

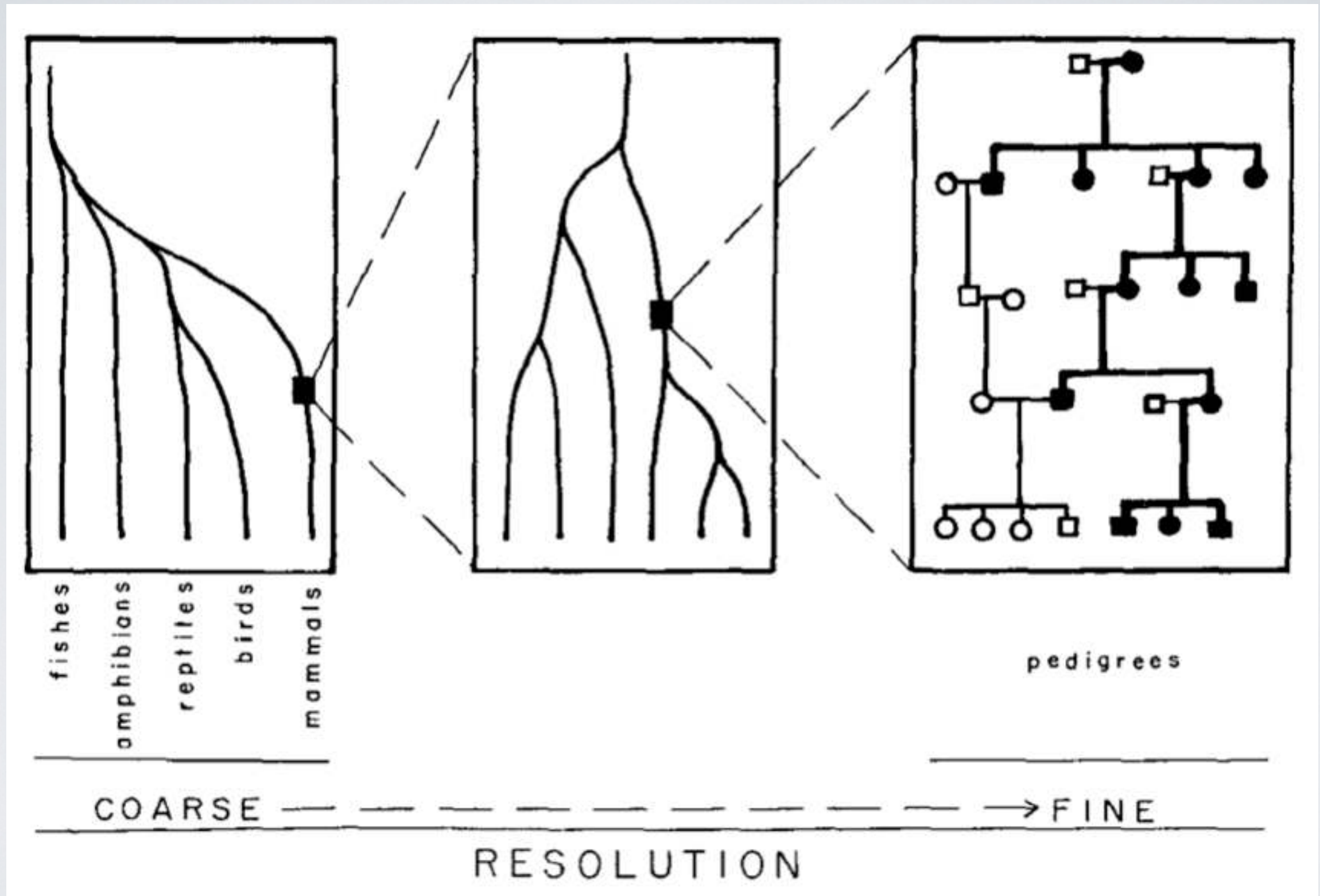
Genomic analyses of non-model organisms with RAD-seq and Stacks

Julian Catchen
jcatchen@uoregon.edu

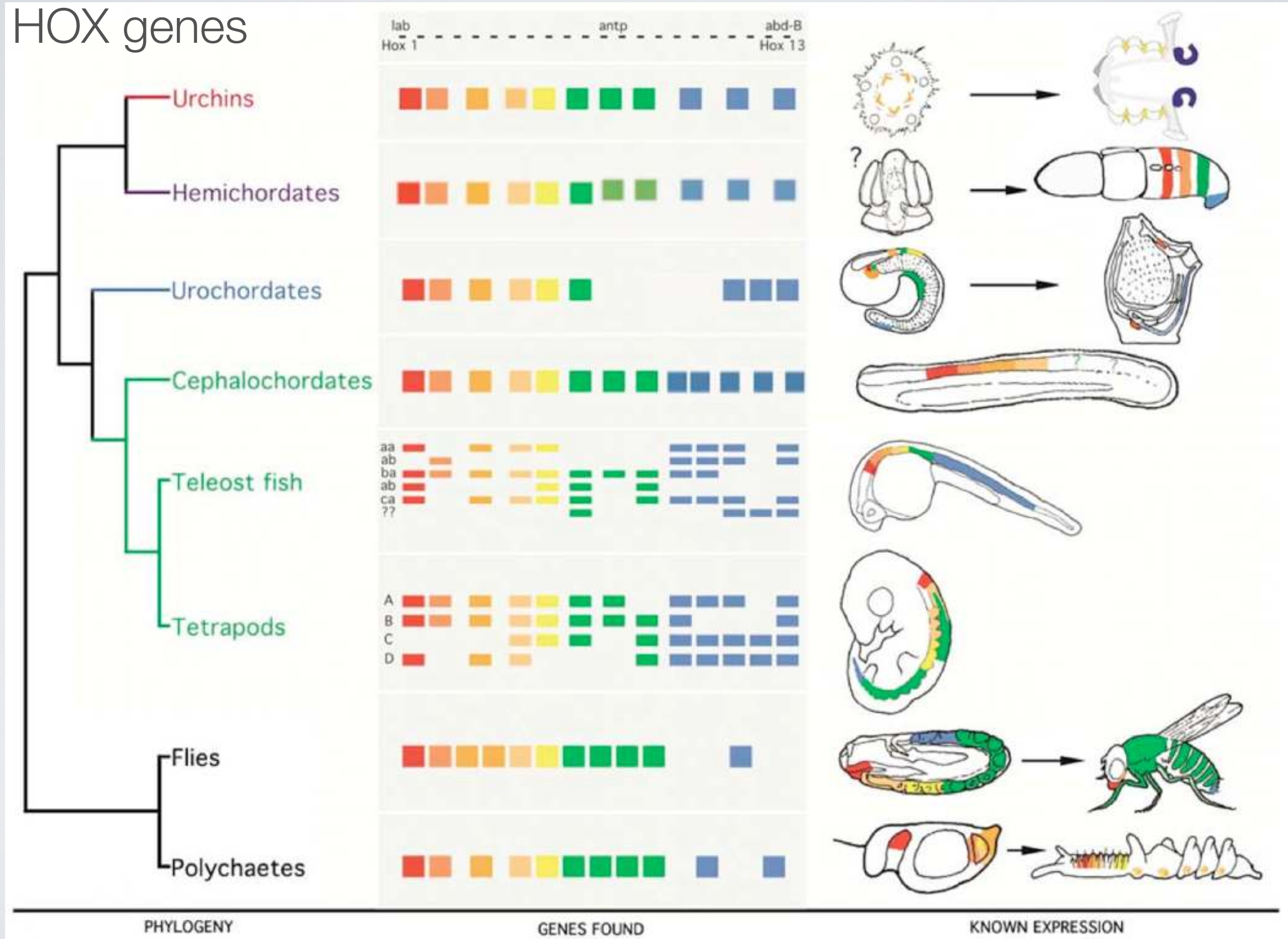


Cresko Lab
Institute of Ecology and Evolution
University of Oregon

“A bridge between population genetics and systematics”



A common genetic toolkit



What makes an organism a model?

classic models: fruit fly, mouse, zebrafish

1. Genetic Maps

(phenotypes, allozymes, microsats)

2. Physical Maps

(random, shotgun Sanger sequencing)

3. Transcriptomic Maps

(EST sequencing)

4. Gene Expression Analyses

(microarrays, RNA-seq)

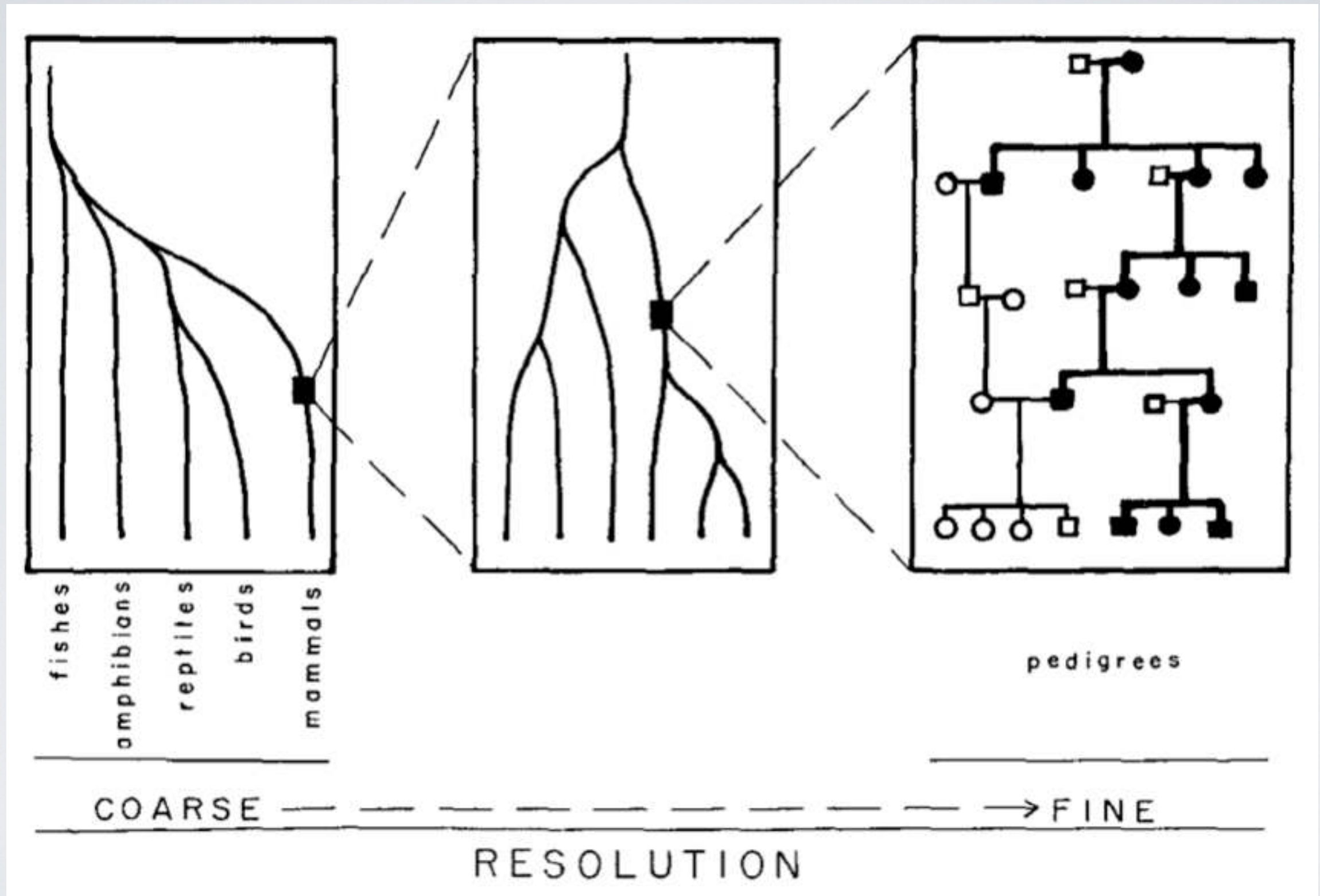


What makes an organism a model?

Organism	Markers	Organism	Markers
Bighorn Sheep	147	Silver carp	483
Shrimp	418	Guppy	790
Coral	420	Barramundi	240
Indian mustard	1,029	Catfish	331
Oilseed rape	13,551	Sea bass	368
Black spruce	1,111	Cichlid	204
Barley	2,111	Platyfish	290
Flounder	1,375	Halibut	604
Turbot	242	Sea bream	204
Human	250,000	Mouse	10,000

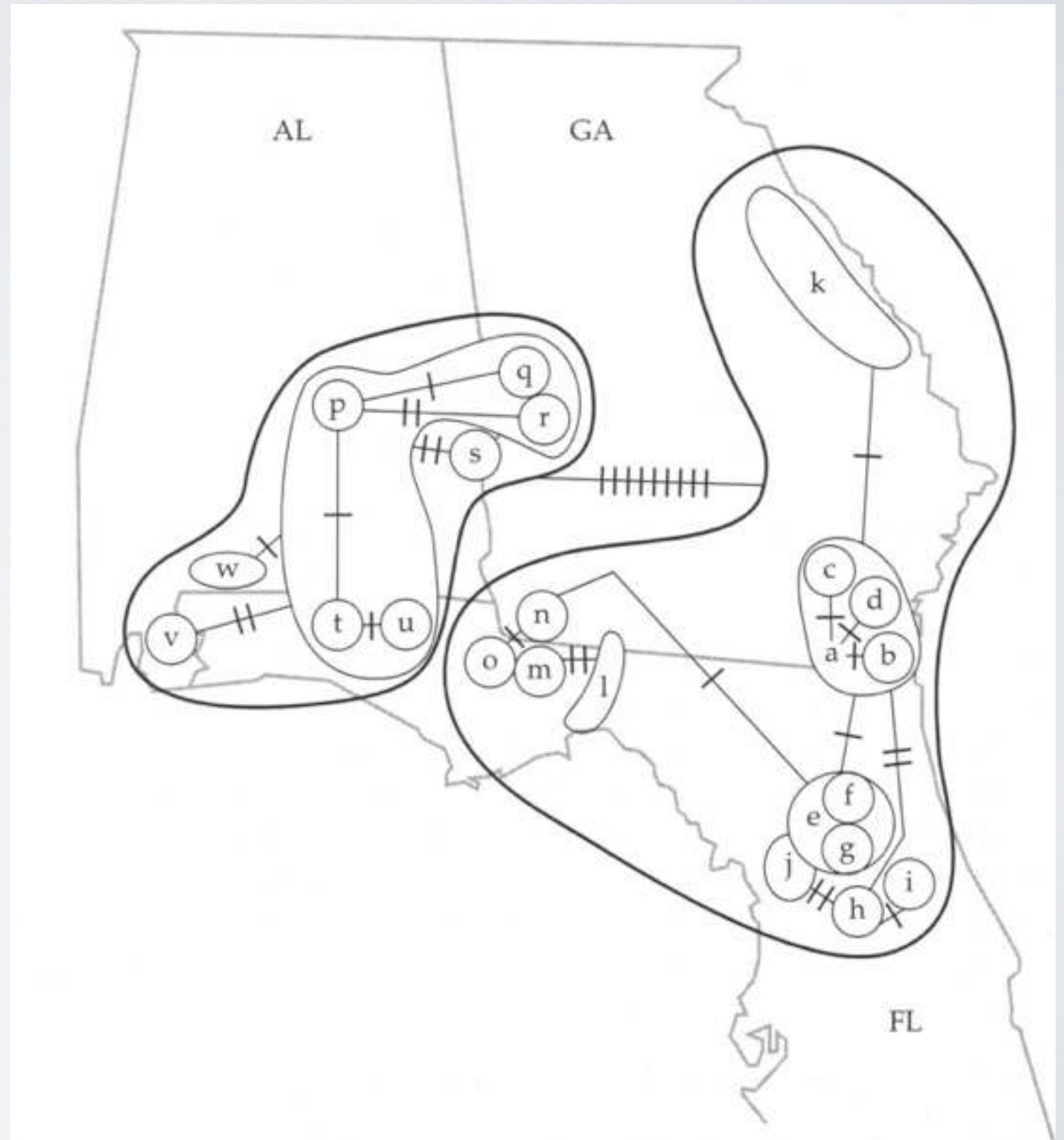
**Short-read sequencing represents a new
“bridge”**

“A bridge between population genetics and systematics”

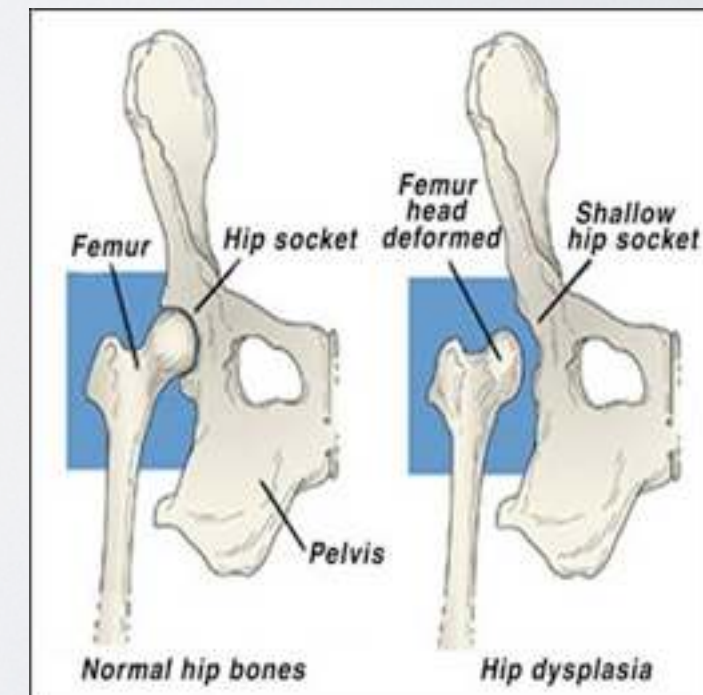


Fine scale estimates of population structure
Estimates of genome-region specific gene flow

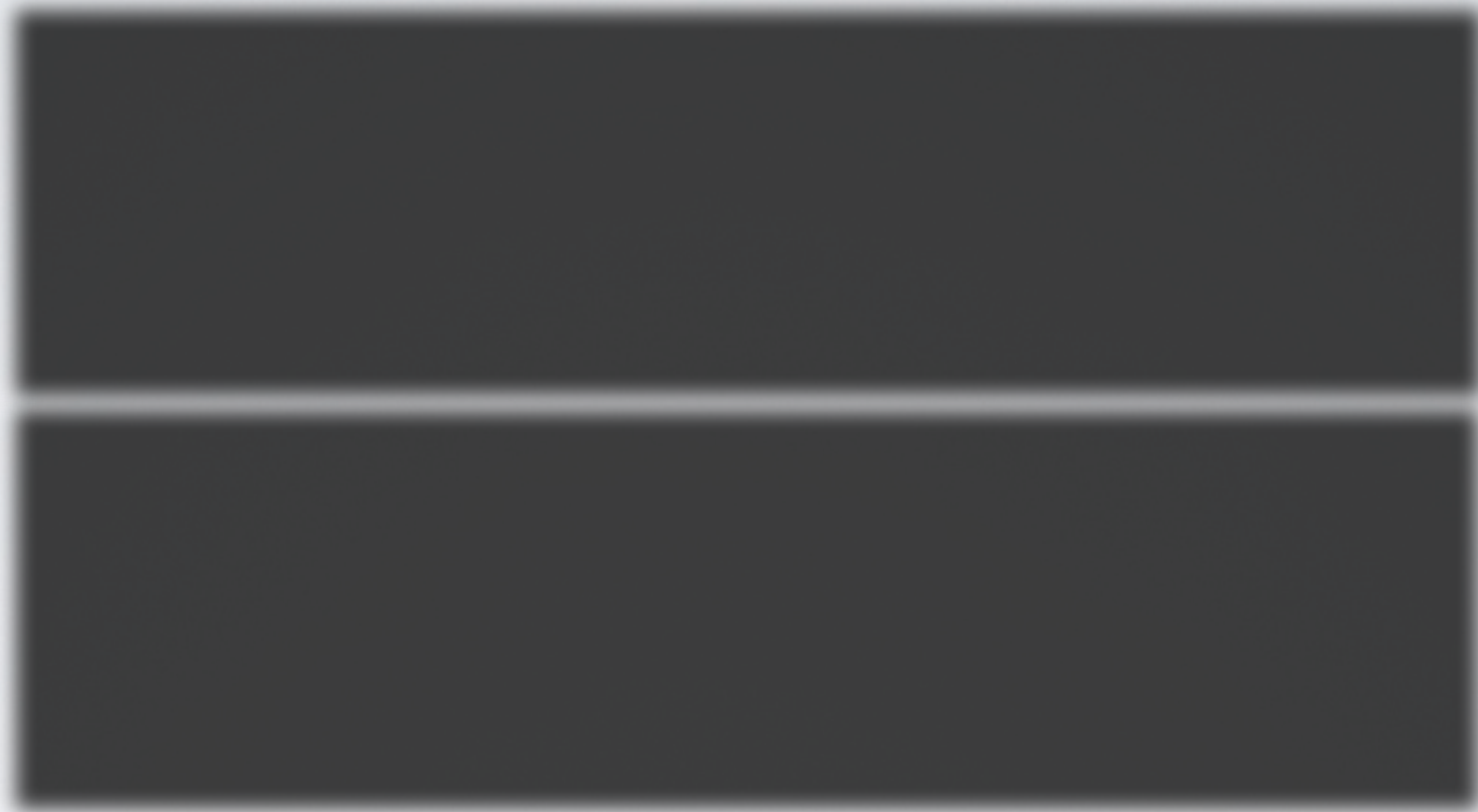
Phylogeography of pocket gophers using mtDNA



Precisely quantify the amount of inbreeding in wild and captive populations



Identify signatures of selection in natural populations
Associate genotype to phenotype using GWAS and QTL



Fill important taxonomic nodes with genetic maps
Quantify genome evolution and structural variation



Sequence everything?

Why not sequence the entire genome?

Sample space has scaled with sequencing capacity

- Human height GWAS; over 15,000 individuals assayed
- Identified many new regions contributing to the variation
- Still only identified a fraction of the heritability

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For many studies, full sequence isn't necessary

- the genomes of many organisms are organized in linkage blocks
- well spaced markers will provide all the necessary coverage
- the cost of genotyping will almost always be a fraction

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Genetic maps are relevant again

- a high density genetic map can facilitate genome assembly
- genomes may be segregating a lot of structural variation
- assaying this variation can presently be done best with genetic maps

Alternative approach - Reduced representation sequencing for genotyping

- Focus the sequencing on a homologous set of tags spread throughout the genome

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- Can allow thousands of genomes to be assayed in just a few weeks
- WHY NOT - some cases complete genomic sequence is necessary
 - when linkage disequilibrium blocks (LD) are very short
 - If the genetic architecture of a trait involves many rare variants
 - Inferring patterns of LD may be easiest with full sequences

Outline

1. RAD-seq and other RRL methods
 - Experimental design considerations
2. Anatomy of a PstI RAD-seq Analysis
3. Signatures of natural selection across the genome with the threespine stickleback
 - SNP detection and kernel Smoothing
4. Building a genetic map of the spotted gar
 - Stacks: the gory details
5. Assembling the genome of the platyfish
6. Phylogeography of the pitcher plant mosquito
7. RAD Synthesis with the gulf pipefish
8. Gene expression quantification with eRAD

Different types of reduced representation sequencing for genotyping

Nature Reviews Genetics, 2011

**Genome-wide genetic marker
discovery and genotyping using
next-generation sequencing**

John W. Davey^{}, Paul A. Hohenlohe[†], Paul D. Etter[§], Jason O. Boone^{||},
Julian M. Catchen[†] and Mark L. Blaxter^{*†}*

Different types
of reduced
representation
sequencing for
genot

CRoPS: similar to RRL
uses AFLP: two restriction enzymes

MSG: similar to GBS
replaces size selection step with very
frequent cutter

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What is RAD-seq?

(**R**estriction-site **A**ssociated **D**NNA)

Eric Johnson



Joe Dunham



Mike Miller



Illumina

2007

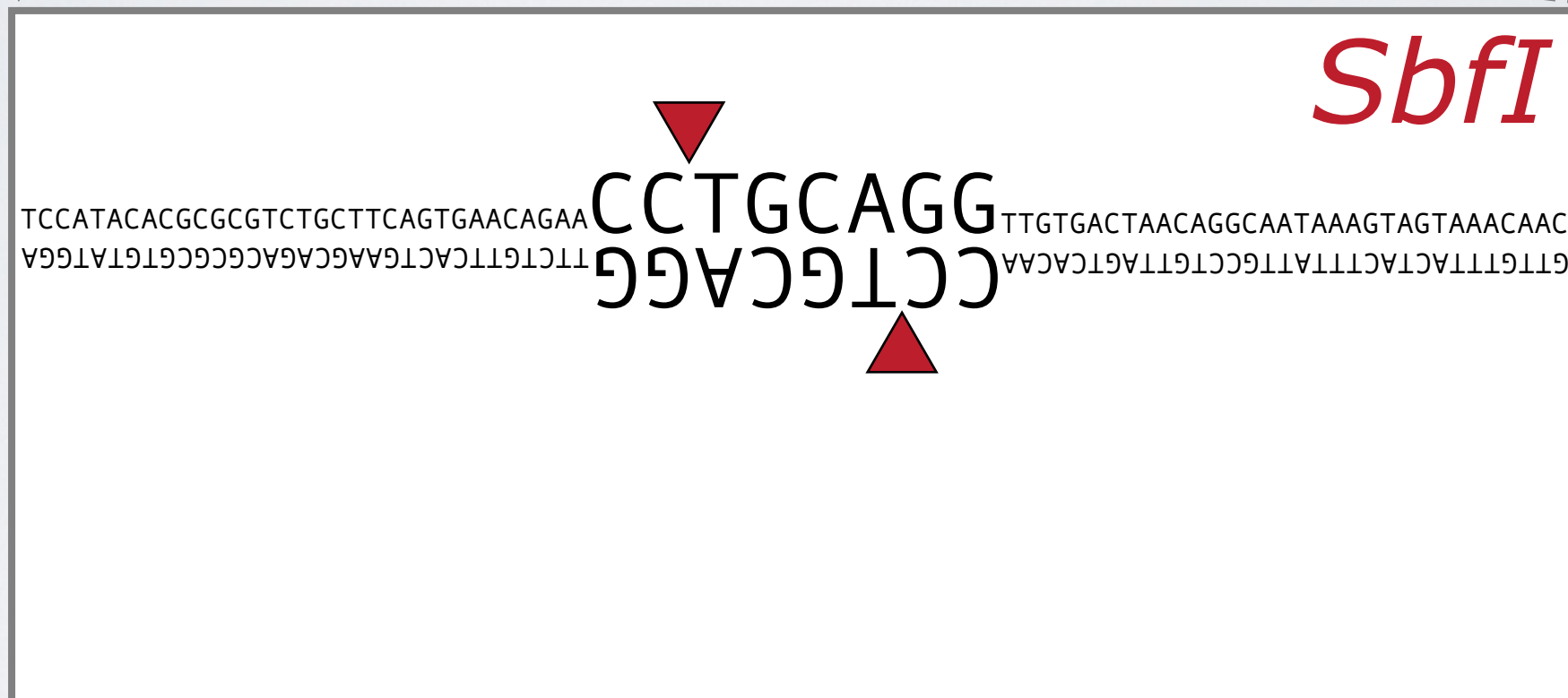


2008



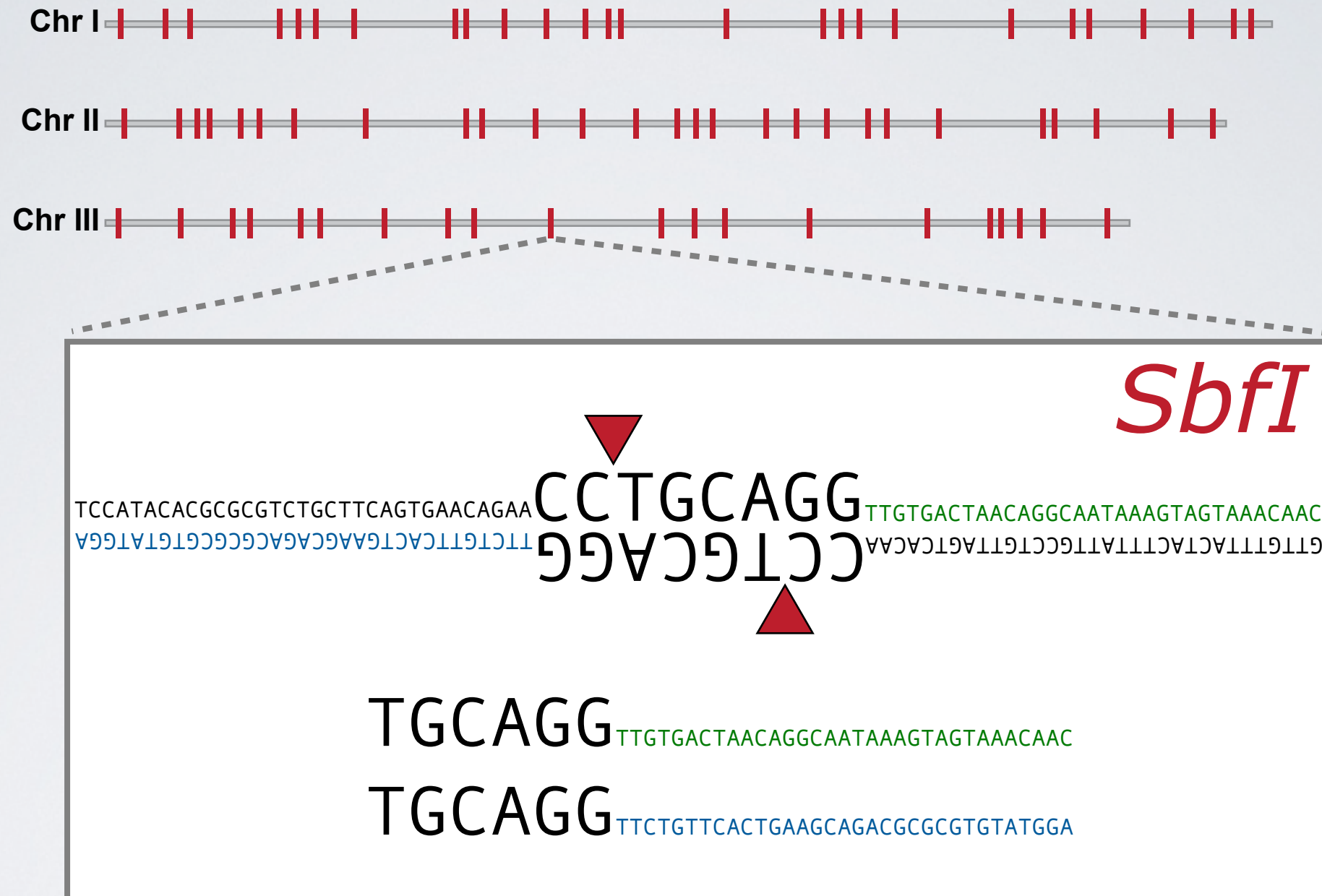
(Restriction-site Associated DNA)

The diagram illustrates the distribution of 100 SNPs across three chromosomes. Chr I has 10 SNPs, Chr II has 20 SNPs, and Chr III has 10 SNPs. The SNPs are represented by red vertical bars along the length of each chromosome.



What is RAD-seq?

(Restriction-site Associated DNA)

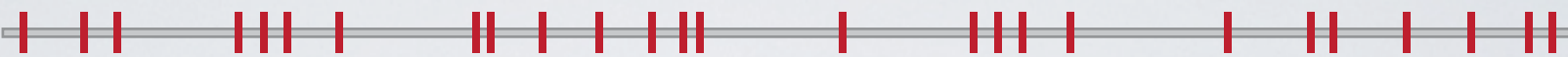


(Restriction-site Associated DNA)



What is RAD-seq?

(Restriction-site Associated DNA)

Chr I 

22,830 *SbfI* sites in Stickleback

~ 45,000 RAD-Tags

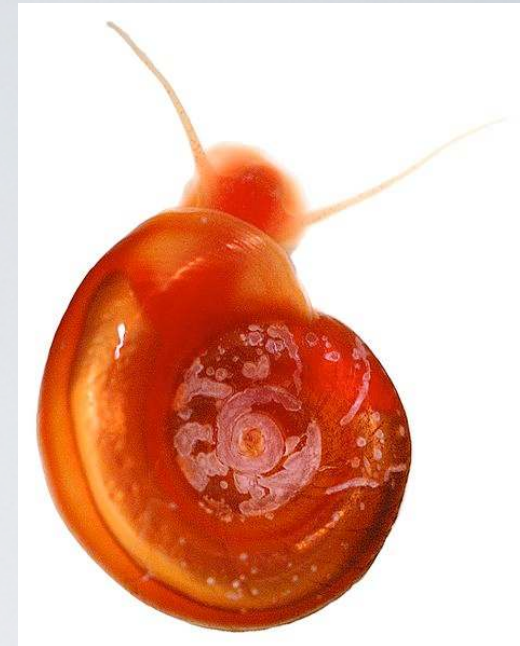
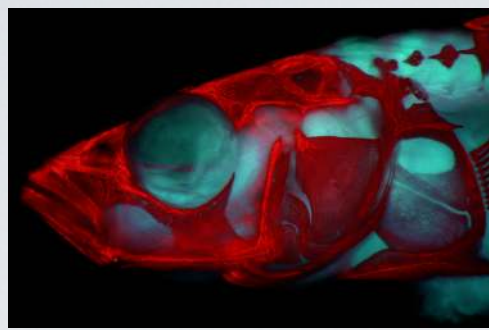
HiSeq Illumina Lane:

100 million reads, 96 barcoded individuals

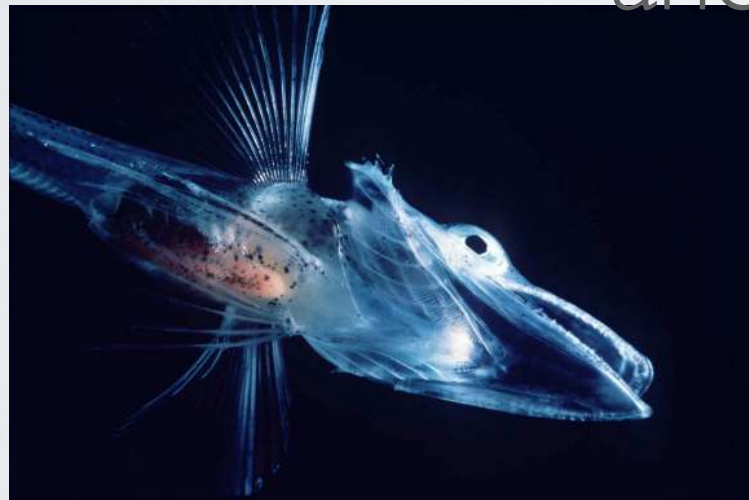
①


TCCATACACGCGCTCTGCTTCAGTGAACAGAA CCTGCAGG TTGTGACTAACAGGCAATAAAGTAGTAAACAAC
GGTGTGACTAACAGGCAATAAAGTAGTAAACAAC
TGCAGG
TGCAGG

②



Wet lab and statistical
considerations for RAD-seq
and other RRL-seq studies



Experimental design considerations for RAD

Tradeoffs:

Number of sites versus **Depth** of sequencing per site versus **Number of samples**

Experimental design considerations for RAD

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Number of sites versus **Depth** of sequencing per site versus **Number of samples**

raw reads / samples / sites = coverage at each locus

1,000,000 / 100 / 1,000 = 10x coverage

Experimental design considerations for RAD

Tradeoffs:

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Don't forget about sequencing error and sample bias

Experimental design considerations for RAD

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Number of sites versus **Depth** of sequencing per site versus **Number of samples**

How many tags do I need?

Experimental design considerations for RAD

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Number of sites versus **Depth** of sequencing per site versus **Number of samples**

How many tags do I need?

Things to consider

Choice of enzyme and genome size $(0.25)^n \times \text{genome size} = \text{expected \# sites}$

Genomes are biased:

expect 112,300 six-cutter sites in stickleback (460 Mb)	actual EcoRI sites = 90,000
expect 7000 eight-cutter sites in stickleback	actual SbfI sites = 22,800
expect 32,900 six-cutter sites in C. remanei (135 Mb)	actual EcoRI sites = 73,200

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Coding regions are GC rich - restriction recognition sequence can bias where tags fall.

e.g., in stickleback coding regions are 10% of the genome, but:
50% of SbfI sites (CCTGCAGG) fall in exons
14% of EcoRI sites (GAATTC) fall in exons



Genome-wide genetic marker discovery and genotyping using next-generation sequencing

John W. Davey^{}, Paul A. Hohenlohe[†], Paul D. Etter[§], Jason Q. Boone^{||}, Julian M. Catchen[‡] and Mark L. Blaxter^{*¶}*

Nature
Reviews
Genetics
2011

Experimental design considerations for RAD

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How many tags do I need?

Things to consider

Choice of enzyme and genome size

Polymorphism and read length

Nucleotide polymorphism rate = 0.01 to 0.001 for most vertebrates

Stickleback populations: 0.01 to 0.02. At least 1 SNP every 100 bp, on average

60 nt single end: sample 22,800 SbfI sites, 2 tags per site, 49 nt = 2.23 Mb = 22,300 SNPs

If $N = 100$, about 48% of tags have a SNP

80 nt single end: 45,600 tags, 69 nt = 3.15 Mb; 60% of tags have a SNP



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Tradeoffs:

Number of sites versus **Depth** of sequencing per site versus **Number of samples**

How many samples should be multiplexed?

Things to consider

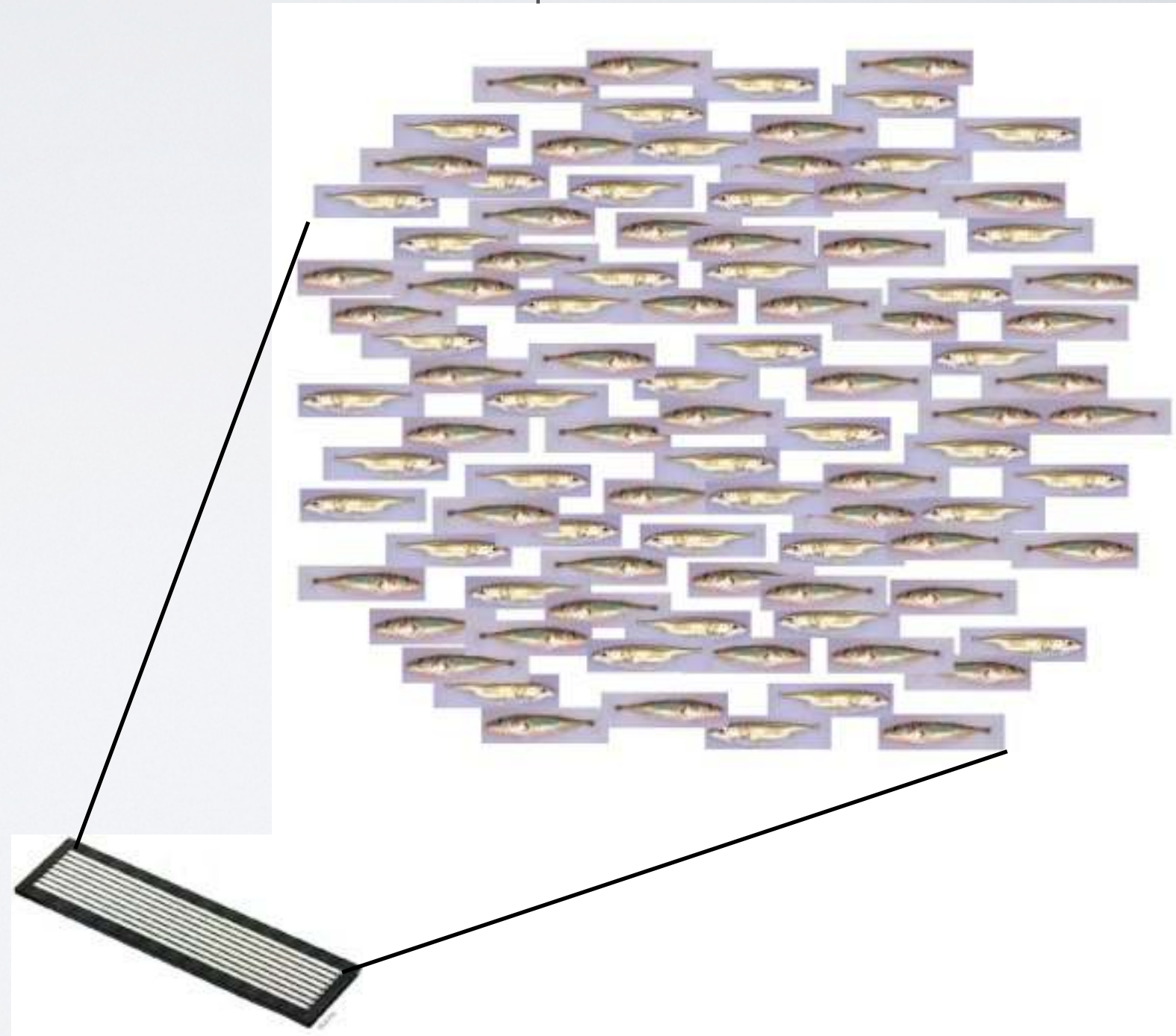
Barcoded adapters

5 nt barcodes

Variable length barcodes

Combinatorial barcodes (PE)

Barcode distance - read recovery



Experimental design considerations for RAD

Tradeoffs:

Number of sites versus **Depth** of sequencing per site versus **Number of samples**

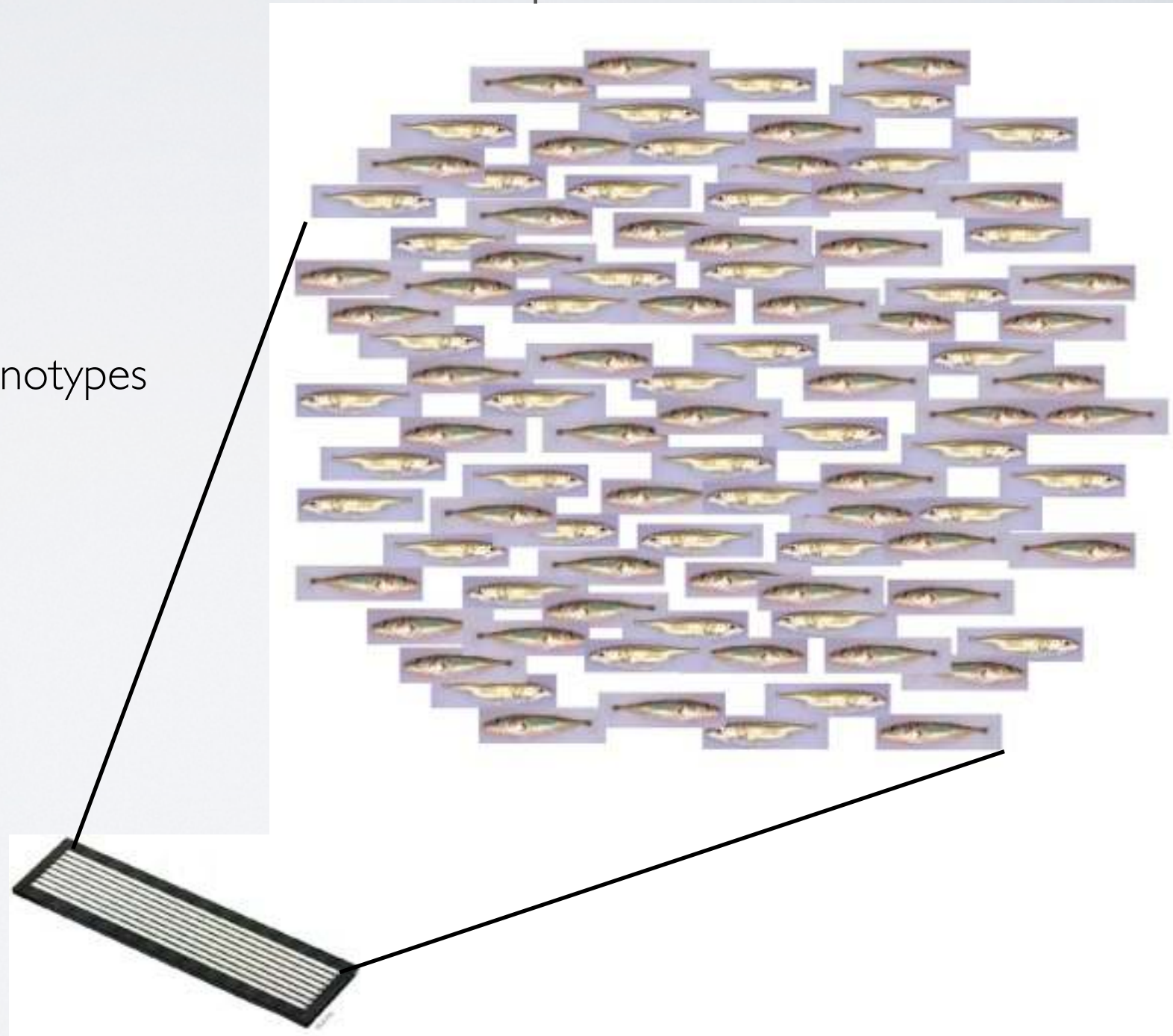
How many samples should be multiplexed?

Things to consider

Barcoded adapters

Depth

Mean depth of 20 to 40 to call genotypes



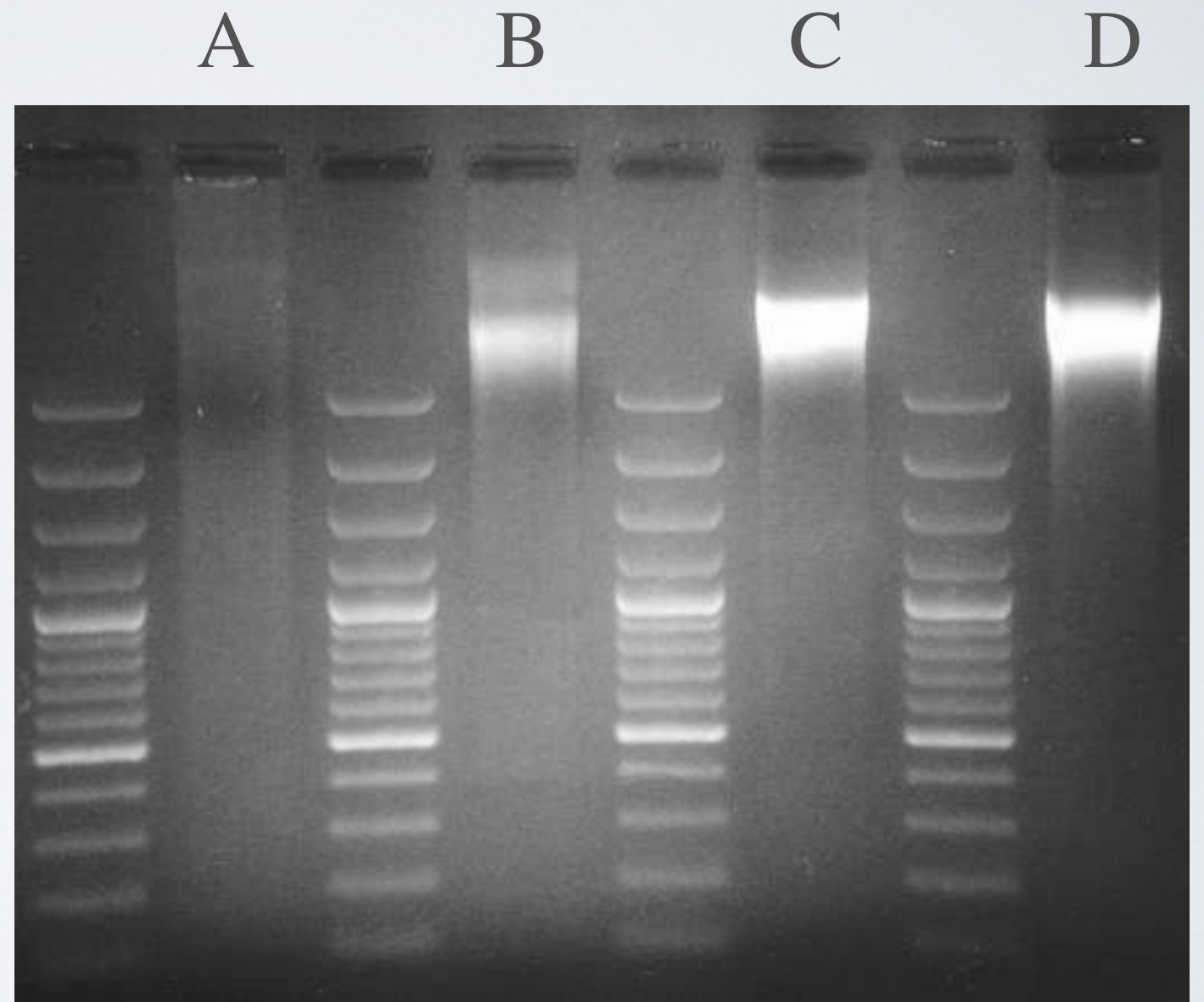
Molecular considerations in library building

How many samples should be multiplexed?

Things to consider

DNA Quality

Multiplex only like samples to help equalize representation of poor quality samples



Molecular considerations in library building

How many samples should be multiplexed?

Things to consider

DNA Quality

Diversify barcodes

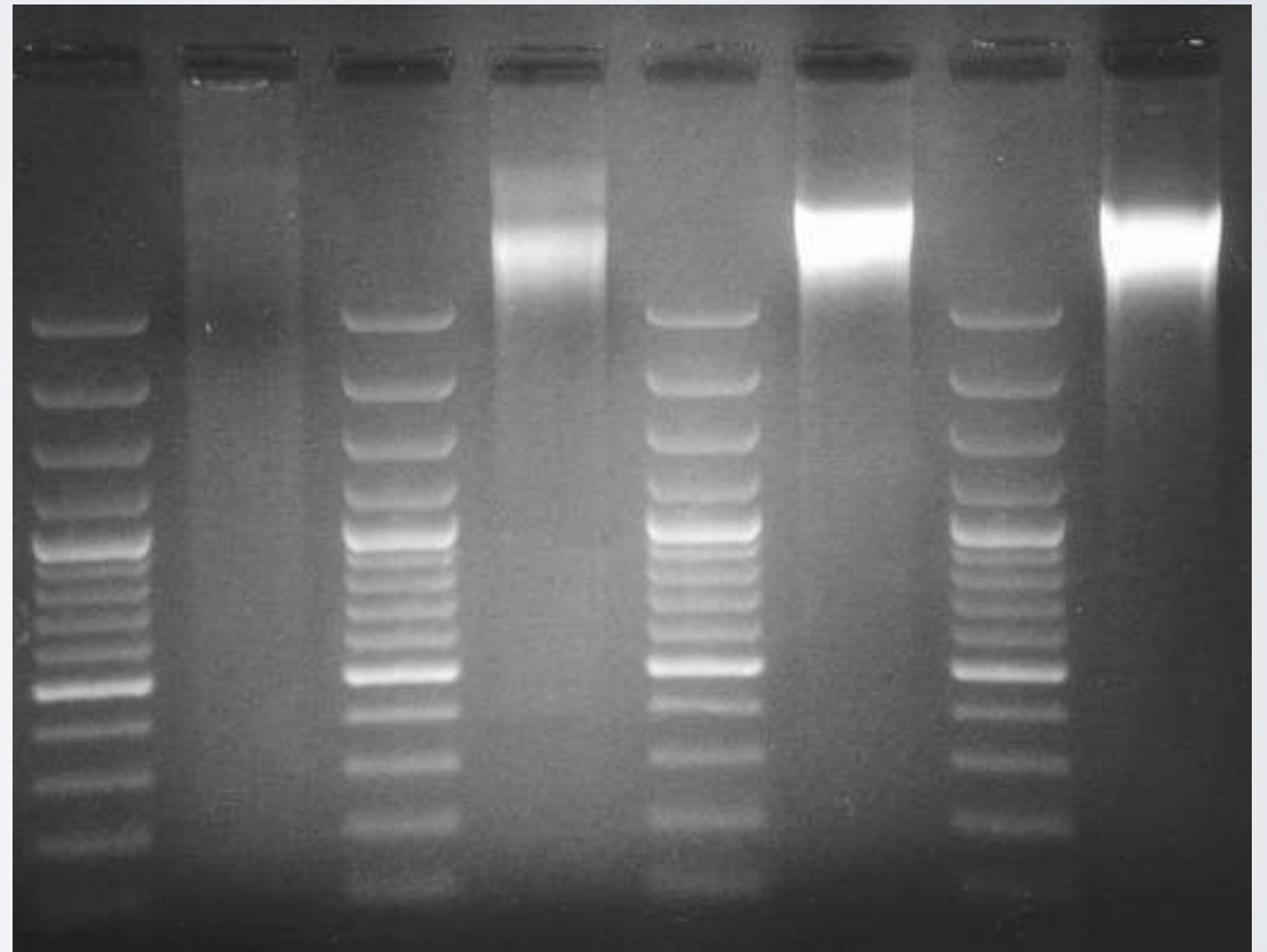
Illumina Genome Analyzer is
confused by repetition in first
4 bases

CGATA

GTACA

TAGCC

ACTGC



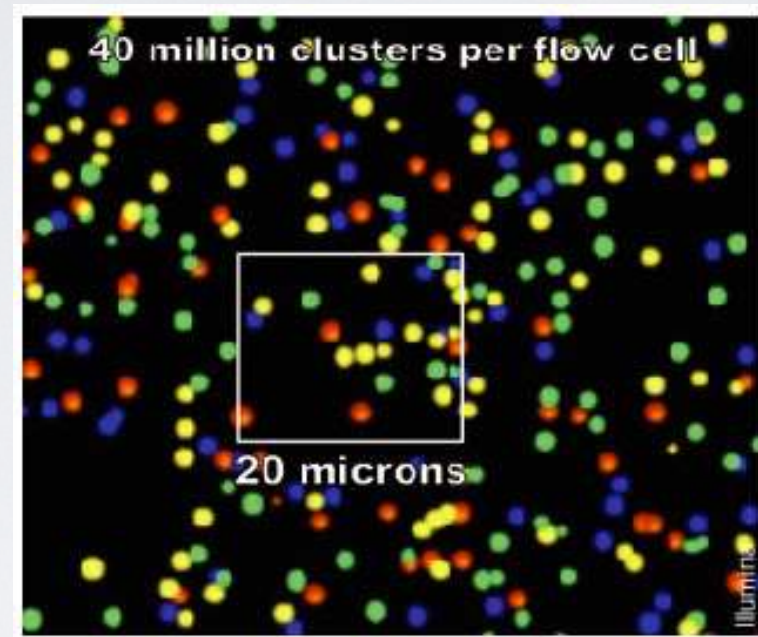
Molecular considerations in library building

How can I get the best depth of coverage?

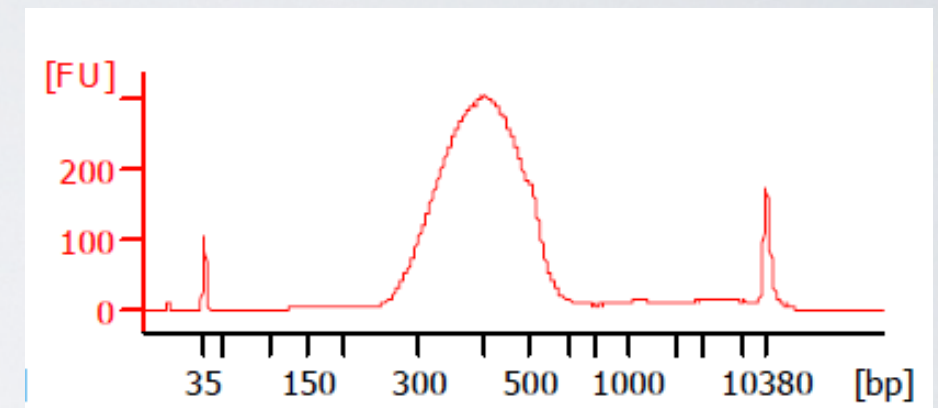
Things to consider

Fragment size

Smaller/tighter is better



Agilent Bioanalyzer



Molecular considerations in library building

How can I get the best depth of coverage?

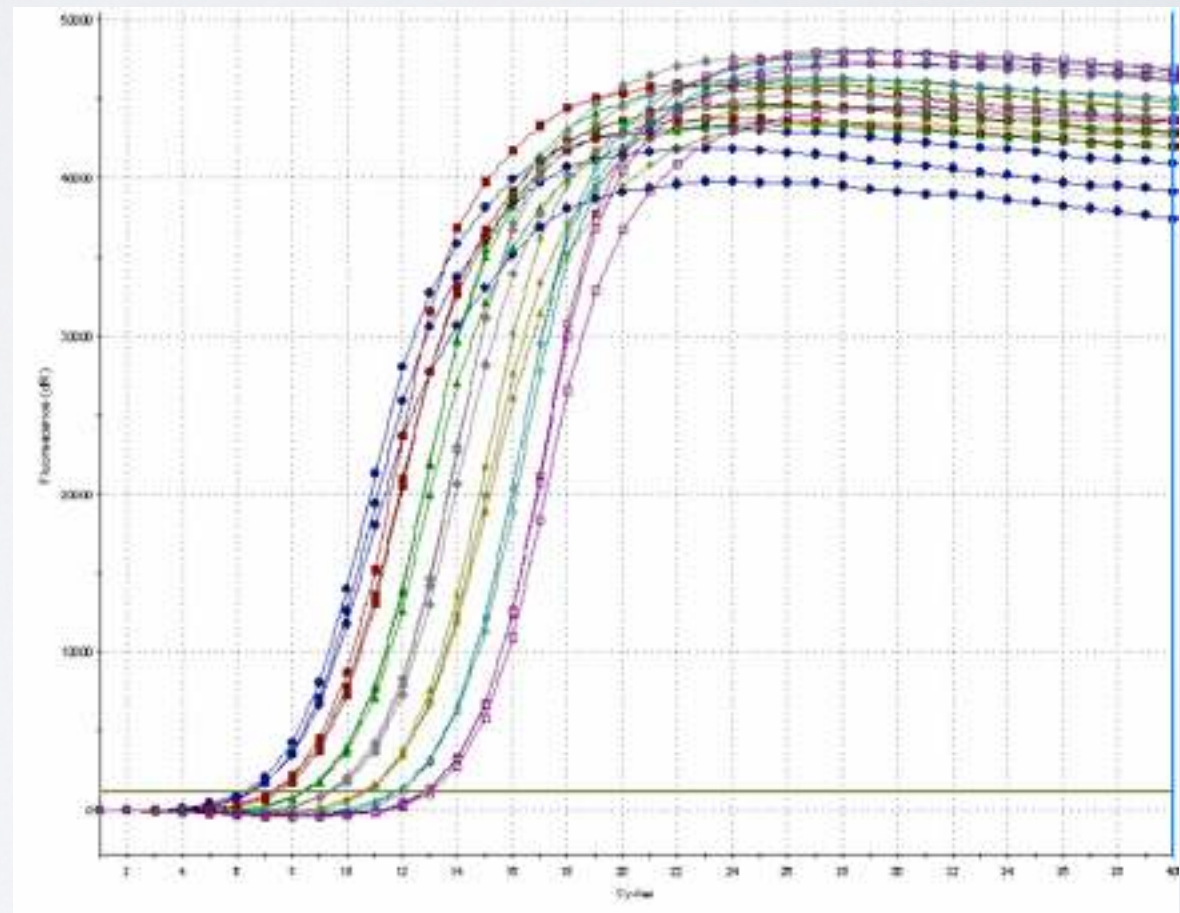
Things to consider

Fragment size

Library quality

qPCR

qPCR control should be similar to measured sample:



Molecular considerations in library building

How can I get the best depth of coverage?

Things to consider

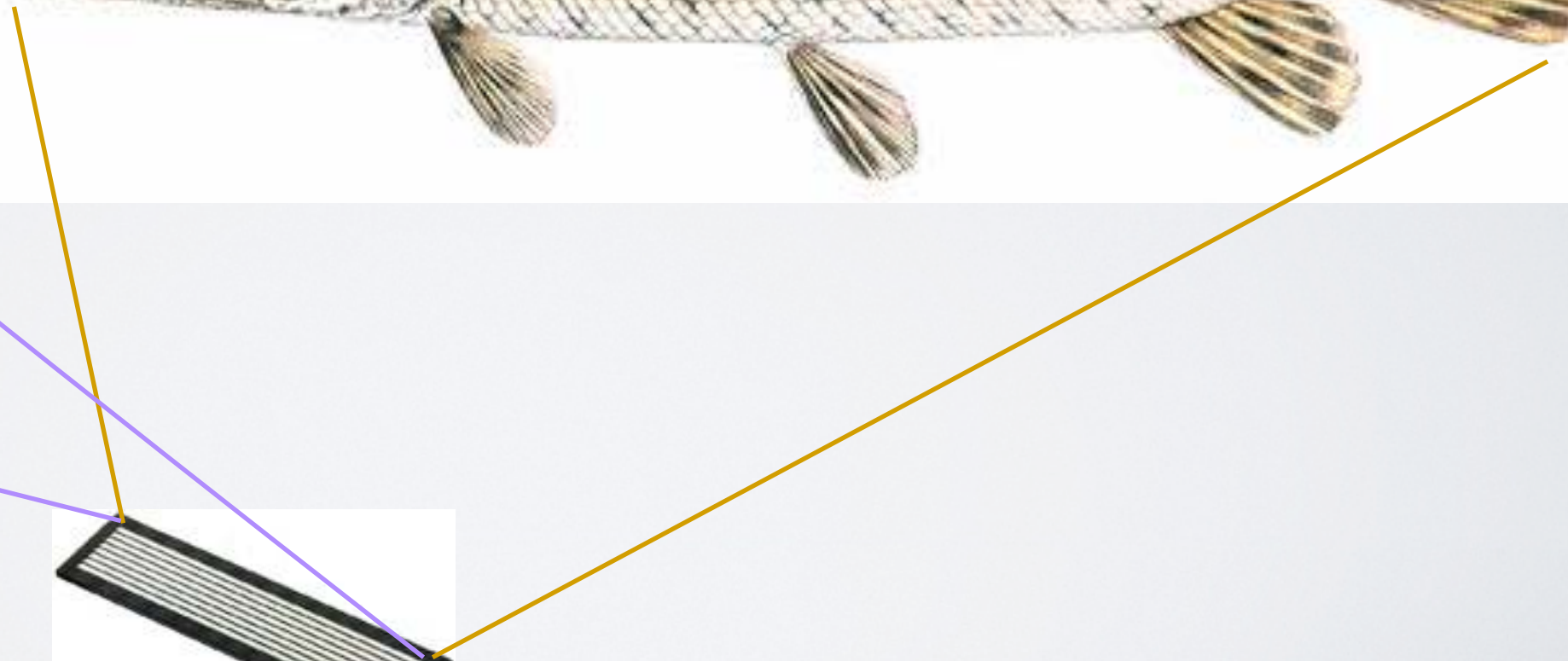
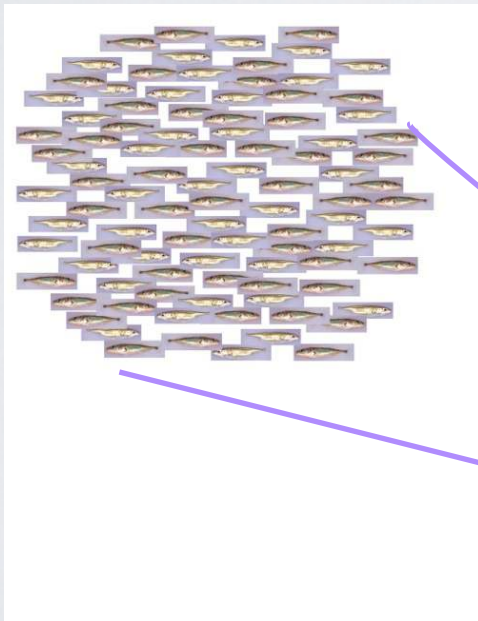
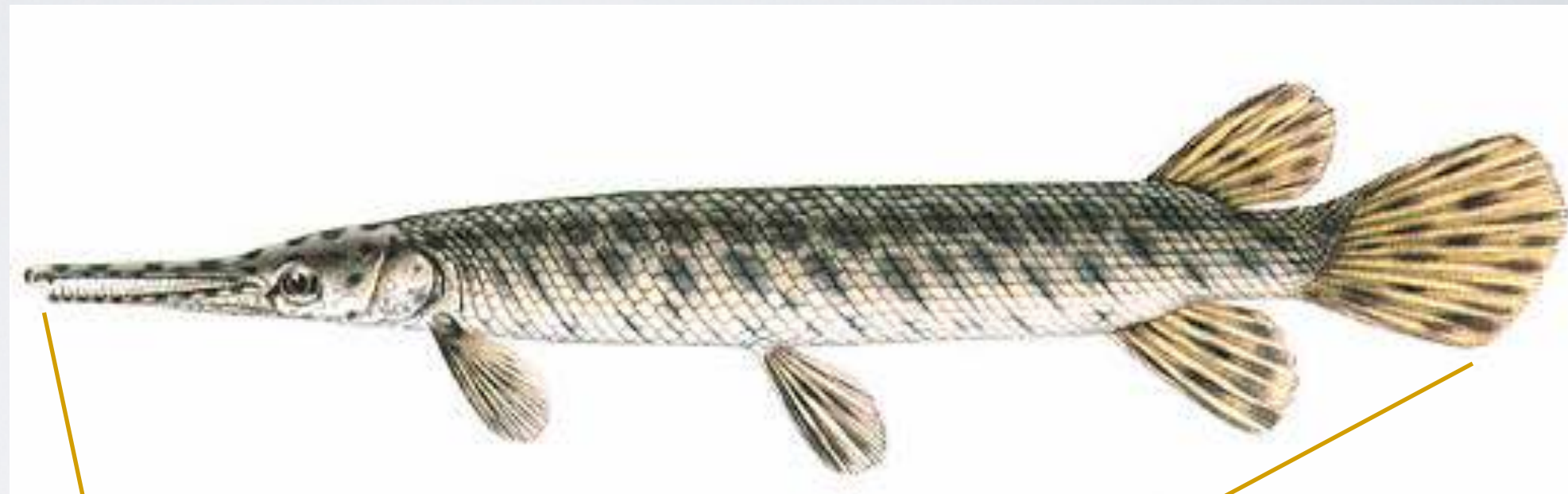
Fragment size

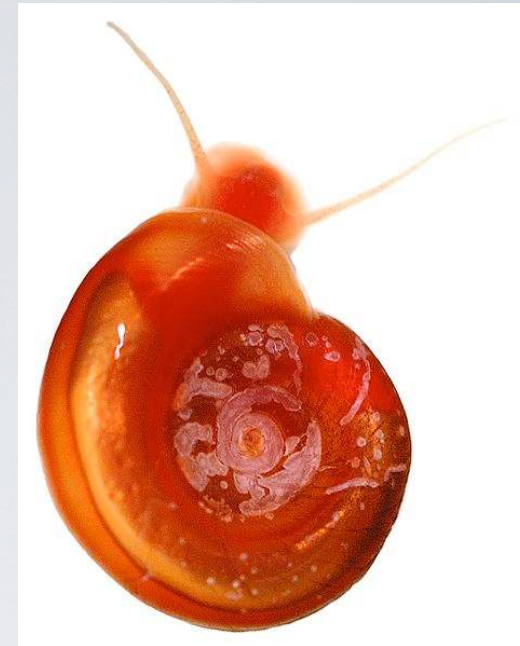
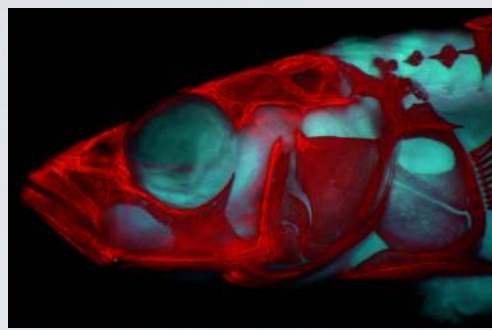
Library quality

qPCR

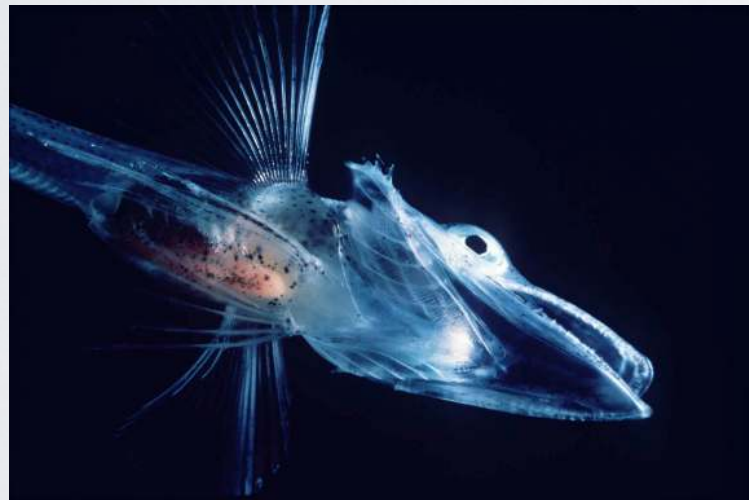
Pilot Experiment:

Spike or split a lane





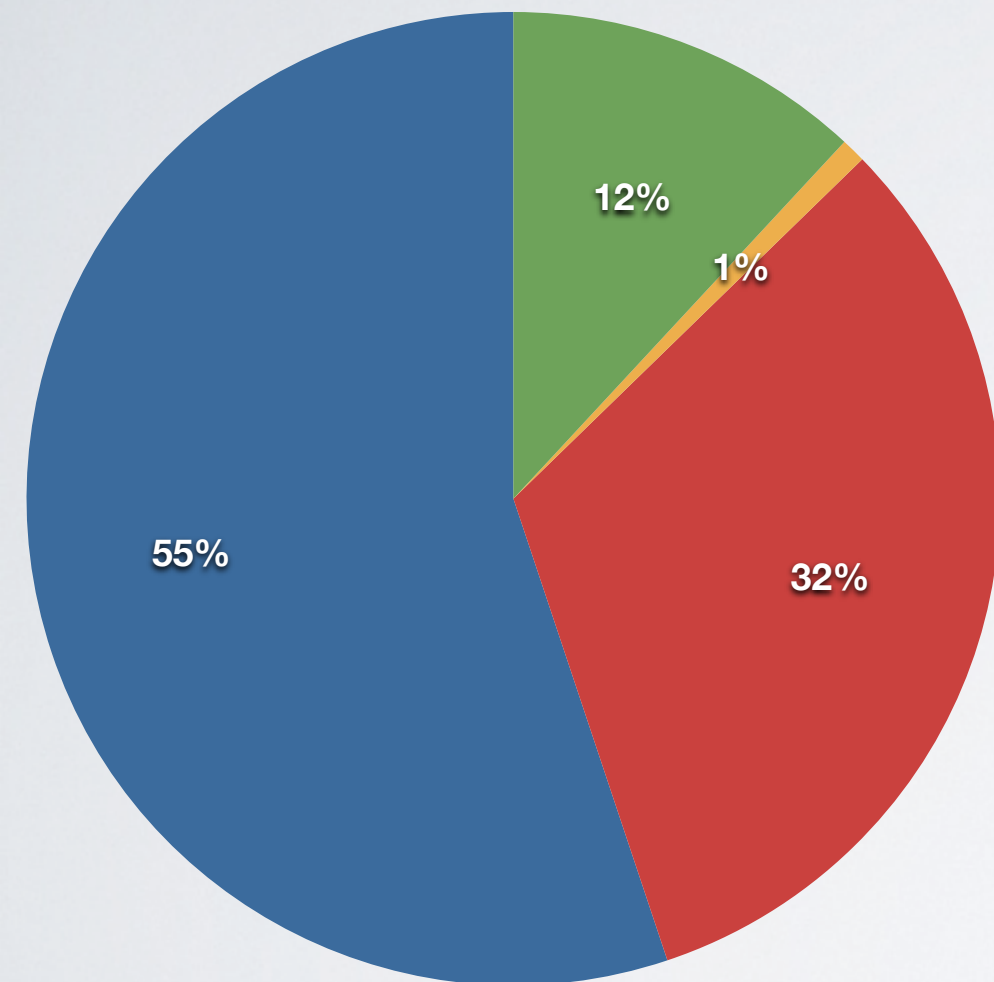
What if you have a reference genome?



Anatomy of a PstI RAD-seq Analysis

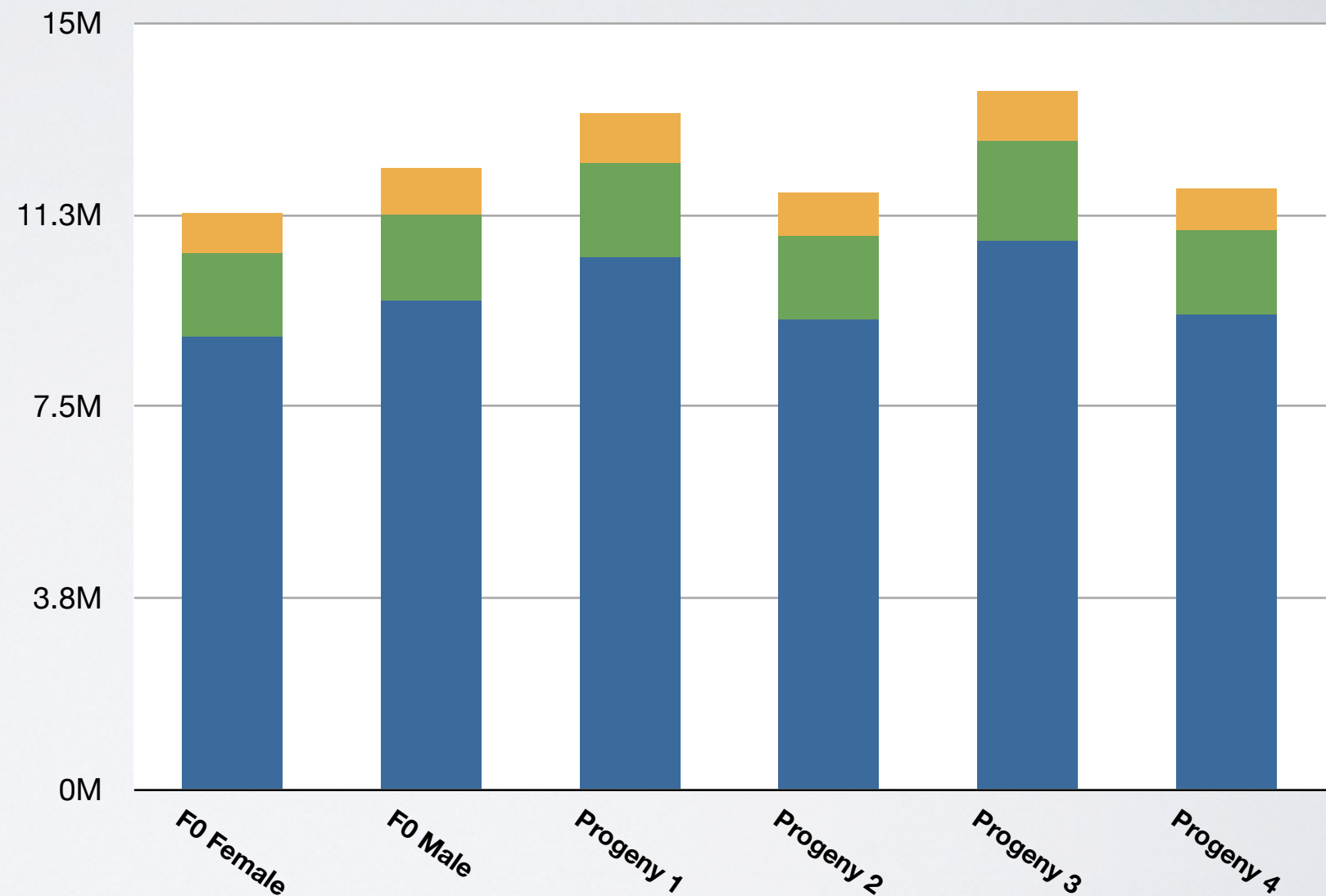
CTGCAAG

133M Illumina Raw Reads



- Ambiguous Barcodes
- Ambiguous RAD site
- Low Quality
- Retained Reads

Aligned to Reference
(Bowtie)

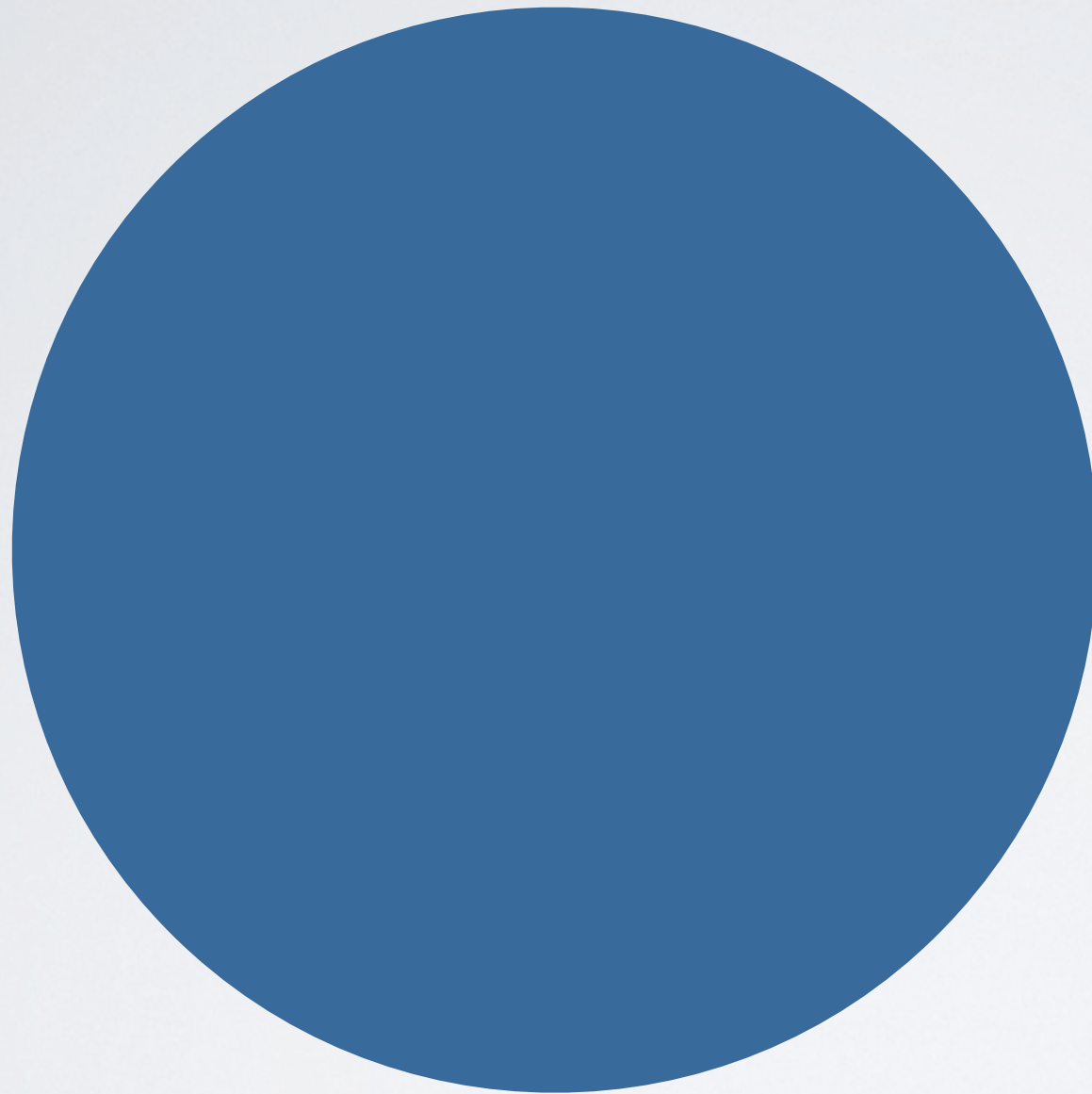


- Repetitive
- Unaligned
- Aligned

Anatomy of a PstI RAD-seq Analysis

548K Reference Sequences

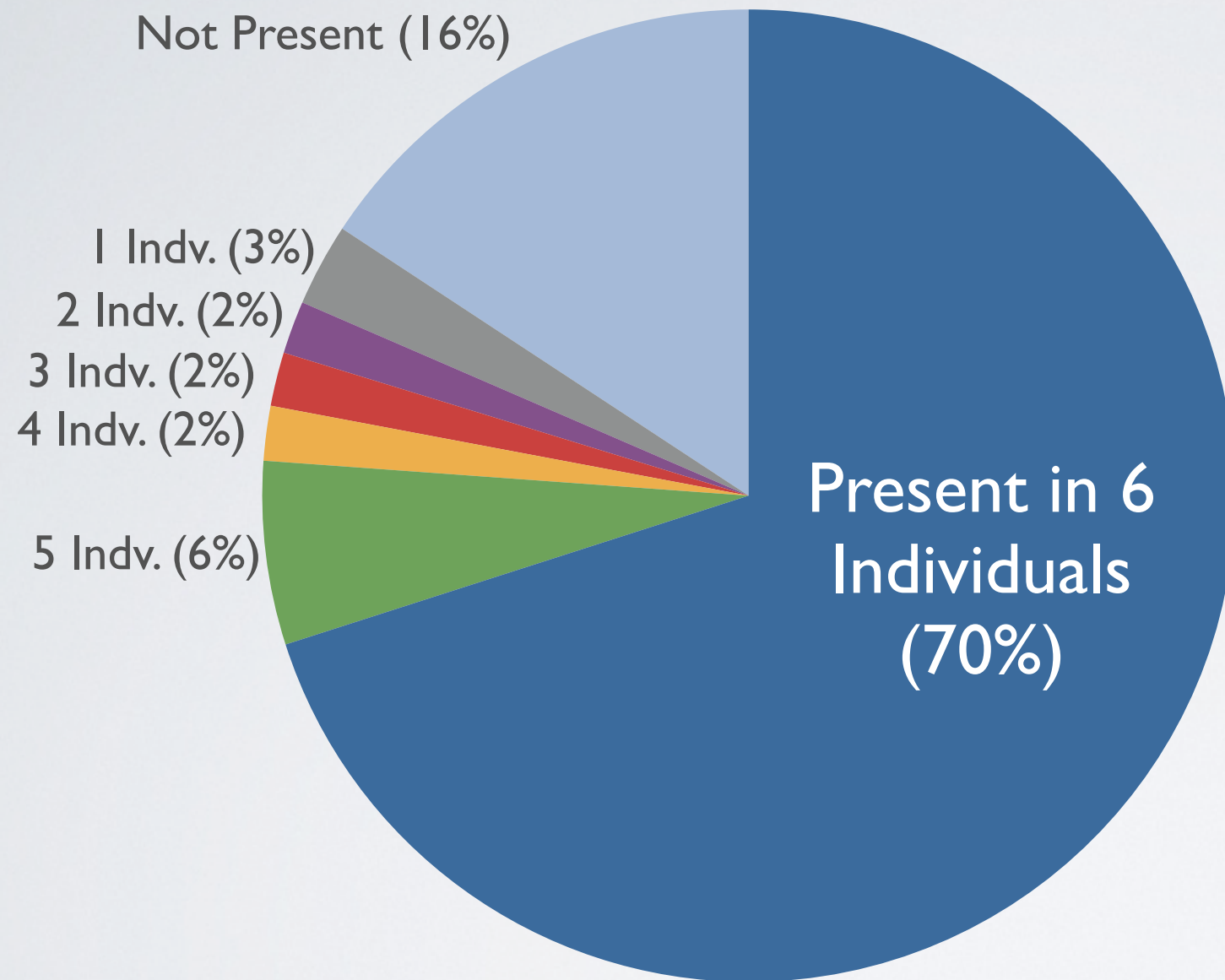
(224K Sites)



Anatomy of a PstI RAD-seq Analysis

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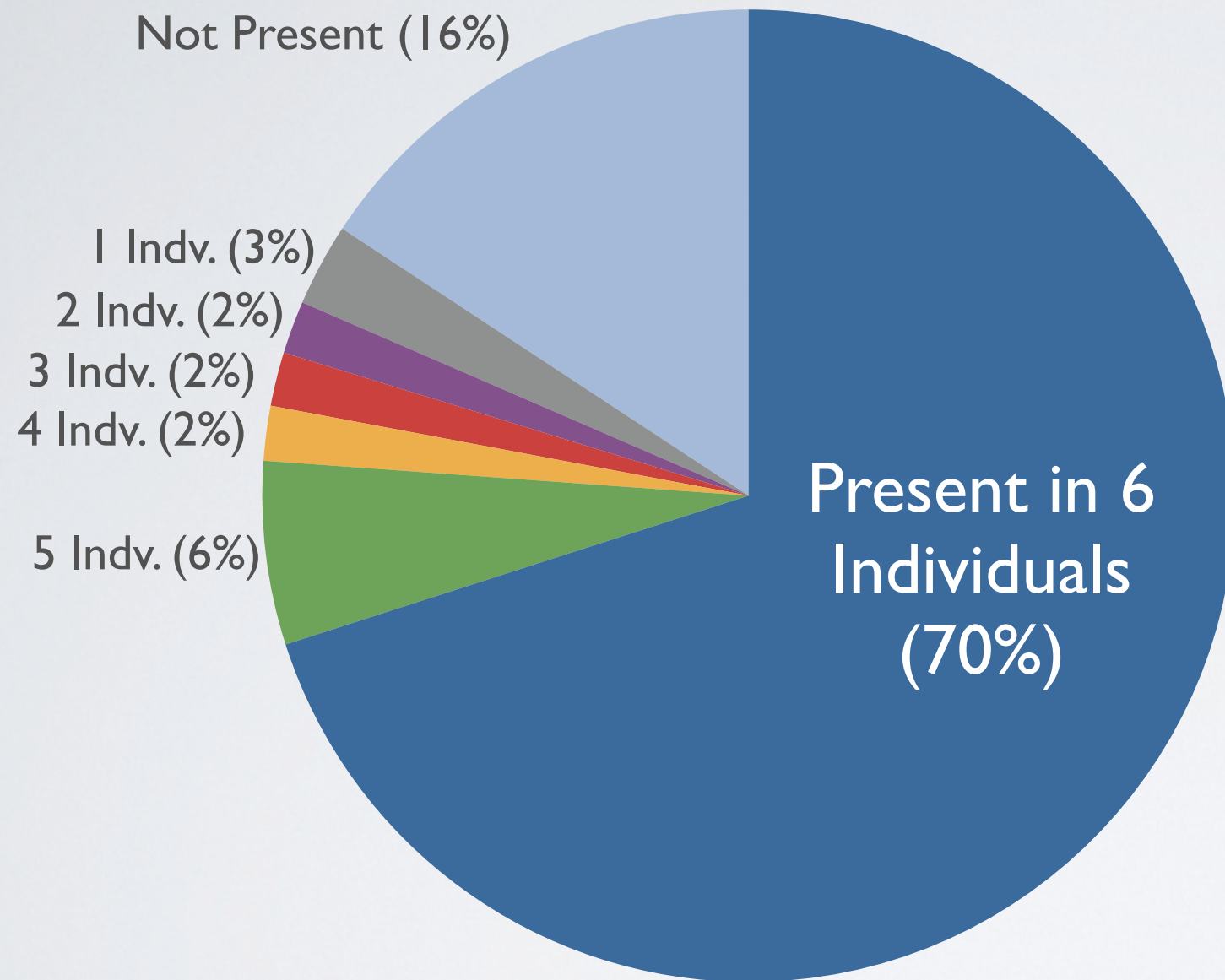
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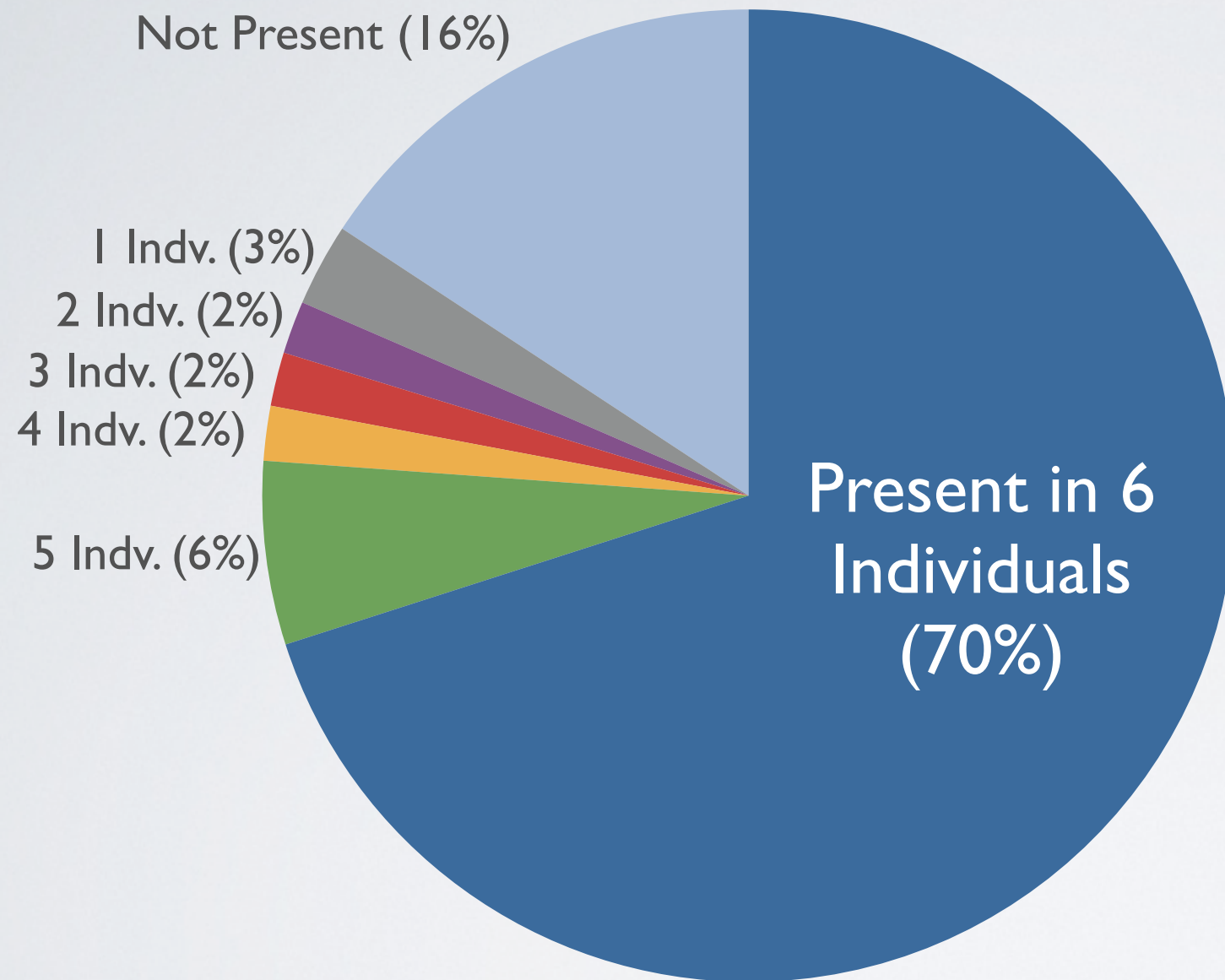
15K Non-Reference Sequences



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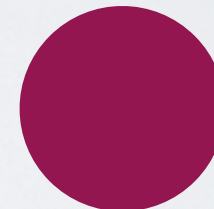
(224K Sites)



15K Non-Reference Sequences



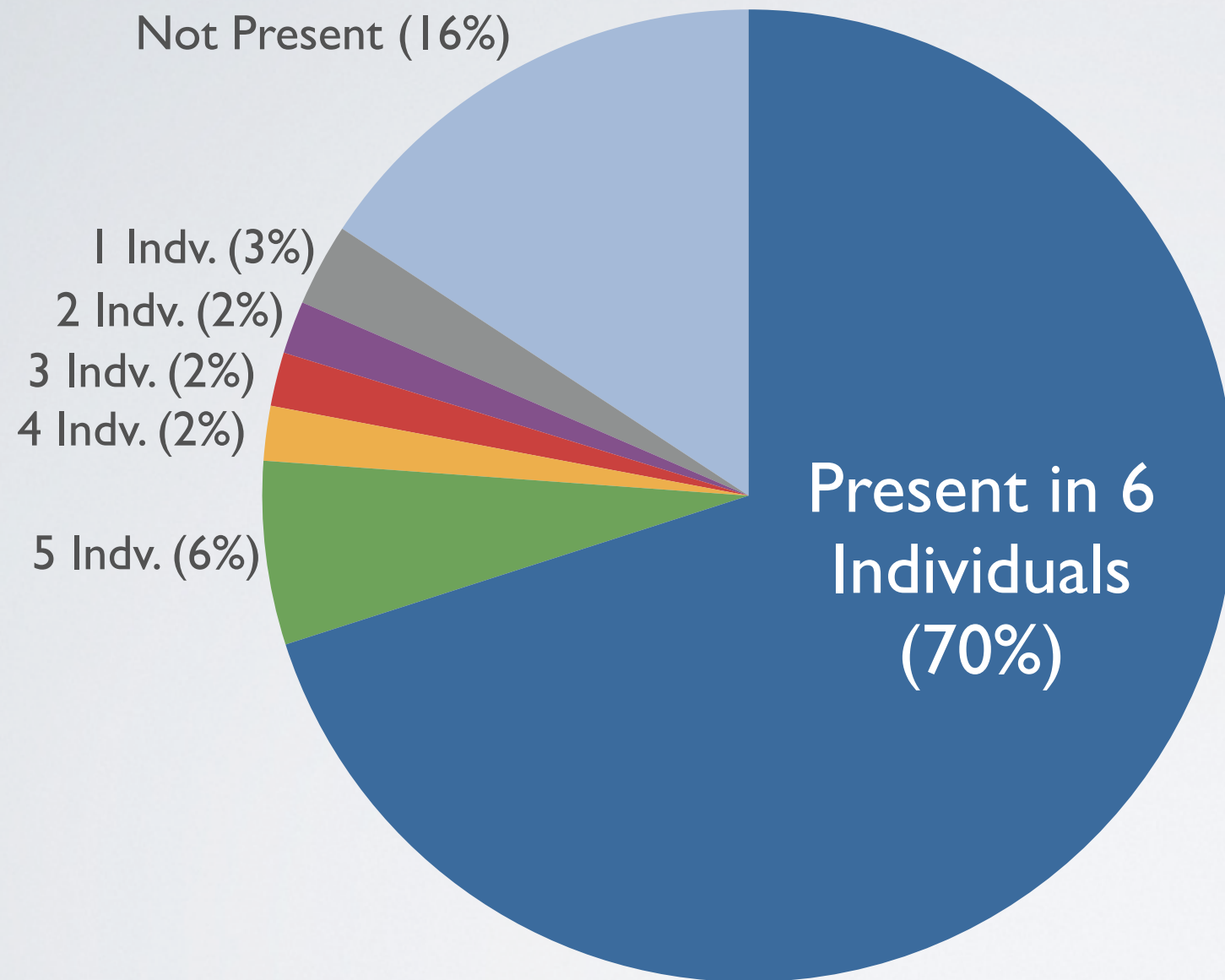
97K *de novo* Stacks



Anatomy of a PstI RAD-seq Analysis

548K Reference Sequences

(224K Sites)



15K Non-Reference Sequences

97K *de novo* Stacks

Genome Size	Method	Cut Sites	Frequency
442Mb	Naive	84K	5.3K
	Empirical	224K	1.5K

Anatomy of a PstI RAD-seq Analysis

Bear Paw Stickleback Reference - LGI



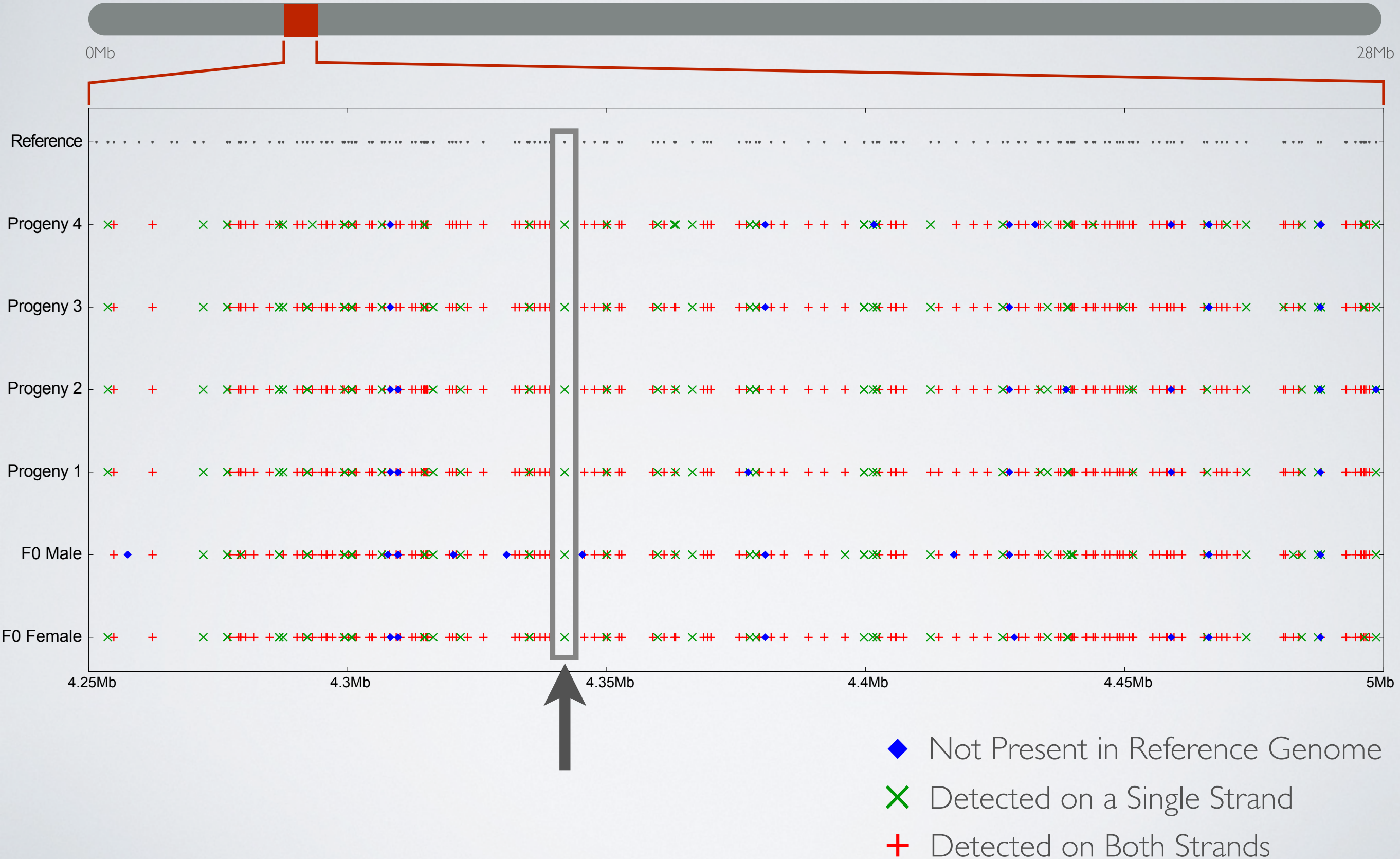
Anatomy of a PstI RAD-seq Analysis

Bear Paw Stickleback Reference - LGI



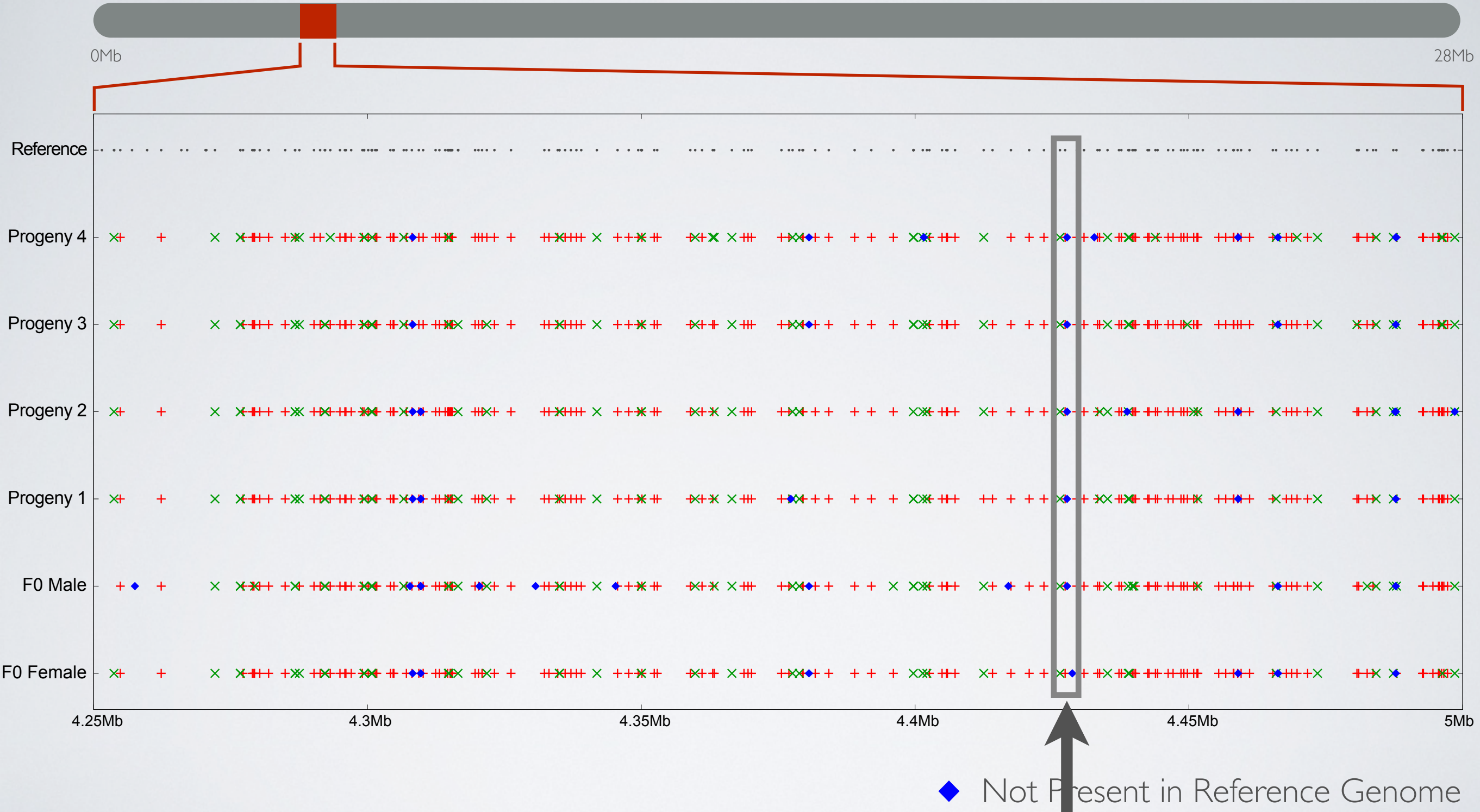
Anatomy of a PstI RAD-seq Analysis

Bear Paw Stickleback Reference - LGI

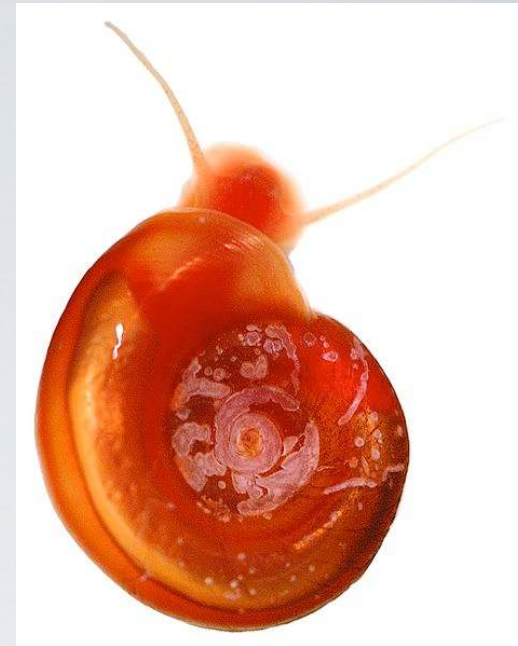
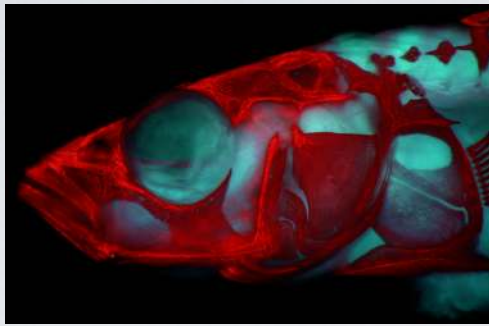


Anatomy of a PstI RAD-seq Analysis

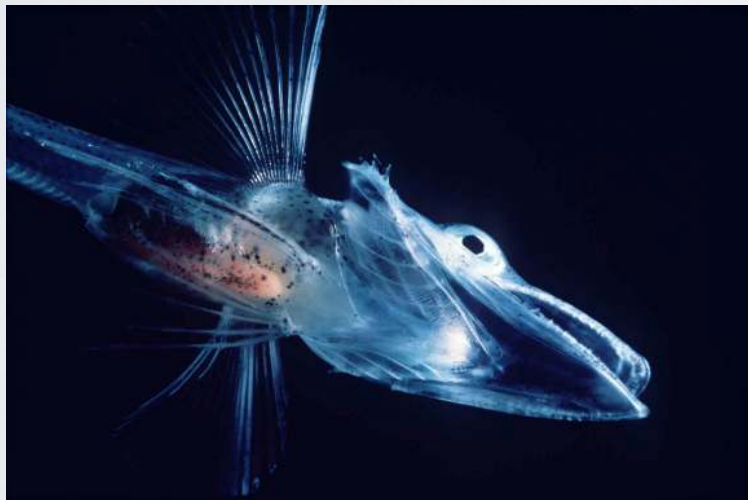
Bear Paw Stickleback Reference - LGI



- ◆ Not Present in Reference Genome
- ✕ Detected on a Single Strand
- ✚ Detected on Both Strands



Signatures of natural
selection across the genome



Threespine Stickleback, *Gasterosteus aculeatus*

● *Ancestral Oceanic Populations*

Marine and Anadromous
Old (> 10 million years)

● *Derived Freshwater Populations*

Lake and stream
Young (< 15,000 years)



Ocean form



Freshwater form

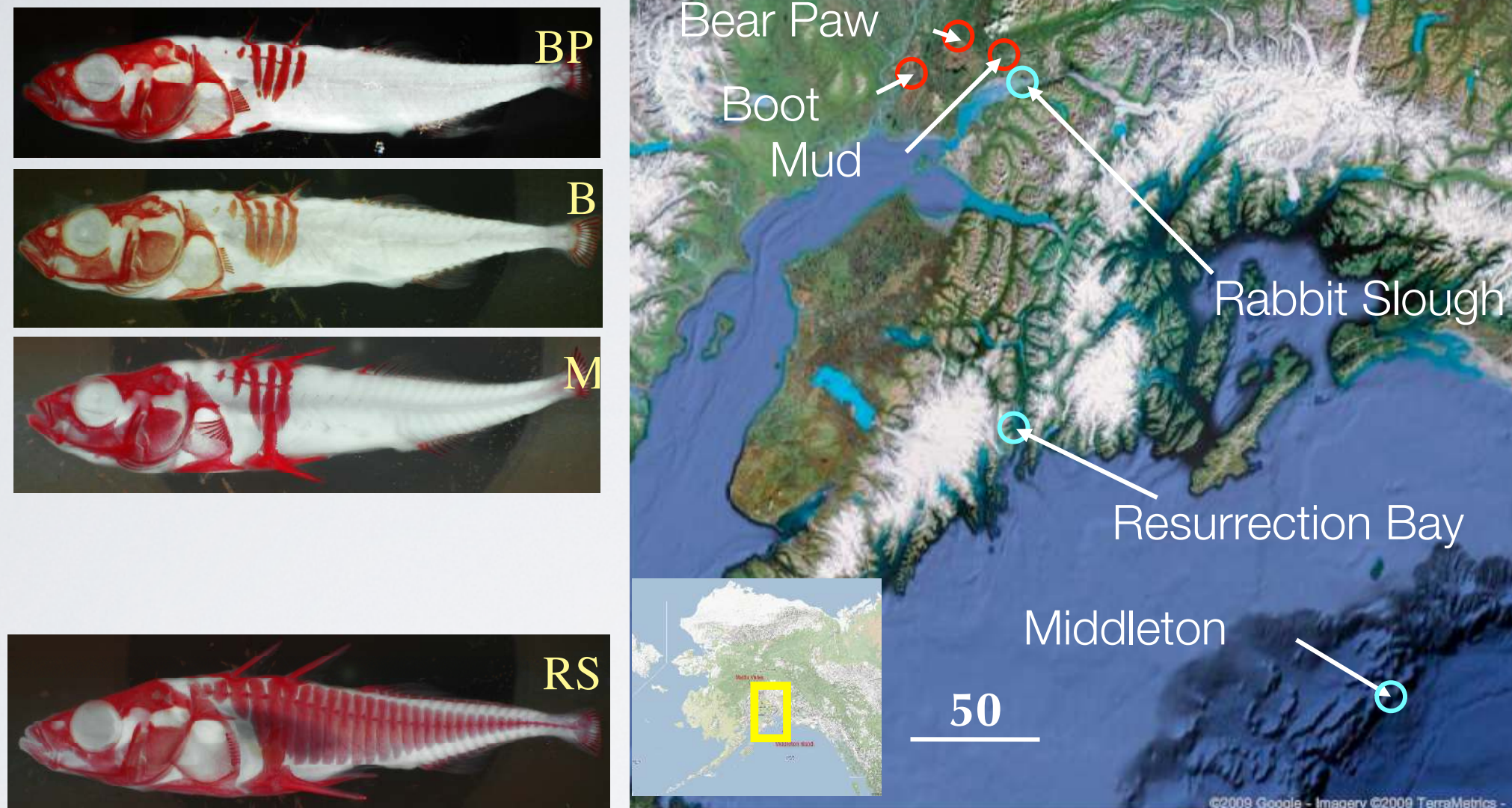


Susan
Bassham



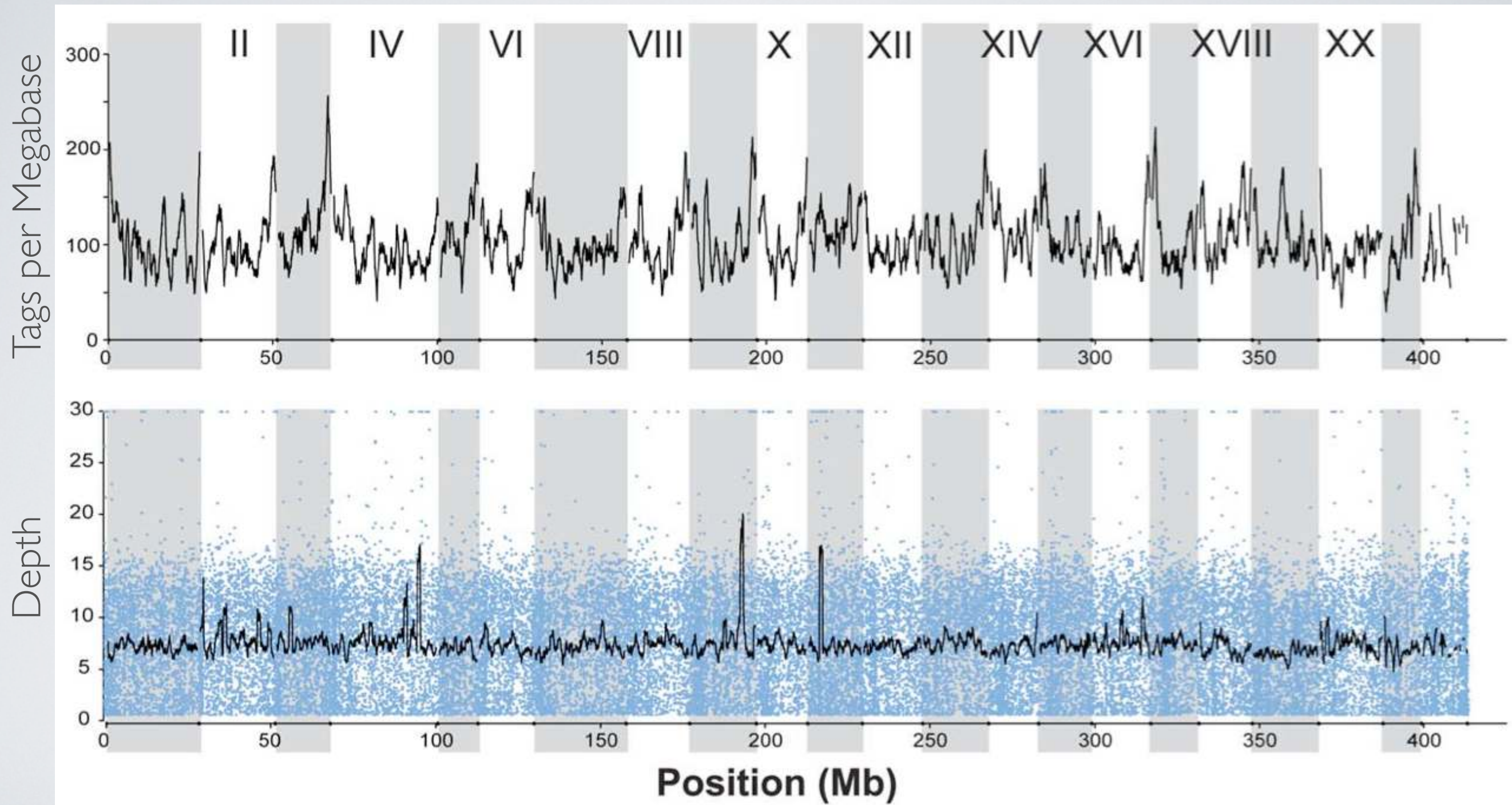
Paul
Hohenlohe

Signatures of natural selection across the genome



- 20 individuals in each of 5 populations
- 2 Ocean & 3 Freshwater
- 45,000 SNPs in each individual

Even coverage of RAD tags across the genome



Concatenated stickleback genome

Differentiating SNPs from error

Restriction enzyme recognition site

Reference
genome
sequence

sequence
reads

```
CTTCAGGTTGGGTGAGTTGTCATCAGTCGGAATGCGCAGGTCACCTTACCTGCAGGCAGCTCTCTGAAGCGCAGGTACTCCATCGACCGGGTGGTGACTAG
NNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNN
ATCAGTCGGAATGCGCAGGTCACCTTACCTGCA
ATCAGTCGGAATGCGCAGGTCACCTTACCTGCA
ATCAGTCgAATGCTCAGGTCaCTTacctGcA
ATCAGTCGGAATGCGCAGGTCACCTTAcctGcA
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cgcagtbggAAtgcGCaggTcaCTTcactgcA
AtCAGTCGGAATGCGCAGGTCACCTTAcctgcA
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ATCAGTCgGAATGCGCAGGTCACCTTAcctgcA
ATCAGTCGGAATGCGCAGGTCACCTTAcctgcA
ATCAGTCGGAATGCGCAGGTCACCTTAcctgcA
ATCAGTCGGAATGCGCAGGTCACCTTAcctgcA
ATCAGTCGGAATGCGCAGGTCACCTTAcctgcA
TgcaggCAGcTCTCTGAAGCGCAGGtACTCCA
TgcaggCAGCTCTCTGAAGCGCAGGtACTCCA
TgCaggCAGCTCTCTGAAGCGCAGGgACTcCA
TGCAGGCAGCTCTCTGAAGCGCAGGtaCTCCA
TgcaggCAGCTCTCTGAAGCGCAGGtACTCcA
tgcaggcAGCtCTCTGAAGagaaGgtaCTCca
TgCaggCAGCTCTCTGAAGCGCAGGtACTcCA
TgcaggCAGCTCTCTGAAGCGCacGtaCtcCa
TgcaggcAGcTATCTGAAGcgCAgGTActcca
TGCaggCAGCTCTCTGAAGCGCAGGtACTCCA
TgCaggCAGCTCTCTGAAGCGCAGGtAcTcca
TGCaGgCAGCTCTCTGAAGCGCAGGTACTCCA
tgcaggCAGCTcTATGAAGcGCAGGtActcca
tgcaggCagCtCTCTgaAGcgAagGgACTcca
ggcaggCAGCtCTCtGaaGcGcAggtactcca
TgCaggCAGCTCTCTGAAGCGCAGGtAcTCCA
```

Differentiating SNPs from error

T
T
G
T
T
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T
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T
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T

```
CTTCAGGTTGGGTGAGTTGTCATCAGTCGGAATGCGCAGGTCACTTACCTGCAGGCAGCTCTCTGAAGCGCAGGTACTCCATCGACCGGGTGGTGACTAG
NNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNN
ATCAGTCGGAATGCGCAGGTCACTTACCTGCA
ATCAGTCGGAATGCGCAGGTCACTTACCTGCA
ATCAGTCgAATGCTCAGGTCaCTTacctGcA
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tgcaggCAGCTcTATGAAGcGCAGGtACTCCA
tgcaggCagCtCTCTgaAGcgAagGtACTCCA
ggcaggCAGctCTCtGaaGcGcAggtACTCCA
TgCaggCAGCTCTCTGAAGCGCAGGtACTCCA
```

Differentiating SNPs from error

T
T
G
T
T
T
T
T
T
T
T
T
T
G
T
T

The reads are 14 T and 2 G:

GT heterozygote?

GG homozygote with error?

AA homozygote with lots of error?

Needed a rigorous method to call genotypes

Differentiating SNPs from error

T
T
G
T
T
T
T
T
T
T
T
T
T
G
T
T

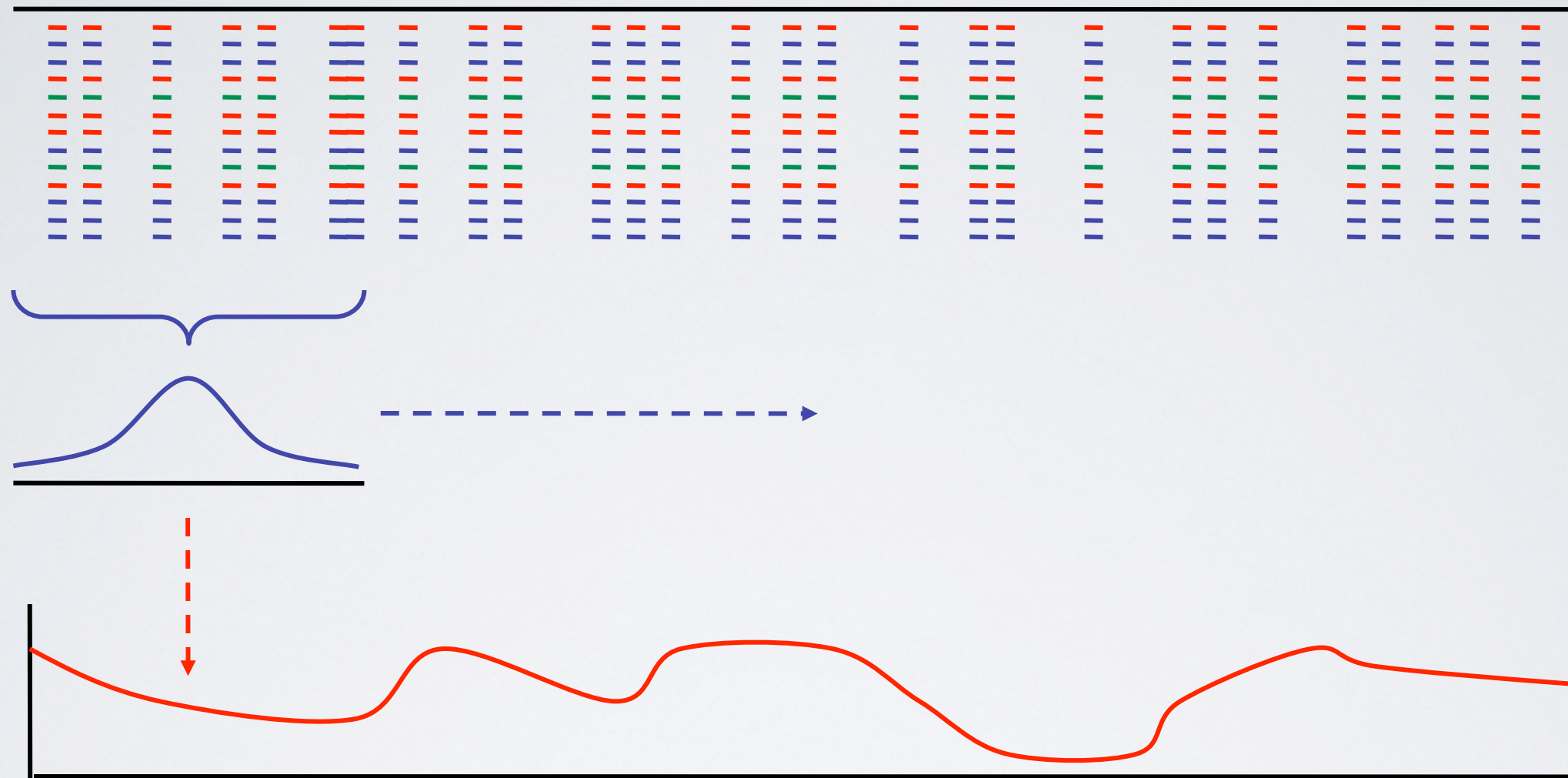
$$L(n_1 \text{ hom}) = P(n_1, n_2, n_3, n_4) = \frac{n!}{n_1! n_2! n_3! n_4!} \left(1 - \frac{3\varepsilon}{4}\right)^{n_1} \left(\frac{\varepsilon}{4}\right)^{n_2} \left(\frac{\varepsilon}{4}\right)^{n_3} \left(\frac{\varepsilon}{4}\right)^{n_4}$$

$$L(n_1 n_2 \text{ het}) = P(n_1, n_2, n_3, n_4) = \frac{n!}{n_1! n_2! n_3! n_4!} \left(0.5 - \frac{\varepsilon}{4}\right)^{n_1} \left(0.5 - \frac{\varepsilon}{4}\right)^{n_2} \left(\frac{\varepsilon}{4}\right)^{n_3} \left(\frac{\varepsilon}{4}\right)^{n_4}$$

Maximum likelihood genotyping based on
multinomial distribution of nucleotide reads

Making statistics continuous across the genome

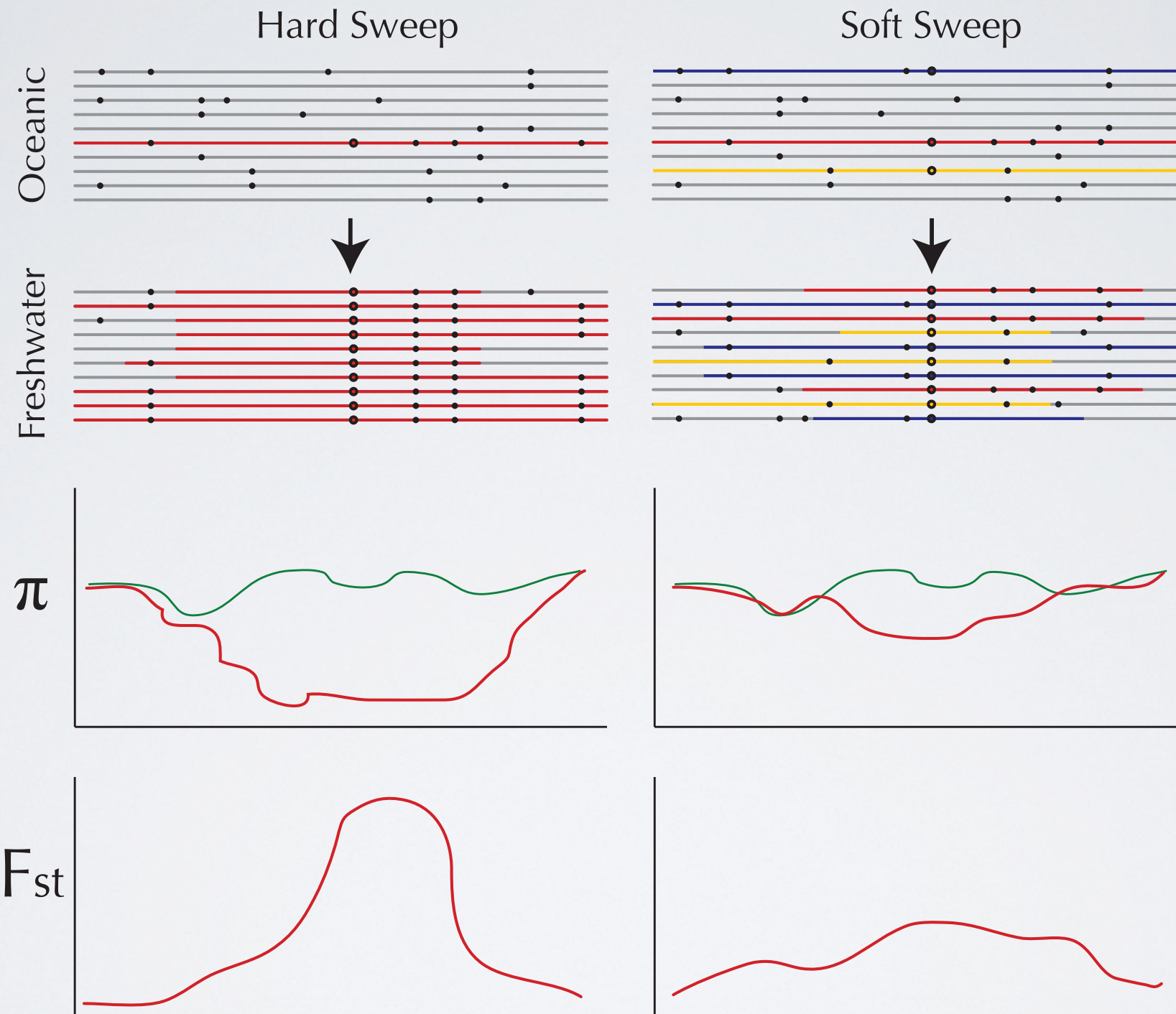
Kernel-smoothing average of summary statistics along genome



Bootstrap re-sampling to estimate significance of moving average

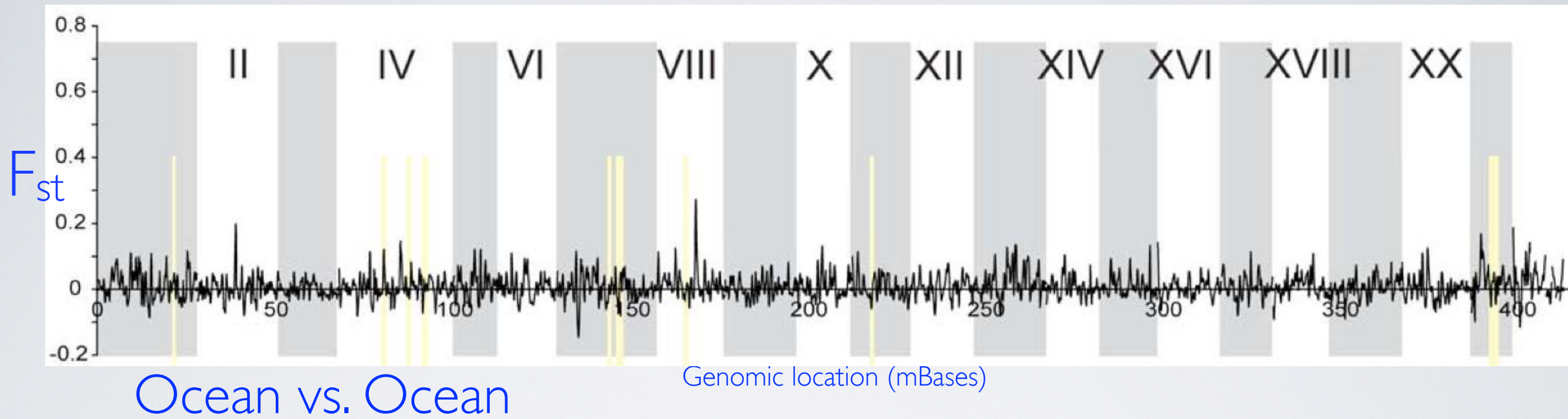
- Re-sample from across genome within the same population
- Significance is relative to genome-wide average, not based on a null model

Signatures of natural selection across the genome

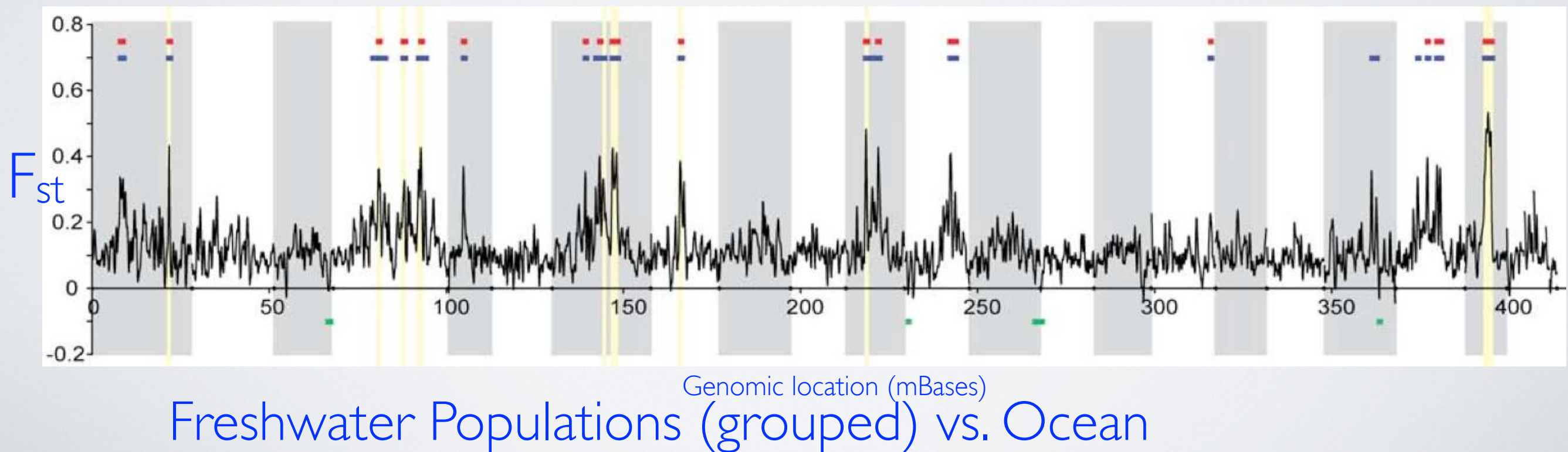
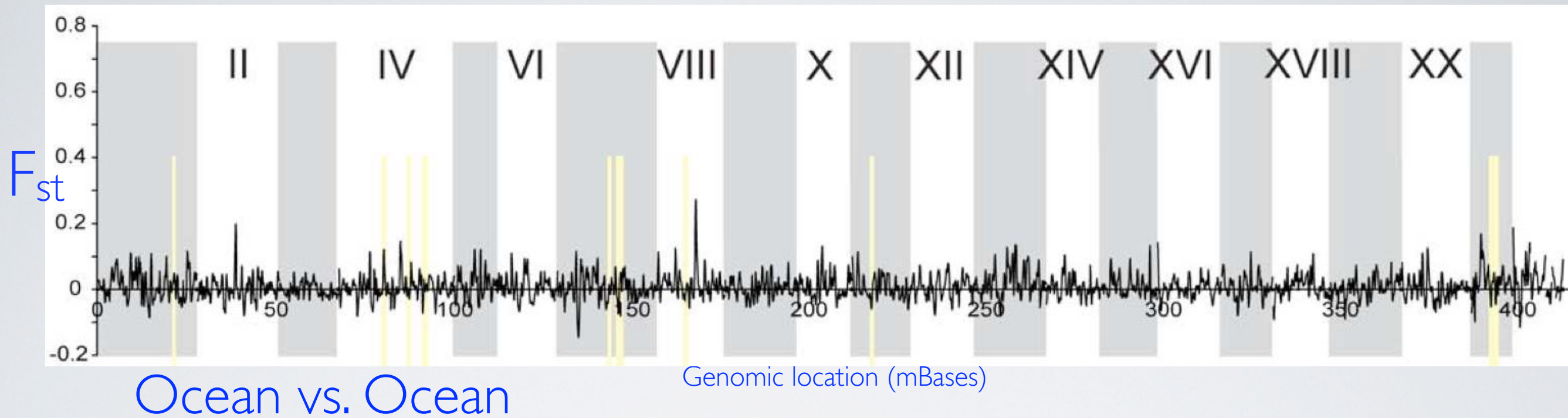


requires a very dense set of markers or complete sequences
assays need to be performed in dozens or hundreds of individuals
hard to develop for non-model organisms

Signatures of natural selection across the genome

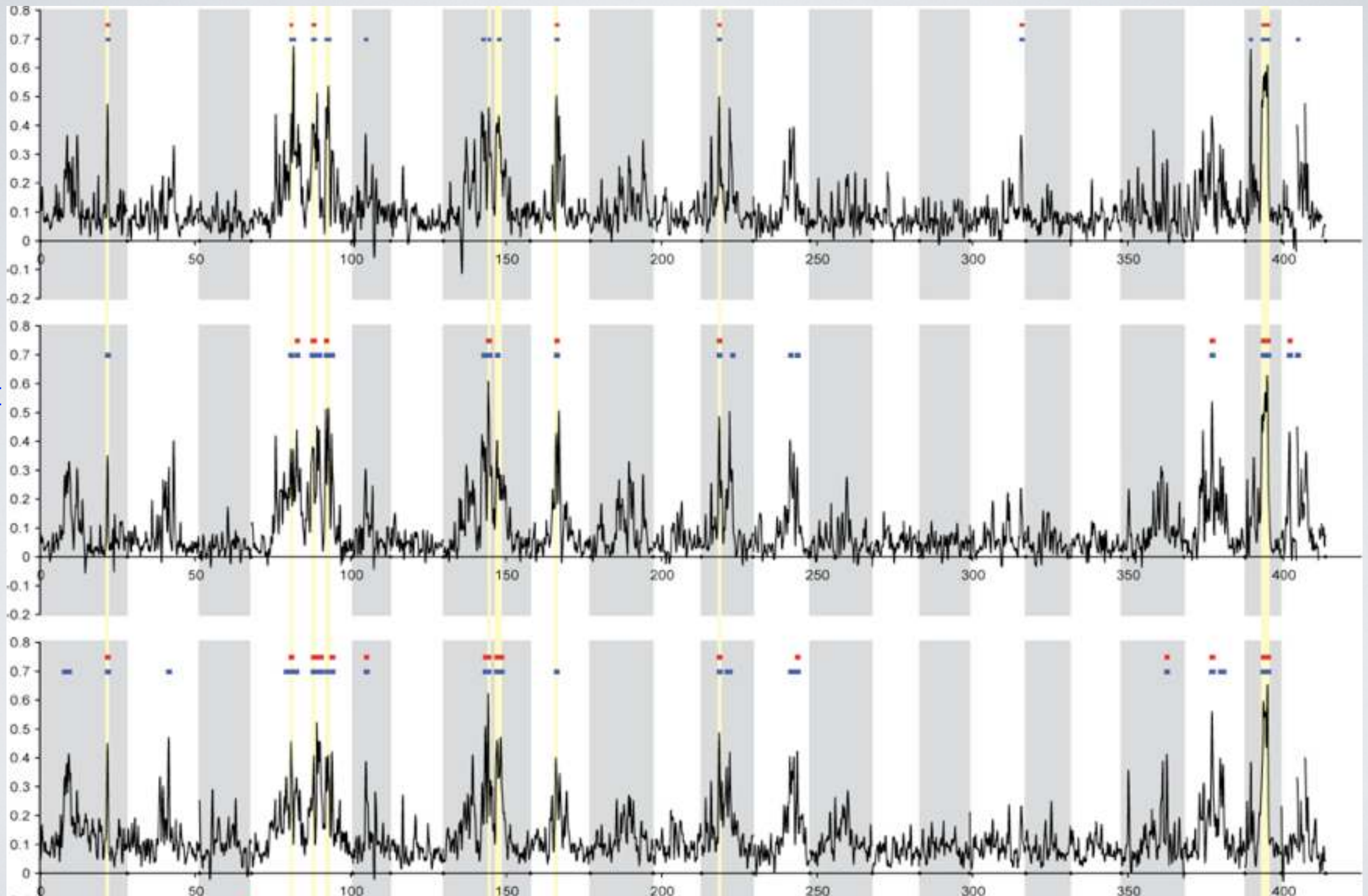


Signatures of natural selection across the genome



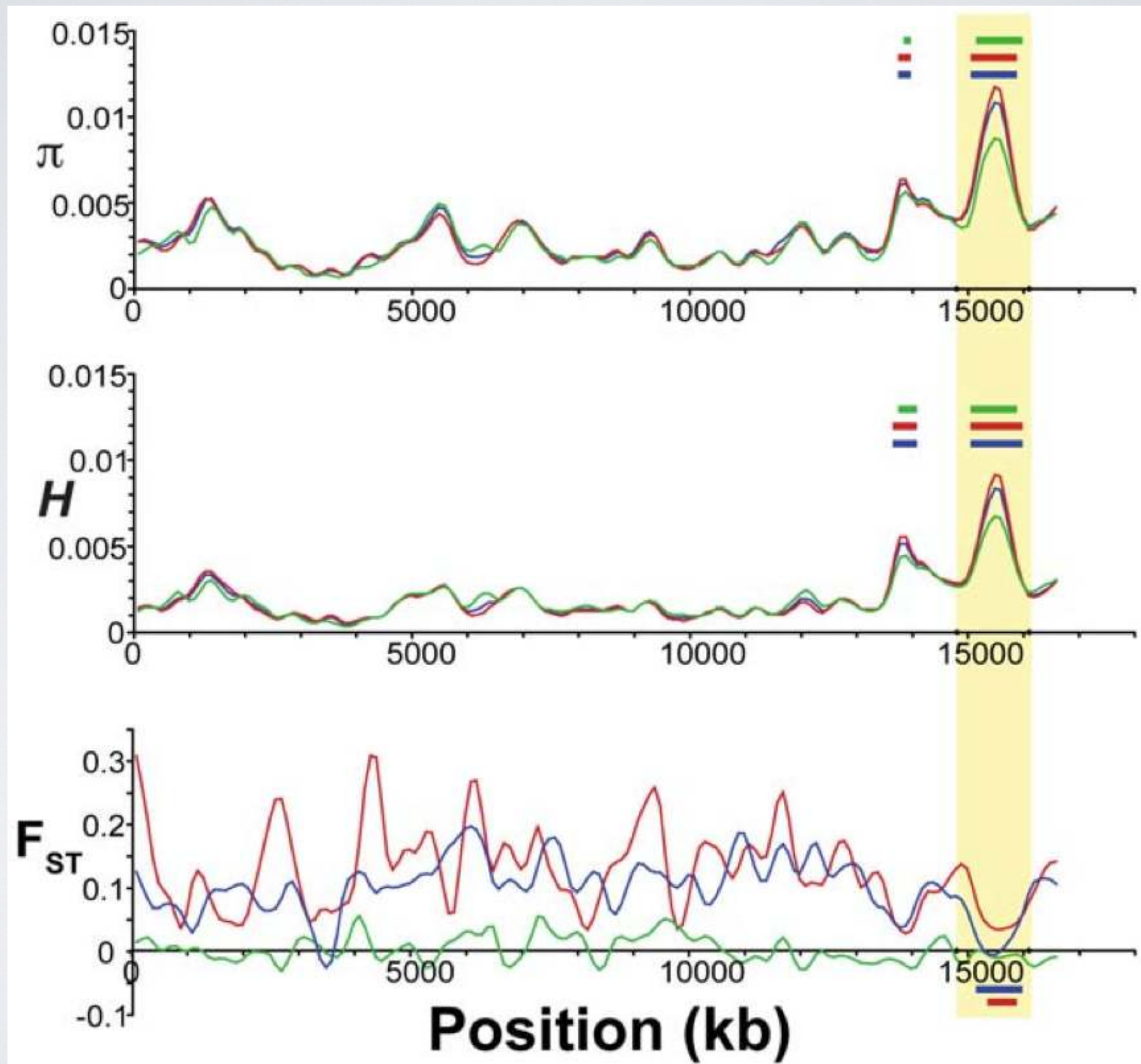
Consistent pattern of parallel selection

F_{st}



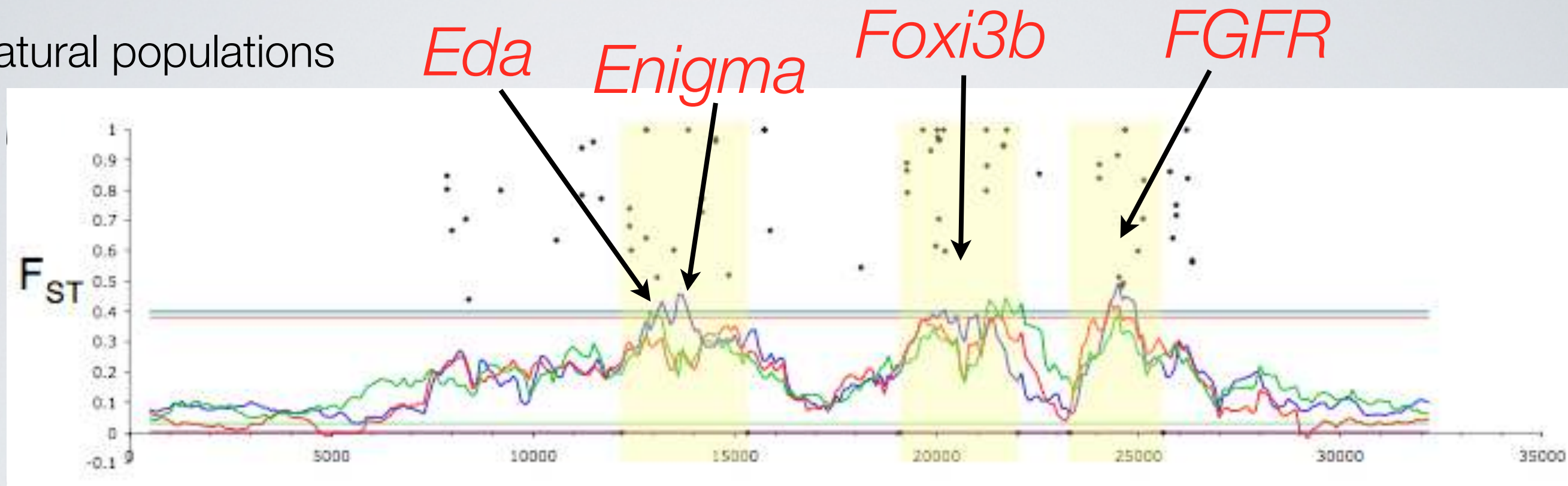
Genomic location (mBases)

Evidence for balancing selection on Linkage Group III

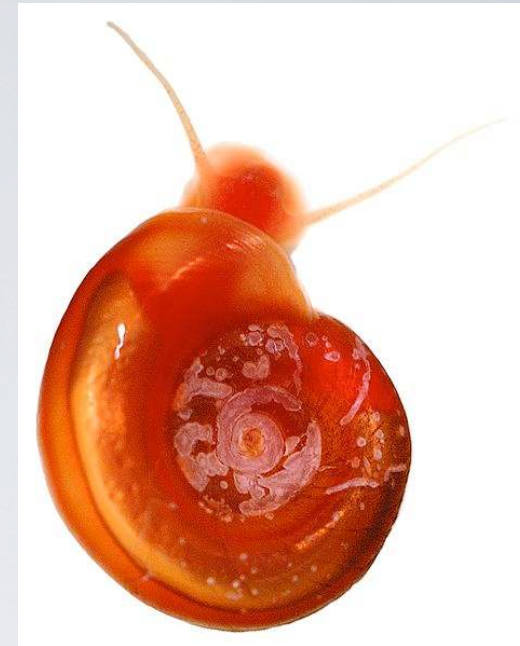
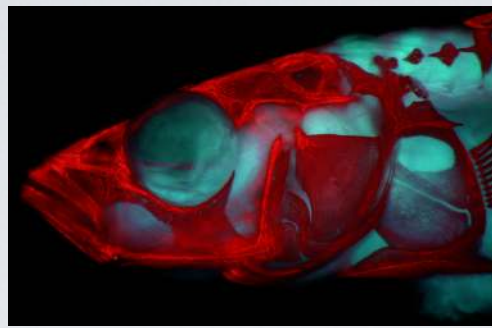


Previously identify quantitative trait loci (QTLs)
are under selection

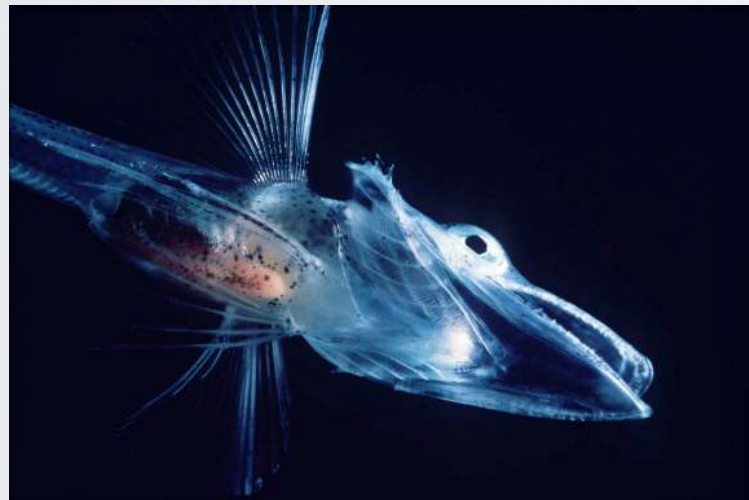
Natural populations



Lateral plate major locus
on LGIV (4000 SNPs)

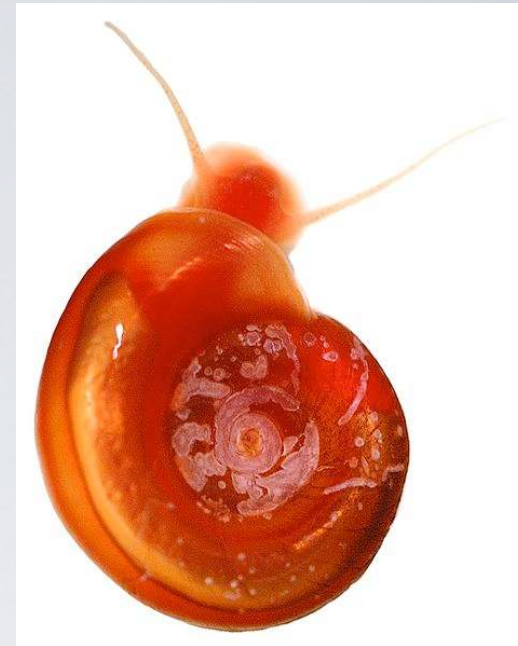
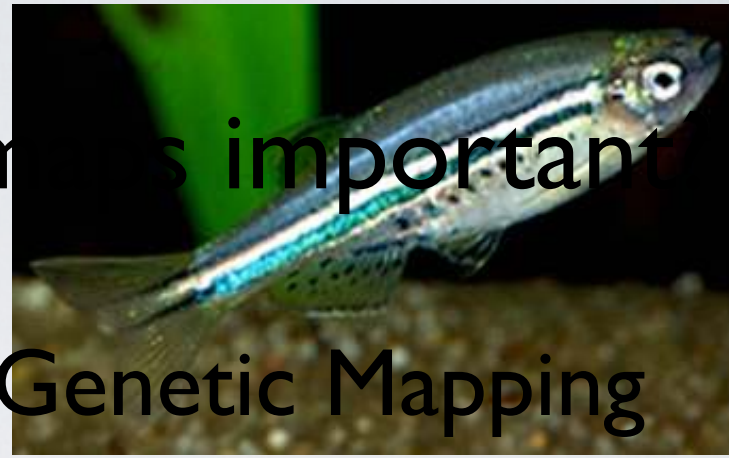


What if you **don't** have a reference genome?



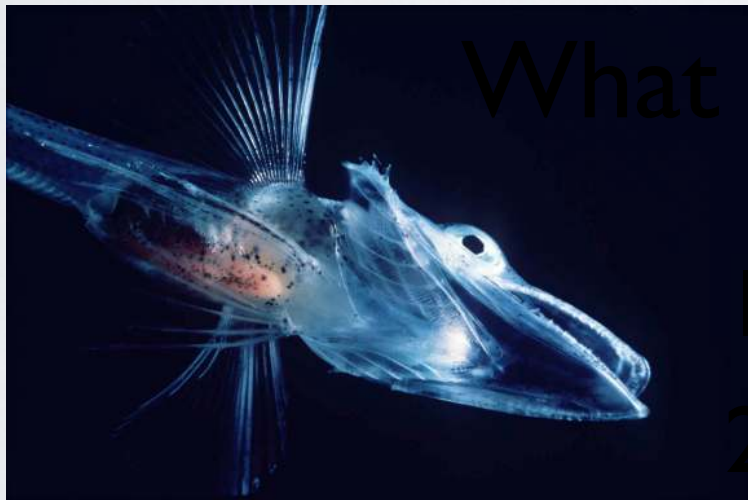
Why are genetic maps important?

- Forward/Reverse Genetic Mapping
- QTL Mapping
- Physical Genome Assembly



What is required to build a map?

1. Generate progeny
2. Recombination
3. Detect polymorphism





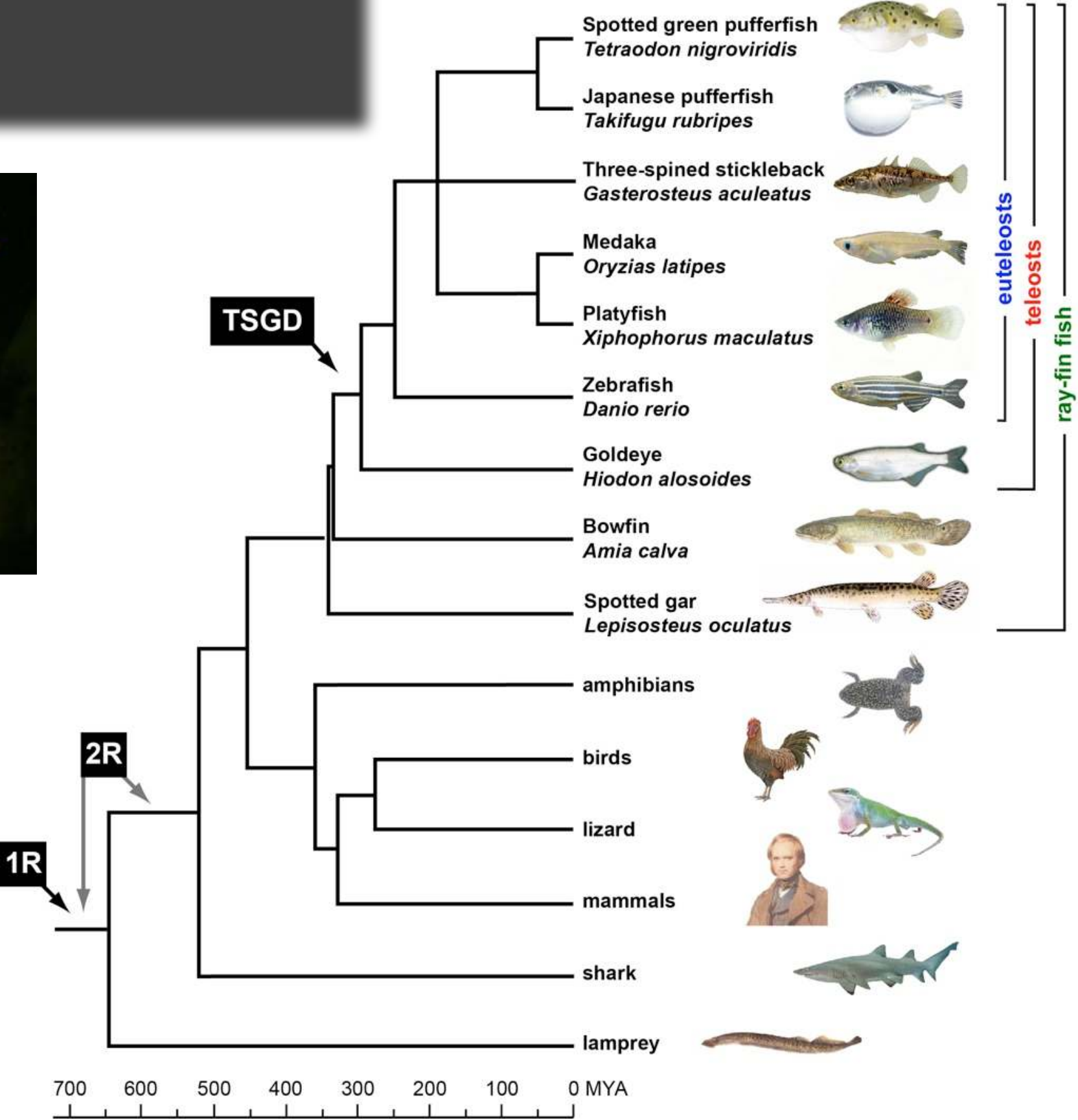
Spotted Gar



Angel Amores



John Postlethwait





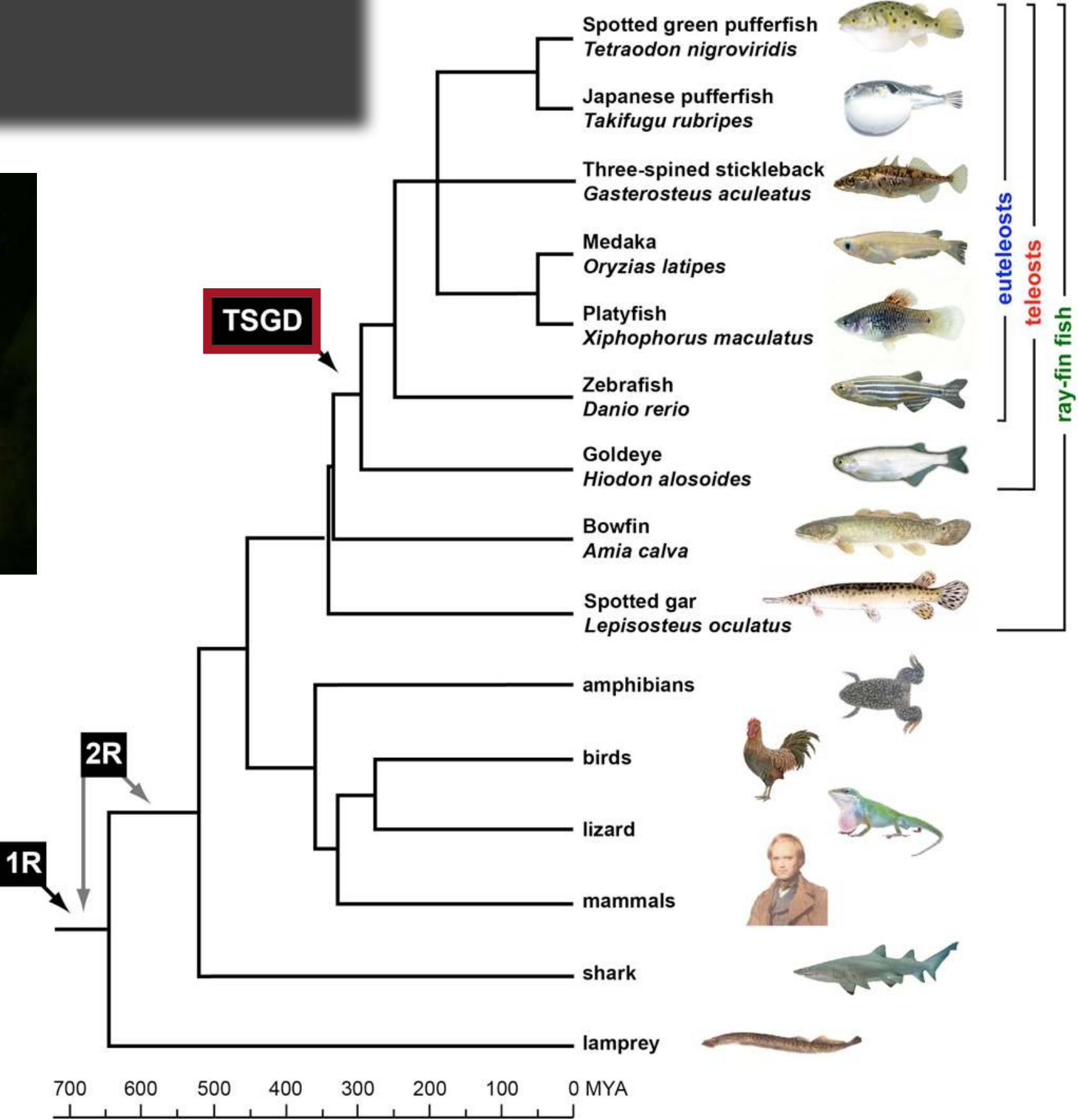
Spotted Gar



Angel Amores

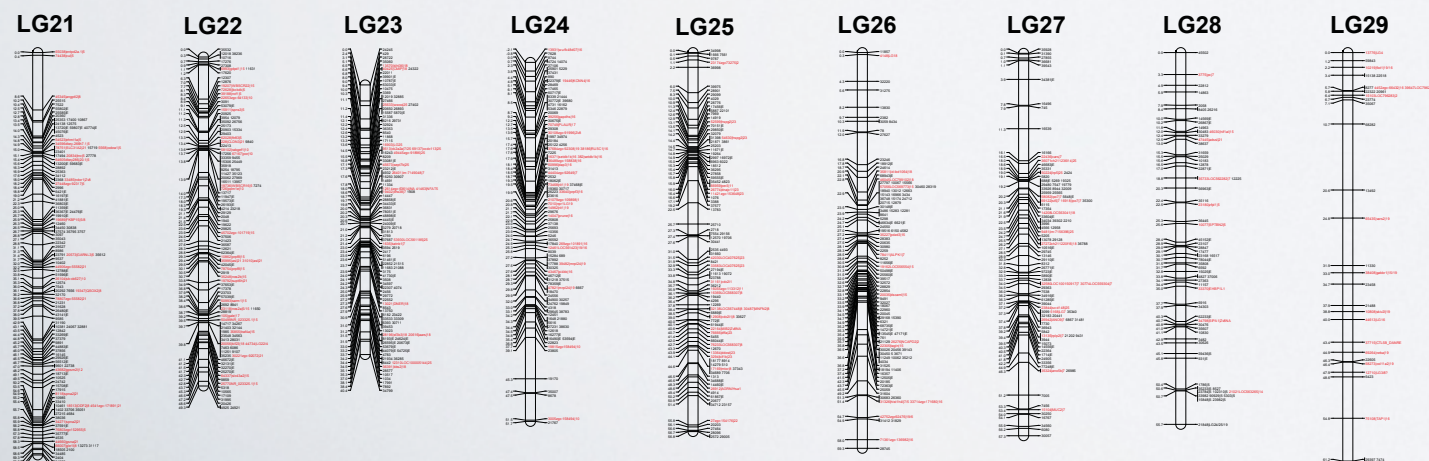
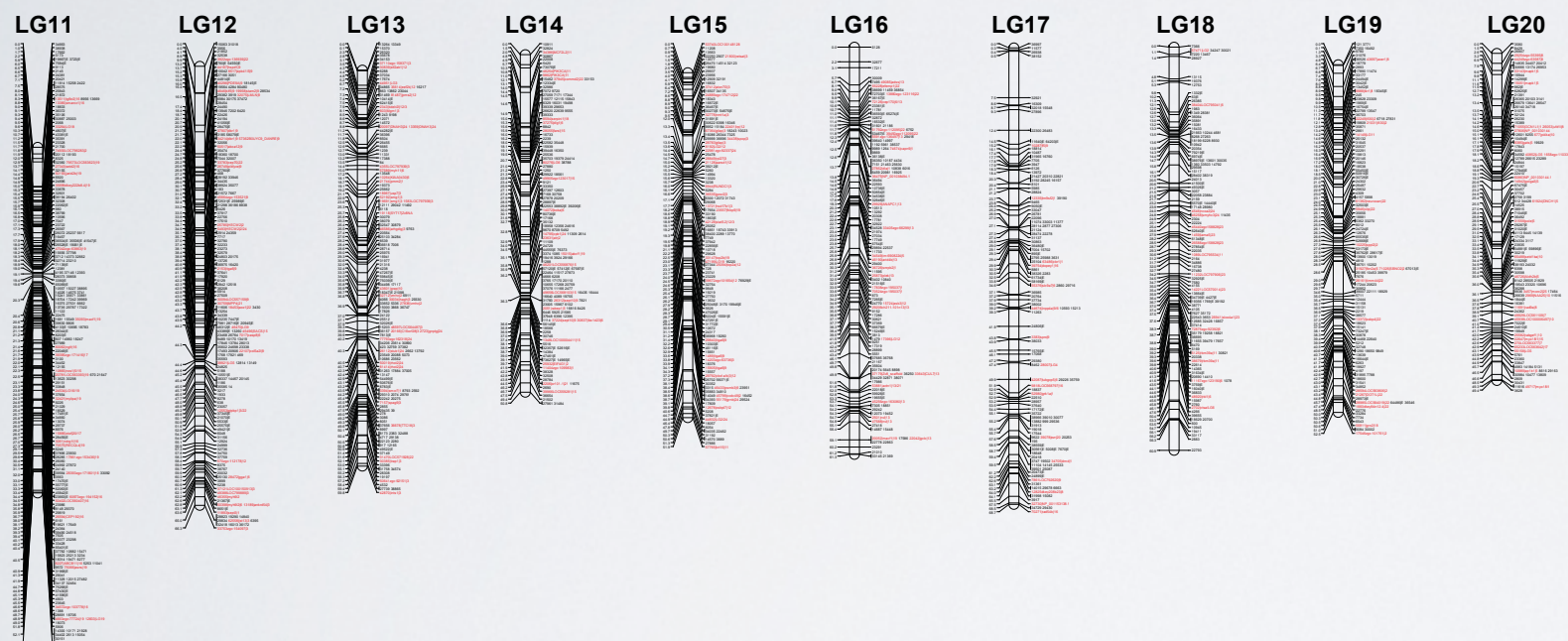
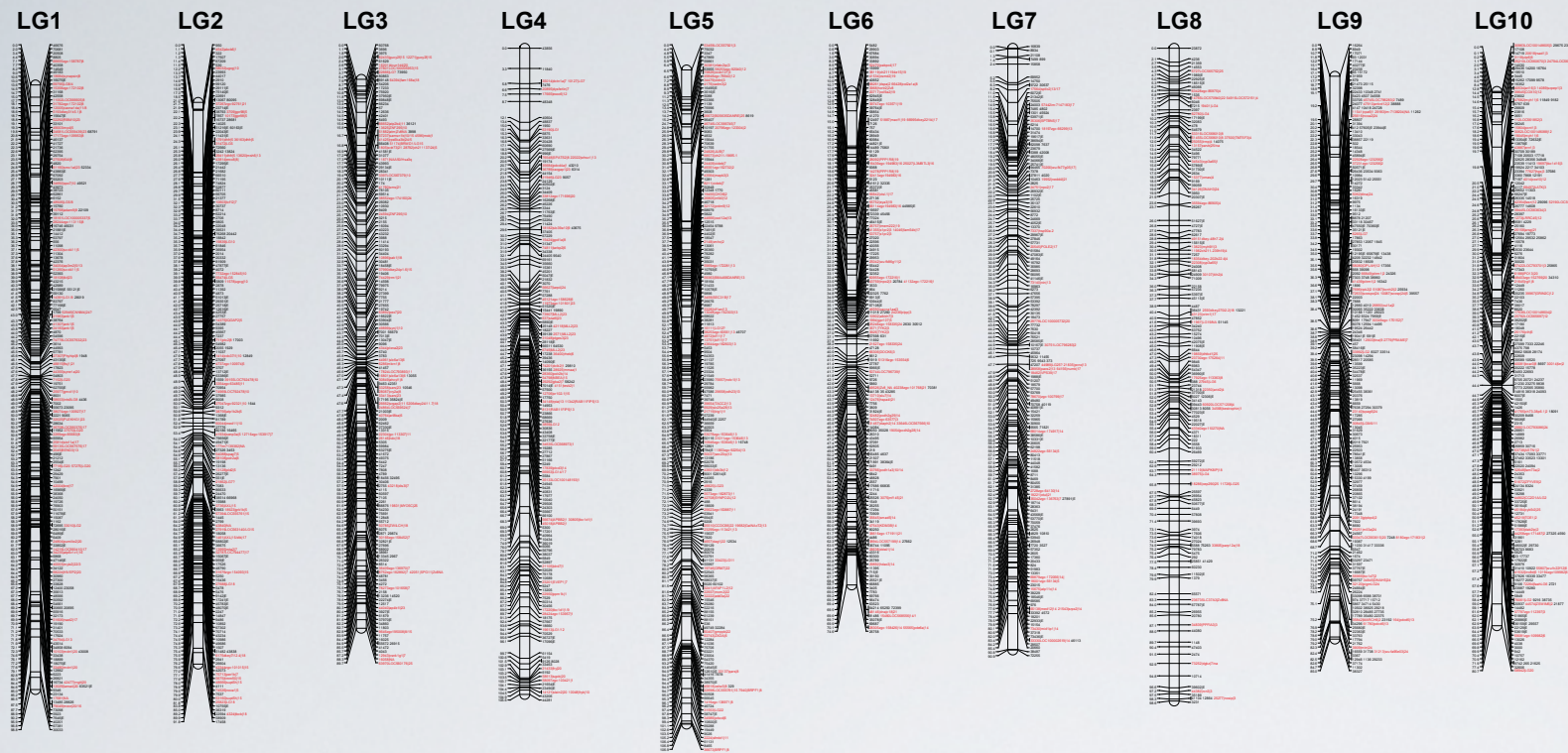


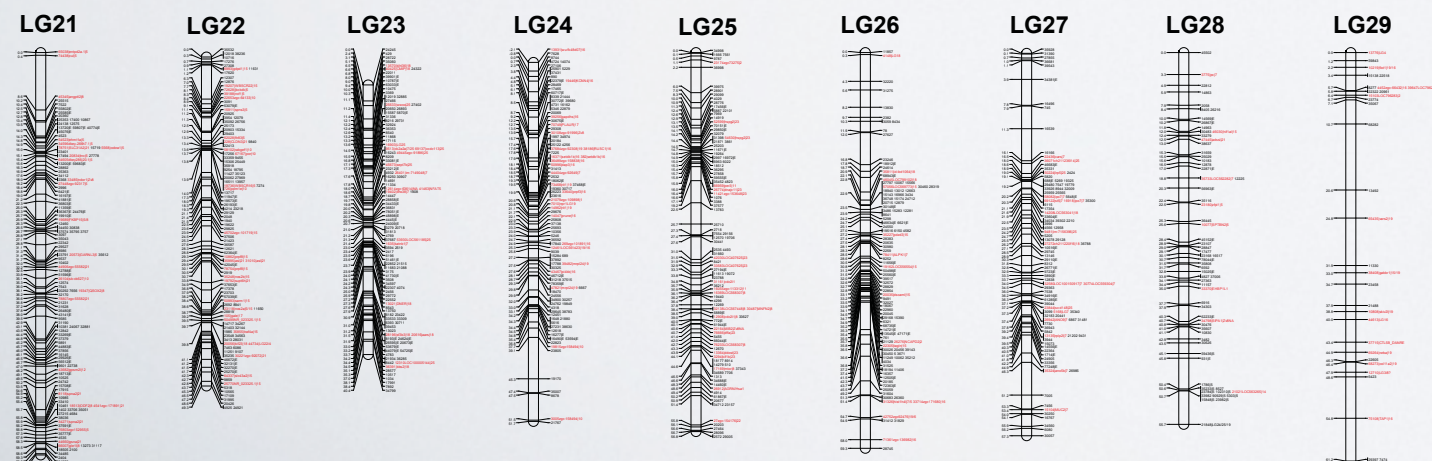
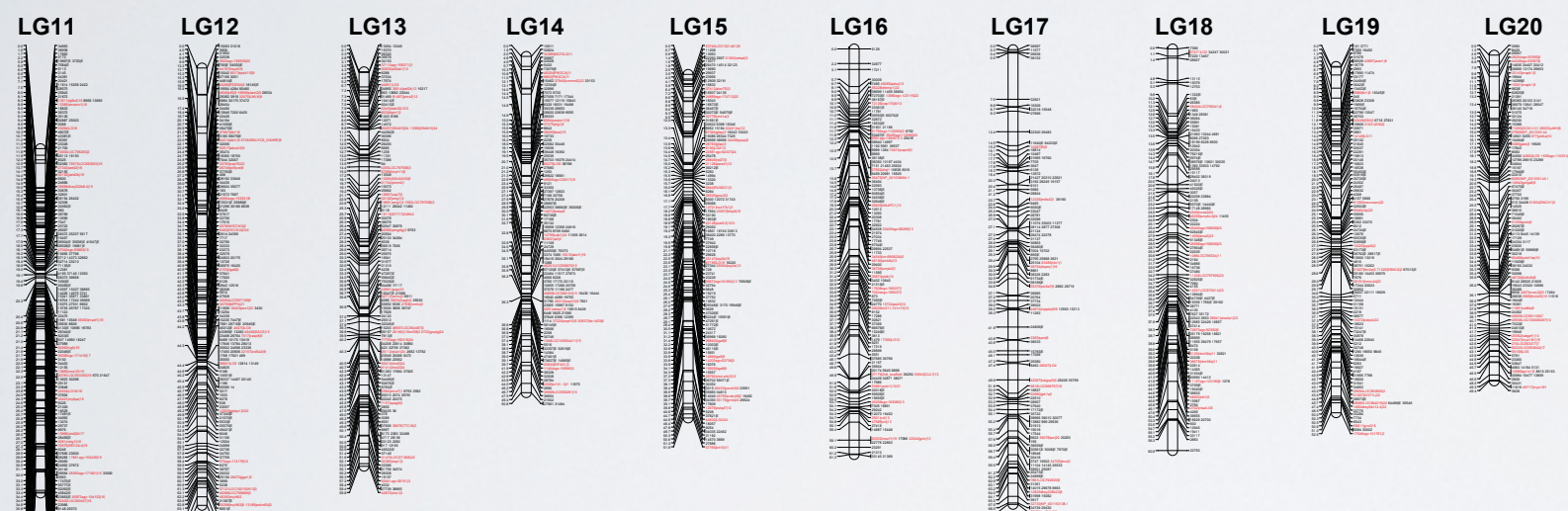
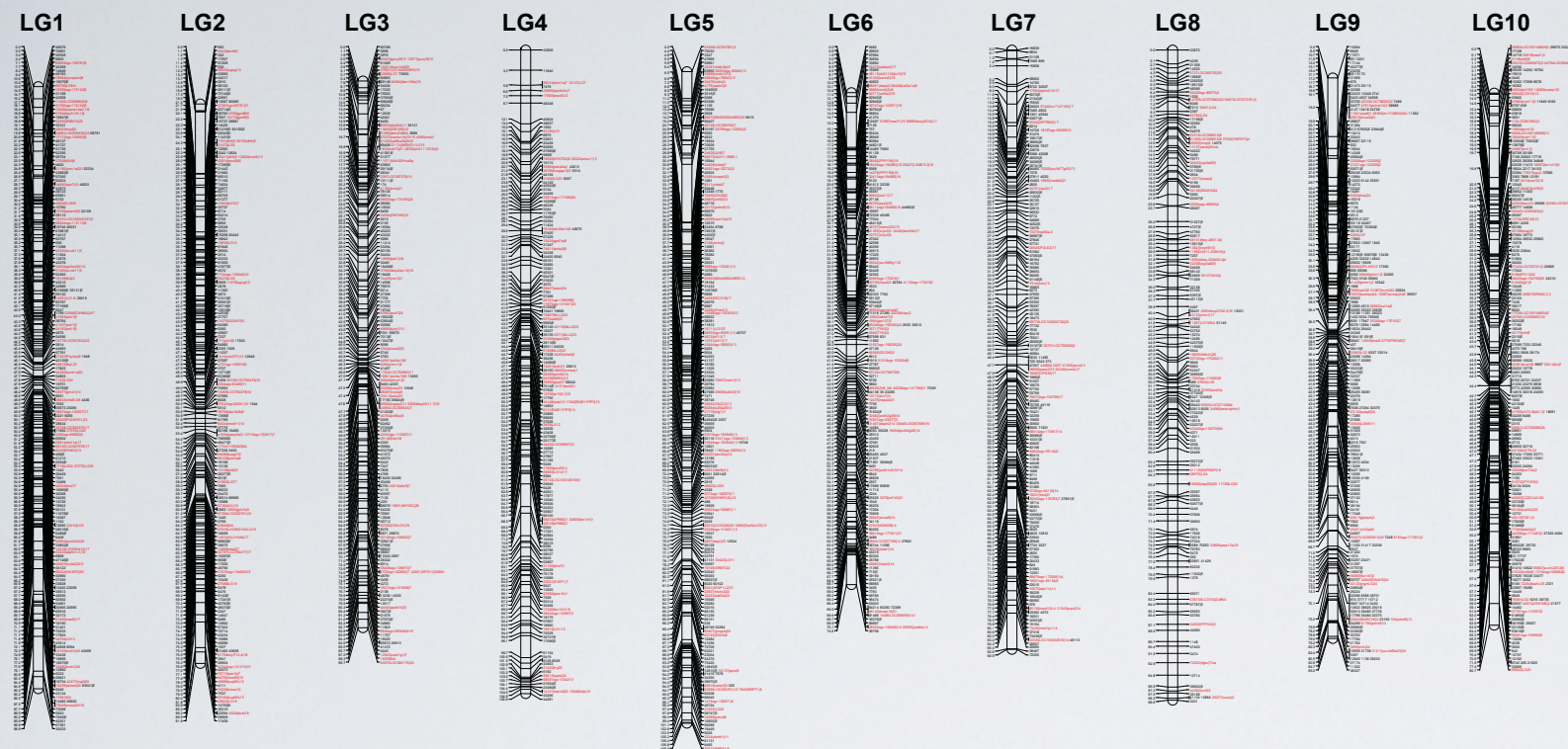
John Postlethwait



F1 Pseudo-Test Cross

94 Individuals
15,076 Markers
8,046 Mapped
974 In Genes

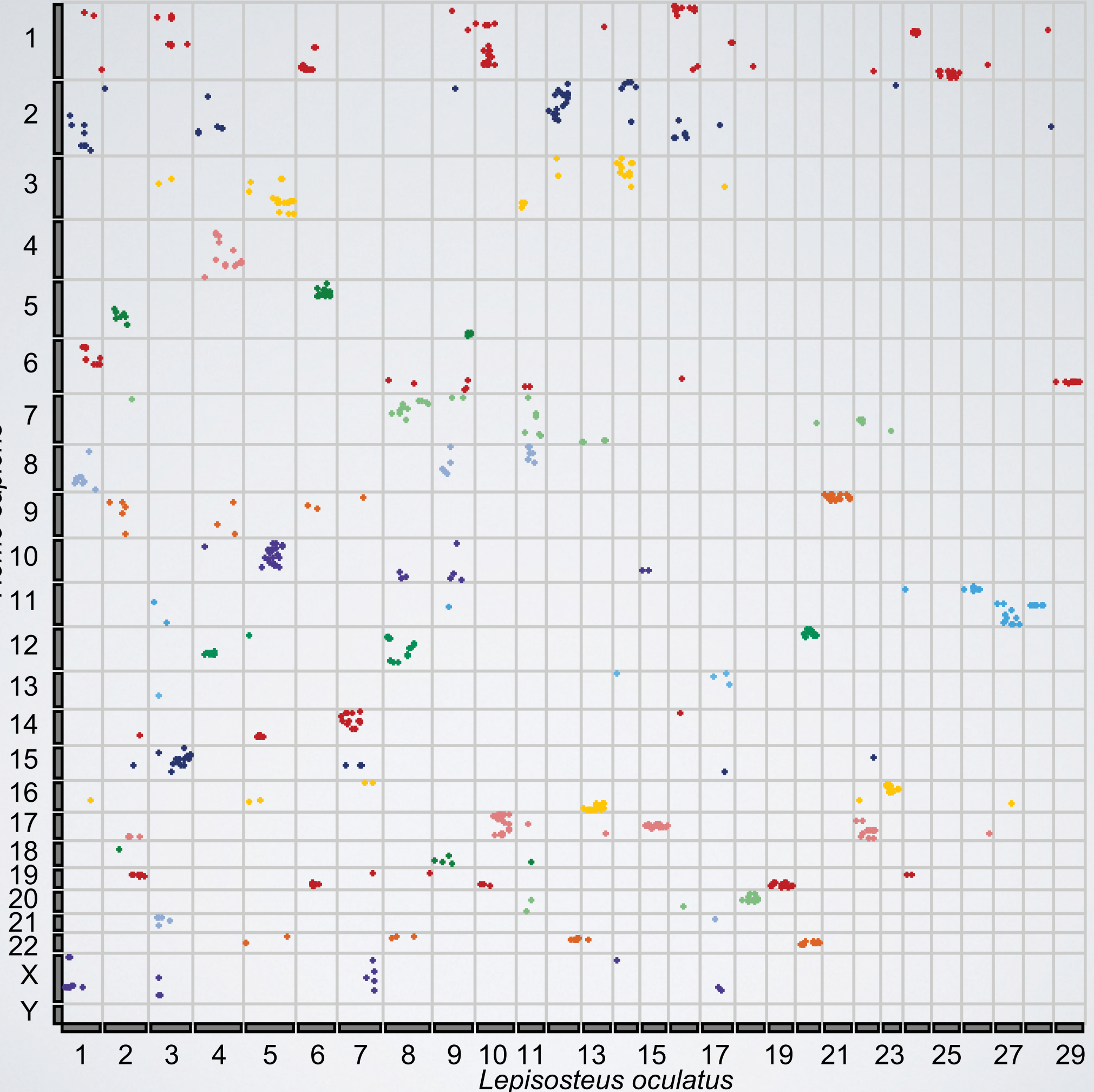


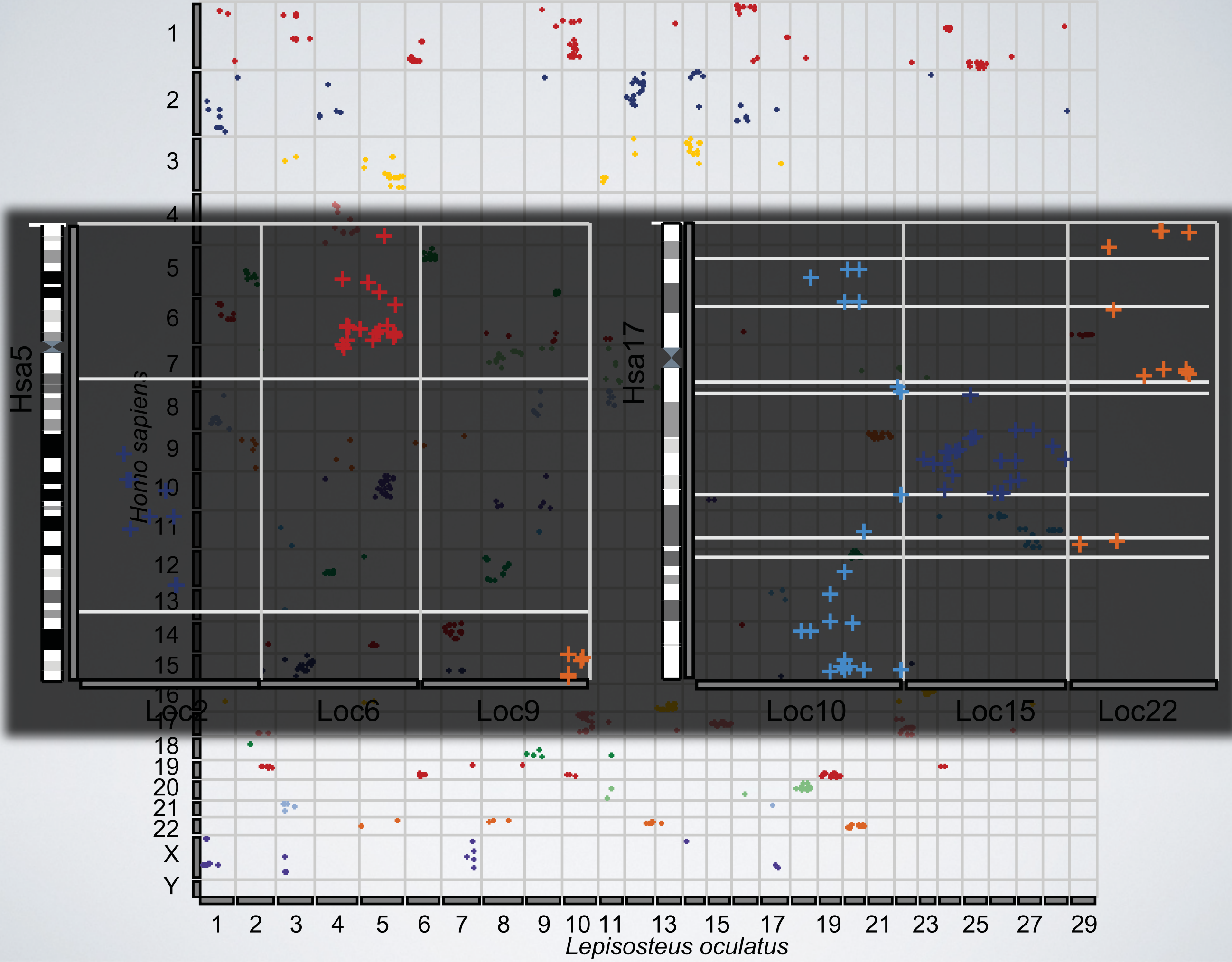


94 Individuals
15,076 Markers
8,046 Mapped
974 In Genes

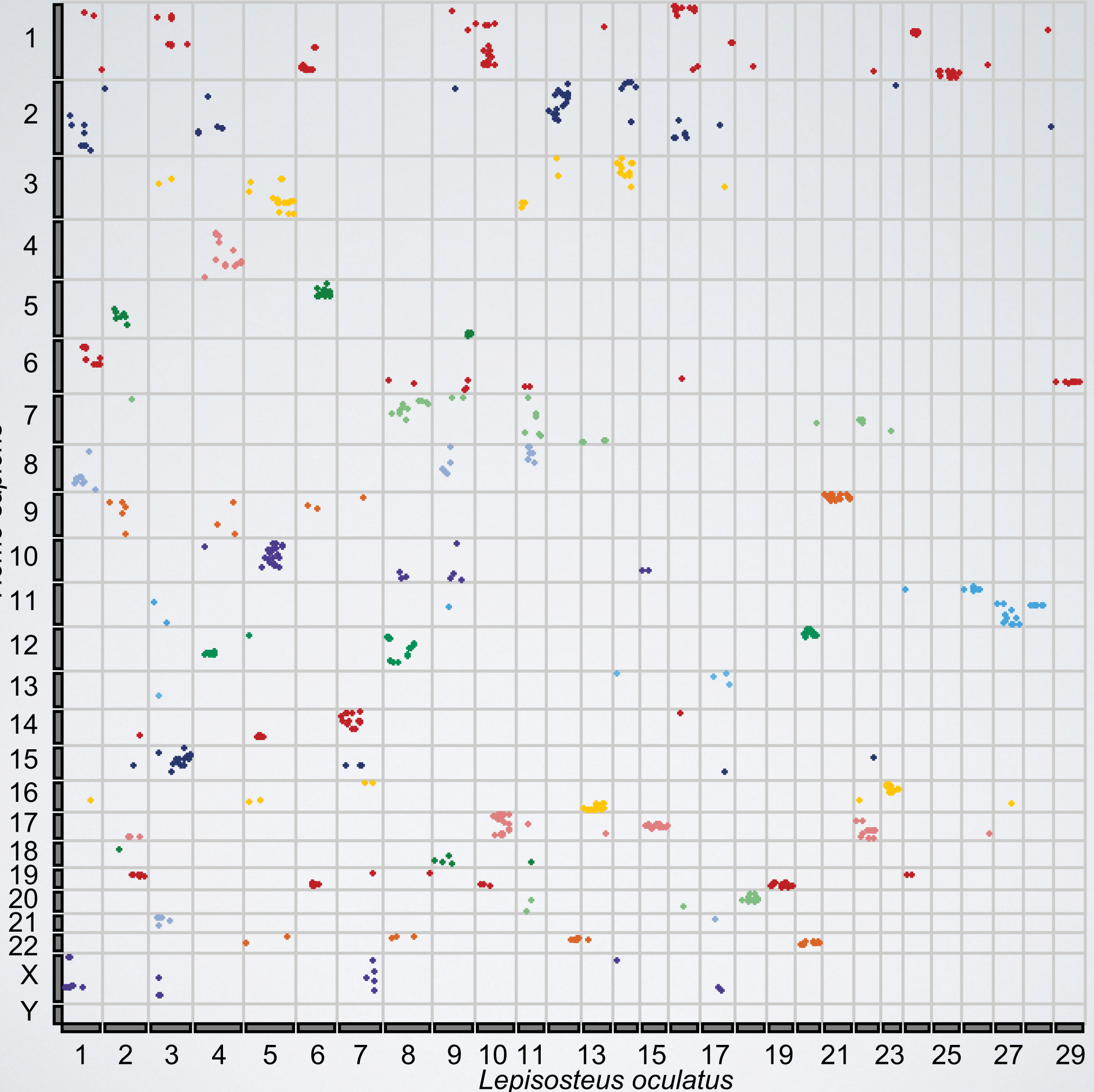
Organism	Markers
Silver carp	483
Guppy	790
Barramundi	240
Catfish	331
Sea bass	368
Cichlid	204
Platyfish	290
Halibut	604
Sea bream	204

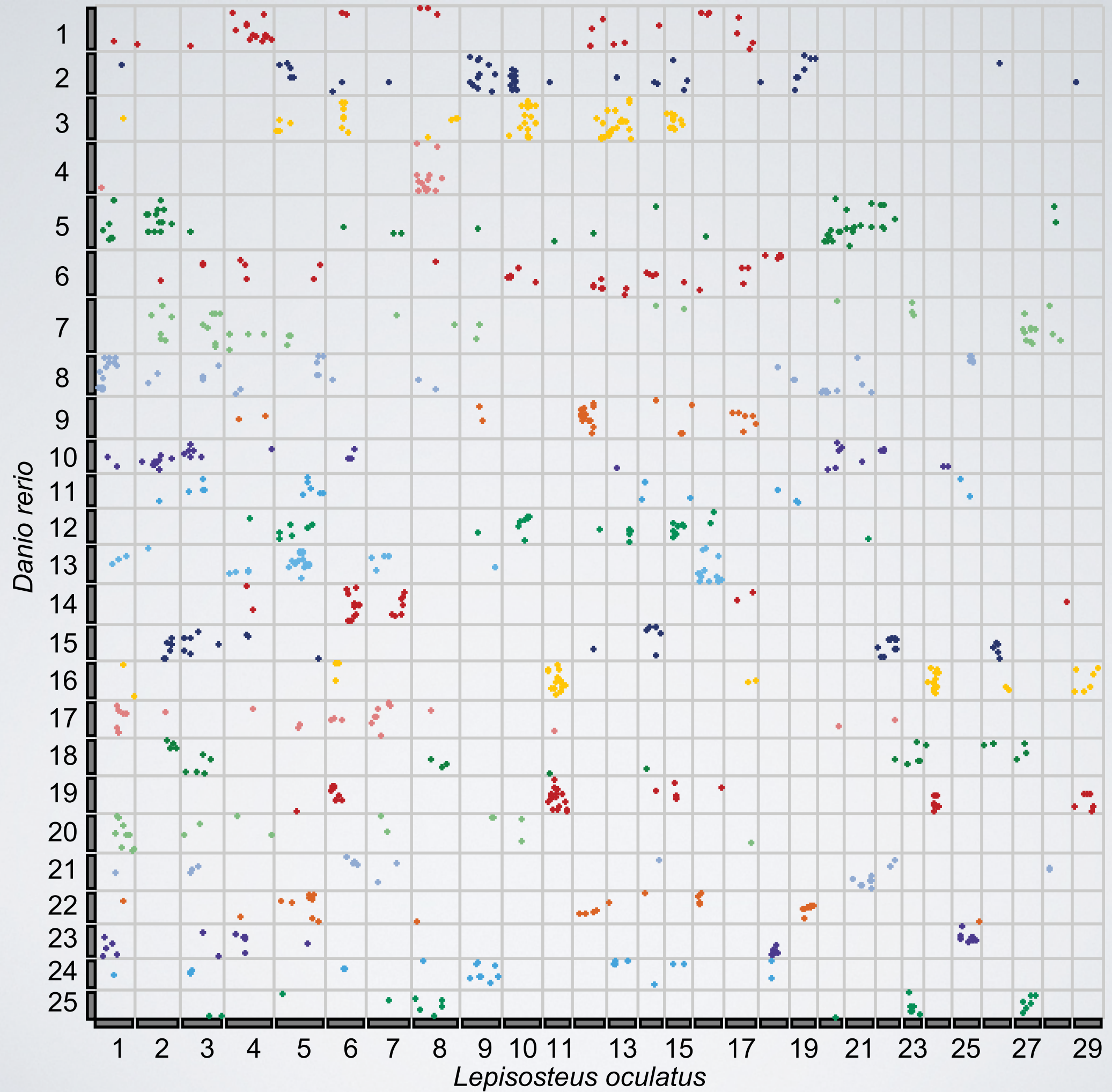
Homo sapiens

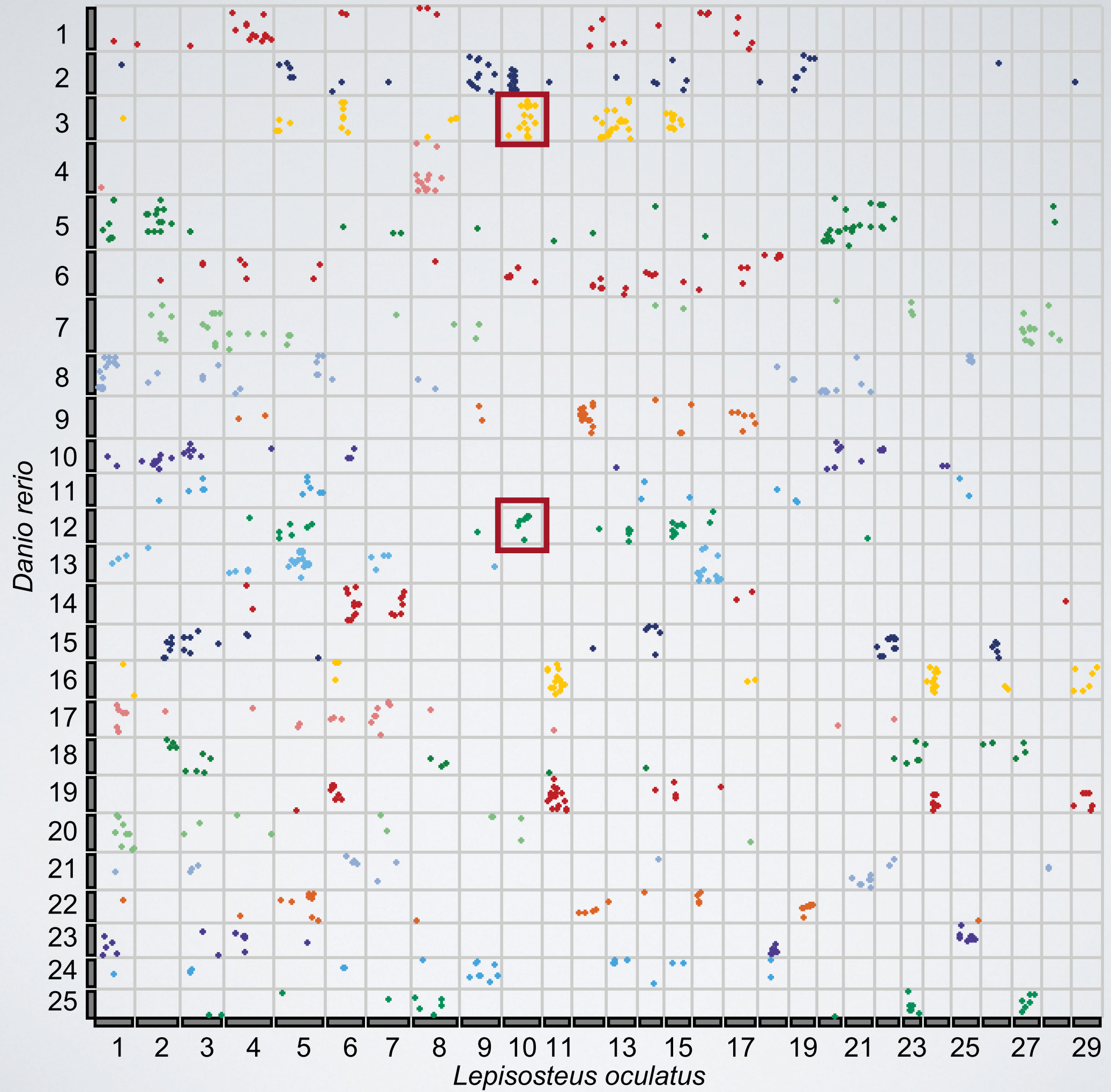


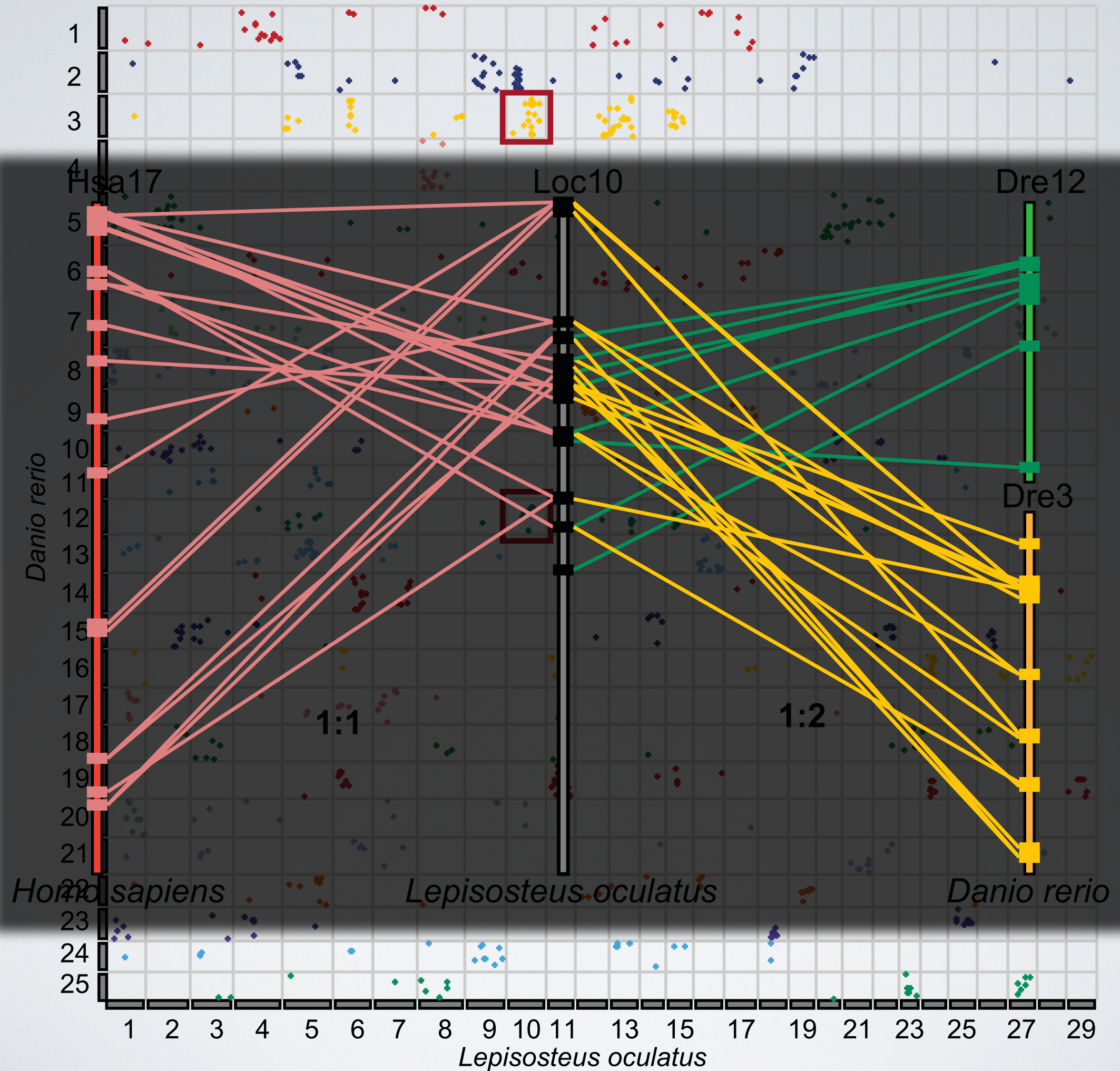


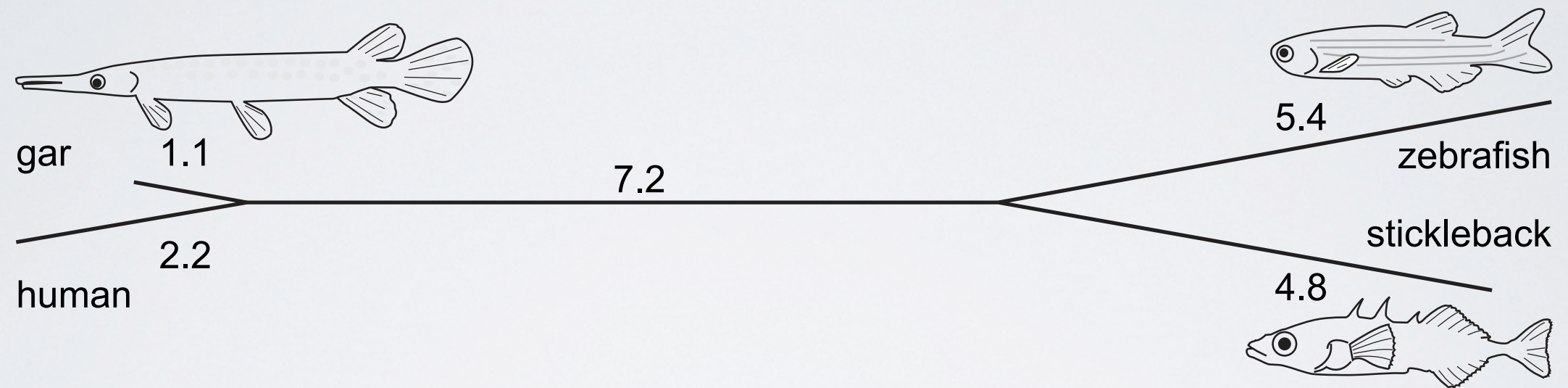
Homo sapiens







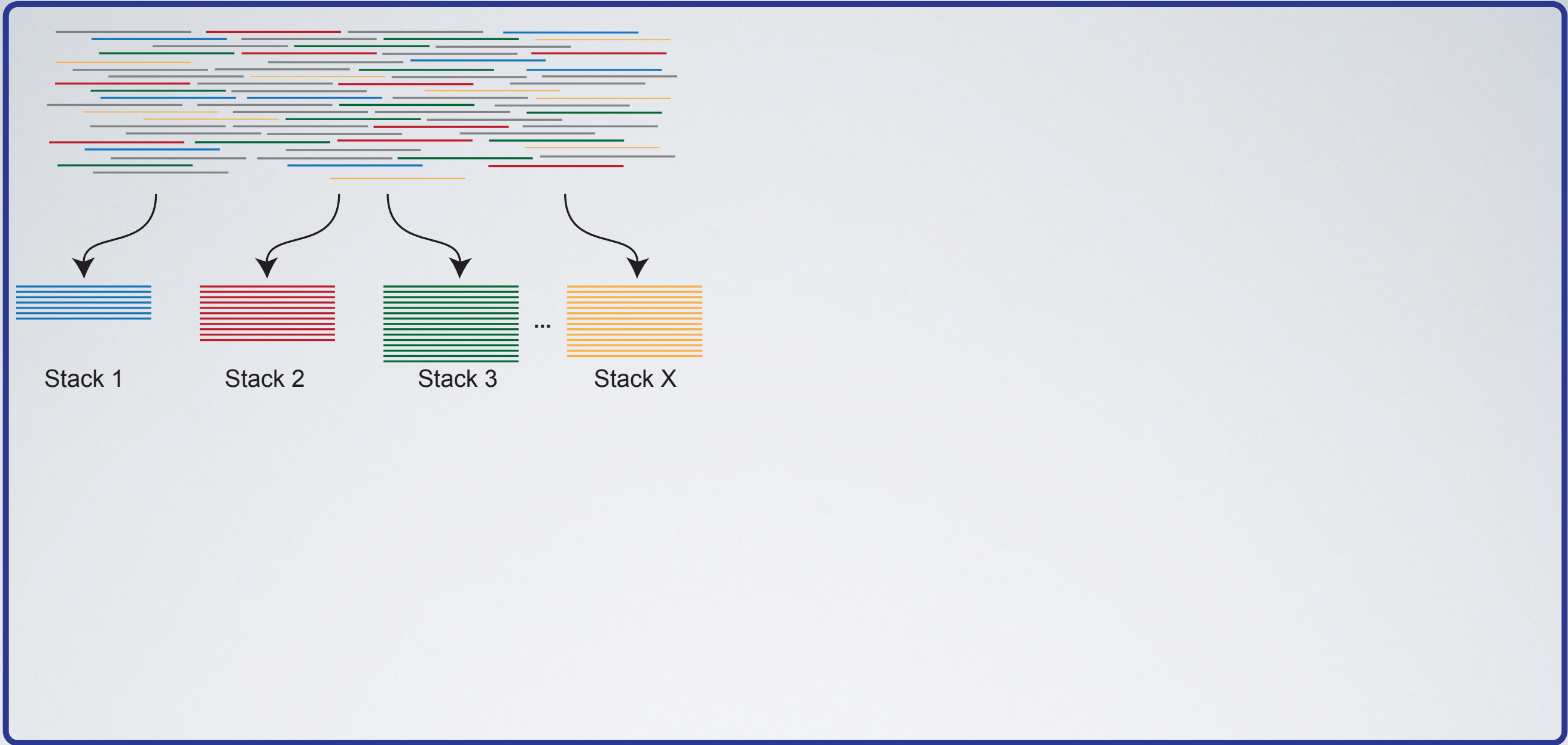






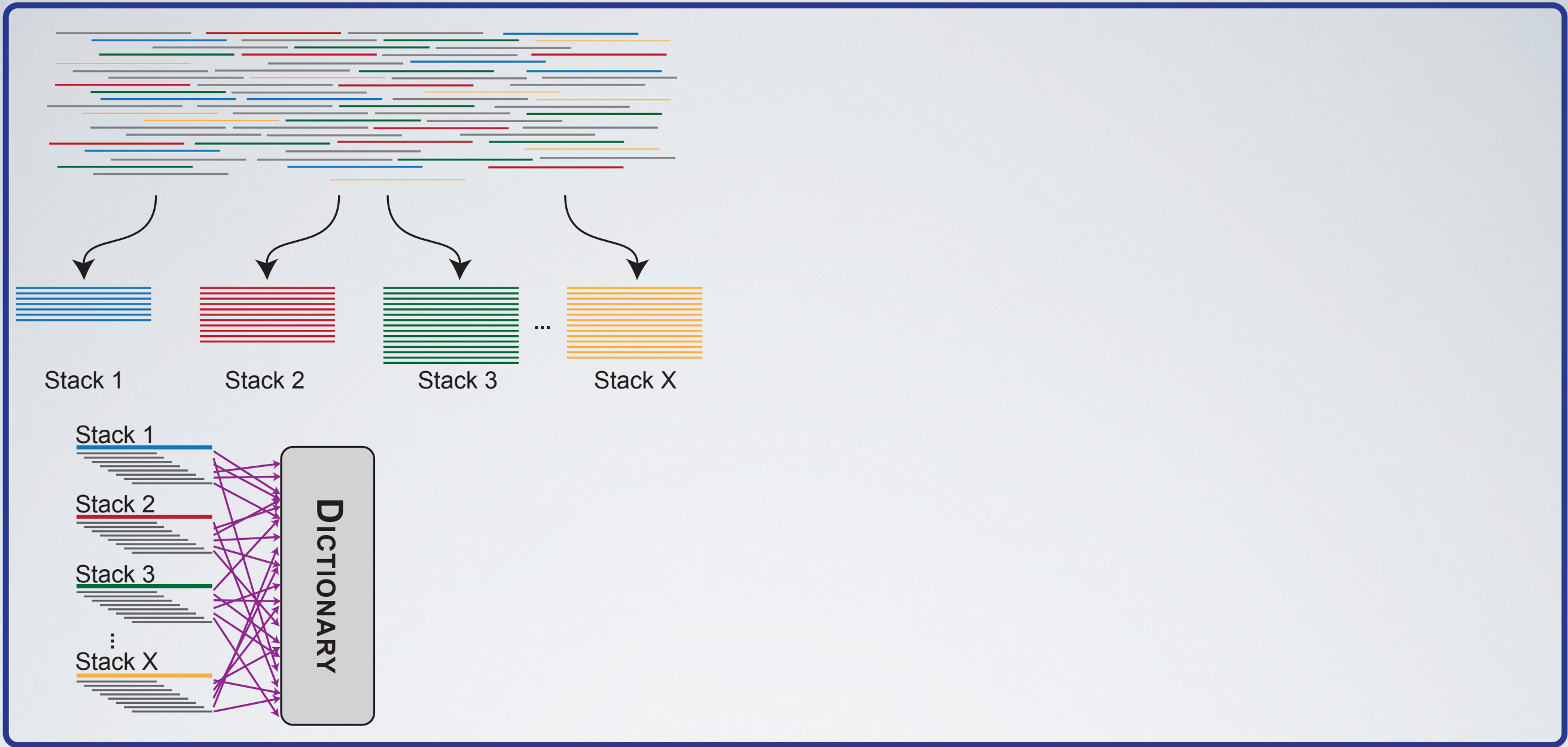
G3: Genes, Genomes, Genetics





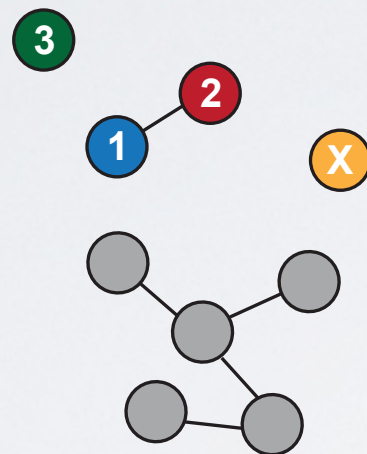
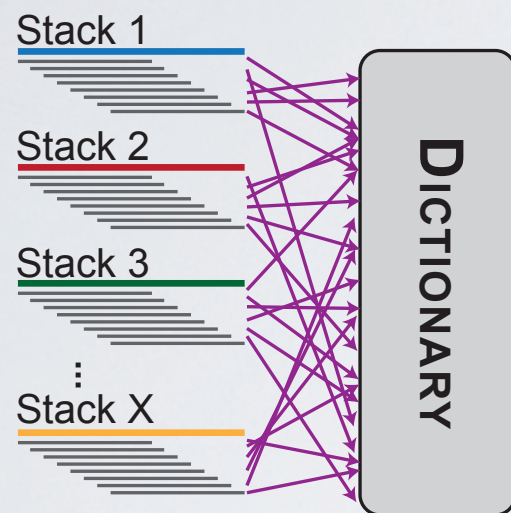
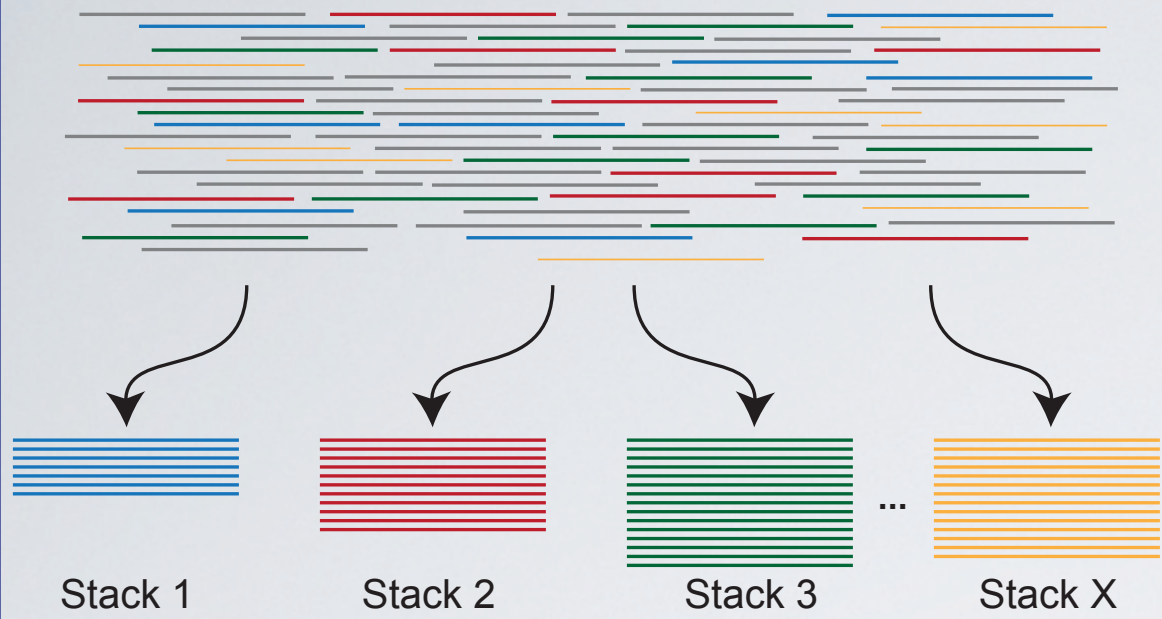
C

CT



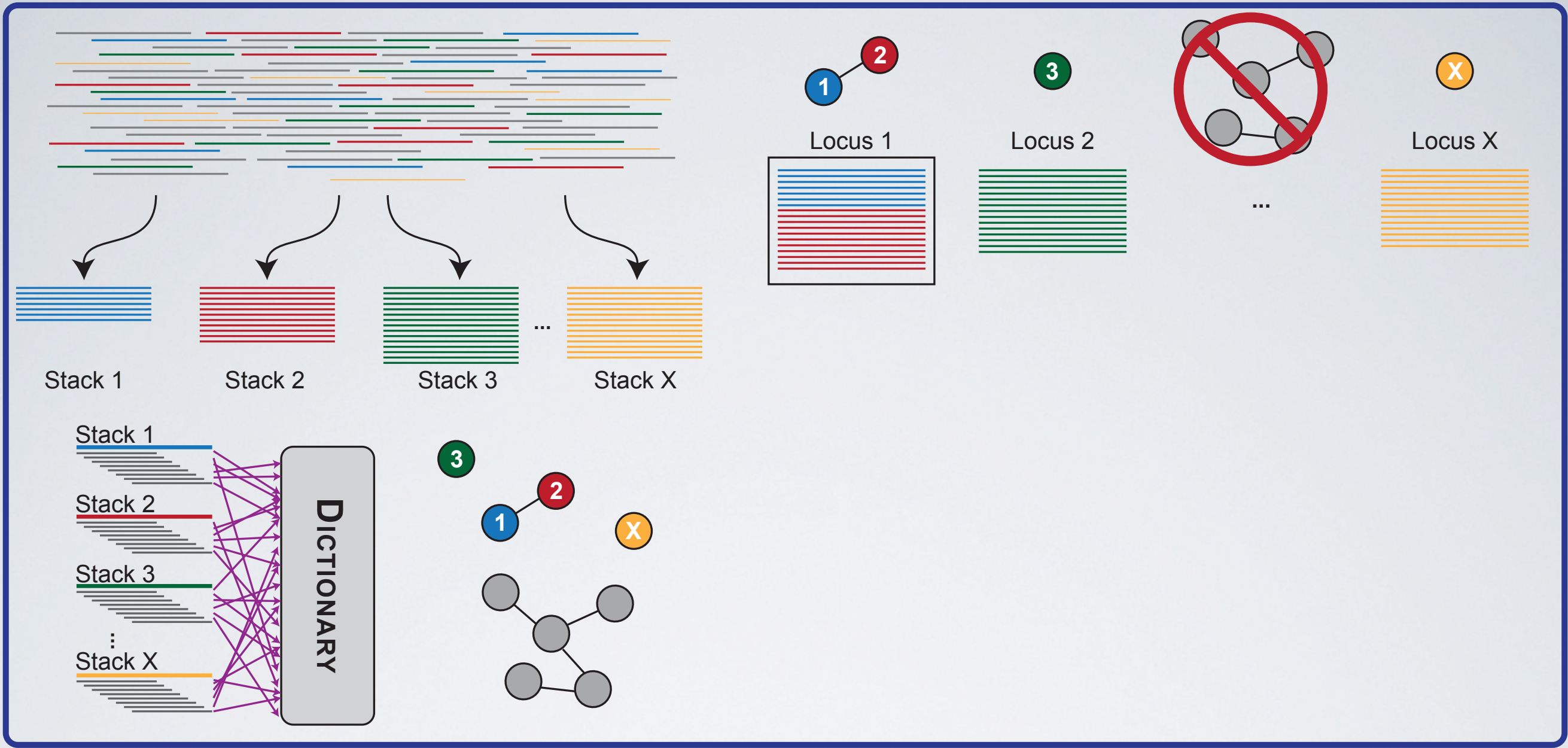
C

CT



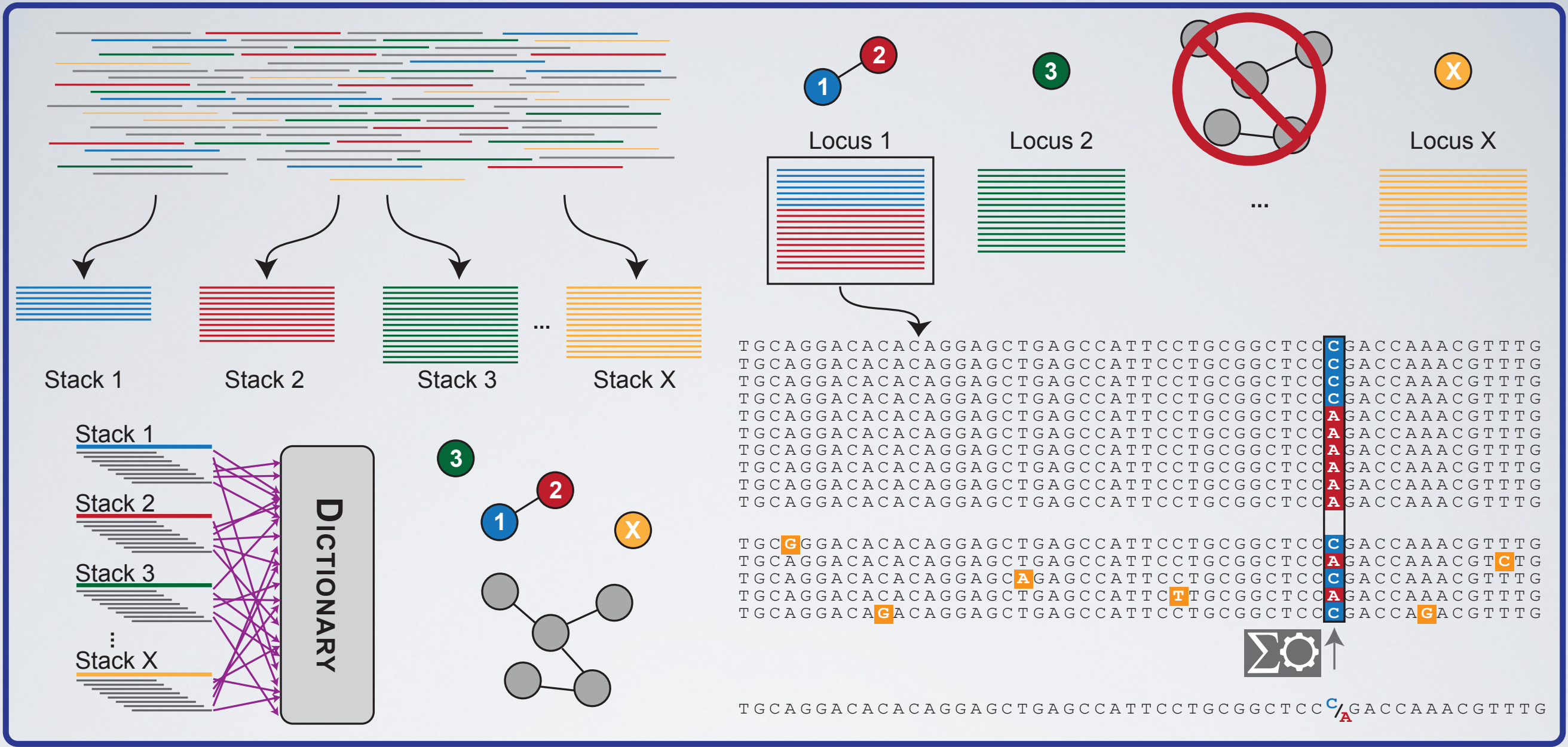
C

CT



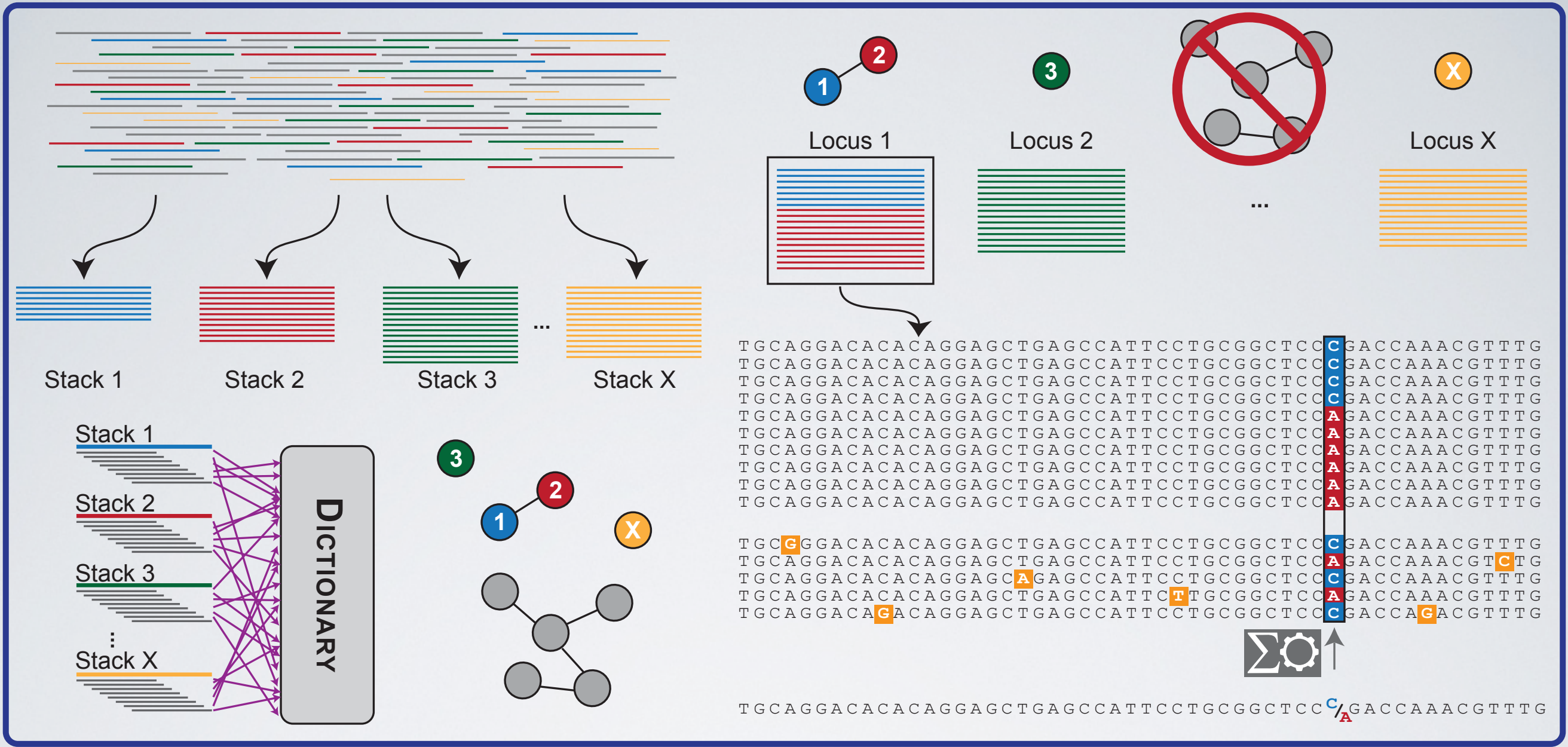
C

CT



C

CT



C

CT



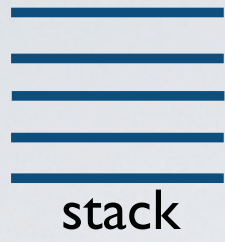
Stacks







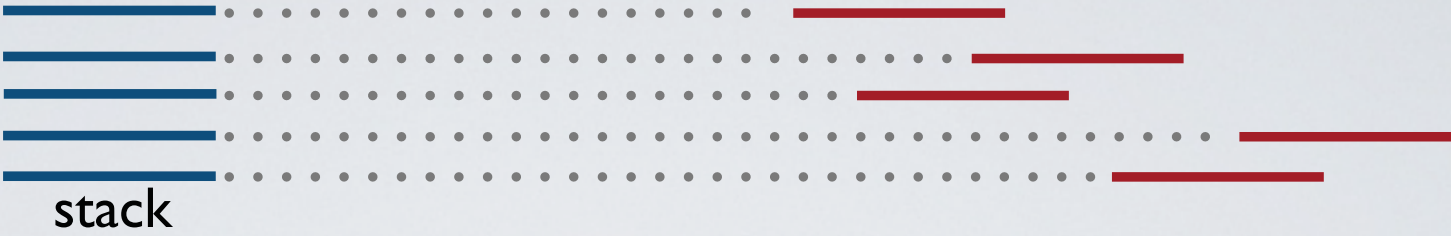




Acquire
paired-end
sequence



Acquire
paired-end
sequence



Match to marker catalog

TGCAGGGGTATTAGCATAA



Collate/Assemble PE reads

AACTAATTTTCACTAGCCATCTTGAATGTGAGTAGCATTTTAAGTAACTATAATTG



Acquire
paired-end
sequence



Match to marker catalog

TGCAGGGGTATTAGCATAA

Collate/Assemble PE reads

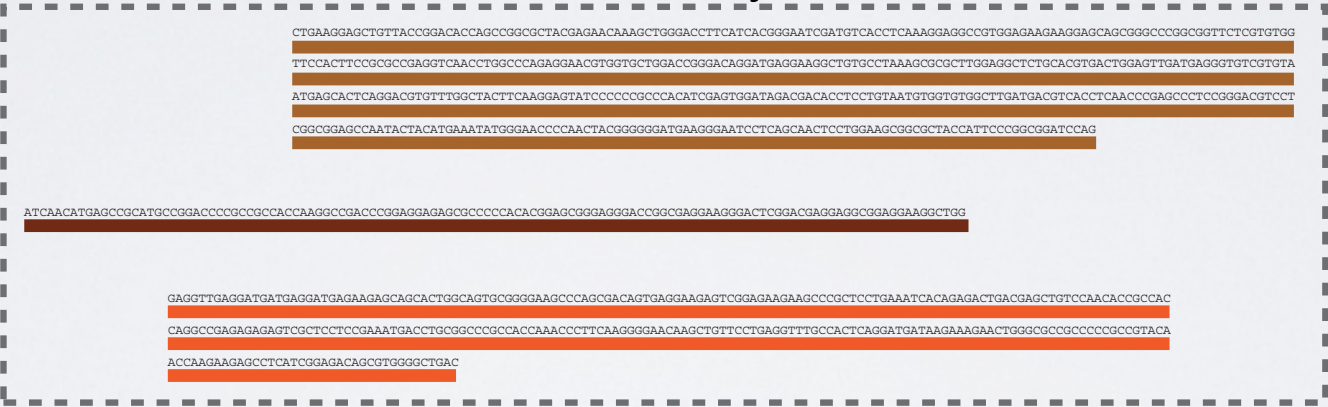
AACTAATTTTCACTAGCCATCTTGAATGTGAGTAGCATTTTAAGTAACTATAATTG

Associate
markers / PE
contigs with
ESTs

BLASTn

BLASTn

EST Library



Acquire
paired-end
sequence



Match to marker catalog

TGCAGGGGTATTAGCATAA

Collate/Assemble PE reads

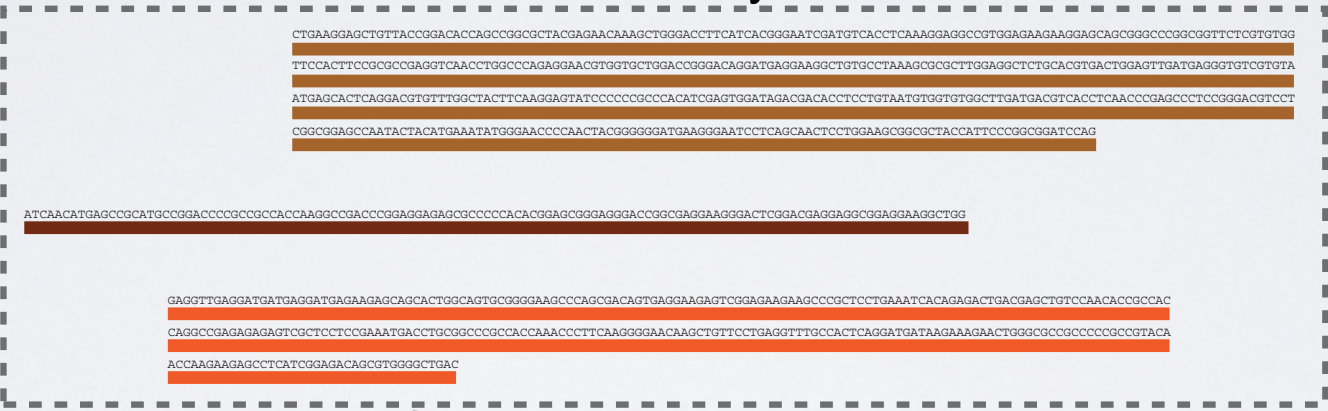
AACTAATTTTTCAGTAGCCATCTTGAATGTGAGTAGCATTTTAAGTAACTATAATTG

Associate
markers / PE
contigs with
ESTs

BLASTn

BLASTn

EST Library



BLASTx

BLASTx

BLASTx

Human
Genome

Zebrafish
Genome

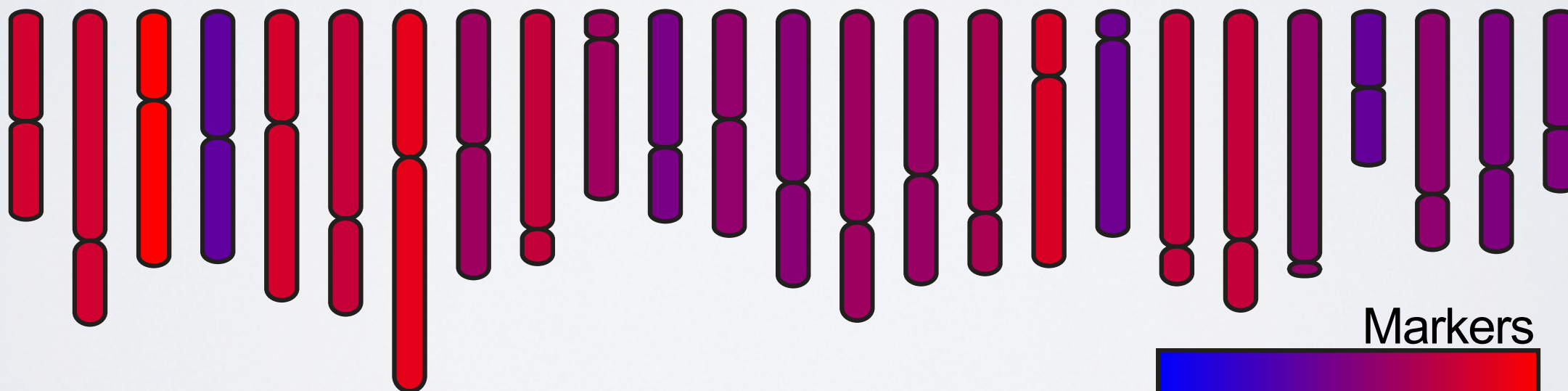
Assign
orthology to:
markers
PE contigs
ESTs

HSMMap

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25

cM

109 157 125 123 144 157 176 135 127 85 101 107 135 153 135 132 125 136 127 158 135 72 135 109 88
103 155 126 124 143 150 188 132 125 93 104 111 136 153 134 130 126 111 135 148 131 76 118 119 89



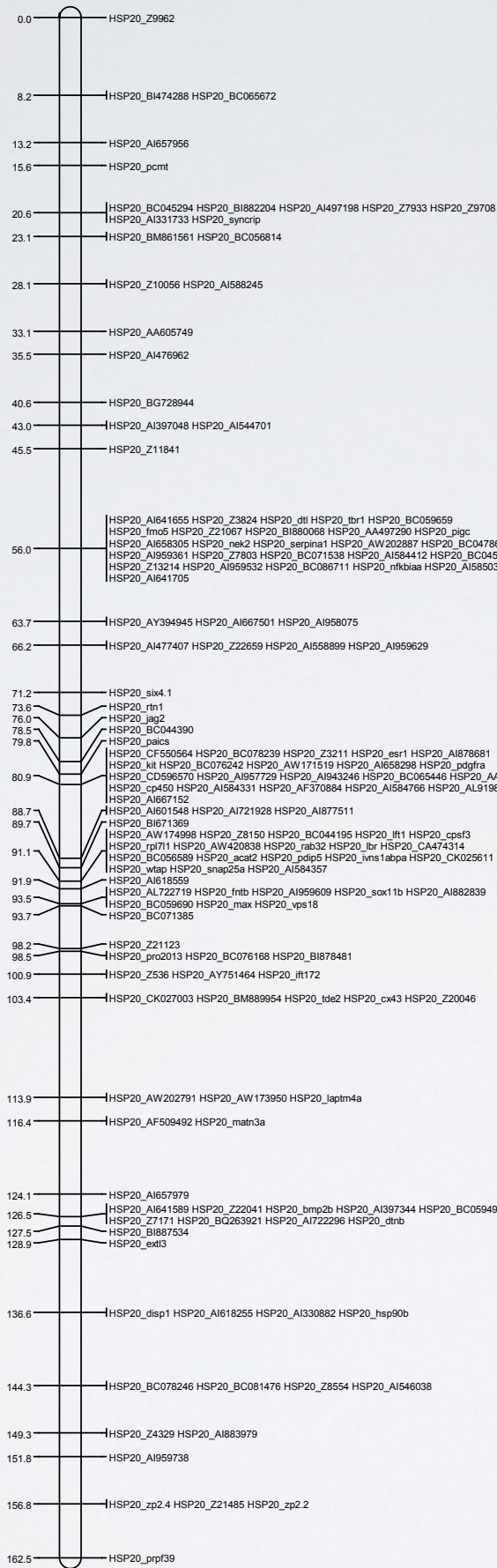
RADMap

Markers

114 153 420

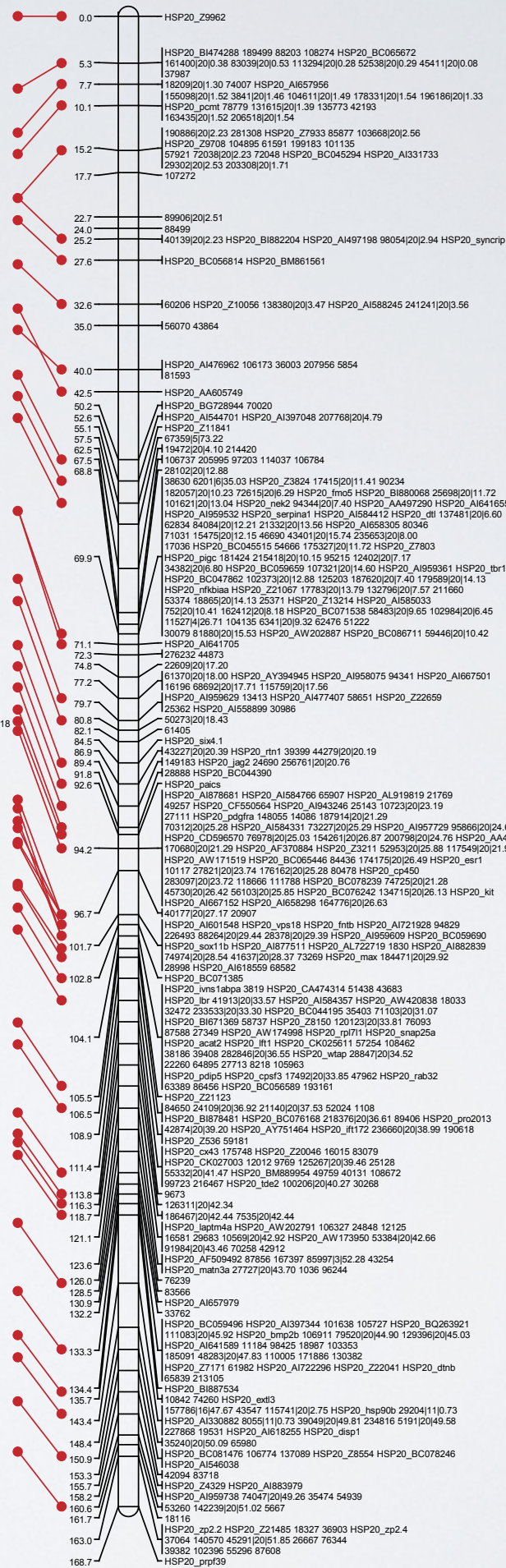
DreLG20

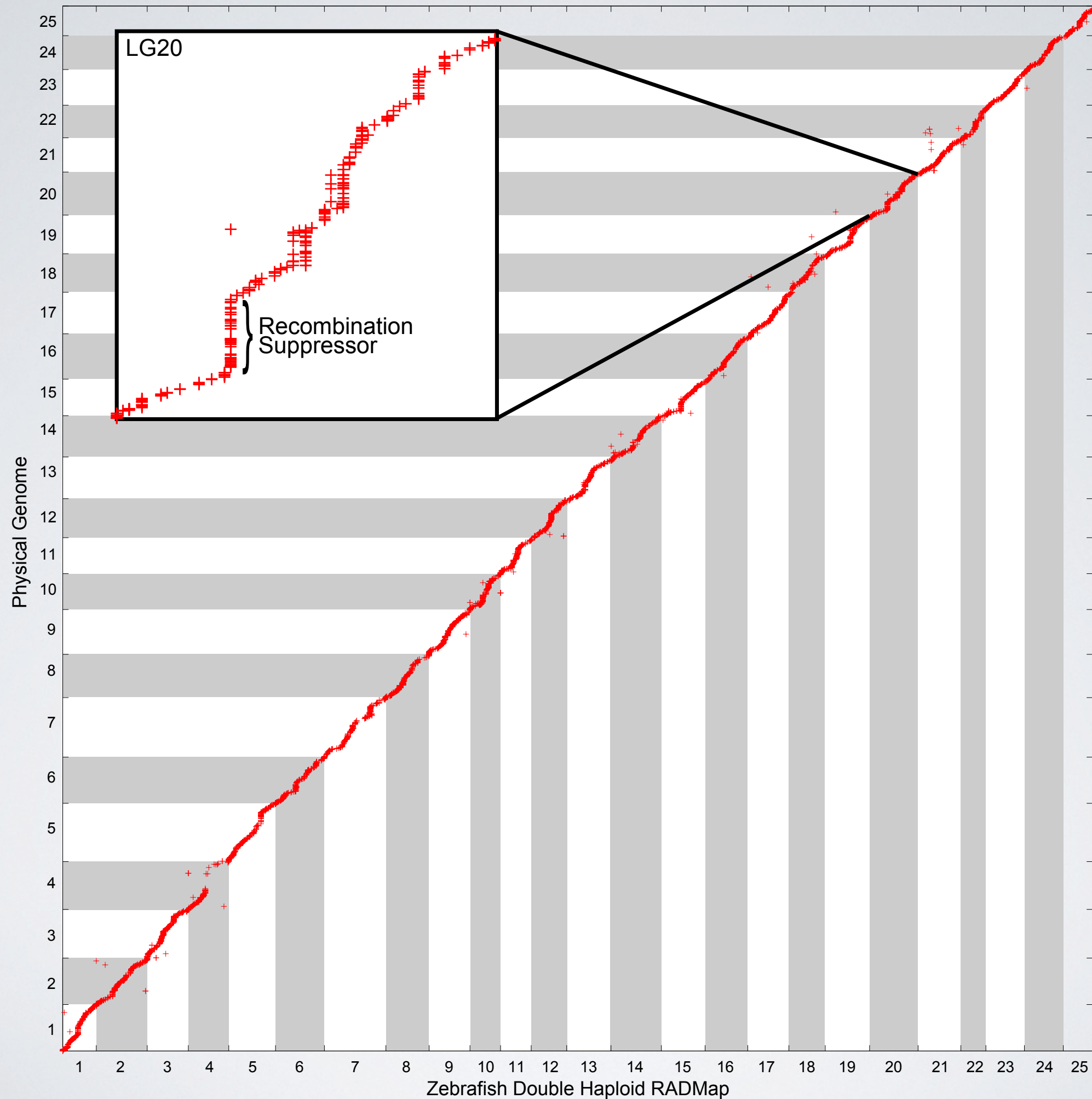
157 EST markers



DreLG20

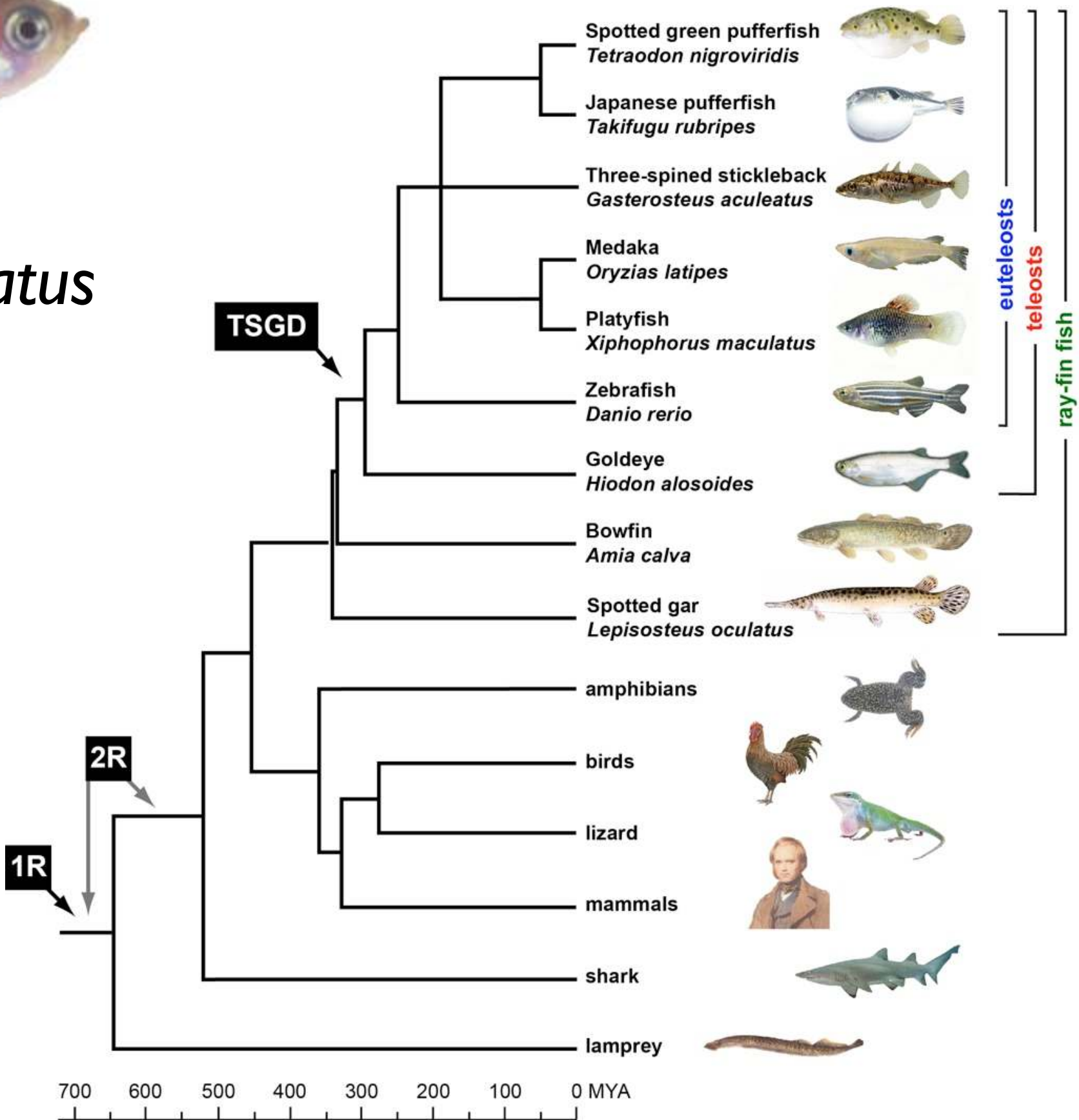
311 RAD markers







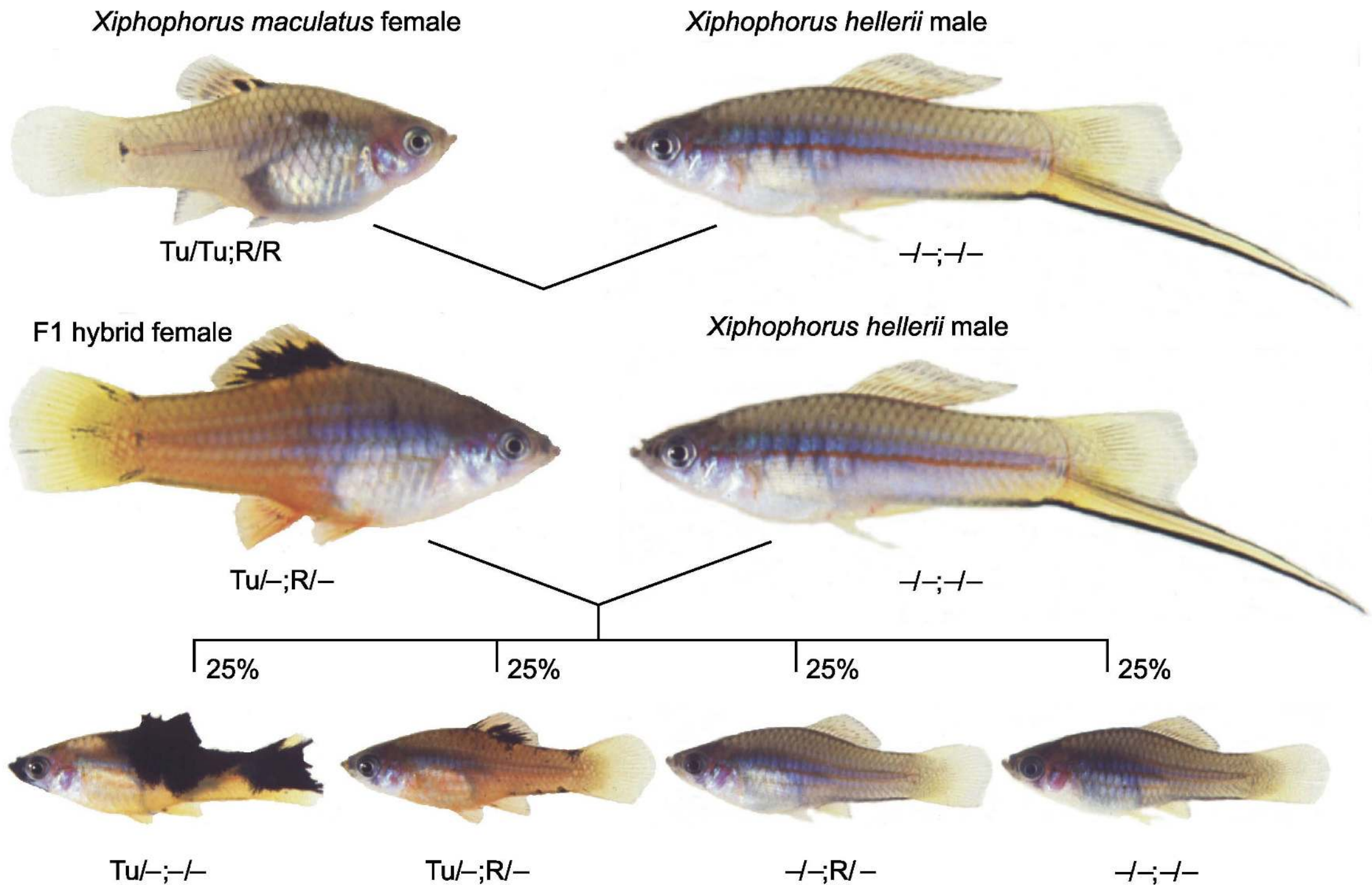
Xiphophorus maculatus



Angel
Amores

John
Postlethwait

Genetic Mapping for Genome Assembly



Version 1.0



290 markers

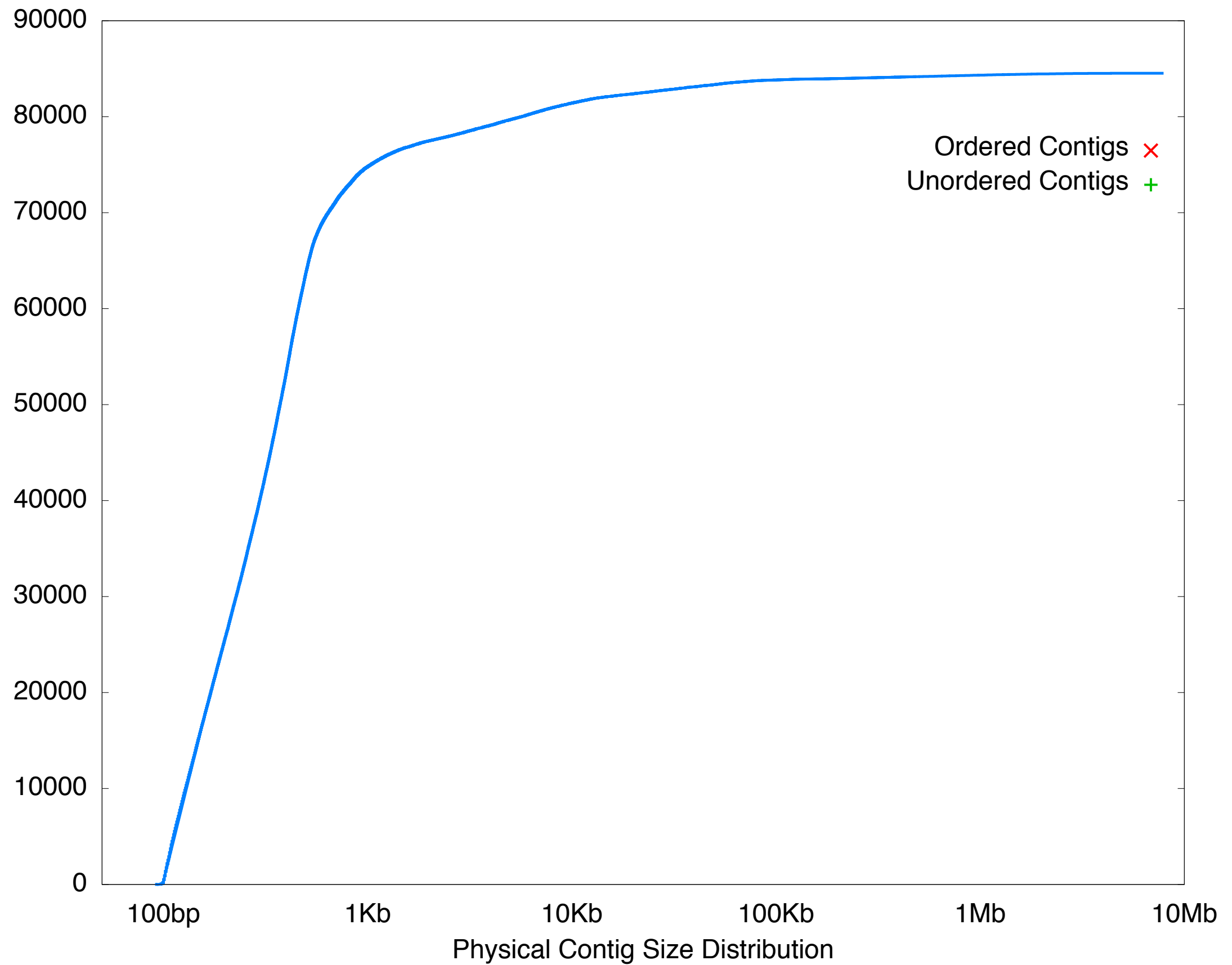
Version 2.0

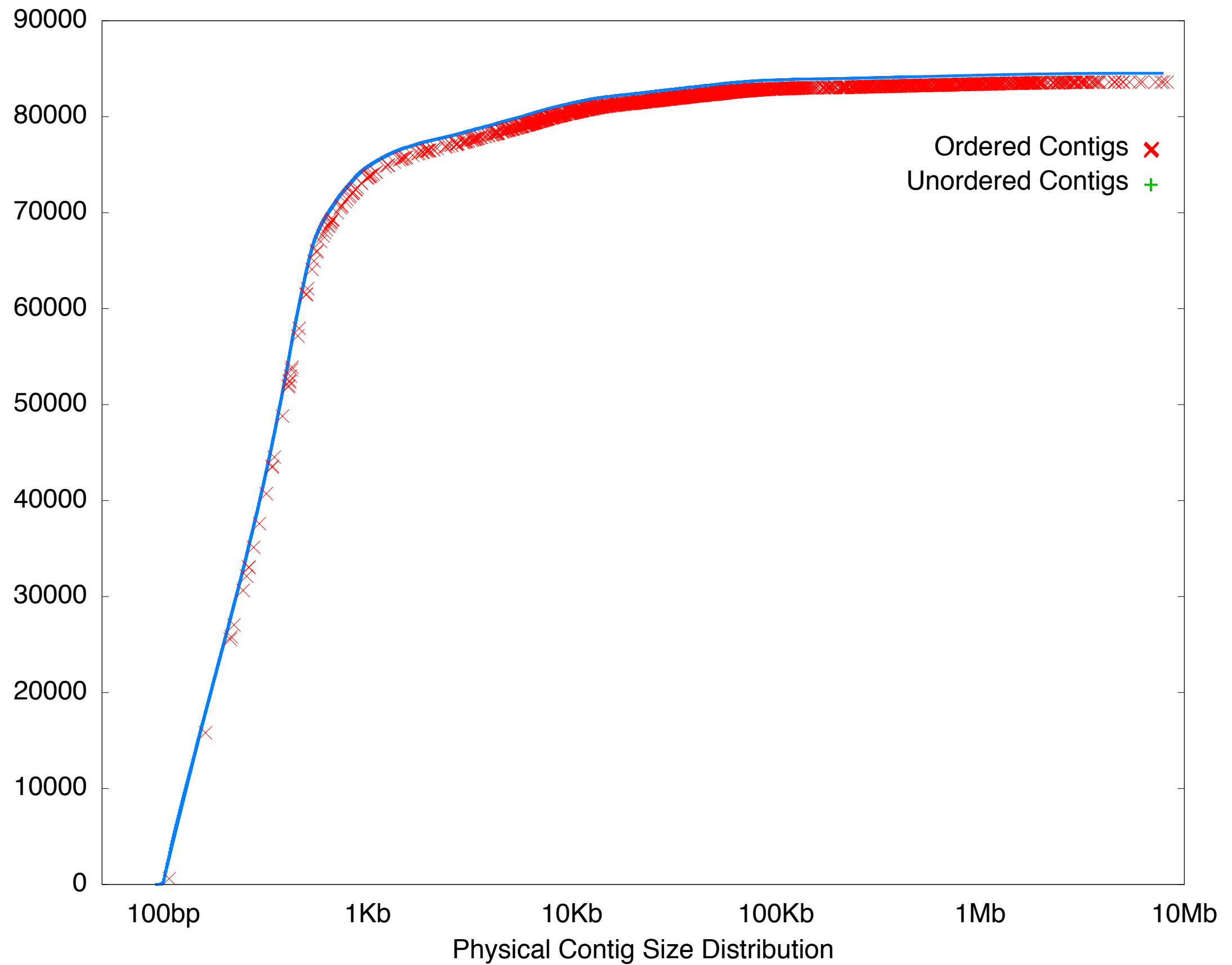
269 total fish

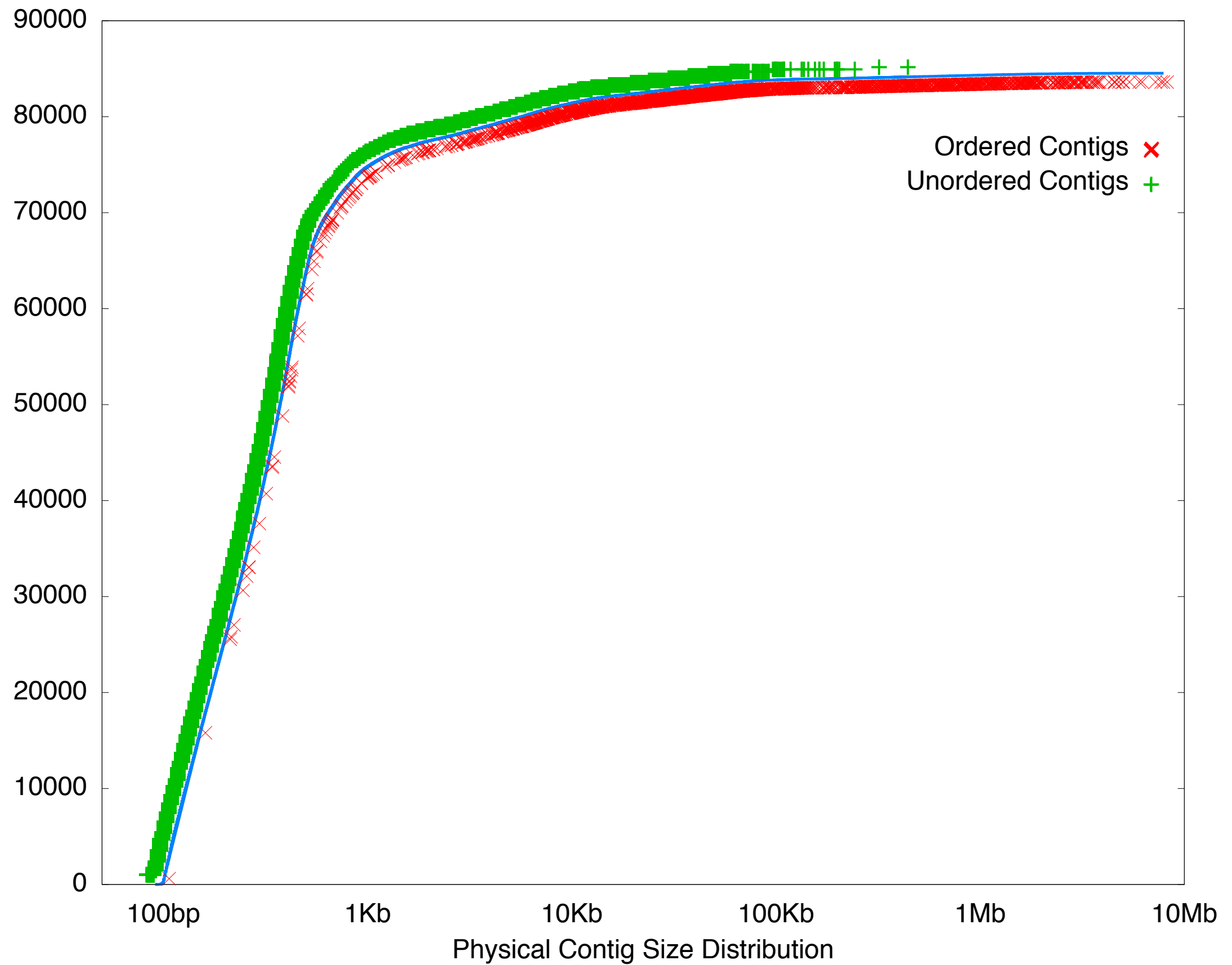
828,373,617 raw reads

613,162,521 incorporated reads

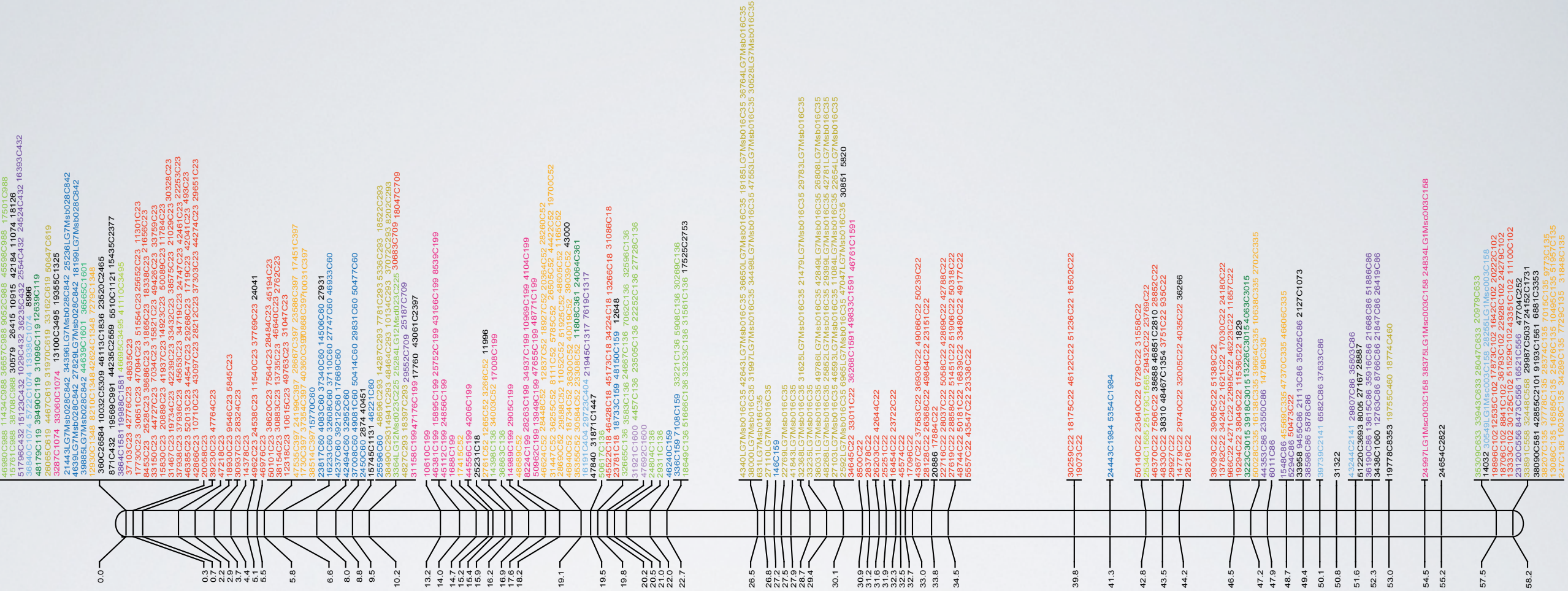
22,144 markers in at least 90 progeny







LG7



Given 80,000 contigs: 86% of genome ordered

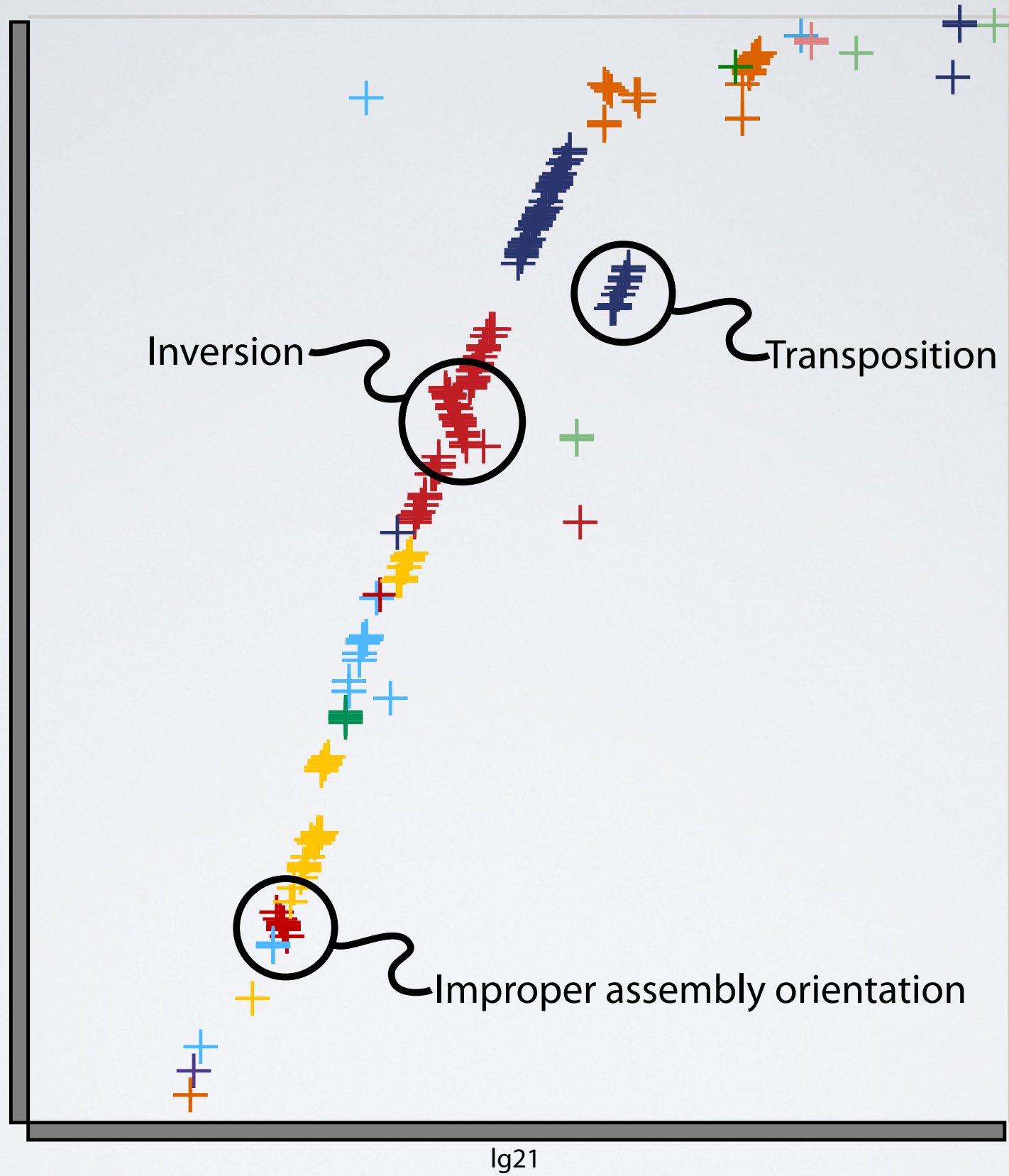


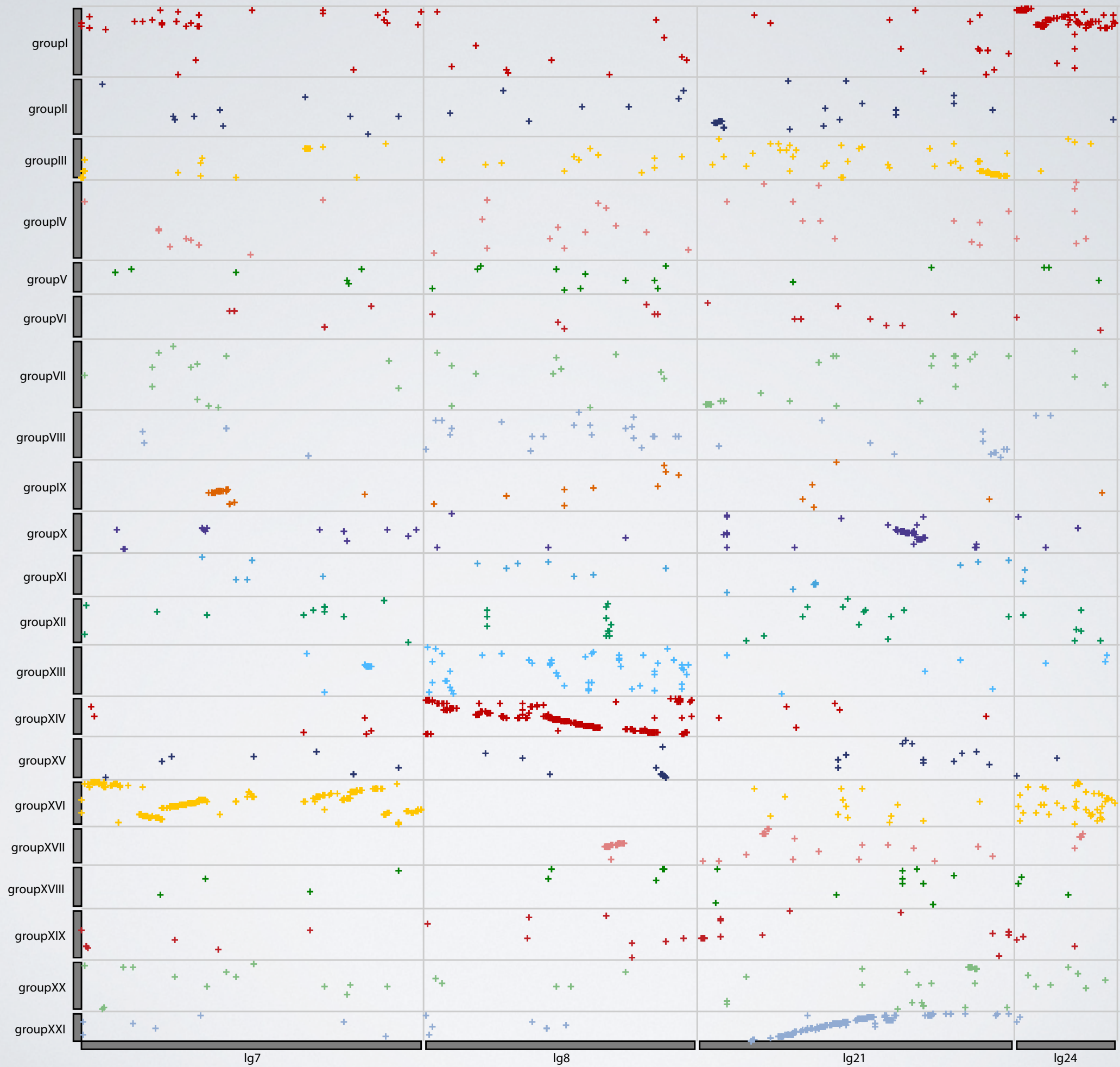
groupXXI

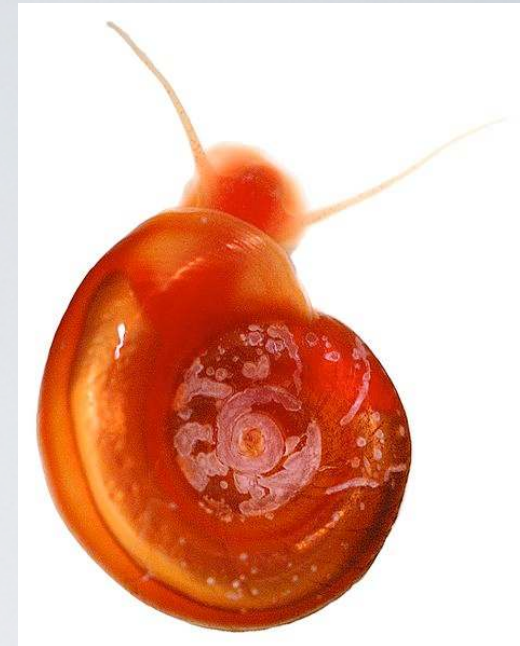
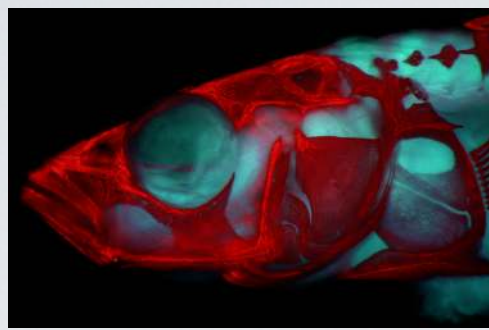




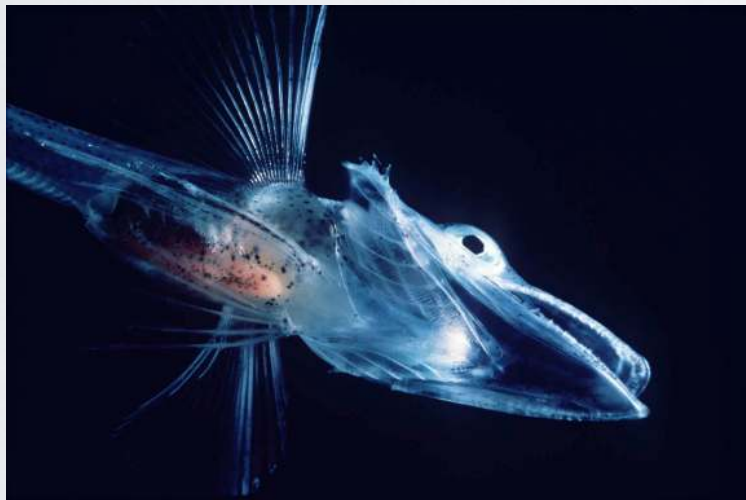
groupXXI

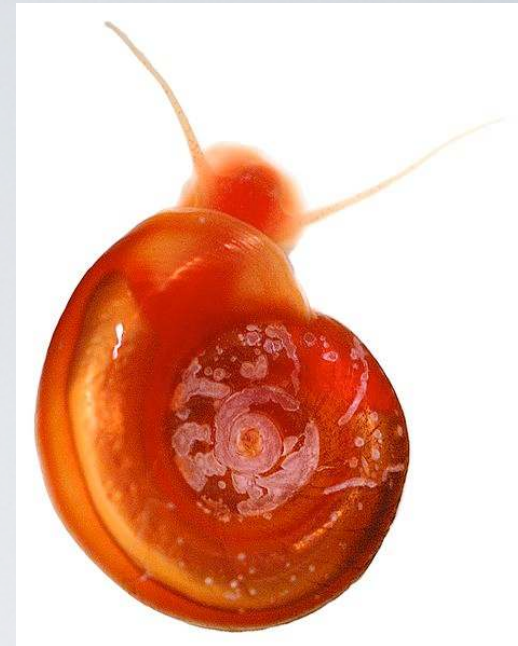
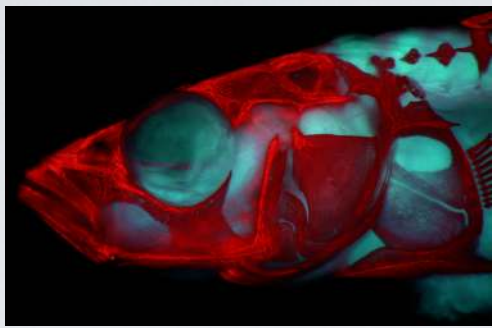




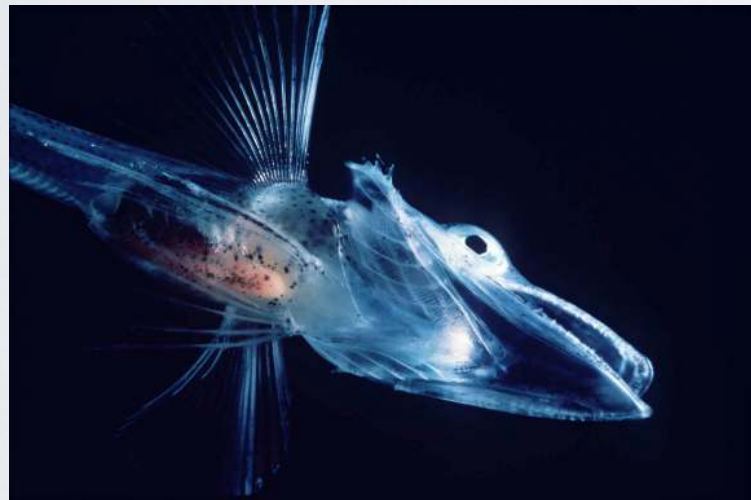


What if you **don't** have a reference genome?

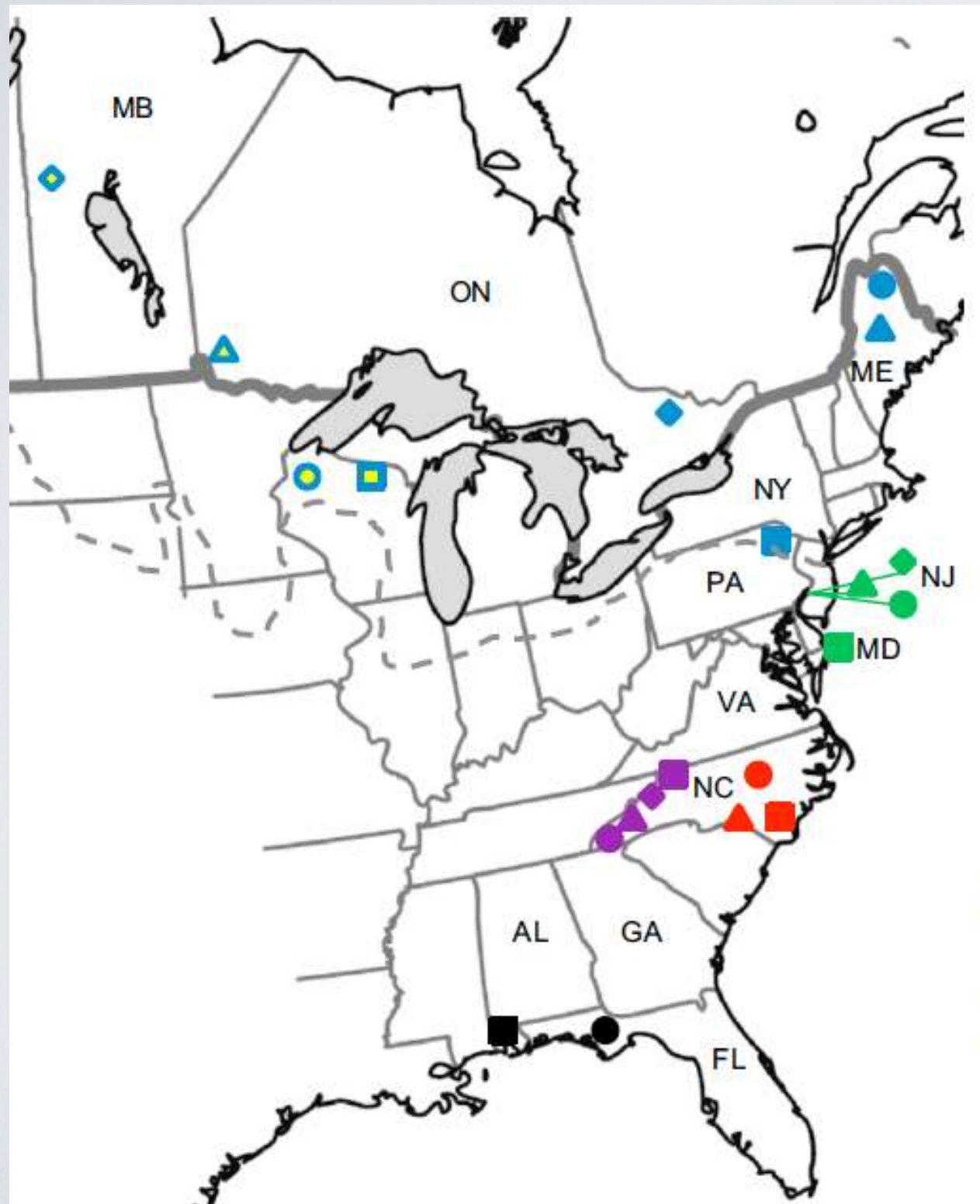




Phylogeography



The pitcher plant mosquito, *Wyeomyia smithii* a model for evolution due to climate change



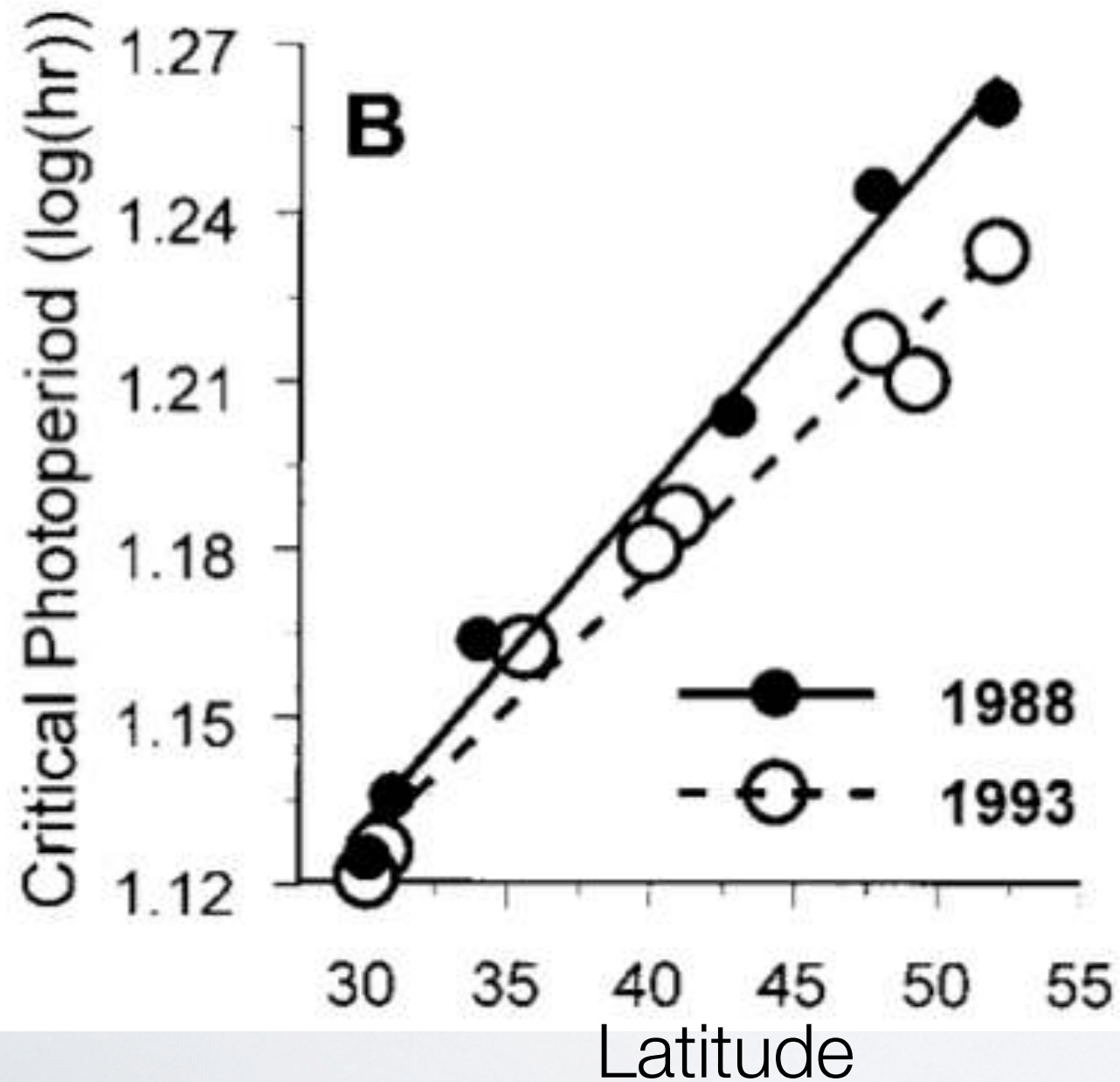
Paul
Hohenlohe



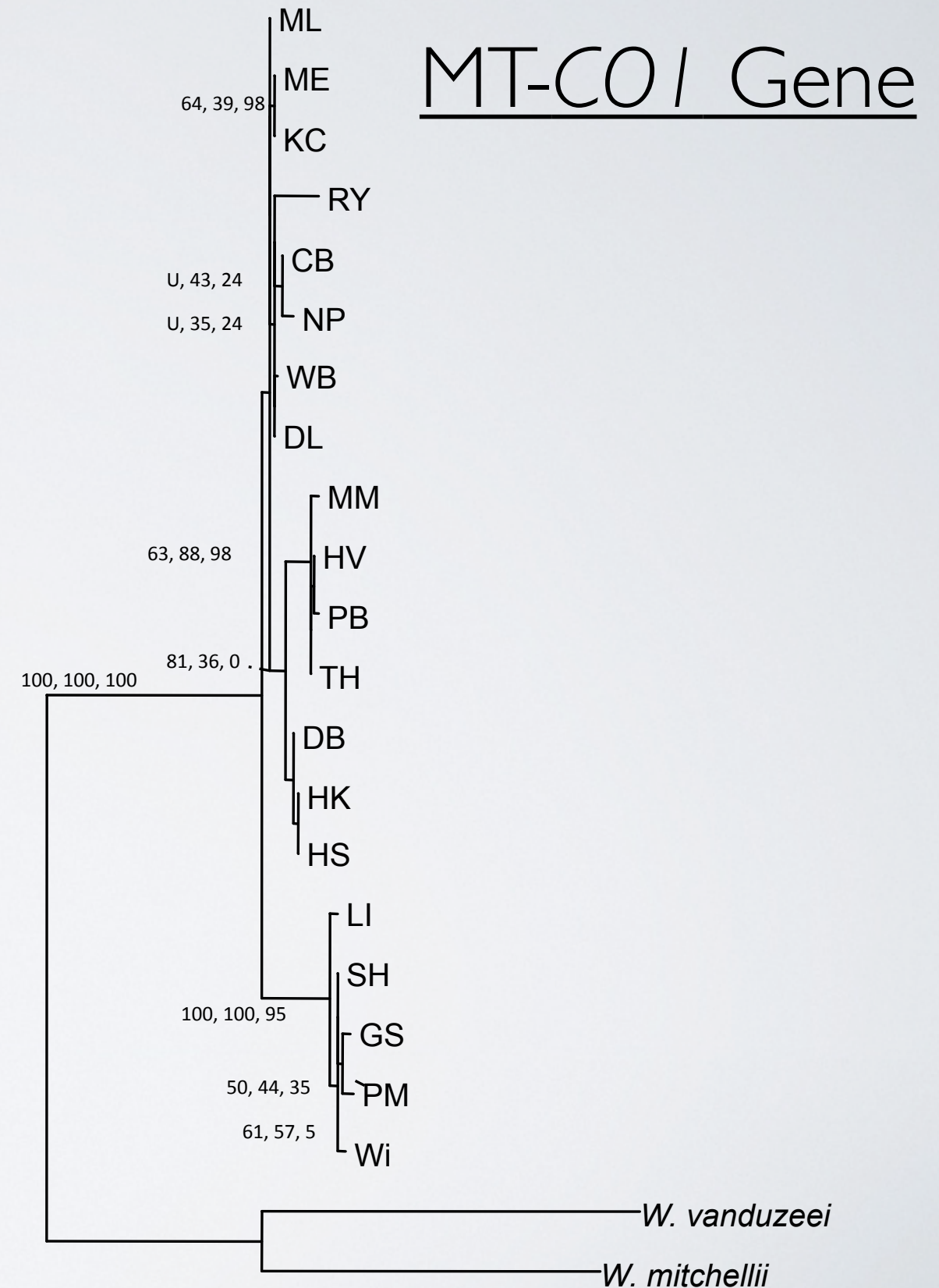
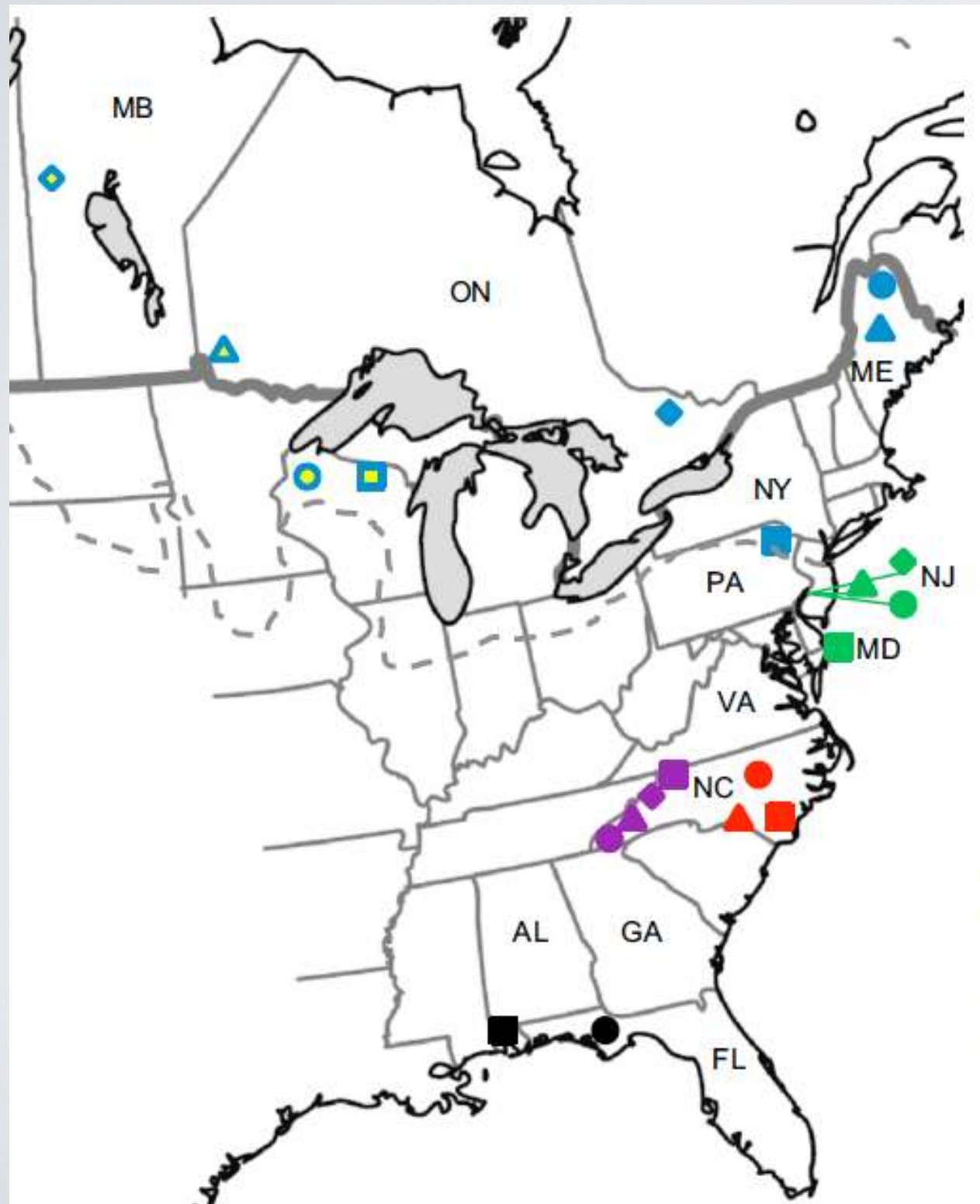
Kevin
Emerson

The pitcher plant mosquito, *Wyeomyia smithii* a model for evolution due to climate change

Bradshaw et al (2001) PNAS

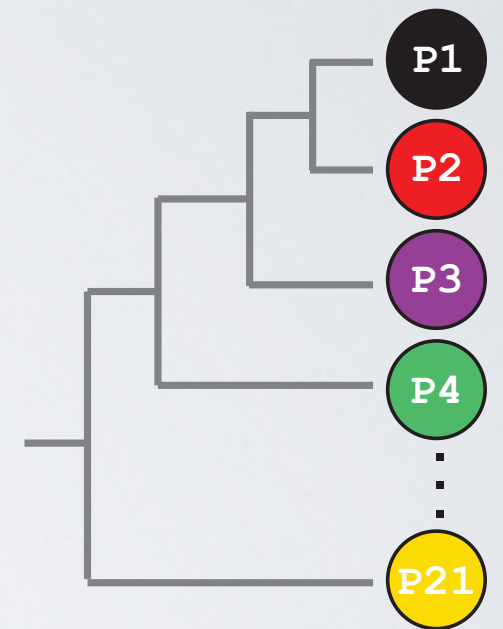


Resolving post-glacial phylogeography in *Wyeomyia smithii*

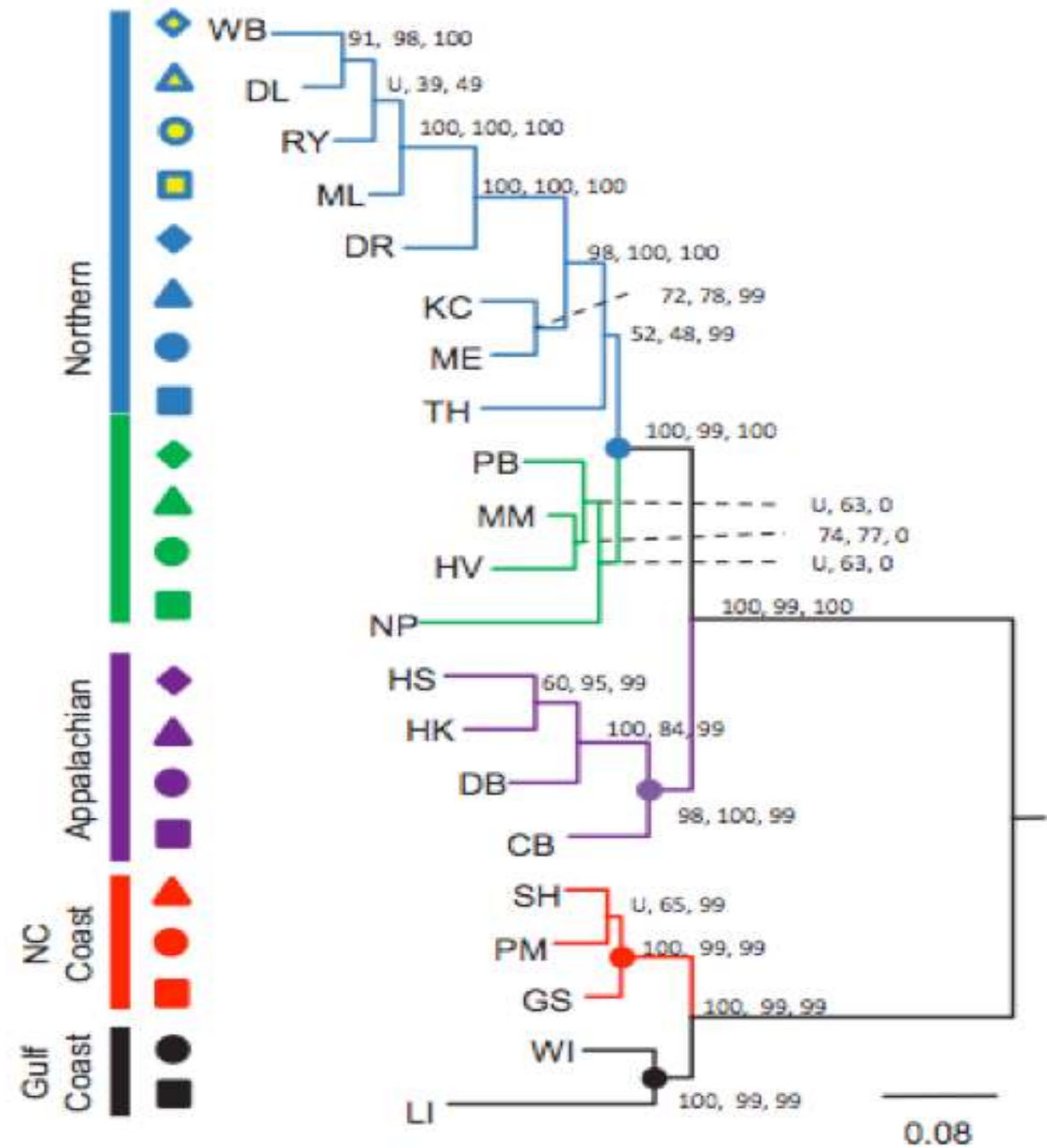
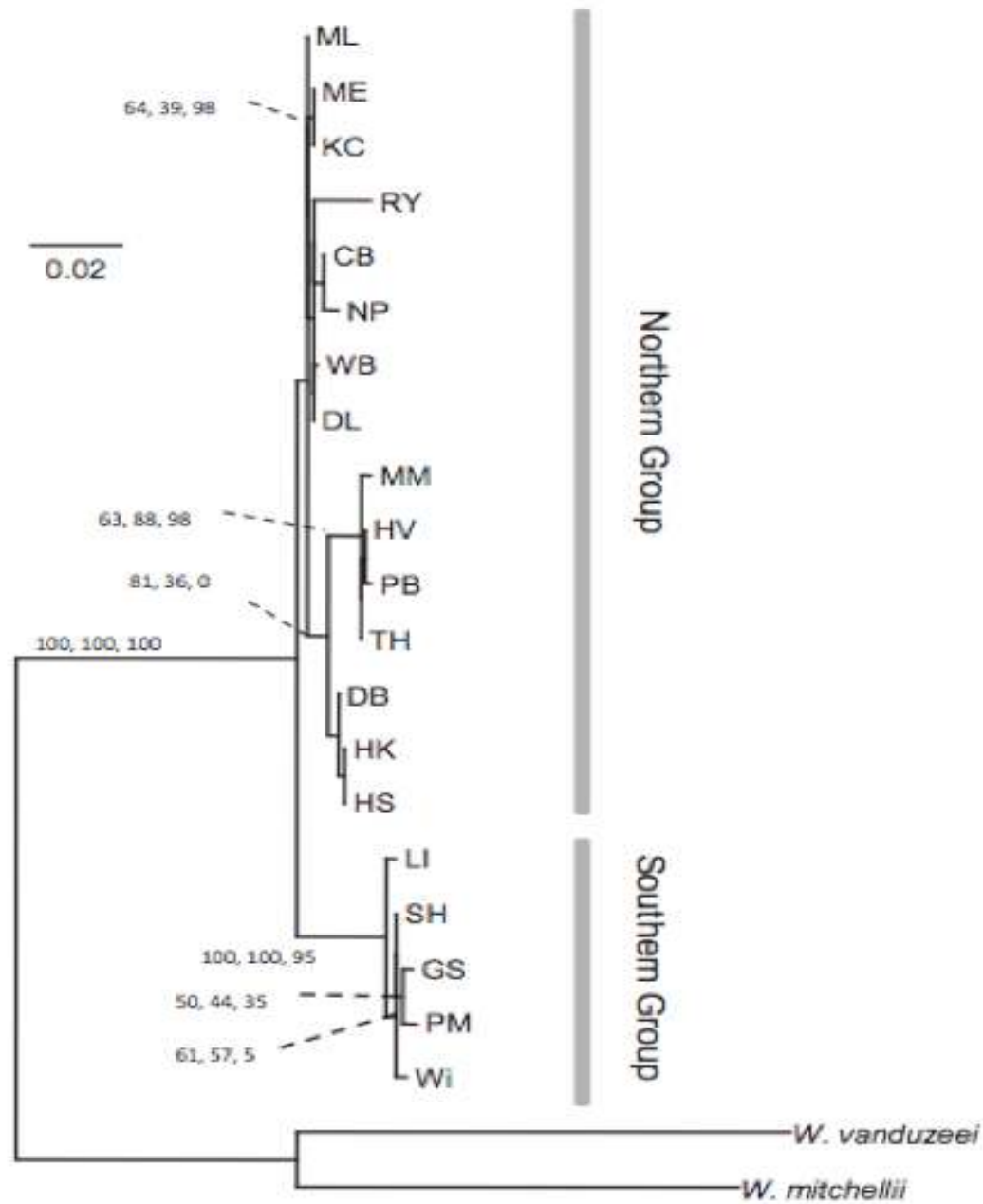


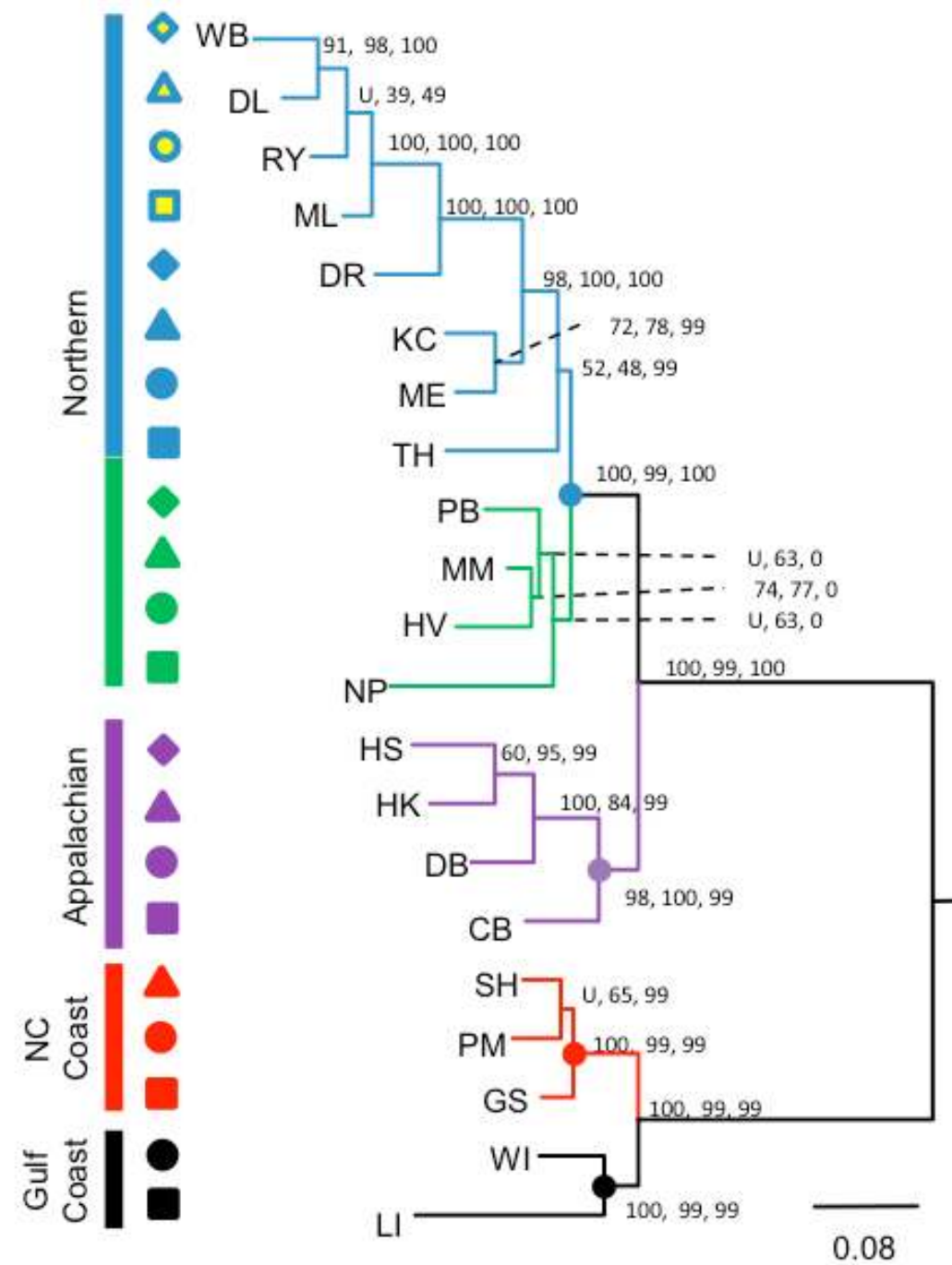
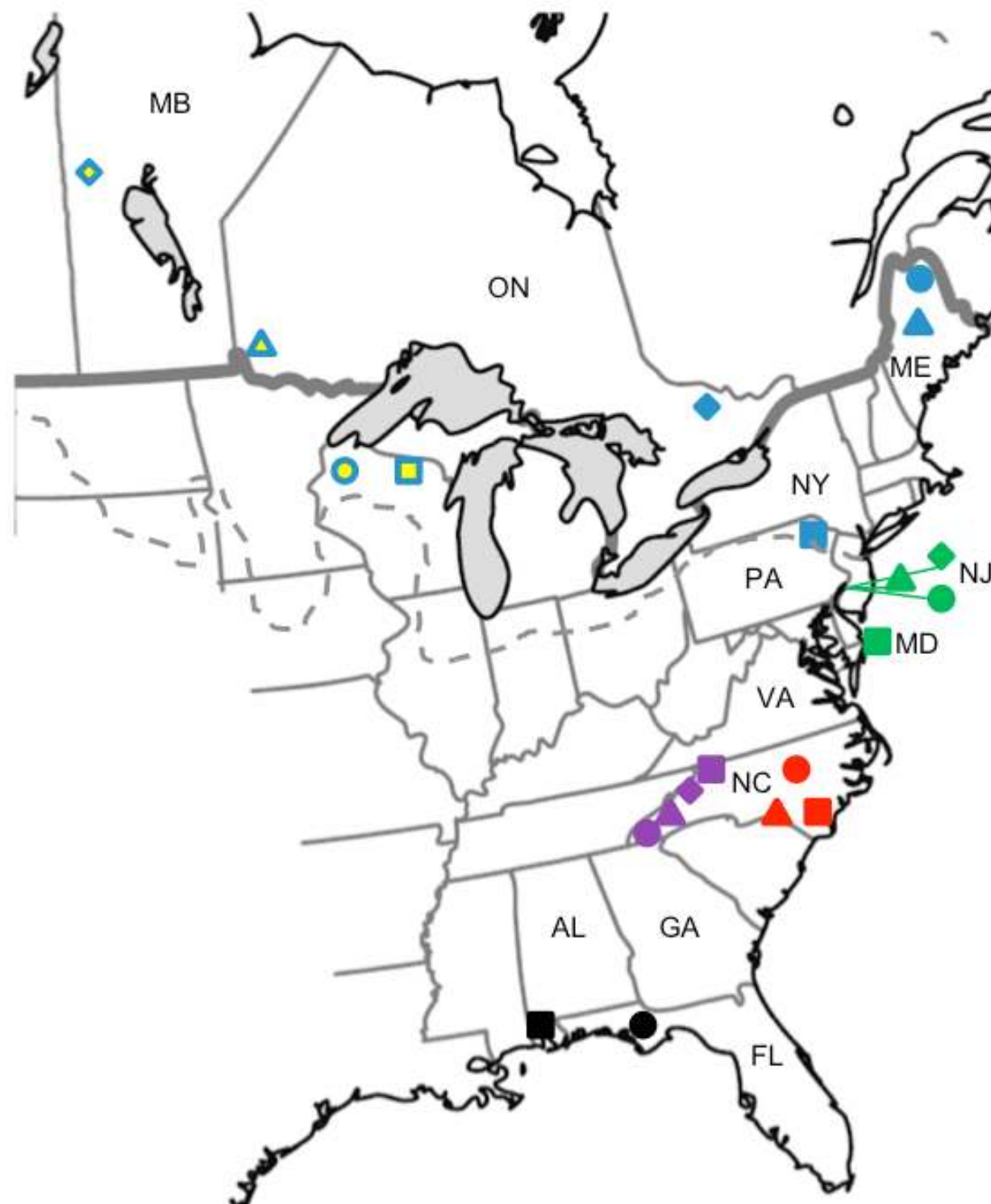
Sampling RAD-Tag markers for population-level phylogenetic analysis

Sampling RAD-Tag markers for population-level phylogenetic analysis



Genome-wide view improves resolution





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