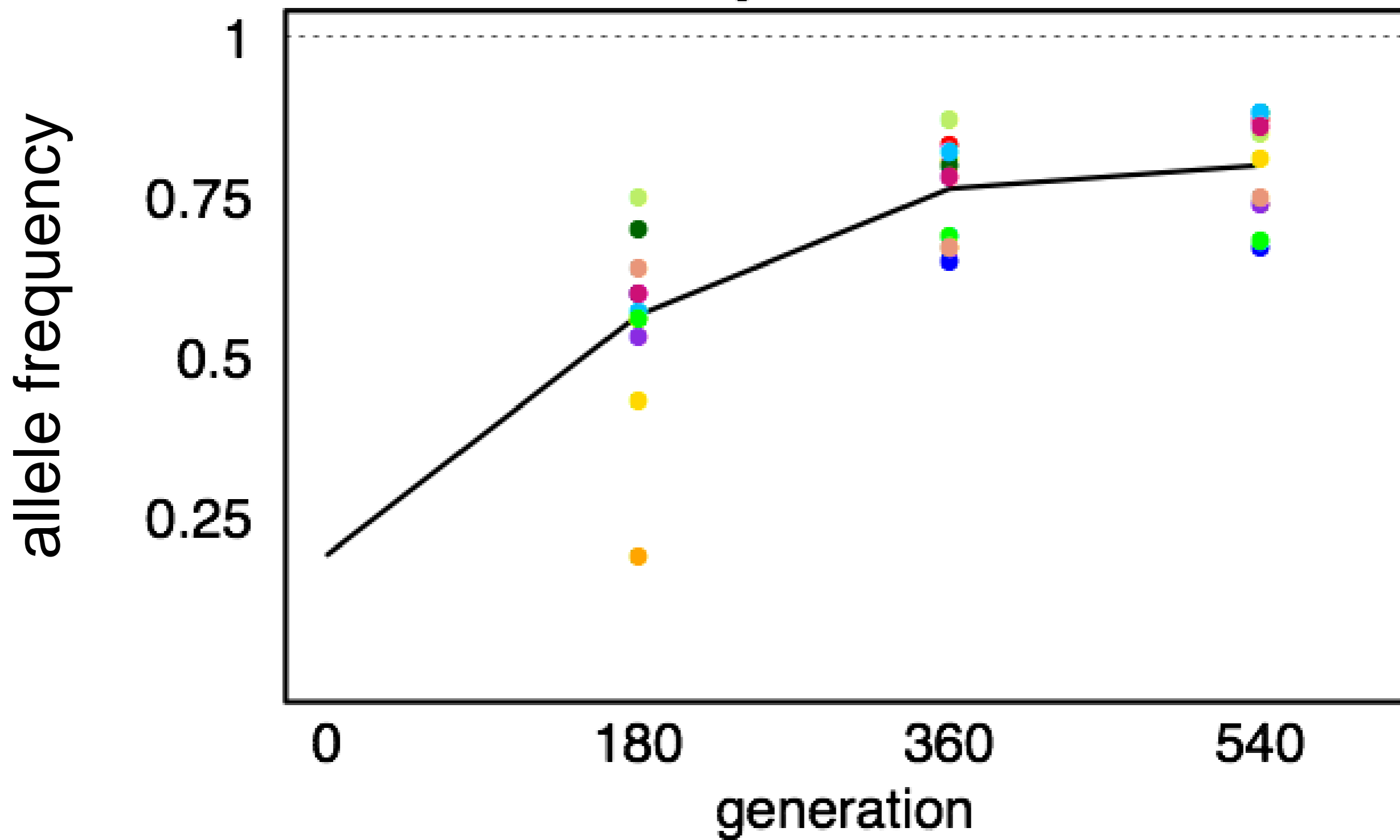
peak C



peak C



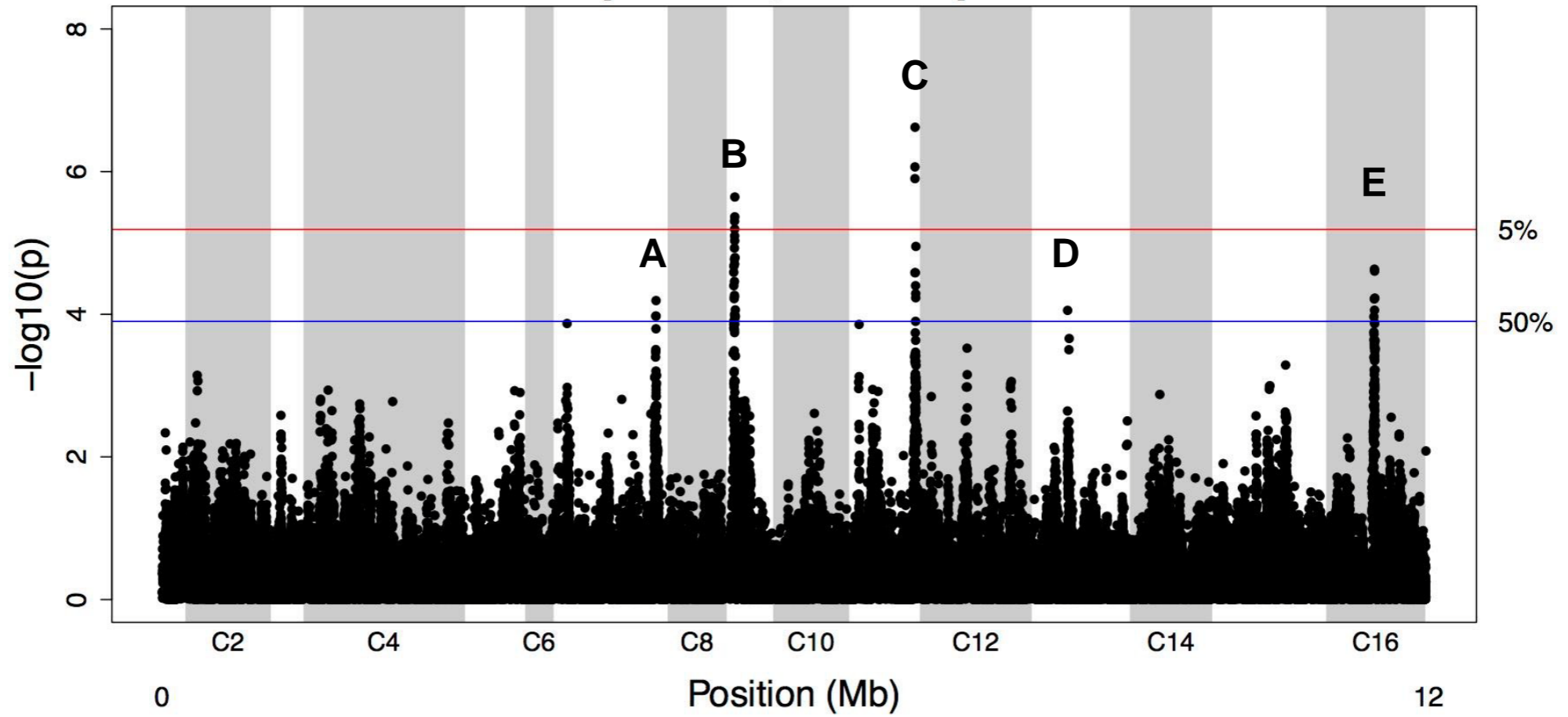
peak C



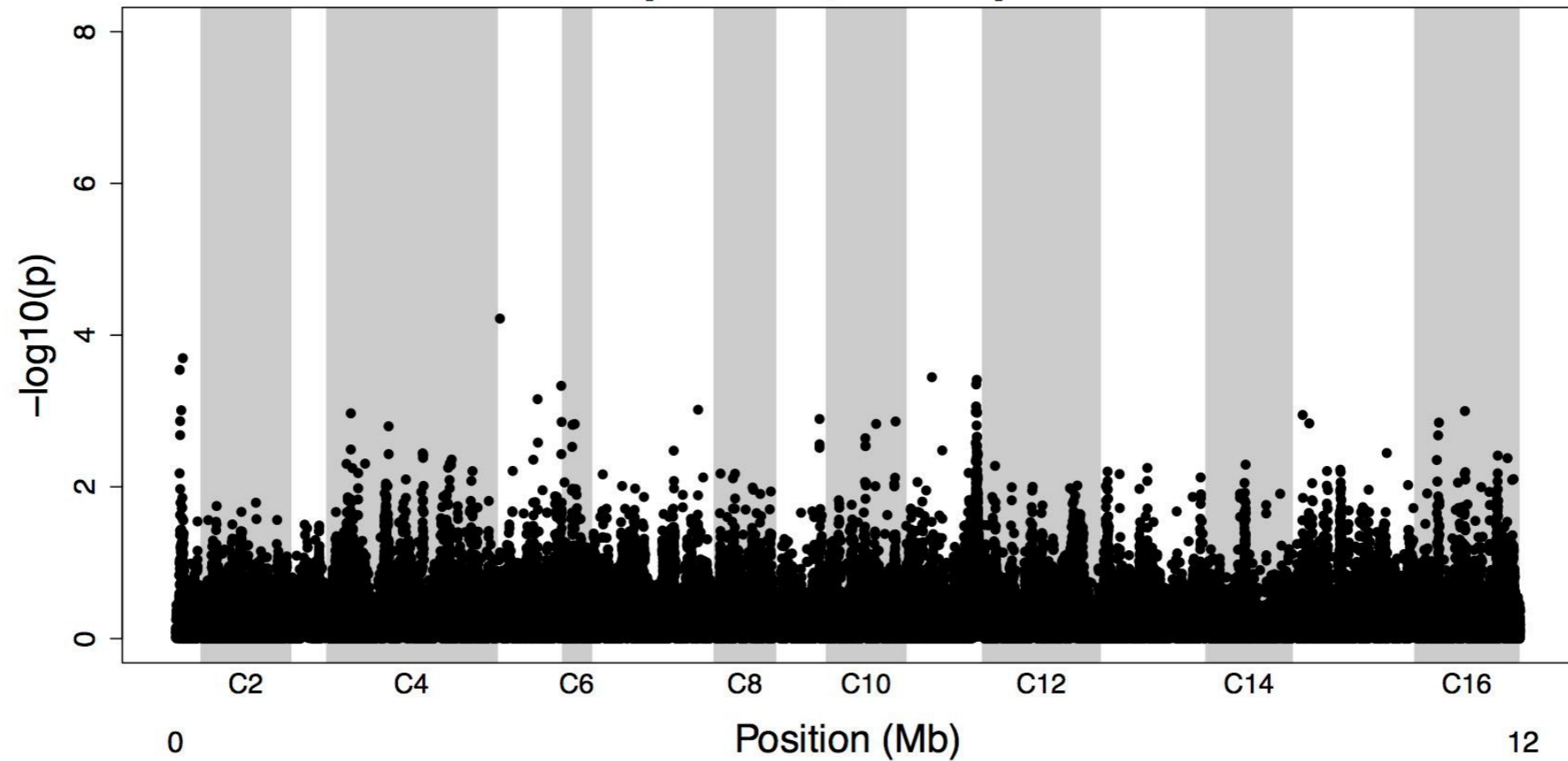
conclusions

- how repeatable is evolution?
- how well can we localize/identify QTL?
- what are the origins and fates of adaptive alleles?

12 replicates, all timepoints



5 replicates, 2 timepoints



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Associate editor: John Parsch

Abstract

Standing genetic variation provides a rich reservoir of potentially useful mutations facilitating the adaptation to novel

environment can also be used in experiments about the divergence of populations. Experimental designs that stratify populations into different replicates are more important for identifying causal variants. Our study identifies a set of trajectories that

Key words: evolution, adaptation, genetic variation, experimental evolution, quantitative trait loci

Introduction

The importance of natural selection in the evolution of quantitative traits has been well established (Falconer & Mackay 1996; Purcell 2003). Experimental evolution studies have shown that natural selection can act on standing genetic variation to drive adaptation (Turner et al. 2011; Remolina et al. 2010; Felsenstein 1981). Experimental evolution studies promise to be a powerful tool for dissecting complex traits and understanding the potential of standing genetic variation (molecular evolution). These loci are often not optimized to the current environment (Barton 1992). Whether a locus is of neutral or deleterious selection, whether E&R studies can identify these loci and evaluate

The Power to Detect Quantitative Trait Loci Using Resequenced, Experimentally Evolved Populations of Diploid, Sexual Organisms

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Associate editor: John Parsch

Abstract

A novel approach for dissecting complex traits is to experimentally evolve laboratory populations under a controlled environment shift, resequence the resulting populations, and identify single nucleotide polymorphisms (SNPs) and/or genomic regions highly diverged in allele frequency. To better understand the power and localization ability of such an evolve and resequence (E&R) approach, we carried out forward-in-time population genetics simulations of 1 Mb genomic regions under a large combination of experimental conditions, then attempted to detect significantly diverged SNPs. Our analysis indicates that the ability to detect differentiation between populations is primarily affected by selection coefficient, population size, number of replicate populations, and number of founding haplotypes. We estimate that E&R studies can detect and localize causative sites with 80% success or greater when the number of founder haplotypes is over 500, experimental populations are replicated at least 25-fold, population size is at least 1,000 diploid individuals, and the selection coefficient on the locus of interest is at least 0.1. More achievable experimental designs (less replicated, fewer founder haplotypes, smaller effective population size, and smaller selection coefficients) can have power of greater than 50% to identify a handful of SNPs of which one is likely causative. Similarly, in cases where $s \geq 0.2$, less demanding experimental designs can yield high power.

Key words: simulation, QTL detection, genomics, adaptive evolution, experimental evolution, evolve and resequence.

Introduction

Quantitative traits are of special interest to biologists. The variation in many traits of medical, agricultural, and evolutionary relevance is due to the concerted action of several genes and the environment. Quantitative trait locus (QTL) mapping has been effective at explaining the majority of the heritability of a trait but is poorly suited to resolving the location of QTL beyond several cM (Mackay et al. 2009). More recently, several groups have attempted to increase the resolution of QTL mapping using advanced generation recombinant inbred lines (cf. Kover et al. 2009; Aylor et al. 2011; King et al. 2012), but resolution is still limited to cM scales. Recently, genome wide association studies (GWAS) have become a major method for investigating the genetic basis for quantitative traits (The Wellcome Trust Case Control Consortium, 2007a, 2007b; Craddock et al. 2010). Although GWAS studies have identified replicable associations between SNPs and complex traits, associated SNPs tend to explain only a small fraction of the heritable variation in the study trait (Manolio et al. 2009), a problem that cannot be solved by increasing sample sizes to tens of thousands of individuals (Signer-Hasler et al. 2012) or replacing SNPchips with complete resequenced genomes (Spencer et al. 2009). Clearly, it is of value to explore novel methods for dissecting complex traits.

In systems that have short generation times and that can easily be reared in the lab in large numbers, an alternative experimental approach to dissecting complex traits has been

to "evolve and resequence" (E&R) populations of organisms. E&R studies have been performed with both asexual (Riehl et al. 2001; Barrick et al. 2009; Kishimoto et al. 2010; Tenallion et al. 2012; Parts et al. 2011) and sexual (Teotónio et al. 2009; Burke et al. 2010; Johansson et al. 2010; Turner et al. 2011; Orozco-Terwengel et al. 2012; Turner and Miller 2012) populations. Because asexual experimental evolution lacks recombination and standing variation in the base population, the footprints of selection in the genome and the means by which an investigator may hope to identify causal variants are different in sexual and asexual systems. Thus, we limit our focus to E&R studies in sexual systems. Under the E&R paradigm, a base population is divided into several replicate populations, half of which are subjected to a well-defined selection pressure, and the other half of which are maintained without selection. Next, the DNA pools from each population are resequenced using NextGen technology and allele frequencies in each pool are estimated. SNPs and/or genomic regions showing consistent differentiation between selected and control population are candidates for harboring causative variants. Studies using this design have claimed to detect numbers of candidate causative sites (CS) from 662 (Burke et al. 2010) to almost 5,000 (Orozco-Terwengel et al. 2012) for various quantitative traits. Currently, the CSs detected by E&R methods have not been validated.

To date, the field of E&R has been almost entirely empirically motivated. Study designs have varied greatly in terms of

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best practices:

$N_e > 1000$

replicates > 25

generations > 500

Article

Article

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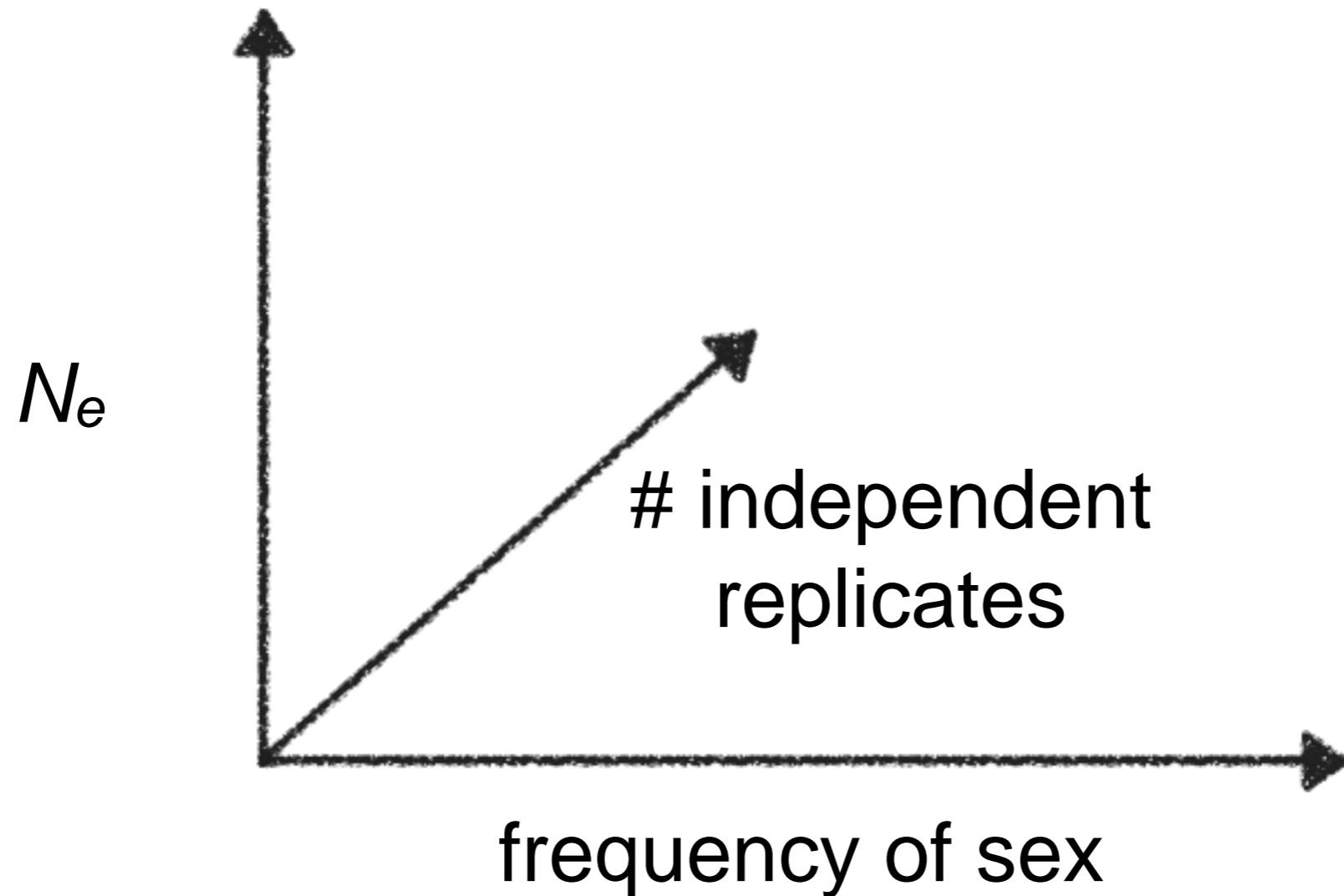
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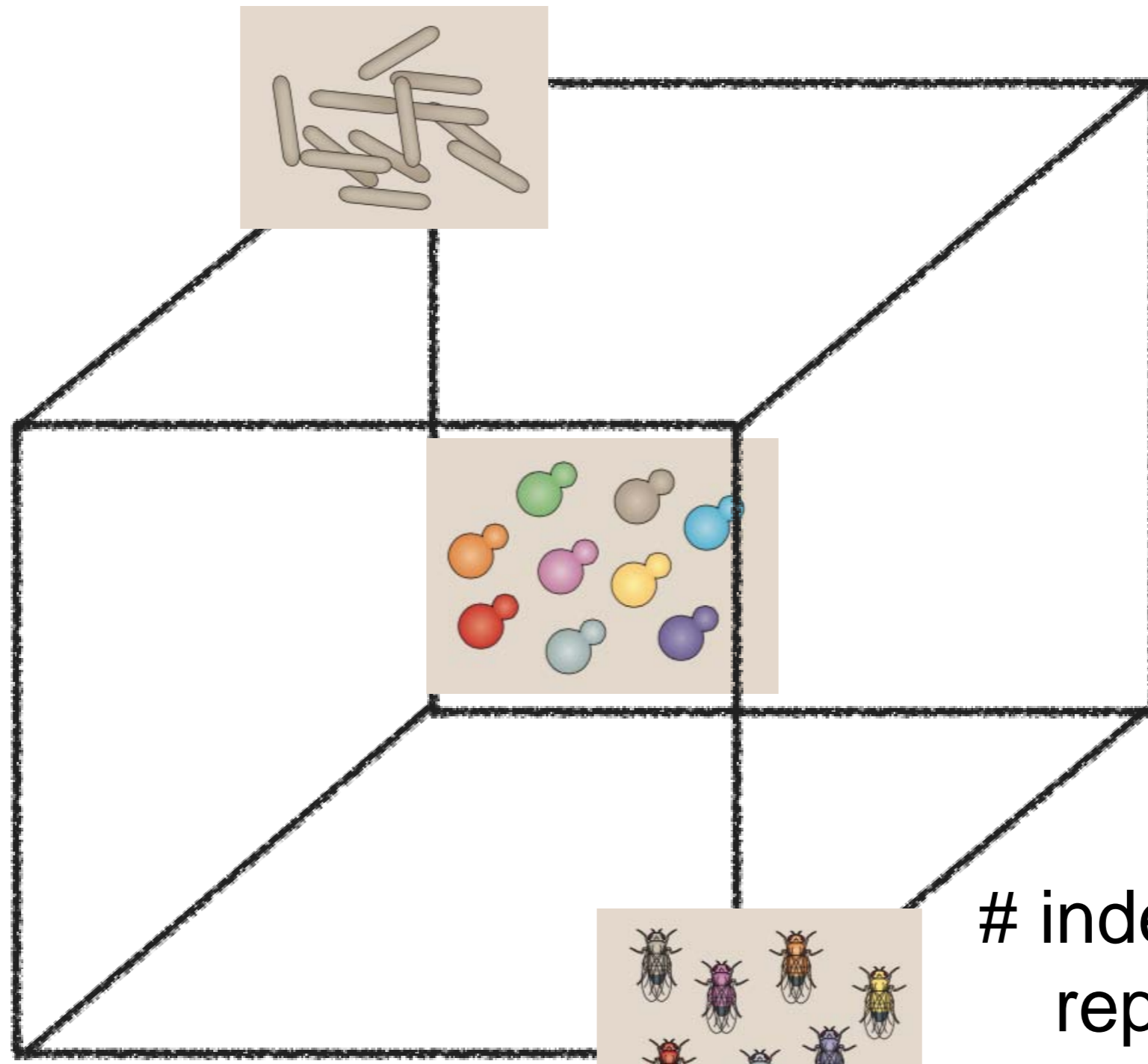
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experimental evolution parameter space



Ne



frequency of sex

independent replicates

acknowledgements

UC Irvine:

Michael Rose

Tony Long

Kevin Thornton

Laurence Mueller

Parvin Shahrestani

Mark Phillips

Larry Cabral

Thomas Barter

USC:

Joseph Dunham

NC A&T State U:

Joseph Graves

UT Tyler:

Kate Hertweck

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Cold
Spring
Harbor
Laboratory