



Ecological & evolutionary genomic analyses in non-model organisms using RAD-seq

2014 Workshop on Genomics

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Department of Biology

University of Oregon



Outline for today's lecture

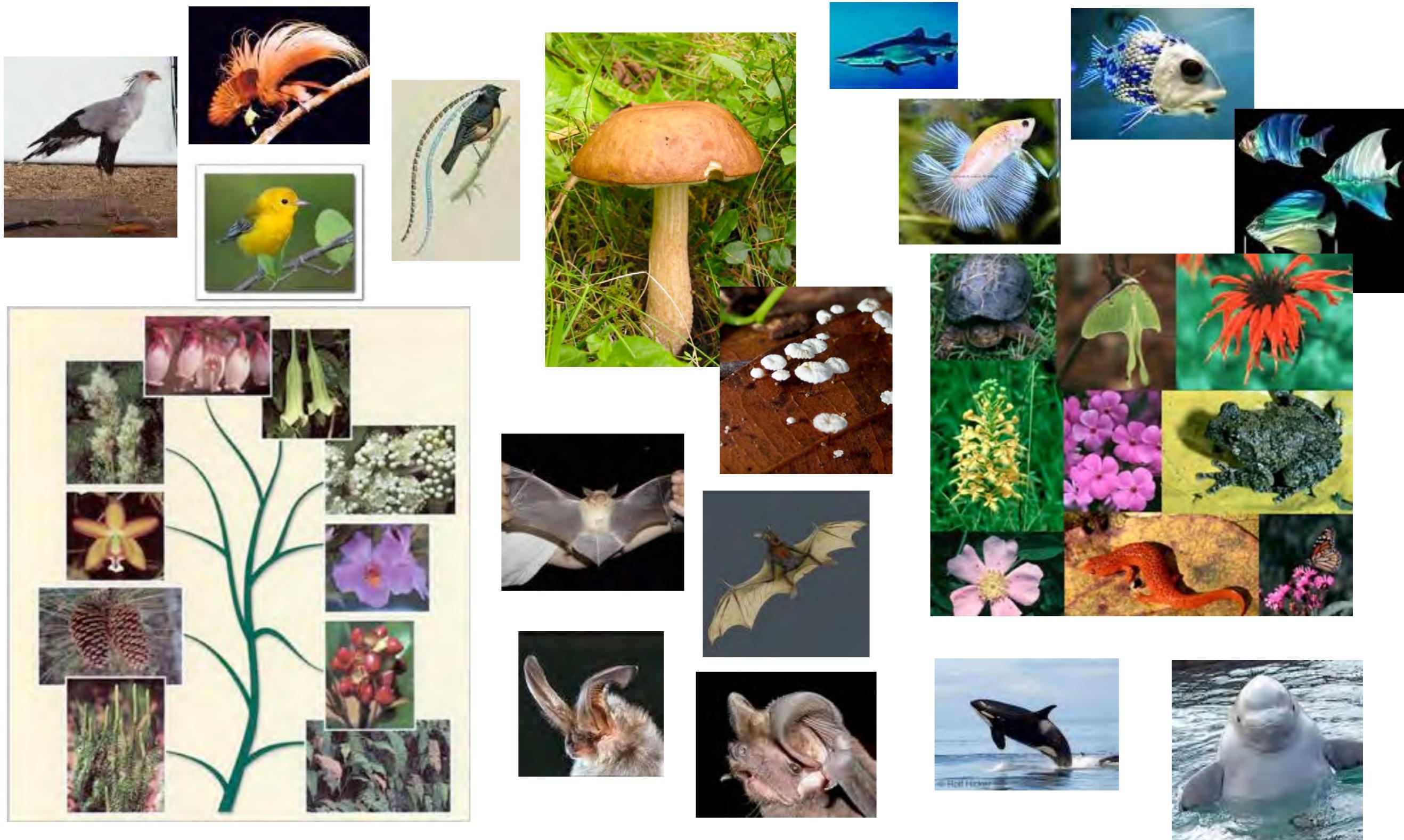
Genomic data and (non-)model organism research

RAD-seq for ecological & evolutionary genomics

Genomically enabling a non-model organism

Stacks software pipeline

Why do organisms look the way that they do?



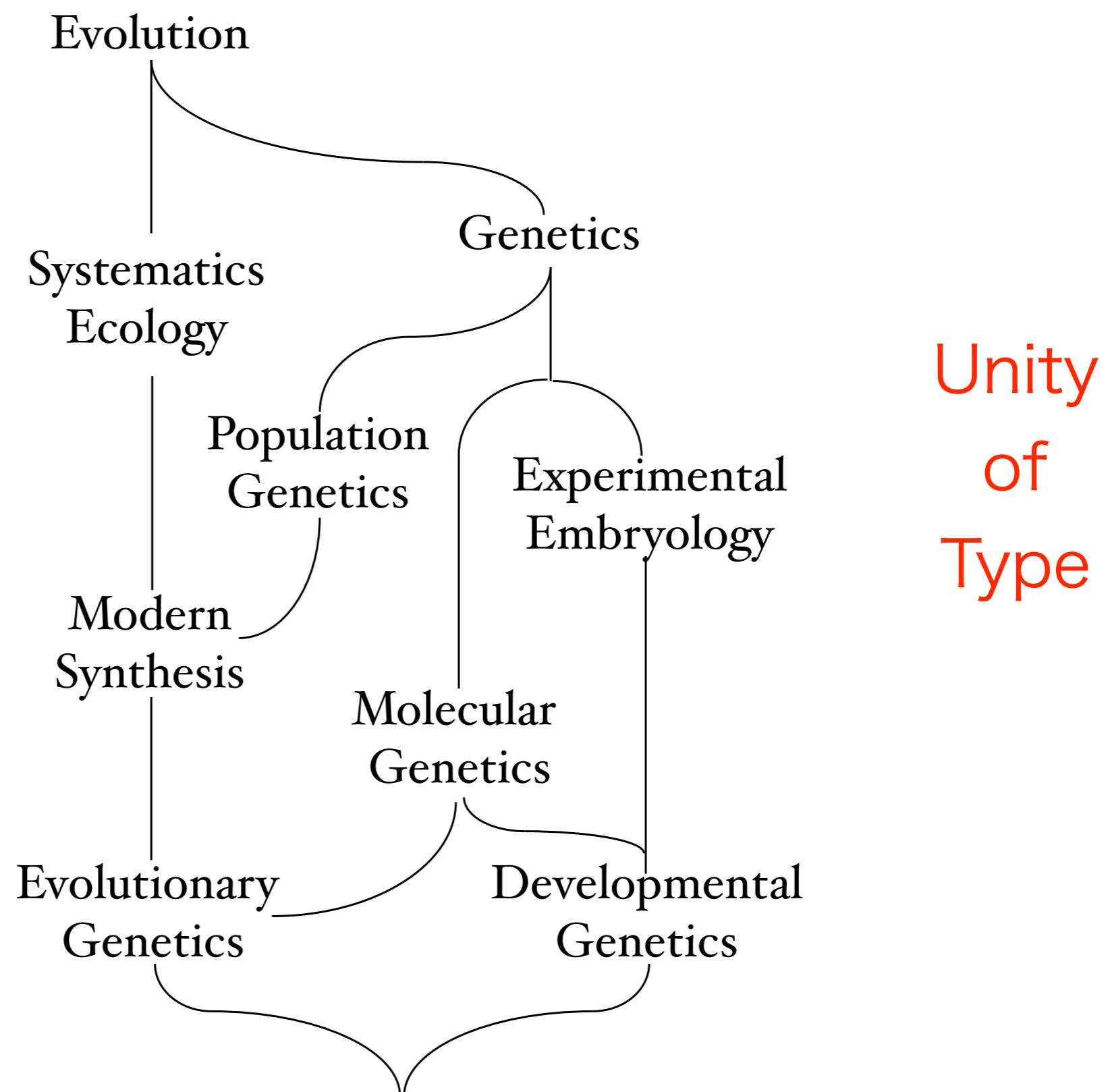
1850

1900

Conditions
of
Existence

1950

2000



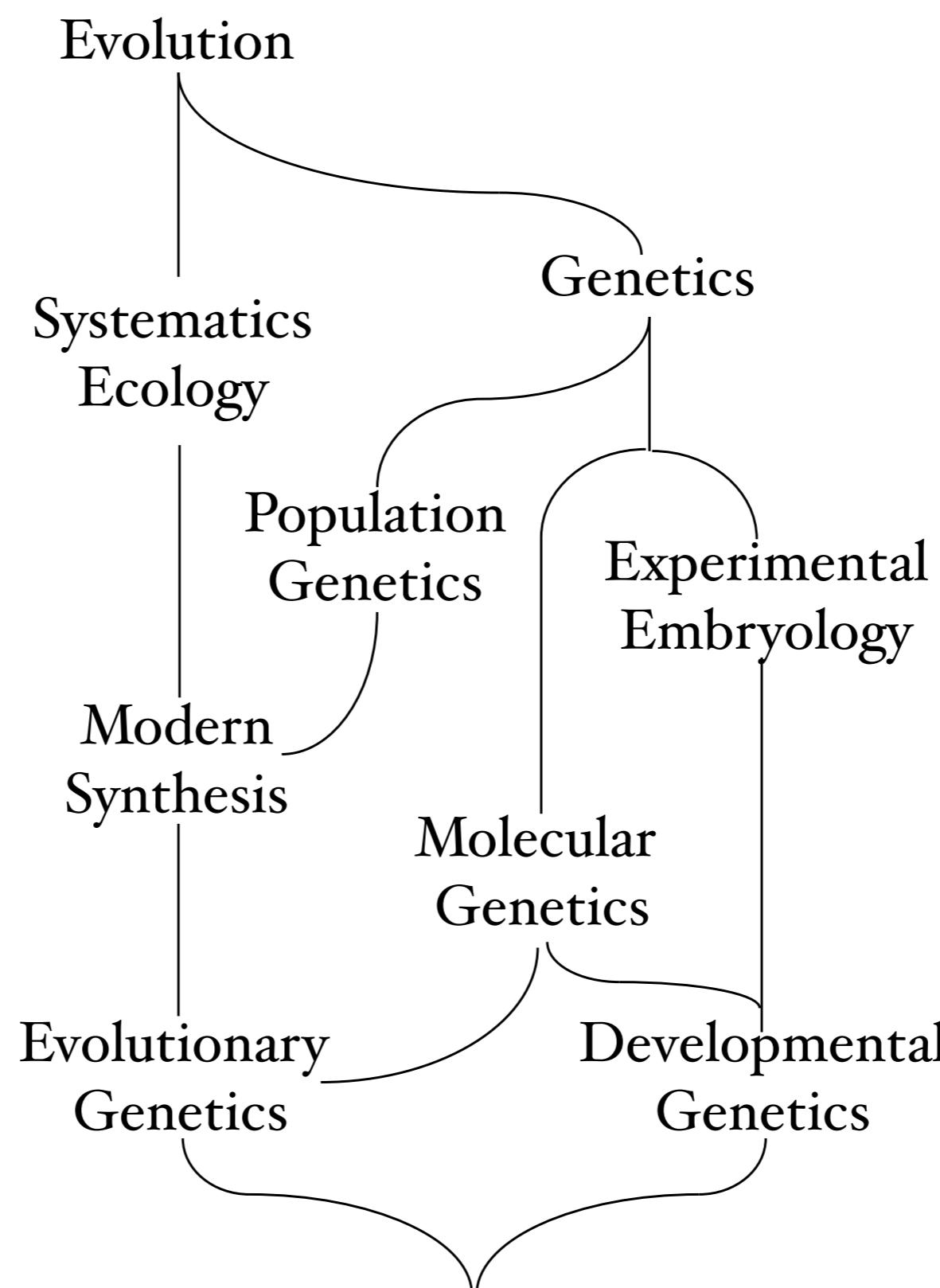
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Model organism research has been very important

Vertebrate **zygotes** or embryos



28 day human



19h zebrafish

Model organism research has been very important

Vertebrate **zygotes** or embryos



28 day human



19h zebrafish

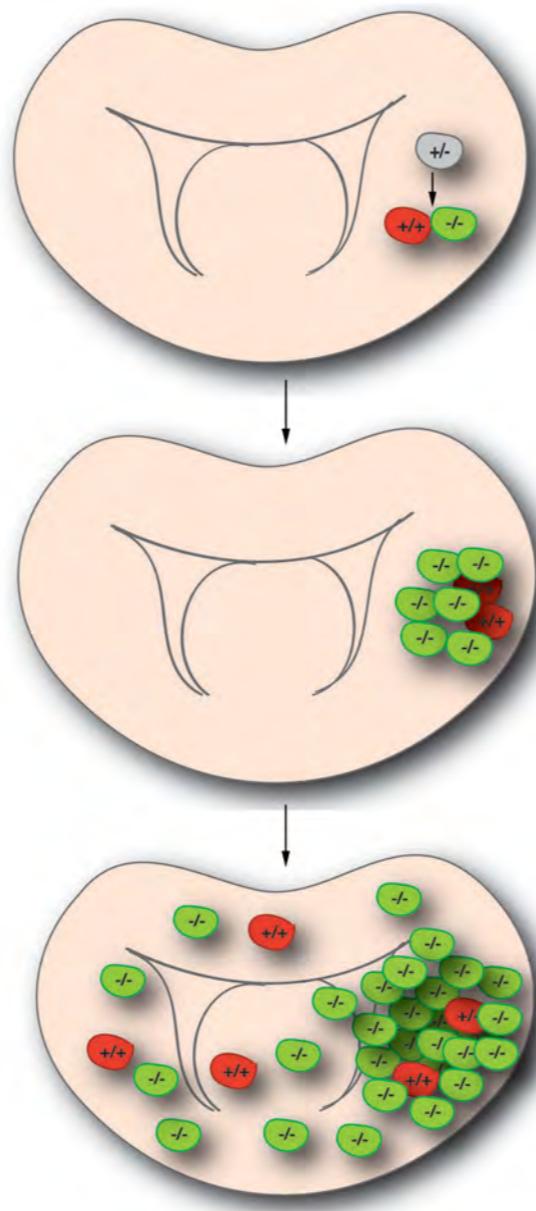


Dr. Catchen in his 'following Phish Phase'

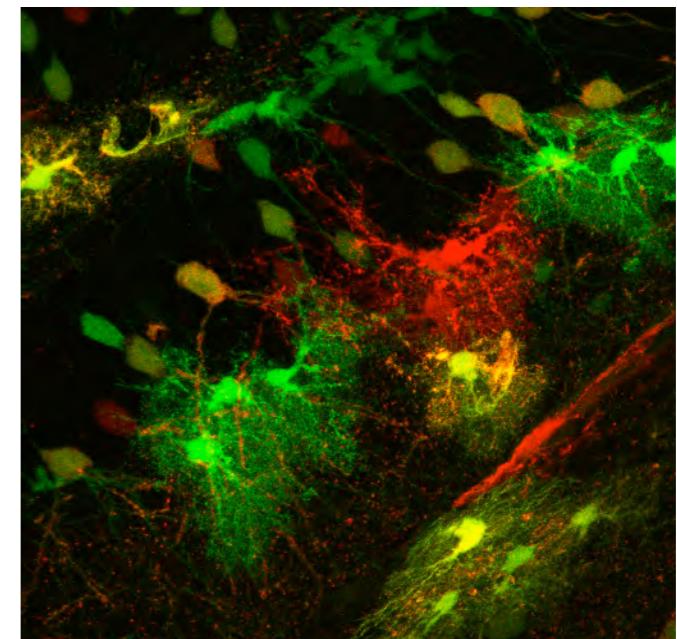


Studying brain cancer using somatic evolutionary genomics in a model organism

pre-cancerous



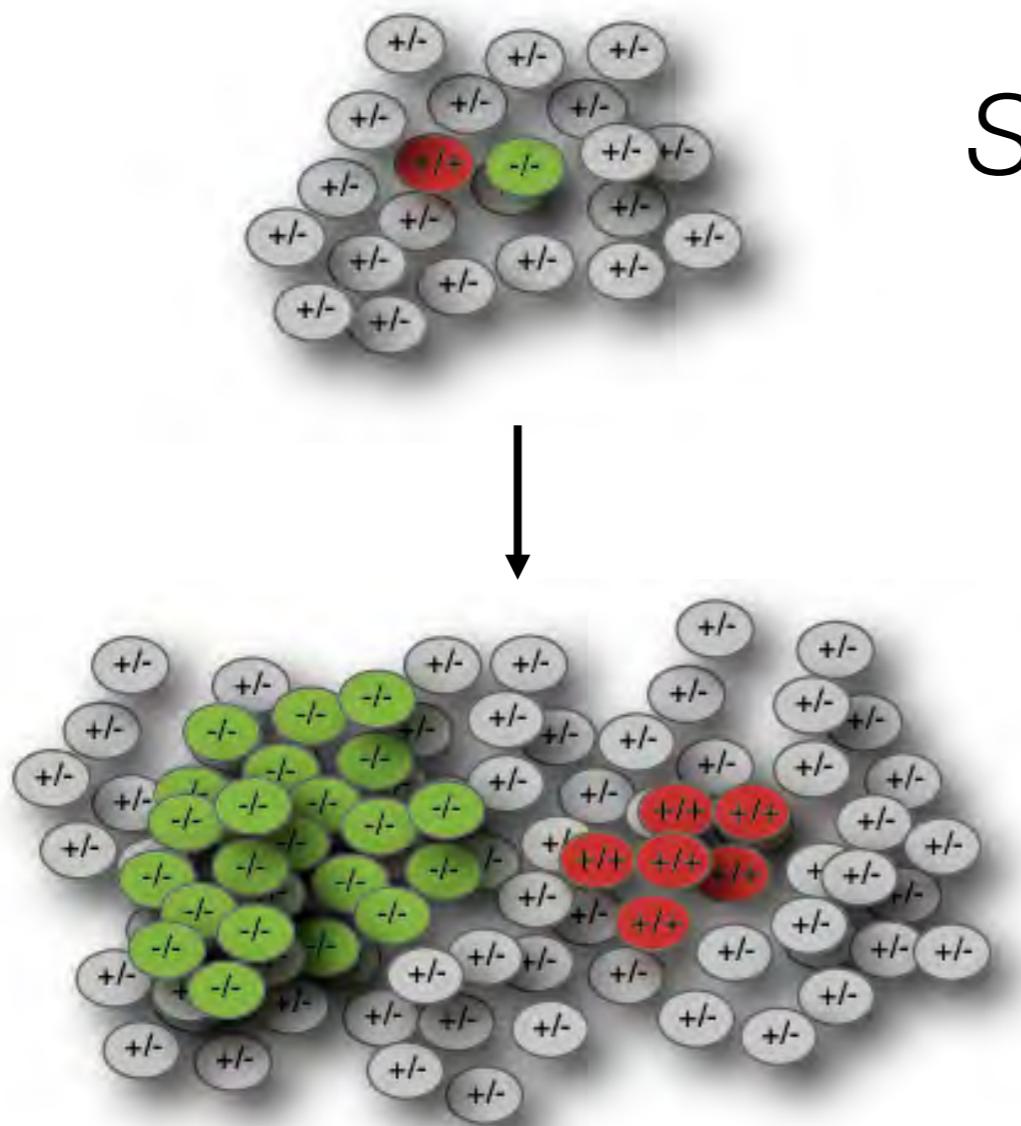
tumor



Laser Capture Microdissection of cells



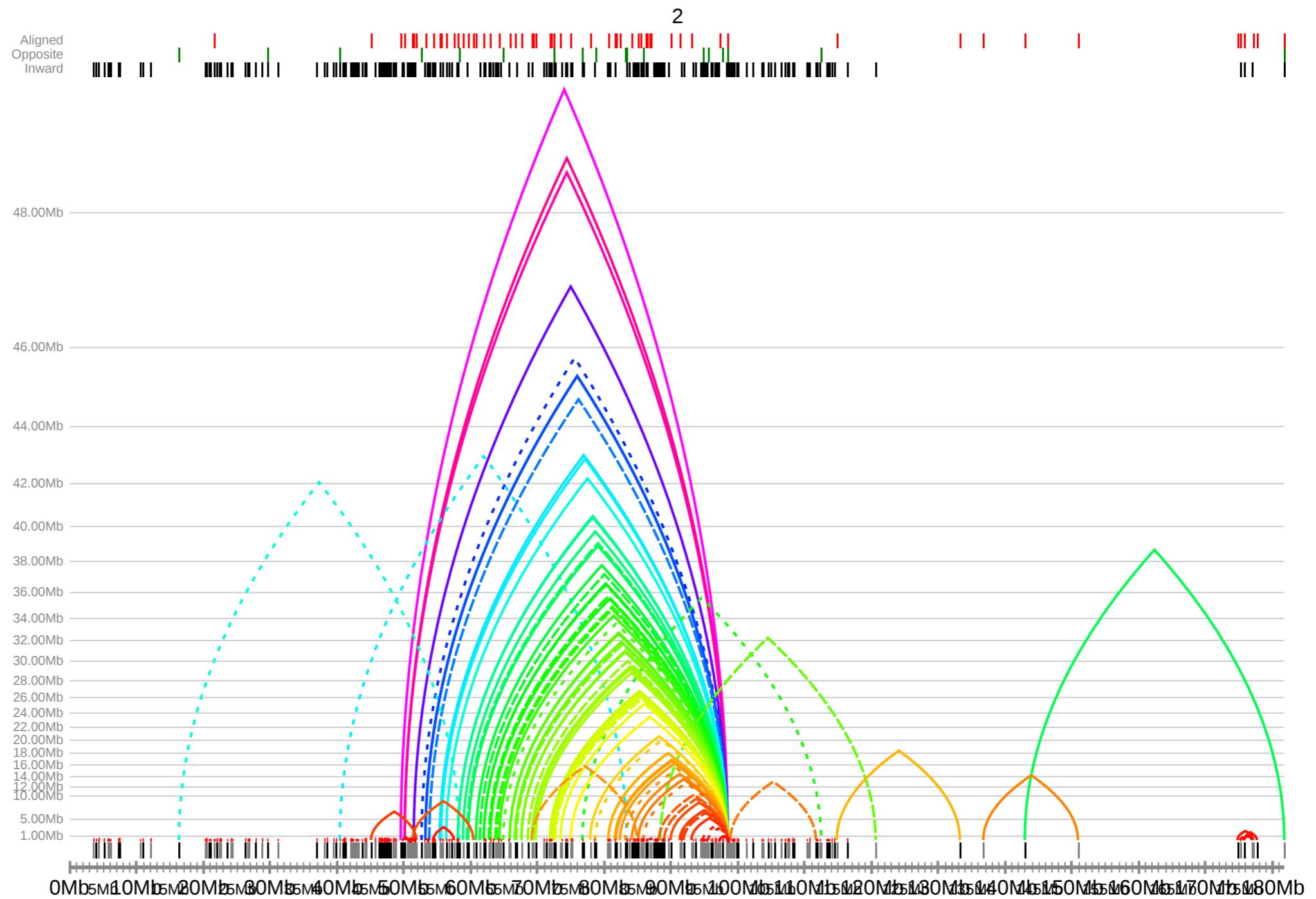
Transcriptomic and genomic analysis of cells



Sequence cells here...

... and here

Genomic rearrangements in cancer cells

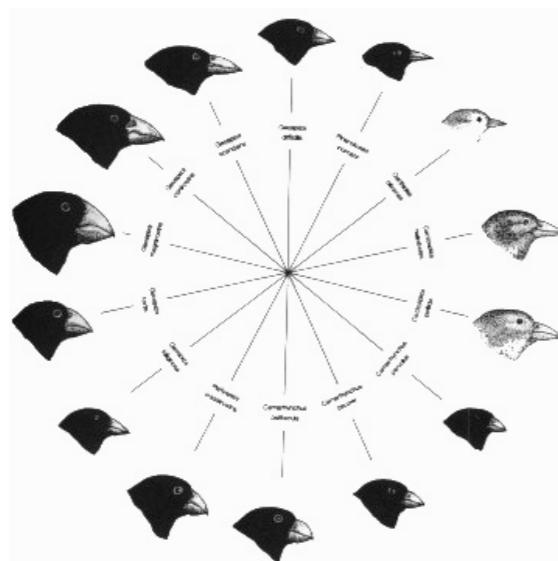


Julian Catchen

1850

1900

Conditions
of
Existence



1950

2000

Evolution

Systematics

Ecology

Modern
Synthesis

Evolutionary
Genetics

Population
Genetics

Genetics

Experimental
Embryology

Molecular
Genetics

Developmental
Genetics

Unity
of
Type



functional evolutionary genomics

How do organisms adapt to novel environments?



from Grant and Grant. 2007. How and why species multiply: The radiation of Darwin's finches. Princeton University Press

How do organisms adapt to novel environments?



How is genetic diversity partitioned across individuals, populations and species?

What genomic regions are important for adaptation to novel environments?

How does the ecology of organisms structure genomic architectures?

How does genome architecture influence rapid evolution?

Where does the basis for evolutionary novelties reside in genomes?



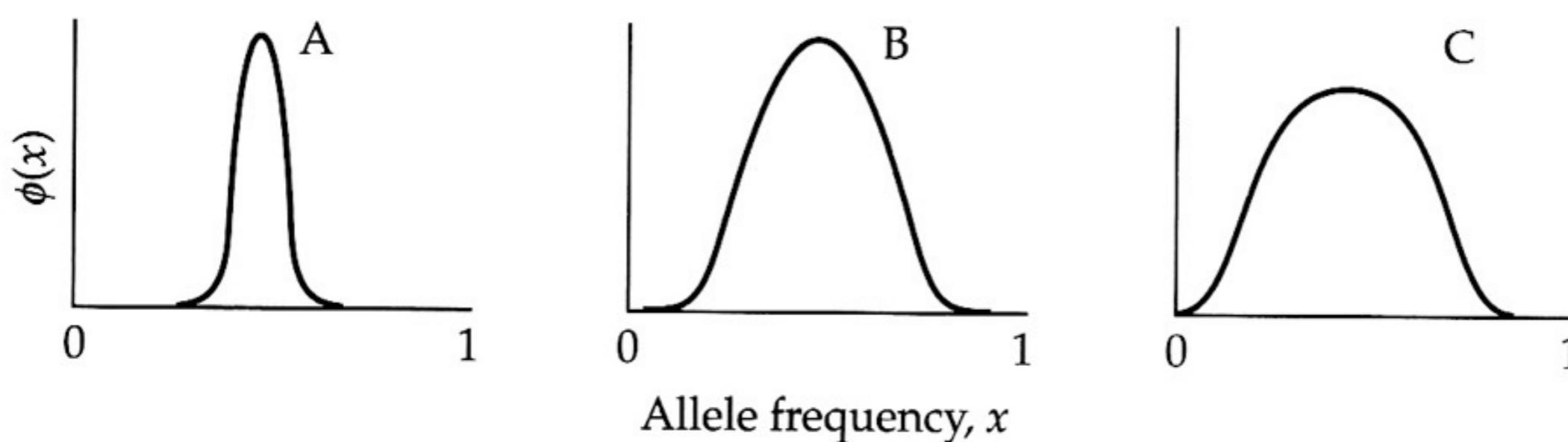
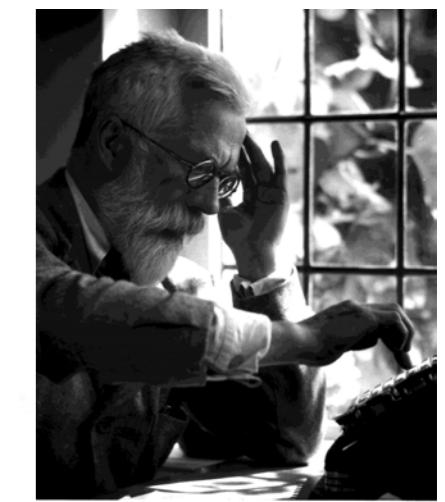
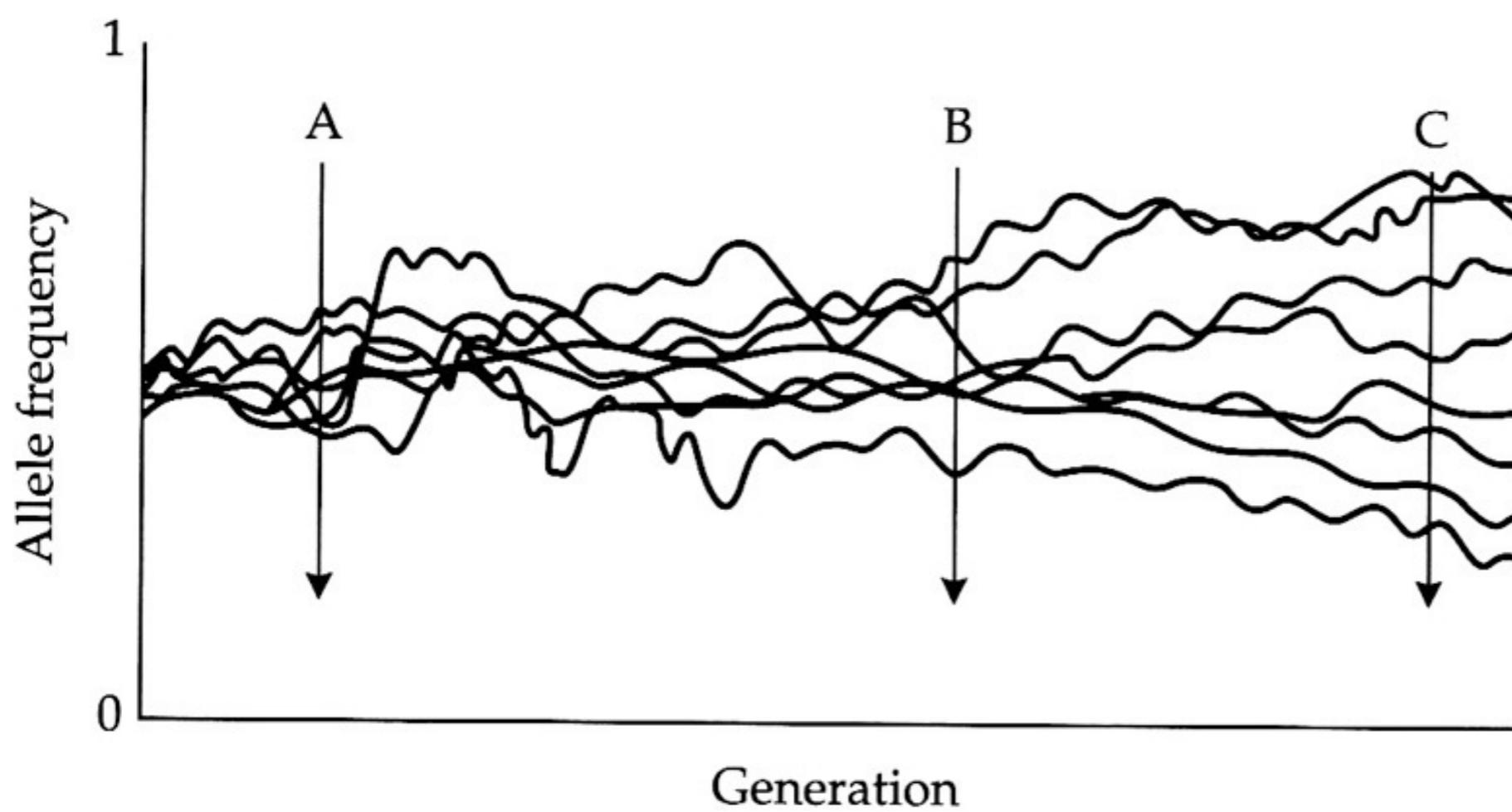
from Grant and Grant. 2007. How and why species multiply: The radiation of Darwin's finches. Princeton University Press

Four fundamental processes in evolution

Origin of genetic variation;
mutation
migration

Sorting of variation;
genetic drift
natural selection

Genetic drift is a null model

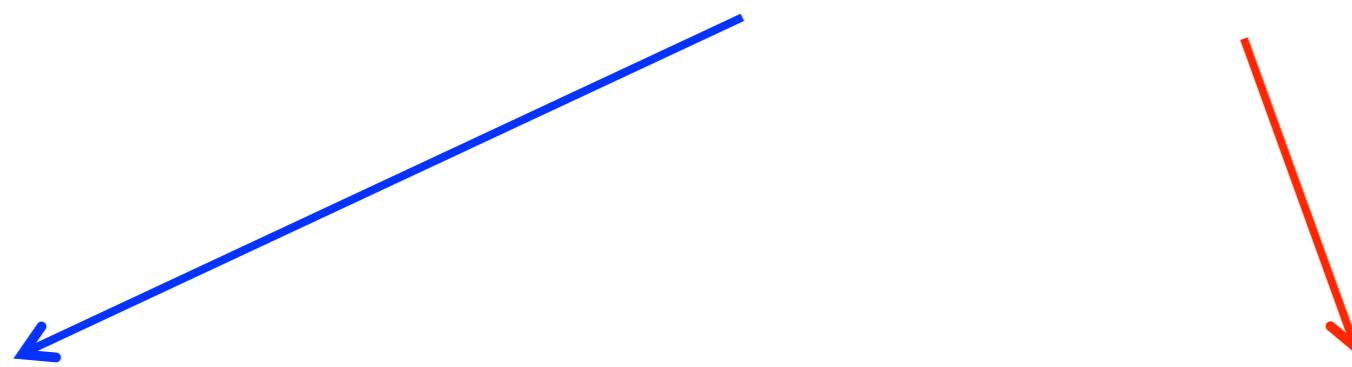


Population genomics

Simultaneous genotyping of **neutral** and **adaptive** loci

Genome-wide background provides more precise estimates:

- Demographic processes (e.g. N_e)
- Phylogeography

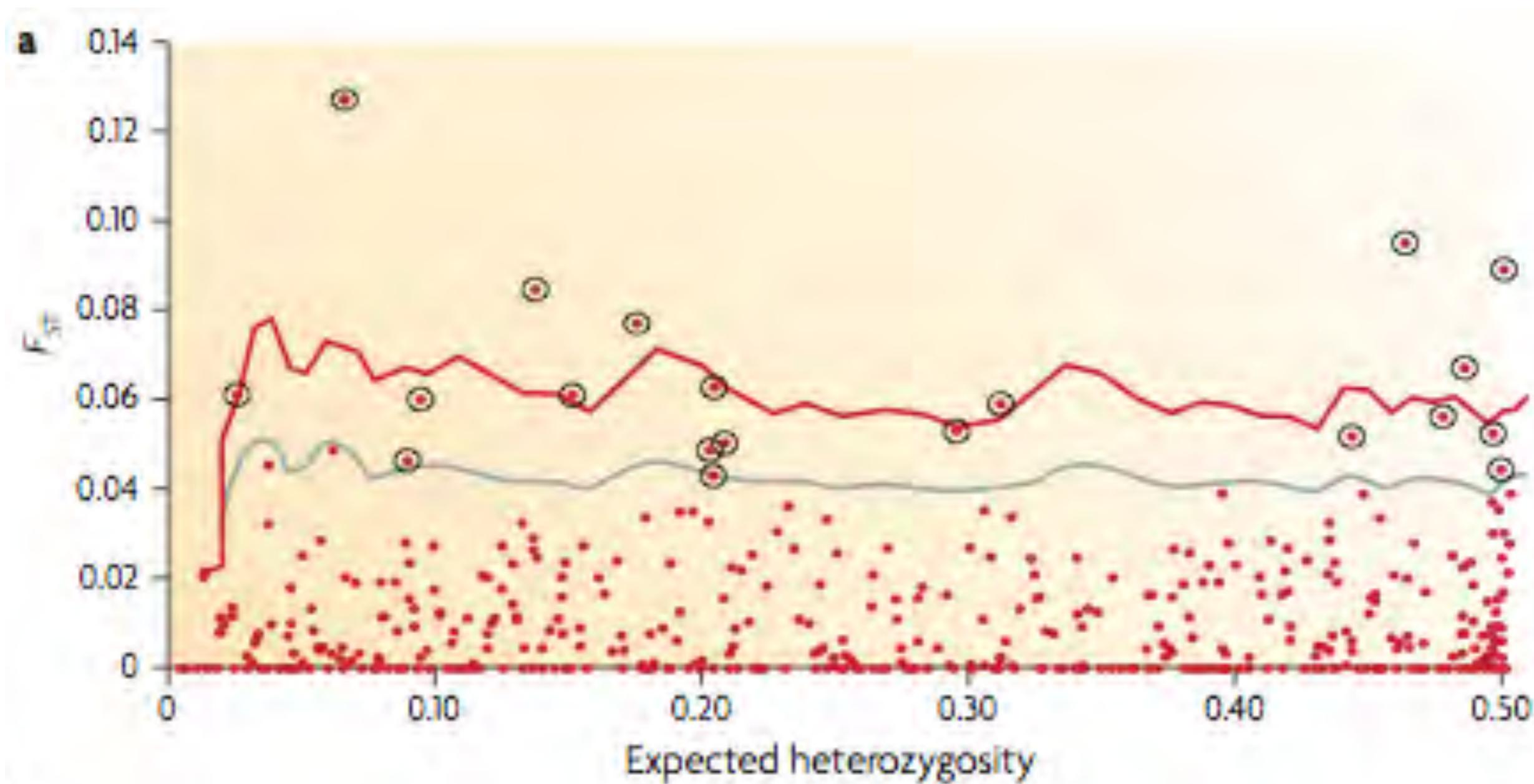


Outliers from background indicate:

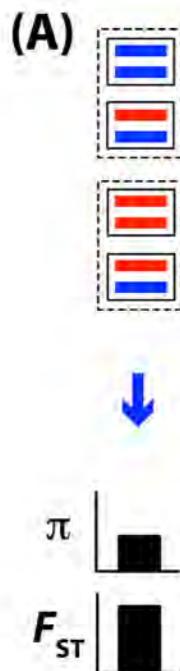
- Selective sweeps
- Local adaptation



Population genomics of unordered markers



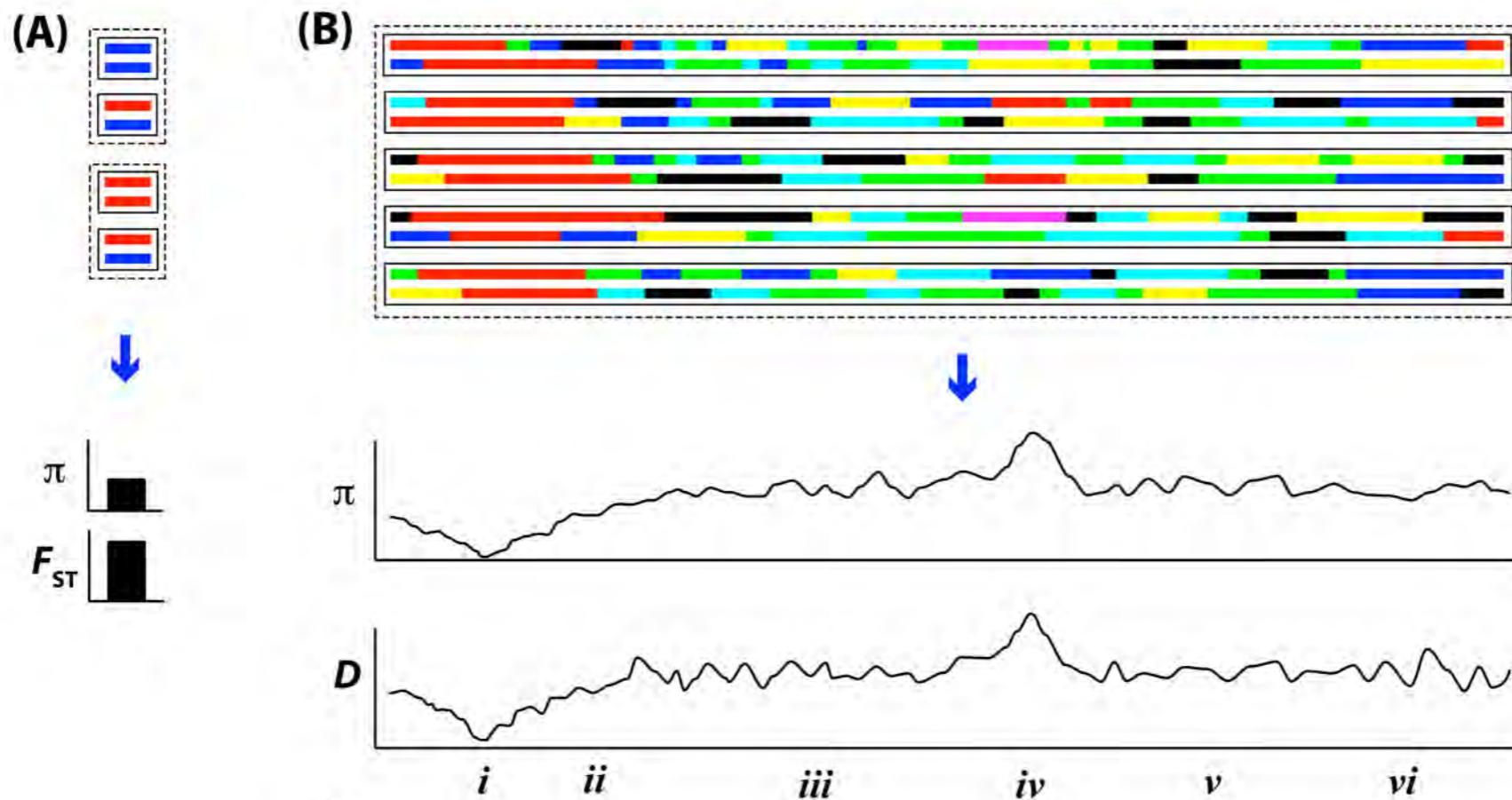
Population Genetics



Population Genetics



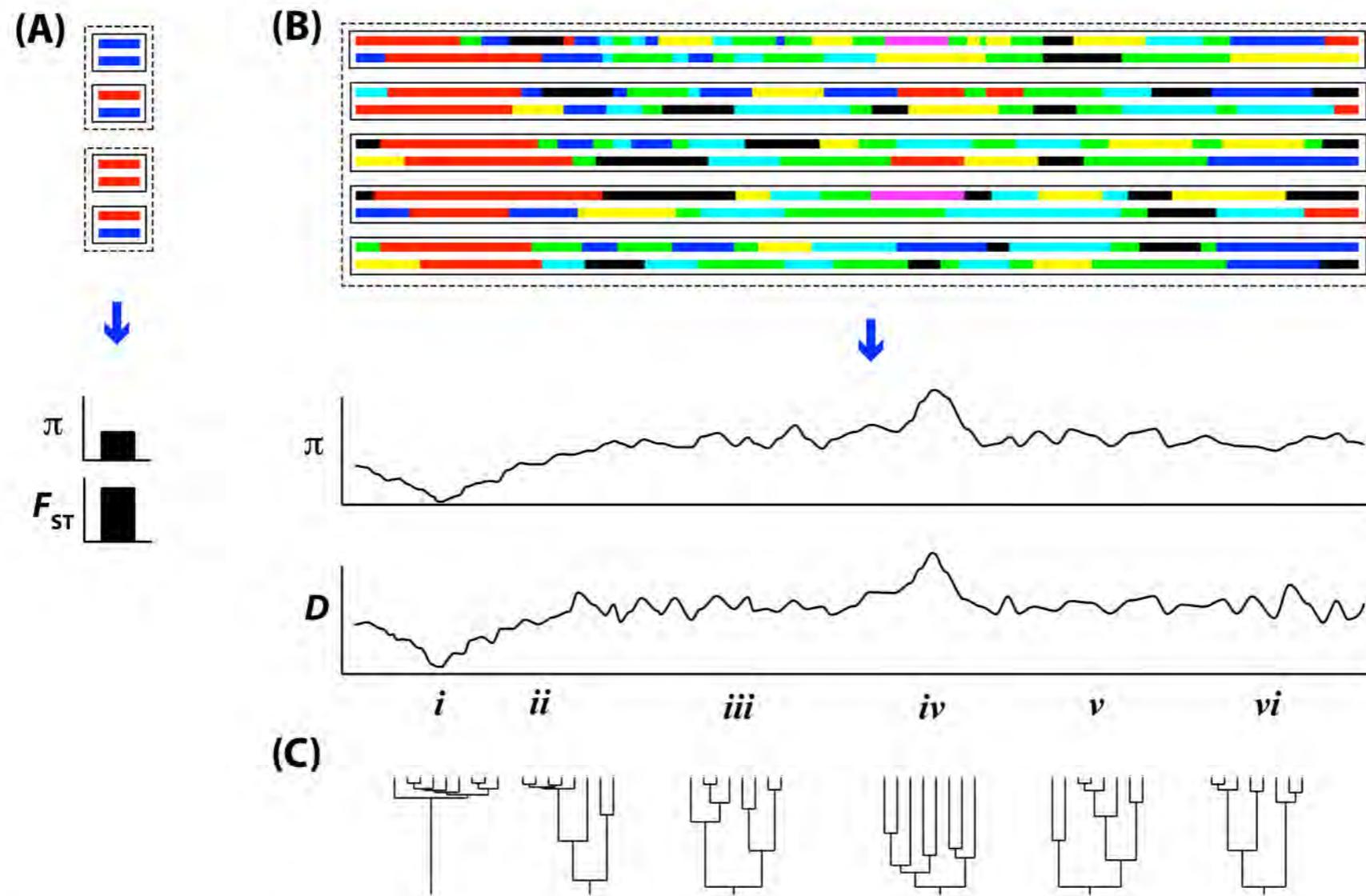
Population Genomics



Population Genetics



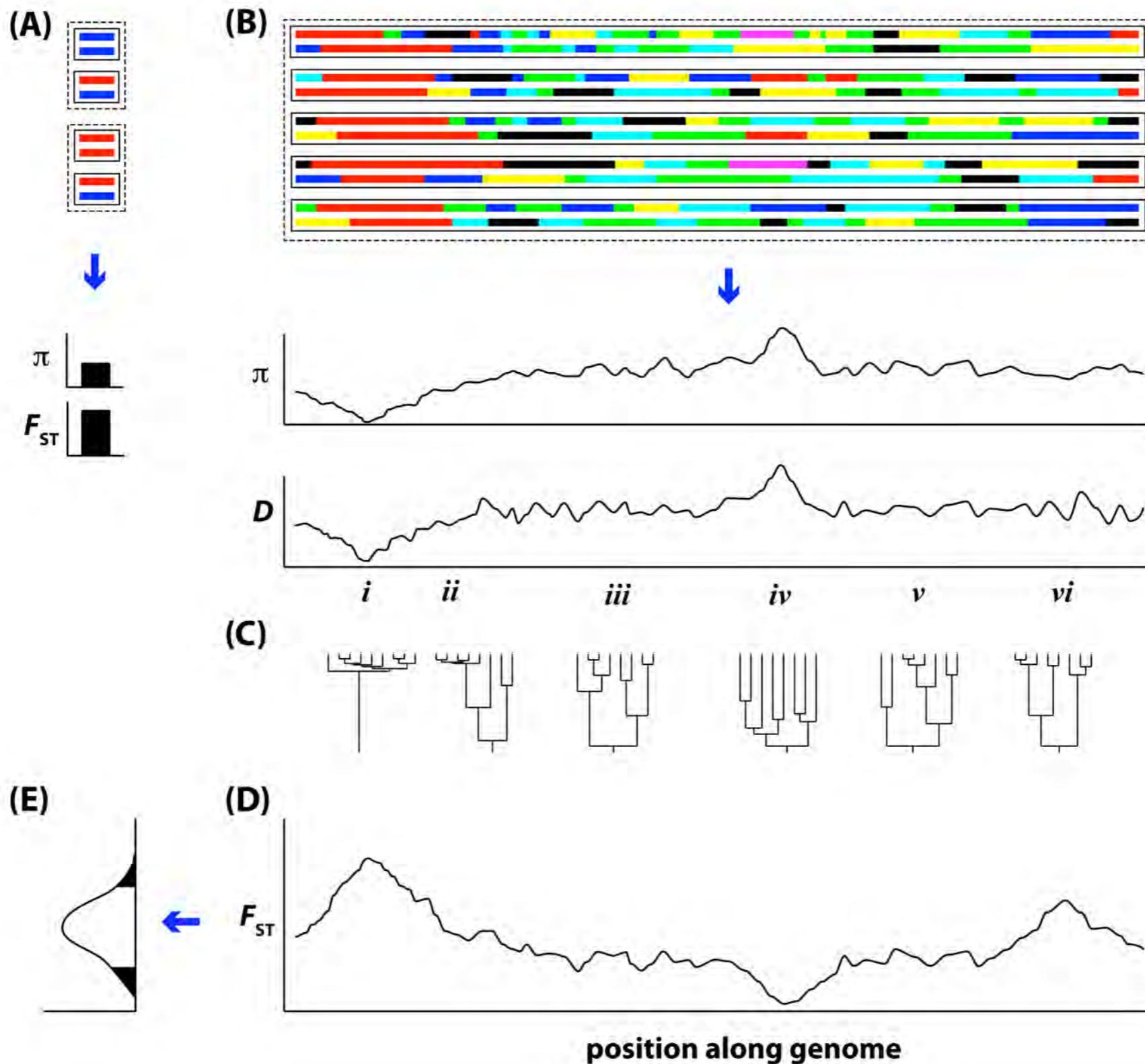
Population Genomics



Population Genetics



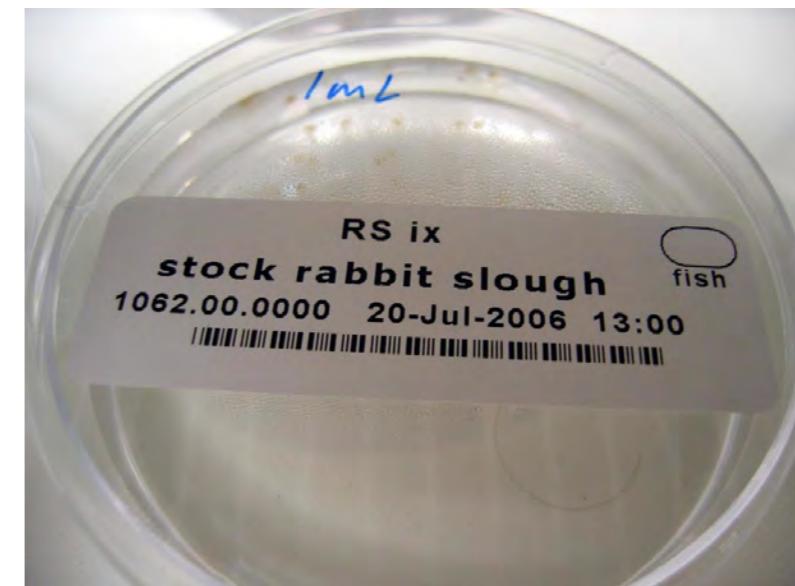
Population Genomics



How do we ‘genomically enable’ research studies of non-model organisms?

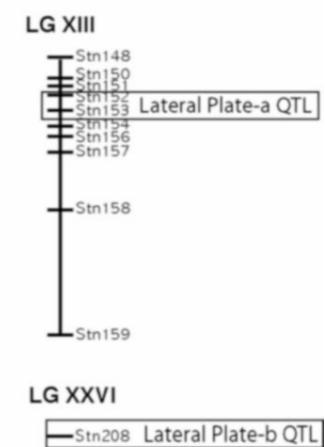
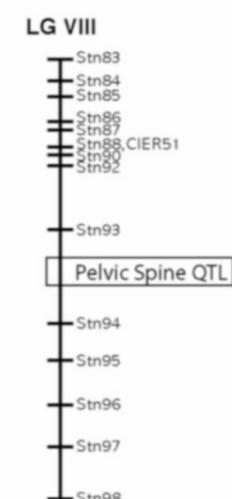
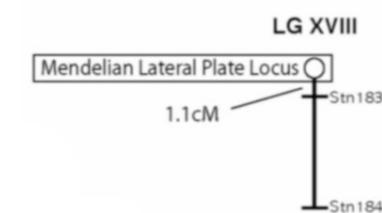
1. Genetic Markers & Genetic Maps
2. Physical Maps
3. Transcriptomes
4. Gene Expression Analyses

In the field and in the lab until a few years ago....



Alaska

British Columbia



The open source genomics breakthrough



Shouldn't we just sequence everything?

(*note* - the answer to this question may be yes soon, and if so I will stop at this slide. But until then....)

Why not sequence the entire genome??

- Still prohibitively expensive for many studies
 - Human height GWAS; over 15,000 individuals assayed
 - Identified many new regions contributing to the variation
 - Still only identified a fraction of the heritability
- For many studies a full sequence isn't necessary
 - genomes of many organisms are organized in linkage blocks
 - well spaced markers will provide the necessary coverage
- Genetic maps are very useful in genomic studies
 - a high density genetic map can facilitate genome assembly
 - genomes may be segregating a lot of structural variation

Alternative approach -

Reduced representation NGS for genotyping

- Focus sequencing on homologous regions across the genome
- Simultaneous identification and typing of single nucleotide polymorphisms (SNPs)
- The cost will be a fraction of the cost of resequencing the genome
 - i.e. 1% genome coverage will be less than 1% the cost
 - often coverage is more even than whole genome sequencing
- Thousands of genomes to be assayed in just a few weeks
- WHY NOT - complete genomic sequence is necessary
 - when linkage disequilibrium blocks (LD) are very short
 - Inferring patterns of LD may be easiest with full sequences

Different flavors of Reduced Representation Library (RRL) Sequencing for genotyping

- Common acronyms
 - **RRL** - Reduced Representation Library
 - **GBS** - Genotyping By Sequencing
 - **CRoPS** - Complexity Reduction of Polymorphic Sequences
 - **MSG** - Multiplex Shotgun Genotyping
 - **RAD** - Restriction site Associated DNA
- All rely on restriction enzyme digestion
- RRL, CRoPS, MSG and GBS use one or two restriction enzymes only
- RAD uses a shearing step to more efficiently capture all restriction sites
- Incorporation of barcodes on adaptors for multiplexing
- Aligned against a reference genome or assembled *de novo*
- Statistical issues
 - new level of sampling variation (sequencing in addition to biological)
 - sequencing error and problems for aligning or clustering

What is RAD-seq?

(Restriction-site Associated DNA)



Illumina

2007

Rapid and cost-effective polymorphism identification and genotyping using restriction site associated DNA (RAD) markers

Michael R. Miller,¹ Joseph P. Dunham,² Angel Amores,³ William A. Cresko,² and Eric A. Johnson^{1,*}

¹Institute for Molecular Biology, University of Oregon, Eugene, Oregon 97403, USA, ²Center for Ecology & Evolutionary Biology, University of Oregon, Eugene, Oregon 97403, USA, ³Institute of Neuroscience, University of Oregon, Eugene, Oregon 97403, USA

2008

OPEN ACCESS Freely available online

plos one

Rapid SNP Discovery and Genetic Mapping Using Sequenced RAD Markers

Nathan A. Baird^{1,2}, Paul D. Etter^{1,2}, Tressa S. Atwood², Mark C. Currey¹, Anthony L. Shiver¹, Zachary A. Lewis¹, Eric U. Selker¹, William A. Cresko², Eric A. Johnson^{1,*}

¹Institute of Molecular Biology, University of Oregon, Eugene, Oregon, United States of America, ²Center for Ecology & Evolutionary Biology, University of Oregon, Eugene, Oregon, United States of America

What is RAD-seq?

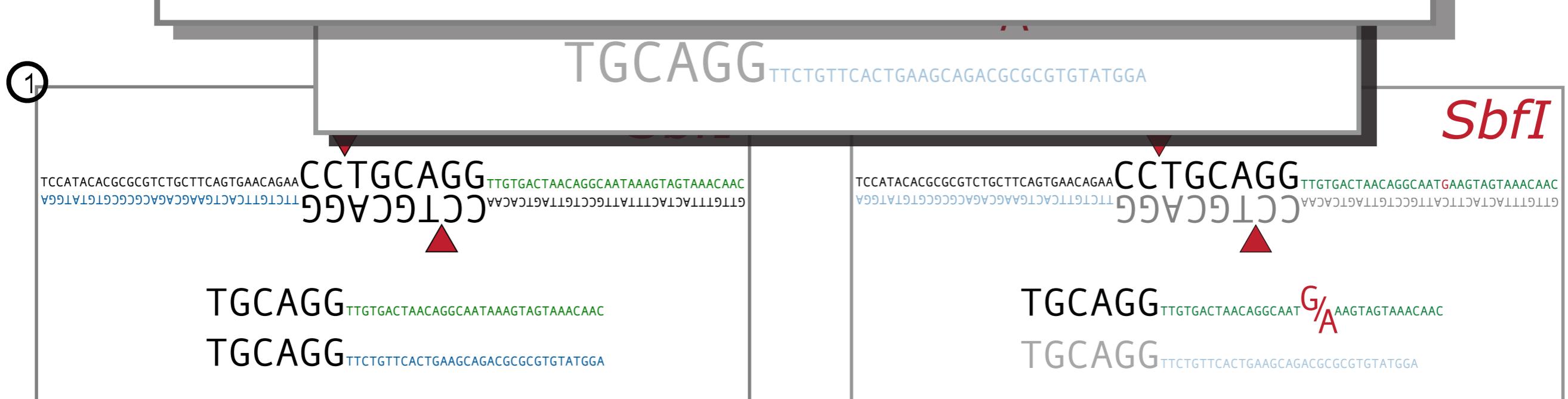
(Restriction-site Associated DNA)

Chr I | | | | | | | | | | | | | | | | | | | |

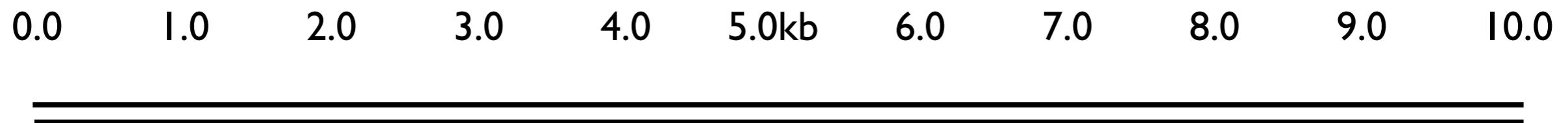
22,830 *SbfI* sites in threespine stickleback

~ 45,000 RAD-Tags

HiSeq Illumina Lane:
| 60 million reads, > 96 barcoded individuals



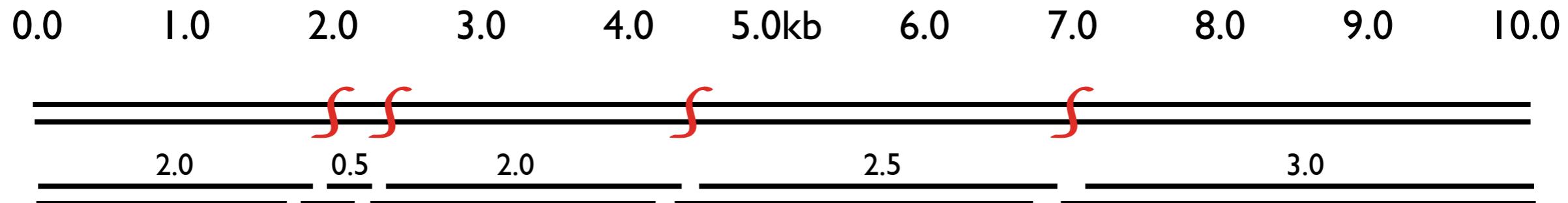
Restriction Enzyme (RE) digestion and first adaptor ligation



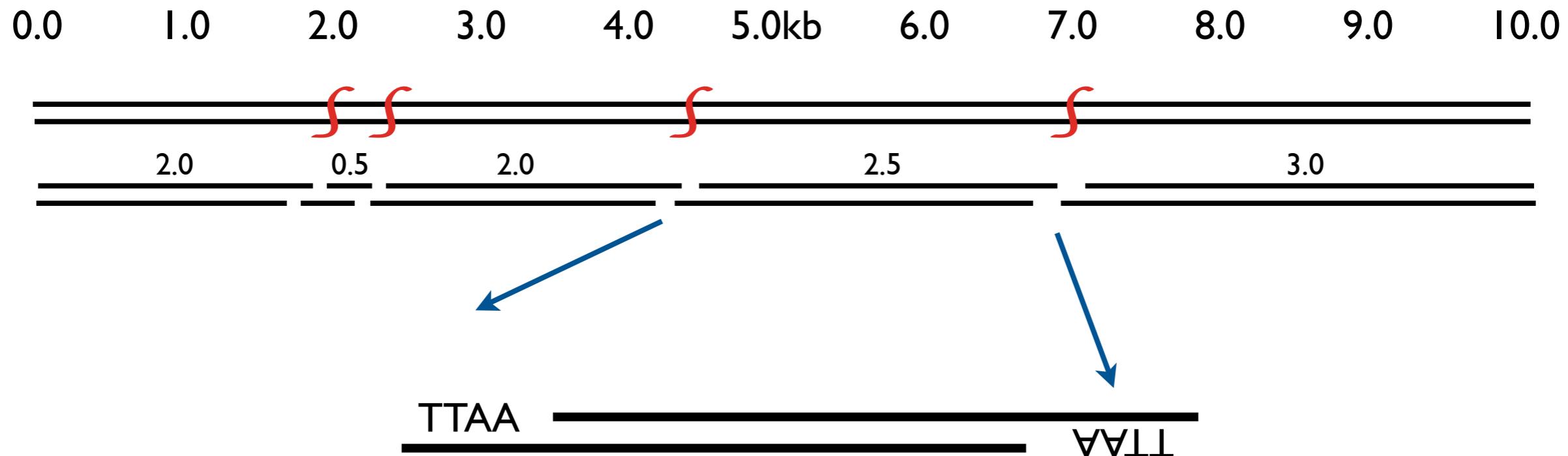
Restriction Enzyme (RE) digestion and first adaptor ligation



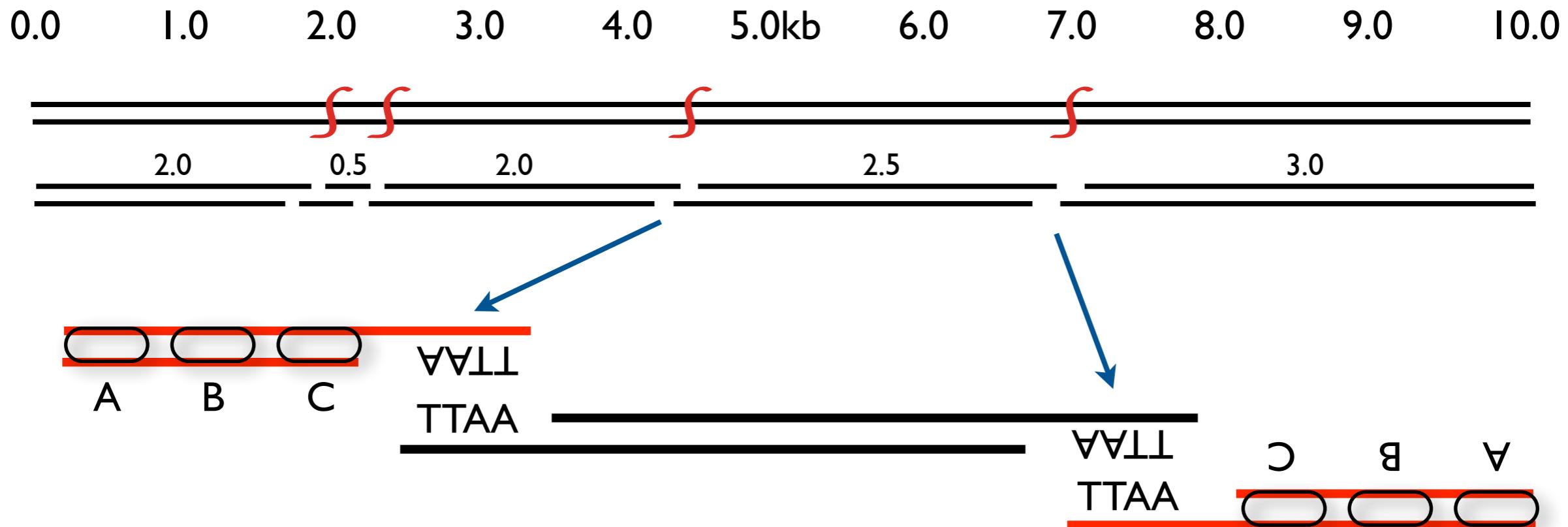
Restriction Enzyme (RE) digestion and first adaptor ligation



Restriction Enzyme (RE) digestion and first adaptor ligation

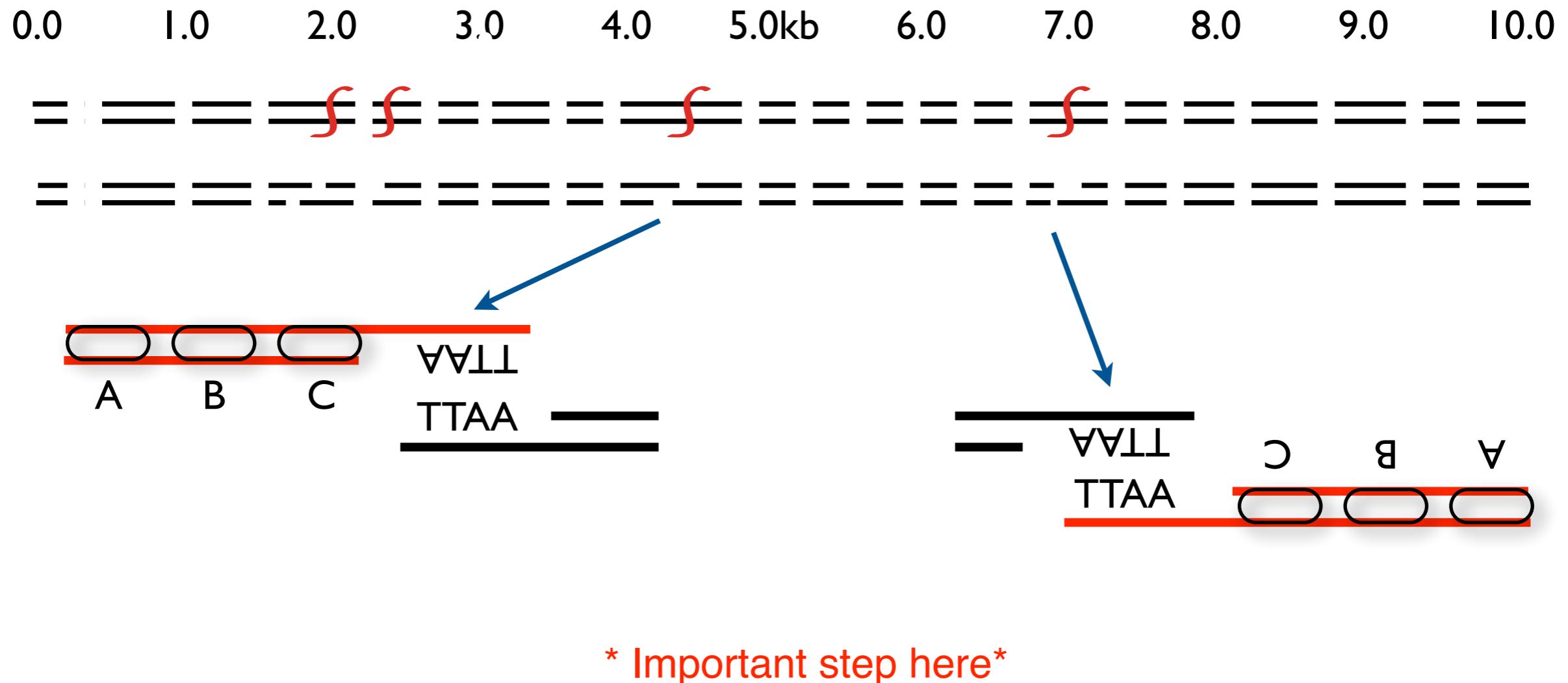


Restriction Enzyme (RE) digestion and first adaptor ligation



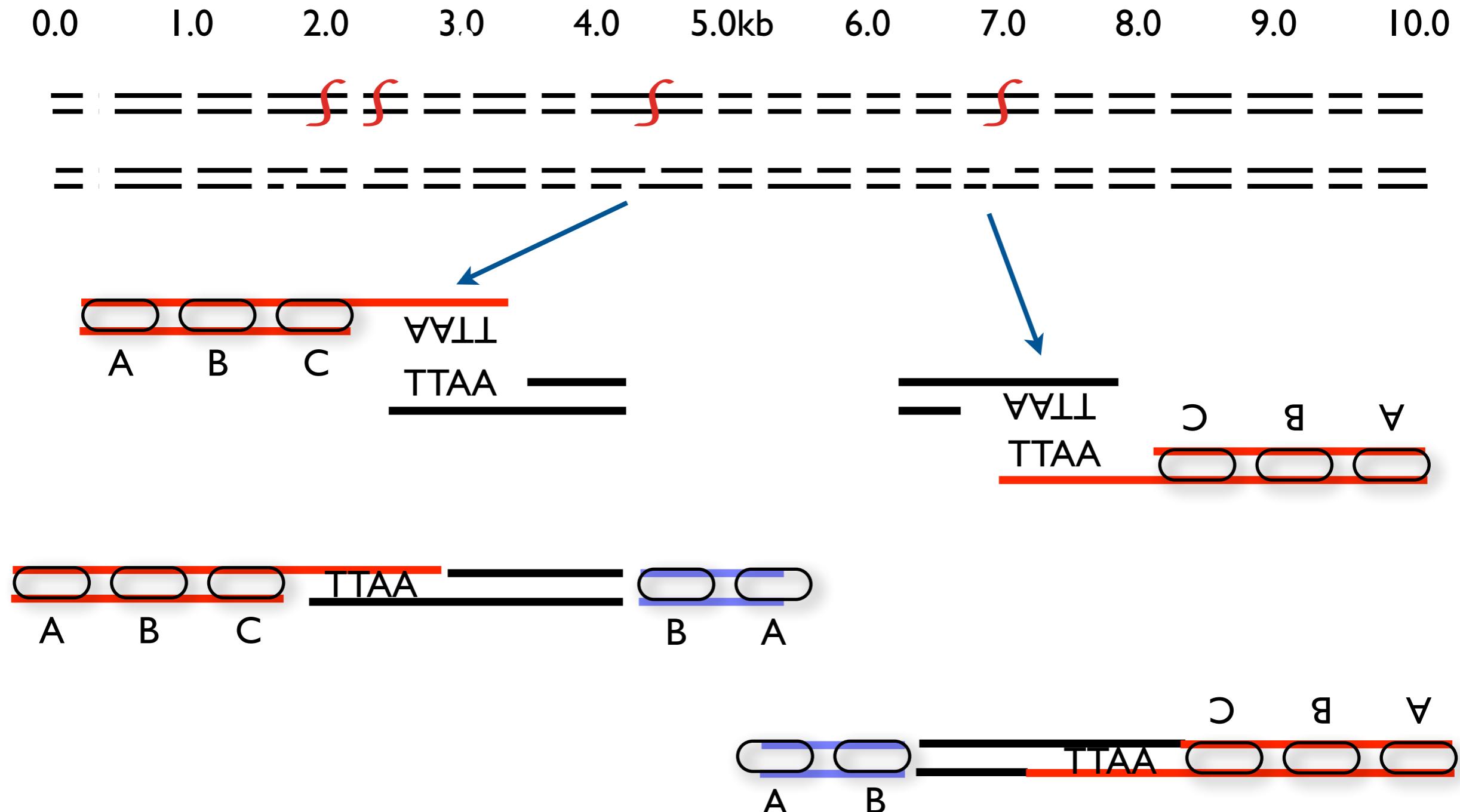
A = Amplification primer
B = Sequencing primer
C = Barcode

Shearing and second adaptor ligation



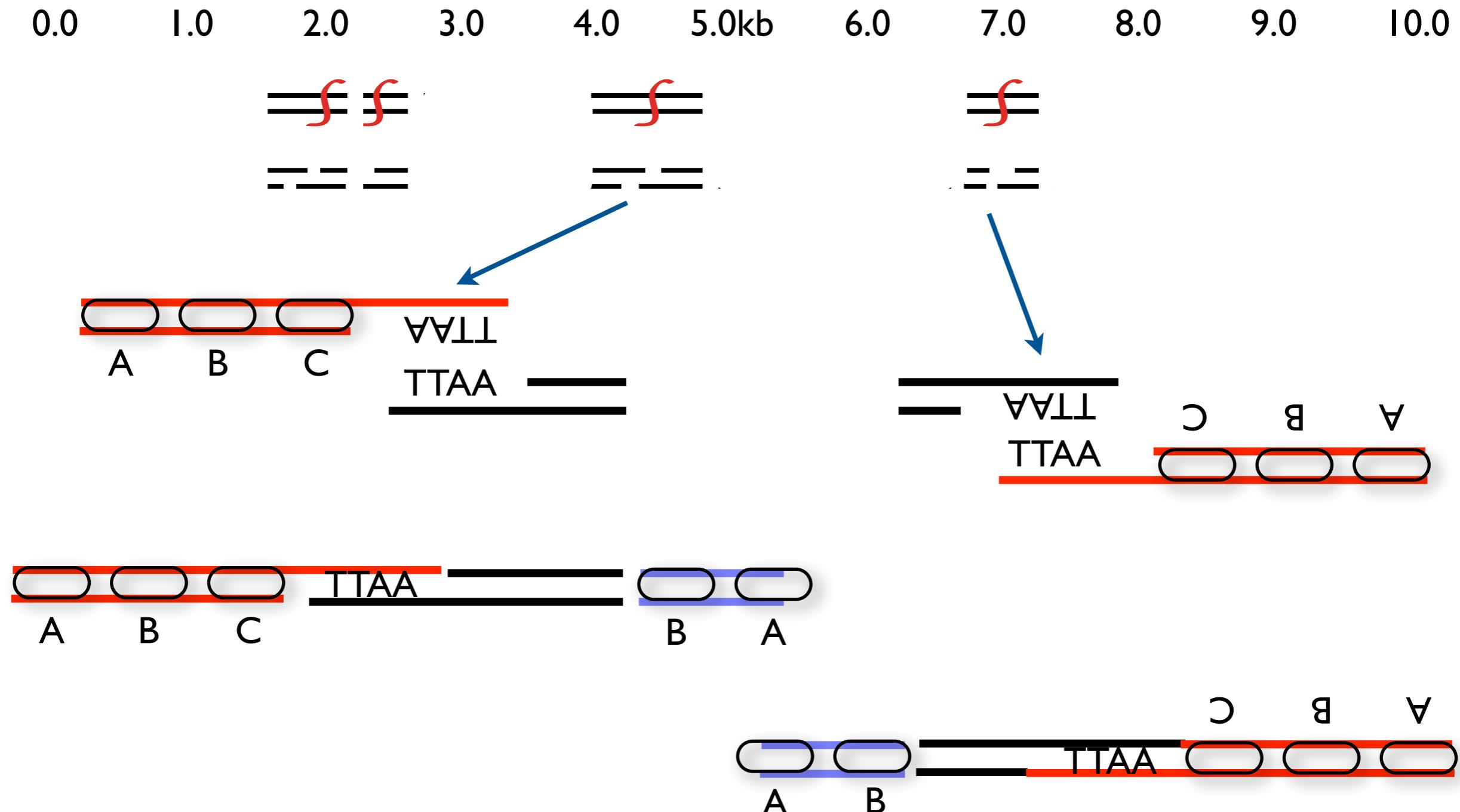
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Shearing and second adaptor ligation



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Shearing and second adaptor ligation

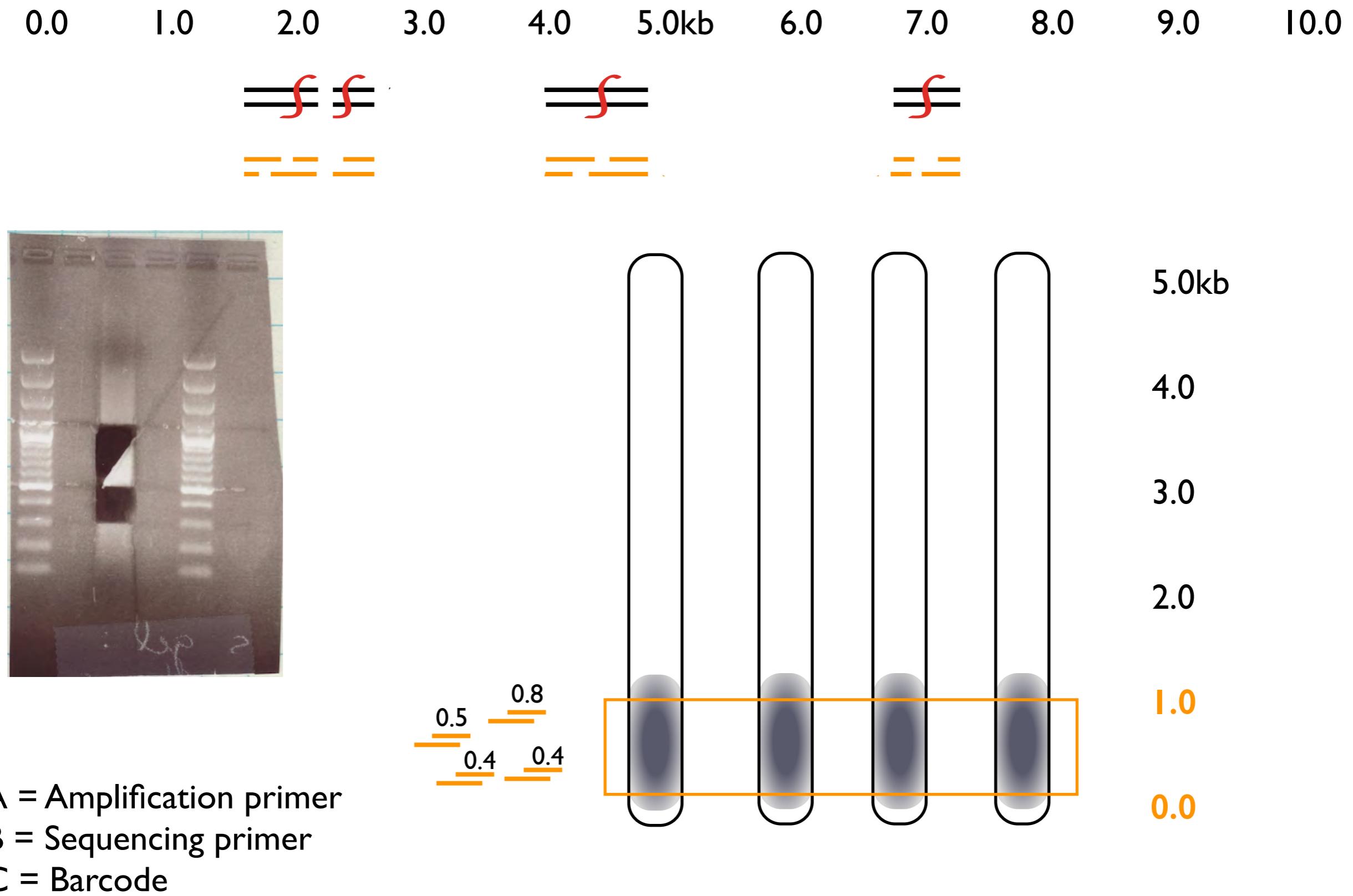


A = Amplification primer

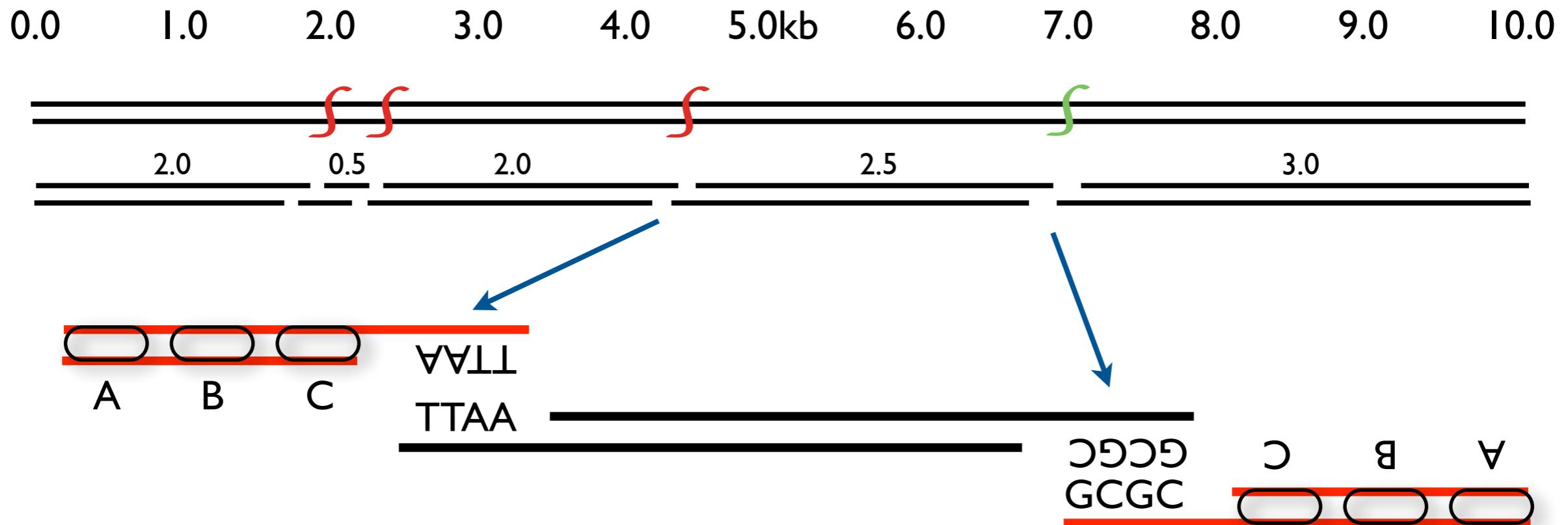
B = Sequencing primer

C = Barcode

Shearing makes consistent fragments for sequencing

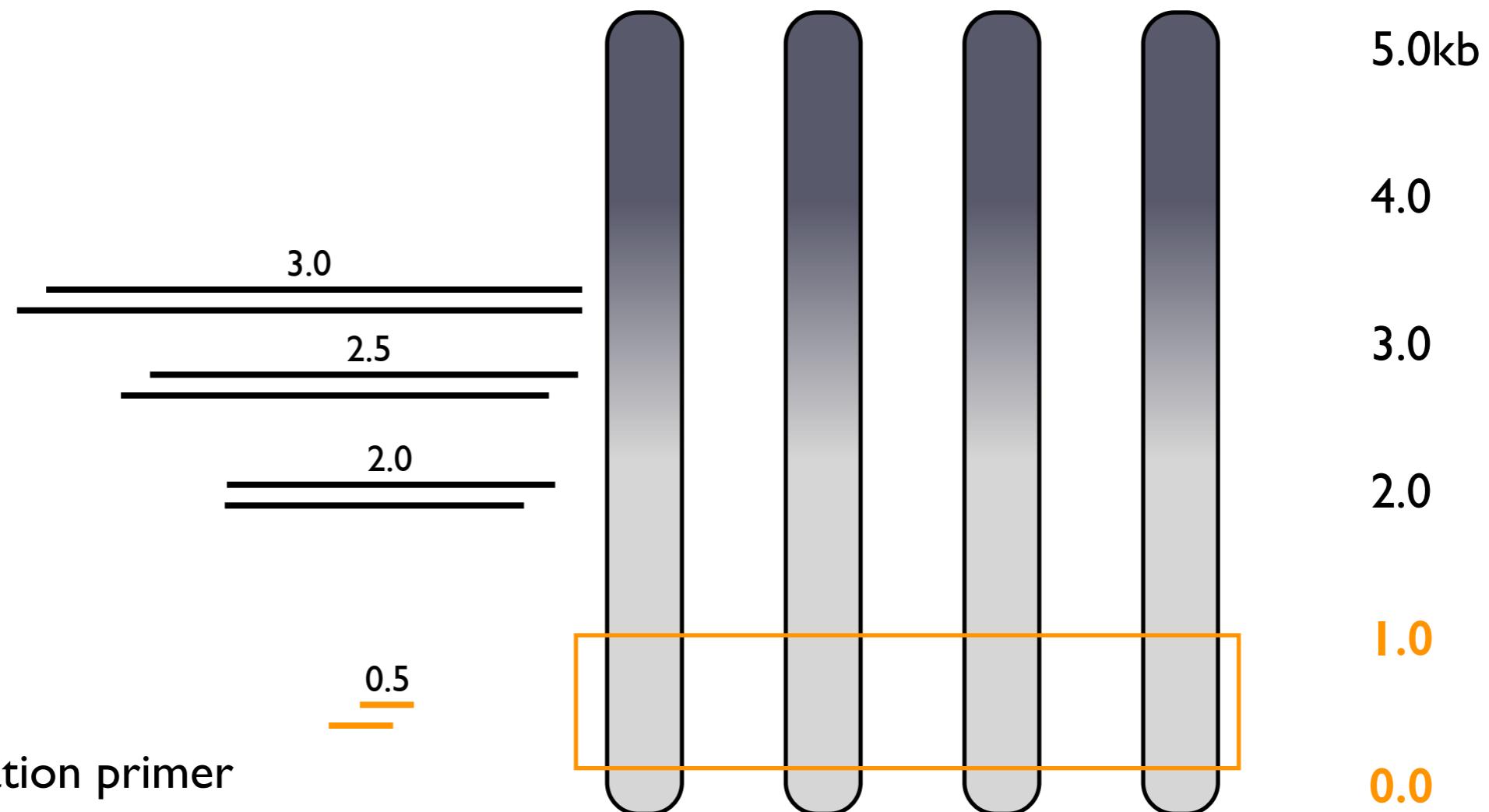
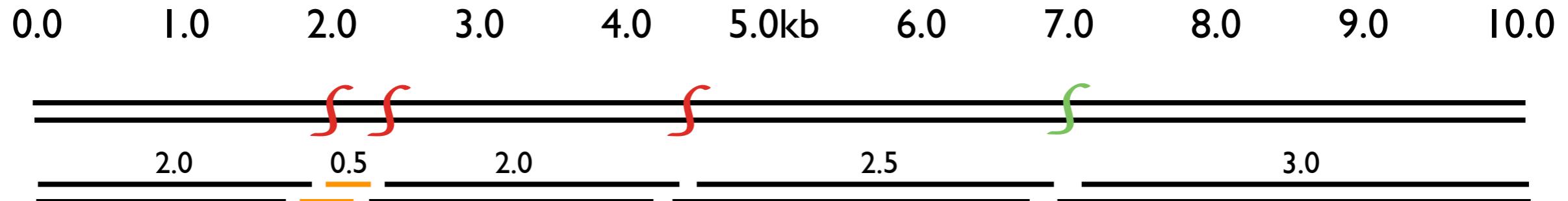


Single (GBS) or Double Digest RAD (ddRAD)



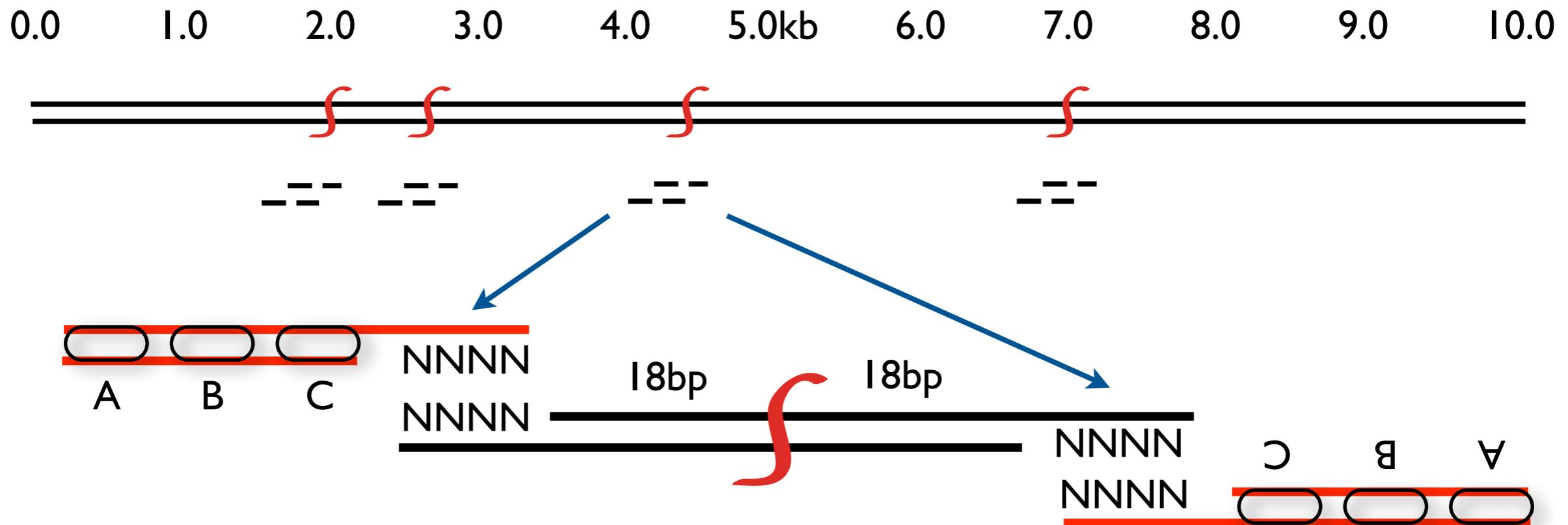
A = Amplification primer
B = Sequencing primer
C = Barcode

Size selection is more problematic without shearing



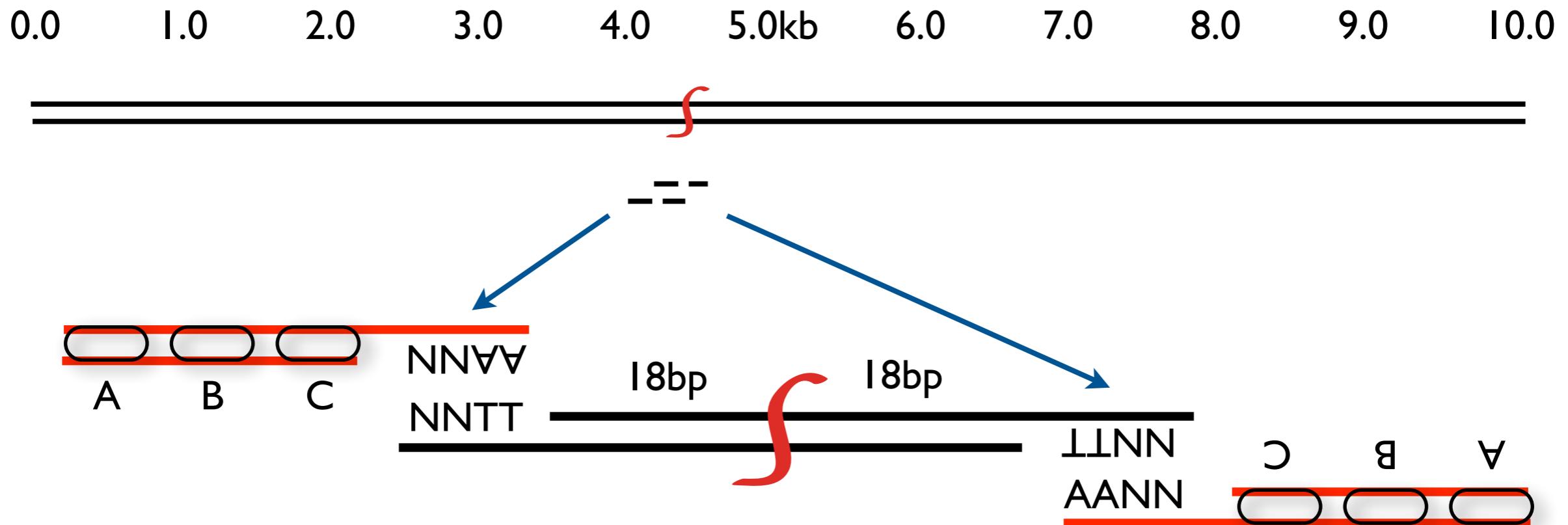
A = Amplification primer
B = Sequencing primer
C = Barcode

2bRAD - type 2b restriction enzyme



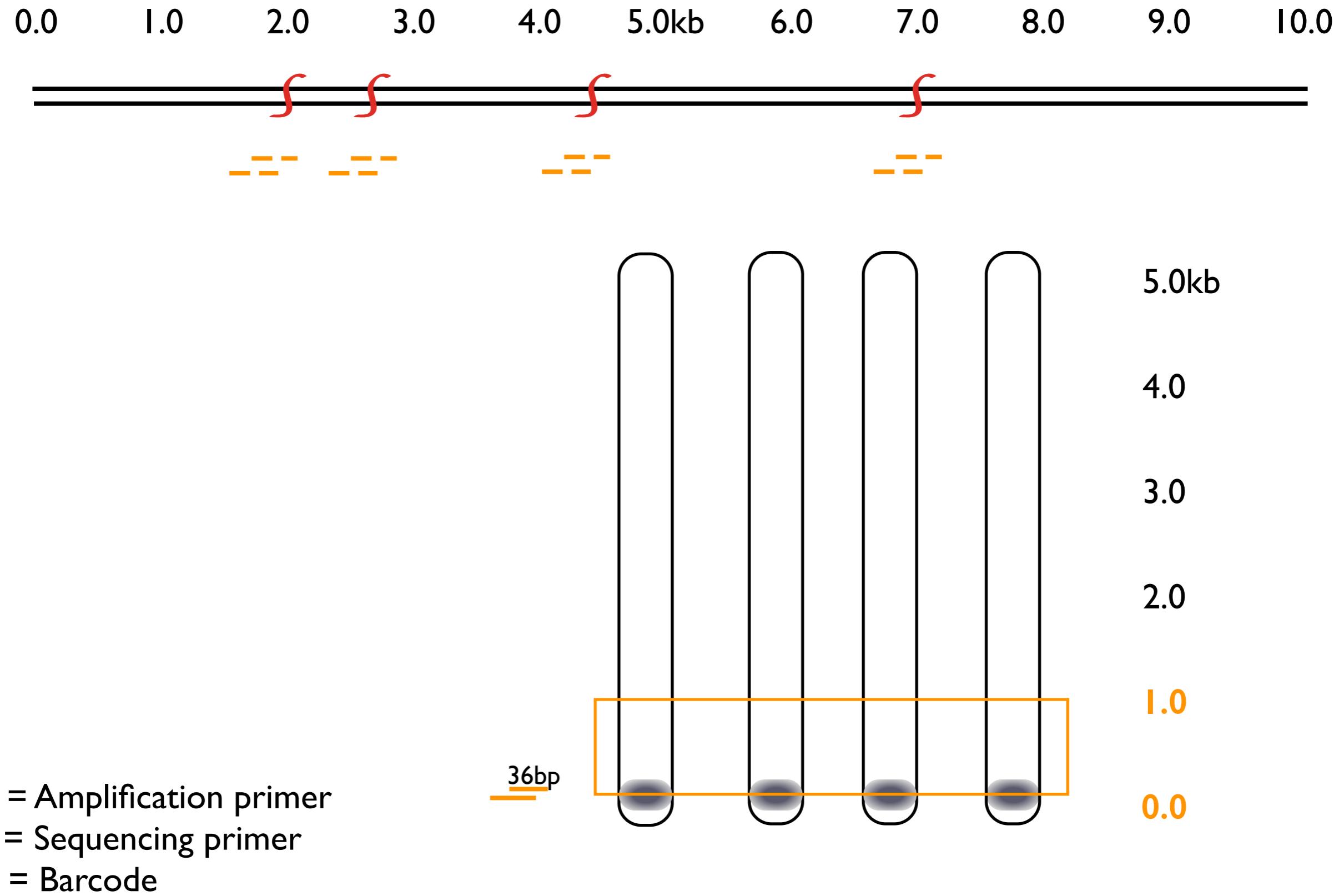
A = Amplification primer
B = Sequencing primer
C = Barcode

2bRAD - can scale number of markers easily



A = Amplification primer
B = Sequencing primer
C = Barcode

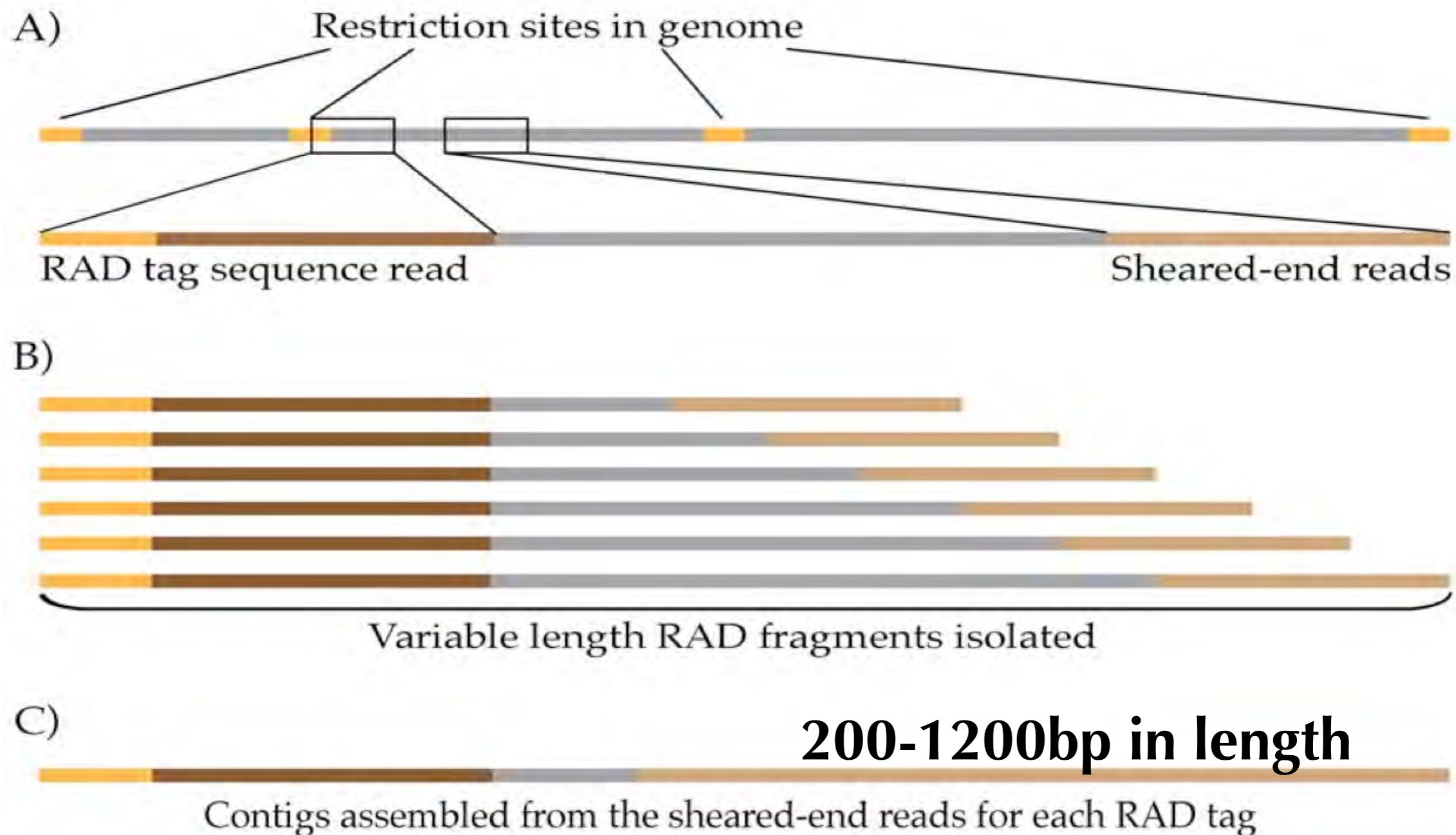
2bRAD - size selection is difficult



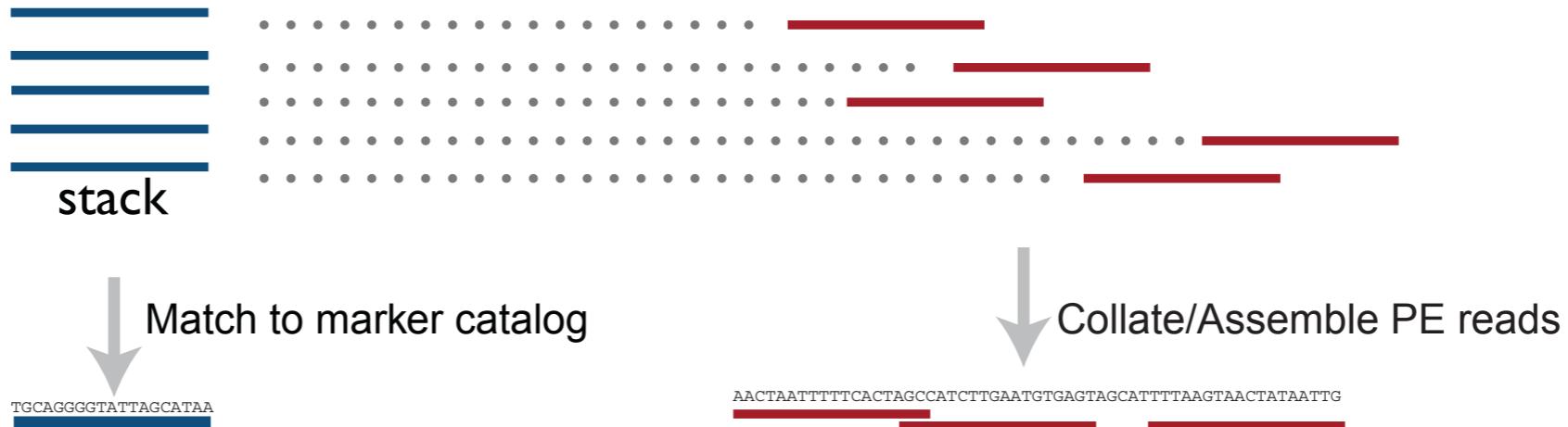
Summary of plusses and minuses of RAD family

	Sheared RAD	Single or ddRAD	2b-RAD
plusses	<ul style="list-style-type: none">- Consistent reads- Local assemblies- Identify PCR duplicates	<ul style="list-style-type: none">- Fewer steps- Easy marker scaling	<ul style="list-style-type: none">- Fewest steps- Easy marker scaling
minuses	<ul style="list-style-type: none">- Shearing step- Scaling requires different enzymes	<ul style="list-style-type: none">- Multiple enzymes- Poor consistency- PCR duplicates	<ul style="list-style-type: none">- Very short reads- PCR duplicates

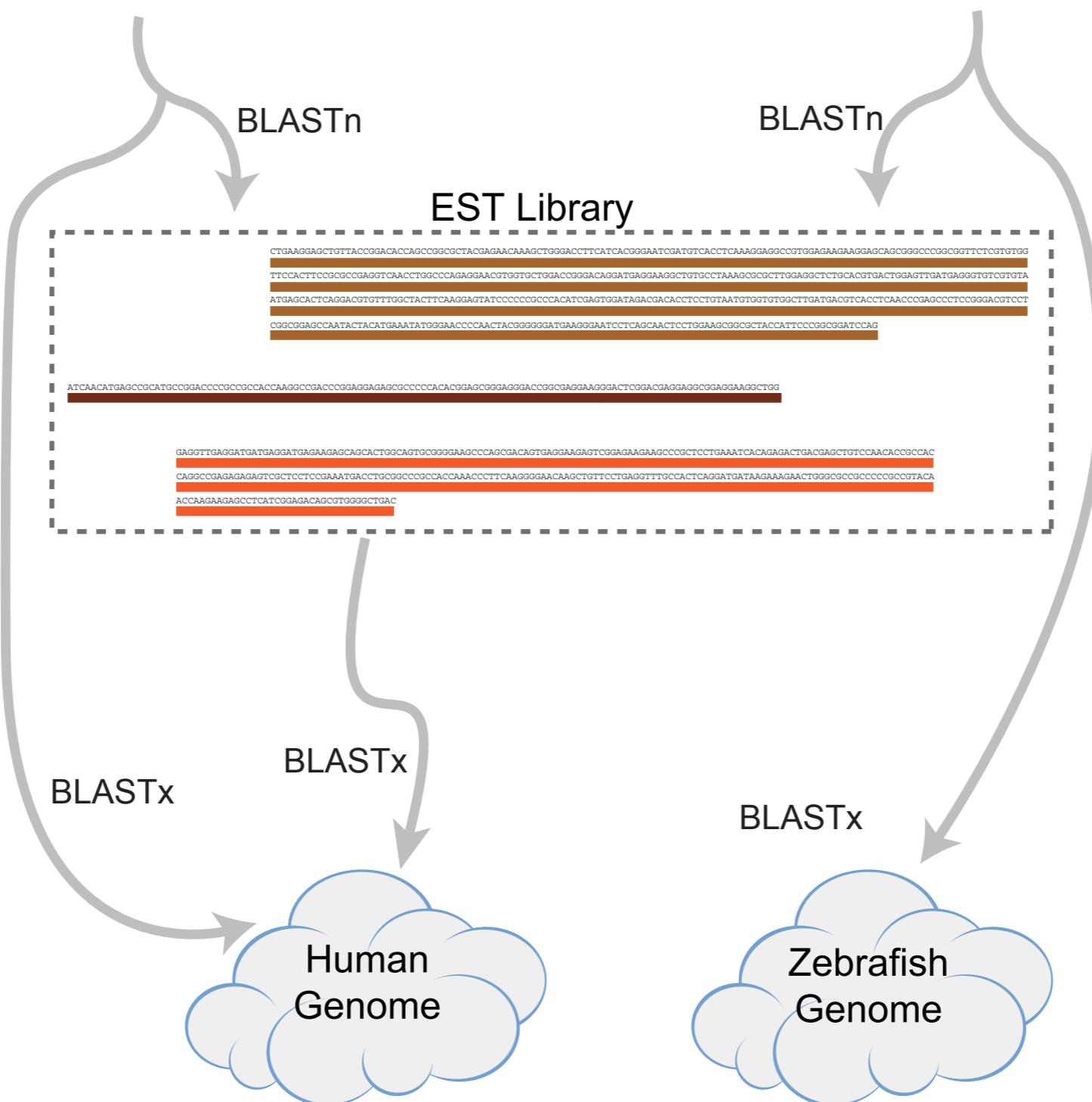
Additional benefits of random shearing in RAD



Acquire paired-end sequence



Associate markers / PE contigs with ESTs



What can you do with the RAD-seq data?
Case studies of using RAD for an organism
with a reference genome: population
genomics of threespine stickleback fish

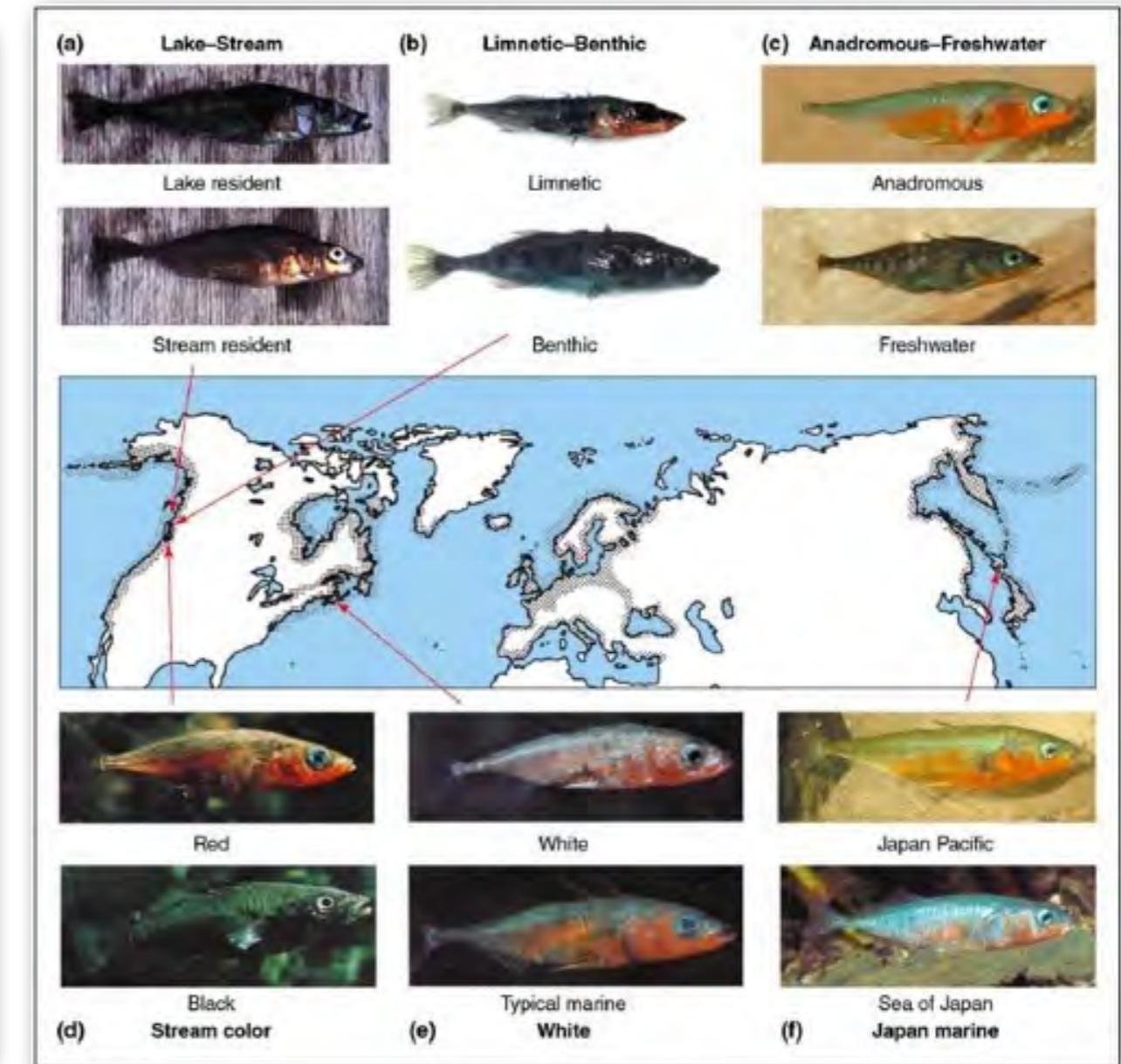


- 1) Population genomic structure of Oregon stickleback
- 2) Population genomics of extremely rapid evolution on new islands

Threespine stickleback, *Gasterosteus aculeatus*



Threespine stickleback, *Gasterosteus aculeatus*



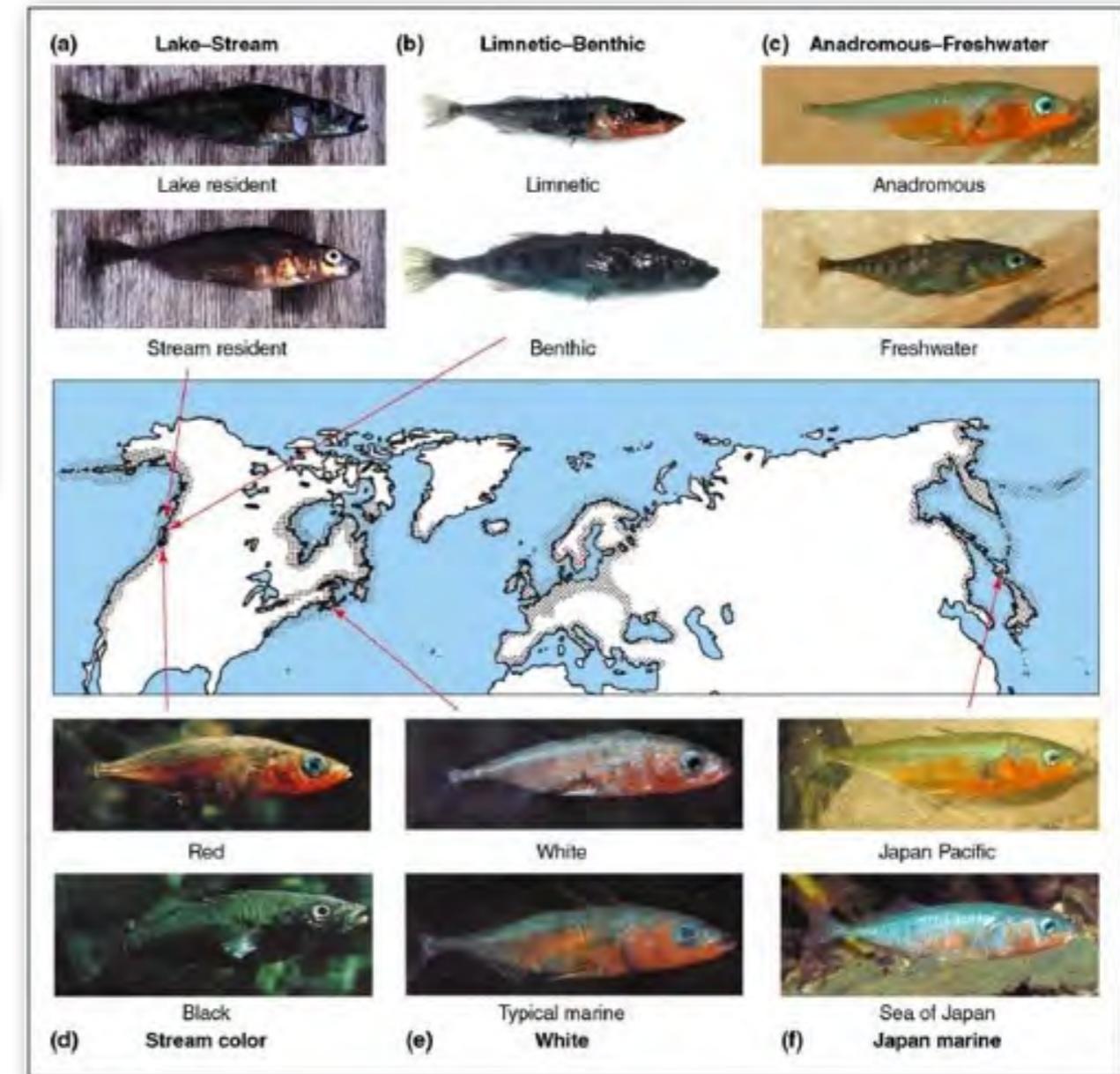
Rundle and McKinnon 2002

Threespine stickleback, *Gasterosteus aculeatus*

Pelvic
Structure



Lateral
Plates



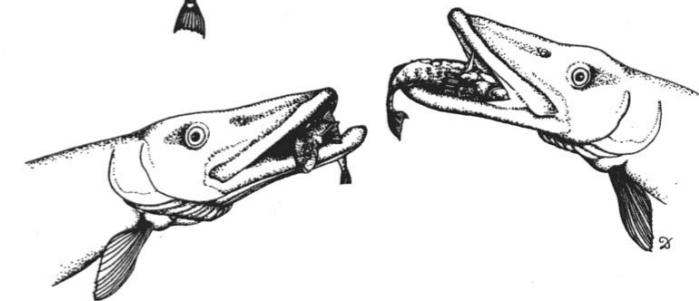
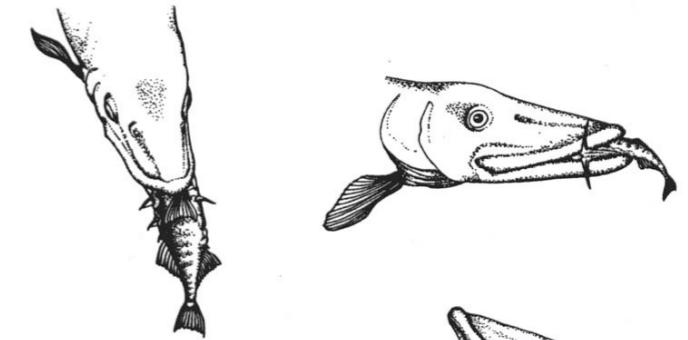
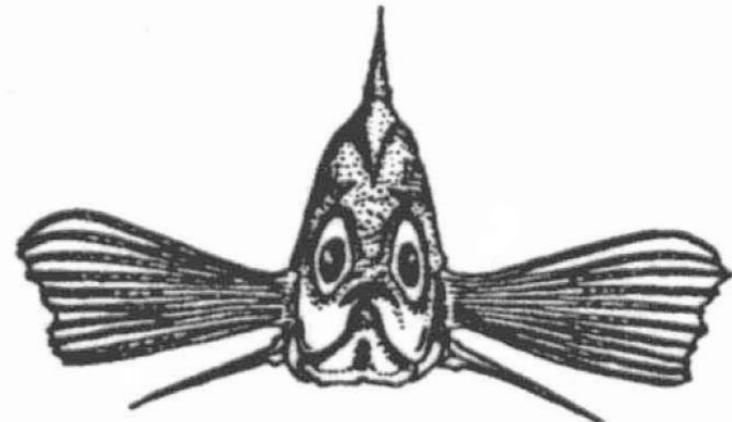
Rundle and McKinnon 2002

Threespine stickleback, *Gasterosteus aculeatus*

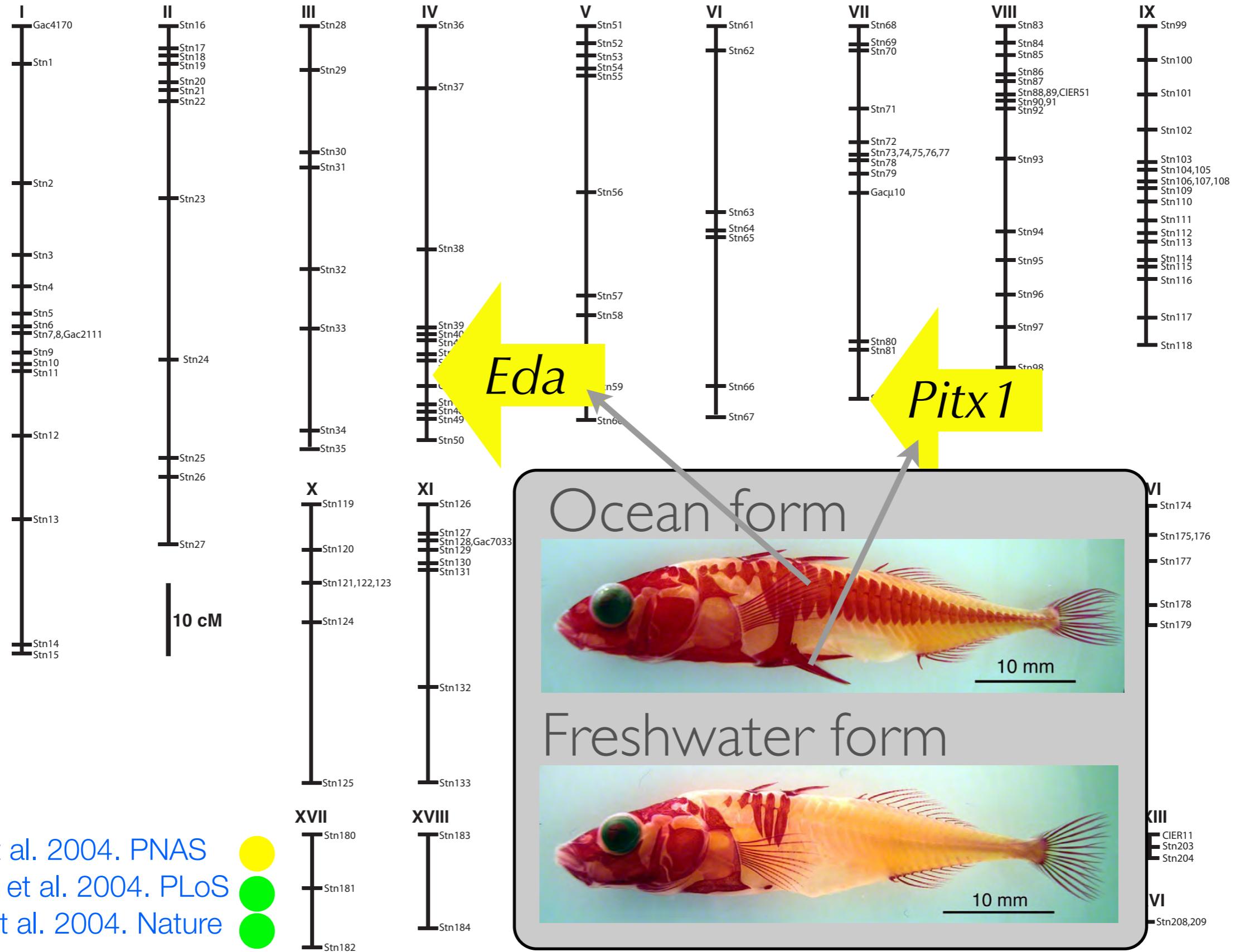
Pelvic
Structure



Lateral
Plates



Laboratory mapping of large effect loci





Stickleback phenotypes mapped in the lab so far....

Pelvic structure size and shape *** (*Eda*)

Lateral plate number *** (*Pitx1*)

Body coloration *** (*KitL*)

Opercle bone shape

Pelvic spine length

Body shape

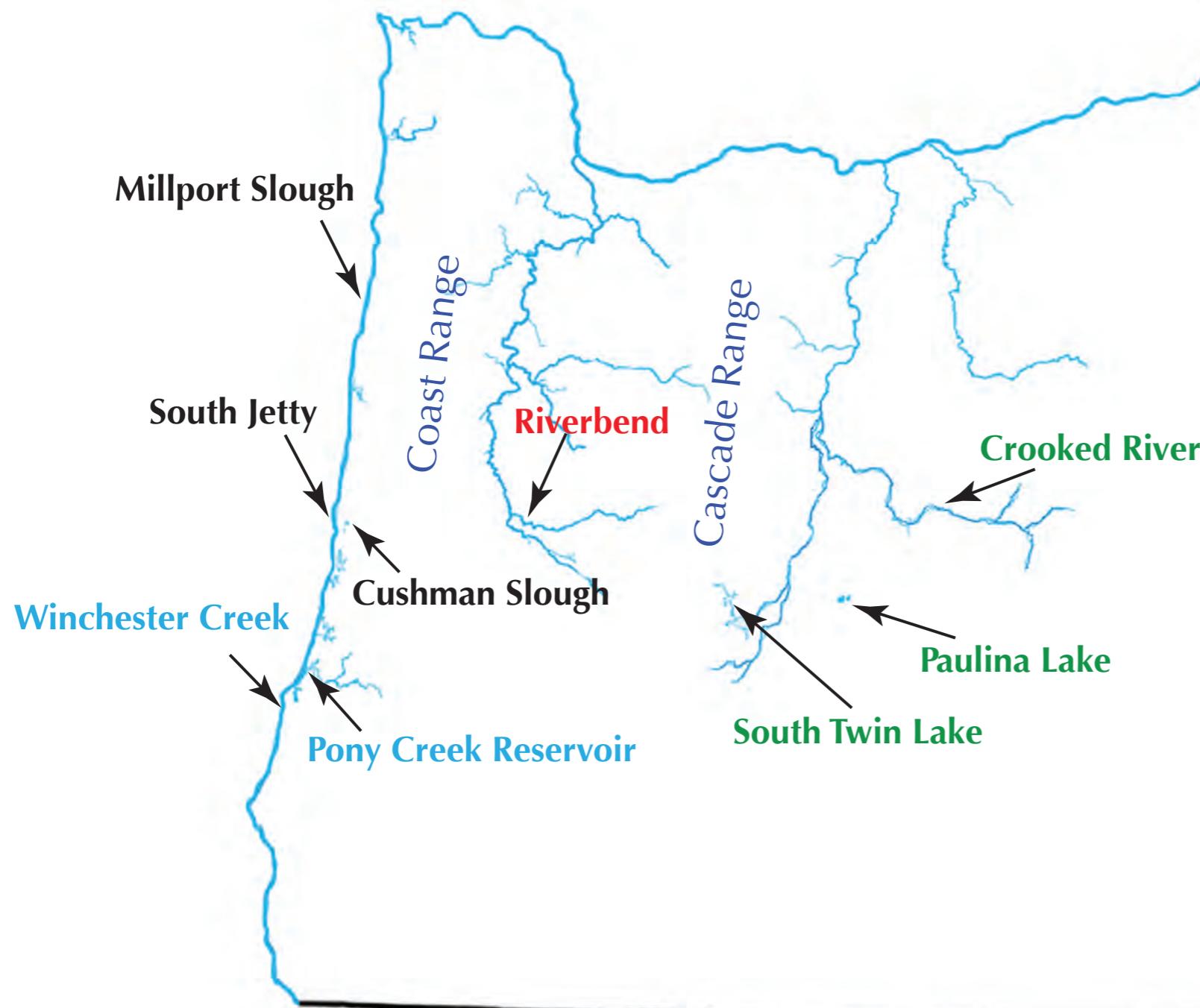
Courtship behavior

Gill raker size

Dorsal spine length

-
- A trend of large effect loci identified in the laboratory
 - Similar genomic regions and sometimes alleles mapped in independent populations
 - A problem is that laboratory mapping approaches are under-powered in stickleback
 - A question is whether population genomics studies can provide complementary or more complete information.

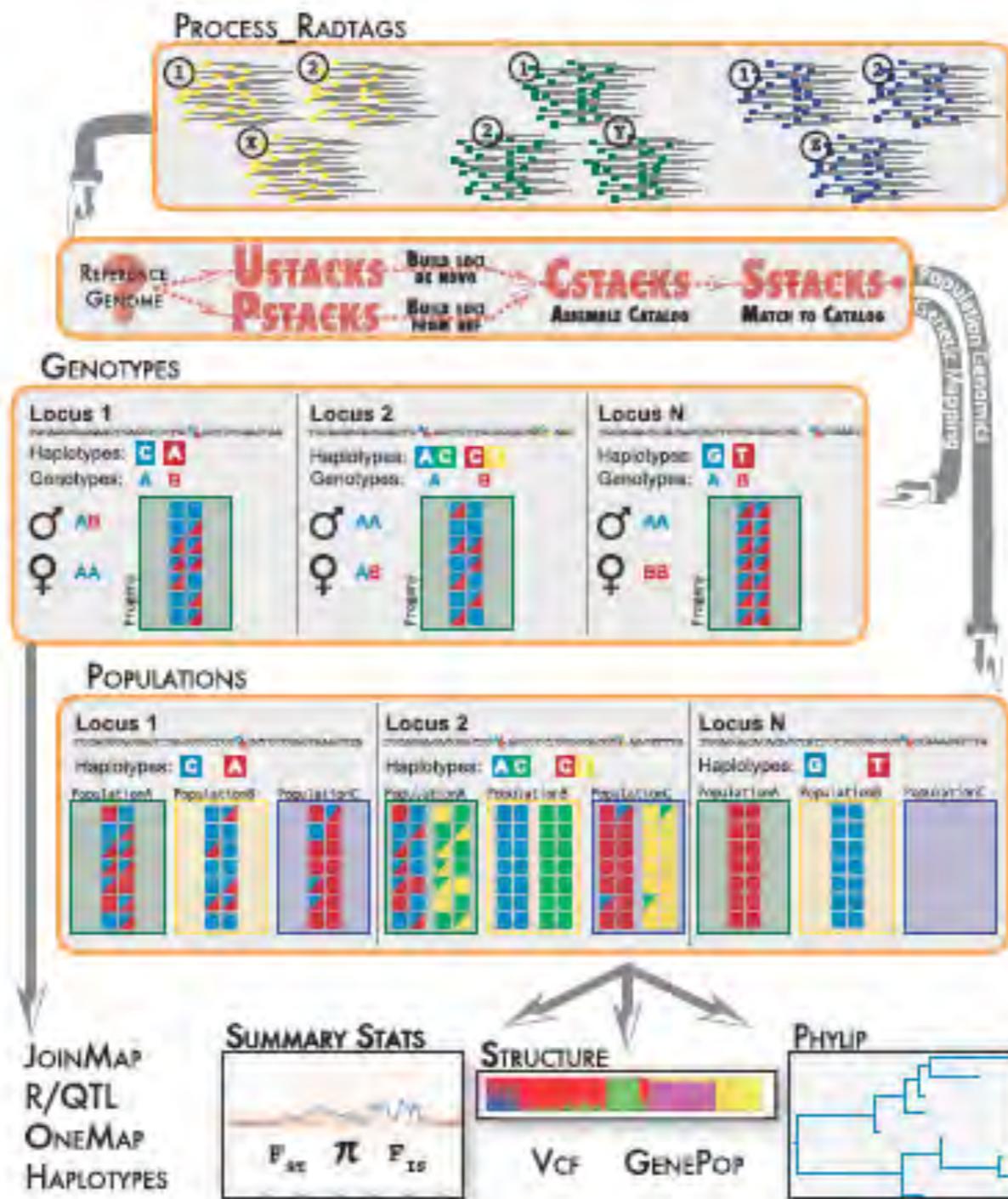
Population genomic structure of Oregon stickleback



590 Individuals
115,000 SNPs each



Stacks analysis pipeline for RAD-seq



Stacks: Building and Genotyping Loci De Novo From Short-Read Sequences

Julian M. Catchen,* Angel Amores,[†] Paul Hohenlohe,^{*} William Cresko,^{*} and John H. Postlethwait^{†,1}

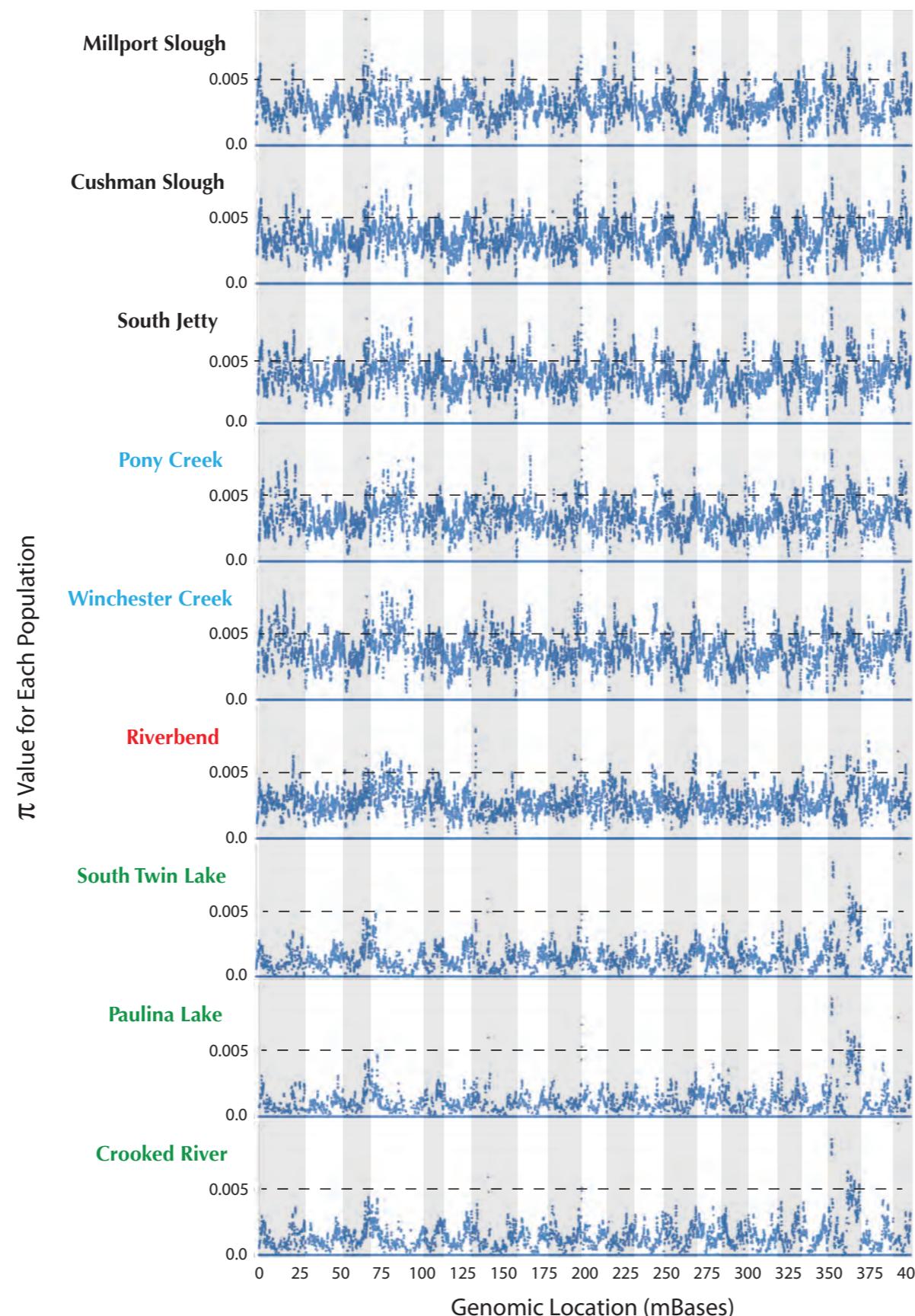
^{*}Center for Ecology and Evolutionary Biology and [†]Institute of Neuroscience, University of Oregon, Eugene, Oregon 97403

Stacks: an analysis tool set for population genetics

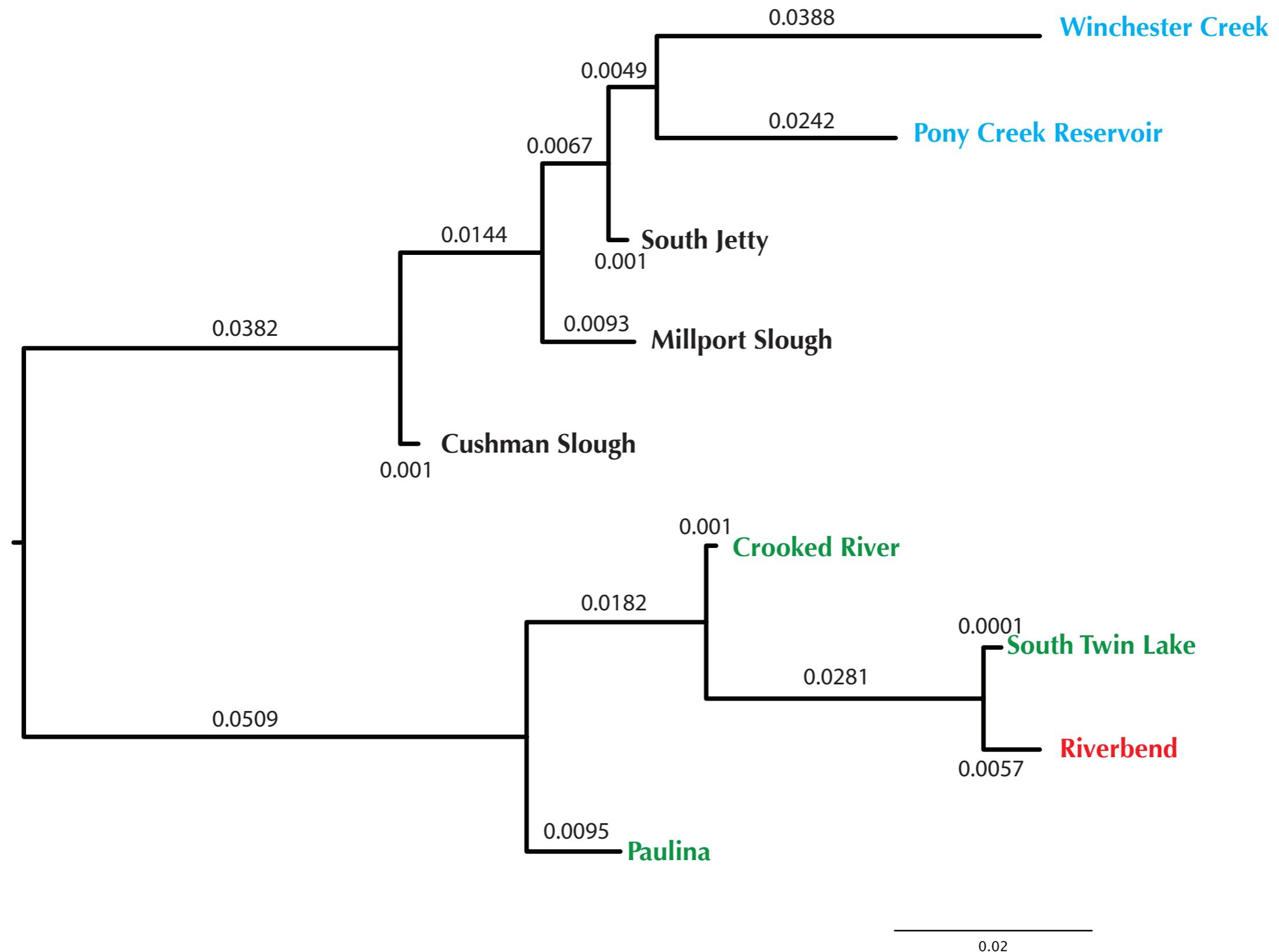
JULIAN CATCHEN,* PAUL A. HOHENLOHE,*[†] SUSAN BASSHAM,* ANGEL AMORES[‡] and WILLIAM A. CRESKO*

^{*}Institute of Ecology and Evolution, University of Oregon, Eugene, OR 97403-5289, USA, [†]Biological Sciences, University of Idaho, Moscow, ID 83844-3051, USA, [‡]Institute of Neuroscience, University of Oregon, Eugene, OR 97403-1254, USA

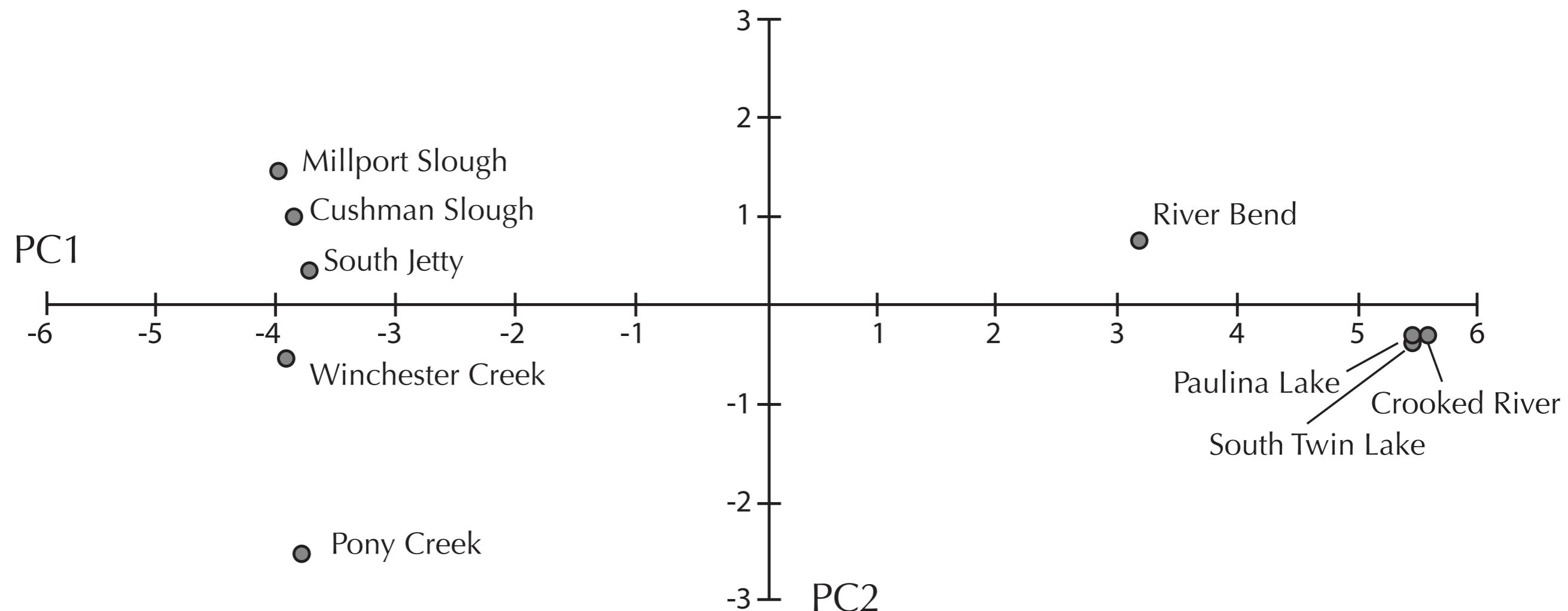
Genetic diversity across populations



Phylogenetic relationship among populations

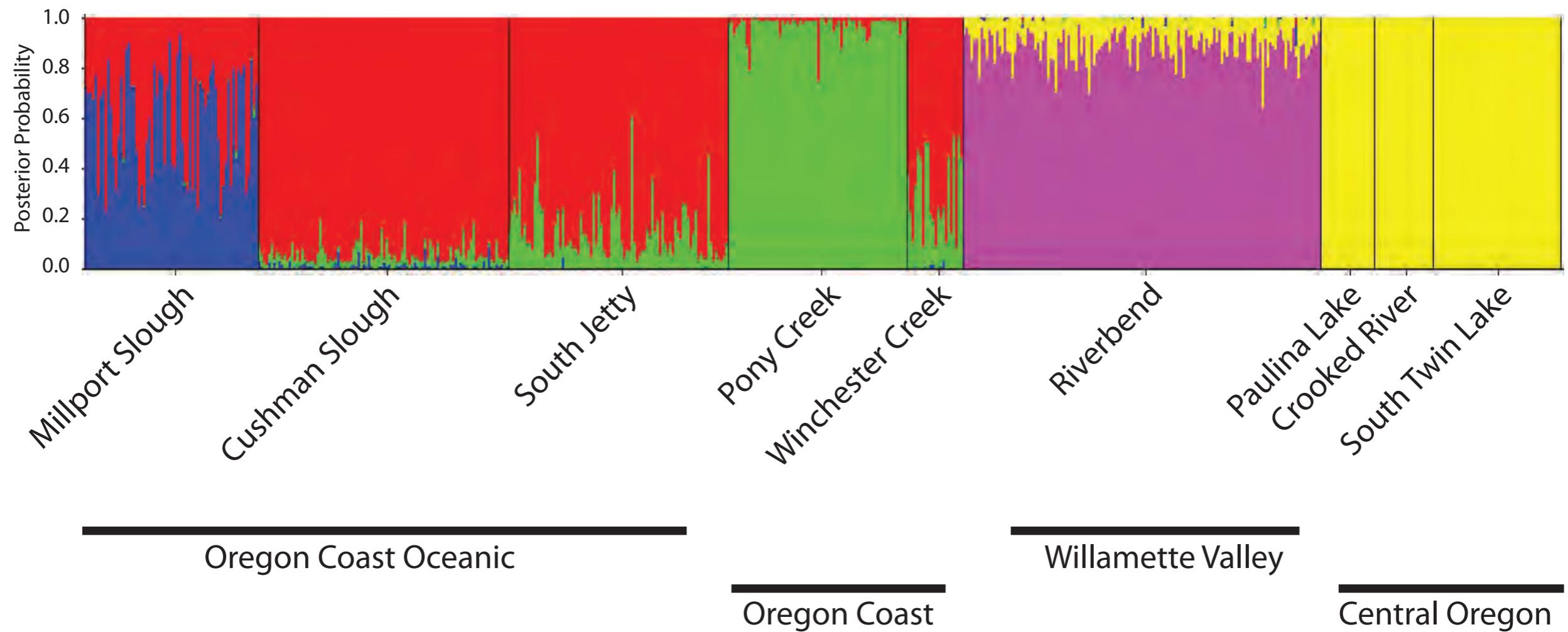


Population structure using PCA



PC 1 explains 89% of the overall variance

Population structure using Bayesian analysis (*Structure*)



What genomic regions are associated with the different habitats?

How quickly can the allele frequencies change?

Shake rattle and evolve in 50 years team earthquake



Susan
Bassham



Julian
Catchen



Emily
Lescak

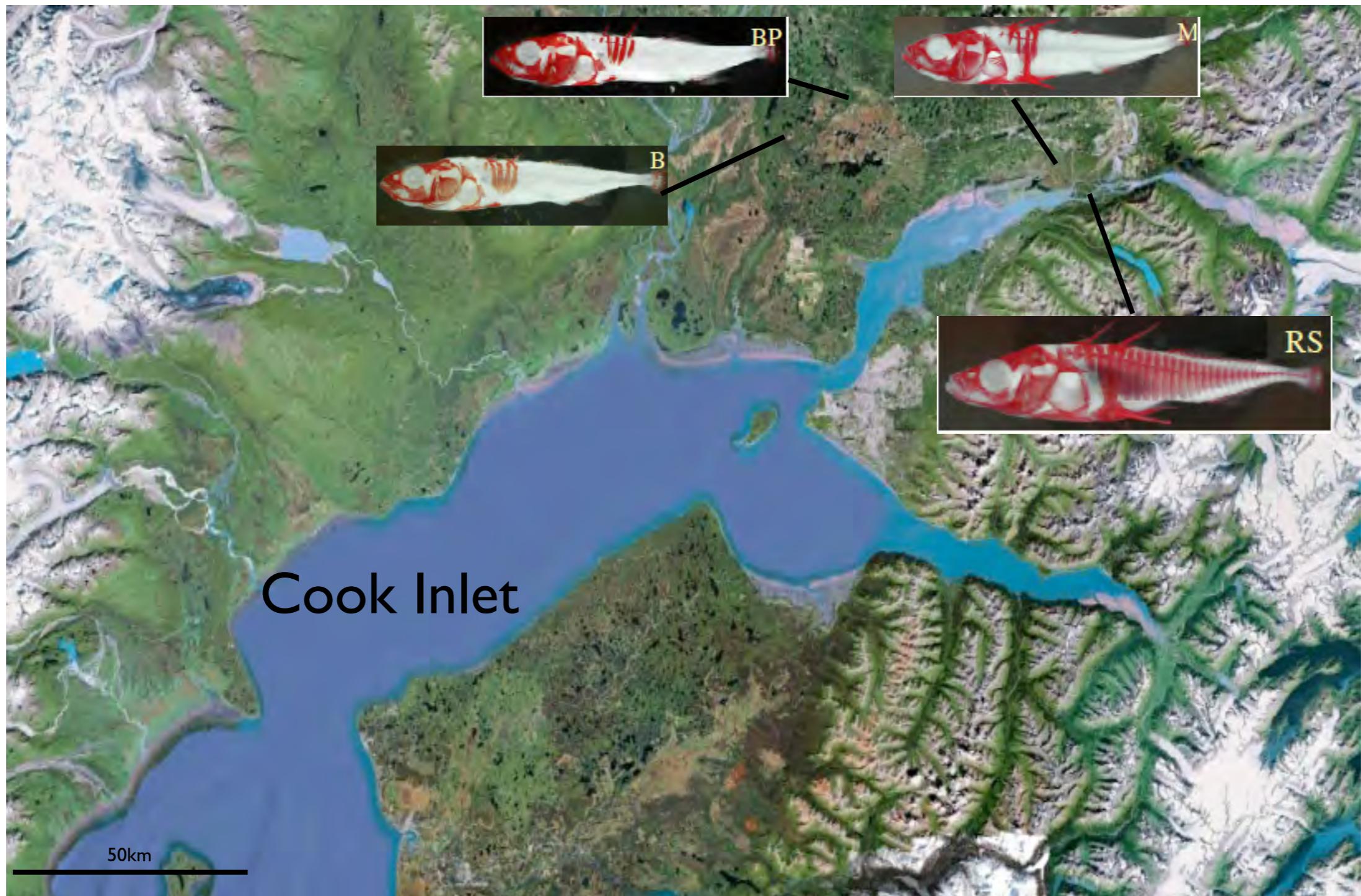


Mary
Sherbick

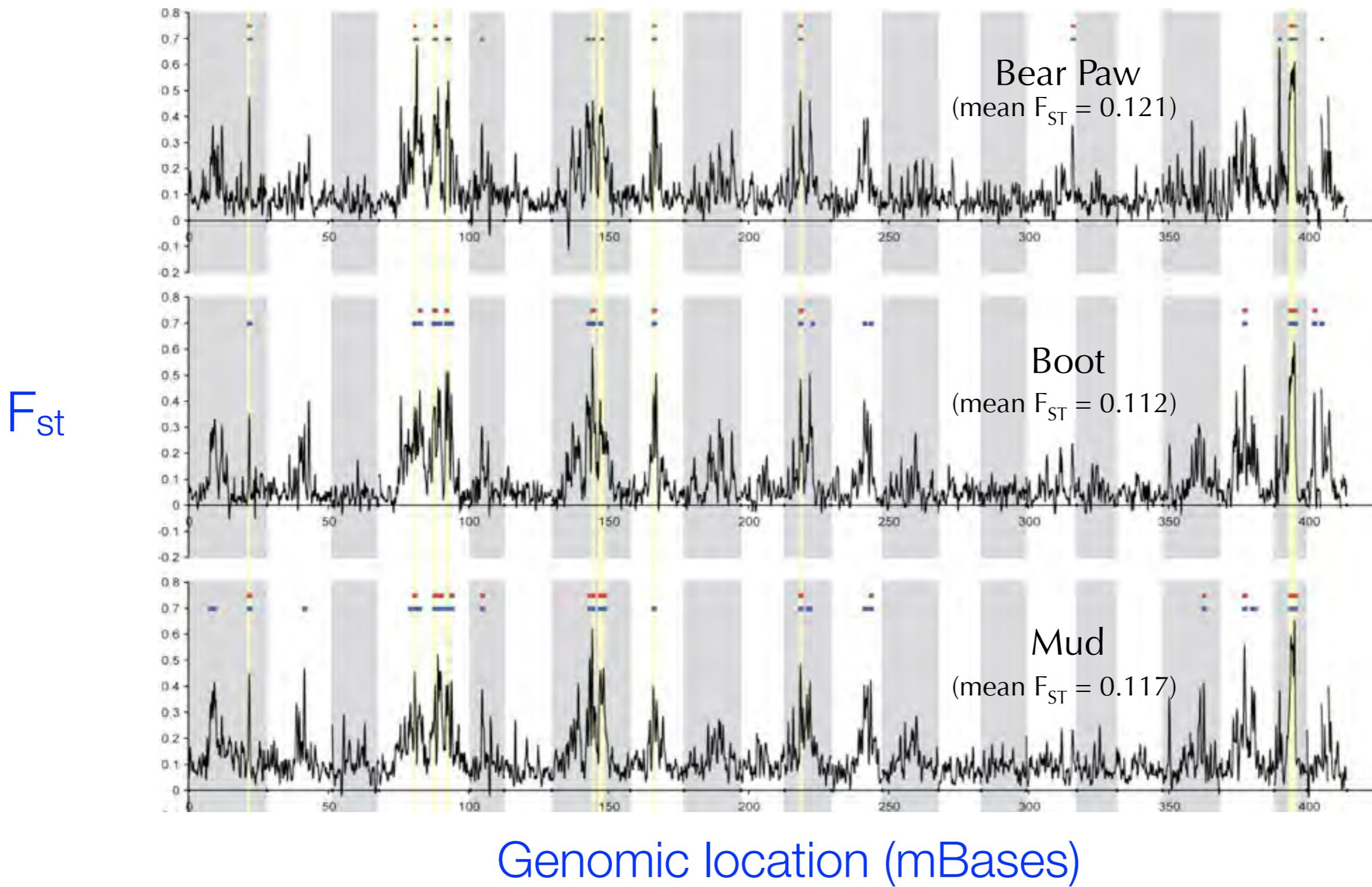


Frank
von Hippel

Signatures of natural selection in 13,000 years

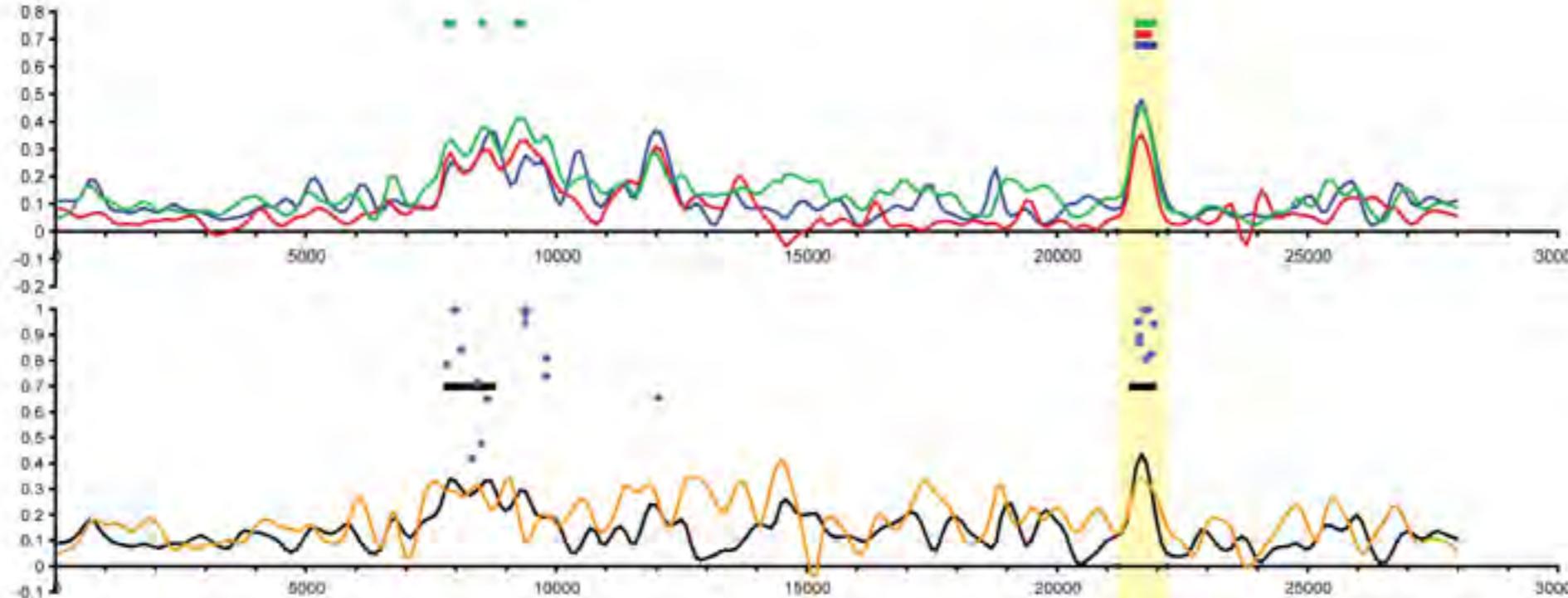


Signatures of natural selection in 13,000 years

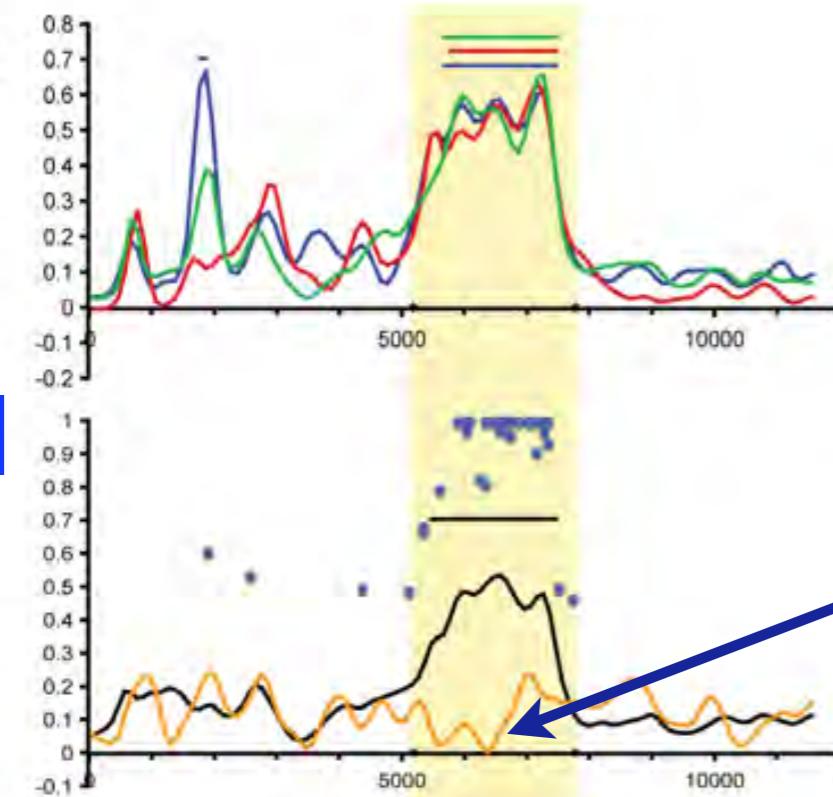


Numerous novel regions identified

LGI

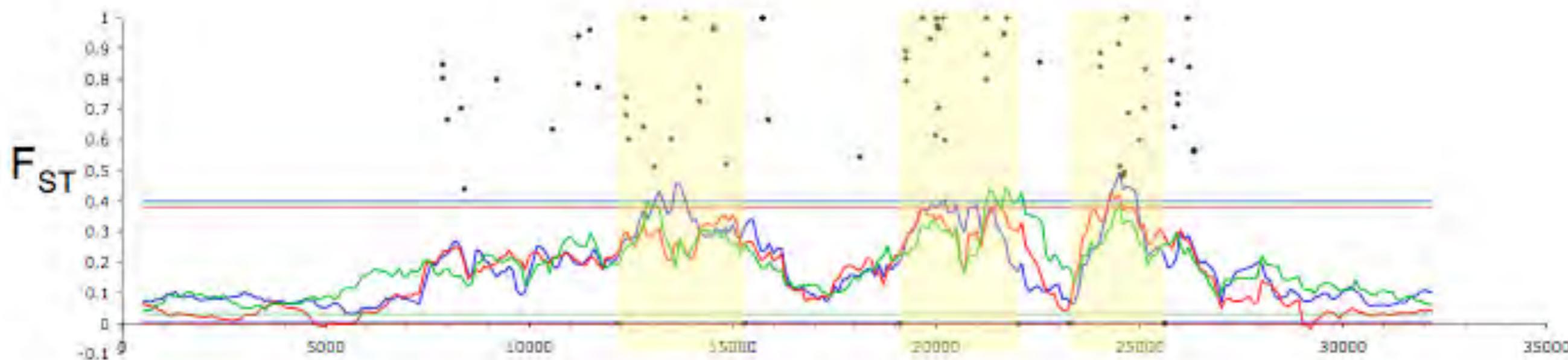


LGXXI



Some previously identify QTLs co-localize with peaks

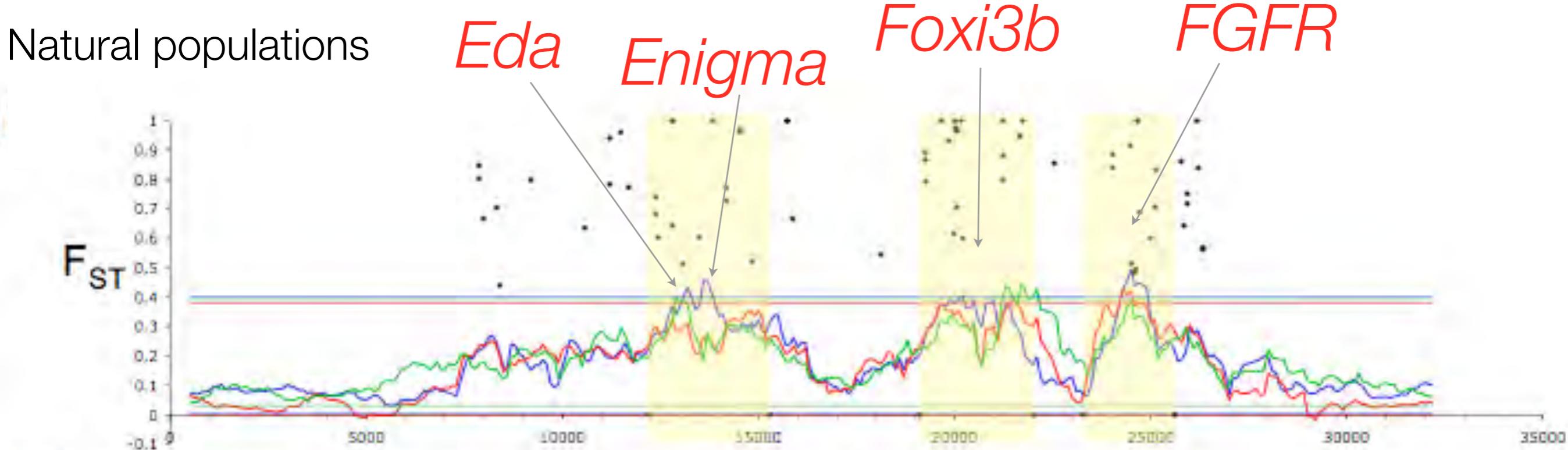
Natural populations



Lateral plate major locus
on LGIV (4000 SNPs)



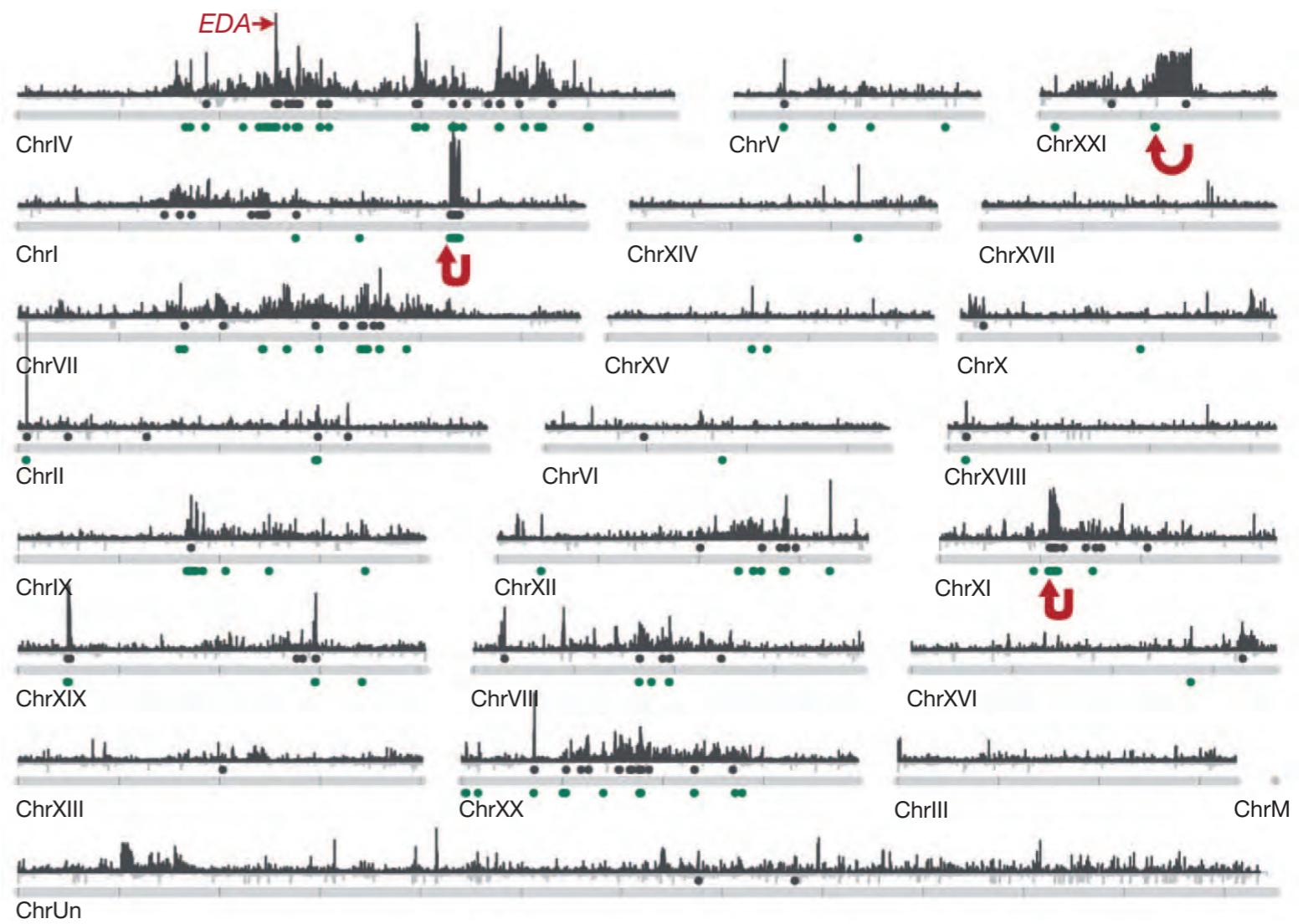
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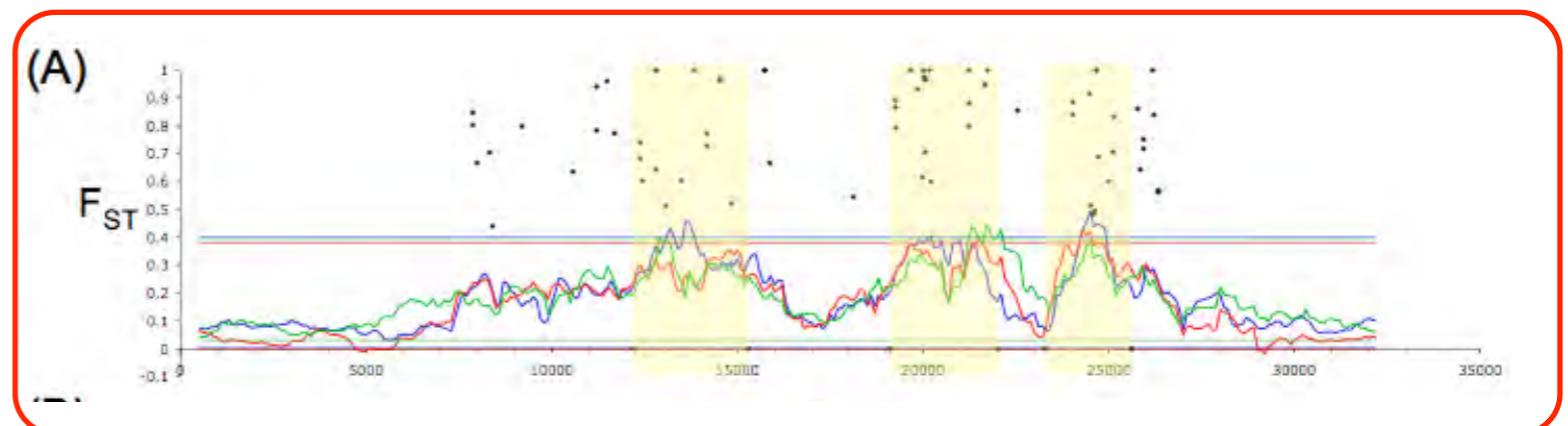
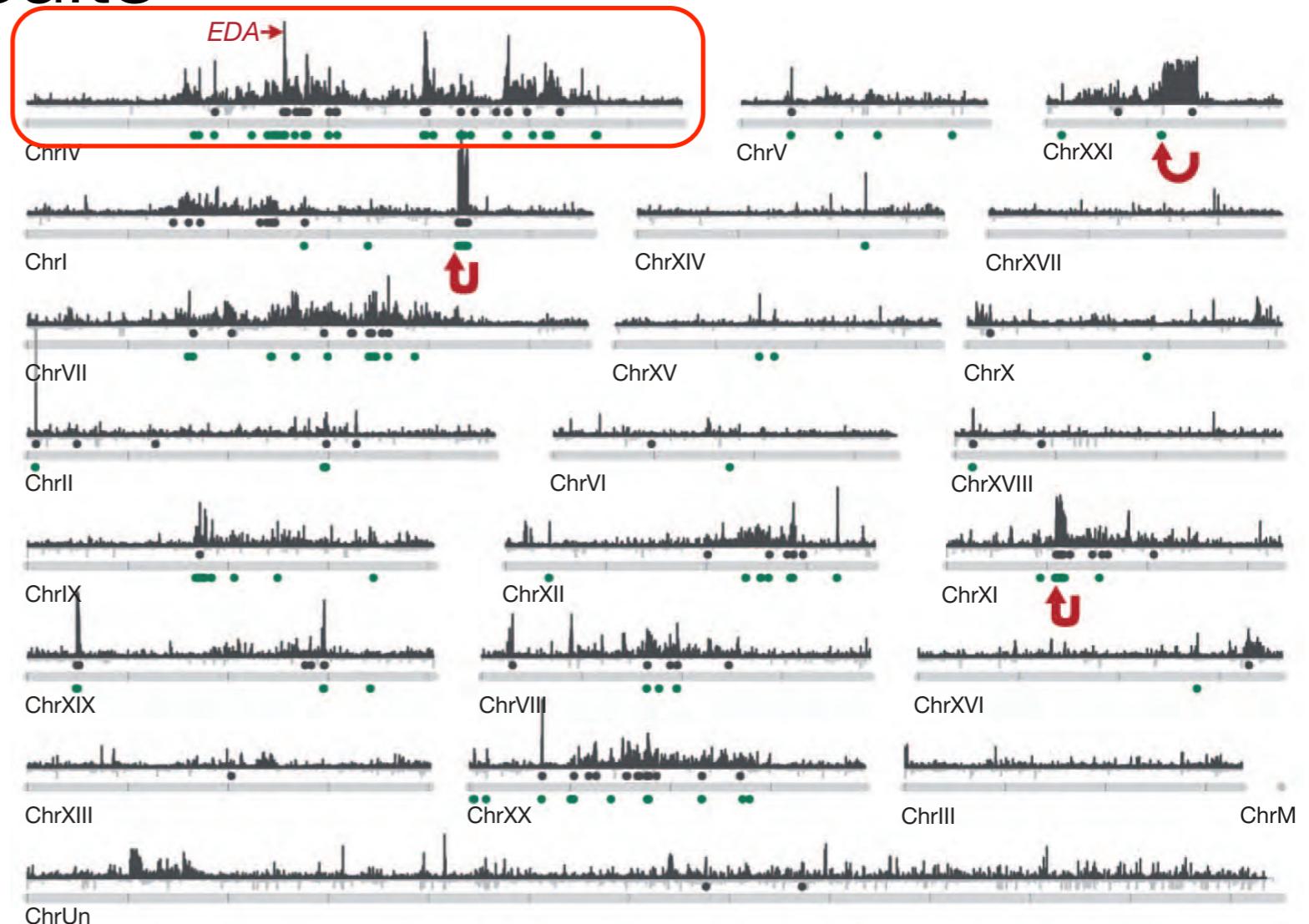
Lateral plate major locus
on LGIV (4000 SNPs)



Global analysis of complete sequencing consistent with the Alaskan results

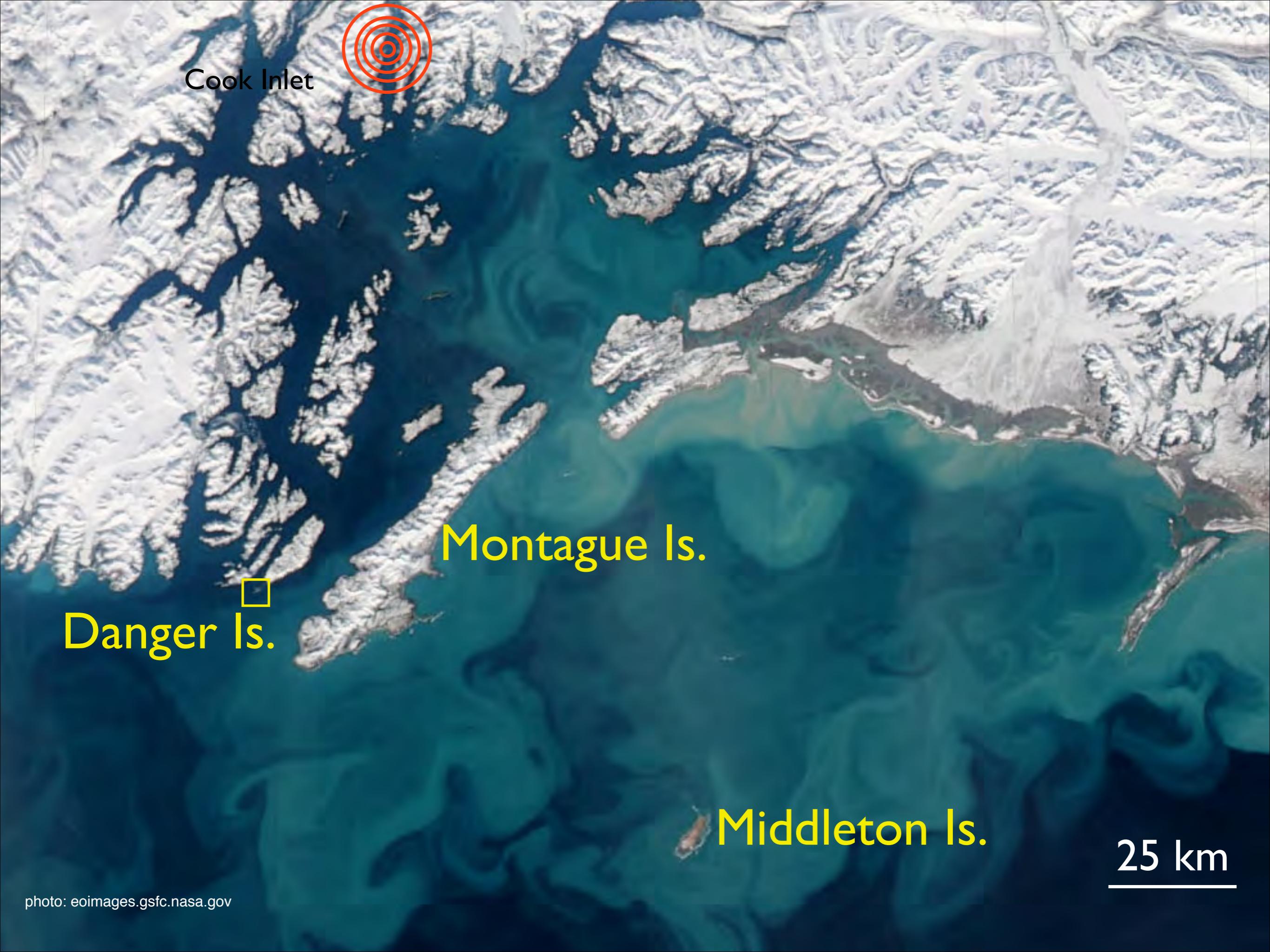


Global analysis of complete sequencing consistent with the Alaskan results



Intermediate conclusions

- Numerous locations throughout the stickleback genome are associated with differences between environments
- Some genomic regions are geographically localized, but many are shared across distant geographic regions
- These results point to segregating genetic variation as being important for rapid evolution
- Question - Can standing genetic and genomic variation allow extremely rapid evolution (<50 years)?



Cook Inlet



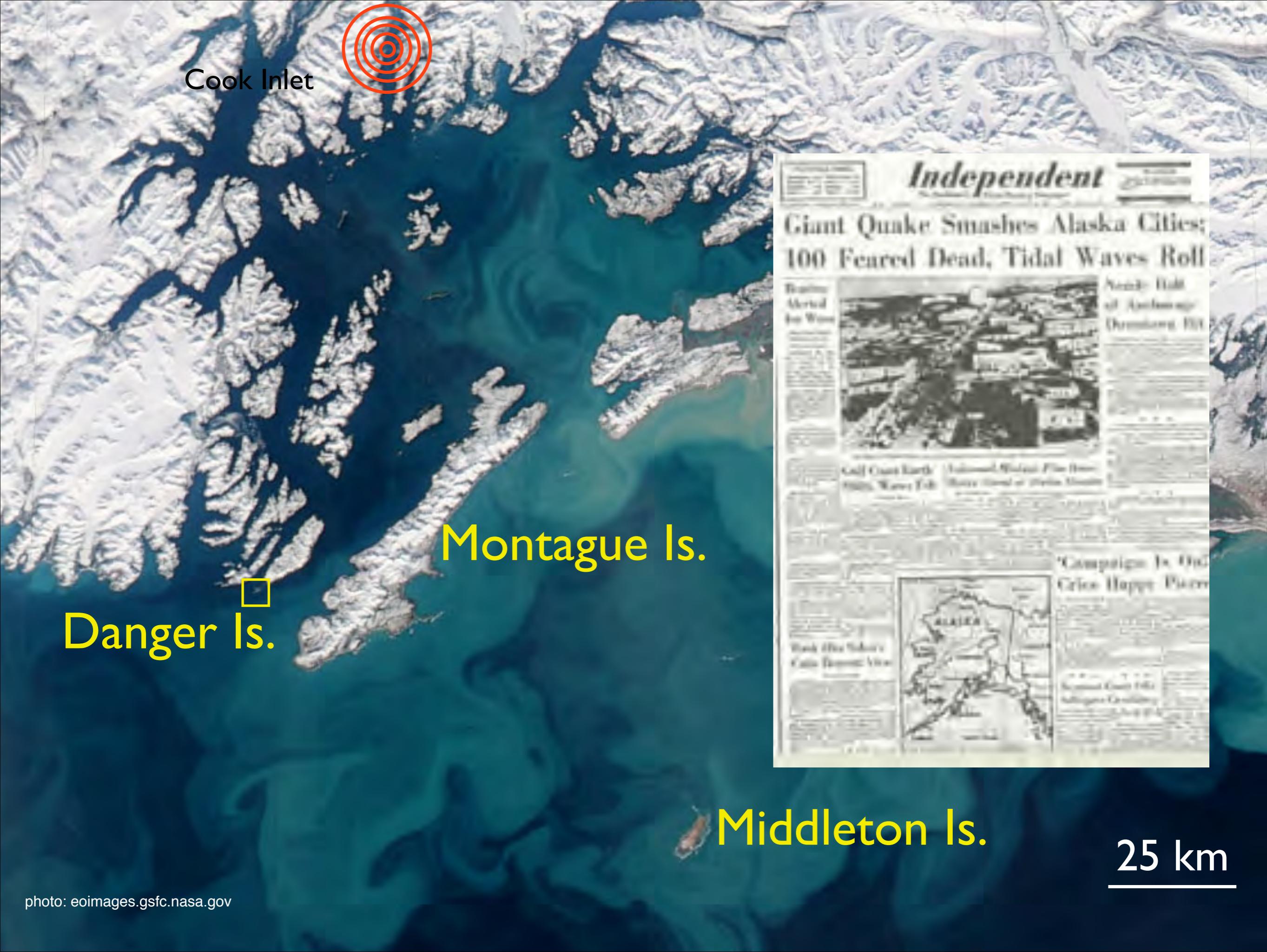
Danger Is.



Montague Is.

Middleton Is.

25 km





Middleton Island

1955



2008

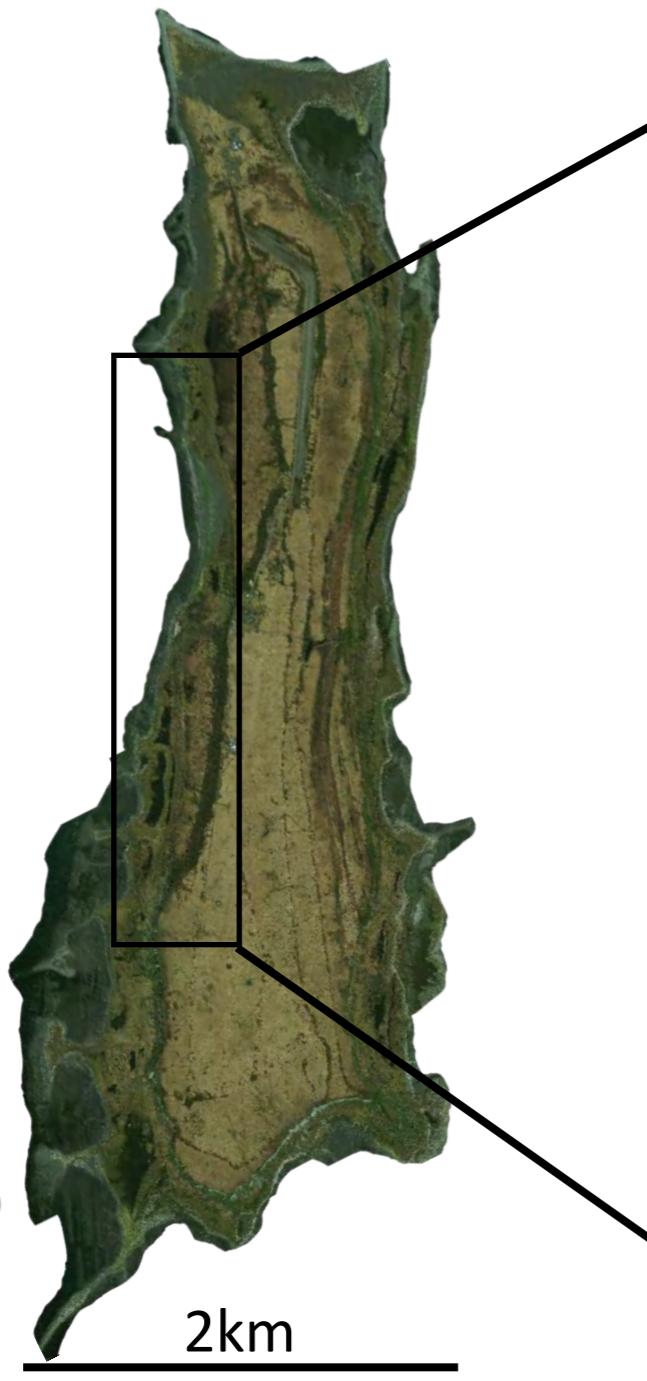


Middleton Island

1955



2008

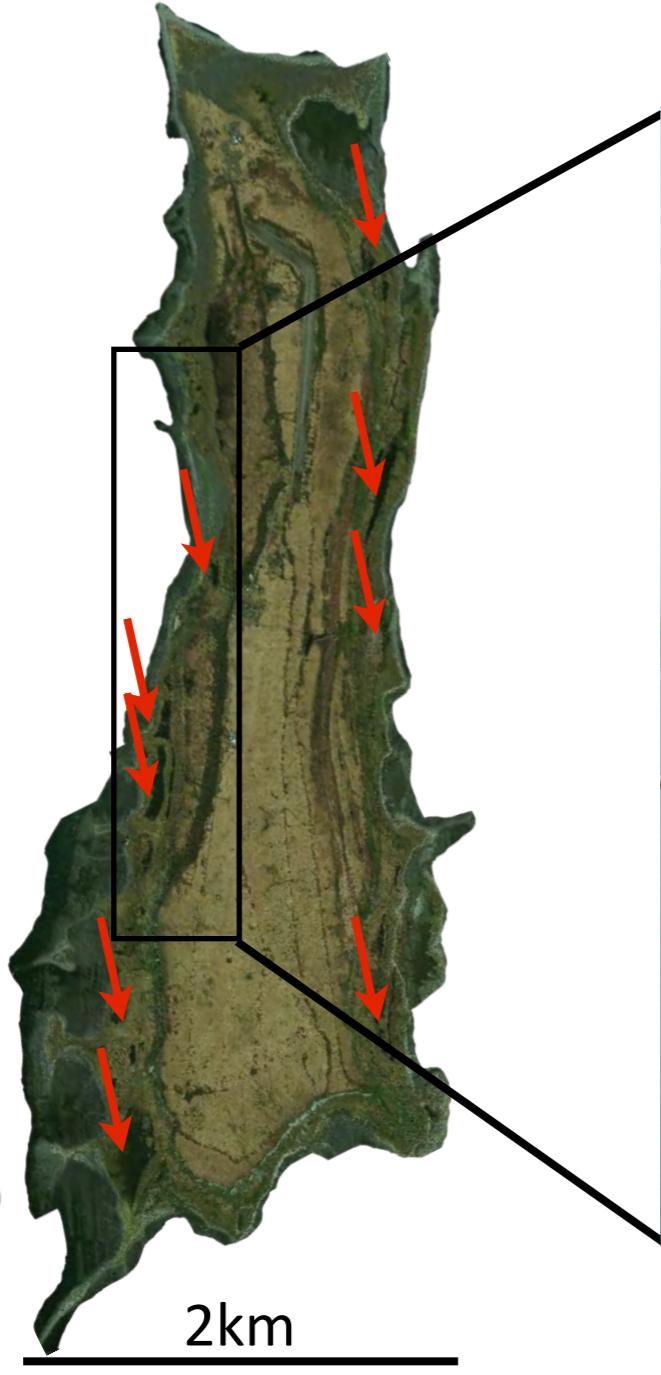


Middleton Island

1955



2008



2km



Tissue Collection and Preservation



Caudal and pectoral
fins clipped for
DNA extraction



Bodies fixed in
formalin, bleached,
stained



Mary Sherick

Tissue Collection and Preservation



Caudal and pectoral
fins clipped for
DNA extraction



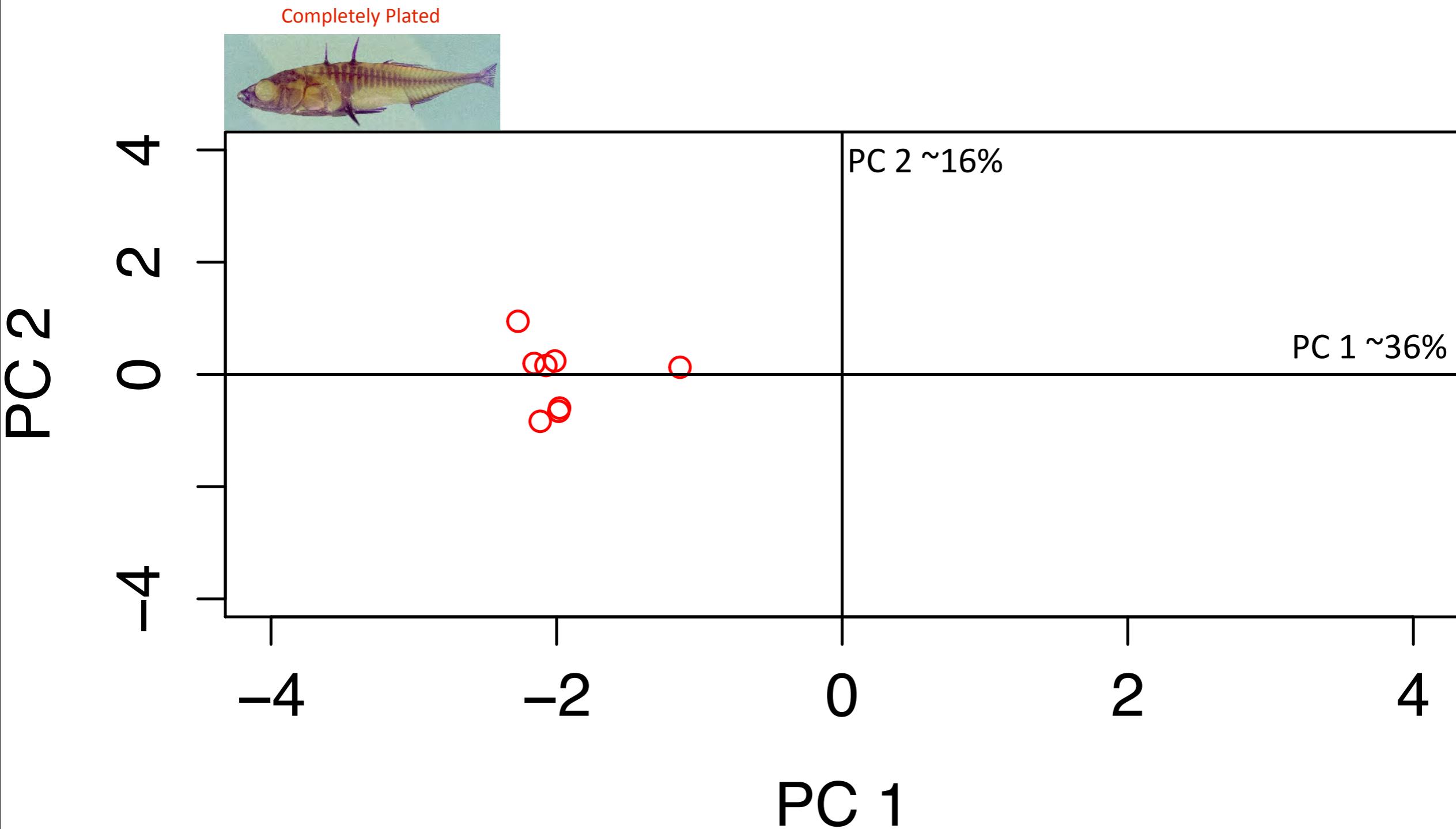
Bodies fixed in
formalin, bleached,
stained

110,000 SNPs per individual
>1000 Individuals
20 million genotypes

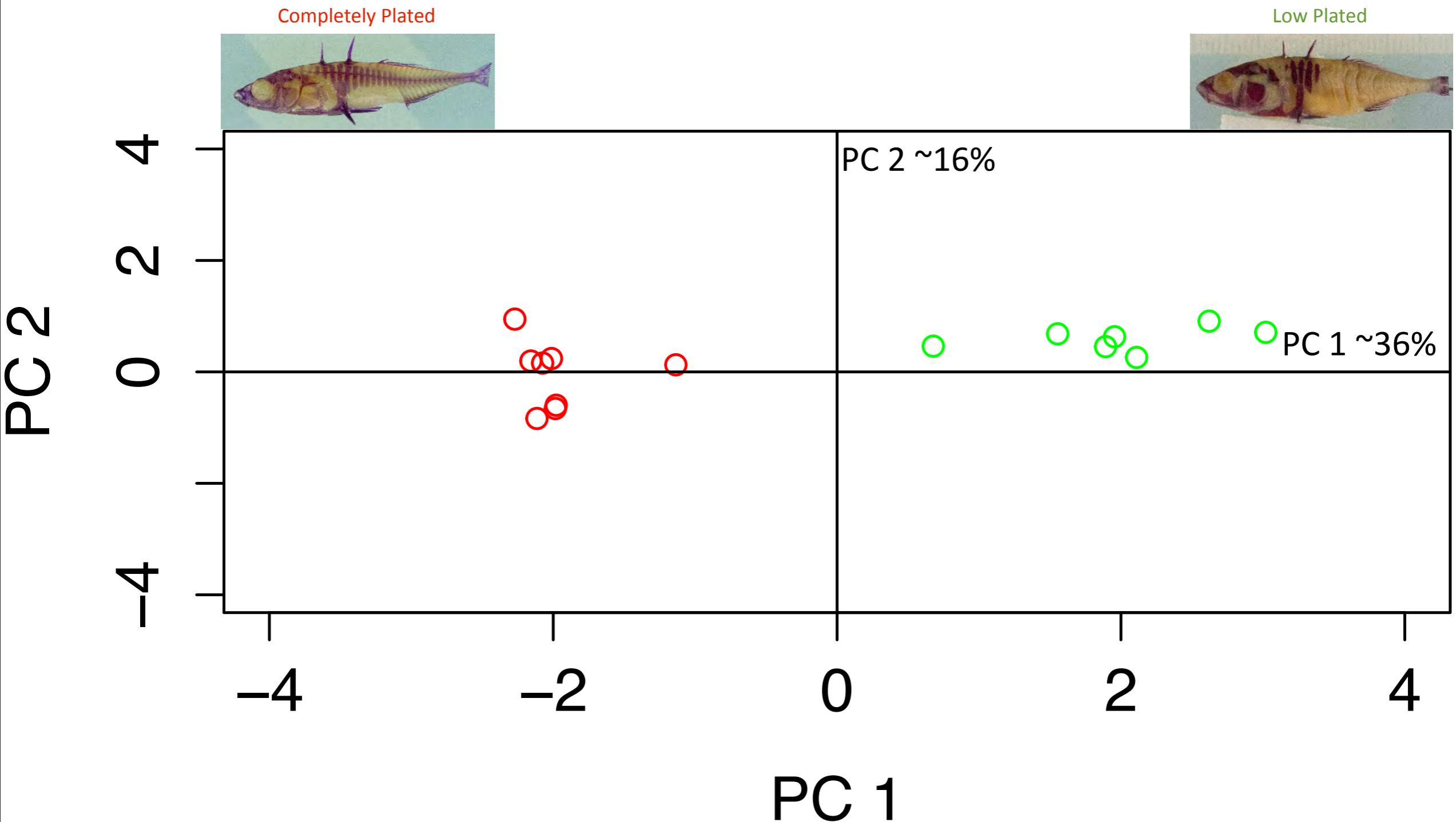
Mary Sherick



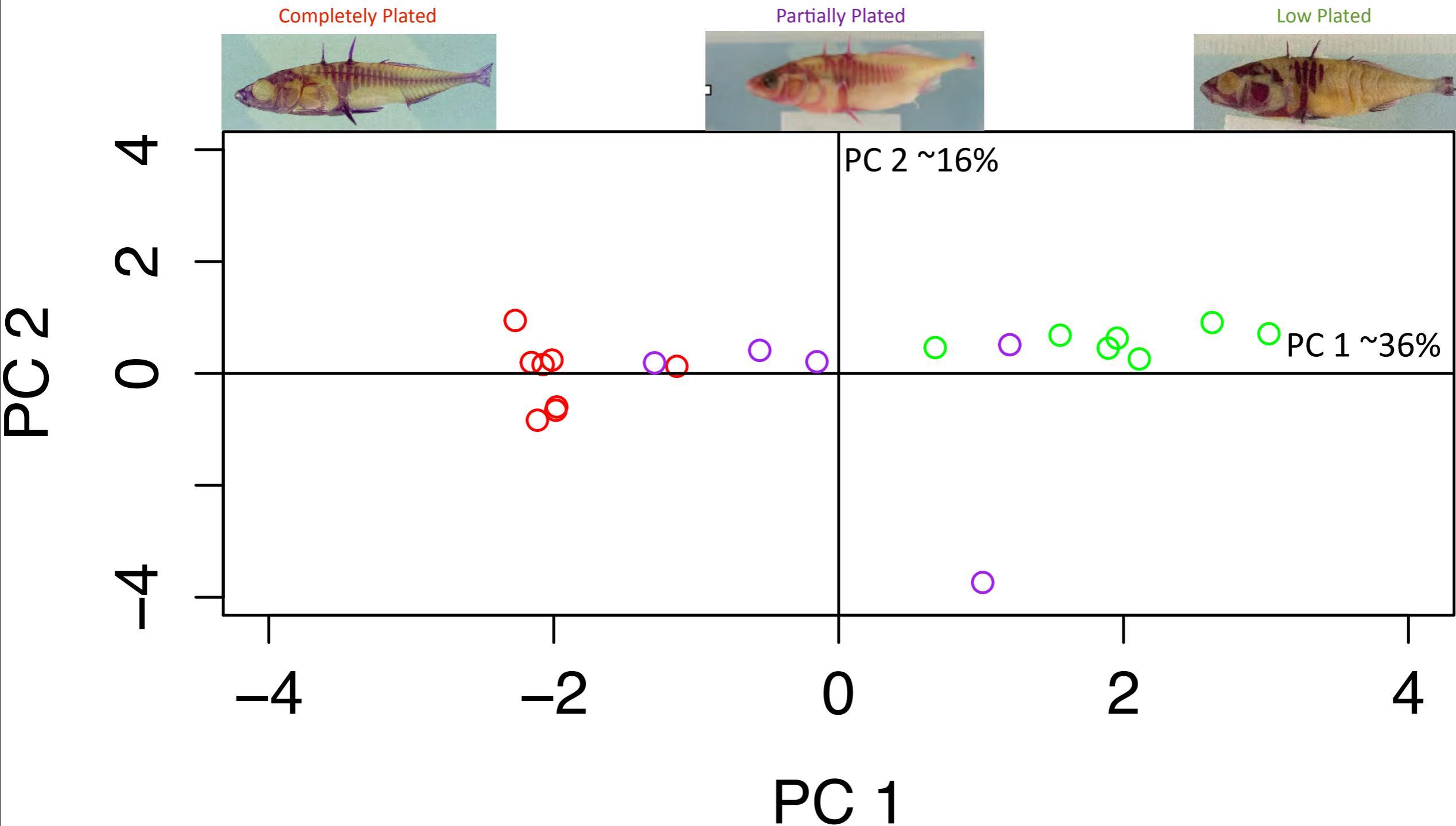
PCA of overall genetic variation



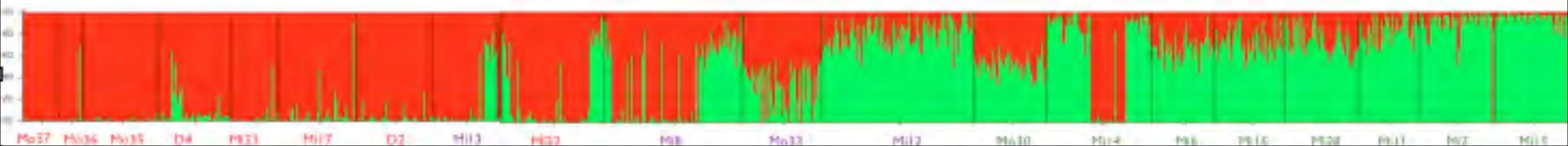
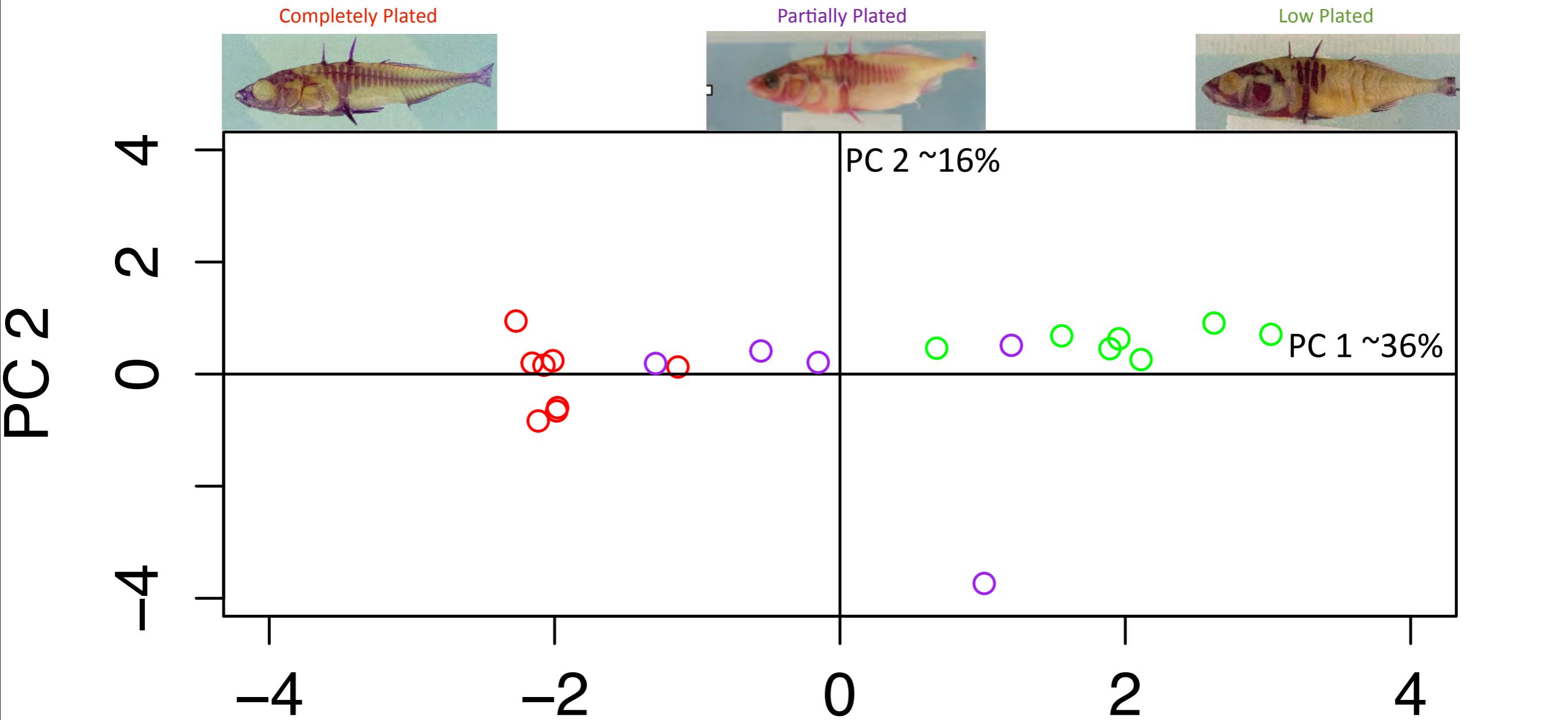
PCA of overall genetic variation



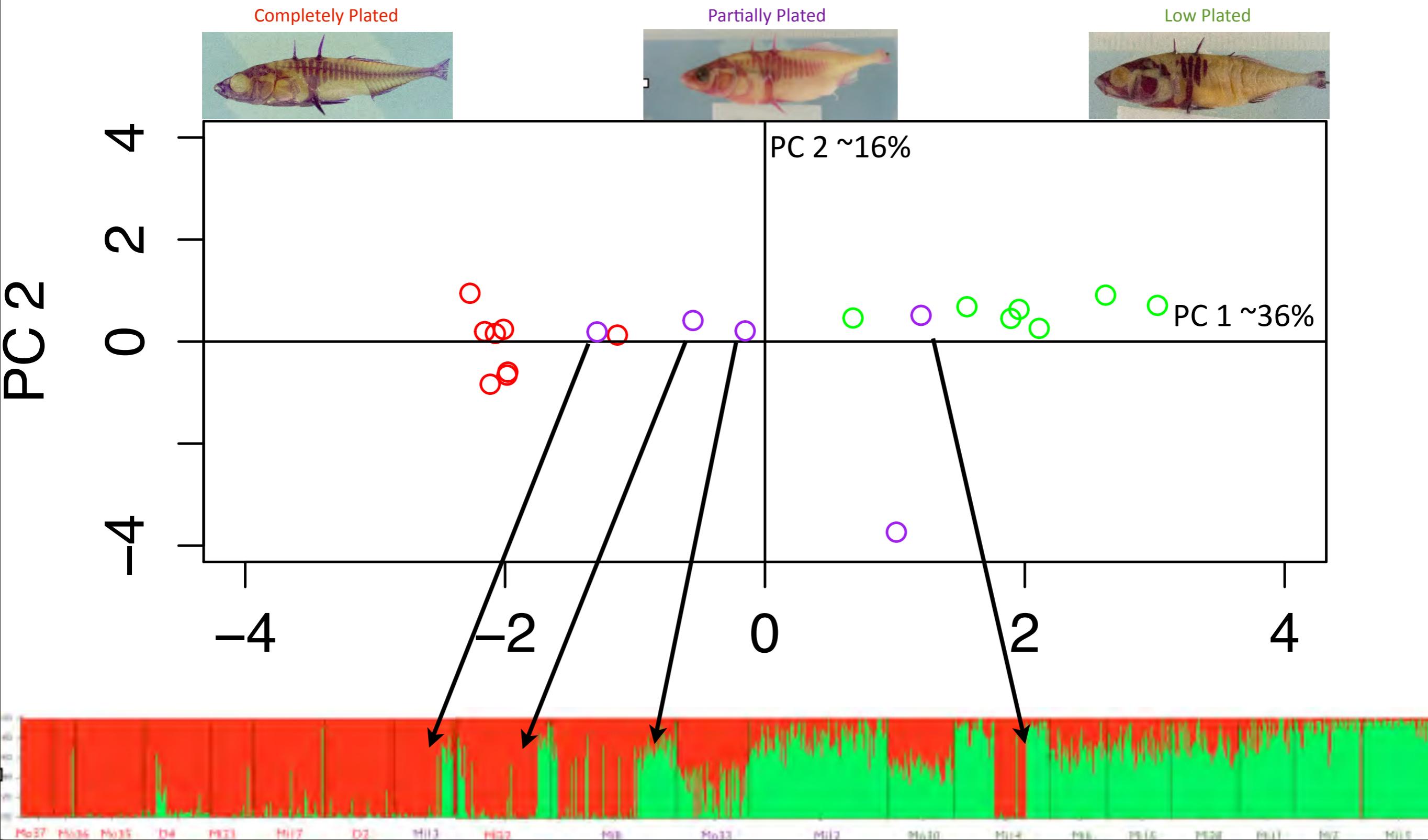
PCA of overall genetic variation



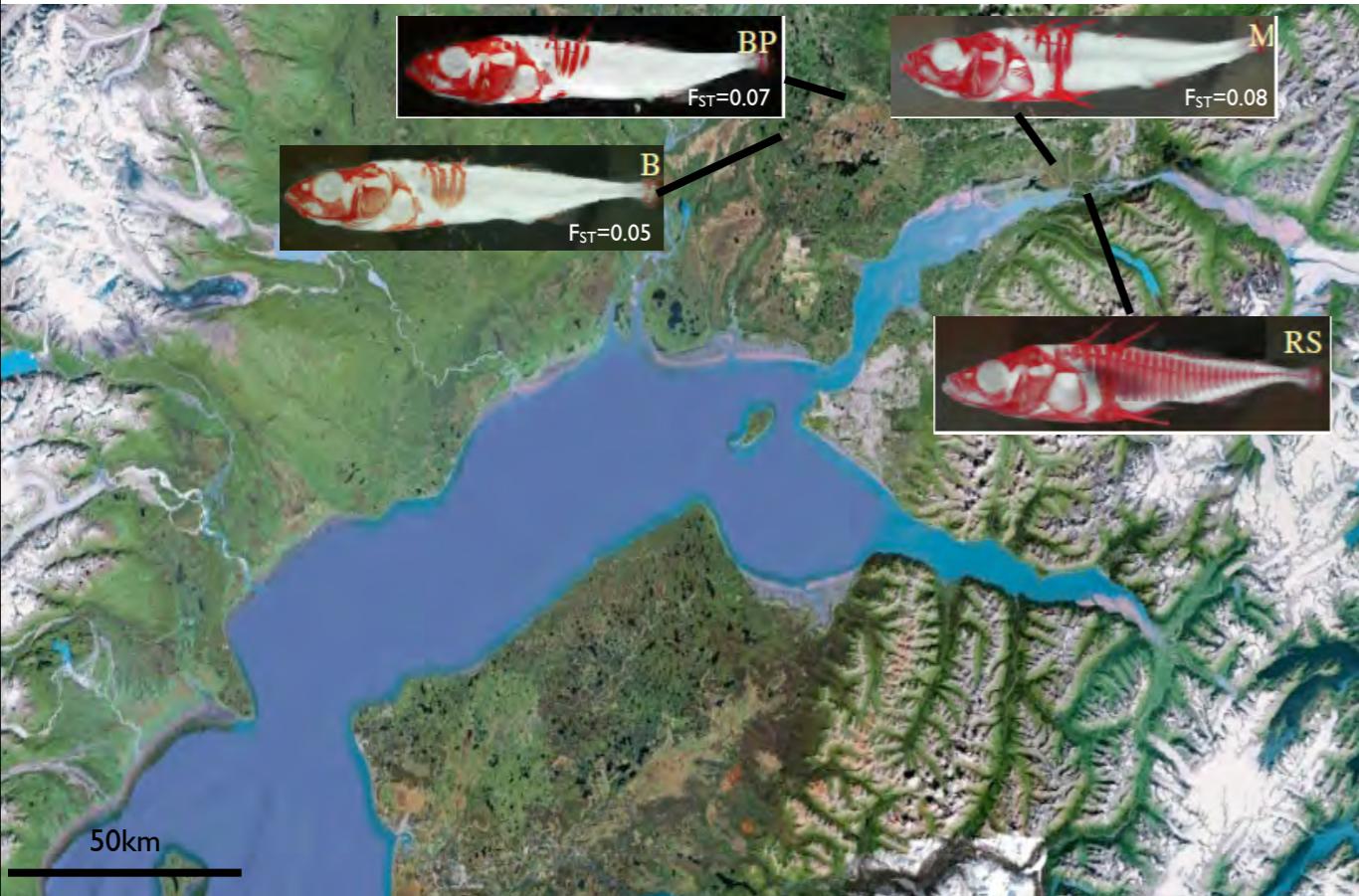
PCA of overall genetic variation



PCA of overall genetic variation



~13,000 Years

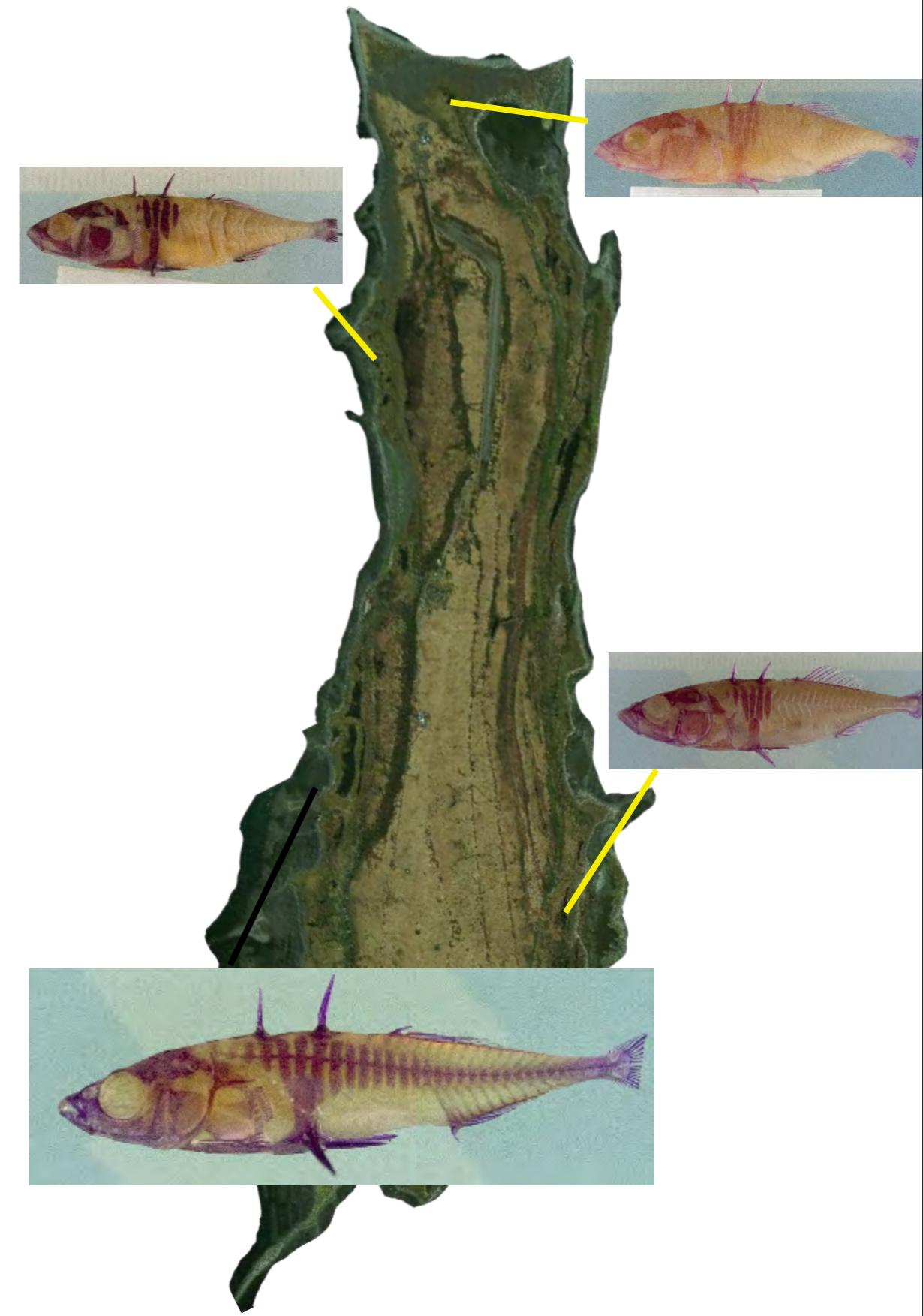


Hohenlohe et al. 2010

~13,000 Years



~50 Years



~13,000 Years



Hohenlohe et al. 2010

~50 Years

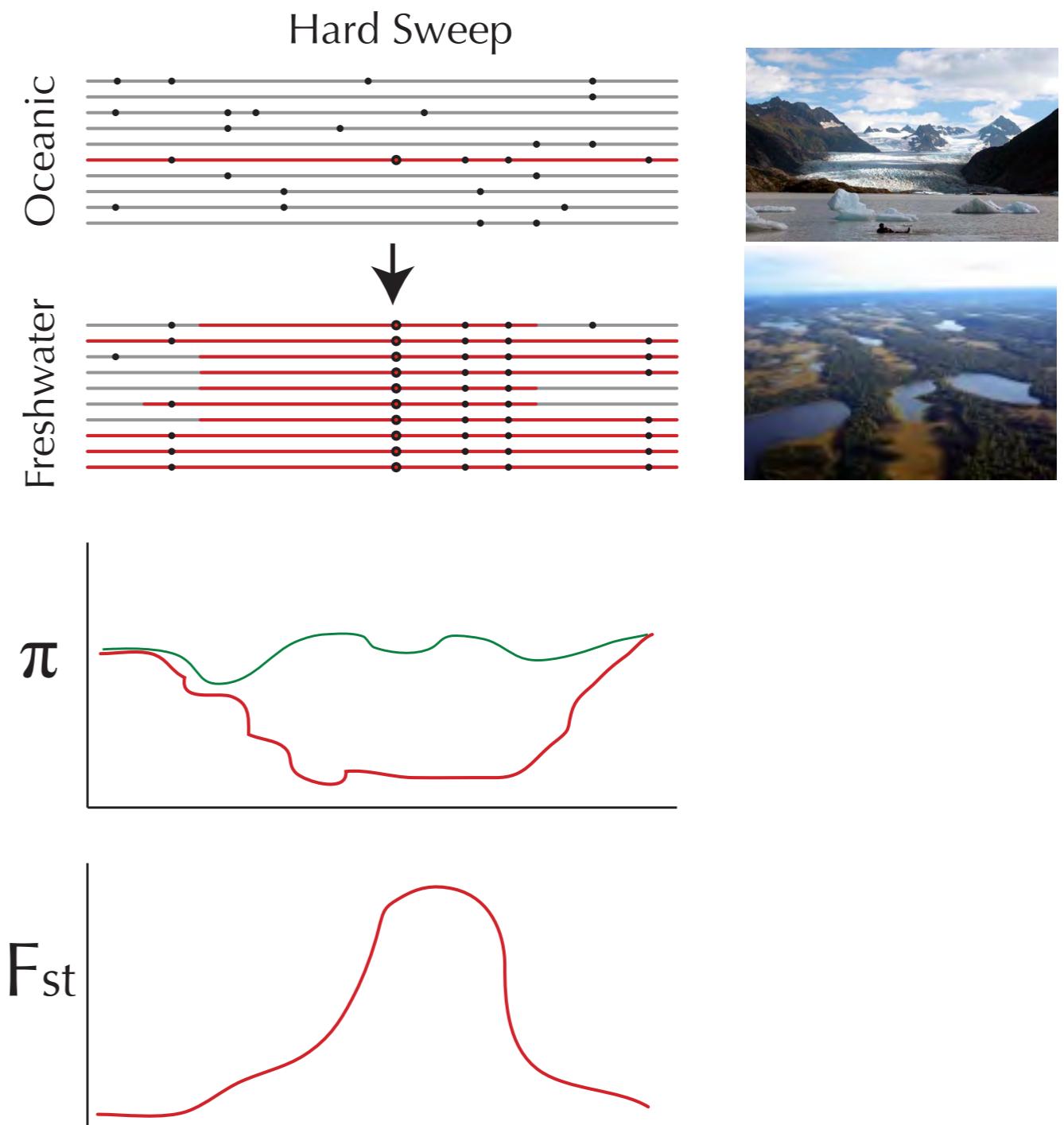


Replicated, independent divergence on two time scales

Cresko et al. 2004

What are the signatures of selection in 50 years across the genome?

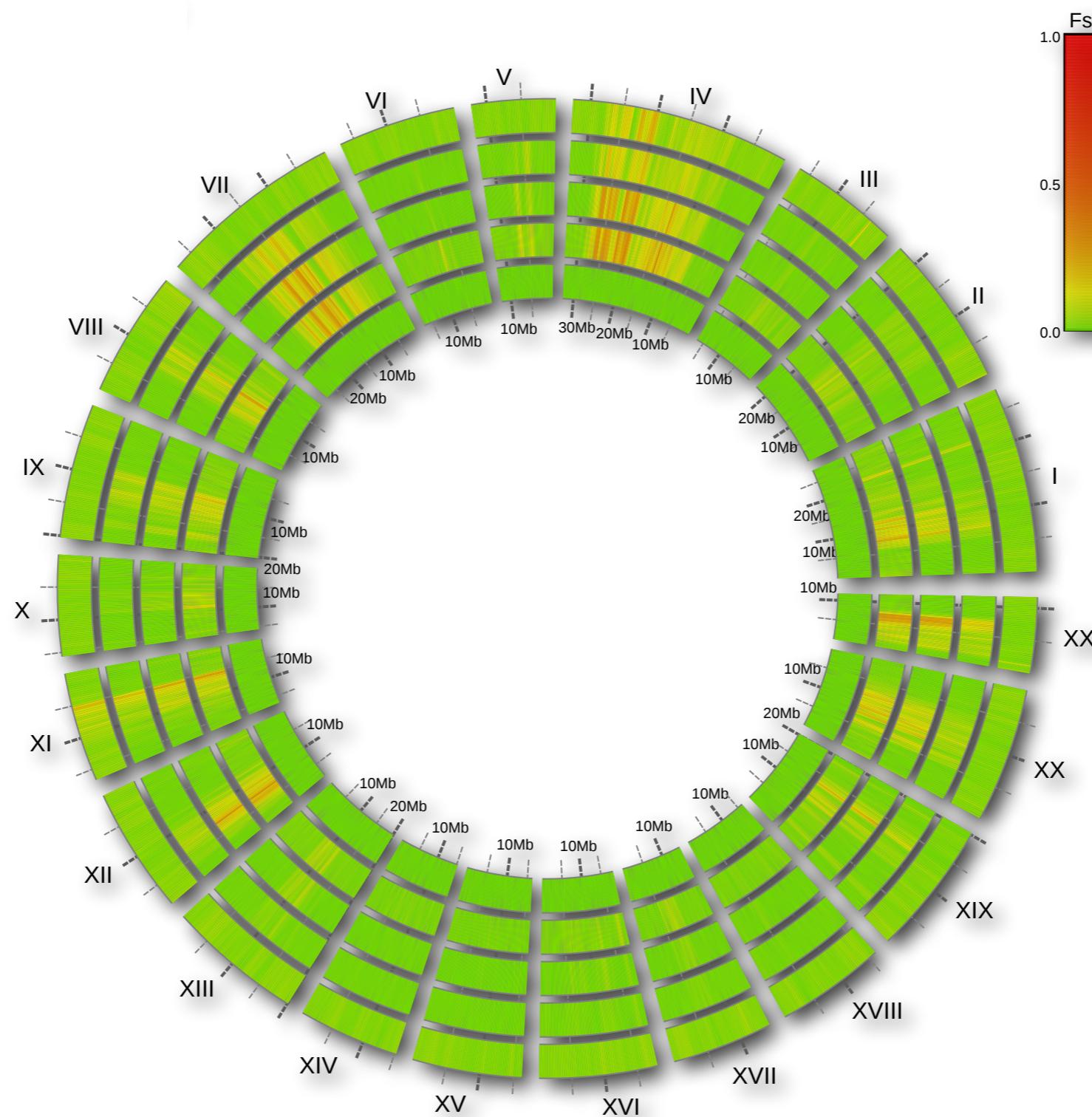
Source of Variation	% Variation
Within Individual	76.4
Among Island	5.8
Fresh vs Salt Water	2.6



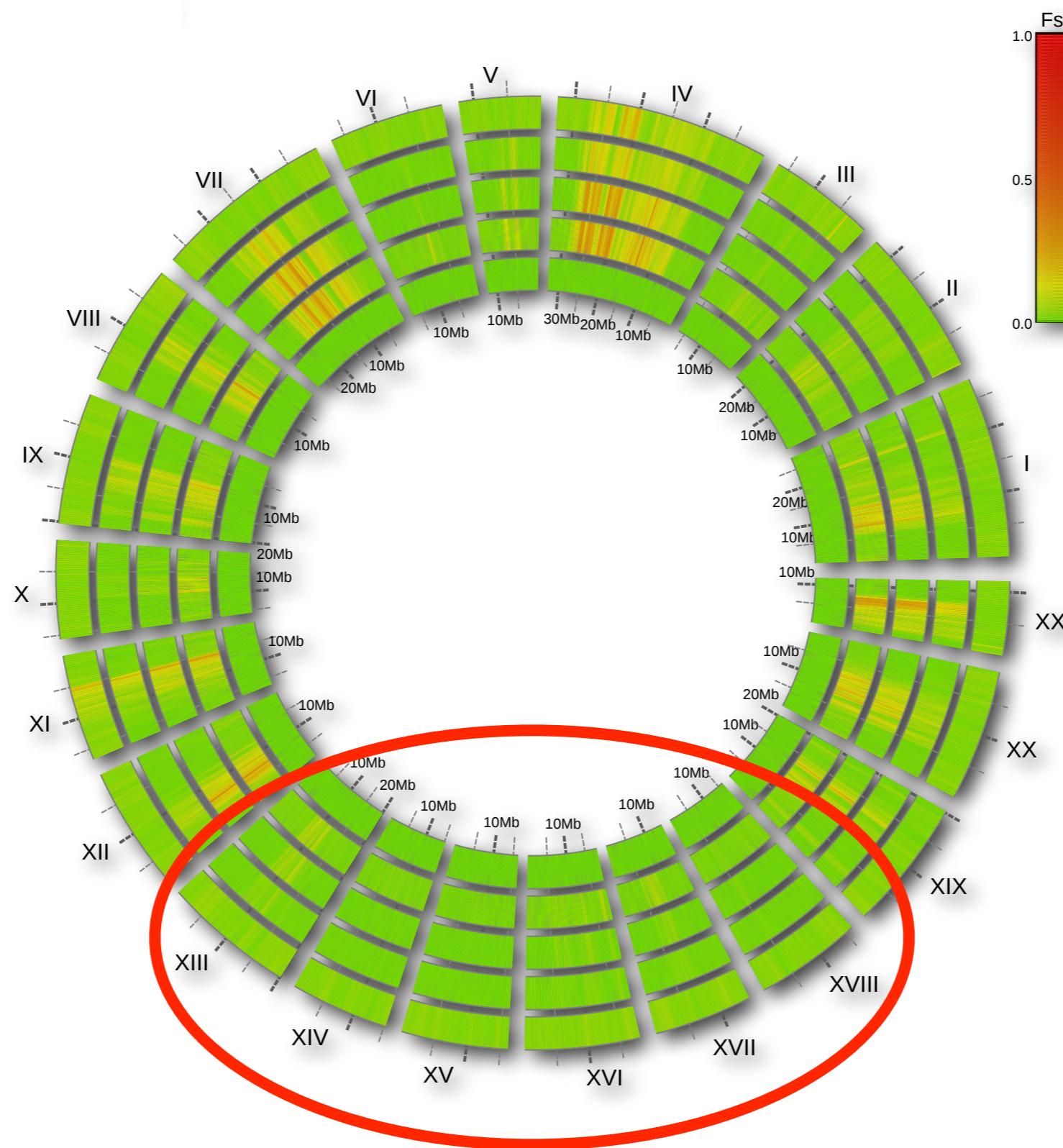
Interpretation?! How do we visualize the results?

	Middleton Island Site 06	Middleton Island Site 07	Middleton Island Site 17	Middleton Island Site 08	Middleton Island Site 11	Middleton Island Site 12	Middleton Island Site 13	Middleton Island Site 14	Middleton Island Site 15	Middleton Island Site 16	Middleton Island Site 22	Middleton Island Site 23	Middleton Island Site 28	Montague Island Site 35	Montague Island Site 36	Montague Island Site 37	Millport Slough	Upper Fire Lake	Danger Island Site 02	Montague Island Site 30	Montague Island Site 33
Danger Island Site 04																					
Middleton Island Site 06																					
Middleton Island Site 07																					
Middleton Island Site 17																					
Middleton Island Site 08																					
Middleton Island Site 11																					
Middleton Island Site 12																					
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Middleton Island Site 28																					
Montague Island Site 35																					
Montague Island Site 36																					
Montague Island Site 37																					
Millport Slough																					
Upper Fire Lake																					
Danger Island Site 02																					
Montague Island Site 30																					

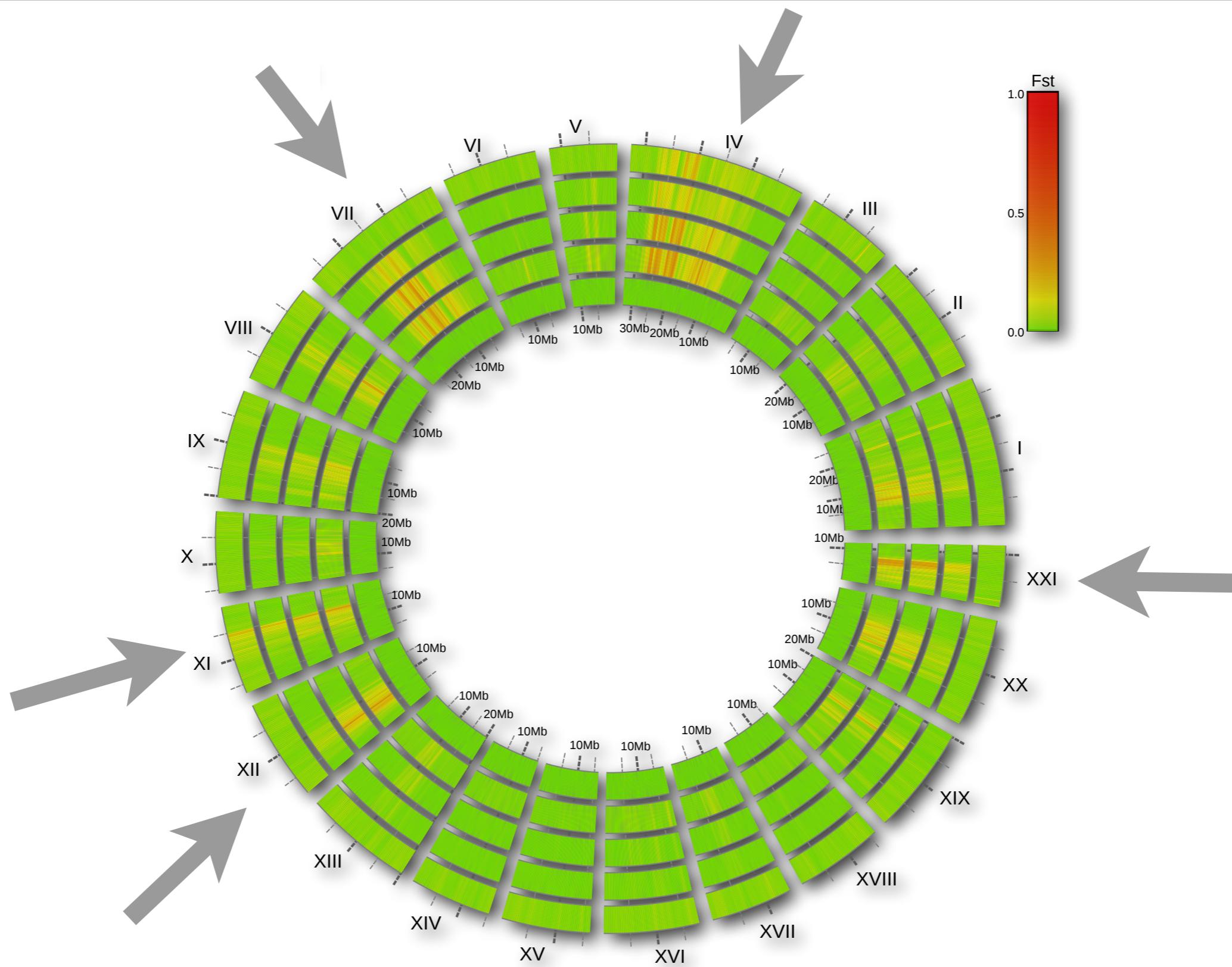
Ocean vs. Freshwater Genomic Comparison



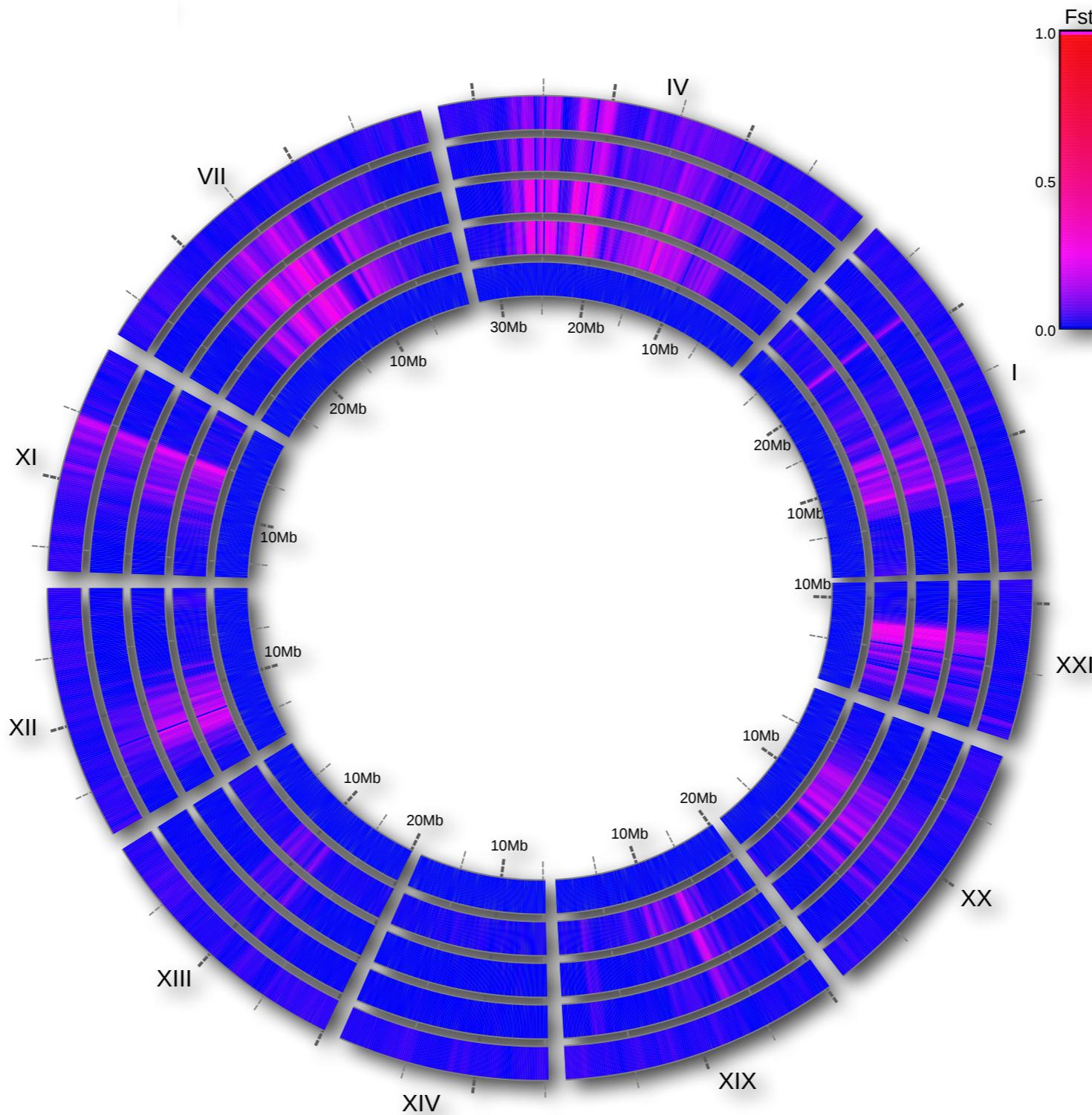
Ocean vs. Freshwater Genomic Comparison



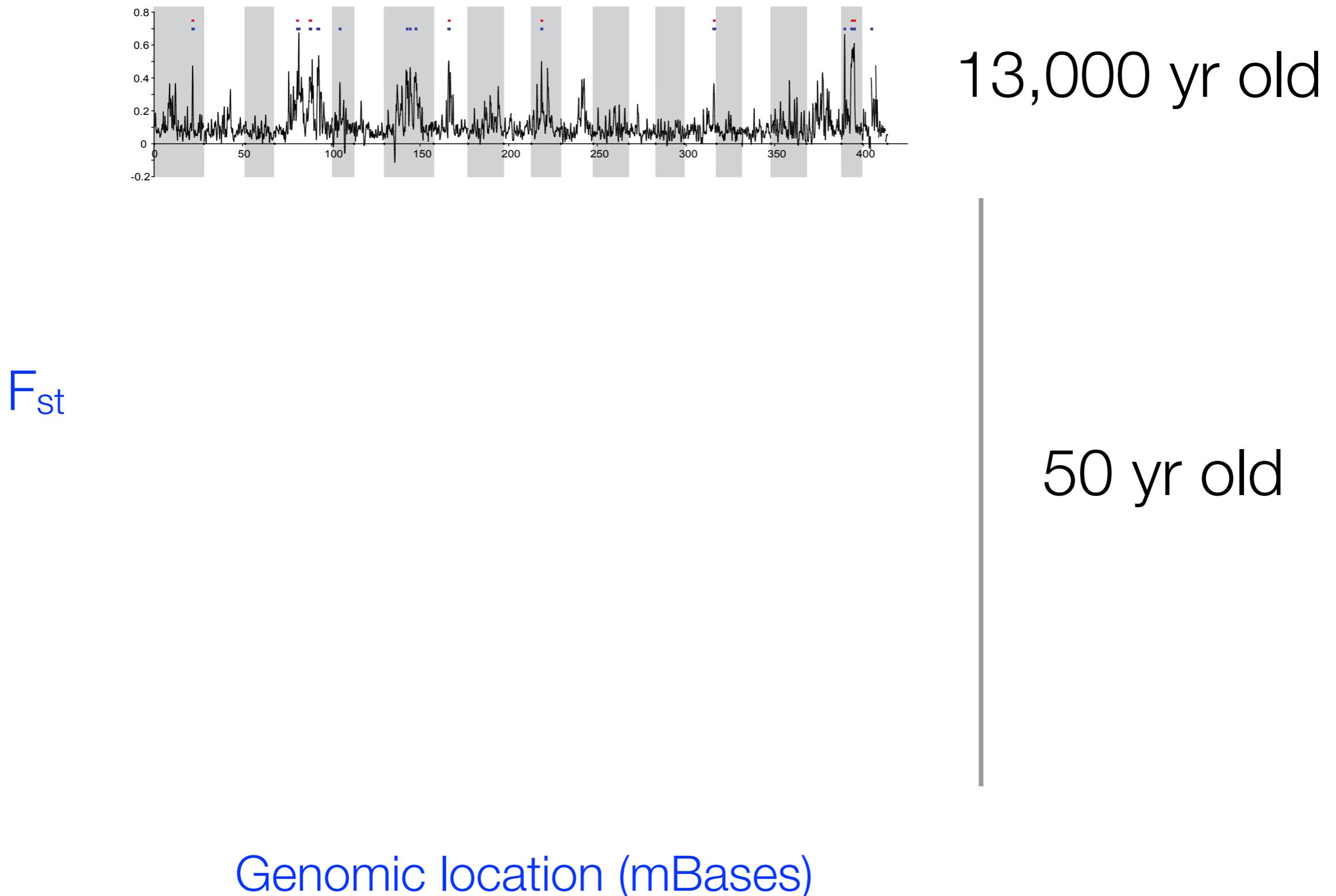
Ocean vs. Freshwater Genomic Comparison



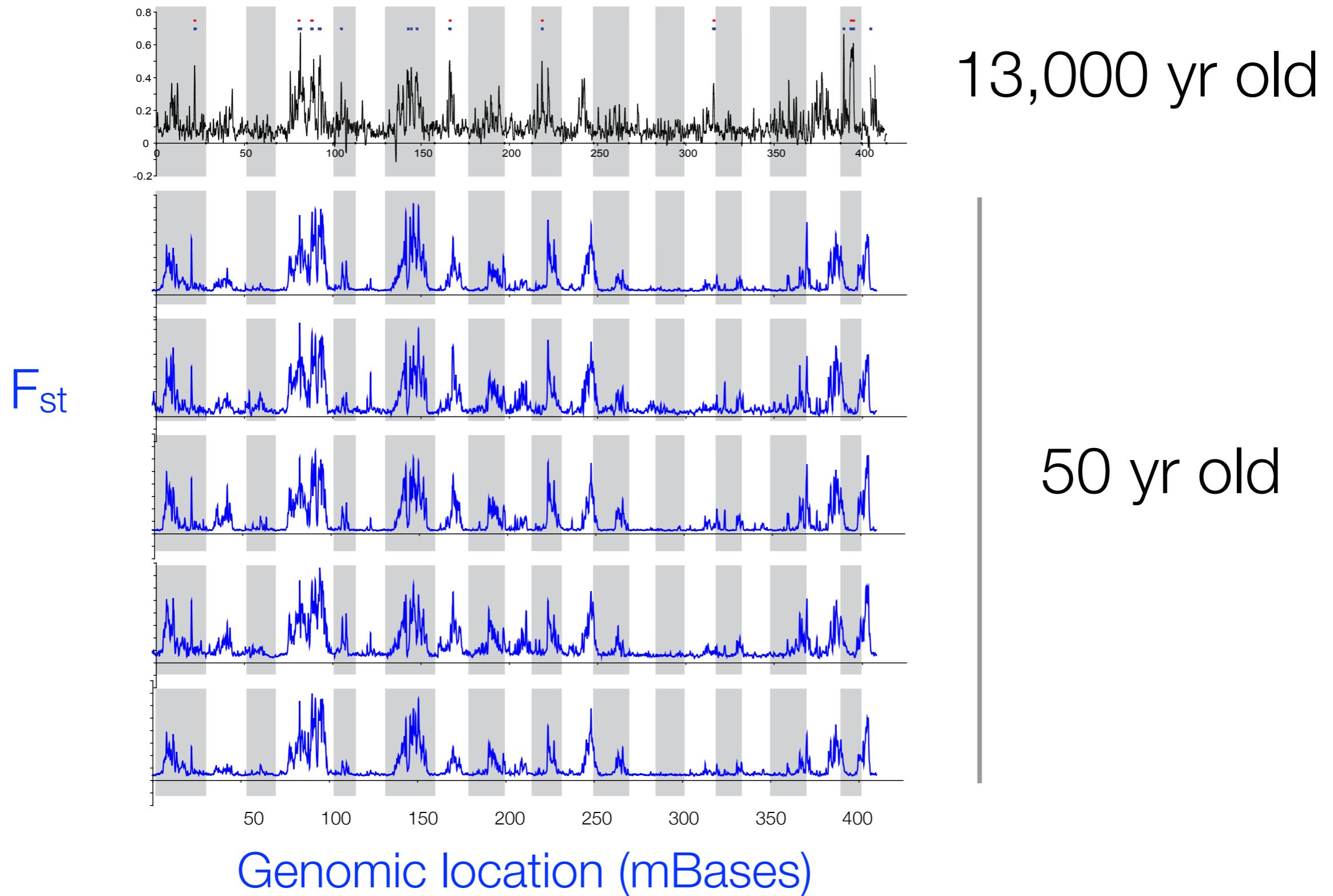
Ocean vs. Freshwater Genomic Comparison



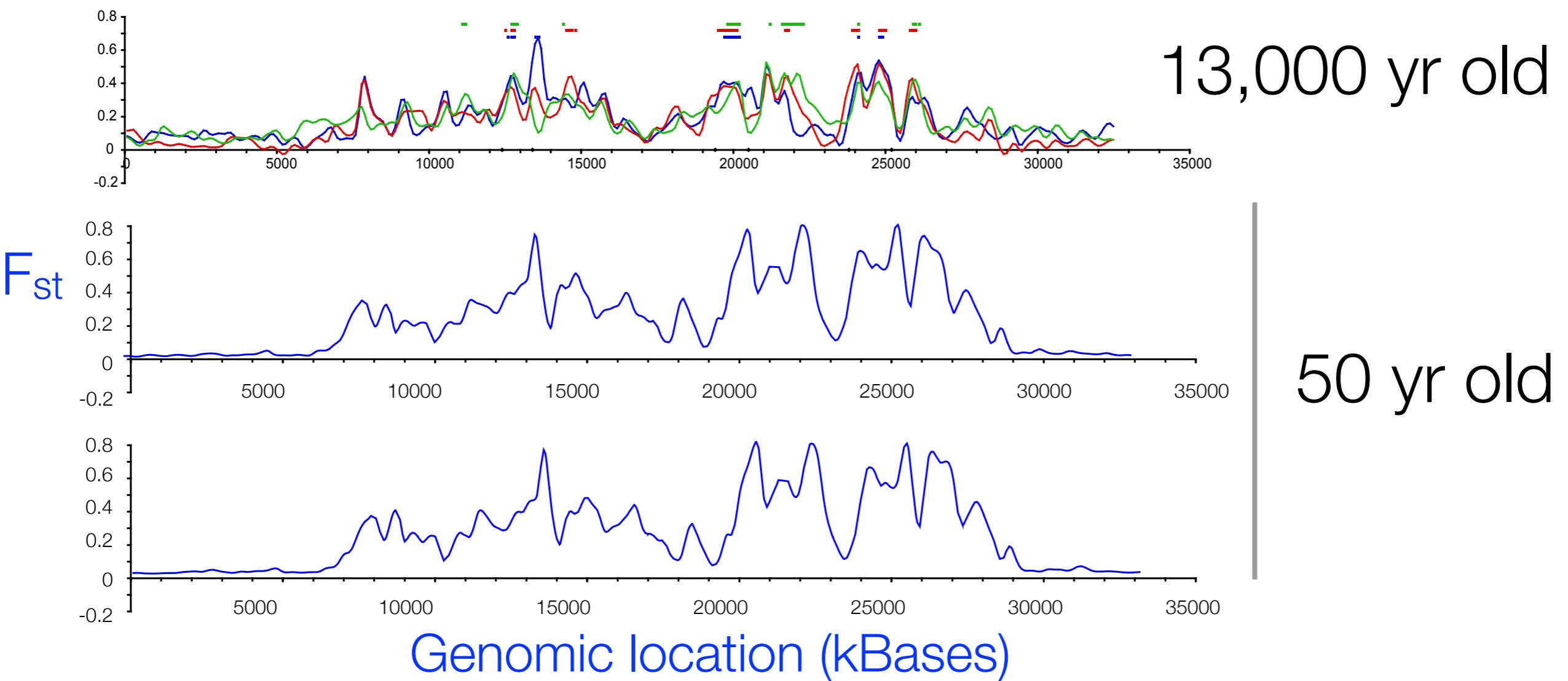
Ocean vs. Freshwater Genomic Comparison



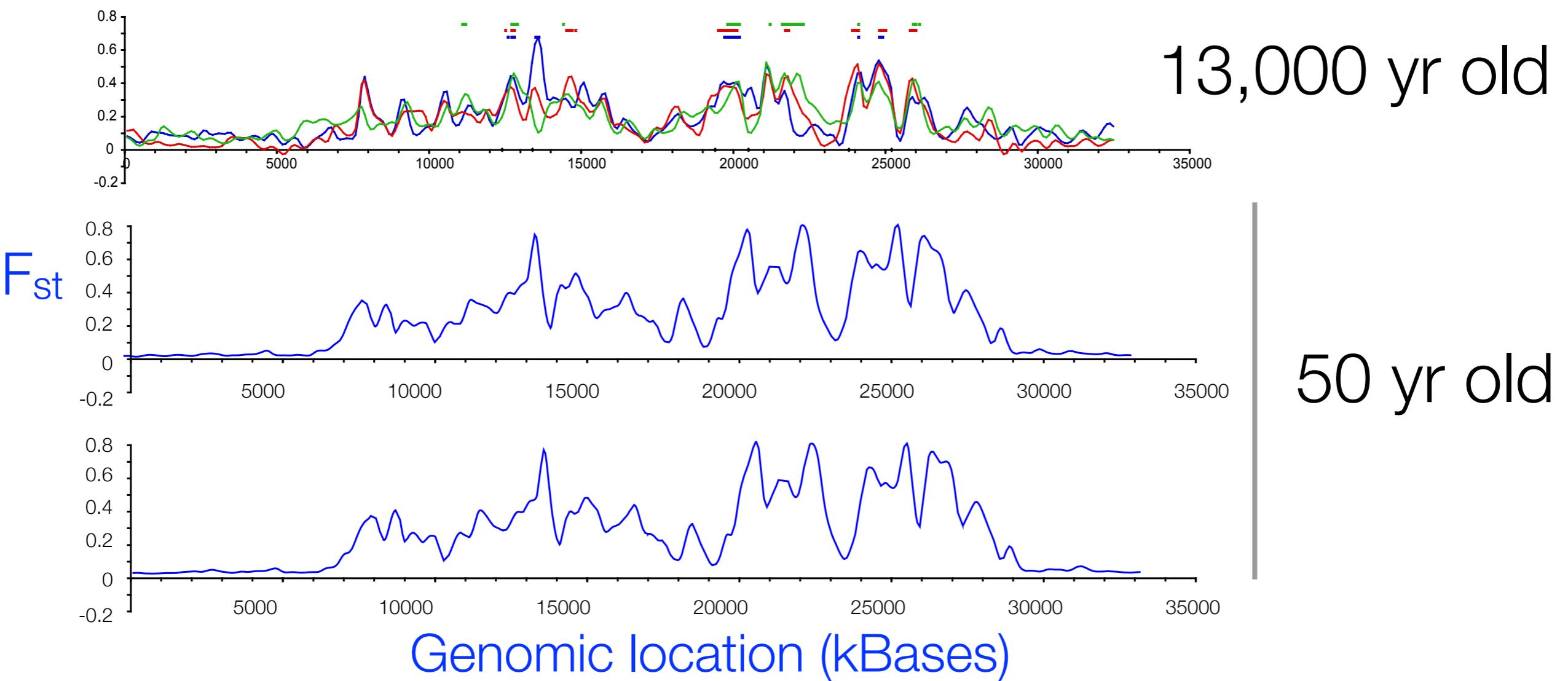
Ocean vs. Freshwater Genomic Comparison



Linkage Group IV comparison

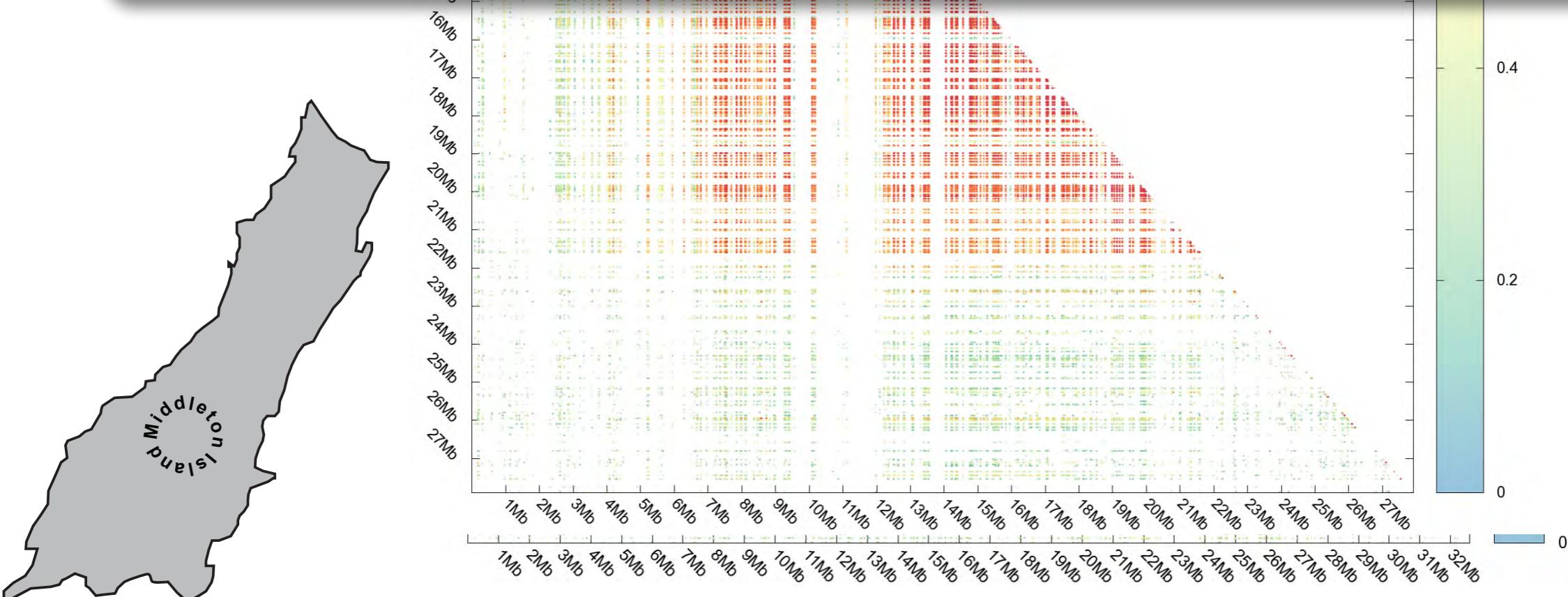
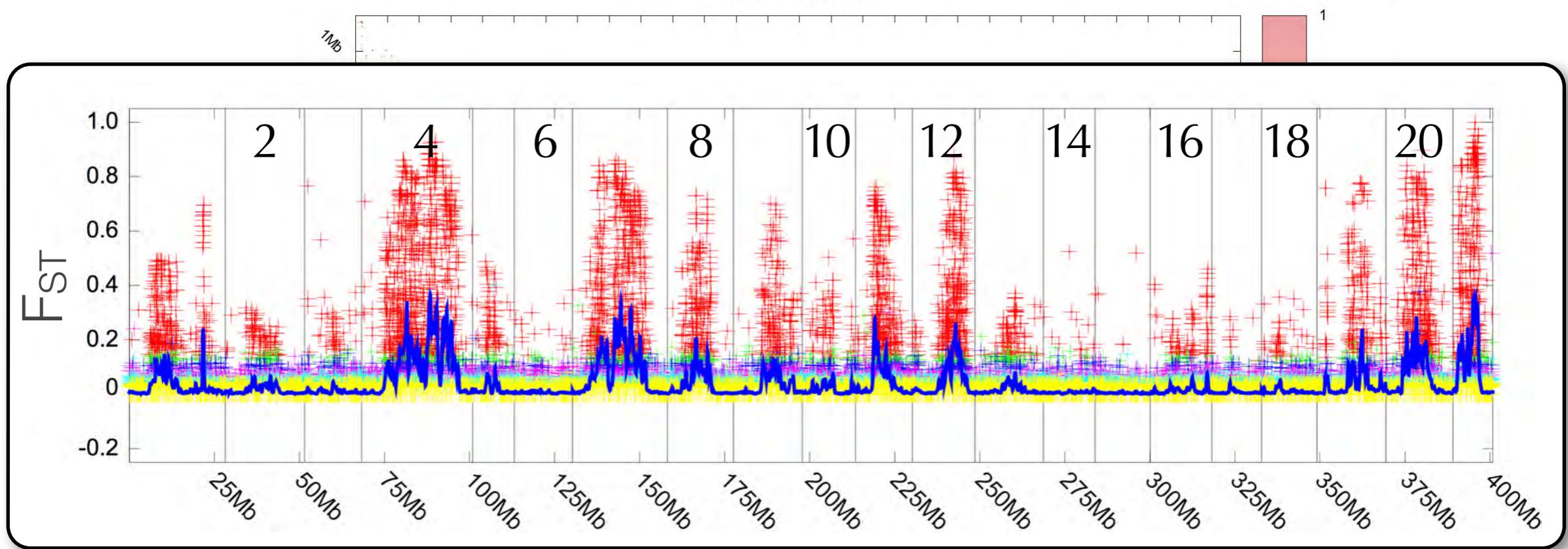


Linkage Group IV comparison

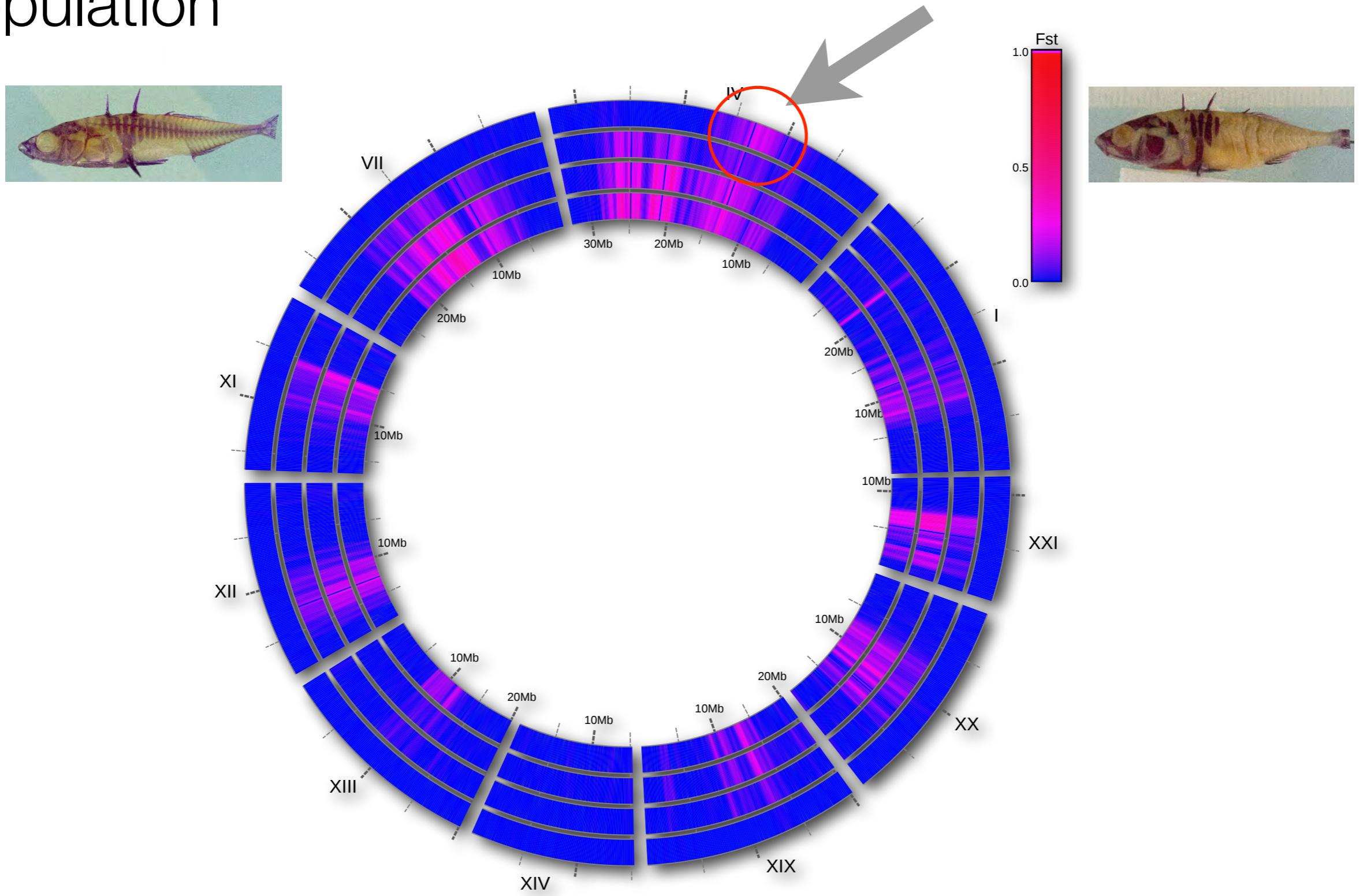


LD on Middleton Island 08, chromosome 4

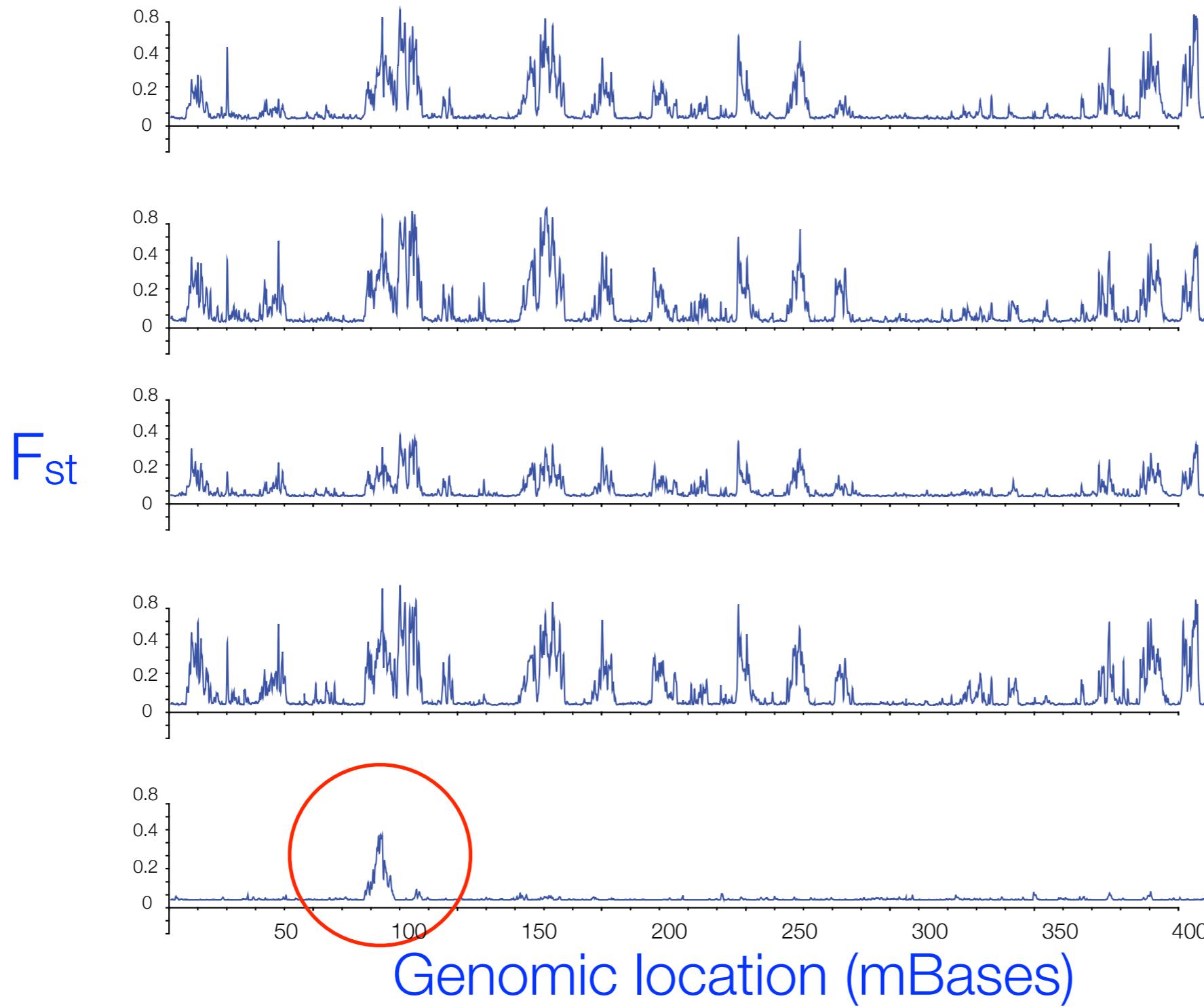
Beagle Mi08 groupVII



Genomic structure in a lateral plate polymorphic population



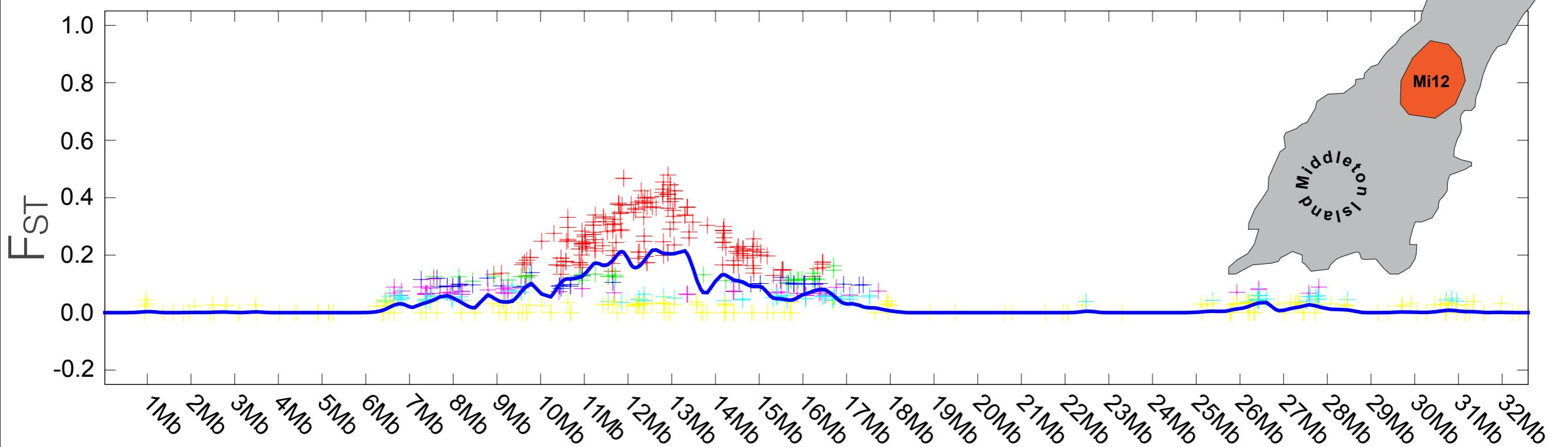
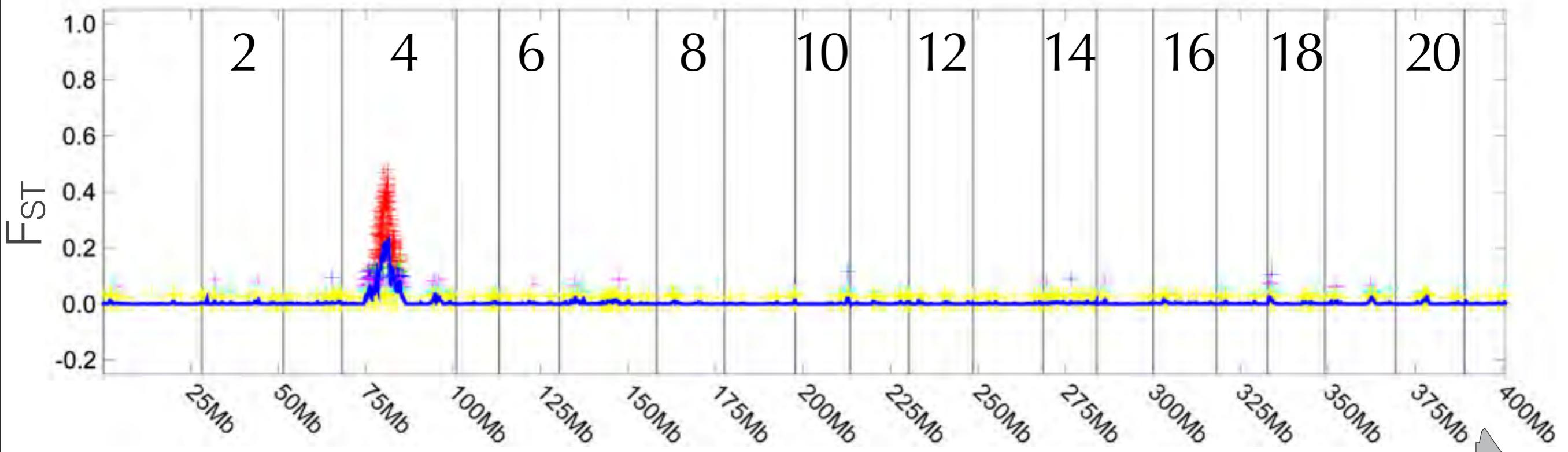
Lateral plate localization to large genomic region



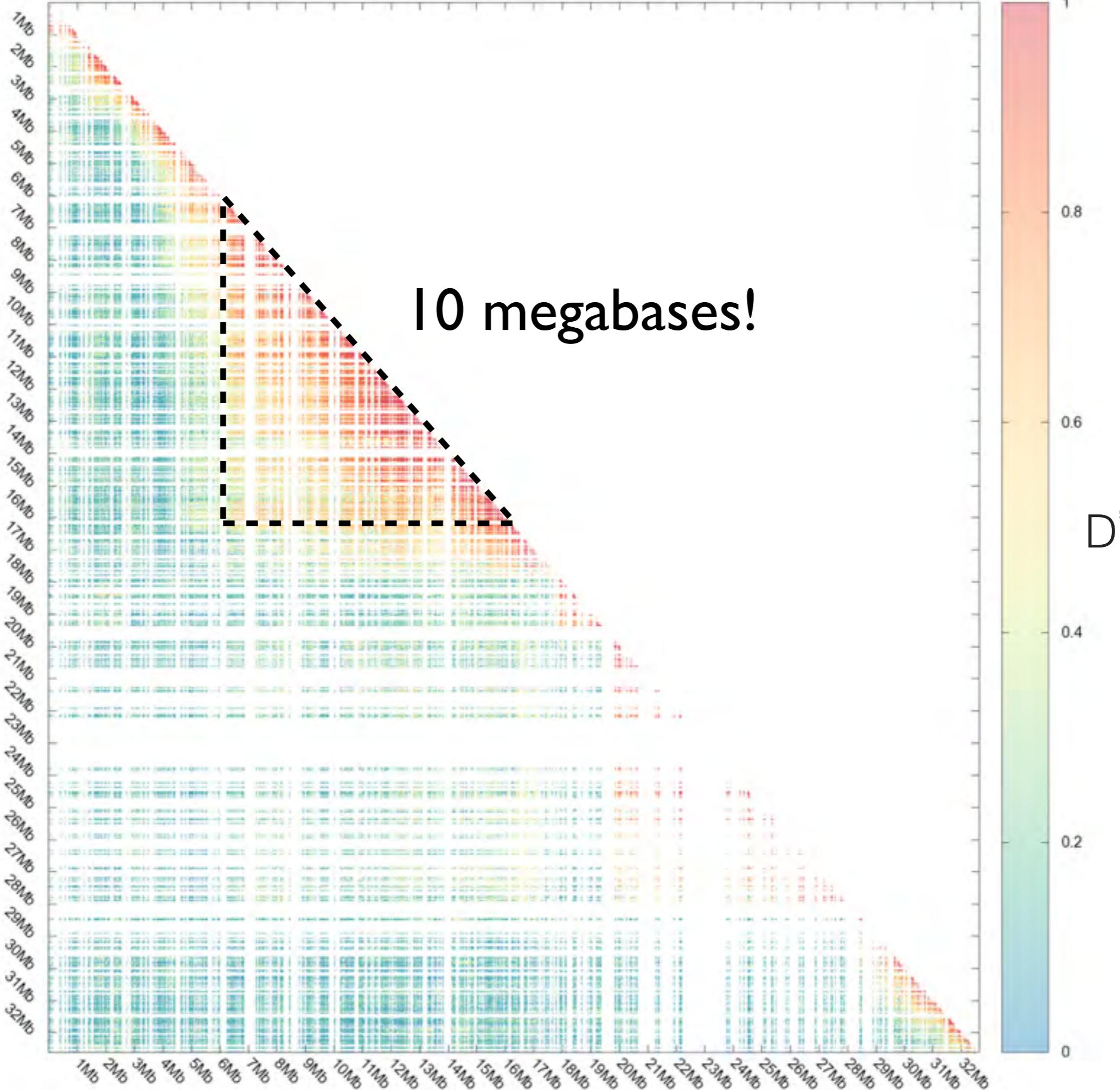
Post-1964

Polymorphic

Mi12: 3000 year old sympatric population



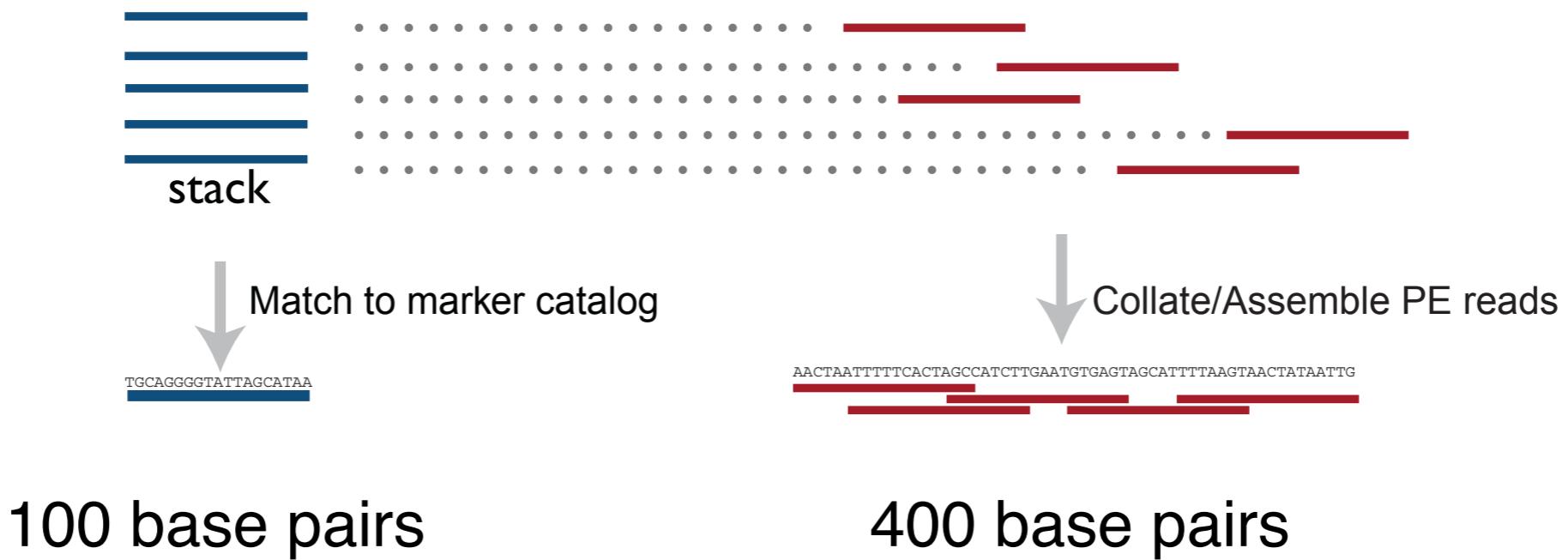
Chromosome 4



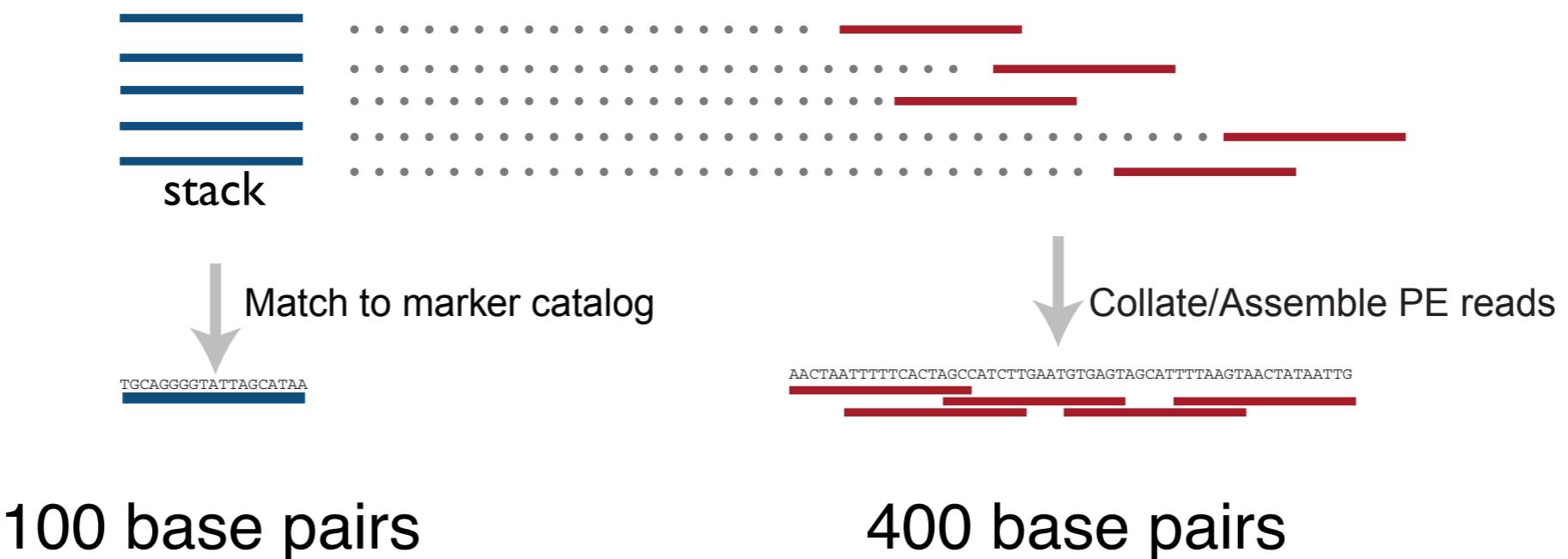
Intermediate Conclusions

- Stickleback can evolve in decades
- Evolution involves the reuse of standing genetic variation
- Signatures of selection appear in divergent habitats
- Loci important for local adaptation are genomically localized
- Linkage patterns of loci begs for the analysis of haplotypes

From SNPs to haplotypes

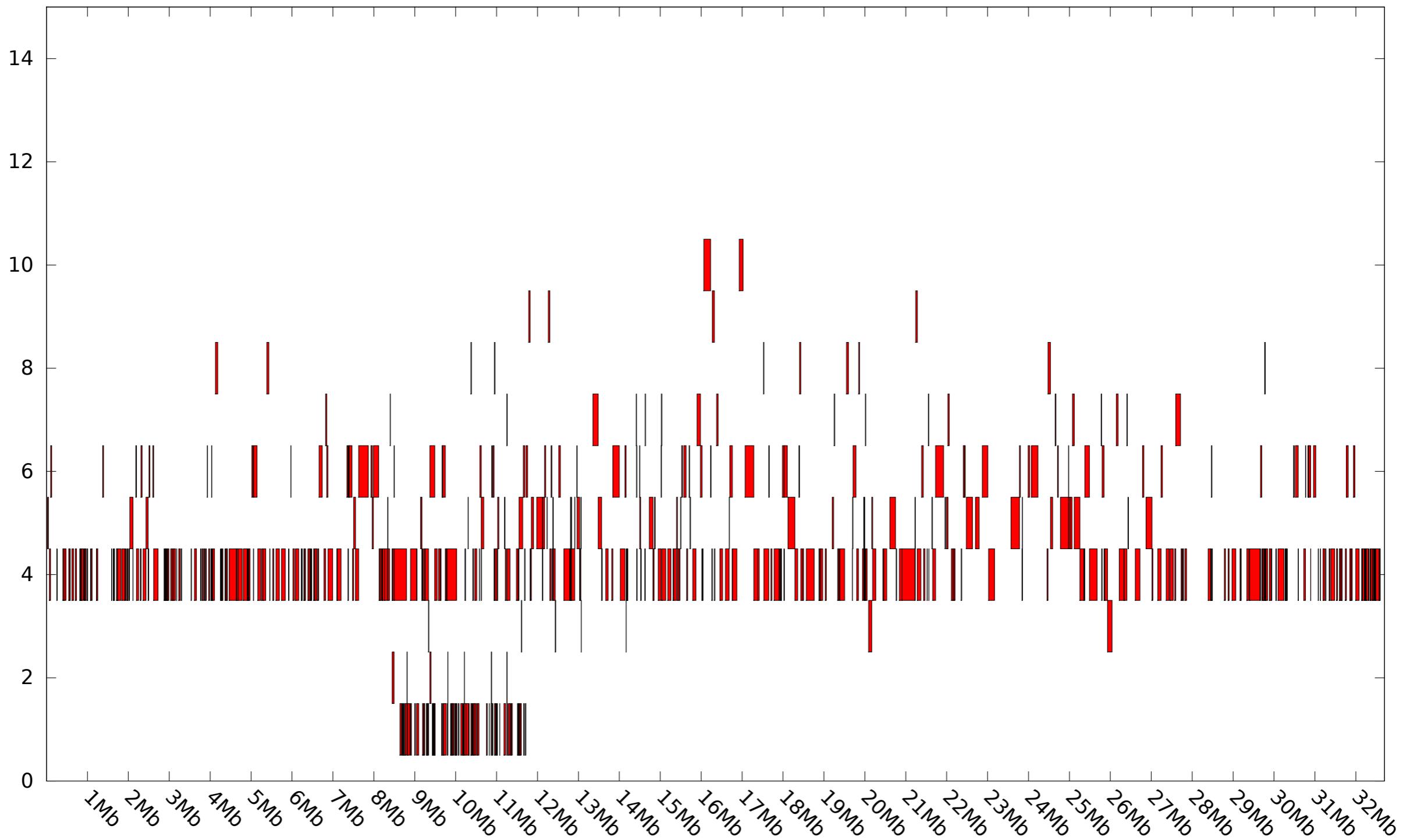


From SNPs to haplotypes

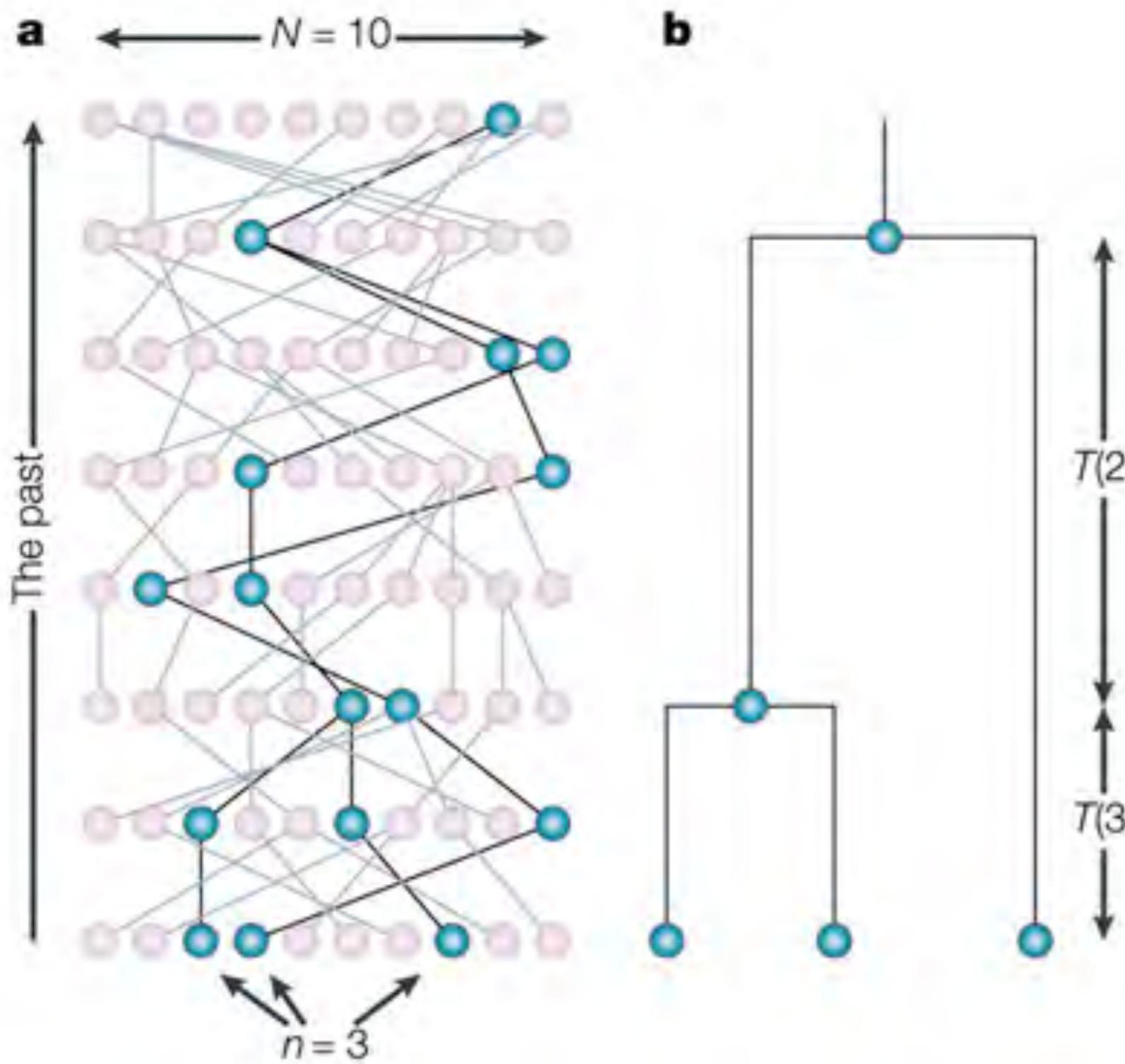


- SNPs can be ordered into haplotypes
- Haplotypes provide deep & shallow evolutionary information
- Phasing genotypes within and among RAD sites
- Genotype imputation for missing SNPs

Haplotype block counts on LGIV

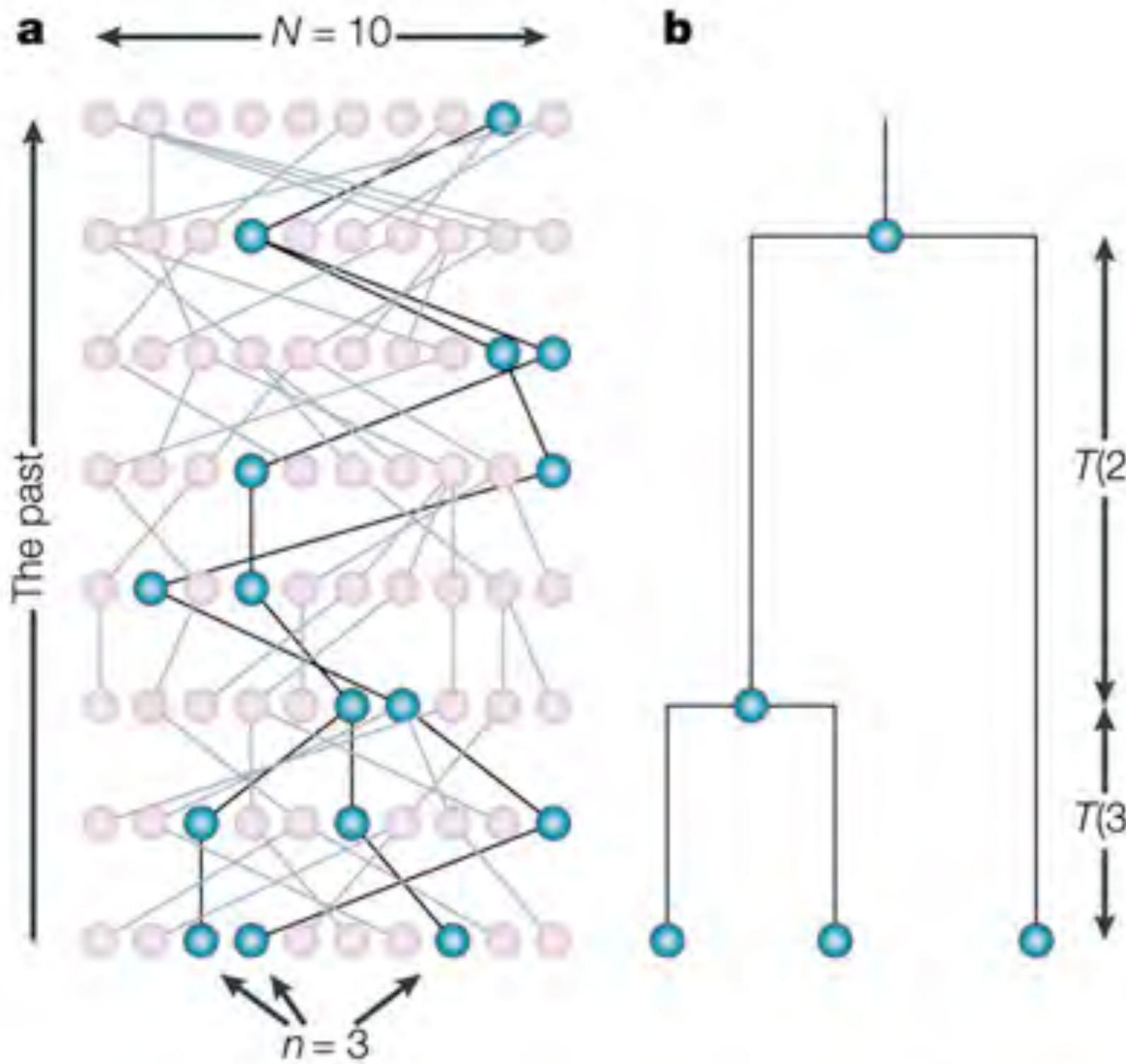


Coalescent analysis using RAD-seq data



Noah A. Rosenberg & Magnus Nordborg
Nature Reviews Genetics 3, 380-390 (May 2002)

Coalescent analysis using RAD-seq data



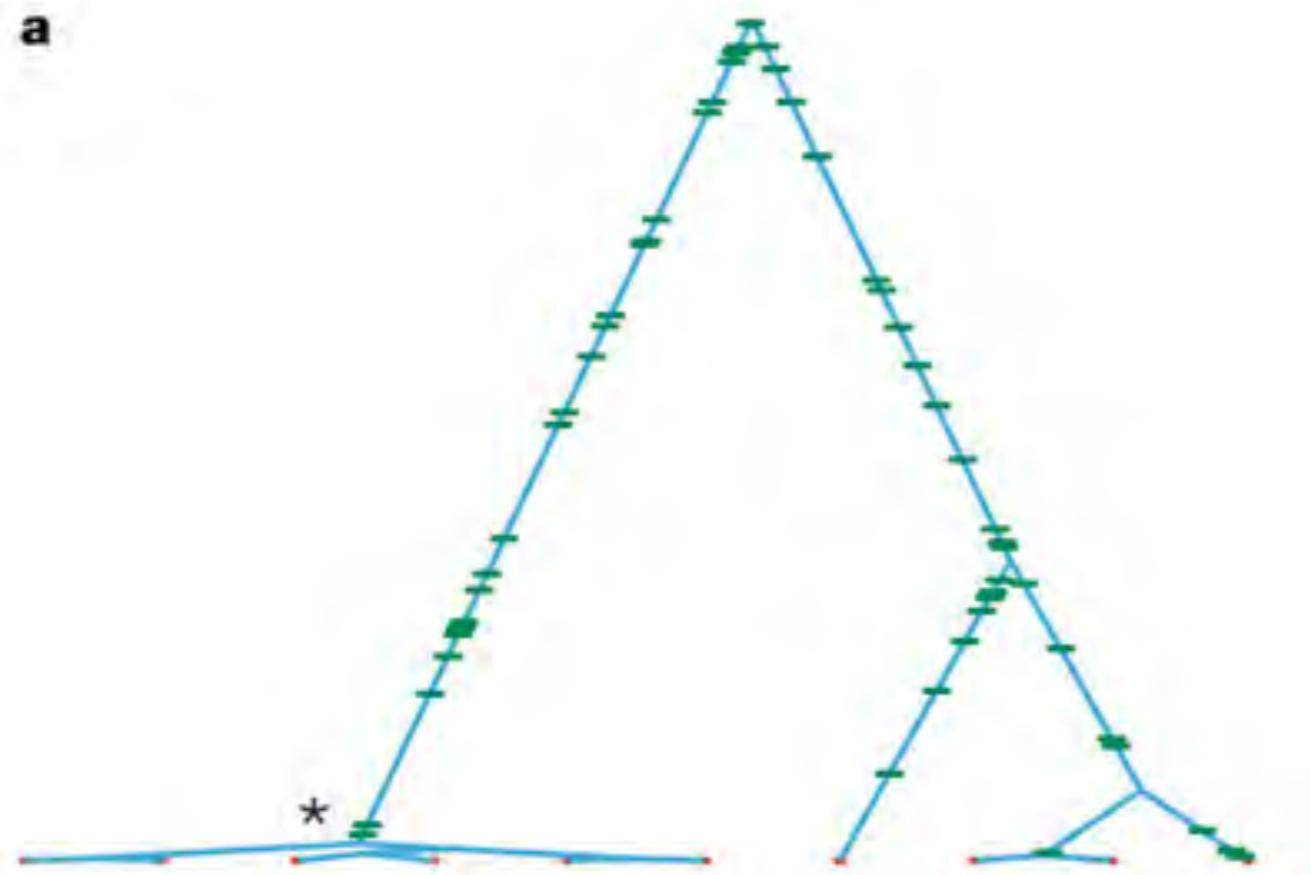
35000

Noah A. Rosenberg & Magnus Nordborg
Nature Reviews Genetics 3, 380-390 (May 2002)

Neutral coalescent expectations

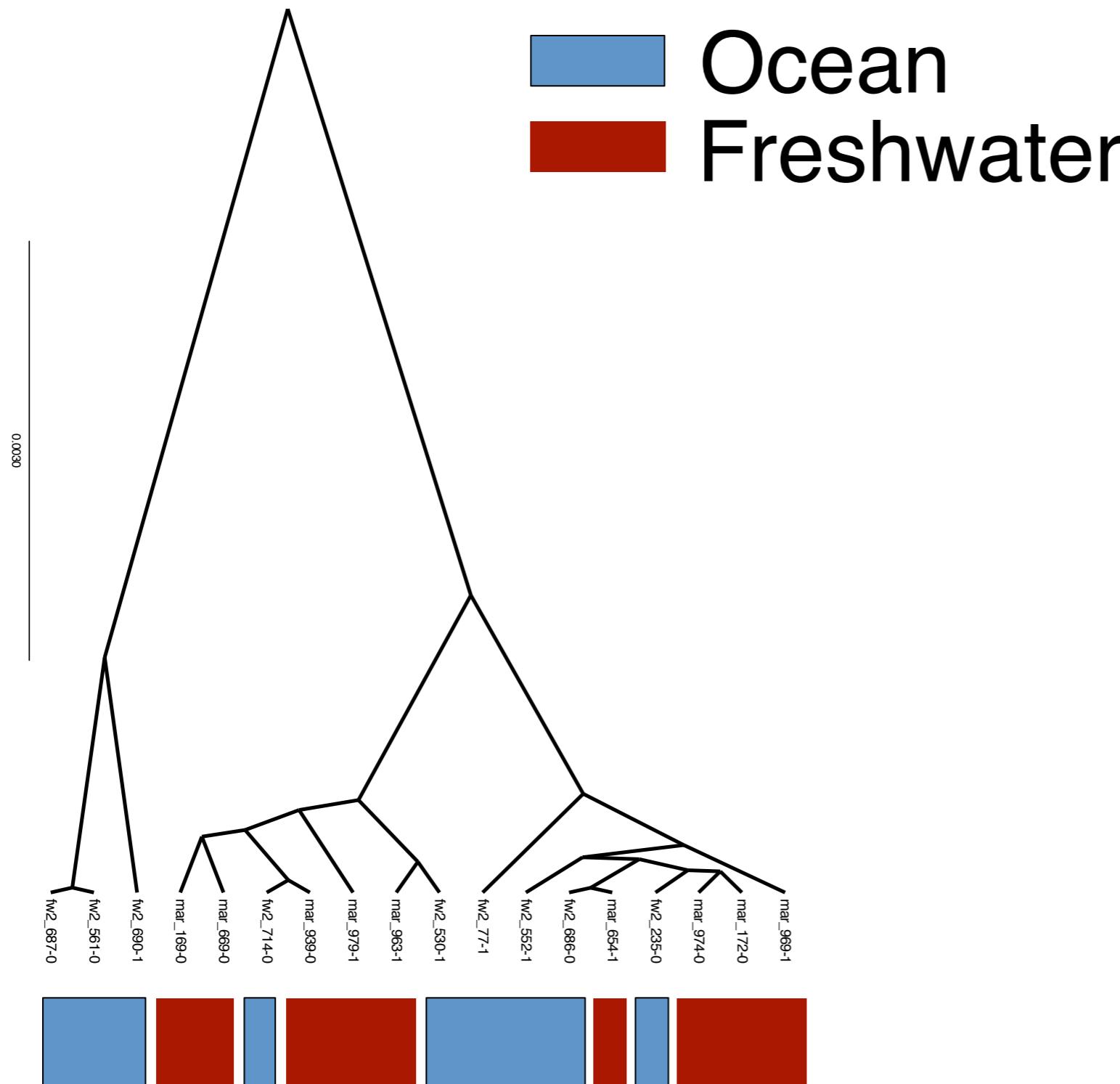


Natural selection and the coalescent



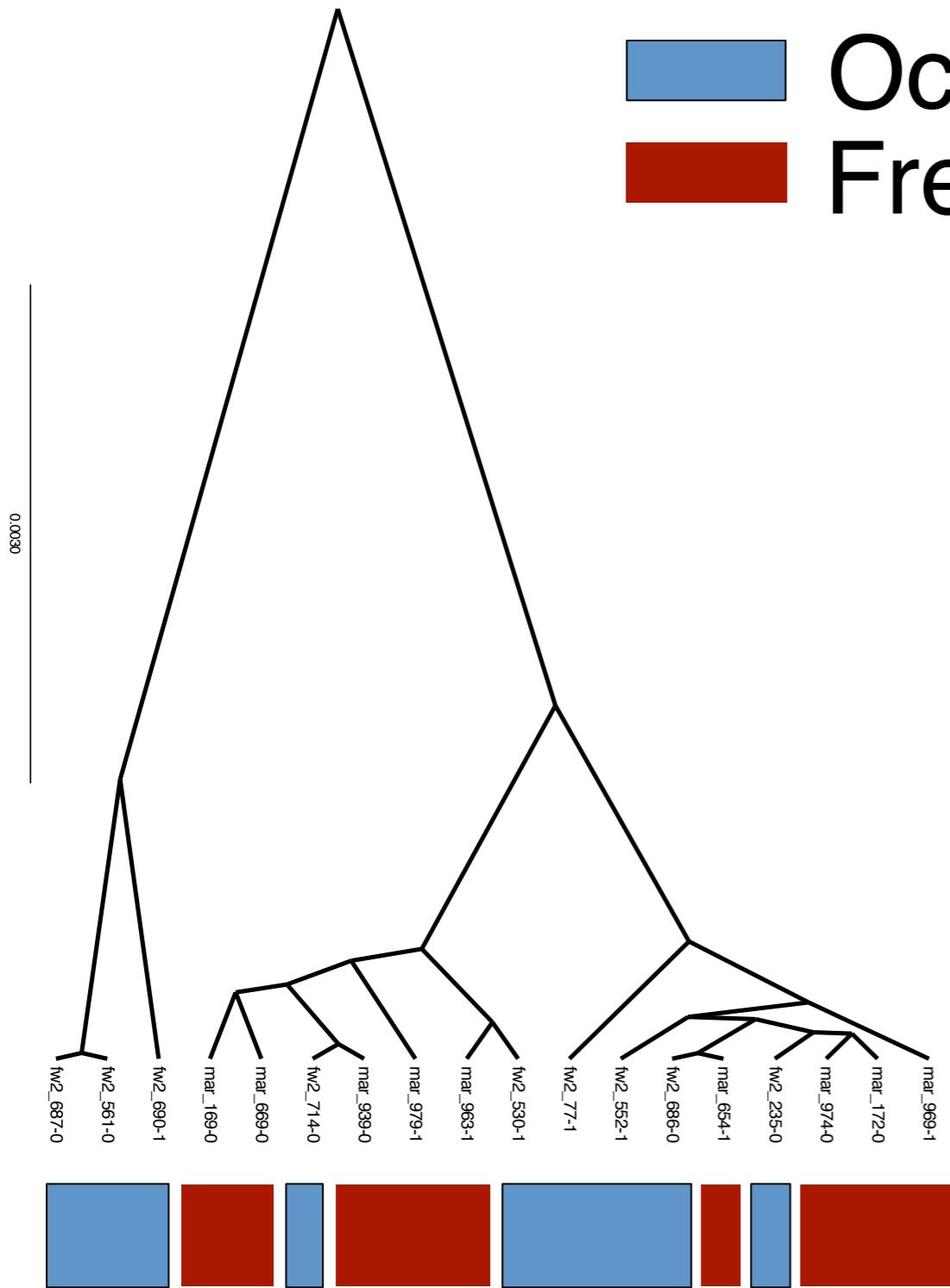
Noah A. Rosenberg & Magnus Nordborg
Nature Reviews Genetics 3, 380-390 (May 2002)

RAD-seq coalescent in stickleback

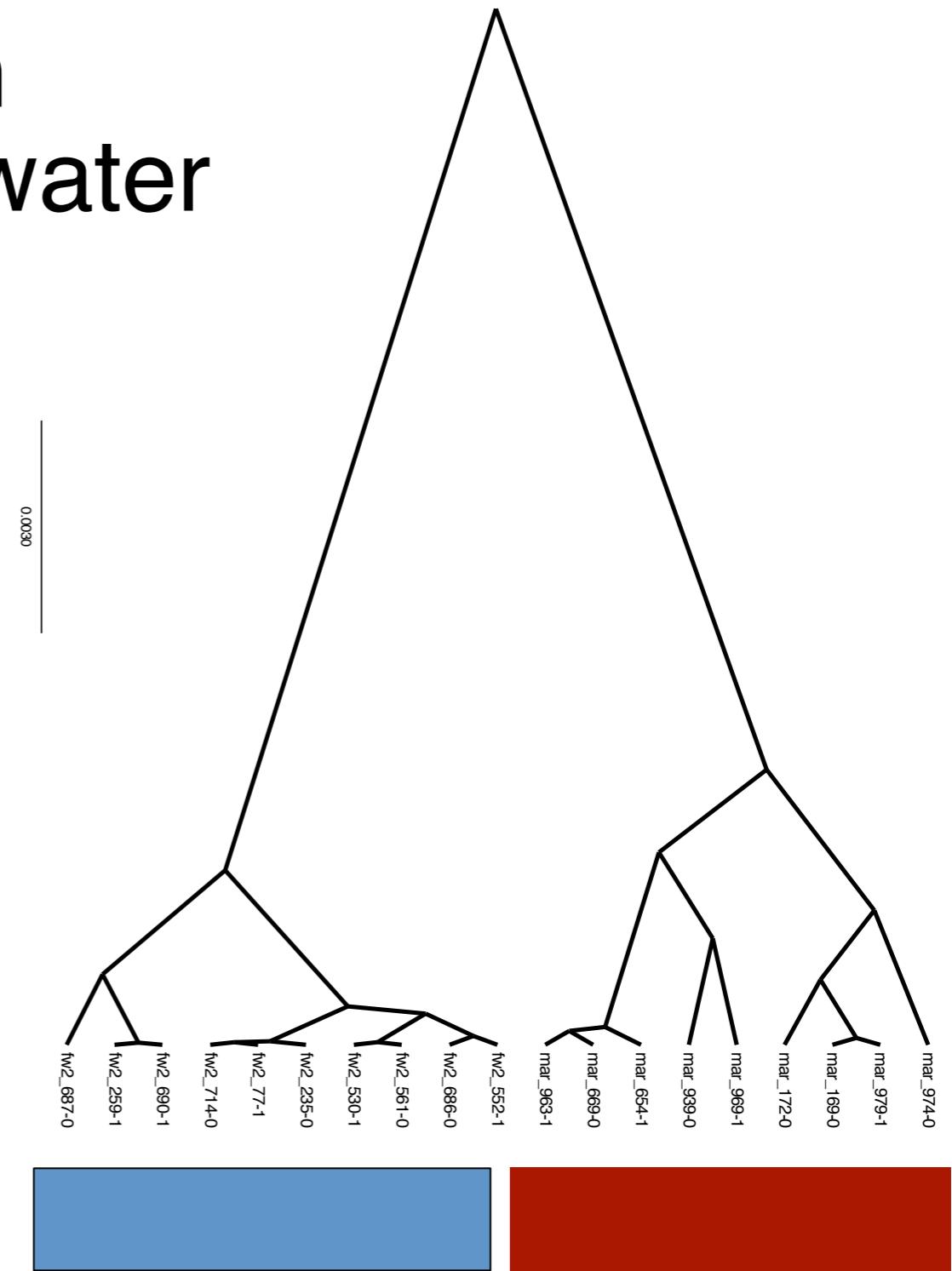


Thom Nelson &
Julian Catchen

RAD-seq coalescent in stickleback

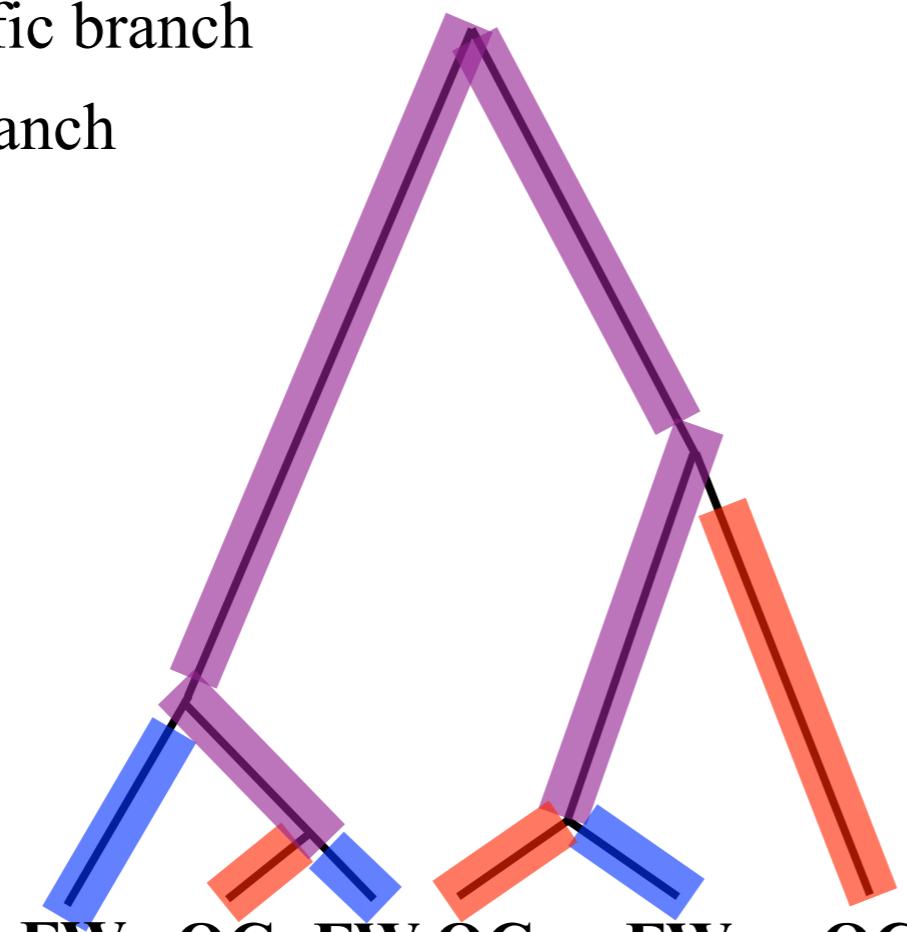


Ocean
Freshwater

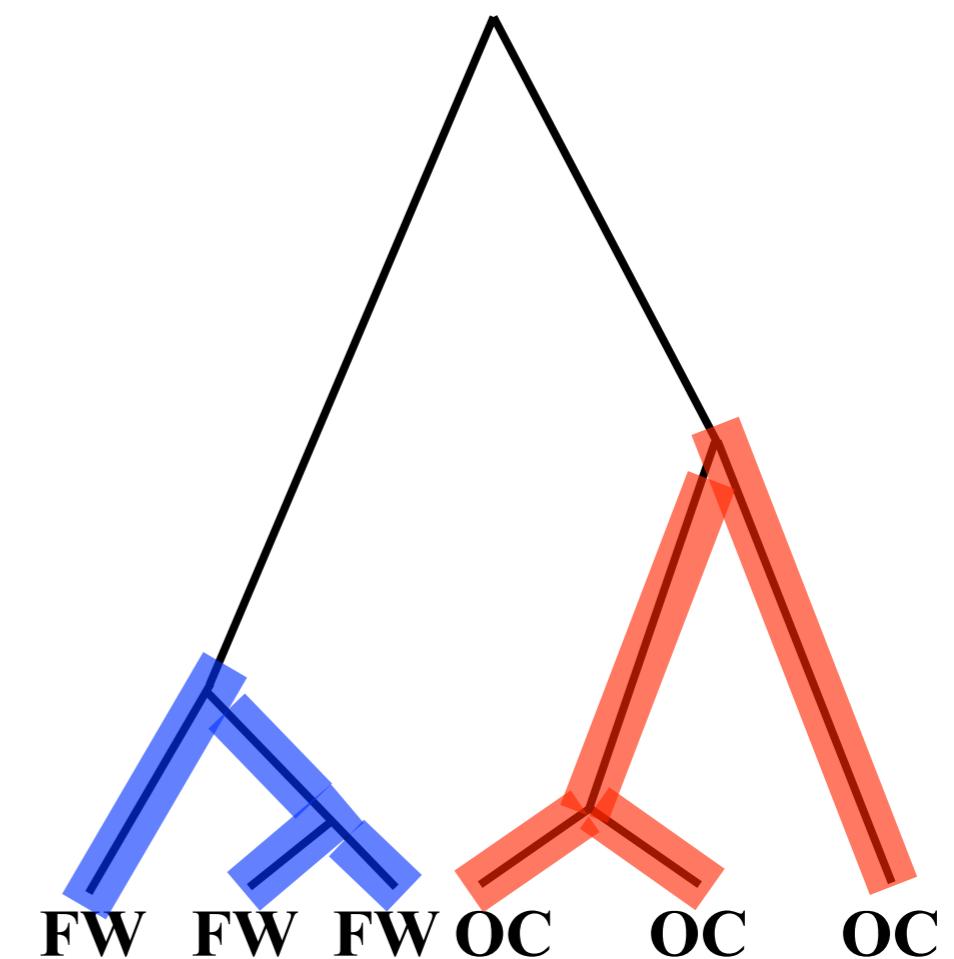


RAD-seq coalescent in stickleback - **UNIFRAC**

- FW-specific branch
- OC-specific branch
- Shared branch

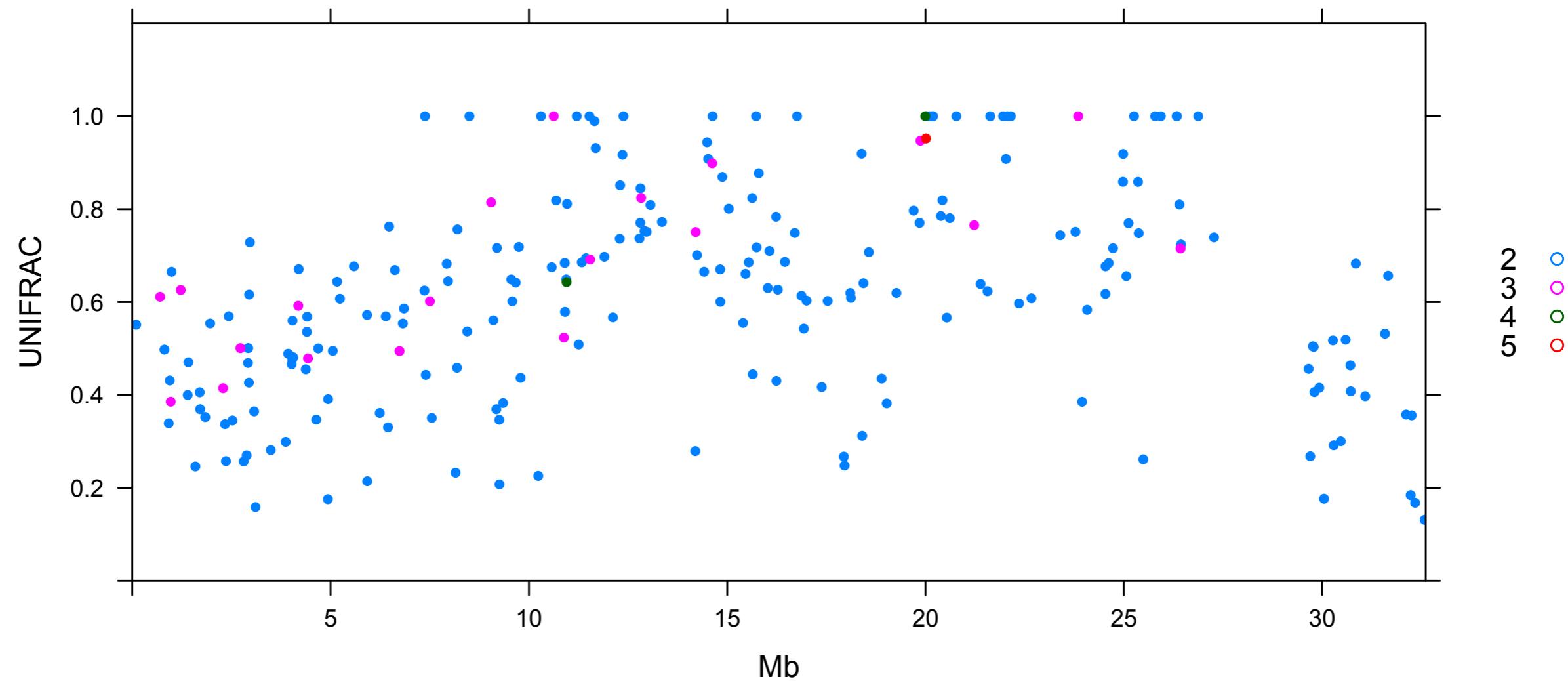


UNIFRAC distance ~ 0.25

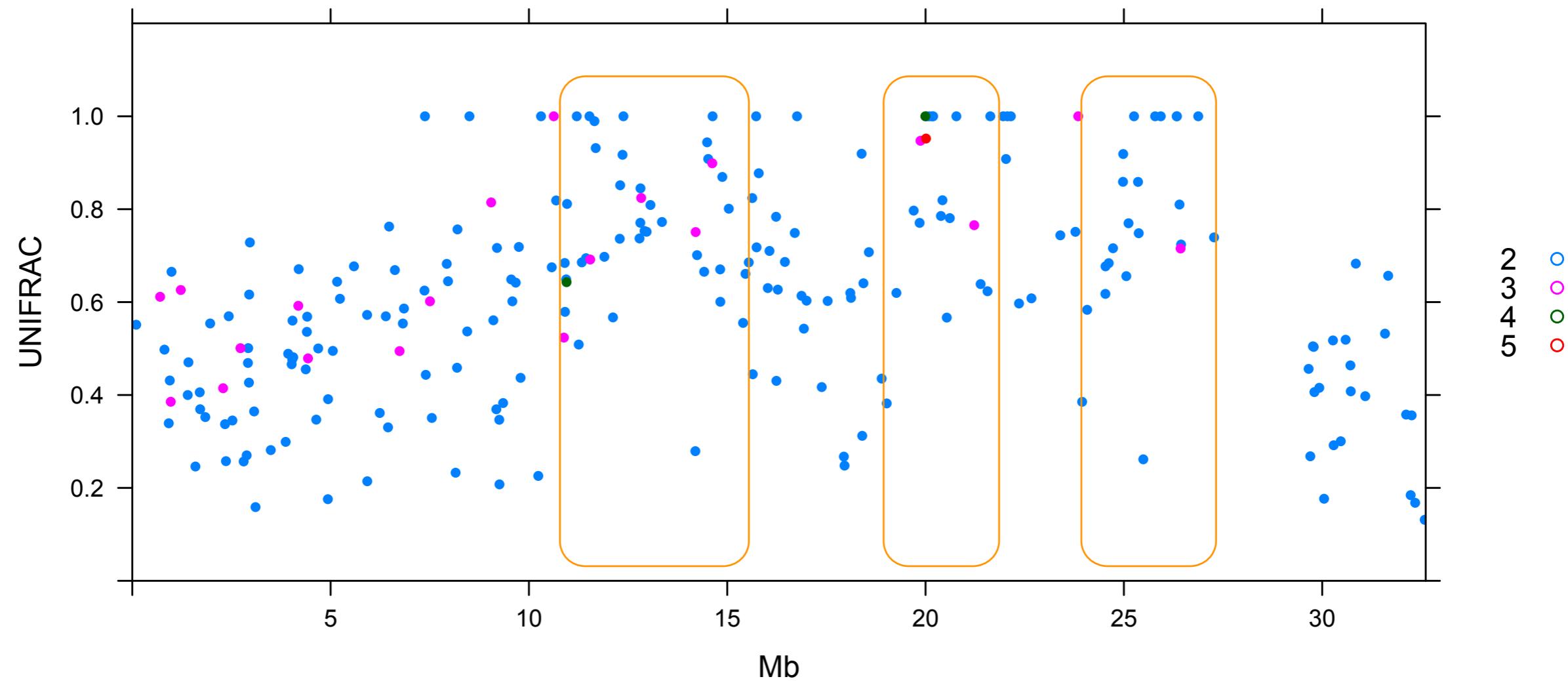


UNIFRAC distance = 1.0

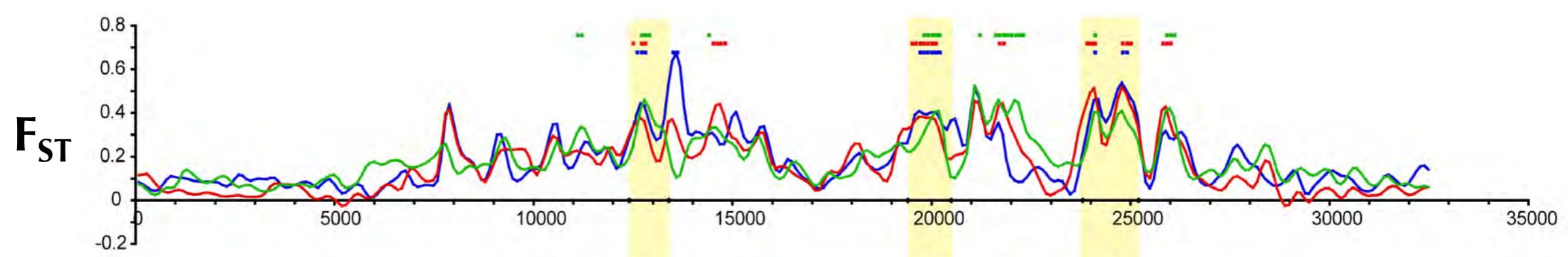
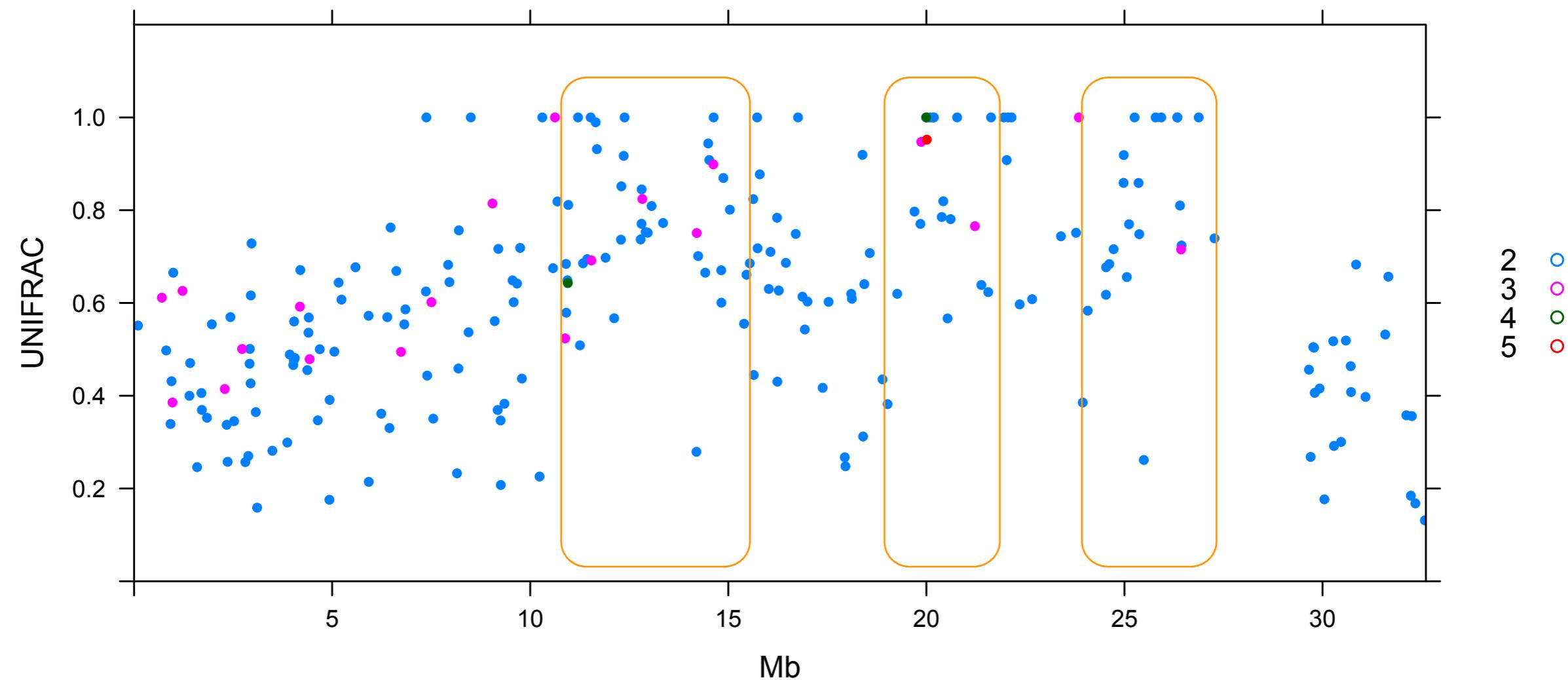
RAD-seq coalescent in stickleback - **UNIFRAC**



RAD-seq coalescent in stickleback - **UNIFRAC**



RAD-seq coalescent in stickleback - **UNIFRAC**

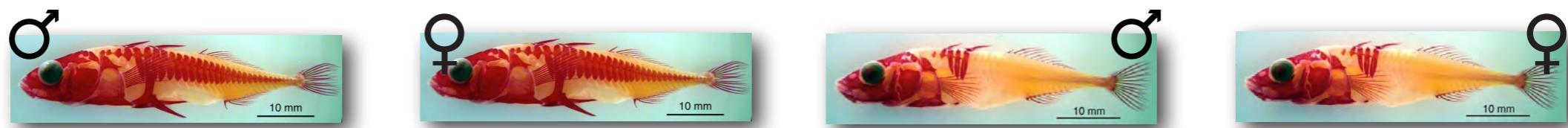
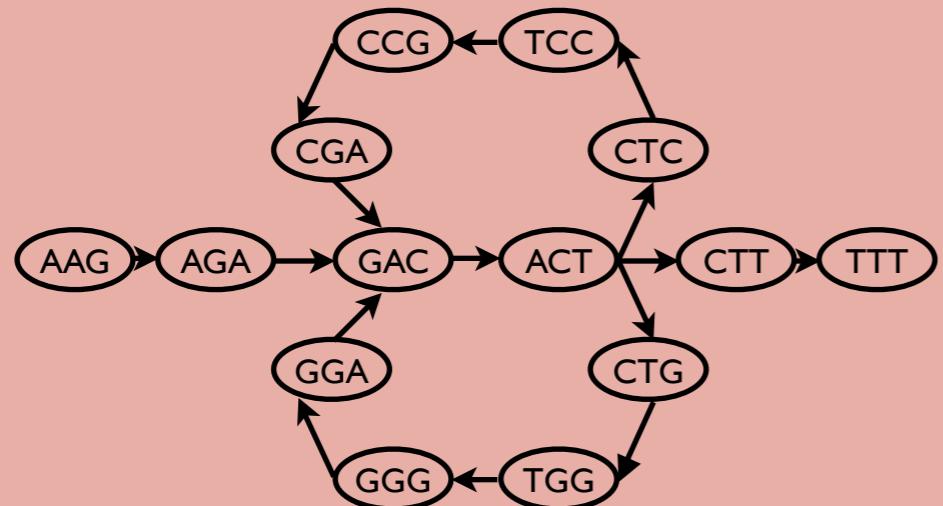


What can explain such rapid evolution and haplotype structure?

Is the stickleback genome architecture partly responsible?

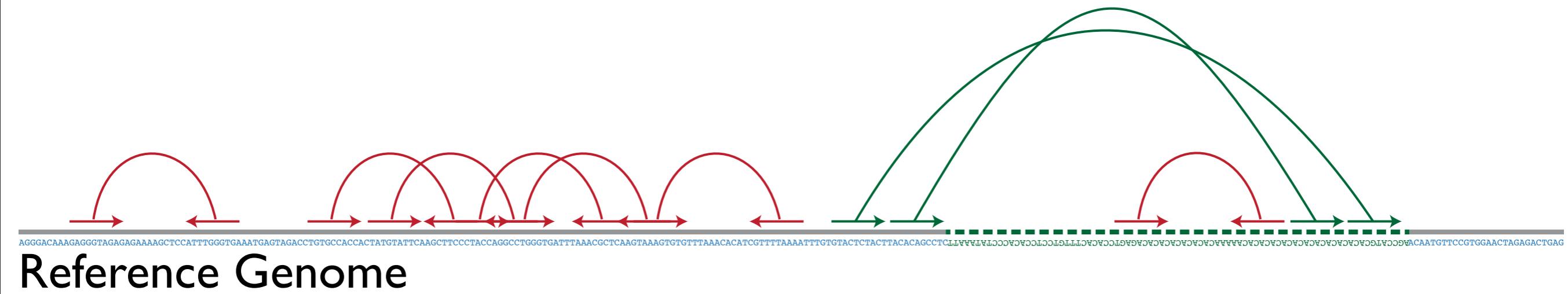
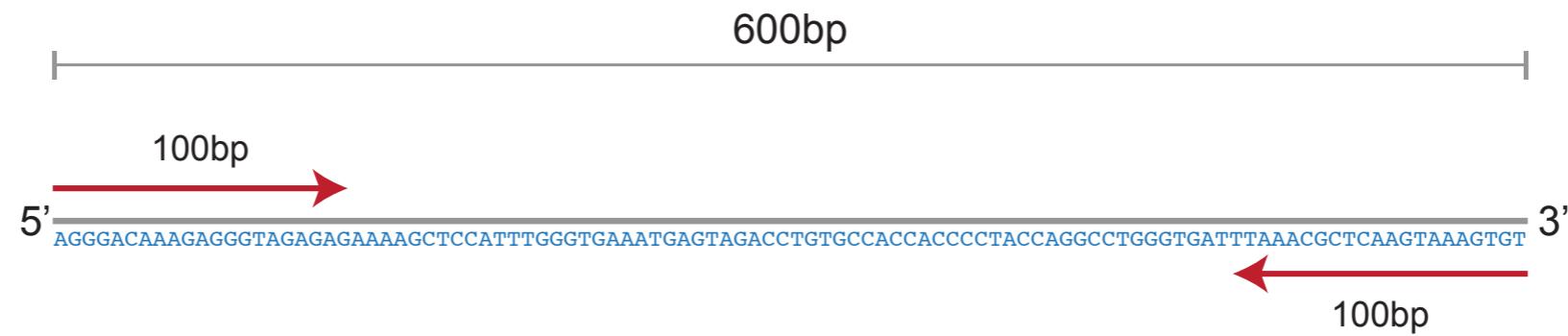
Julian Catchen, Susie Bassham
and Thom Nelson

Genome Assembly



N50	17,417 bp	18,982 bp	15,555 bp	15,534 bp
Max	199,905 bp	192,283 bp	238,768 bp	254,734 bp
Total	488.8 Mb	472.5 Mb	456.4 Mb	473.4 Mb
Median Coverage	24.6x	26.5x	24.1x	25.8x

Illumina Paired-end Reads



Reference Genome



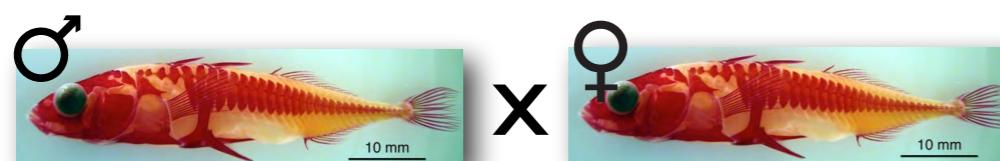
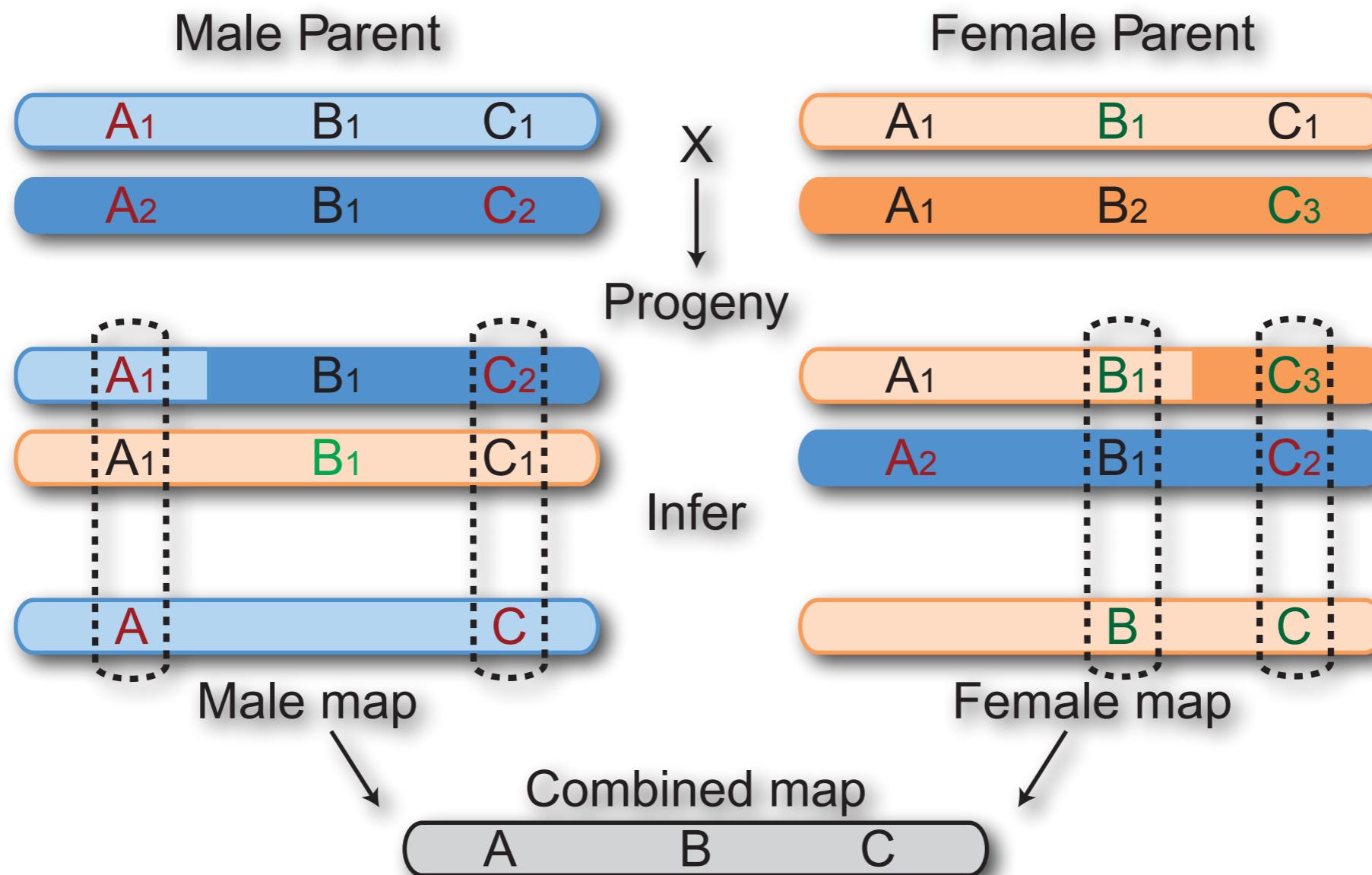
161,305,595 pairs

144,396,898 pairs

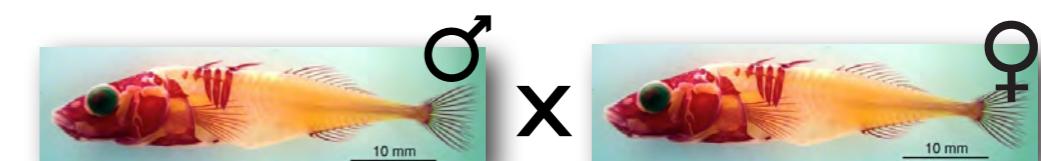
131,471,548 pairs

150,786,462 pairs

FI Pseudo-testcross

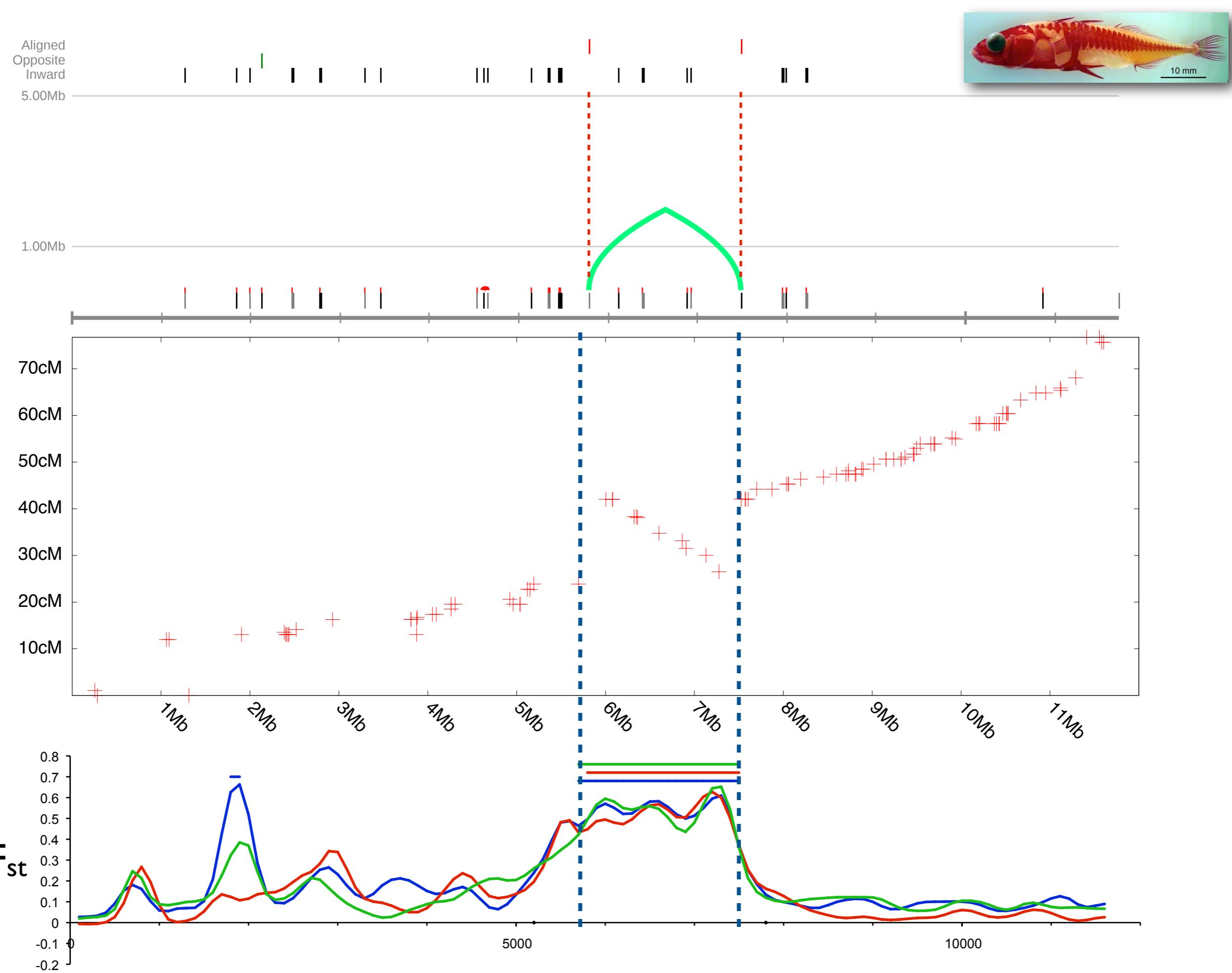


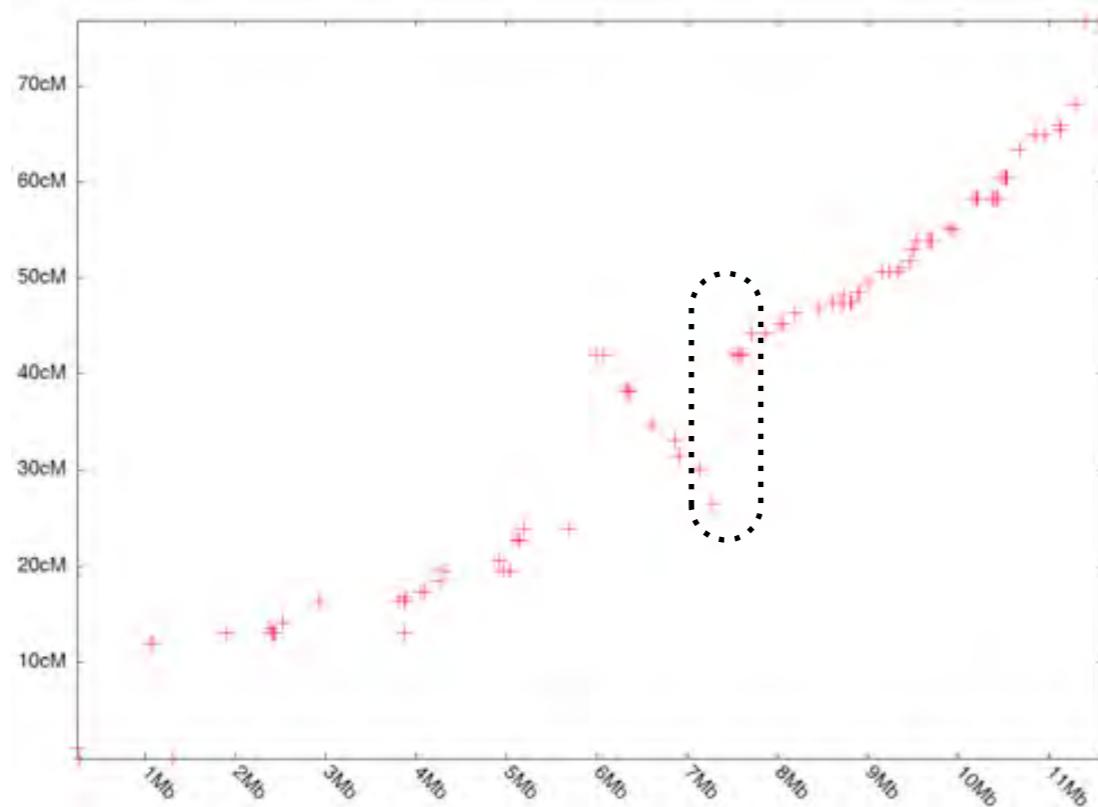
93 progeny
66,071 loci
5,351 markers



93 progeny
45,301 loci
3,927 markers

LGXX



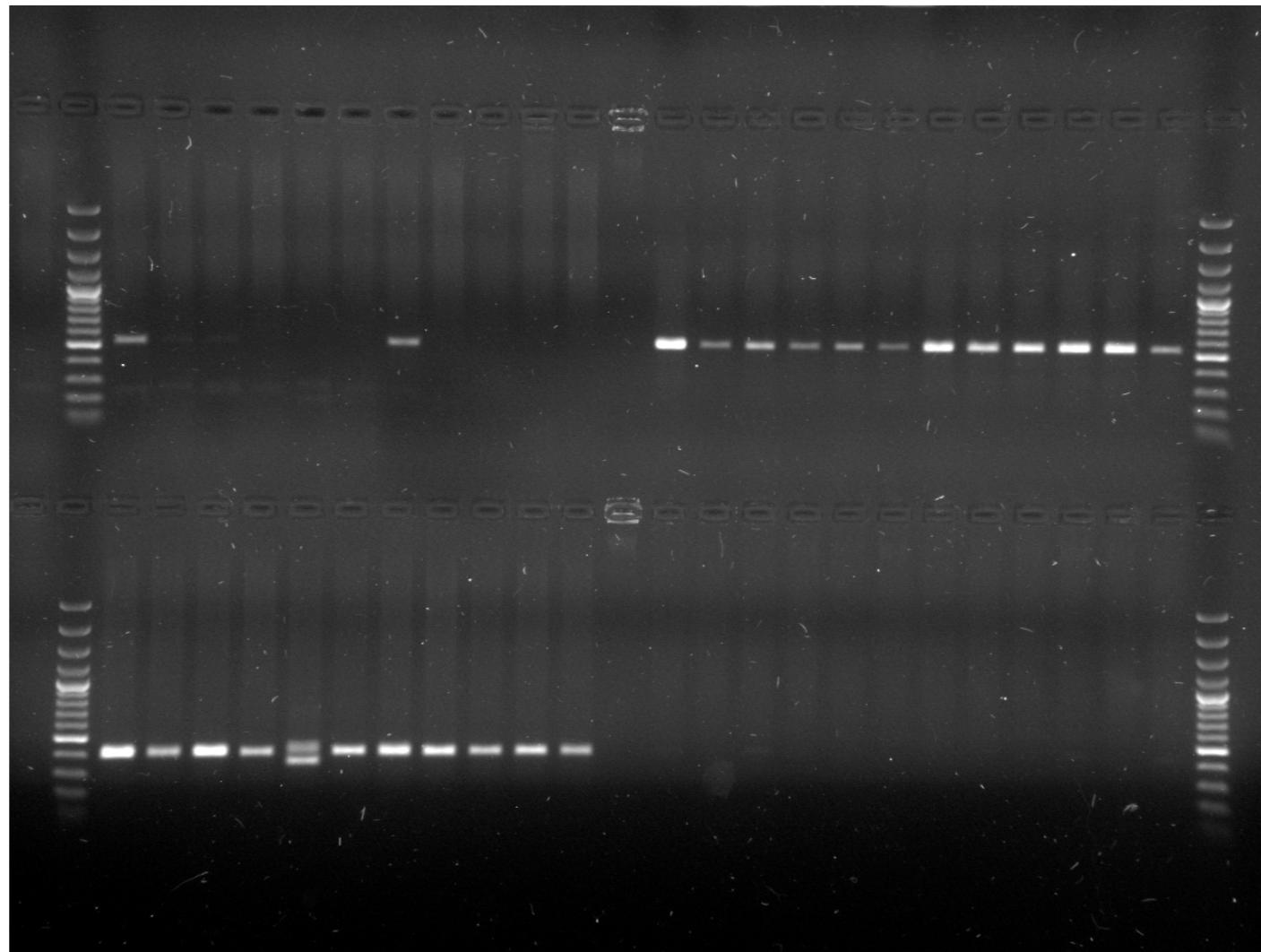


Linkage Group XXI

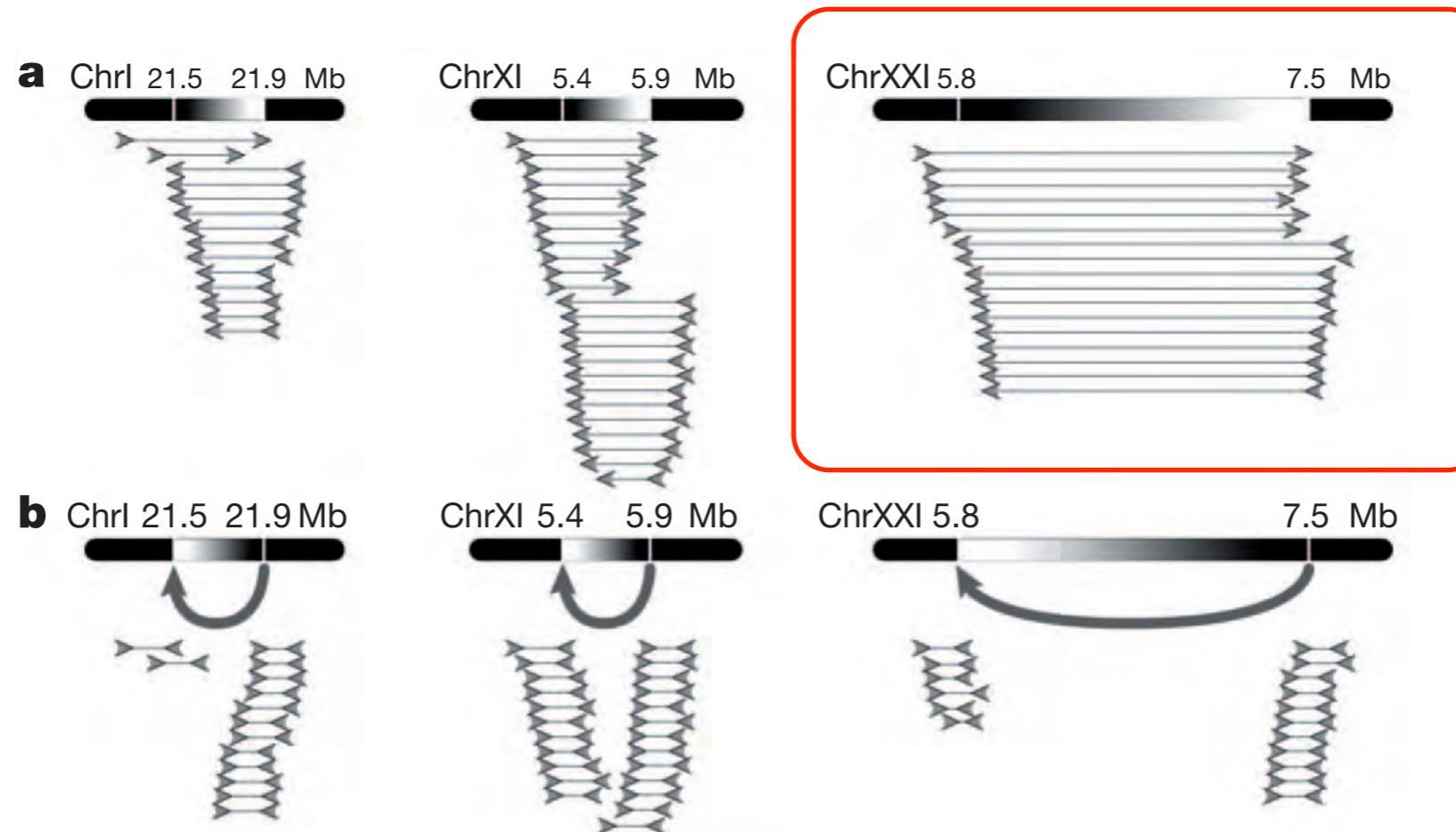
RS (Marine) Boot (Freshwater)

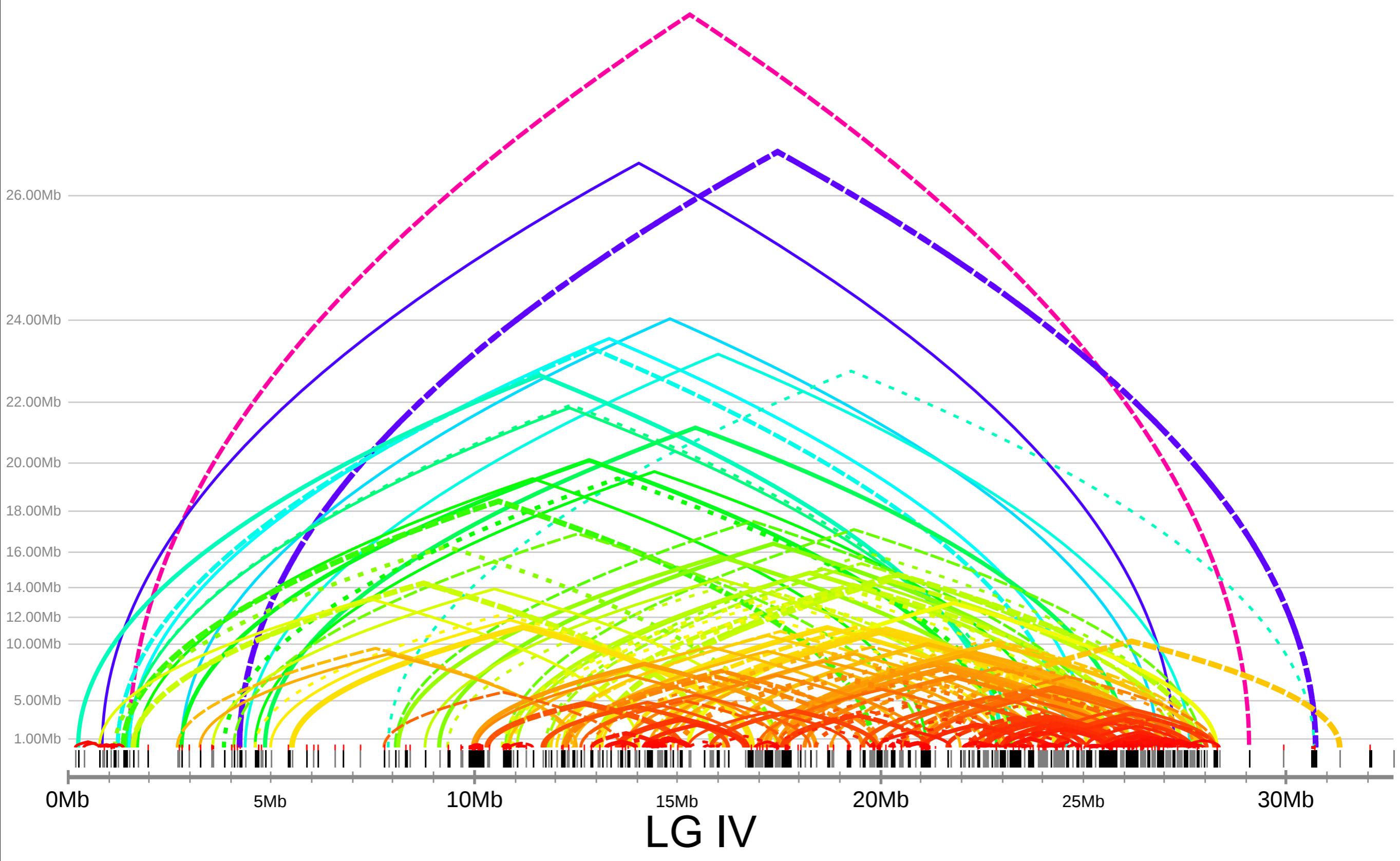
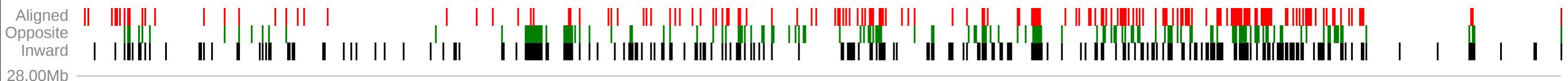
Genome
Arrangement

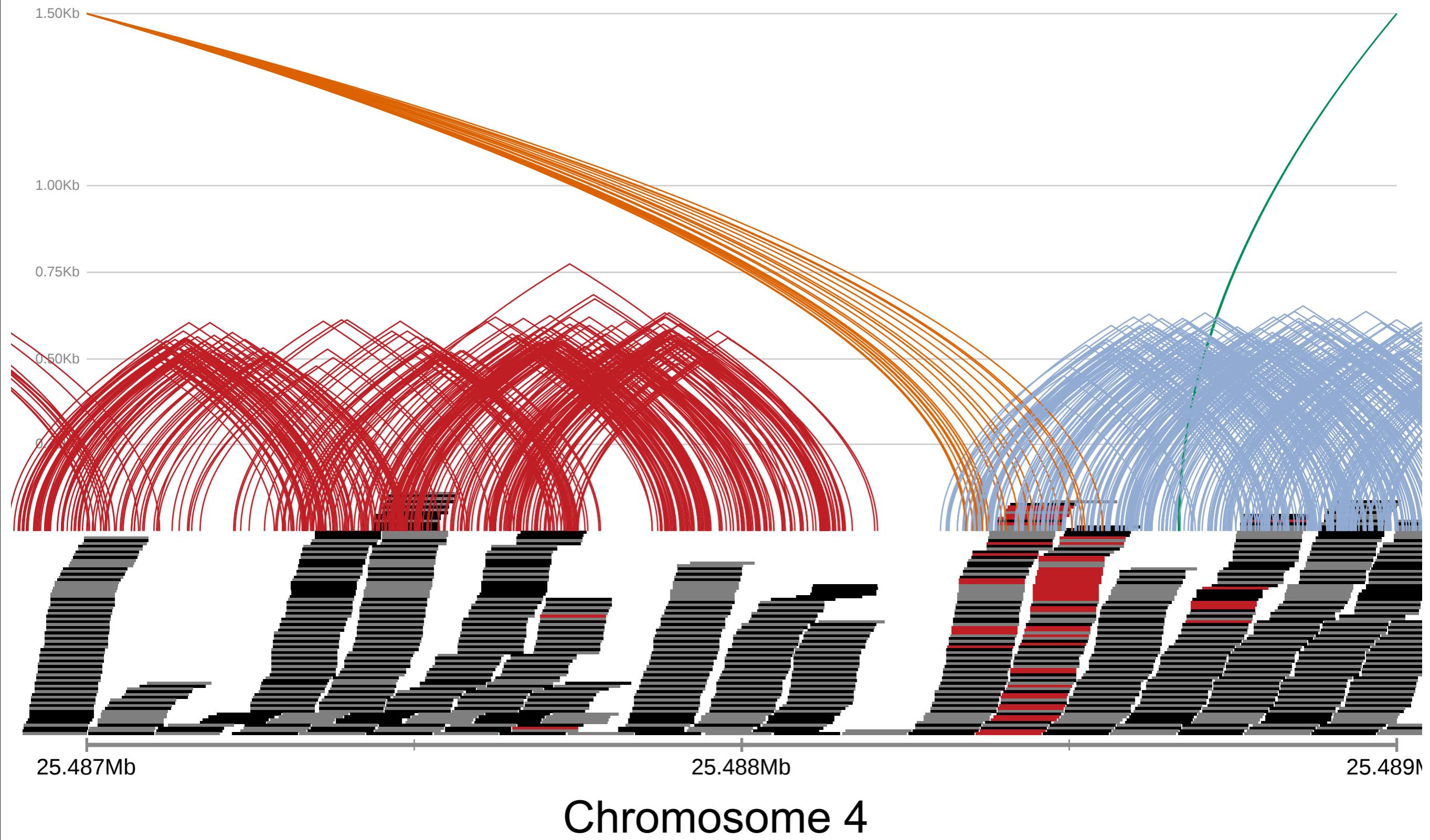
Inverted



Global analysis also identified these inversions

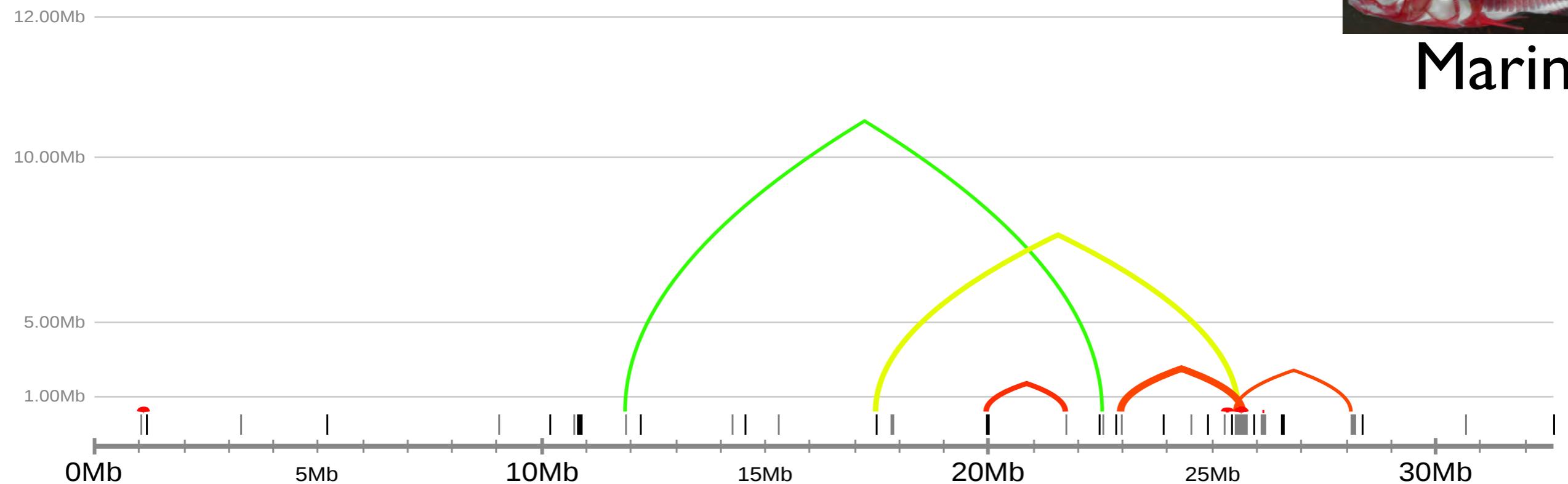




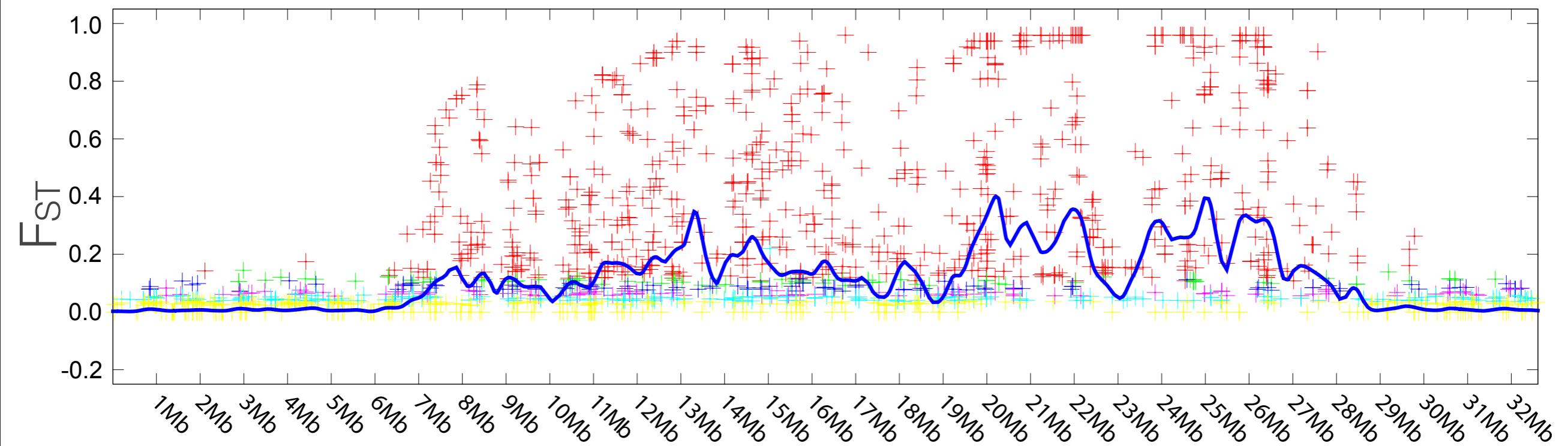




Marine ♀

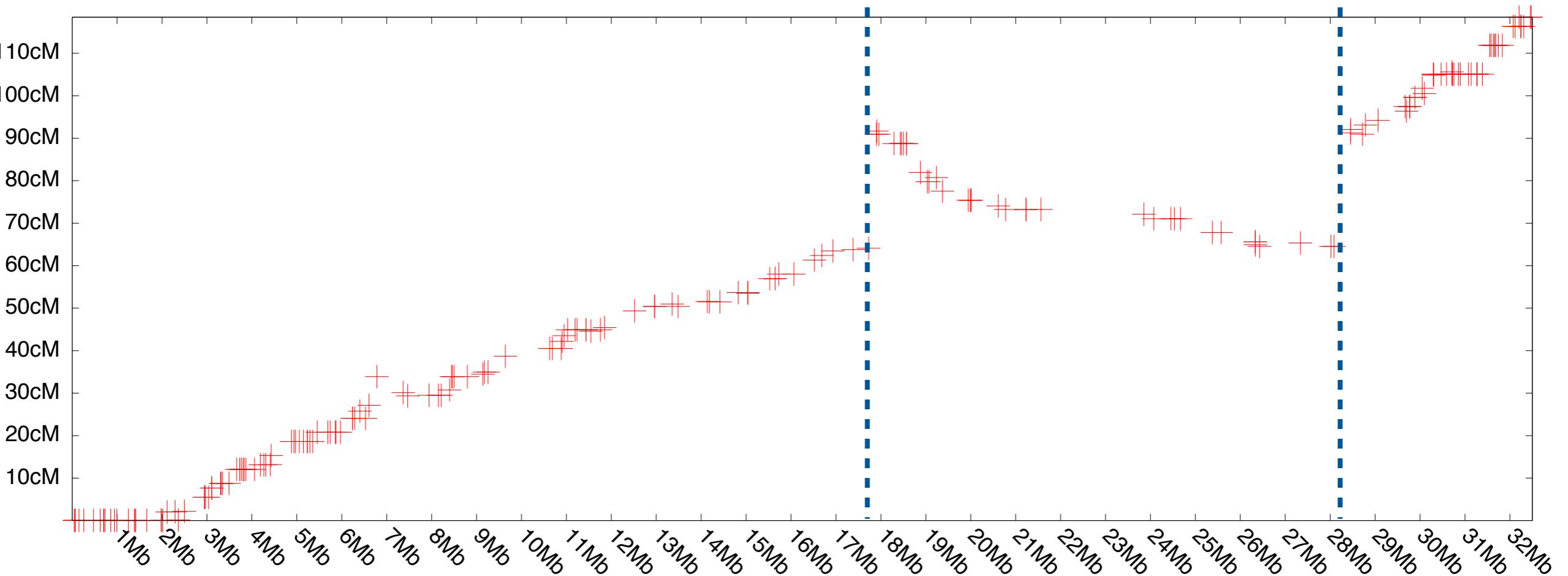
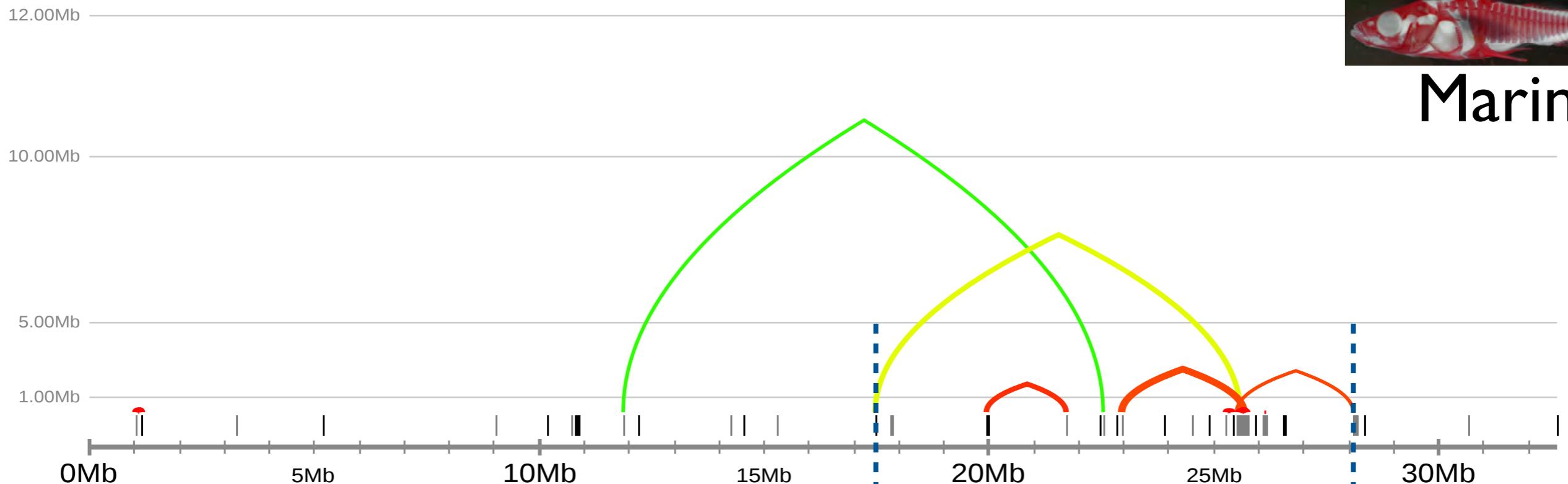


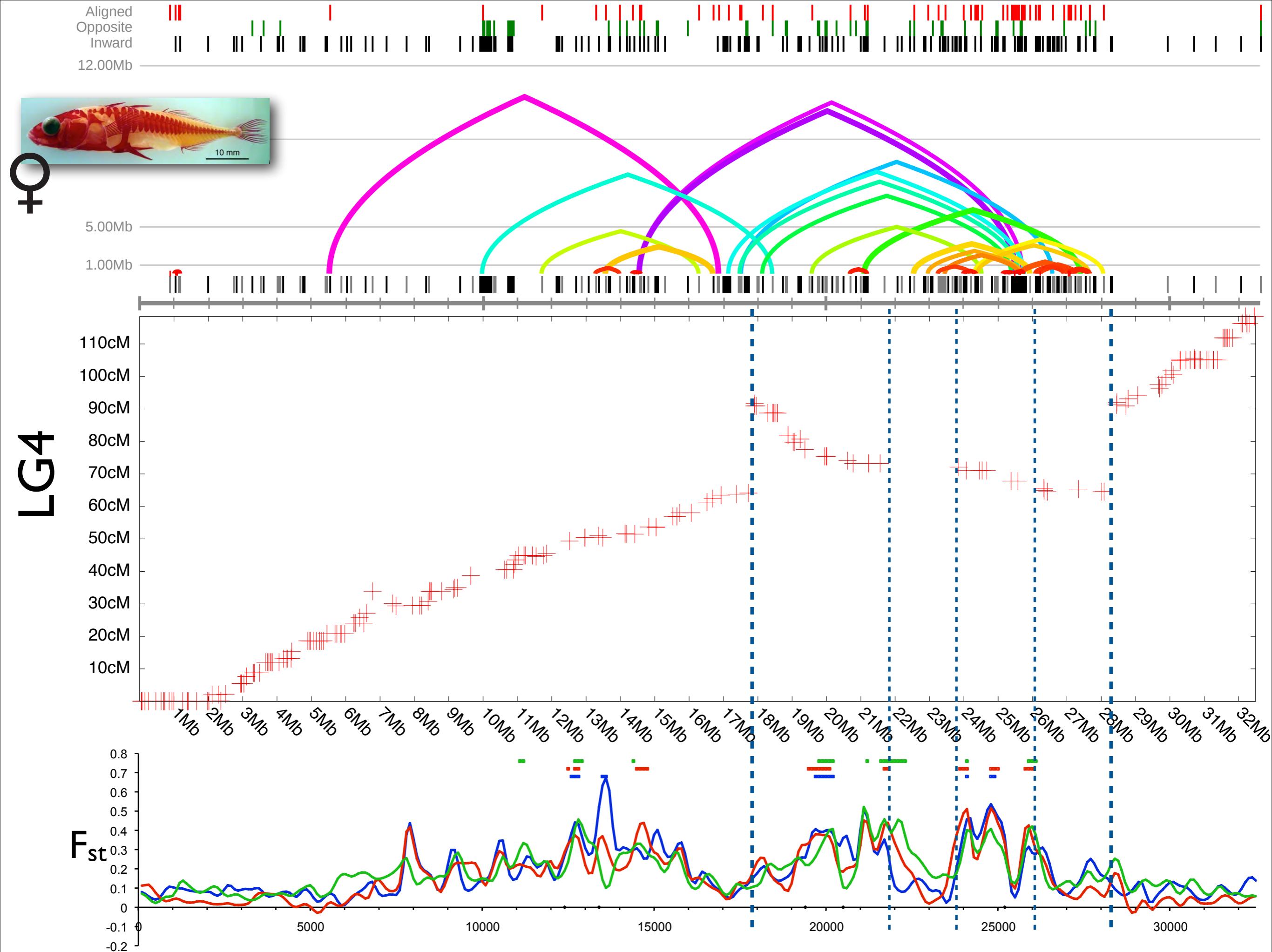
Chromosome 4





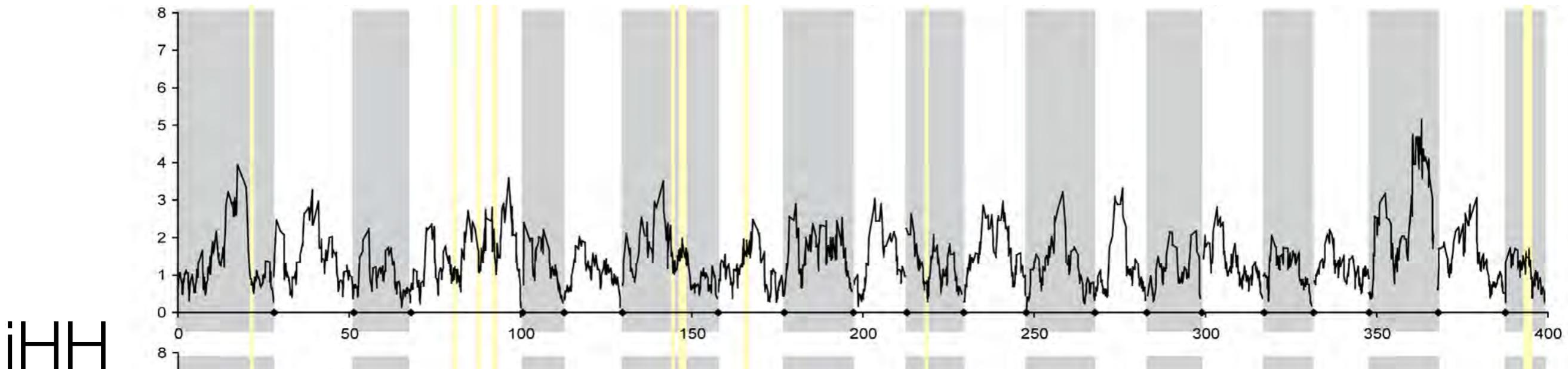
Marine ♀



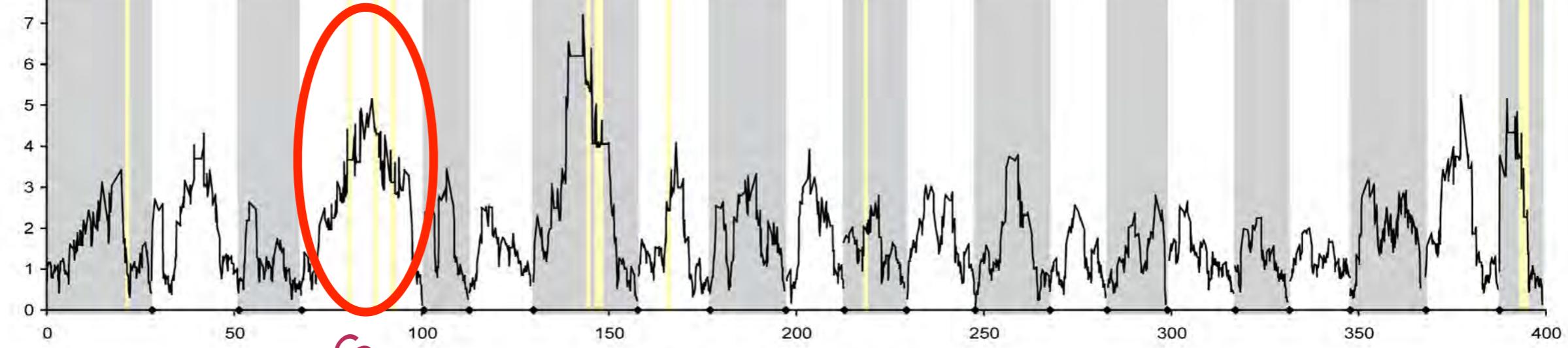


Inferred inversions correlate with LD patterns

Freshwater



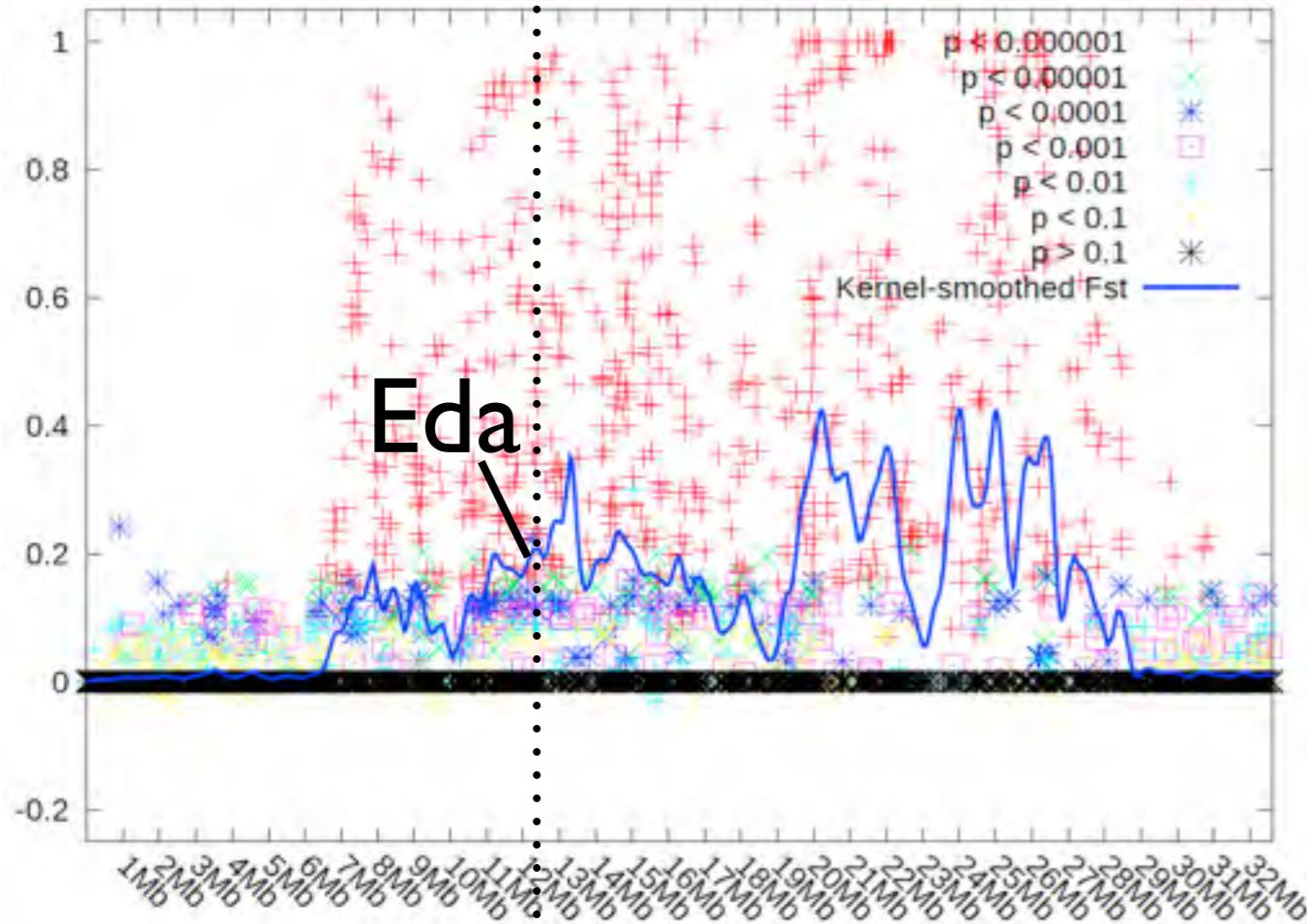
iHH



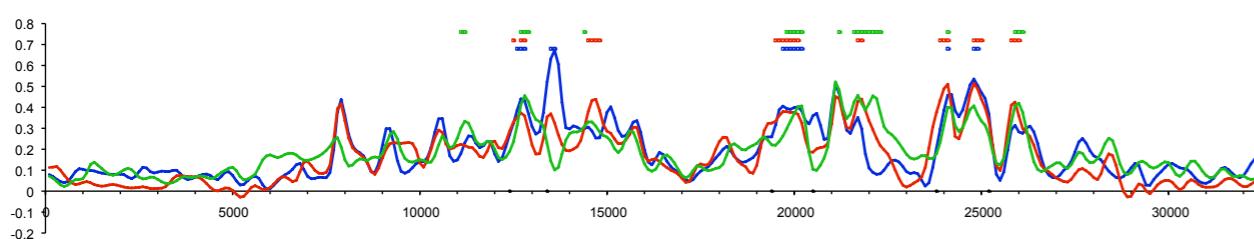
Ocean

Plates

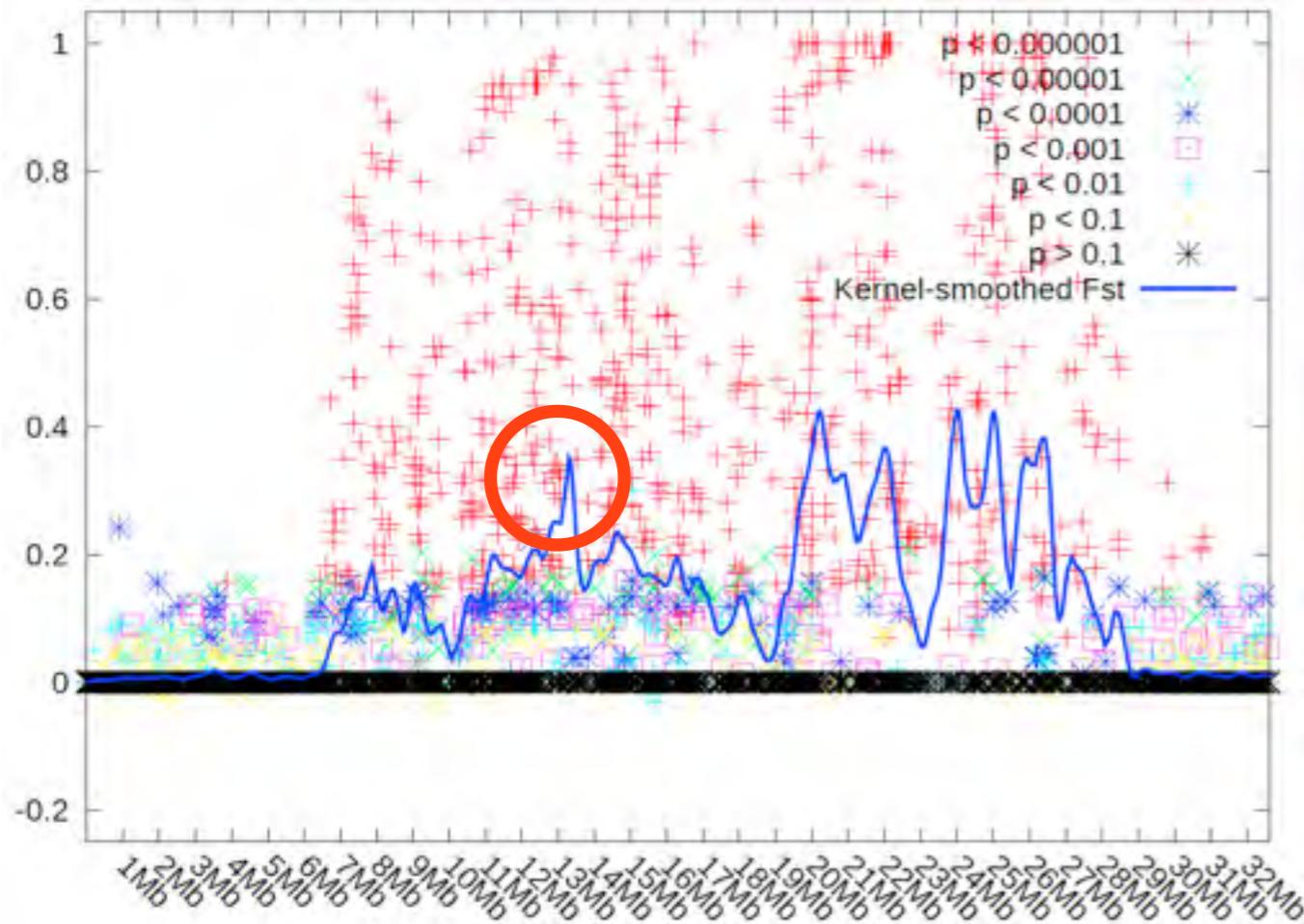
Position (Mb)



Fresh vs Marine



Bear Paw Lk
Boot Lk vs Marine
Mud Lk



HBEGF -
renal/cardiac response to
hyperosmotic conditions

Enigma -
involved in dermal bone
development

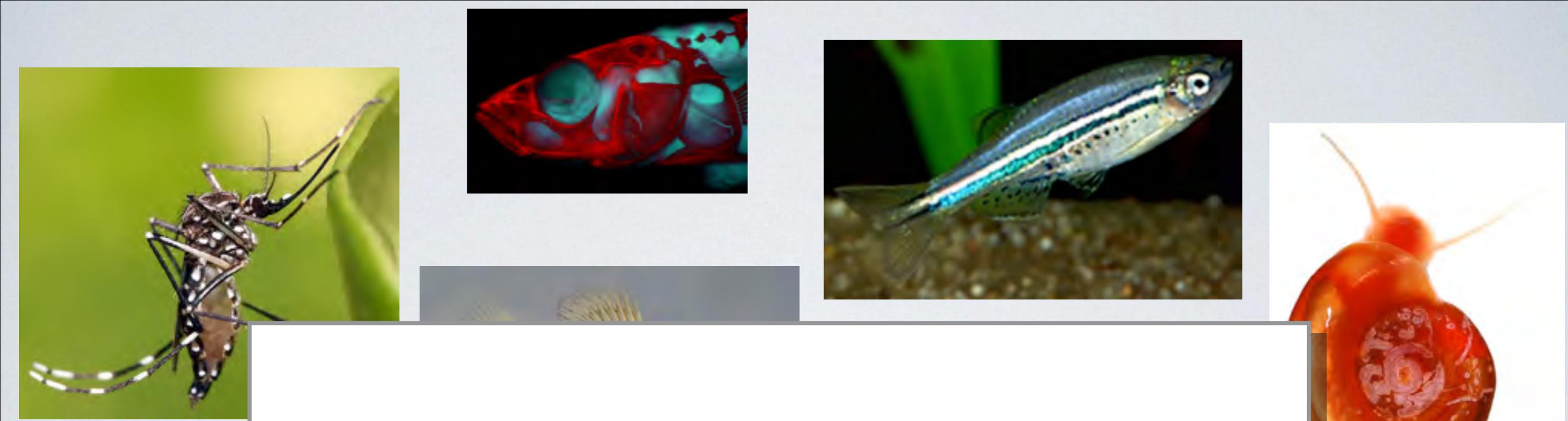


Overall Conclusions

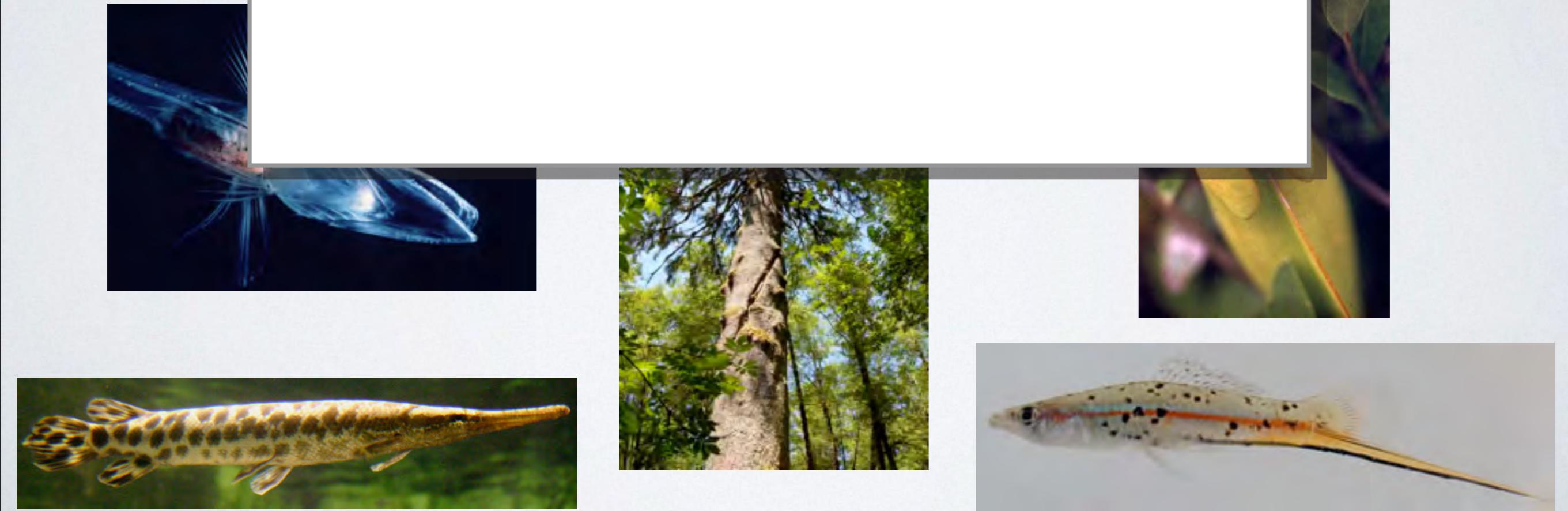
- Stickleback can evolve in decades largely through the reuse of standing genetic variation and geographically mediated balancing selection
- Signatures of selection are heterogeneous across the genome, but strikingly similar across populations
- Genome architecture varies extensively across stickleback and is associated with signatures of selection in divergent habitats
- Loci important for local adaptation appear to be genomically localized due to the segregating genomic architecture variation

Implications

- Ecological factors are very important for the tempo and mode of rapid adaptation and genome evolution
- The standing genetic variation is a product of a long evolutionary history and is associated with standing genomic architecture variation
- Present alleles of large effect are likely the product of many mutations across linked loci
- The evolved genetic and genomic architecture may significantly influence present patterns (e.g. parallel evolution) and future evolvability (e.g. speciation)



Considerations for RAD-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

Tradeoffs:

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

raw reads / samples / sites = coverage at each RAD locus

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

25 to 50x average coverage per RAD locus is a good goal

Experimental design considerations for RAD

Tradeoffs:

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

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 <i>C. remanei</i> (135 Mb)	actual EcoRI sites = 73,200

Experimental design considerations for RAD

Tradeoffs:

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

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

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

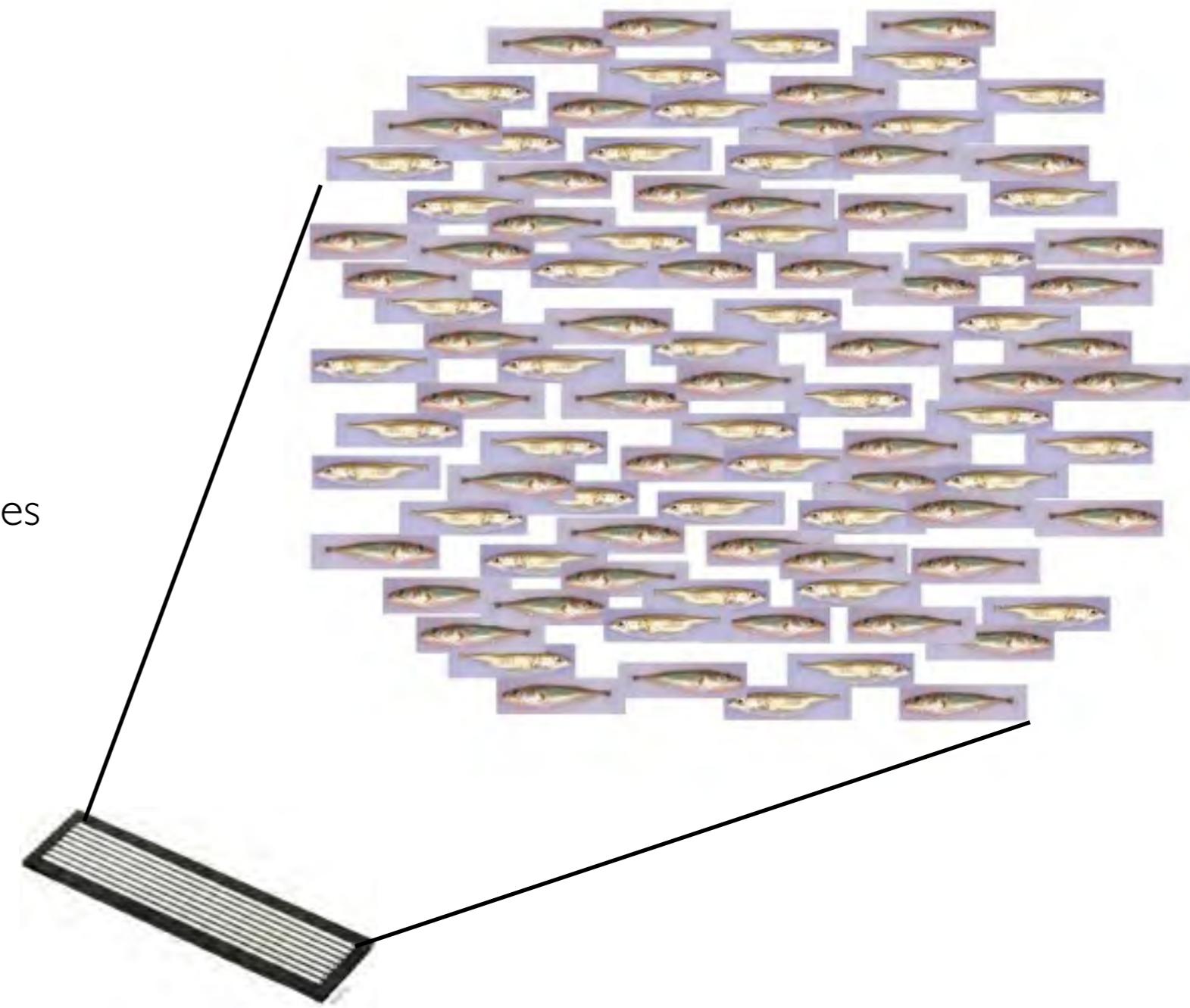
Barcoded adapters

5 to 8nt barcodes

Variable length barcodes

Combinatorial barcodes (PE)

Barcode distance - two mismatches



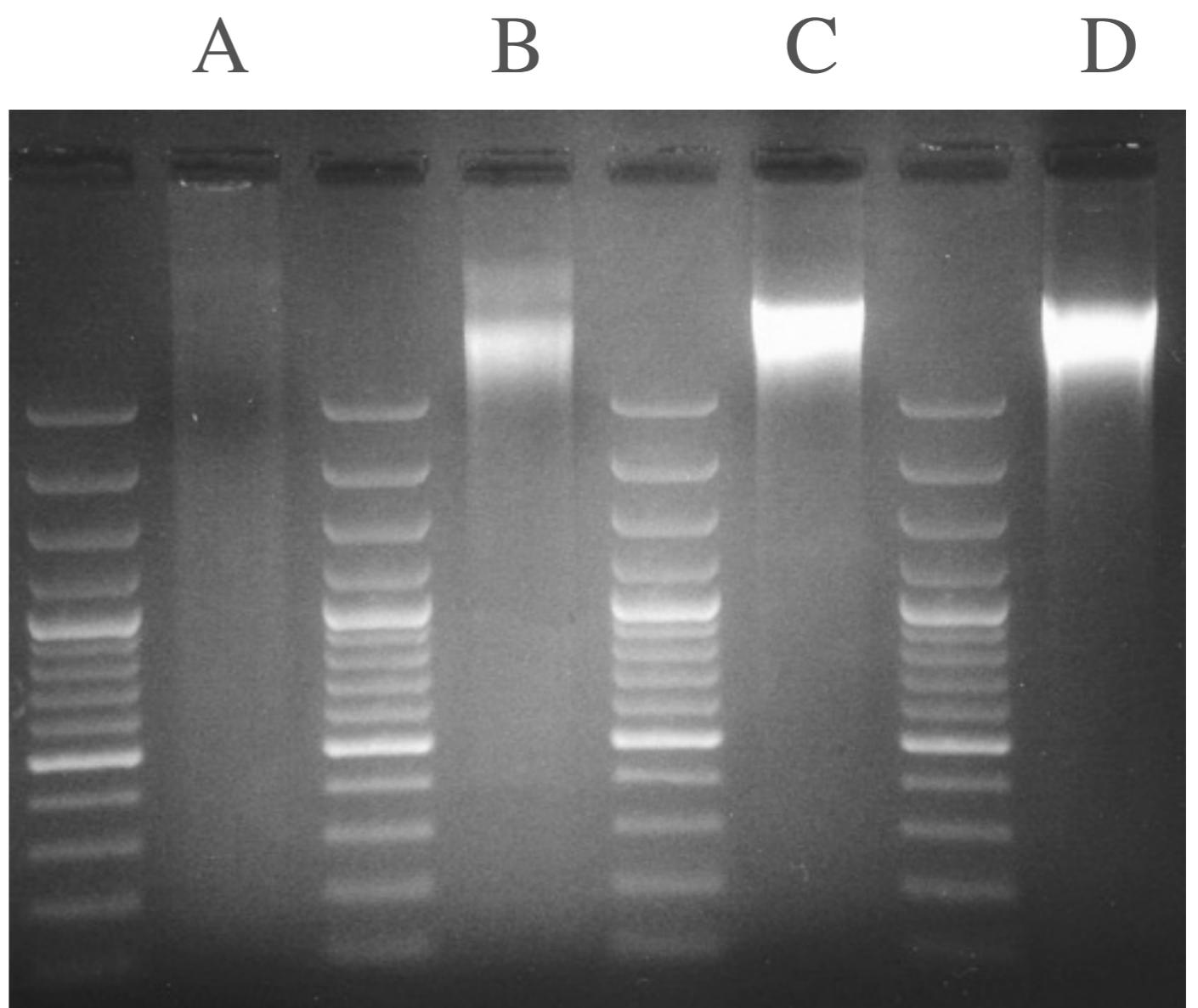
Molecular considerations in library building

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?

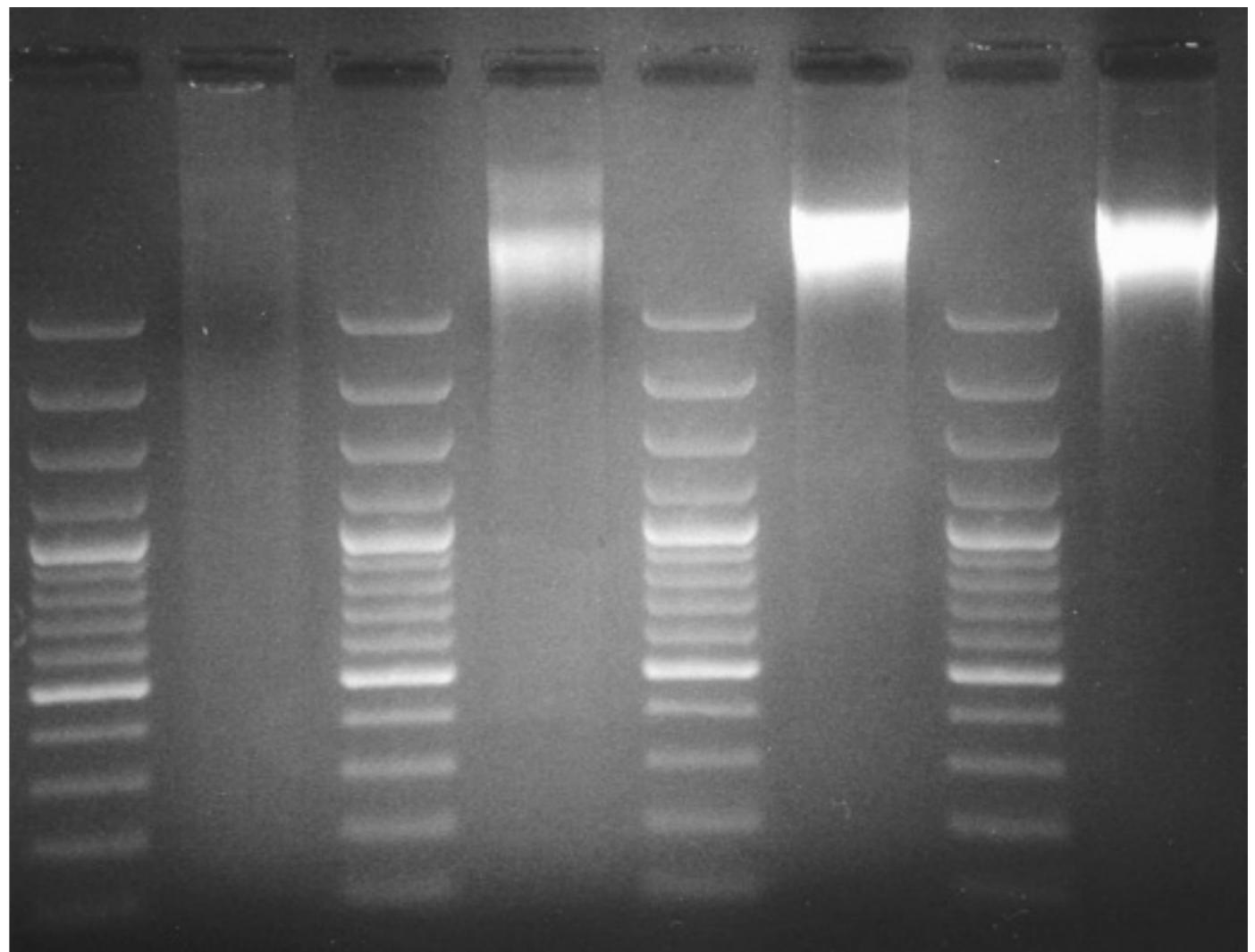
Things to consider

DNA Quality

[Diversify barcodes](#)

Illumina cluster calling is confused by repetition in first 4 bases - can offset barcodes

CGATA GTACA TAGCC ACTGC



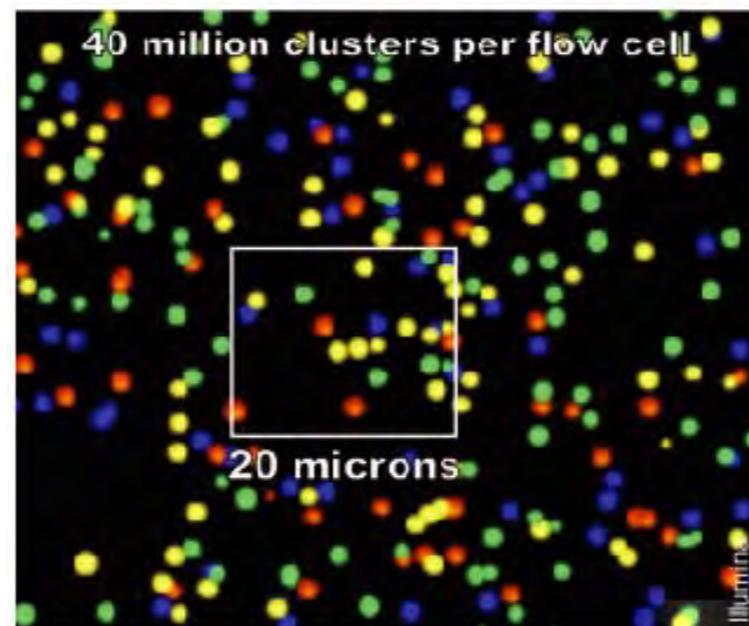
Molecular considerations in library building

How can I get the best depth of coverage?

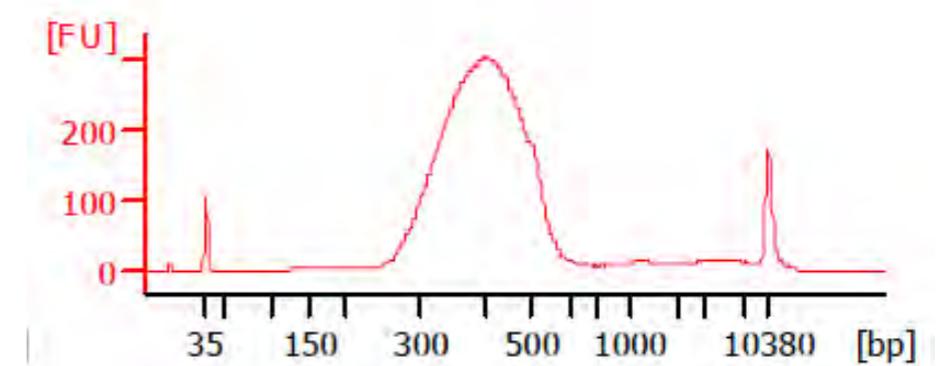
Things to consider

Fragment size

Smaller/tighter is better



Agilent Bioanalyzer



Molecular considerations in library building

How can I get the best depth of coverage?

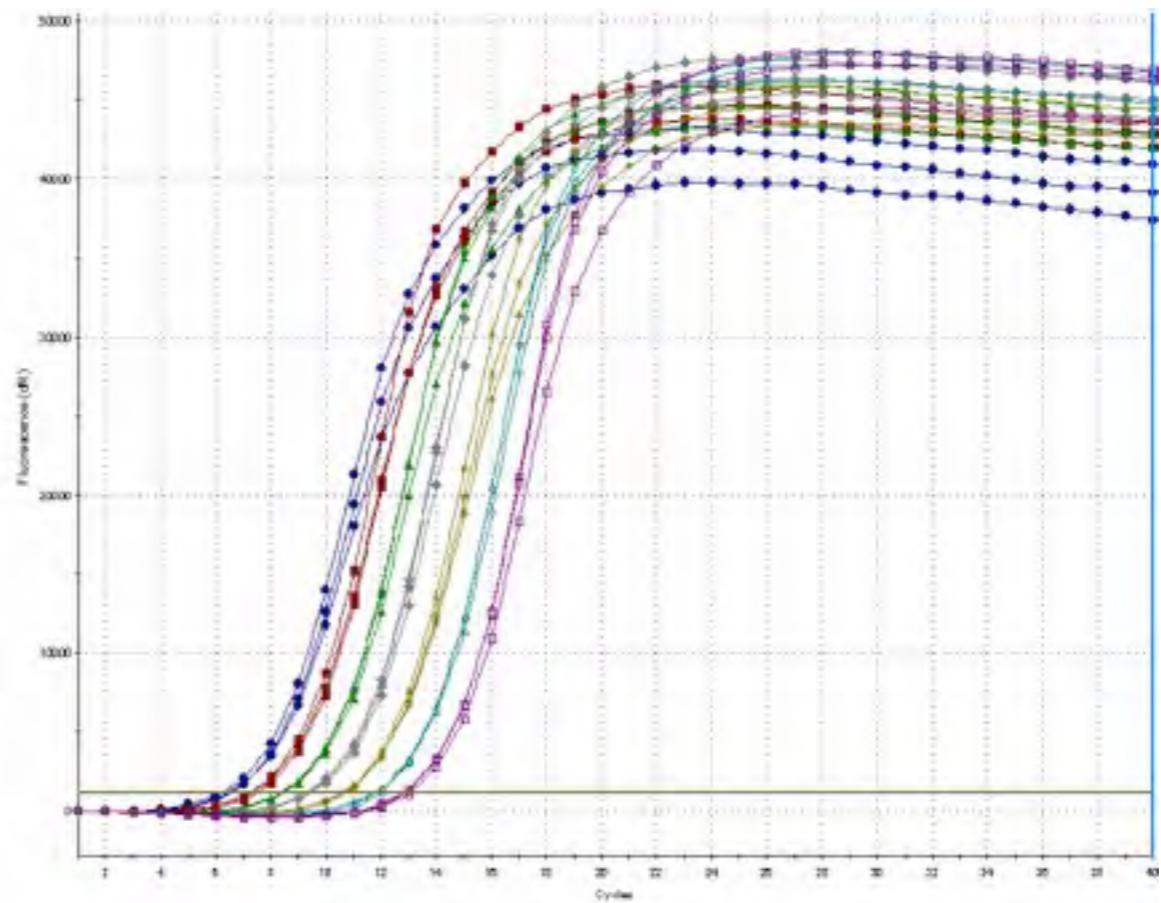
Things to consider

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

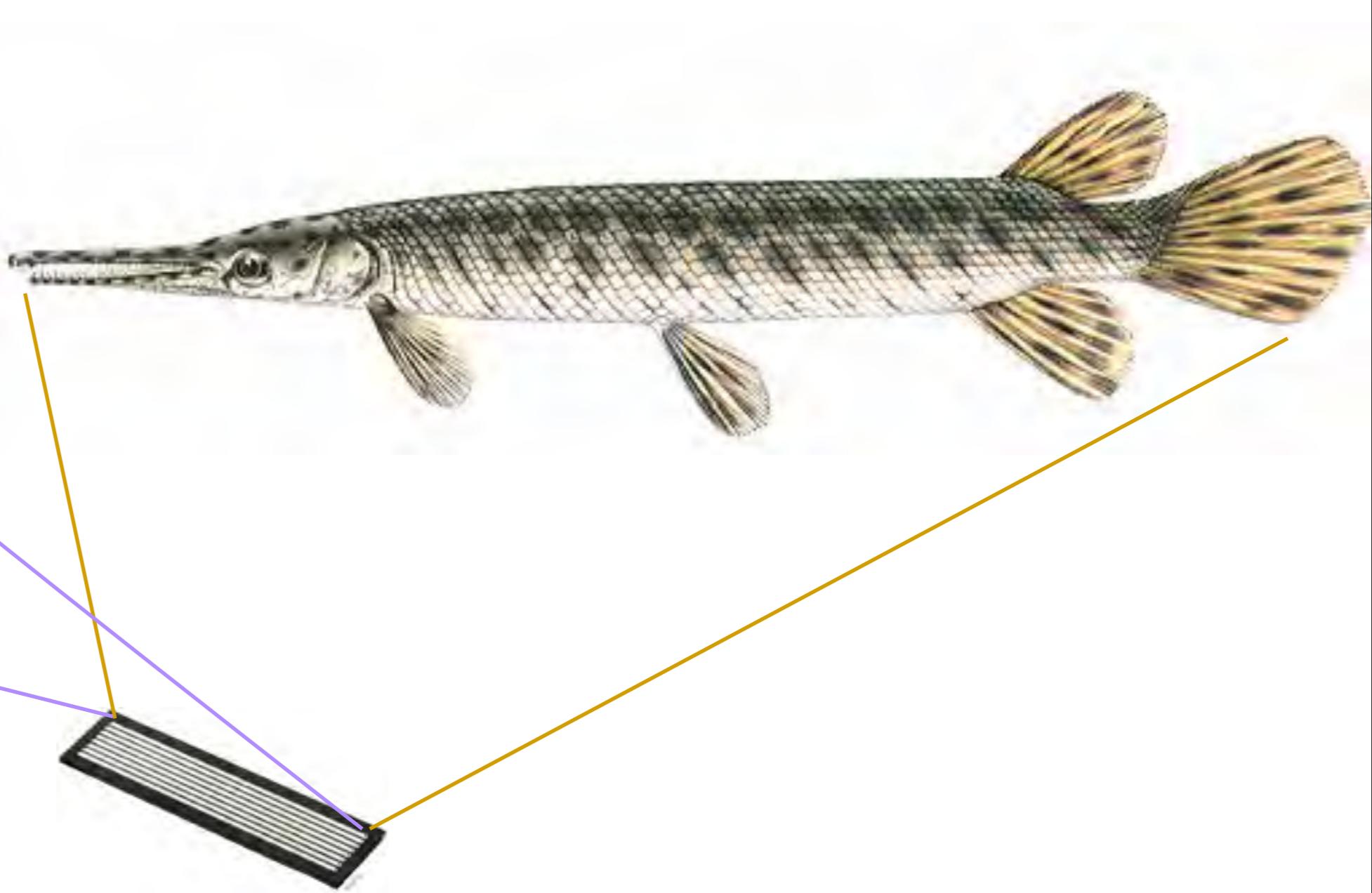
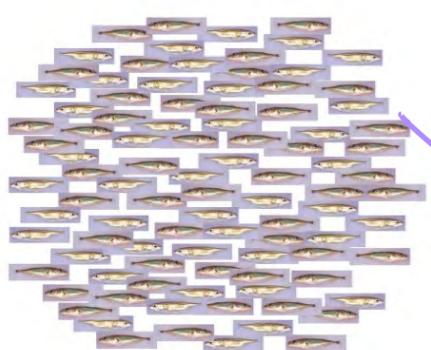
Fragment size

[Library quality](#)

qPCR

Pilot Experiment:

Spike or split a lane



Statistical considerations in RAD-seq

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

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

The reads are 14T and 2G:

GT heterozygote?

GG homozygote with error?

AA homozygote with lots of error?

Needed a rigorous method to call genotypes

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

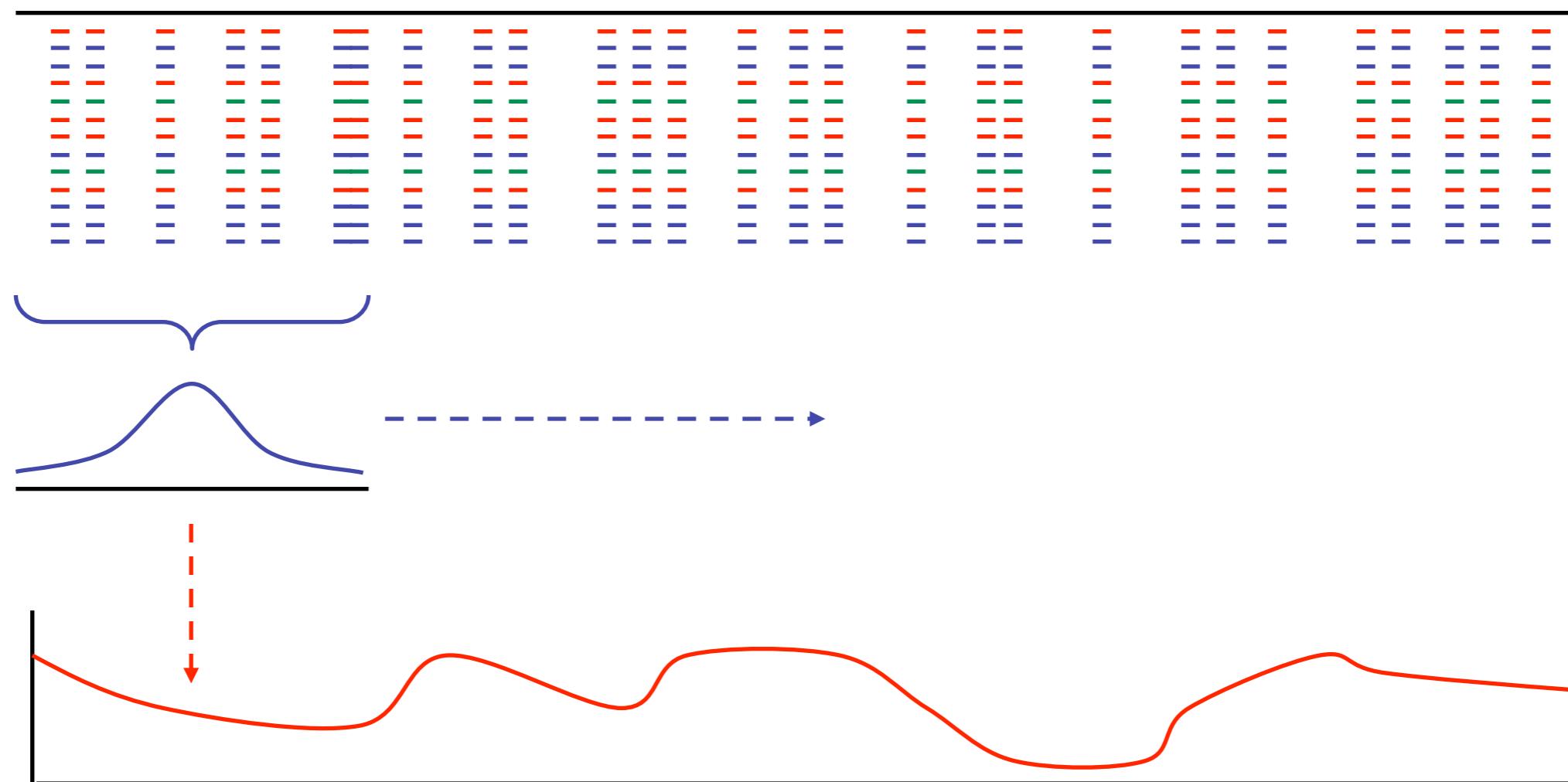
$$L(n_1 \text{ hom}) = P(n_1, n_2, n_3, n_4) = \frac{n!}{n_1! n_2! n_3! n_4!} \left(1 - \frac{3\epsilon}{4}\right)^{n_1} \left(\frac{\epsilon}{4}\right)^{n_2} \left(\frac{\epsilon}{4}\right)^{n_3} \left(\frac{\epsilon}{4}\right)^{n_4}$$

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

Maximum likelihood genotyping based on multinomial distribution of nucleotide reads

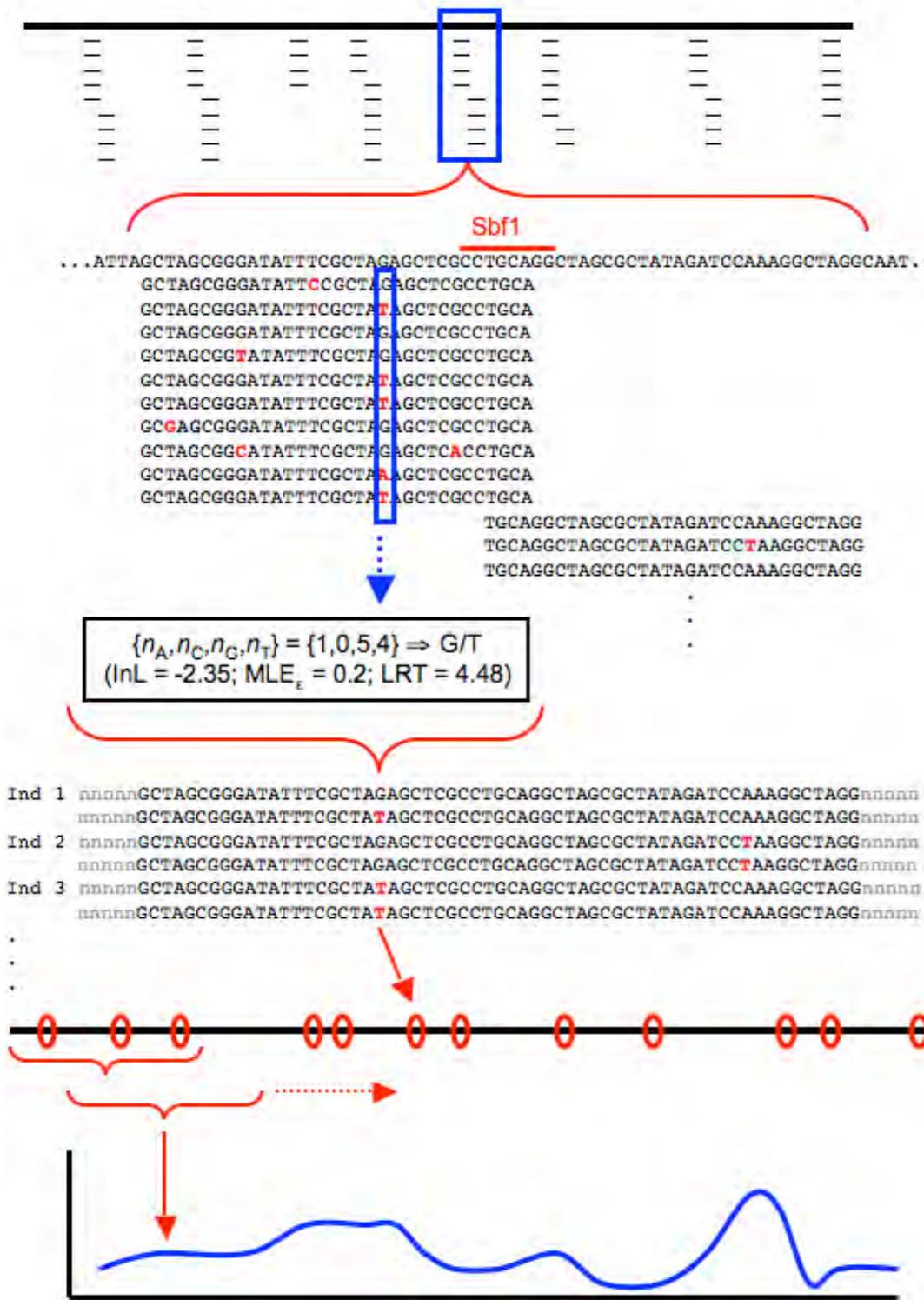
Making statistics continuous across the genome

Kernel-smoothing average of summary statistics along genome

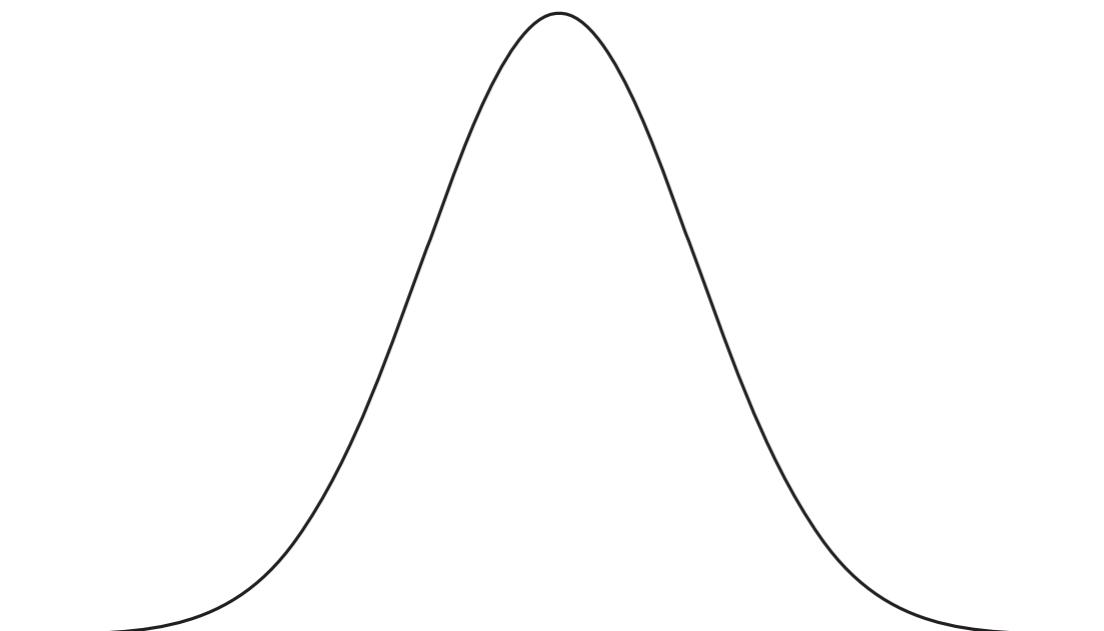


Bootstrap re-sampling to estimate significance of moving average

Overall pipeline



‘Bias’ in RAD-sequencing

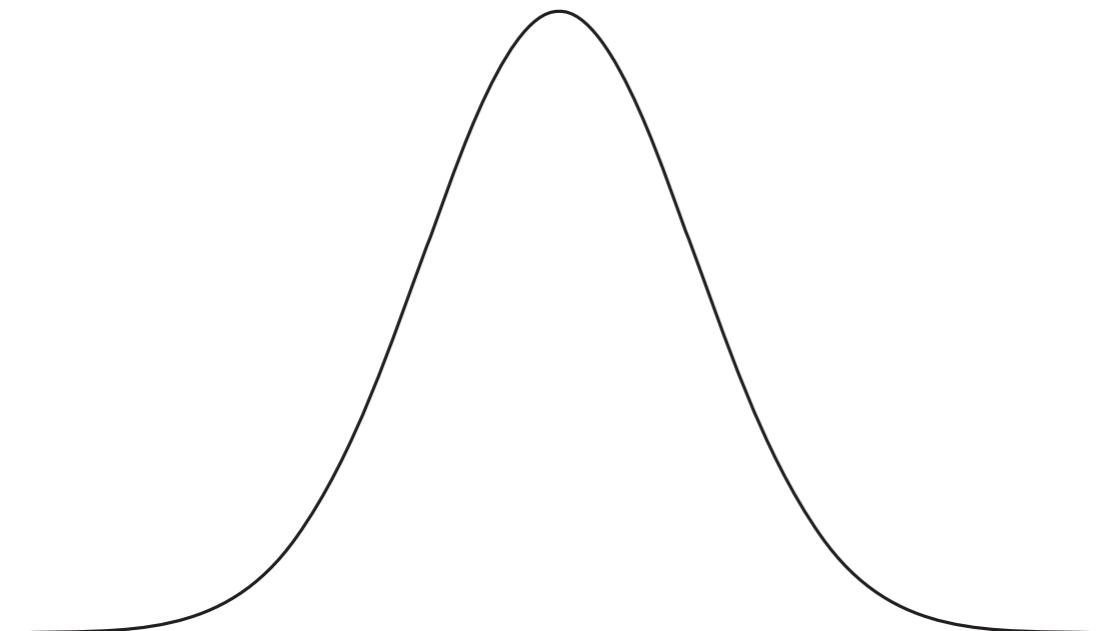


$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

$e = 2.7182\dots$

$\pi = 3.1415\dots$

‘Bias’ in RAD-sequencing



$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

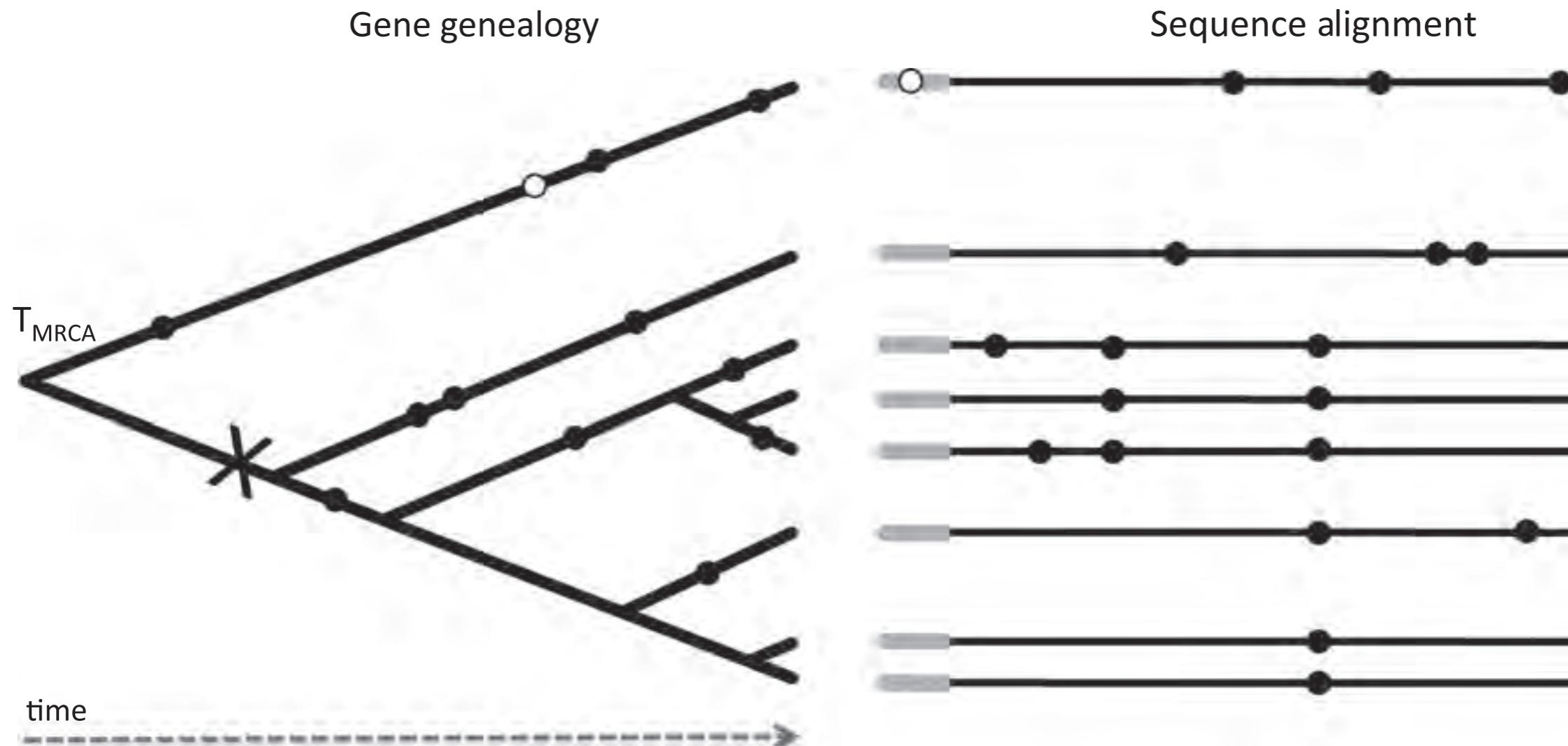
$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$$

Bias in RAD-sequencing

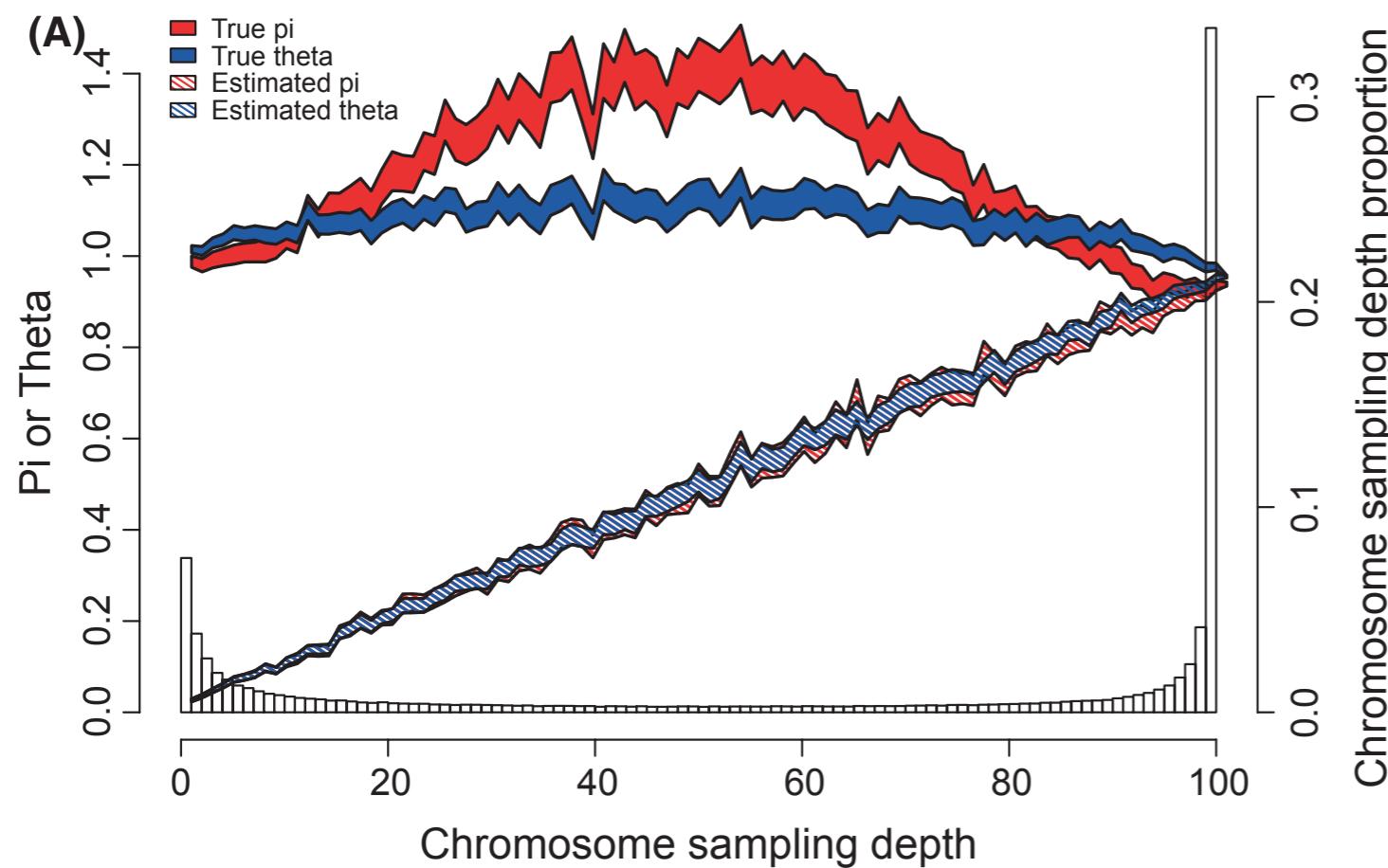
RADseq underestimates diversity and introduces genealogical biases due to nonrandom haplotype sampling

B. ARNOLD,¹ R. B. CORBETT-DETIG,¹ D. HARTL and K. BOMBLIES

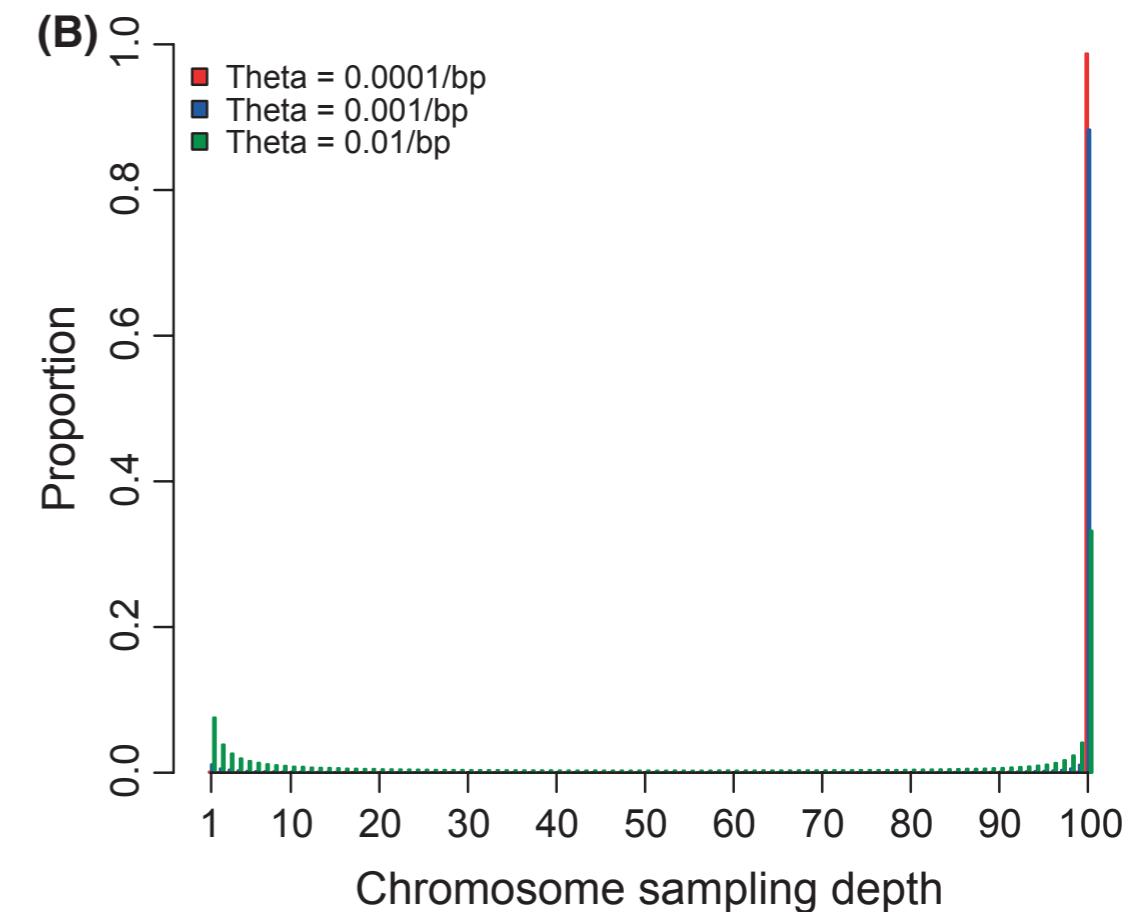
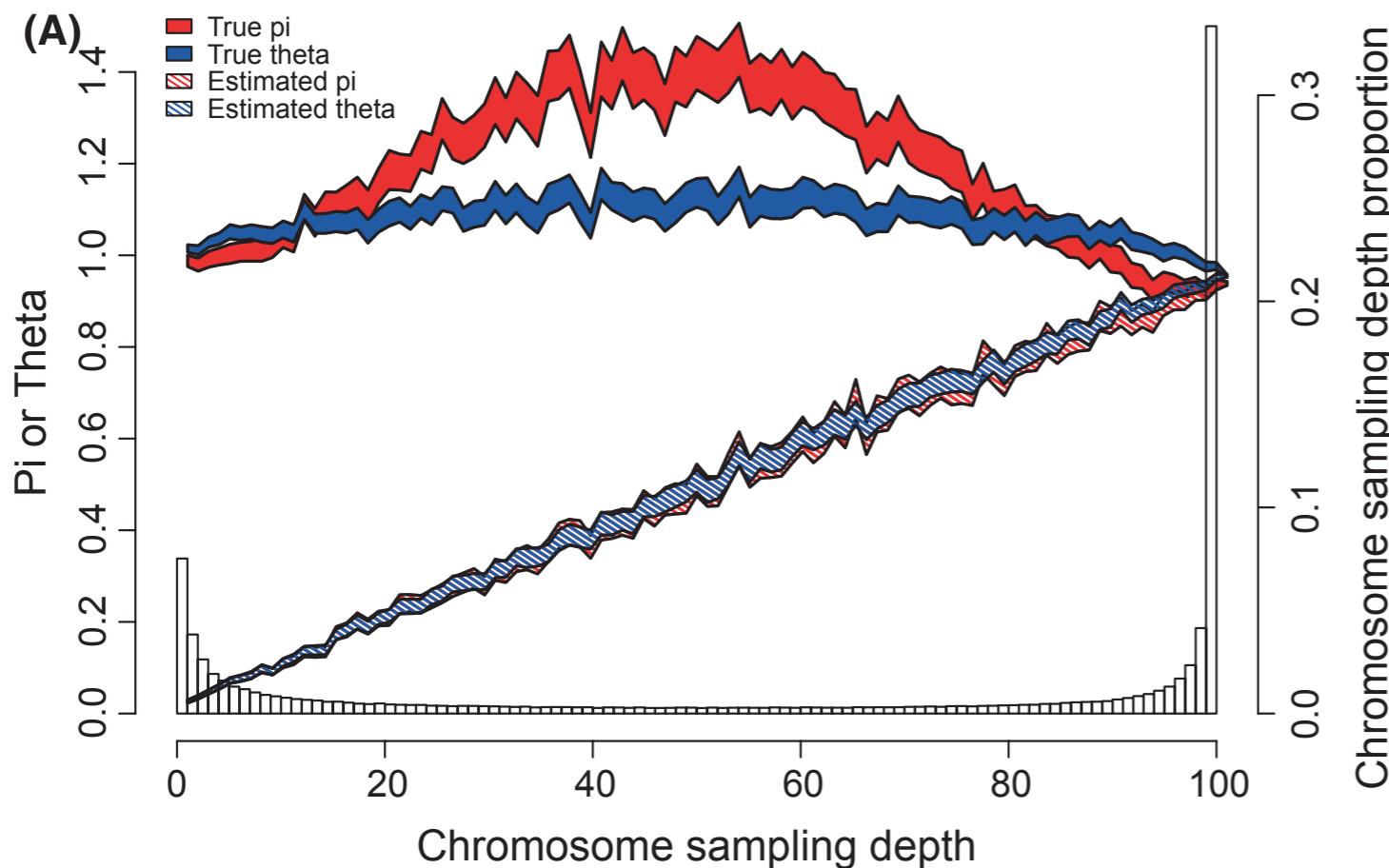
Department of Organismic and Evolutionary Biology, Harvard University, Cambridge, MA 02138, USA



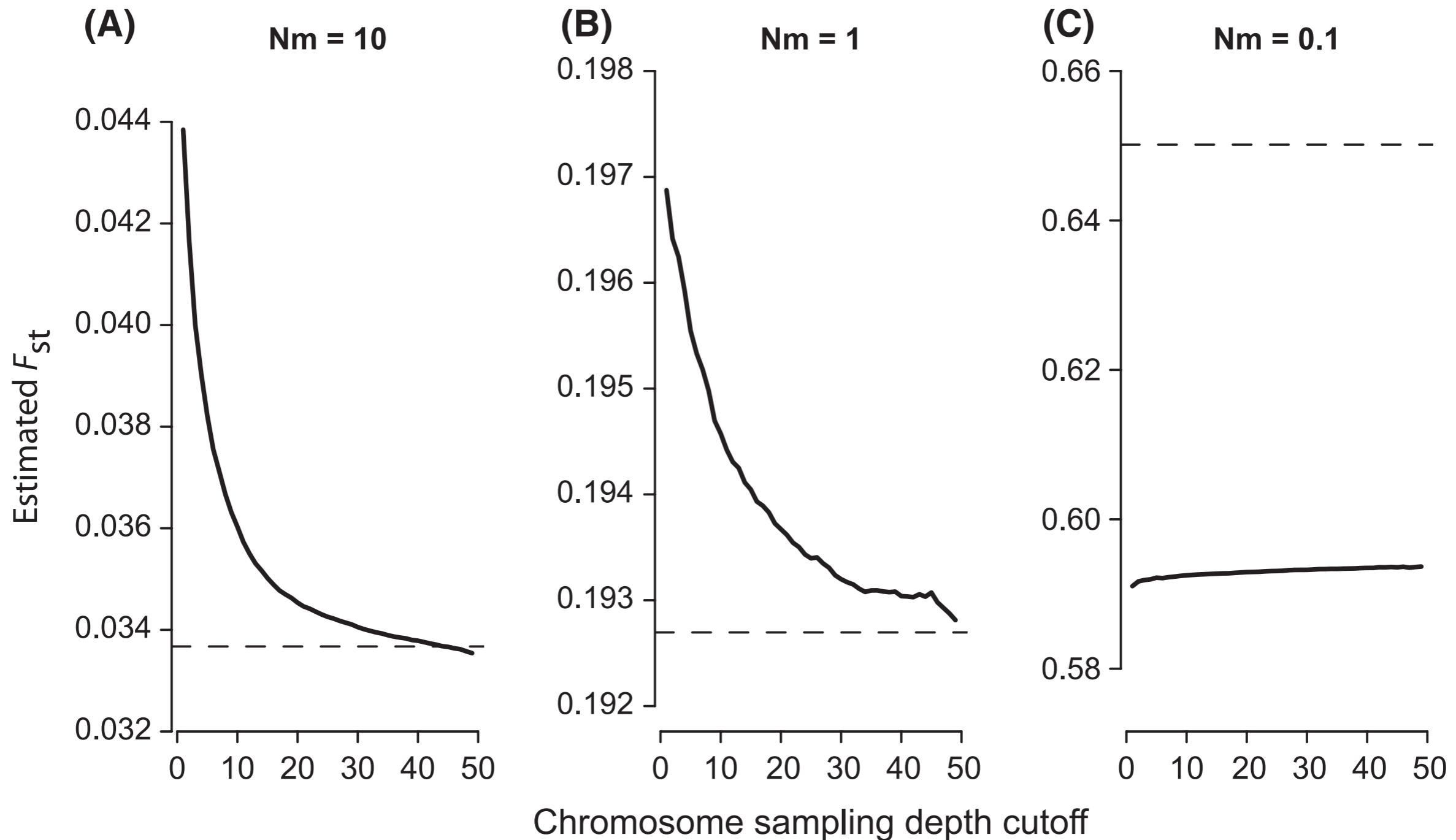
Bias in RAD-sequencing; genetic diversity



Bias in RAD-sequencing; genetic diversity



Bias in RAD-sequencing; Fst



Bias in RAD-sequencing summary

Protocol	θ per bp	Mean	
		θ_{we}/θ_{wa}	π_e/π_a
Standard	0.0001	0.994	0.995
	0.001	0.987	0.982
	0.01	0.956	0.933
Double digest	0.0001	0.835	0.836
	0.001	0.858	0.851
	0.01	0.829	0.797

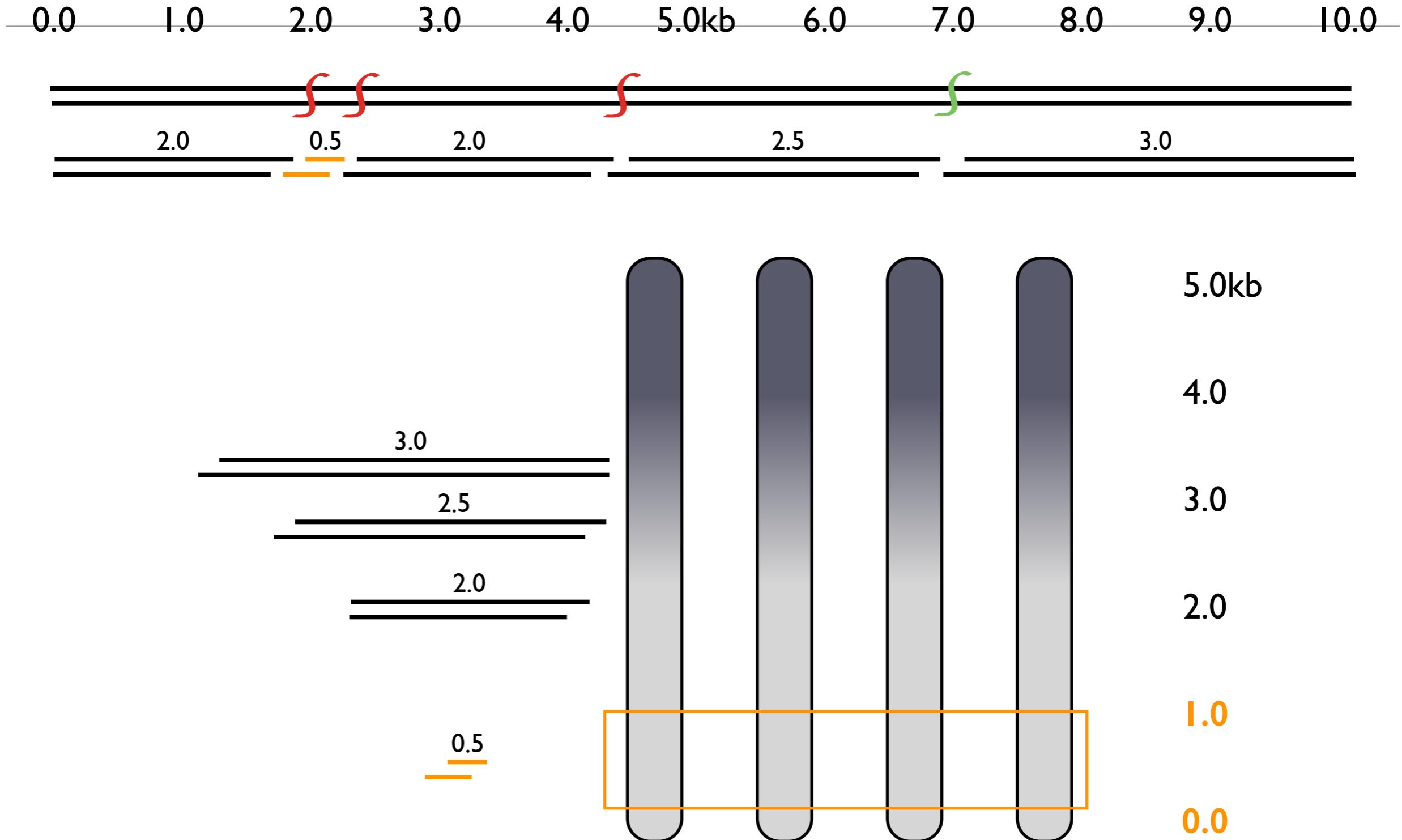
Bias in RAD-sequencing summary

Protocol	θ per bp	Mean	
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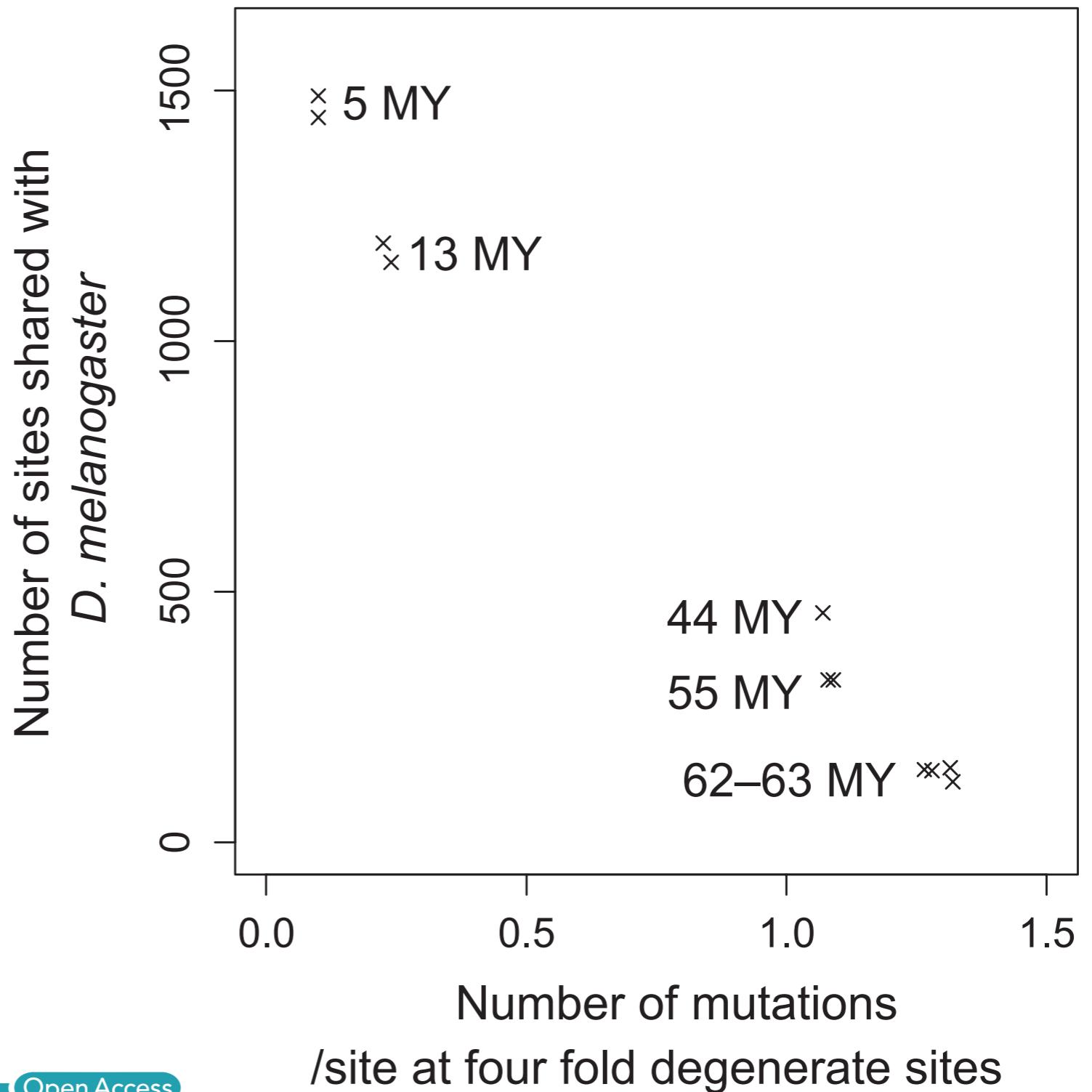
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Double digest	0.0001	0.835	0.836
	0.001	0.858	0.851
	0.01	0.829	0.797

Why is ddRAD so much more biased?



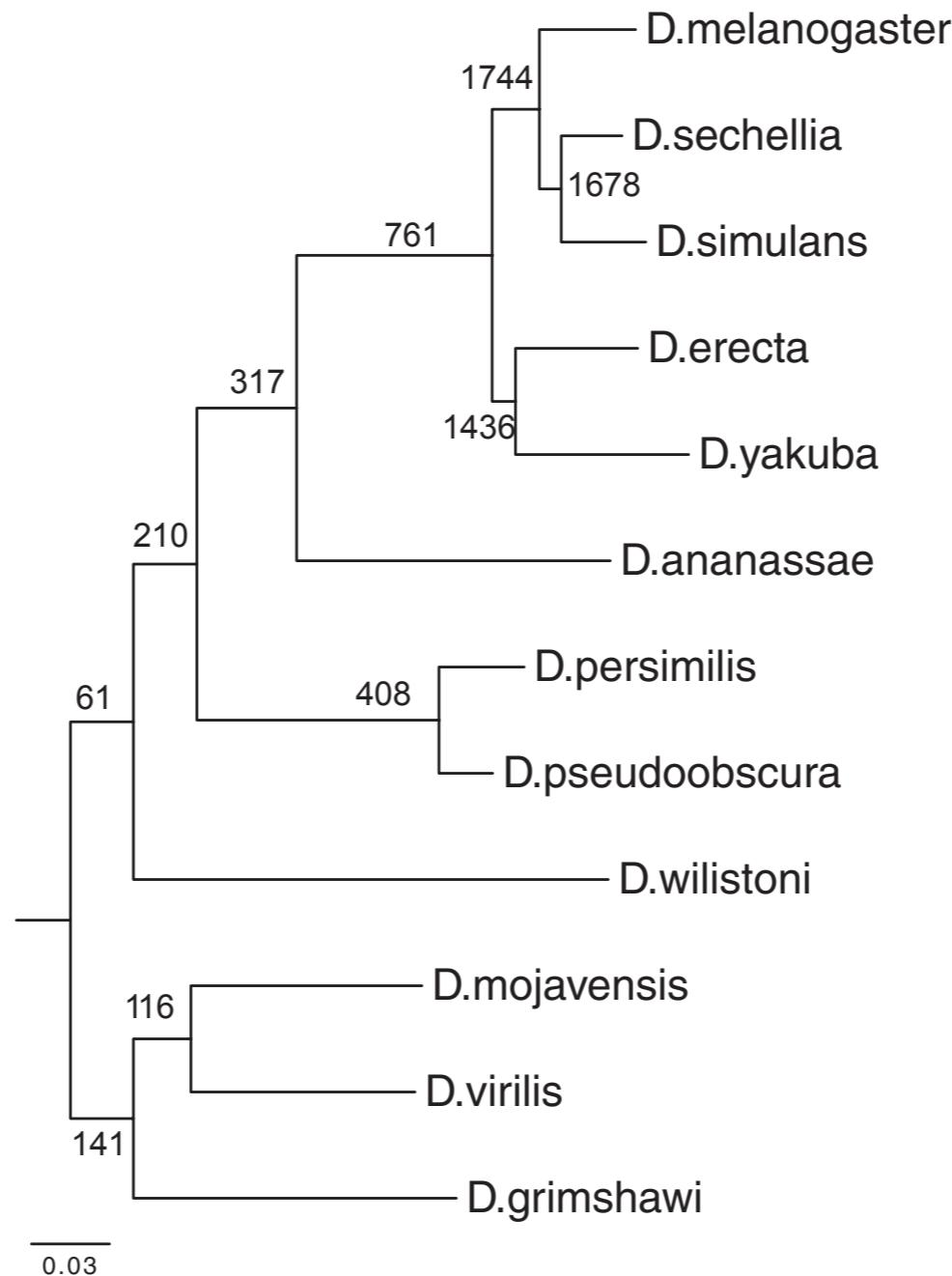
RAD-seq and phylogenetics of divergent species



RAD-seq and phylogenetics of divergent species

Species pair <i>D. melanogaster</i>	Node depth (My)	Orthologous tags	Retrieved orthologous tags (%)	In clusters including paralogs (%)
<i>D. sechellia</i>	5.4	2978	99	5
<i>D. simulans</i>	5.4	2892	99	4
<i>D. erecta</i>	12.6	2390	97	3
<i>D. yakuba</i>	12.8	2314	97	8
<i>D. ananassae</i>	44.2	916	68	9
<i>D. persimilis</i>	54.9	648	65	9
<i>D. pseudoobscura</i>	54.9	648	66	9
<i>D. wilistoni</i>	62.2	242	49	6
<i>D. grimshawi</i>	62.9	290	60	8
<i>D. virilis</i>	62.9	286	59	5
<i>D. mojavensis</i>	62.9	298	59	8

RAD-seq and phylogenetics



Ecology and Evolution

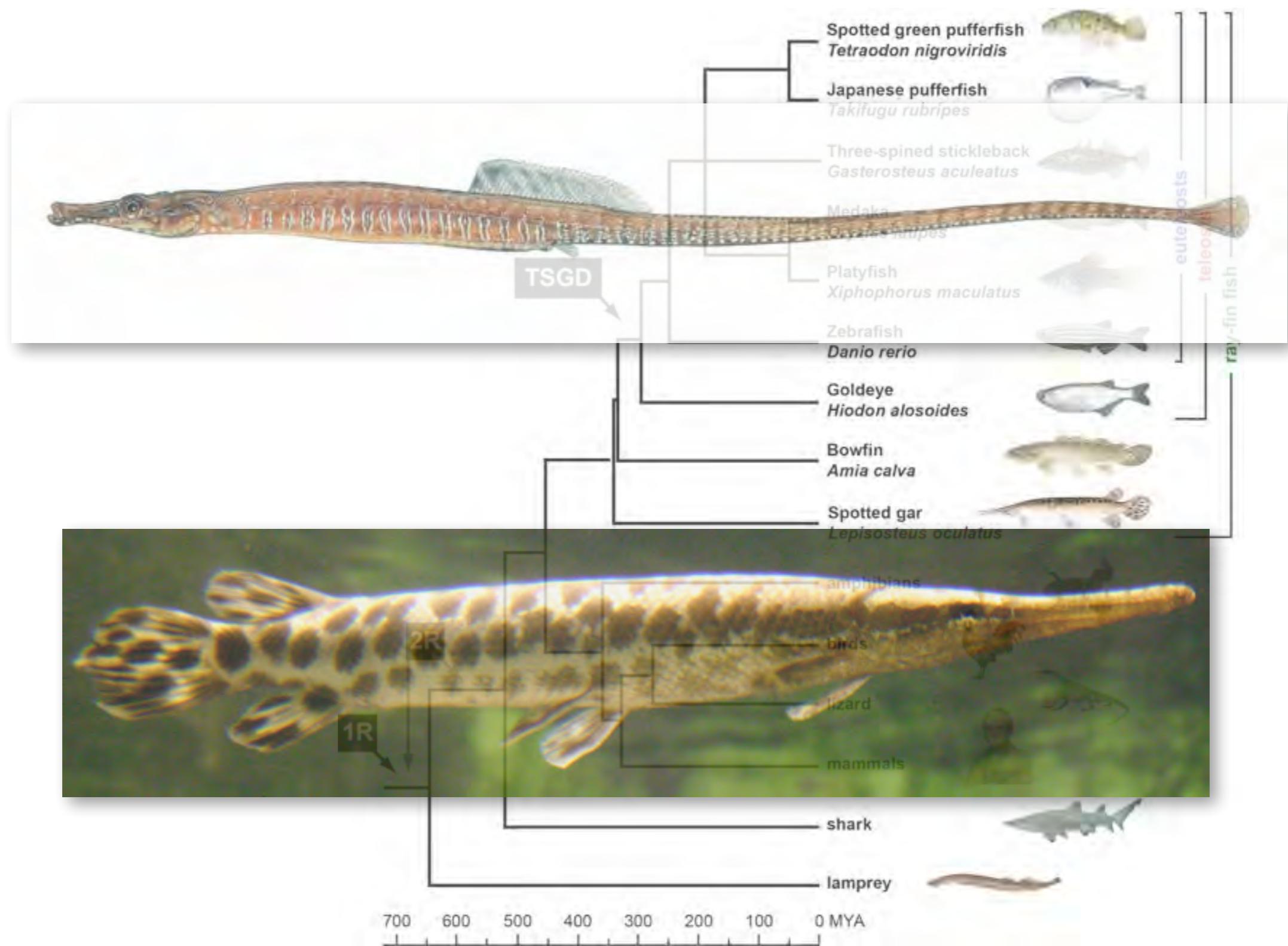
Open Access

Is RAD-seq suitable for phylogenetic inference? An *in silico* assessment and optimization

Marie Cariou, Laurent Duret & Sylvain Charlat

What if you don't have a genome sequence?

Genomically enabling very non-model organisms

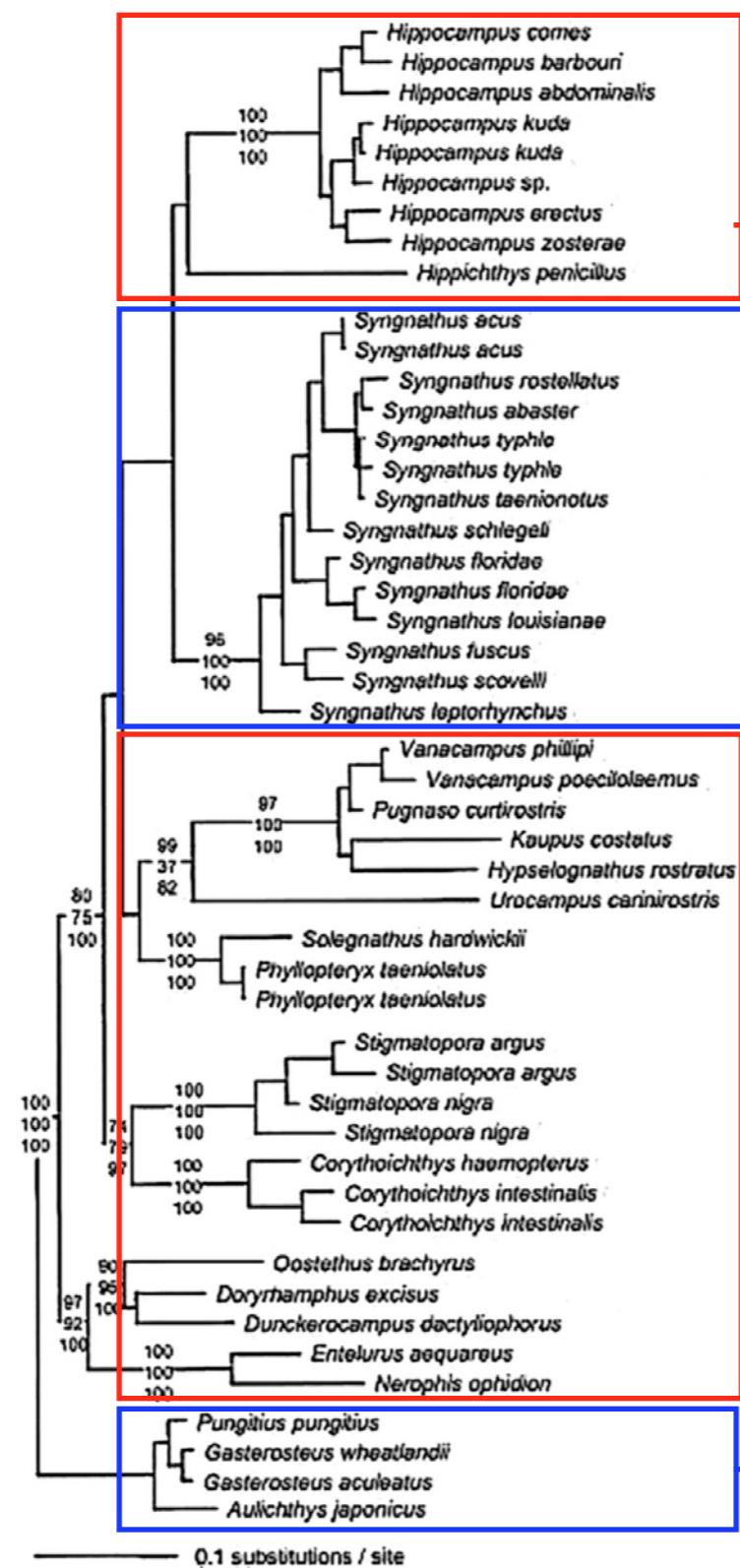


Andrew Nishida, Julian Catchen, Susie Bassham,
Clay Small and Adam Jones

Seahorses, sea dragons and pipefishes



Gasterosteidae and Syngnathidae are historically considered to be closely related



Seahorses



Pipefish



Seadragons



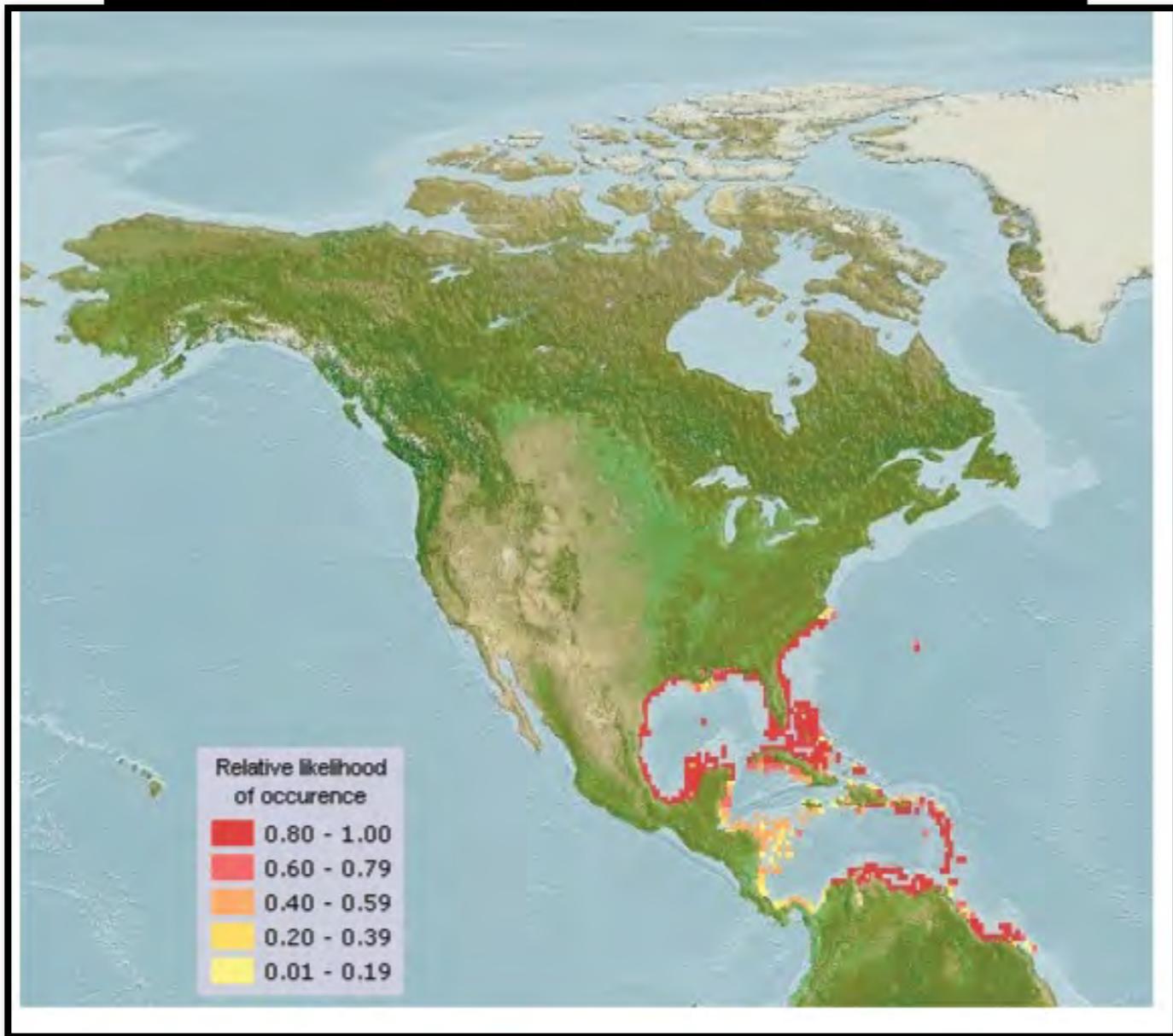
Stickleback





Gulf Pipefish

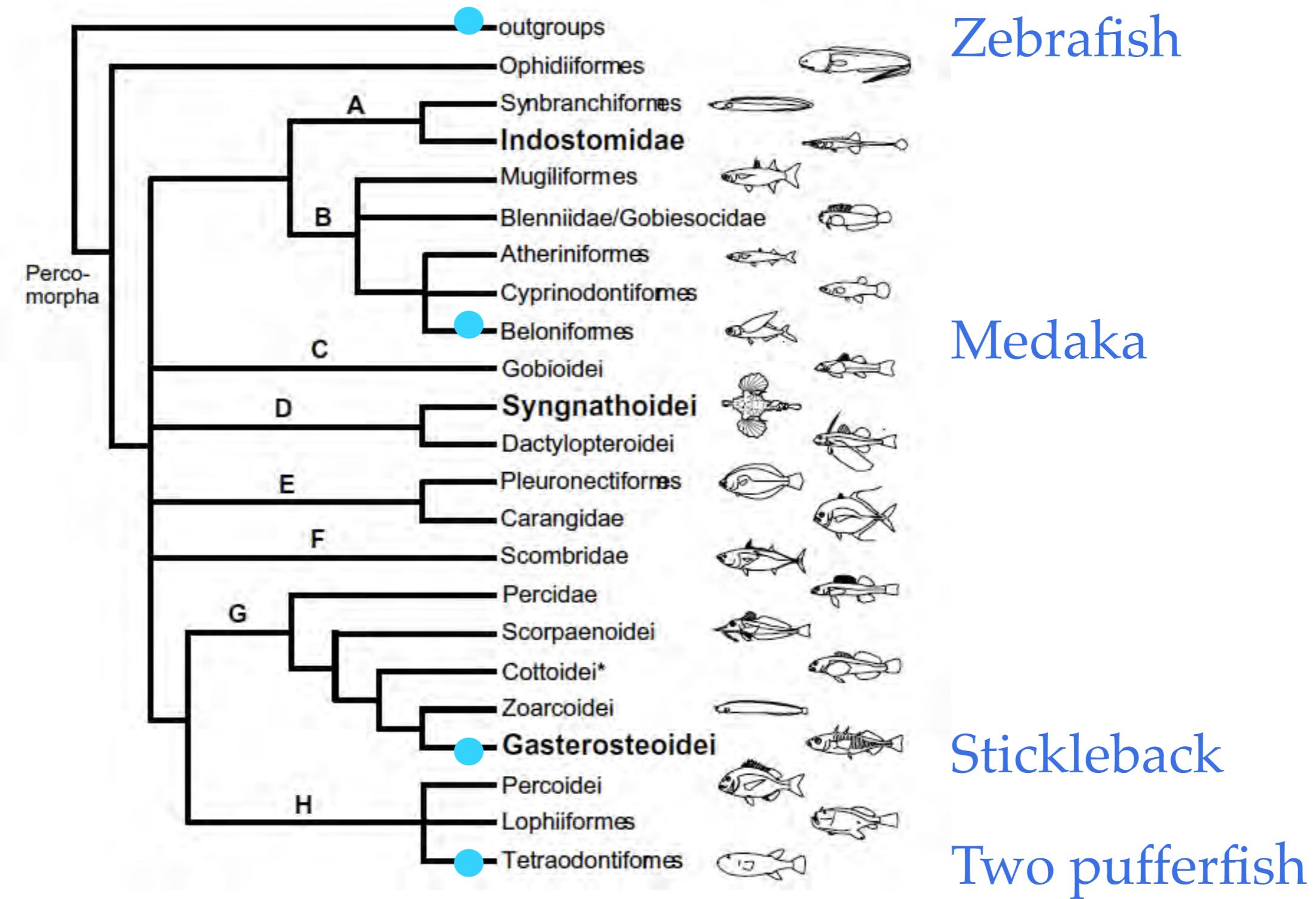
Syngnathus scovelli



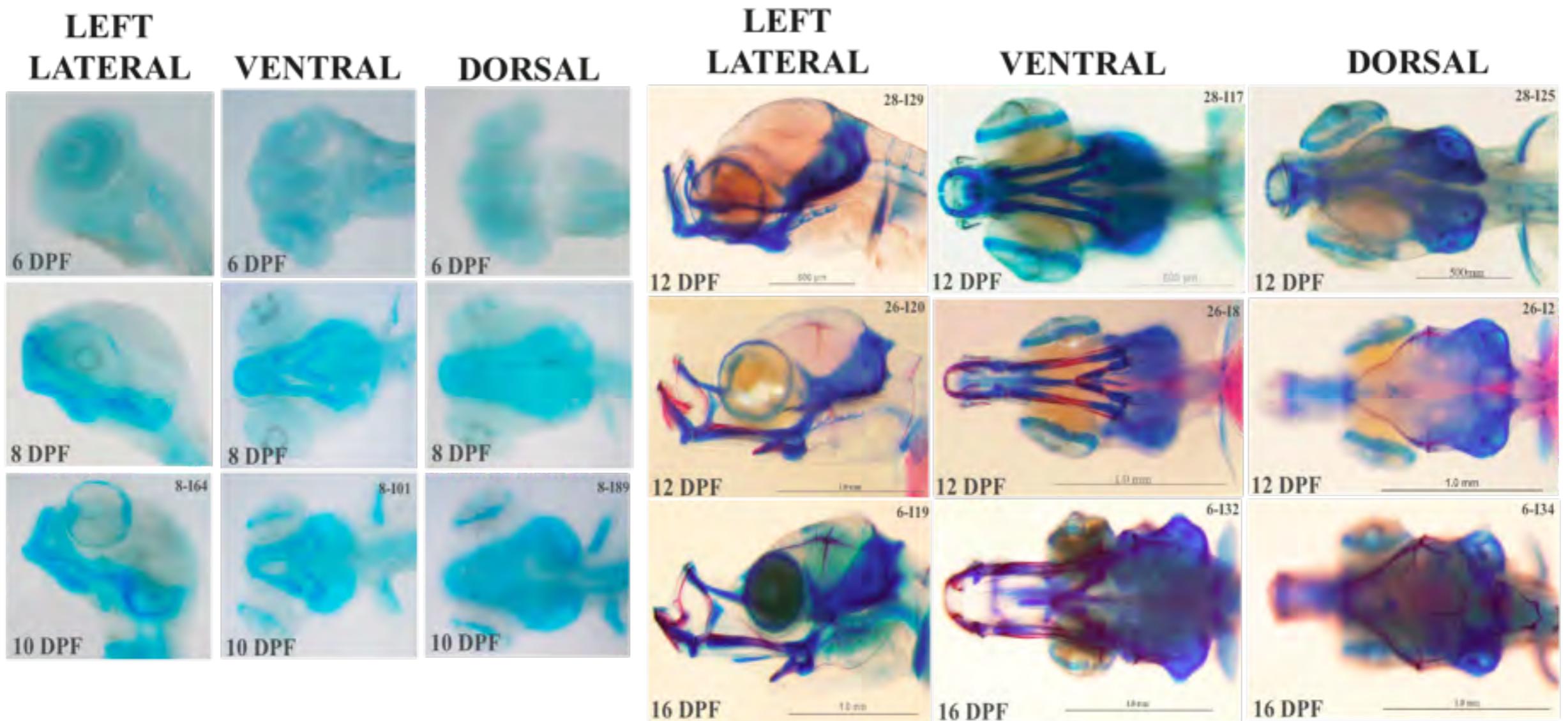
- 160 mm (6.3")
- reversed sex roles
- sexual dimorphism
- specialized suction feeding
- no sequences in international databases

Few teleost genomes are available

Gasterosteiformes: only stickleback



We're really interested in the head and body axis



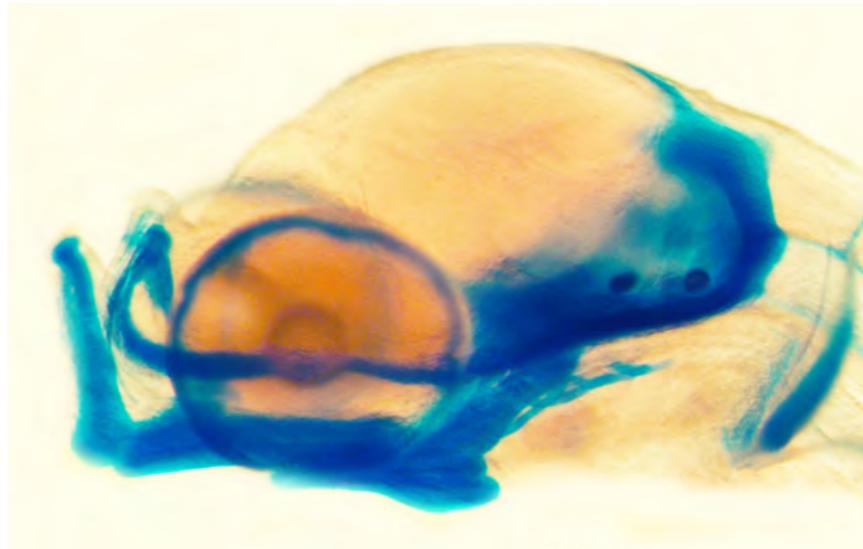
Solution: ‘genomically enable’ pipefish

- 1) A high quality transcriptome
- 2) Very dense RAD genetic map
- 3) Deep coverage shotgun sequencing of genome
- 4) Order genomic and transcriptomic contigs against
the RAD reference map

Pipefish Transcriptome



Building an EST database in pipefish



Pipefish embryonic mRNA



Illumina sequencing:
100 nt, paired-end

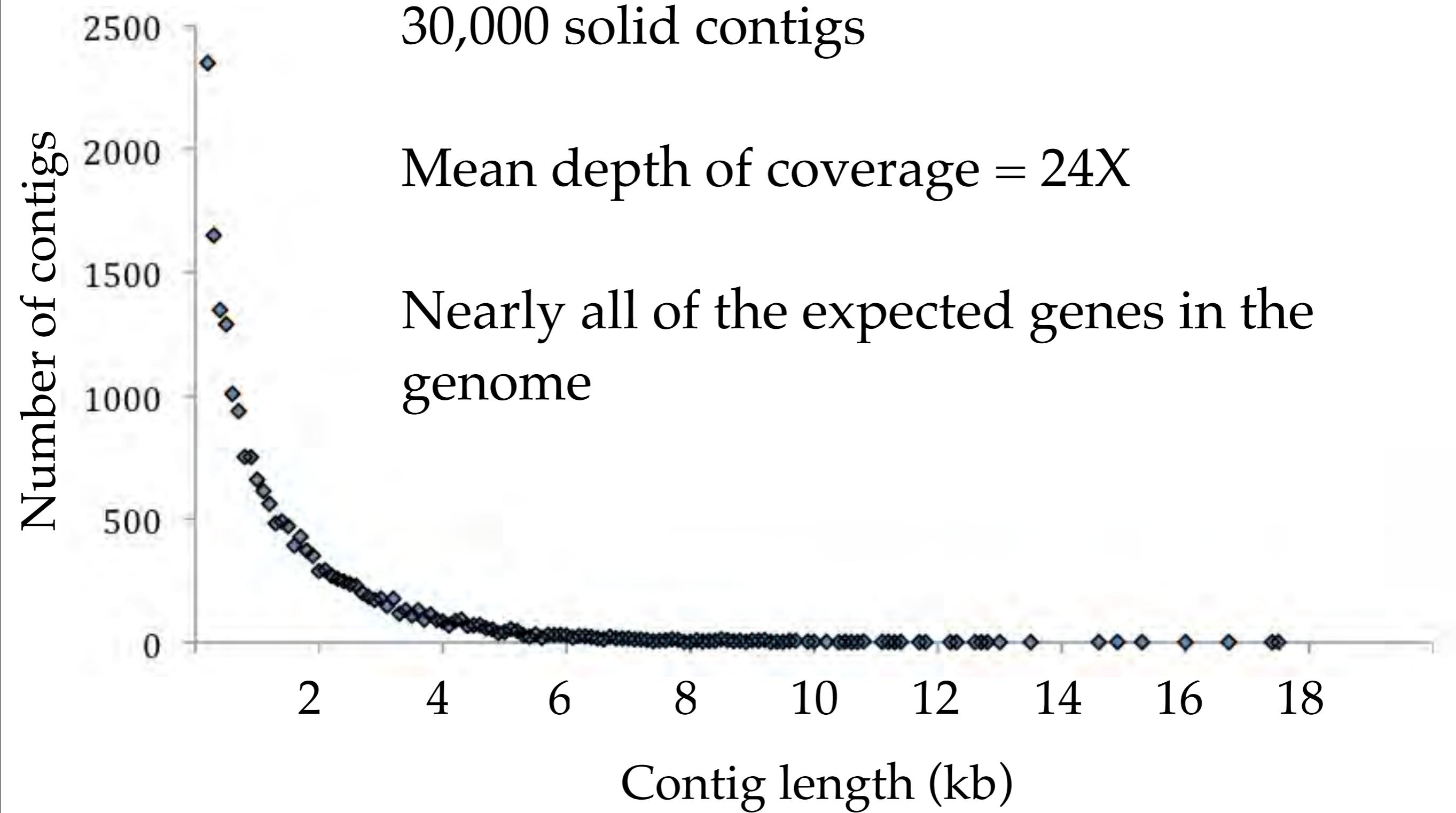


200 million reads (two lanes)

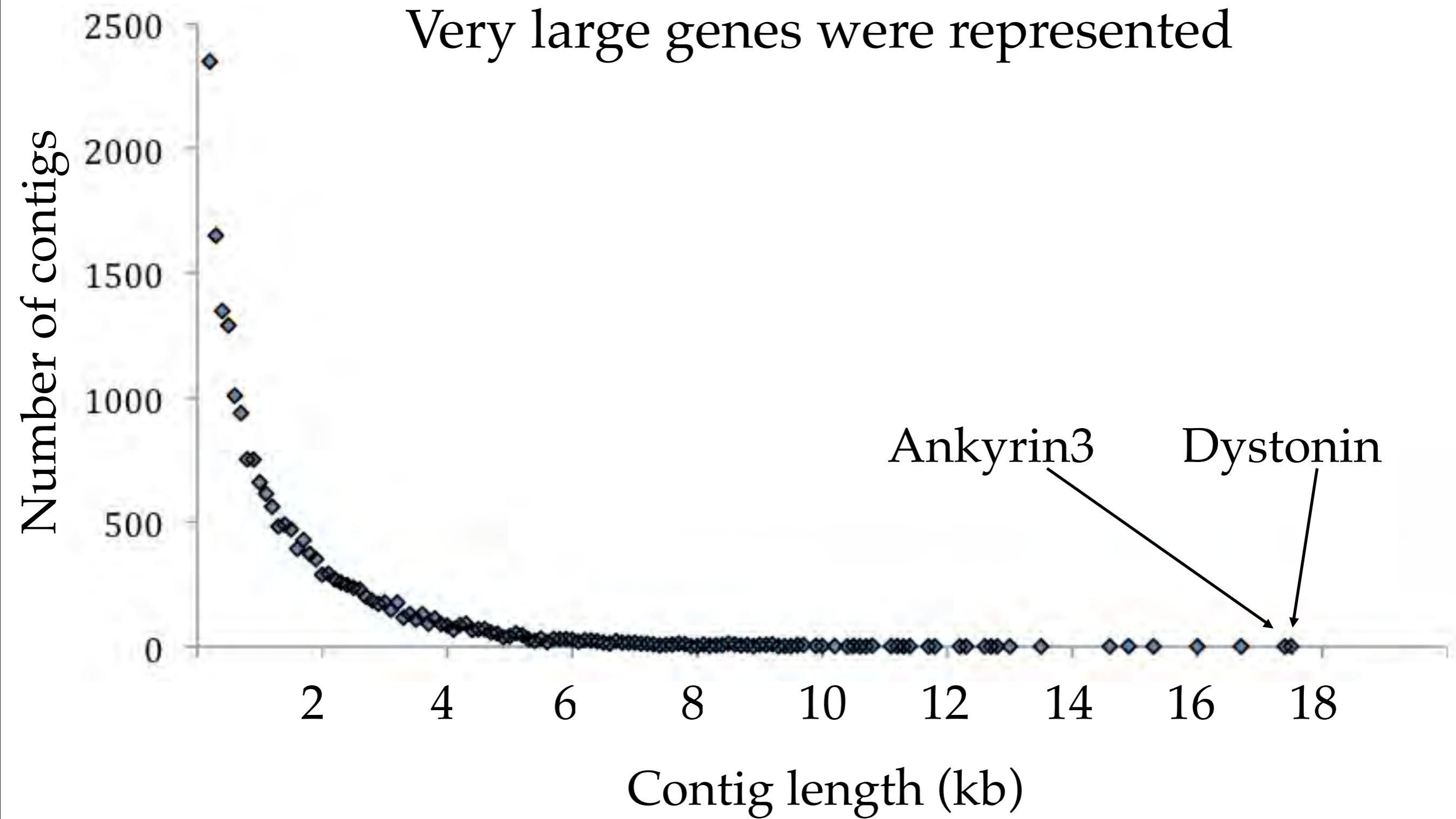


Assembly of transcripts

Transcriptome

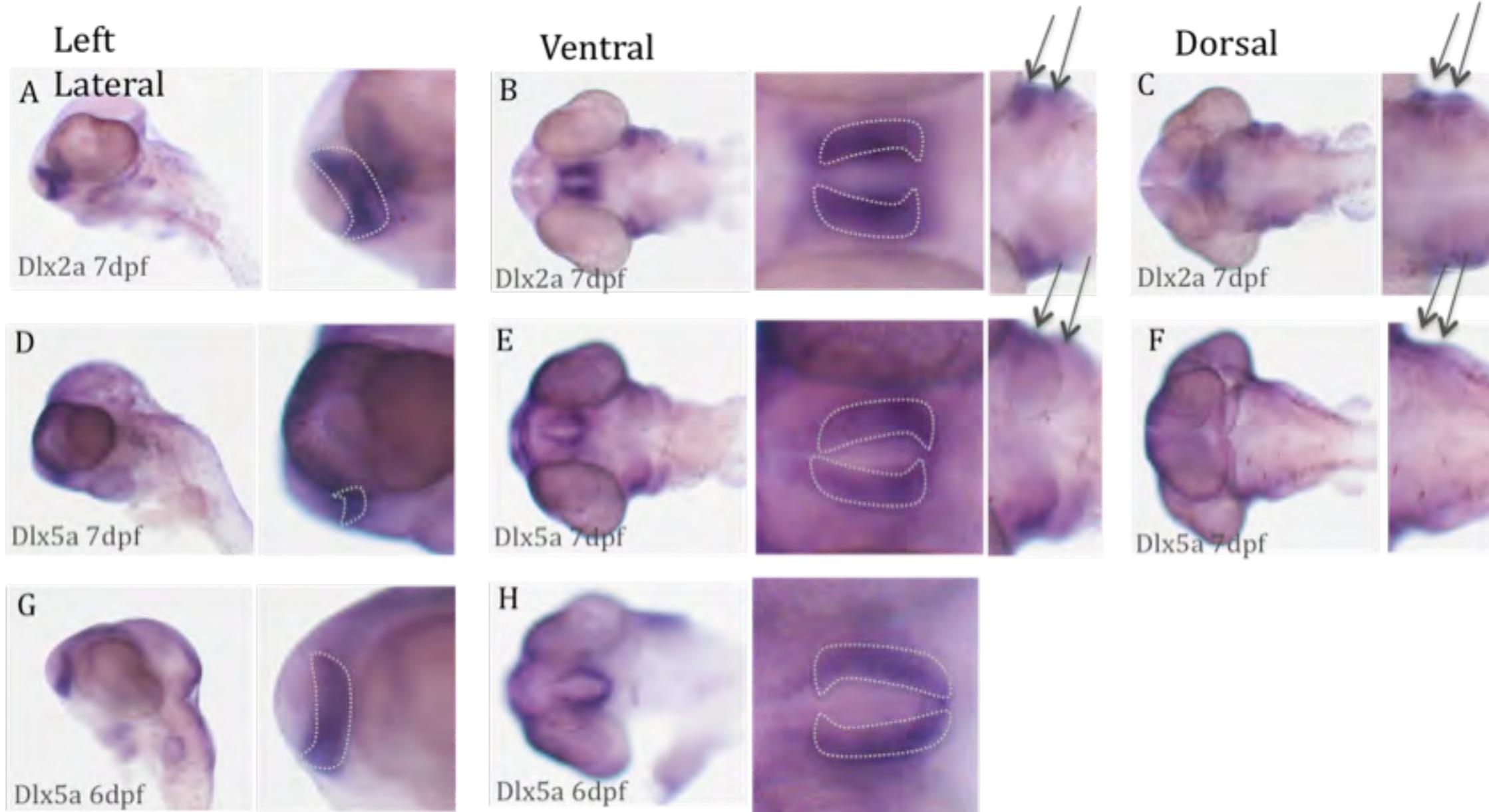


Transcriptome



We could use these genes right away

Dlx2a and *Dlx5a* expression in pipefish



Pipefish Genetic Map



Genetic map workflow

Generated an F1 family of 103 individuals

RAD sequenced the parents and offspring

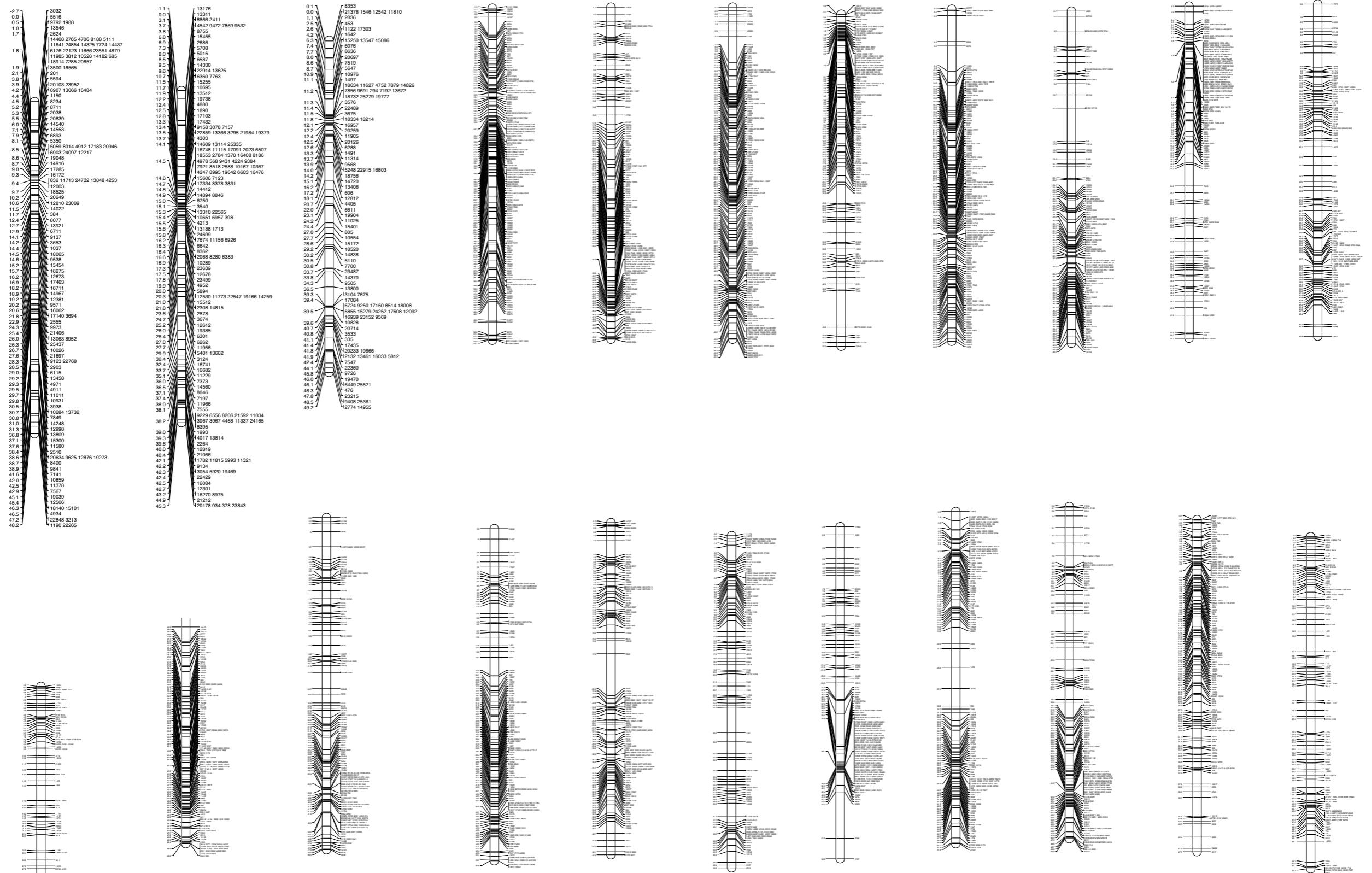
Analyzed the data using *Stacks*

Paired end local assemblies

Output to JoinMap format

Created Linkage map

The pipefish genetic map is closed; 22 LGs 6000 segregating SNPs; 30,000 RAD sites



Pipefish Genome Project



Genome workflow

Generated DNA from a single individual

Random Illumina shotgun sequencing

Removed highly repetitive kmers

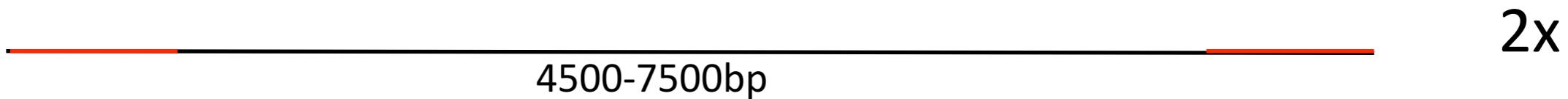
Produced several different genome assemblies

Illumina genomic libraries for pipefish genome

paired end 101bp



mate pair

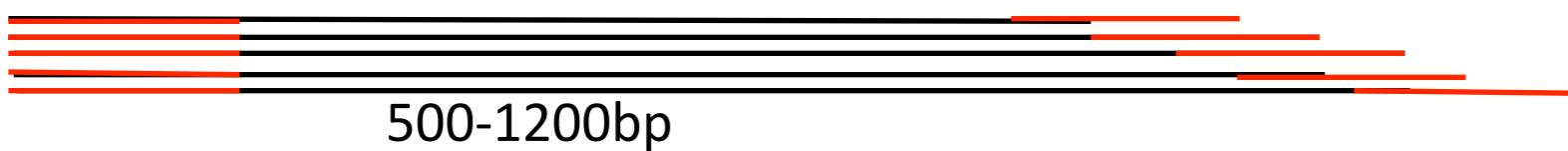


overlapping



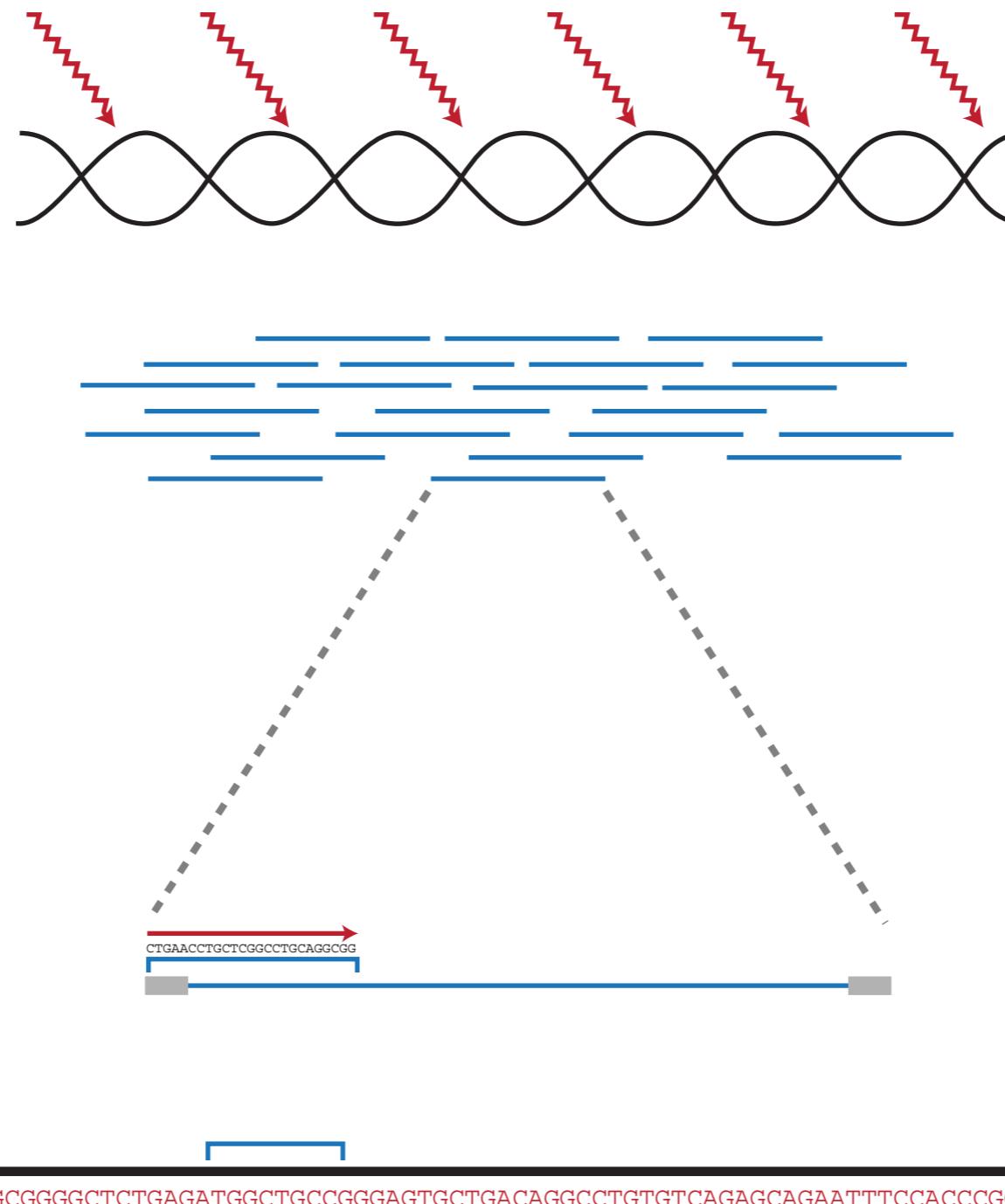
paired end RAD

ACTCTC



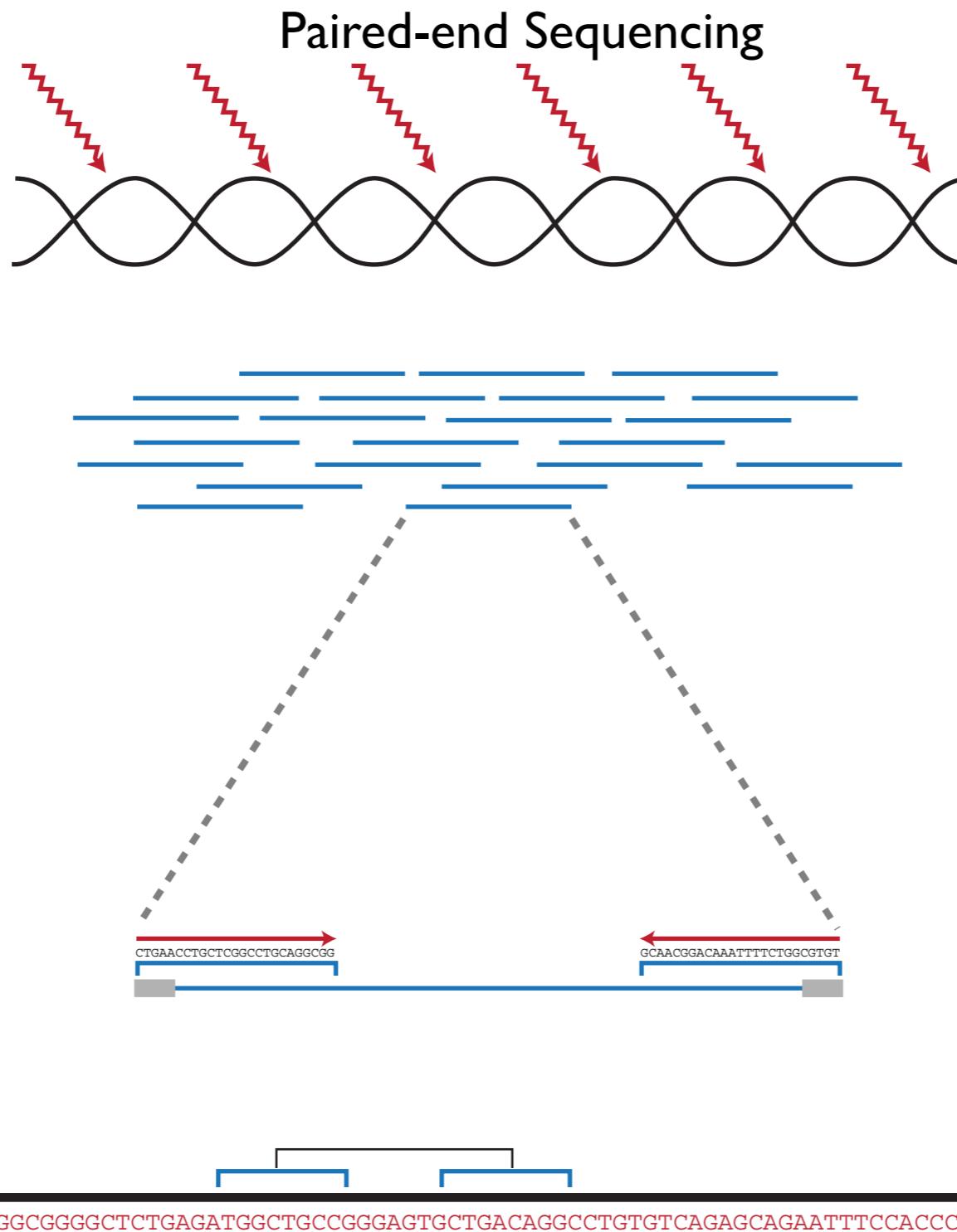
de novo Genome Sequencing

Single-end Sequencing

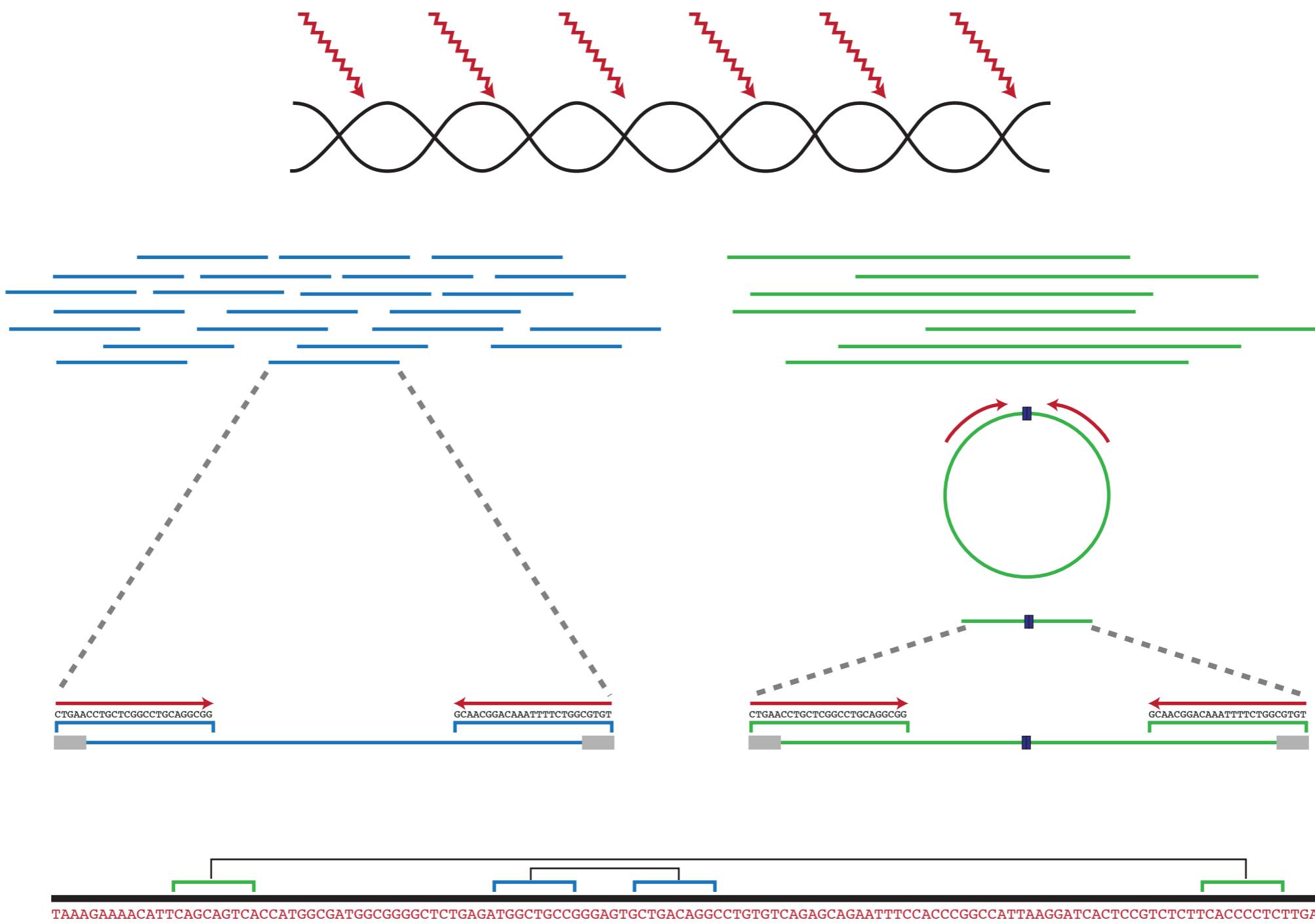


TAAAGAAAACATTCAAGCAGTCACCATGGCGATGGCGGGCTCTGAGATGGCTGCCGGAGTGCTGACAGGCCTGTGTAGAGCAGAATTCCACCCGGCCATTAAGGATCACTCCGTCTTCACCCCTTTGA

de novo Genome Sequencing



Mate-pair Sequencing

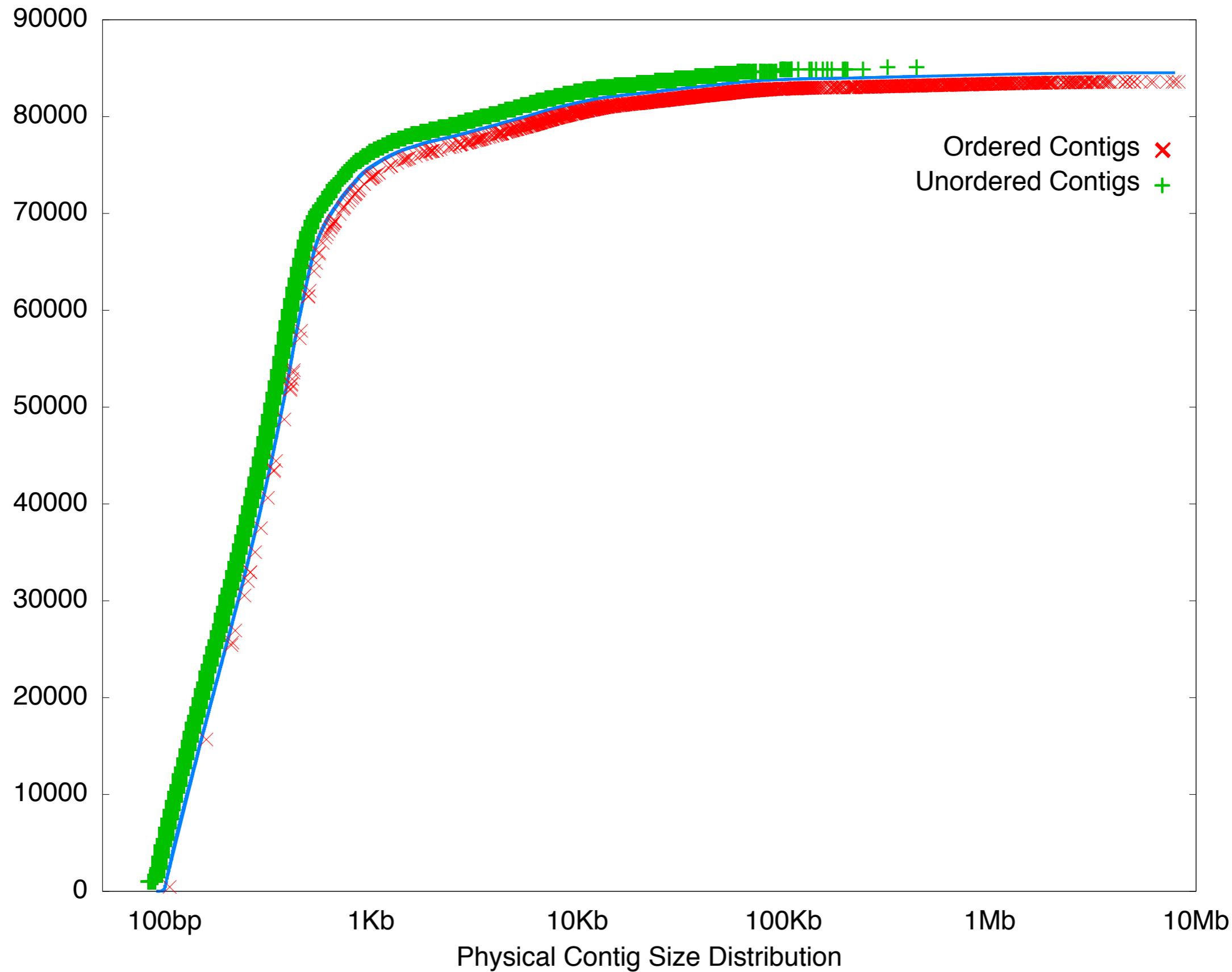


Pipefish genome assembly version 0.99

Nearly the whole genome is covered

Coverage	Scaffolds	Contigs	Scaffold N50	Contig N50
All (66.6x)	33,911	307,317	26,109	1,840

Max	Average Length	Total Length	Gap Length	%
198,155	9,916.35	336,273,415	38,303,839	(11.39%)



Overall Conclusions

Genomics can be a tool for enabling new ecology and evolution research

- documenting patterns of genetic variation
- identifying the molecular genetic basis of important phenotypic variation
- assessing how ecological processes structure this genetic variation in genomes
- RAD-seq is a powerful tool for SNP identification and genotyping
- analytical and computational approaches are challenging but manageable

Not your father's genome assembly

- a mixture of data types can be efficiently combined
- a genetic map is extremely useful for pulling it all together
- having a tiled genome is good enough - it doesn't have to be completely closed

Open Source Genomics provides a suite of breakthrough technologies

- the molecular approaches are not as daunting as they first appear
 - analytical and computational approaches are challenging
- New software tools can help, but knowledge of Unix and Scripting is essential**

Acknowledgments



- *Past and present lab members* **Paul Hohenlohe, Thom Nelson, Joe Dunham, Nicole Nishimura & Mark Currey**
- *Collaborators* **Eric Johnson, Patrick Phillips, Chuck Kimmel, John Postlethwait**
- *Funding from NSF & NIH, as well as Keck & Murdock Foundations*



TUTORIAL - USING STACKS

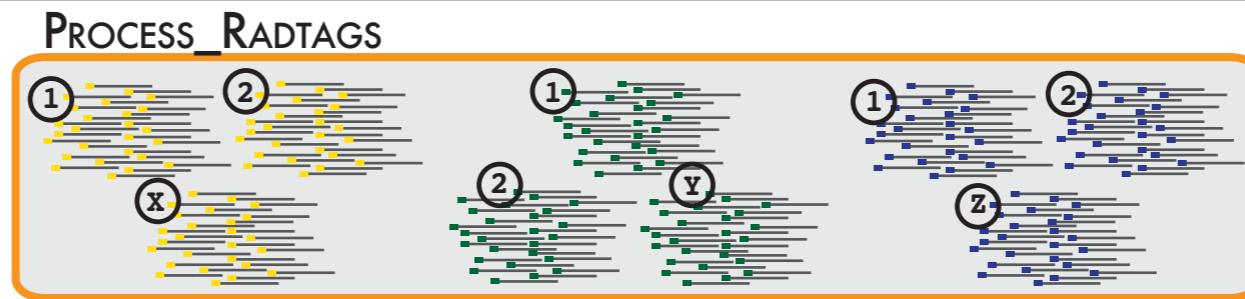


G3: Genes, Genomes, Genetics

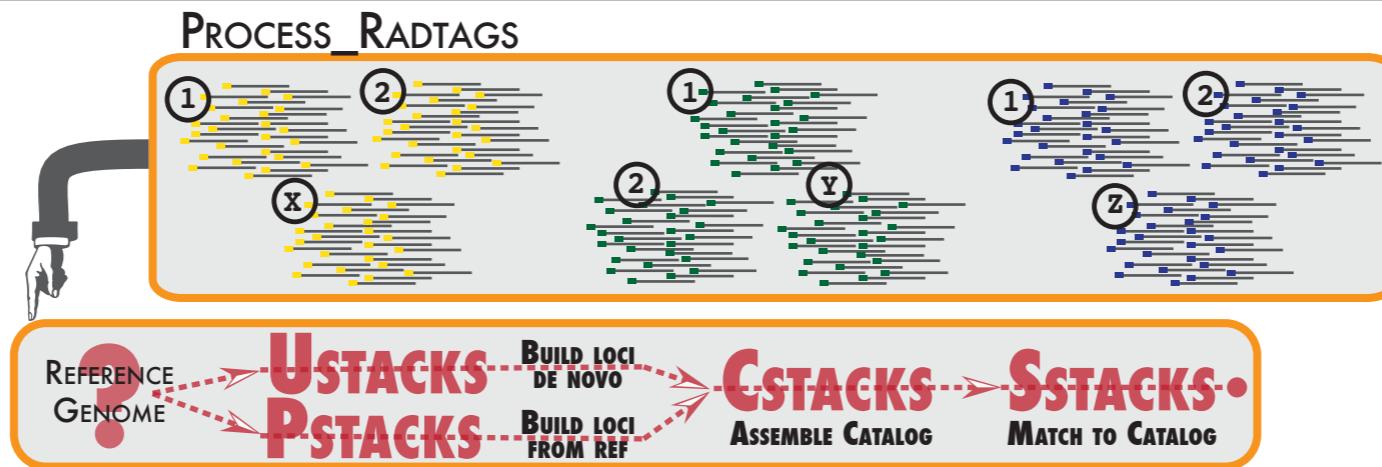
Stacks: Building and Genotyping Loci De Novo From Short-Read Sequences

Julian M. Catchen,* Angel Amores,[†] Paul Hohenlohe,^{*} William Cresko,^{*} and John H. Postlethwait^{†,1}
^{*}Center for Ecology and Evolutionary Biology and [†]Institute of Neuroscience, University of Oregon, Eugene, Oregon 97403

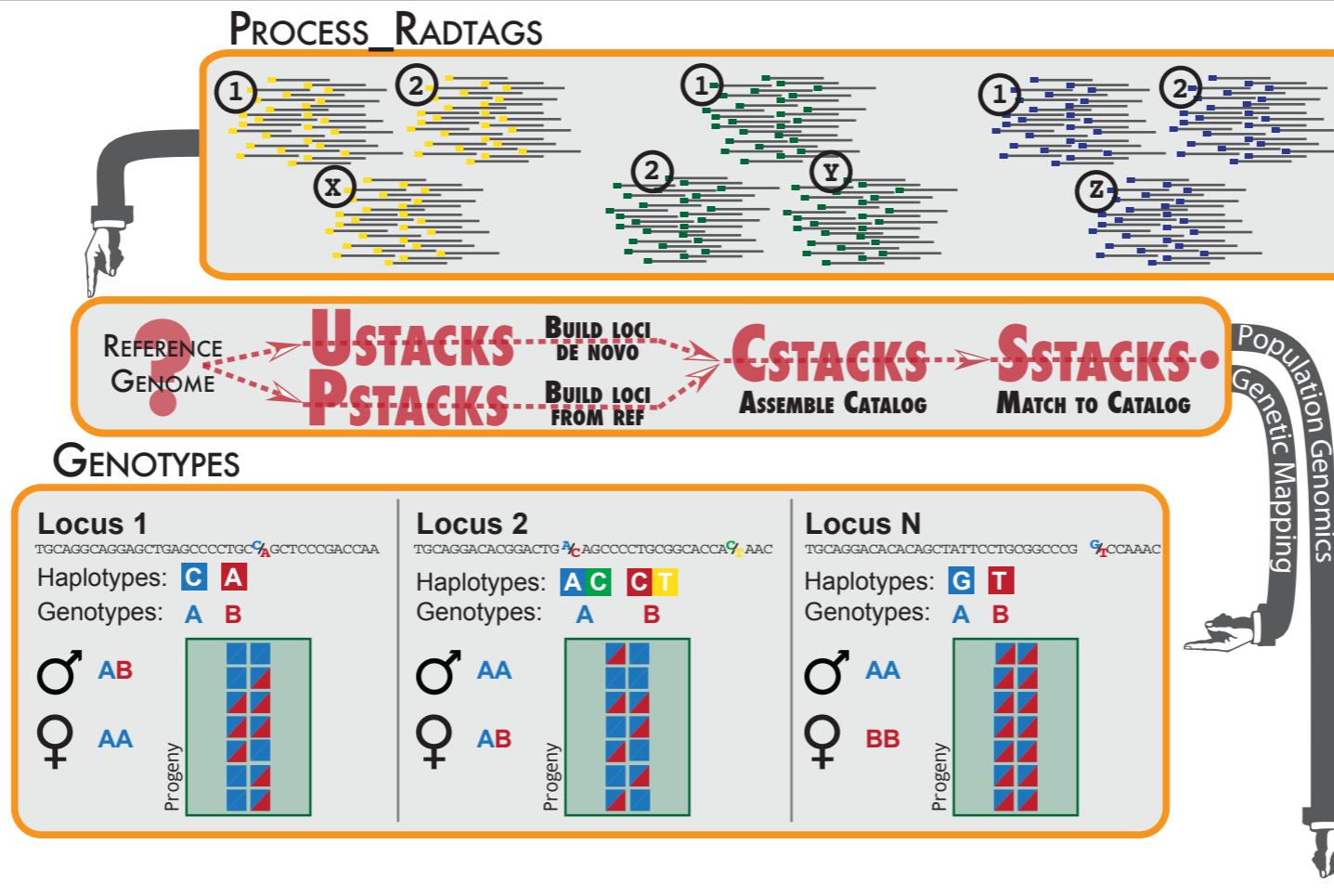
Stacks workflow



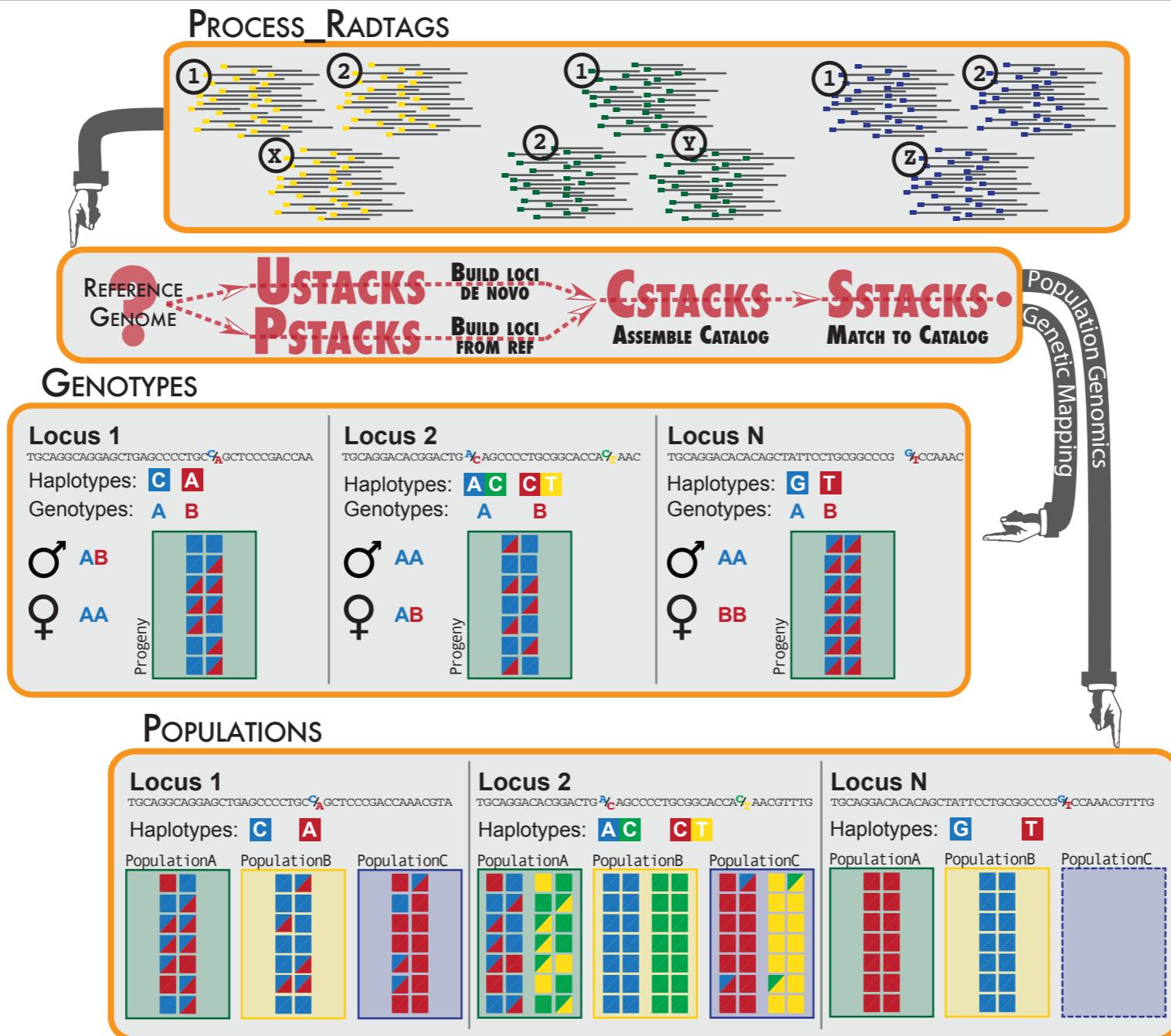
Stacks workflow



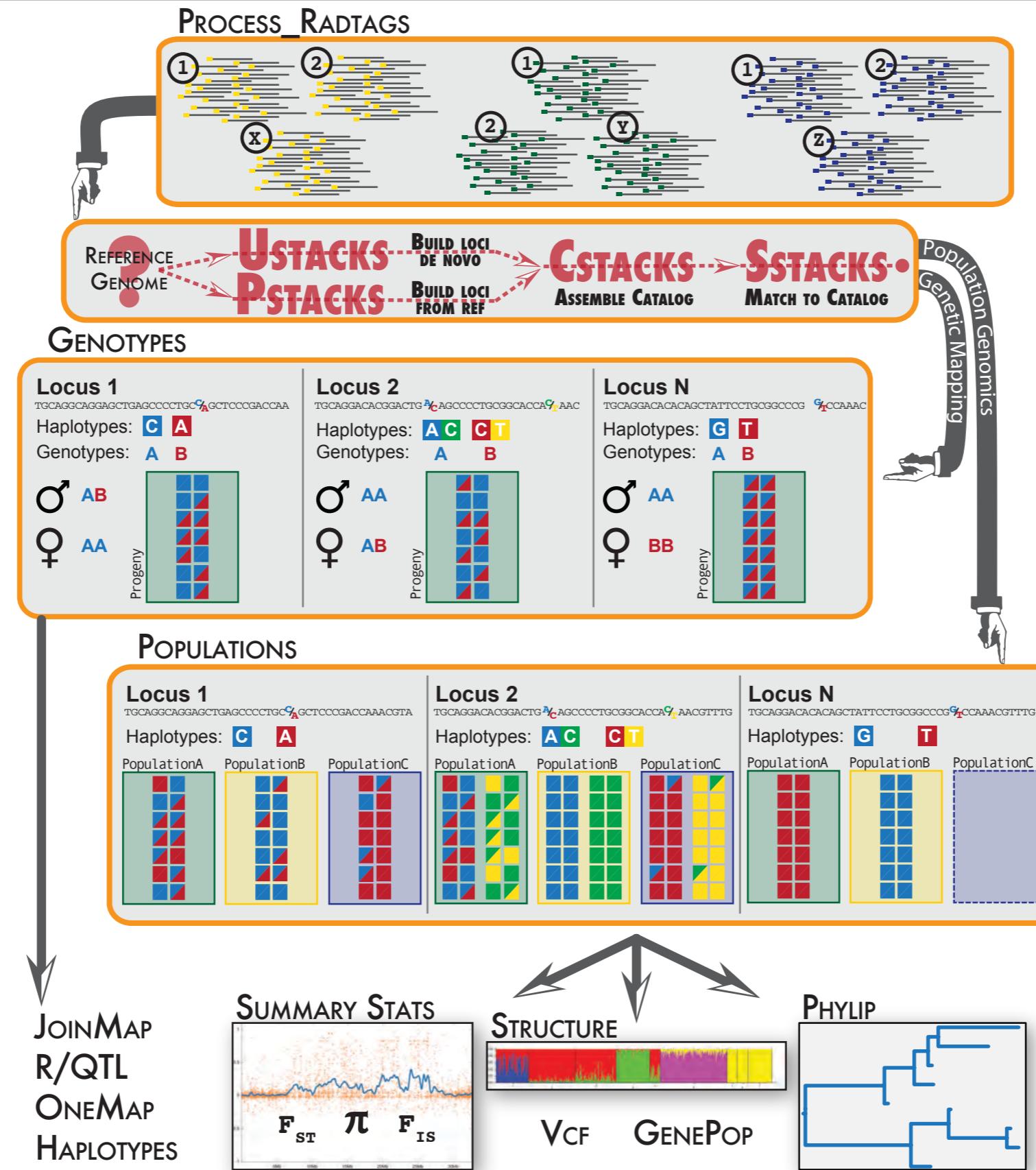
Stacks workflow



Stacks workflow



Stacks workflow



Stacks Analysis Pipeline: RAD-Tag Catalog Viewer

[http://genome.uoregon.edu/stacks/catalog.php?id=1&db=gartut_radtags&p=1&pp=10&filter_type\[\]=%cata&filter_cata=103&filter_alle_l=1&l](http://genome.uoregon.edu/stacks/catalog.php?id=1&db=gartut_radtags&p=1&pp=10&filter_type[]=%cata&filter_cata=103&filter_alle_l=1&l)

Q+ Google

1 (1 tags)

tags per page 10

Id	SNP	Consensus	Matching Parents	Progeny	Marker	Ratio	Genotypes
~103 annotate	Yes [2nuc]	TGCAGGAGCCCTCCCACTCGCTGATGCCACTCCATTCACTGACCCAGAGC G CAAAGCAACACTTCACAT T CCC	2	<u>92 / 91</u>	ab/ac	aa: 25 (27.5%) ab: 24 (26.4%) ac: 18 (19.8%) bc: 24 (26.4%)	91

SNPs**Alleles**Column: 52; G/A
Column: 70; T/Ga : GT
b : GG
c : AG**Matching Samples**View: Haplotypes Allele Depths Genotypes

Male	Female	Progeny 1	Progeny 2	Progeny 3	Progeny 4	Progeny 5	Progeny 6	Progeny 7	Progeny 8
GT / GG	AG / GT	GT	AG / GG	GG / AG	GG / GT	GG / AG	AG	GT / GG	AG / GT
GT	GT	GG / GT	GT / AG	GG / AG	GT / AG	GT / GG	GG / GT	GG / AG	GT
GT / AG	AG / GG	GT / AG	AG / GT	GG / AG	GG / AG	GT	GG / GT	GG / AG	GT
GT / GG	GT	GT	GT	GT	GT / GG	GT	GT / AG	GT	AG / GT
GT	GT	GT	GT	GT	GT	GT	GG / GT	GG / AG	GT / GG
GT	GT	GT / AG	GG / GT	GT / GG	GG / GT	GT	GG / AG	GT	GT / GG
GT	GT	GT / AG	GG / GT	GT / GG	GG / GT	GT	AG / GT	GT / AG	GG / GT
GT / GG	GT / GG	GT / AG	GG / AG	GG / GT	GT	GT	GG / GT	GT	GG / AG
GG / AG	AG / GG	GT	GG / AG	GT / GG	GT	GT	GG / AG	GT / GG	GT
GT / AG	GT / AG	GG / AG	GT	GT / GG	GT / GG	GT	GG / AG	GT	GG / AG
AG / GG	GT / AG	AG / GG	GG / AG	GG / AG	GT	GT	GT	GT	GT

1 (1 tags)

tags per page 10

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GT / GG 34 / 13	AG / GT 12 / 14	GT 7	AG / GG 8 / 16	GG / AG 26 / 14	GG / GT 15 / 11	GG / AG 14 / 8	AG 29	GT / GG 22 / 11	AG / GT 12 / 5
Progeny 9 GT 25	Progeny 10 GT 23	Progeny 11 GG / GT 32 / 14	Progeny 12 GT / AG 22 / 7	Progeny 13 GG / AG 7 / 8	Progeny 14 GT / AG 7 / 8	Progeny 15 GT / GG 2 / 3	Progeny 16 GG / GT 19 / 14	Progeny 17 GG / AG 9 / 4	Progeny 18 GT 15
Progeny 19 GT / AG 6 / 3	Progeny 20 AG / GG 6 / 9	Progeny 21 GT / AG 18 / 9	Progeny 22 AG / GT 4 / 5	Progeny 23 GG / AG 7 / 6	Progeny 24 GG / AG 8 / 10	Progeny 25 GT 7	Progeny 26 GG / GT 10 / 16	Progeny 27 GG / AG 3 / 3	Progeny 28 GG / GT 4 / 5
Progeny 29 GT / GG 8 / 5	Progeny 31 GT 11	Progeny 32 GT 10	Progeny 33 GT 17	Progeny 34 GT 20	Progeny 35 GT / GG 7 / 3	Progeny 36 GT 8	Progeny 37 GT / AG 12 / 4	Progeny 38 GT 9	Progeny 39 AG / GT 12 / 7
Progeny 40 GT 9	Progeny 41 GT 5	Progeny 42 GT 9	Progeny 43 GT / GG 9 / 12	Progeny 44 GG / GT 3 / 6	Progeny 45 GT 6	Progeny 46 GG / GT 4 / 11	Progeny 47 GG / AG 3 / 7	Progeny 48 GT 18	Progeny 49 GT / GG 5 / 6
Progeny 50 GT 18	Progeny 51 GT 9	Progeny 52 GT / AG 8 / 5	Progeny 53 GG / GT 10 / 8	Progeny 54 GT / GG 5 / 6	Progeny 55 AG / GG 8 / 10	Progeny 56 GT 22	Progeny 57 AG / GT 17 / 16	Progeny 58 GT / AG 23 / 24	Progeny 59 GG / GT 25 / 13
Progeny 60 GT / GG 12 / 18	Progeny 61 GT / GG 22 / 29	Progeny 62 GT / AG 7 / 23	Progeny 63 GG / AG 15 / 11	Progeny 64 GG / GT 13 / 20	Progeny 65 GT 44	Progeny 66 GT 27	Progeny 67 GG / GT 23 / 17	Progeny 68 GT 30	Progeny 69 GG / AG 14 / 13
Progeny 71 GG / AG 15 / 7	Progeny 72 AG / GG 9 / 6	Progeny 73 GT 42	Progeny 74 GG / AG 31 / 29	Progeny 75 GT / GG 15 / 22	Progeny 76 GT 41	Progeny 77 GG / AG 14 / 17	Progeny 78 GG / AG 25 / 17	Progeny 79 GT / GG 29 / 14	Progeny 80 GT 34
Progeny 81 GT / AG 17 / 29	Progeny 82 GT / AG 29 / 24	Progeny 83 GG / AG 16 / 25	Progeny 84 GT 41	Progeny 85 GT / GG 14 / 24	Progeny 86 GT / GG 5 / 4	Progeny 87 GT 15	Progeny 88 GG / AG 5 / 11	Progeny 89 GT 18	Progeny 90 GG / AG 5 / 17
Progeny 91 AG / GG 14 / 13	Progeny 92 GT / AG 12 / 6	Progeny 93 AG / GG 7 / 7	Progeny 94 GG / AG 3 / 2						

1 (1 tags)

tags per page 10

Stacks

Batch #1 [2011-08-10; 80bp *Lepisosteus oculatus* F1 Genetic Map RAD-Tag Samples]

RAD-Tag Sample #2 [female]

• Sequence #73

Catalog ID	Depth	SNPs		Alleles	Deleveraged?	Lumberjackstack?	Blacklisted?
#103	26x	Column: 52	G/A	AG 46.15%	False	False	False
		Column: 70	T/G	GT 53.85%			