



Genomic analysis in non-model organisms

2013 Workshop on Genomics
Český Krumlov

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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

1850

Evolution

1900

Conditions
of
Existence

Systematics
Ecology

Genetics

Unity
of
Type

1950



Population
Genetics
Modern
Synthesis

Experimental
Embryology

Molecular
Genetics

2000

Evolutionary
Genetics

Developmental
Genetics



functional evolutionary genomics

Model organism research has been very important

Vertebrate **zygotes** or embryos



28 day human



19h zebrafish



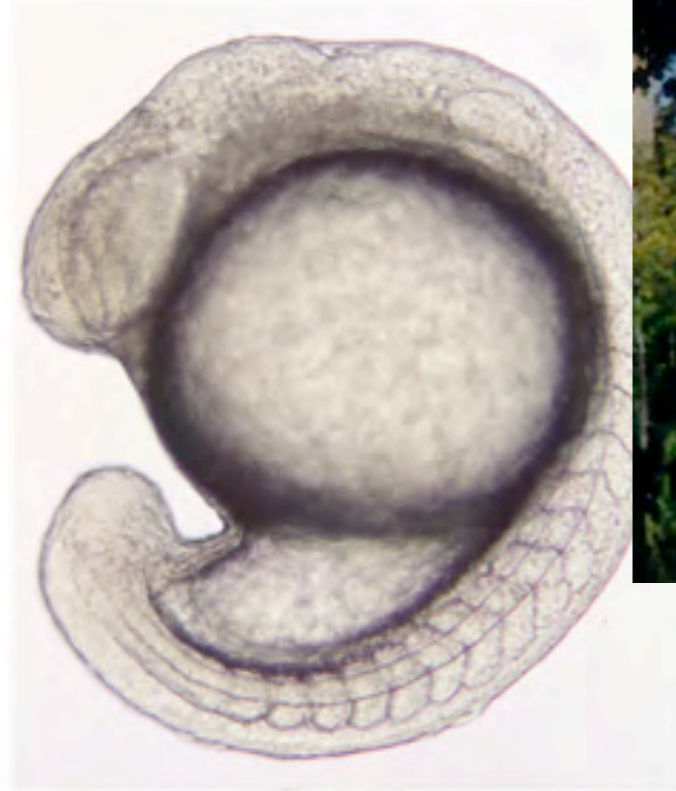
Video by Don Kane

Model organism research has been very important

Vertebrate **zygotes** or embryos



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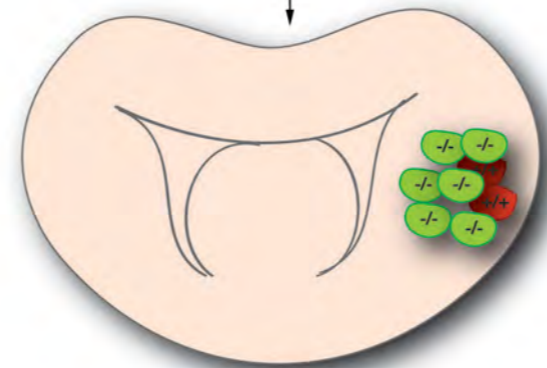
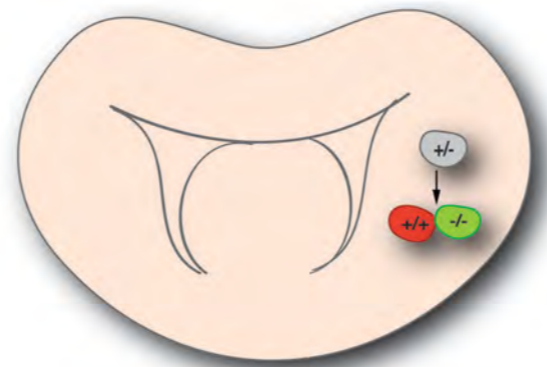


19h zebrafish

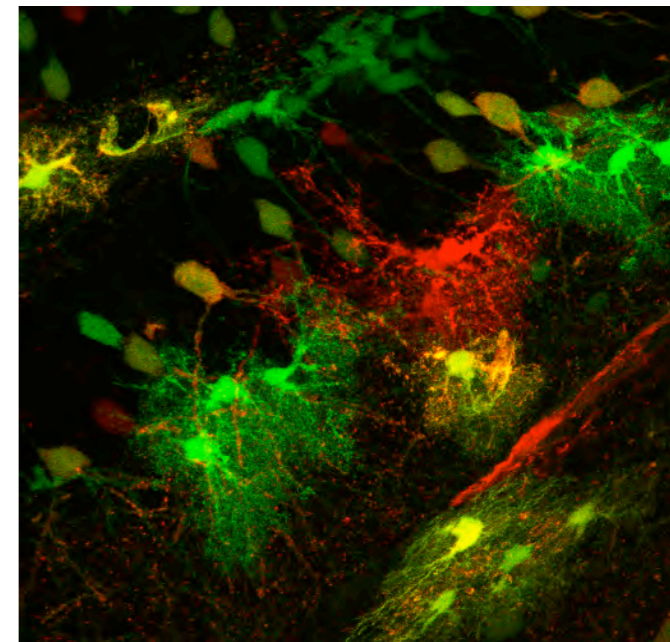
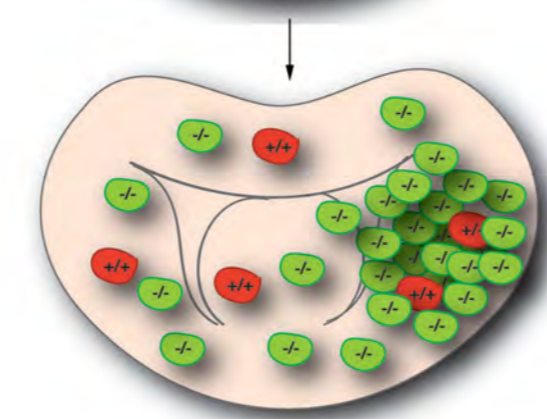


Studying brain cancer using somatic evolutionary genomics in a model organism

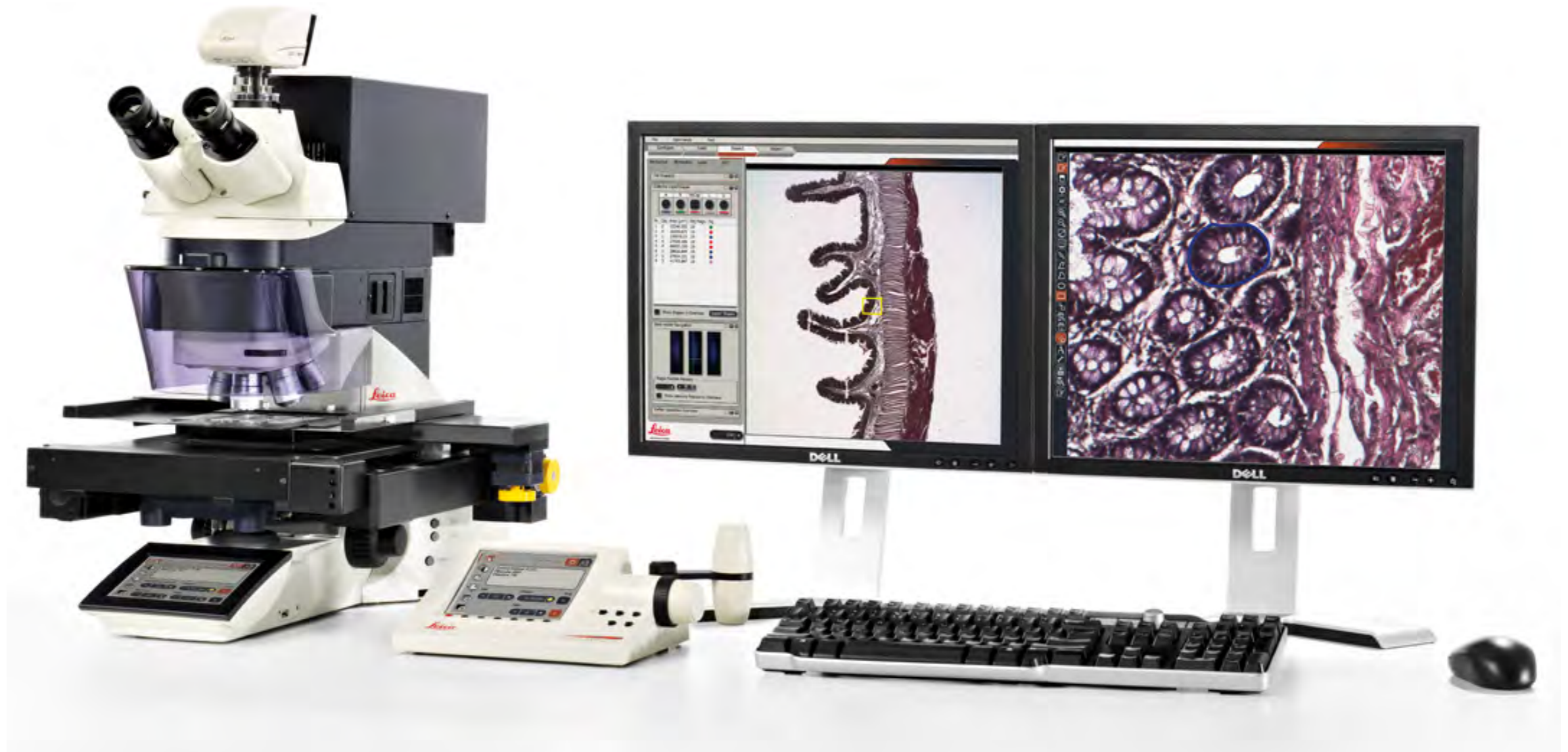
pre-cancerous



tumor



Laser Capture Microdissection of cells



Transcriptomic and genomic analysis of cells



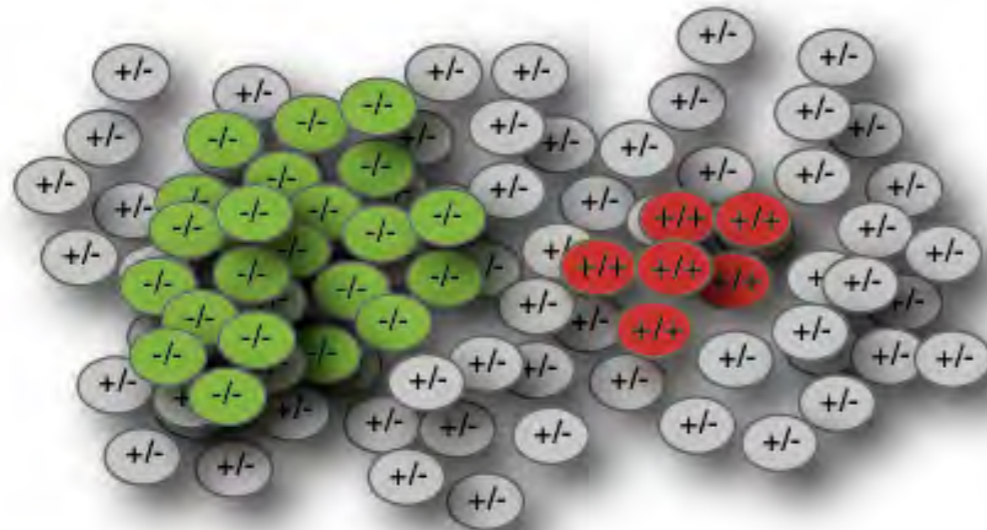
Sequence cells here...

... and here

Transcriptomic and genomic analysis of cells



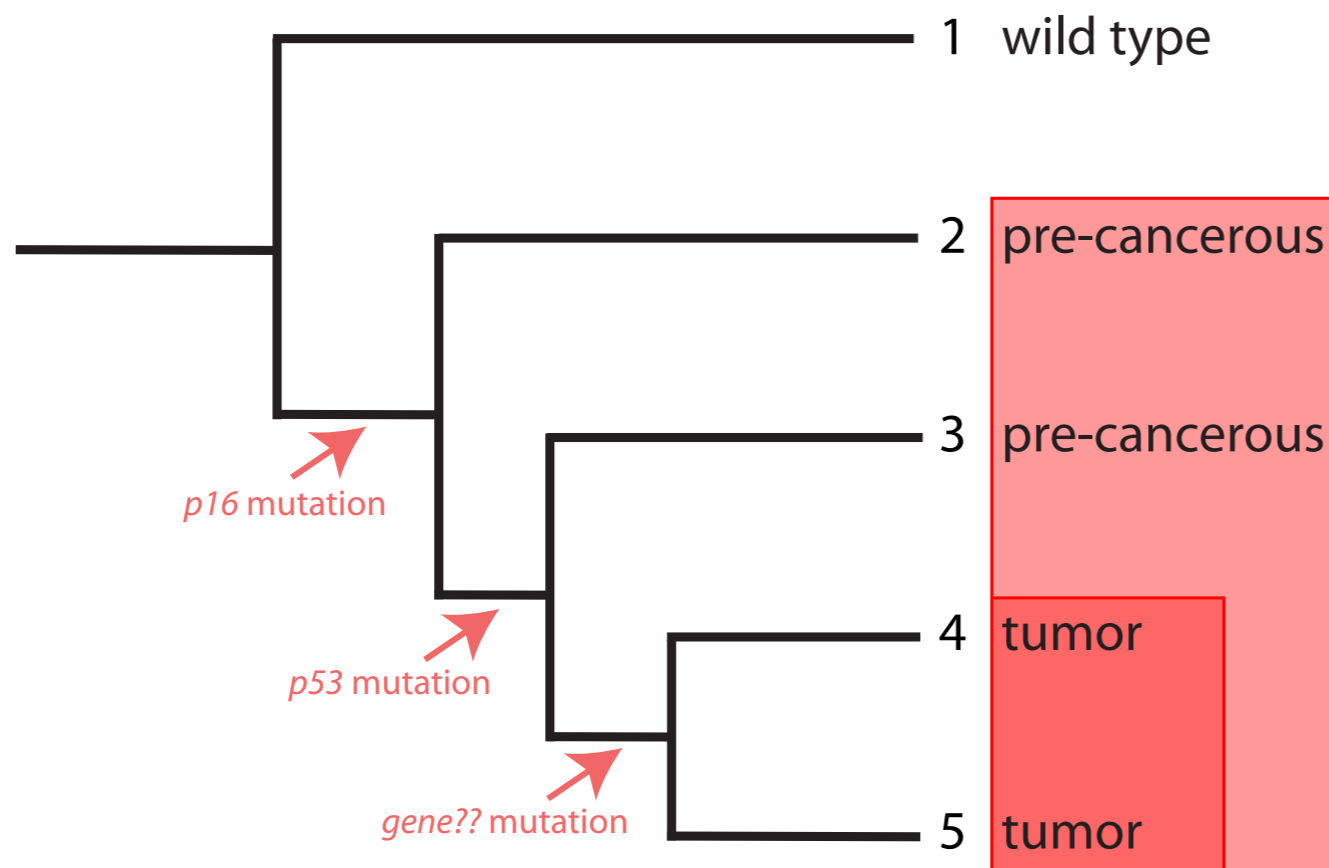
Sequence cells here...



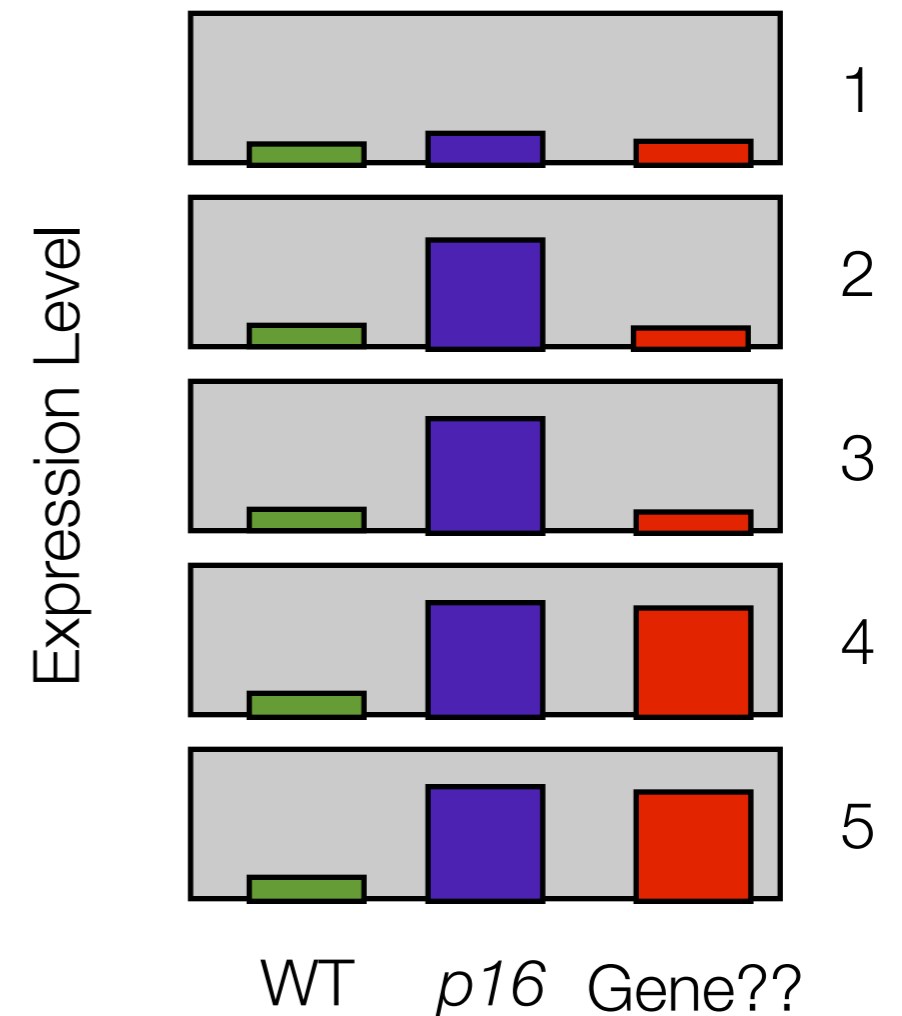
... and here

Multiple lines of genetic evidence for causative mutations

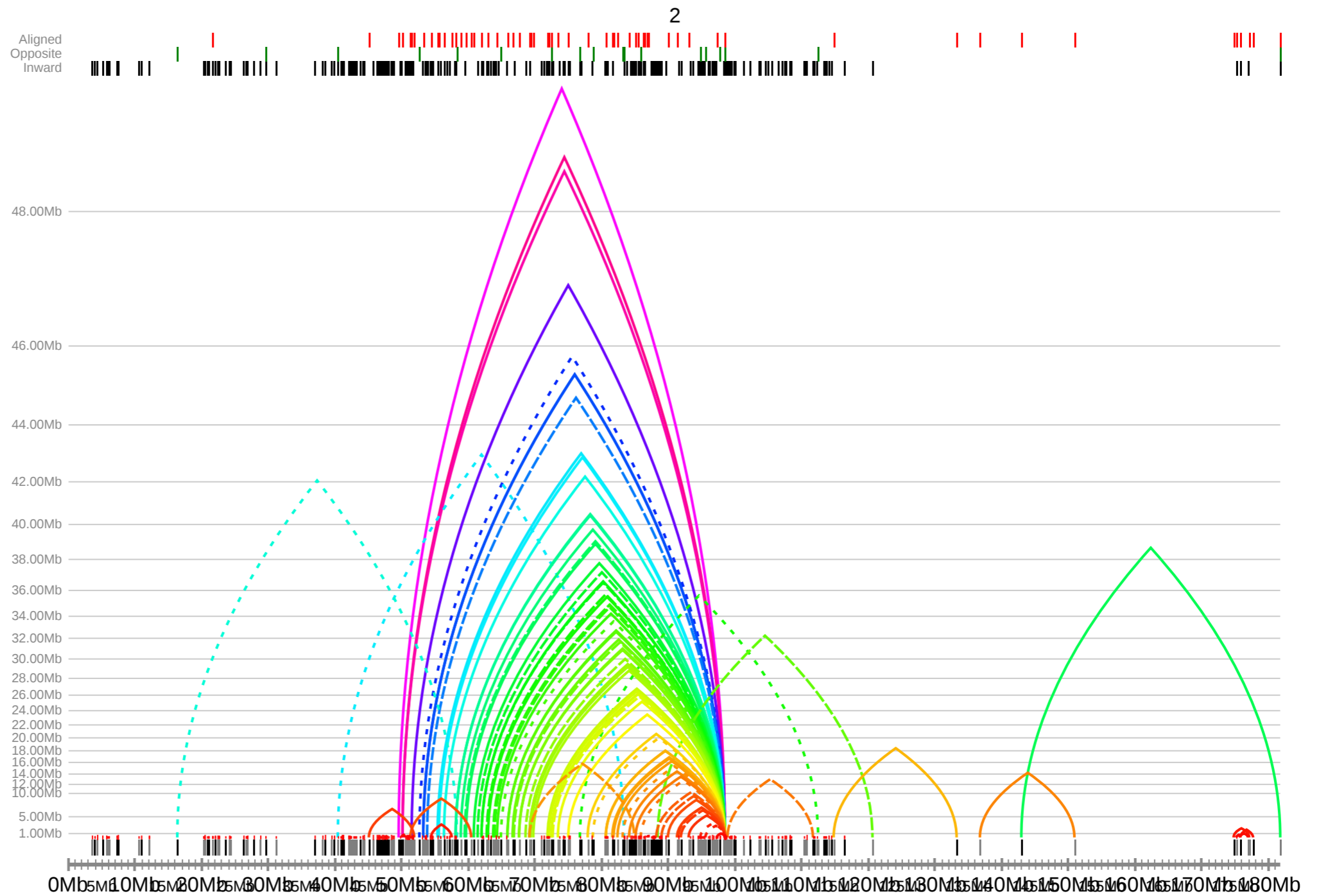
Gene Mutations



Gene Expression

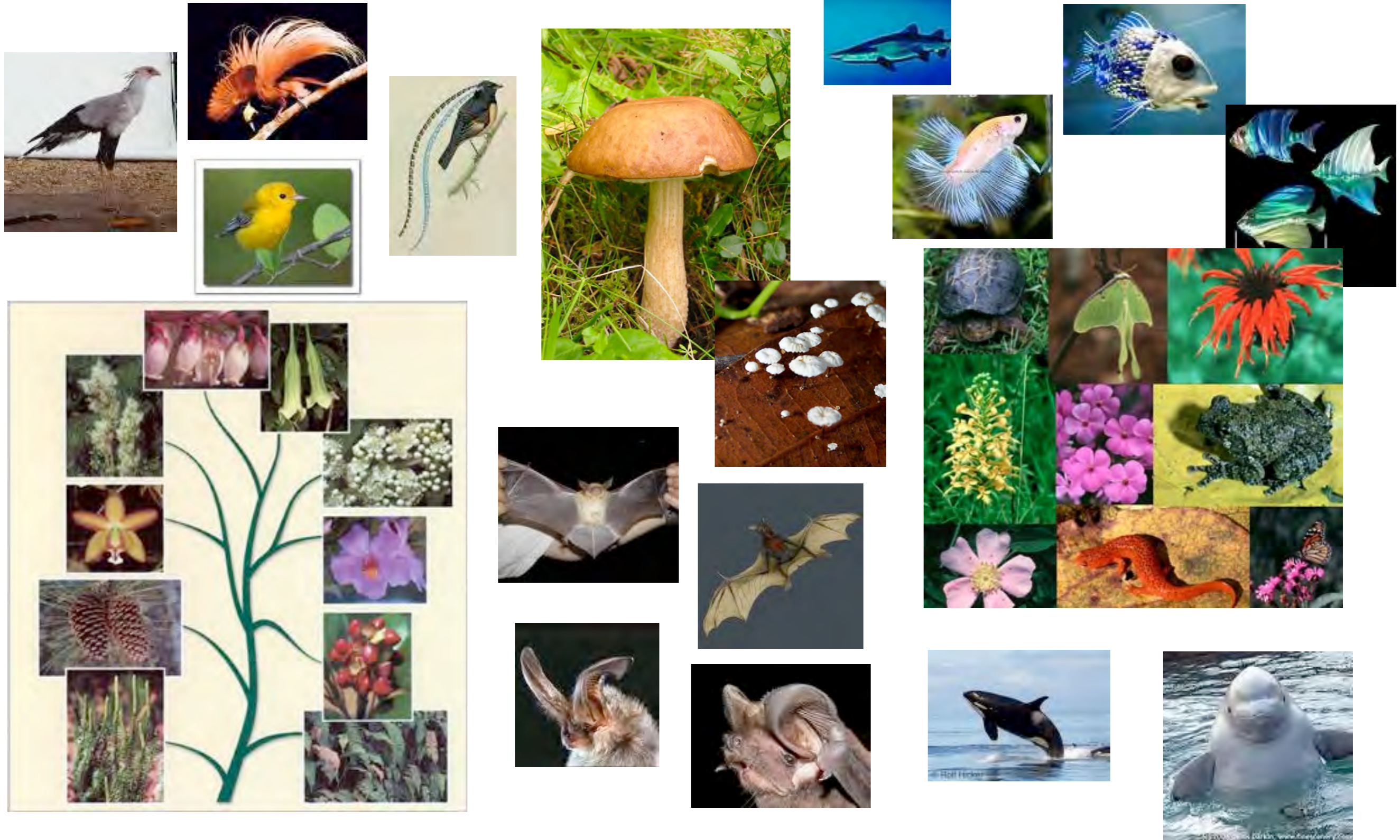


Genomic rearrangements in cancer cells

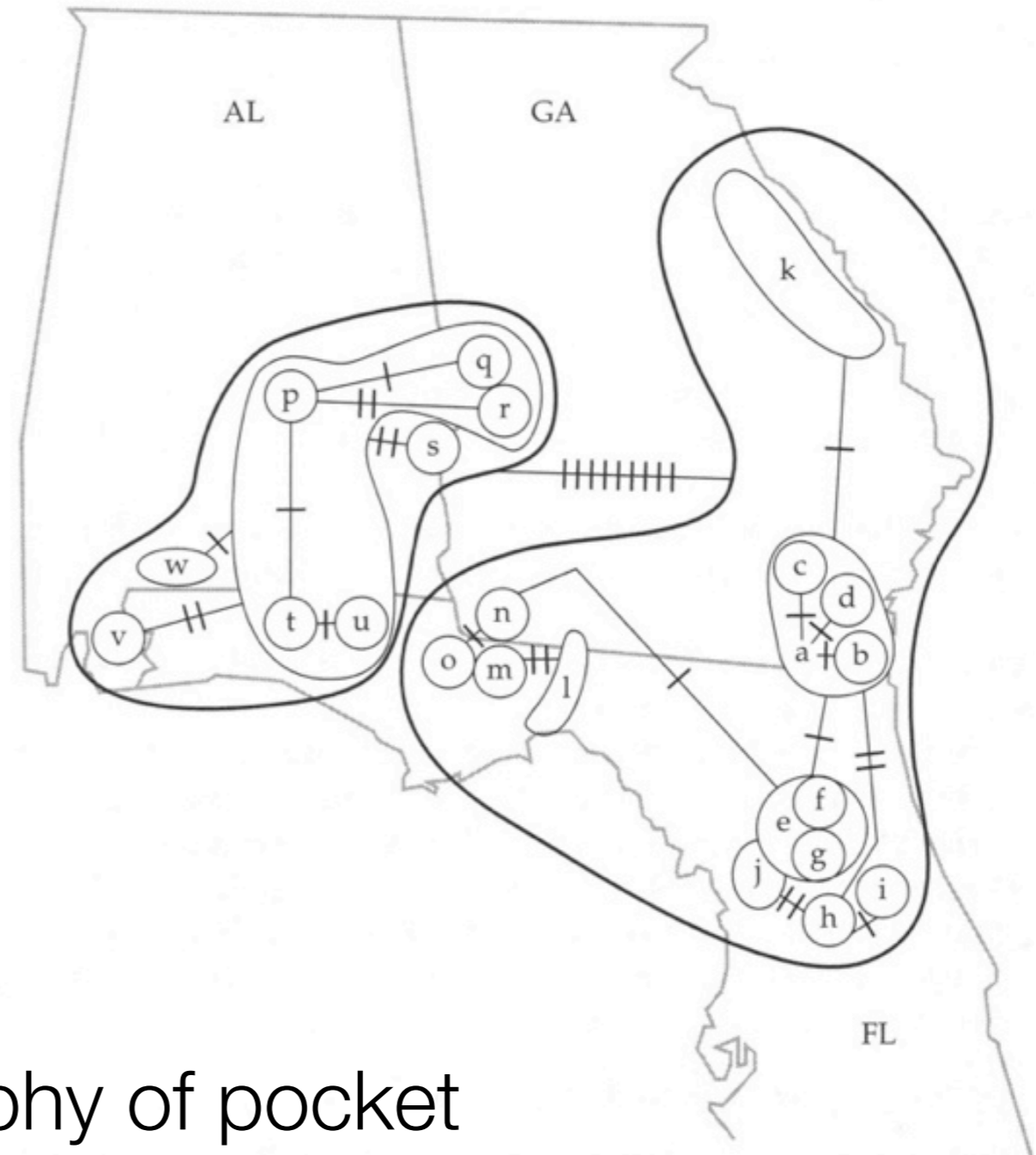


How can modern genomics improve studies
of non-model organisms??

How do the major differences among lineages evolve?



How are organisms related to one another?



Phylogeography of pocket gophers using mtDNA

Avise, 1979

How do organisms adapt to novel environments?



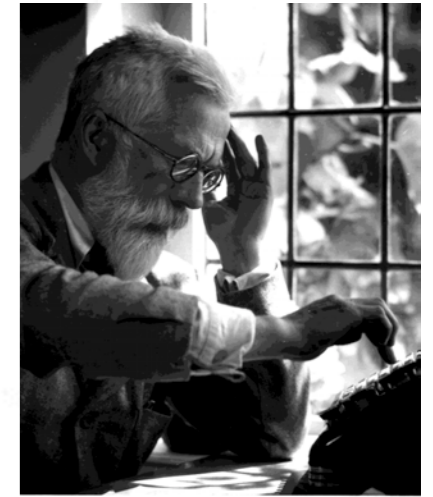
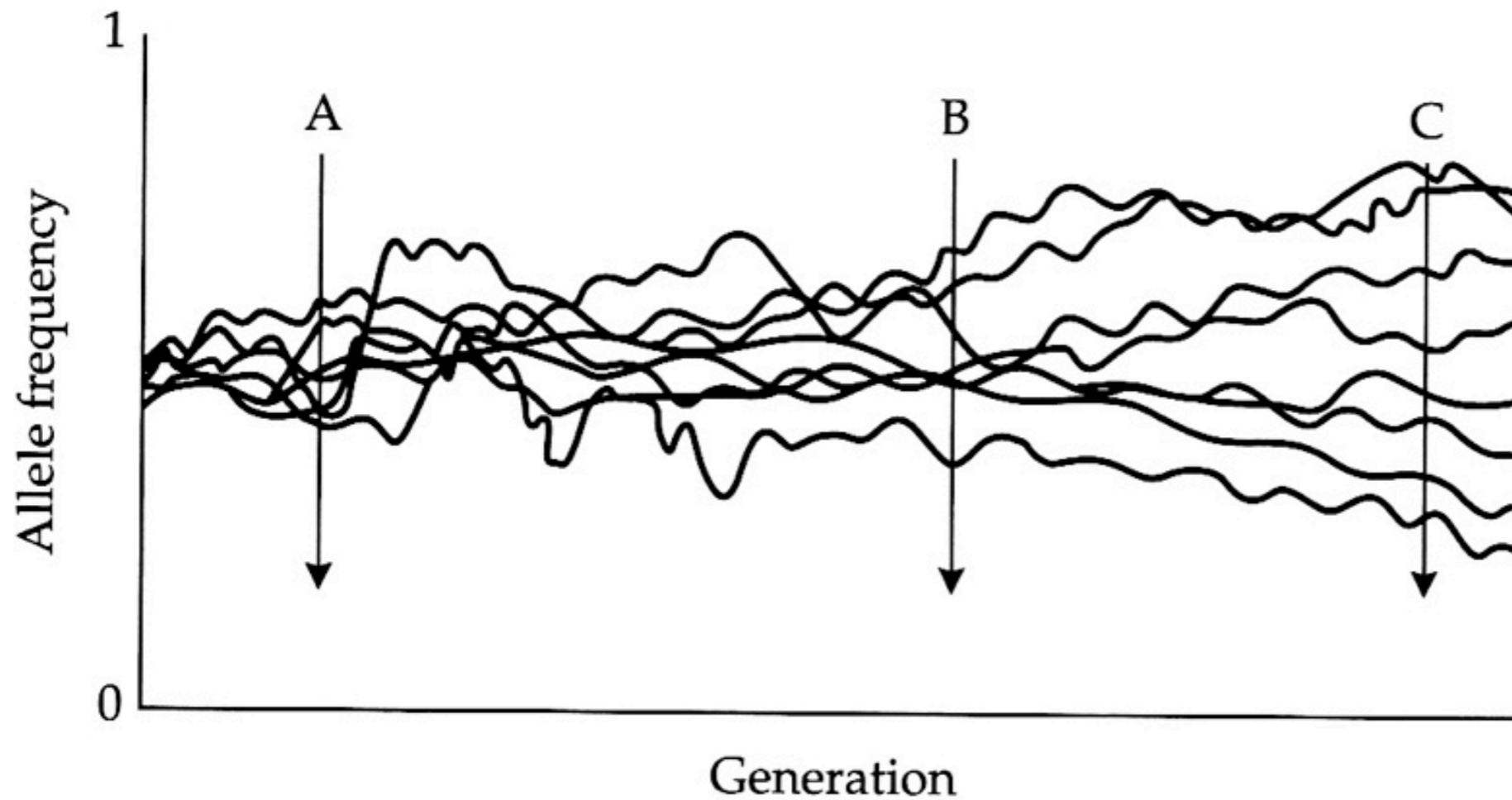
from Grant and Grant. 2007. How and why species multiply: The radiation of Darwin's finches. Princeton University Press

Four fundamental processes in evolution

Origin of genetic variation;
mutation
migration

Sorting of variation;
genetic drift
natural selection

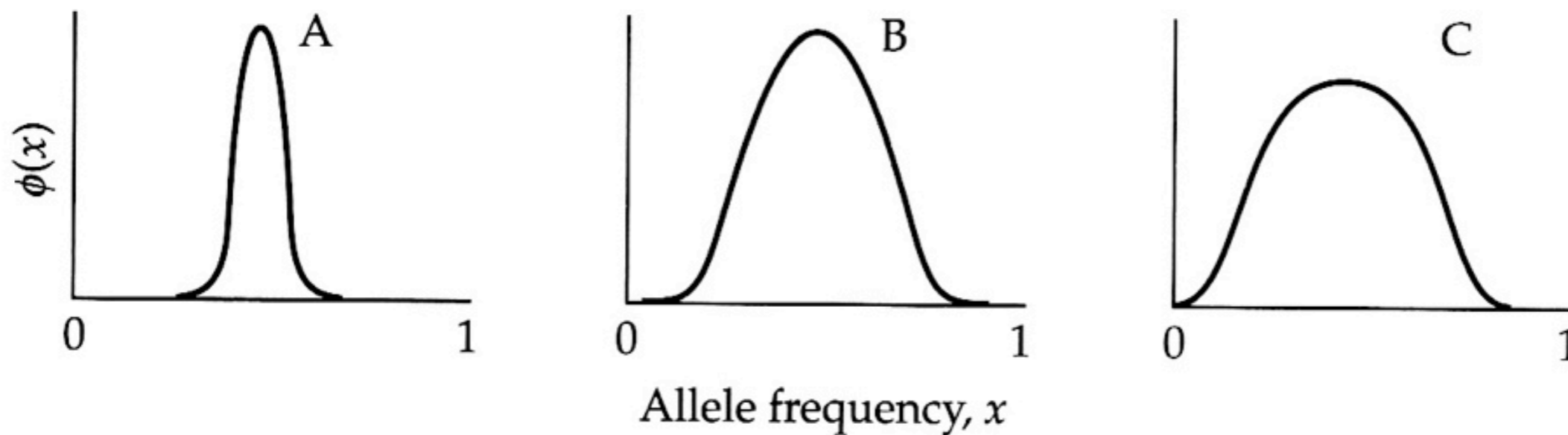
Genetic drift is a null model



R.A. Fisher



Sewall Wright



Population genomics

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



Genome-wide background provides more precise estimates:

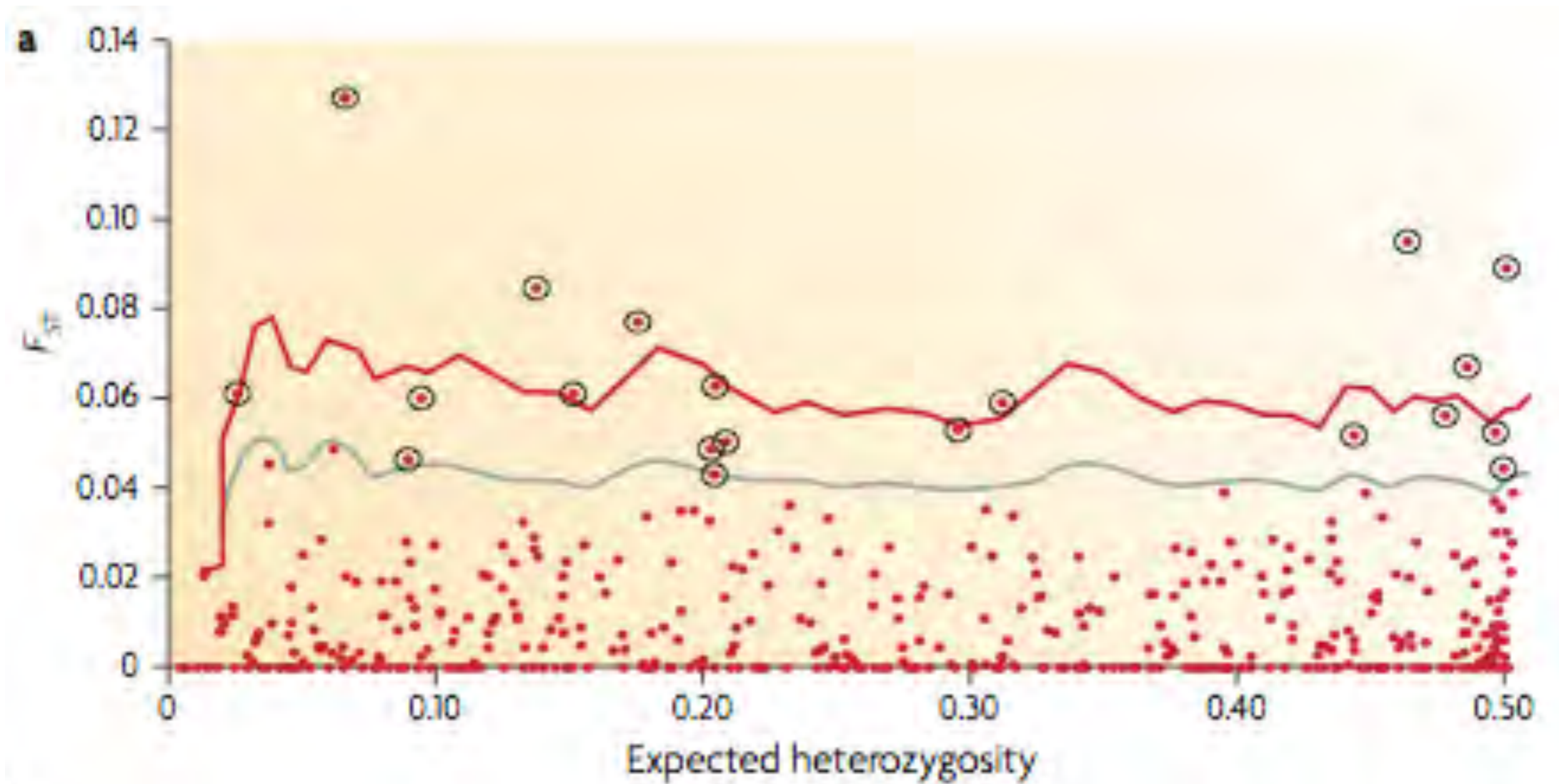
- Demographic processes (e.g. N_e)
- Phylogeography



Outliers from background indicate:

- Selective sweeps
- Local adaptation

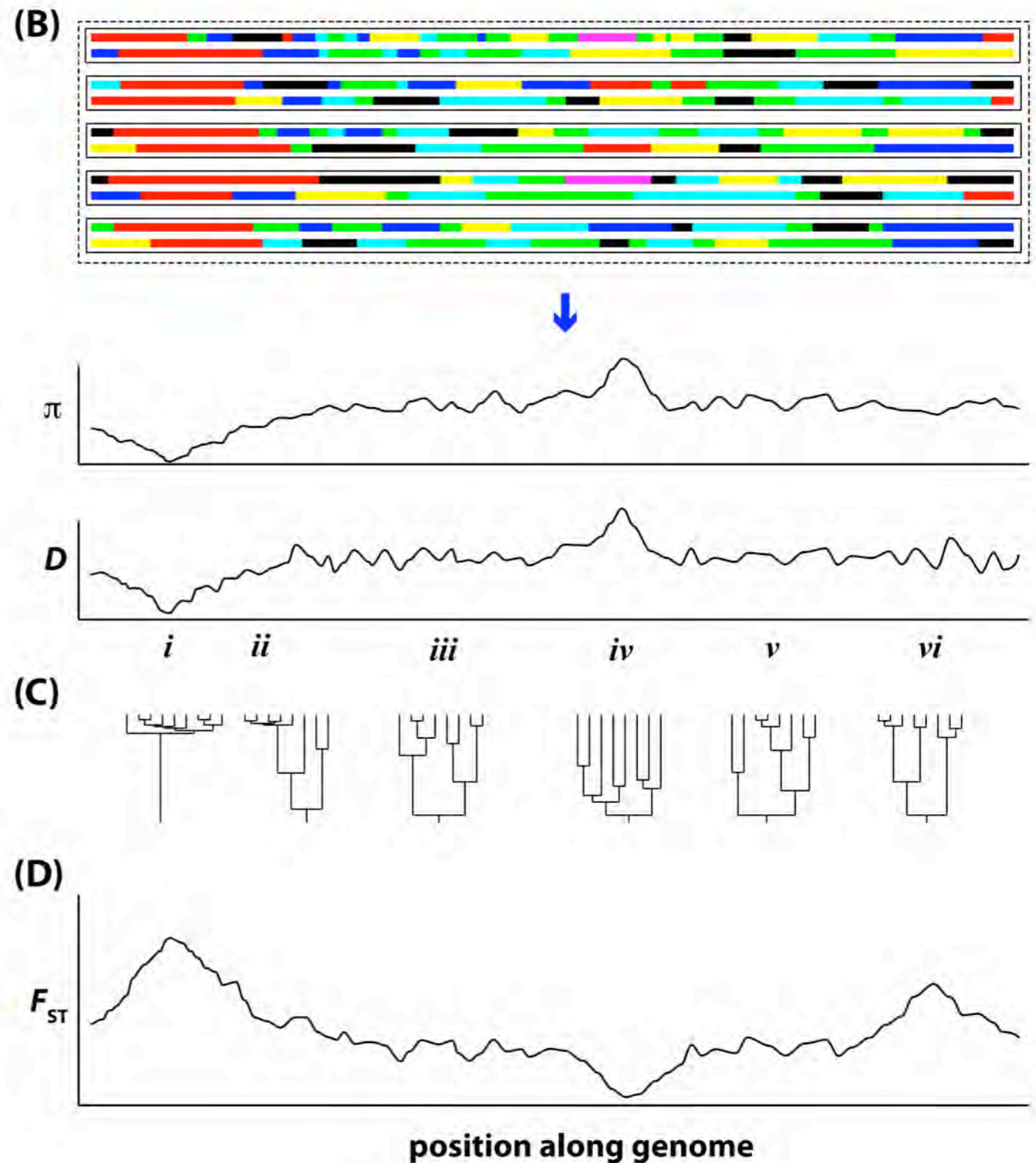
Population genomics of unordered markers



Population genomics of ordered markers

Genomic architecture:

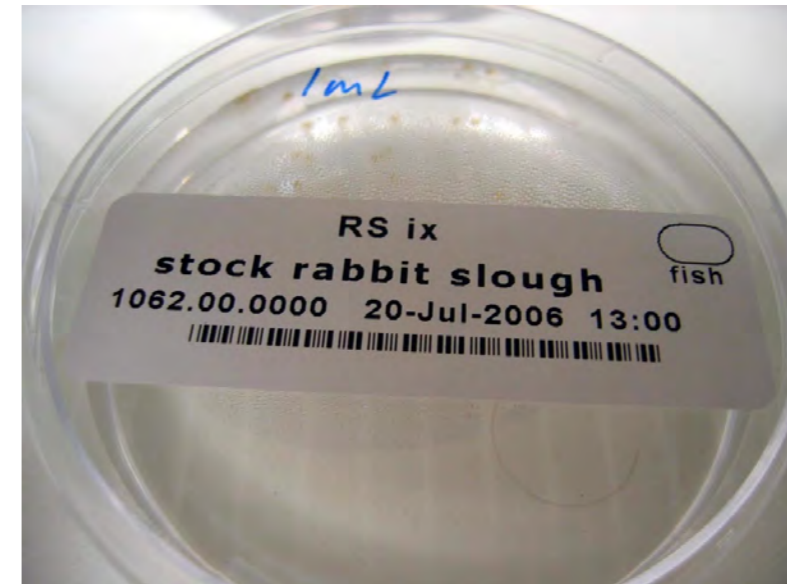
- Distribution adaptive variation across the genome
- Correlations among genomic regions (linkage disequilibrium)
- Interactions among genomic regions (e.g. epistasis)
- Recombination rates and chromosomal inversions



How do we 'genomically enable' research on non-model organisms?

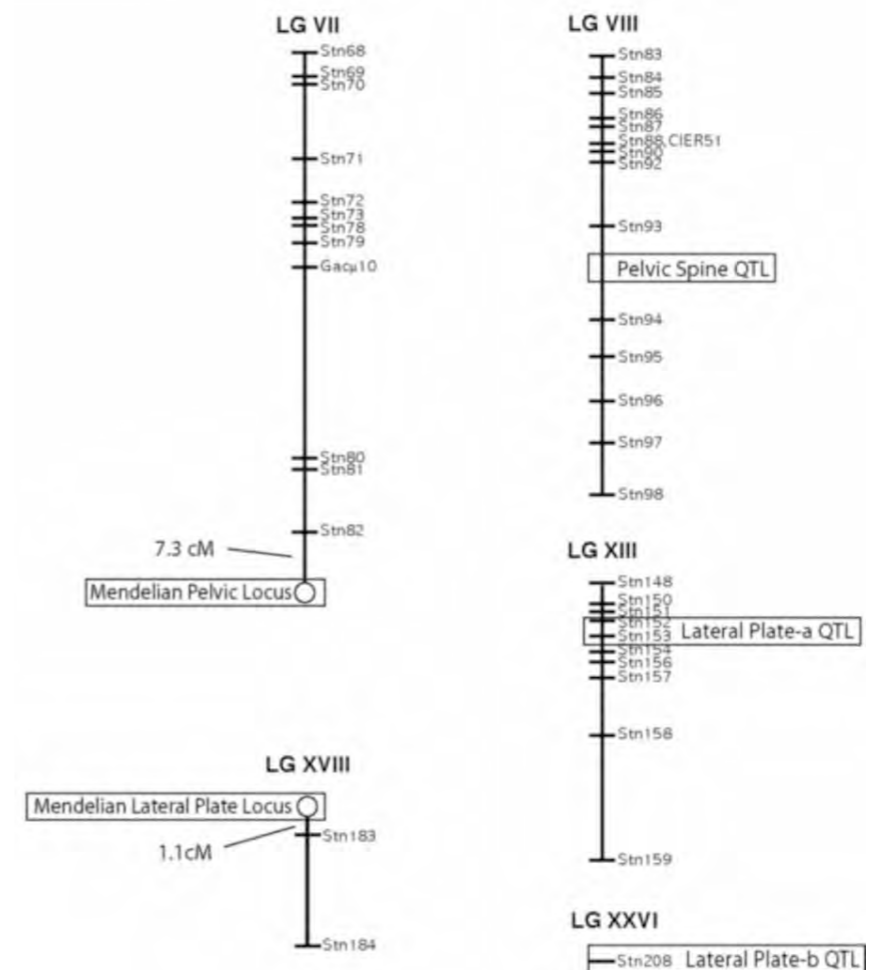
1. Genetic Markers & Maps
2. Physical Maps
3. Transcriptomes
4. Gene Expression Analyses

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



Alaska

British Columbia



The open source genomics breakthrough



Next generation sequencing, high performance computing and new analytical approaches have fundamentally changed the scope of studies of non-model organisms

Should we just sequence everything?

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
 - the genomes of many organisms are organized in linkage blocks
 - well spaced markers will provide the necessary coverage
 - the cost of genotyping will almost always be a fraction of full sequencing
- 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 the sequencing on a homologous set of tags spread throughout the genome
- Can lead to the simultaneous identification and typing of single nucleotide polymorphisms (SNPs)
- The cost will always be a fraction of the cost of resequencing the genome
 - i.e. 1% genome coverage will be less than 1% the cost
 - often the coverage is more even than whole genome sequencing
- 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
 - Inferring patterns of LD may be easiest with full sequences

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

- Common acronyms
 - **RRL** - **R**educed **R**epresentation **L**ibrary
 - **GBS** - **G**enotyping **B**y **S**equencing
 - **CRoPS** - **C**omplexity **R**eduction of **P**olymorphic **S**equences
 - **MSG** - **M**ultiplex **S**hotgun **G**enotyping
 - **RAD** - **R**estriction site **A**ssociated **D**N
- All rely on restriction enzyme digestion
- RRL, CRoPS, MSG and GBS use one or two restriction enzymes only
- RAD uses an extra shearing step to 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^{1,4}

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

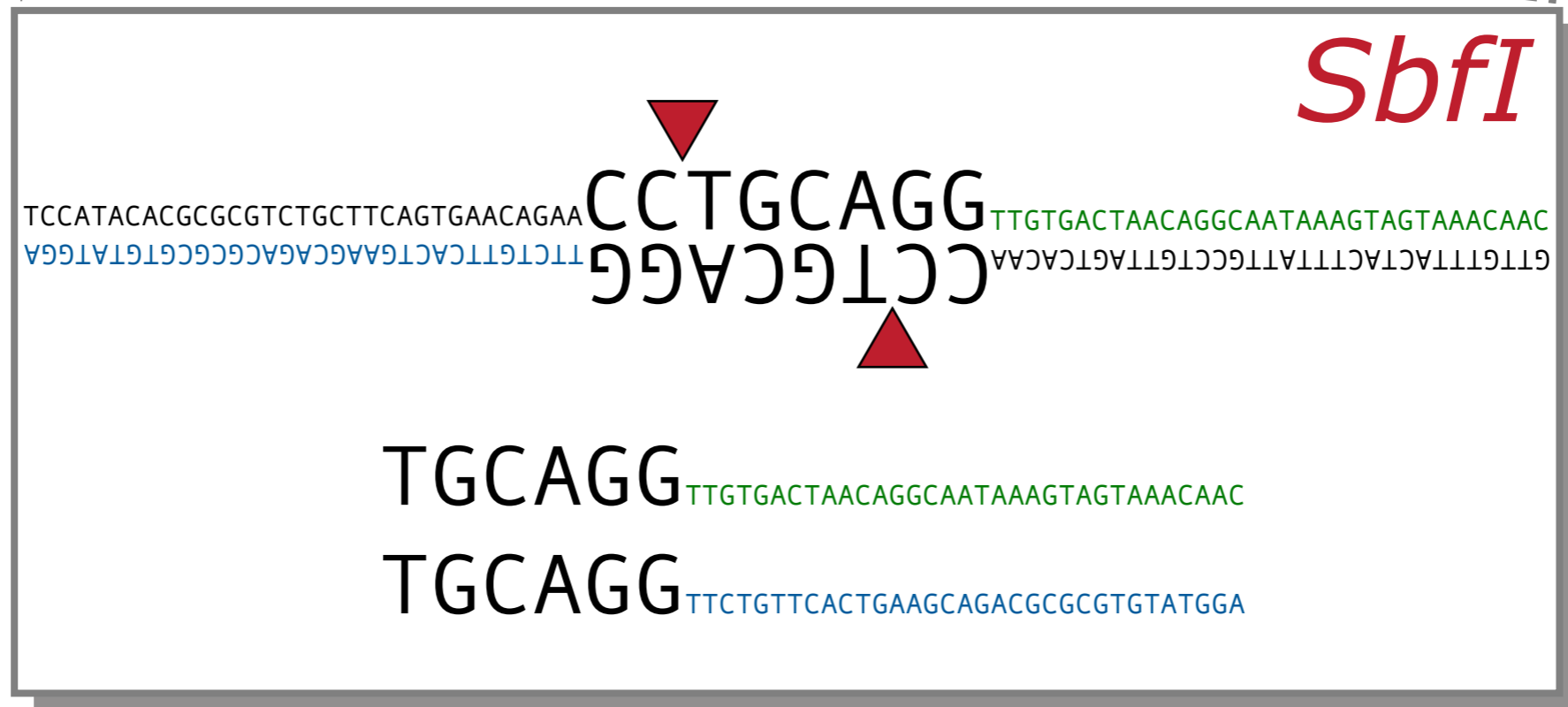
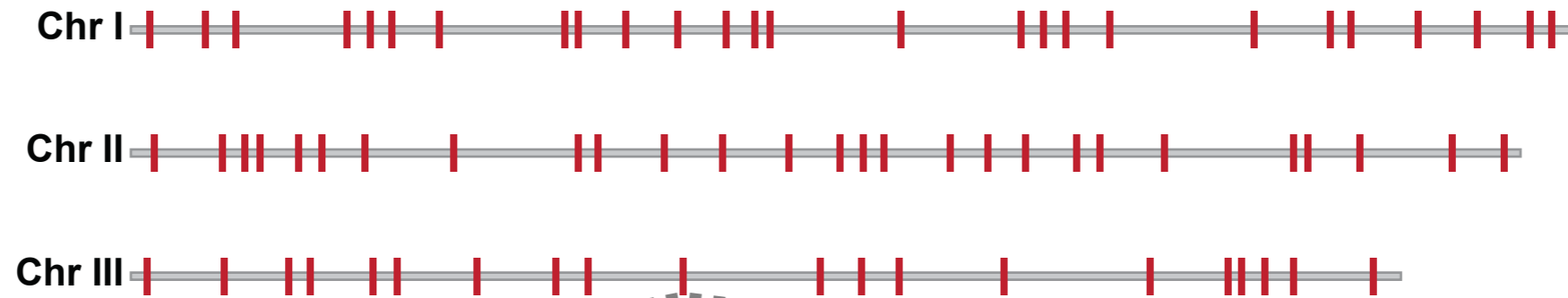
Rapid SNP Discovery and Genetic Mapping Using Sequenced RAD Markers

Nathan A. Baird^{1,2}, Paul D. Etter^{1,2}, Tressa S. Atwood², Mark C. Currey², Anthony L. Shiver¹, Zachary A. Lewis¹, Eric U. Selker¹, William A. Cresko², Eric A. Johnson^{1,4}

¹Institute of Molecular Biology, University of Oregon, Eugene, Oregon, United States of America, ²Illinois, Eugene, Oregon, United States of America, ³The Center for Ecology and Evolutionary Biology, University of Oregon, Eugene, Oregon, United States of America

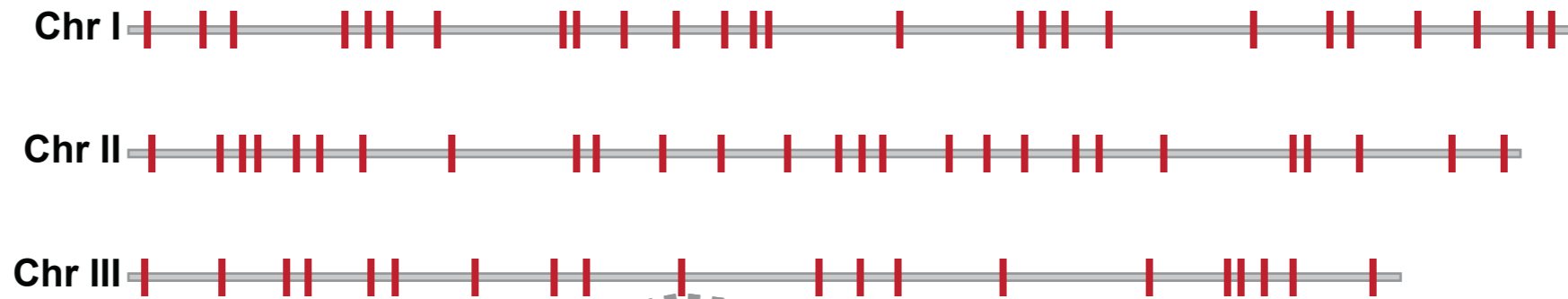
What is RAD-seq?

(Restriction-site Associated DNA)



What is RAD-seq?

(Restriction-site Associated DNA)



2

SbfI

TCCATACACGCGCGTCTGCTTCAGTGAACAGAA CCTGCAGG TTGTGACTAACAGGCAATGAAGTAGTAAACAAC
TTCTGTTCACTGAAGCAGACGCGCGTGTATGGA CCTGCAGG TTCTGTTCACTGAAGCAGACGCGCGTGTATGGA

TGCAGG TTGTGACTAACAGGCAAT ^{G/A} AAGTAGTAAACAAC

TGCAGG TTCTGTTCACTGAAGCAGACGCGCGTGTATGGA

1

TCCATACACGCGCGTCTGCTTCAGTGAACAGAA CCTGCAGG TTGTGACTAACAGGCAATAAAGTAGTAAACAAC
TTCTGTTCACTGAAGCAGACGCGCGTGTATGGA CCTGCAGG TTCTGTTCACTGAAGCAGACGCGCGTGTATGGA

TGCAGG TTGTGACTAACAGGCAATAAAGTAGTAAACAAC

TGCAGG TTCTGTTCACTGAAGCAGACGCGCGTGTATGGA

What is RAD-seq?

(Restriction-site Associated DNA)



22,830 *SbfI* sites in threespine stickleback

~ 45,000 RAD-Tags

HiSeq Illumina Lane:

160 million reads, 96 barcoded individuals

①

SbfI



TGCAGG TTGTGACTAACAGGCAATAAAGTAGTAAACAAC
TGCAGG TTCTGTTCACTGAAGCAGACGCGGTGTATGGA

②

SbfI



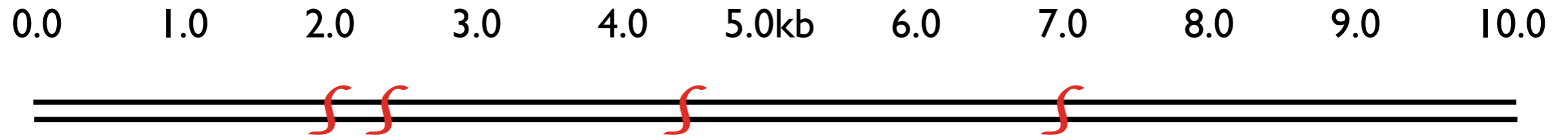
TGCAGG TTGTGACTAACAGGCAAT ^{G/A} AAGTAGTAAACAAC
TGCAGG TTCTGTTCACTGAAGCAGACGCGGTGTATGGA

Restriction Enzyme (RE) digestion and first adaptor ligation

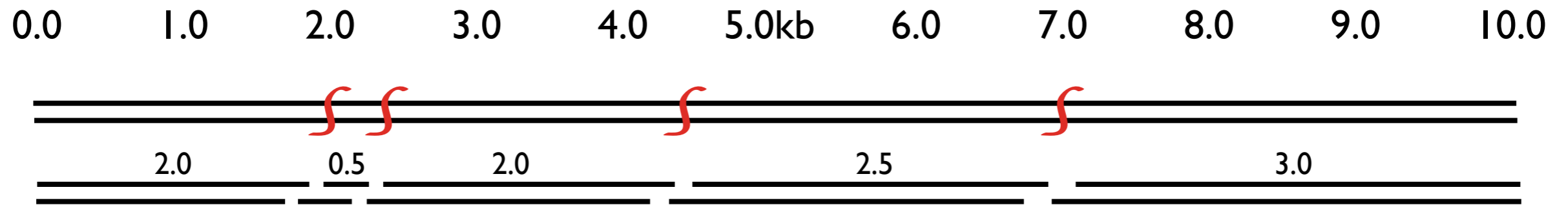
0.0 1.0 2.0 3.0 4.0 5.0kb 6.0 7.0 8.0 9.0 10.0



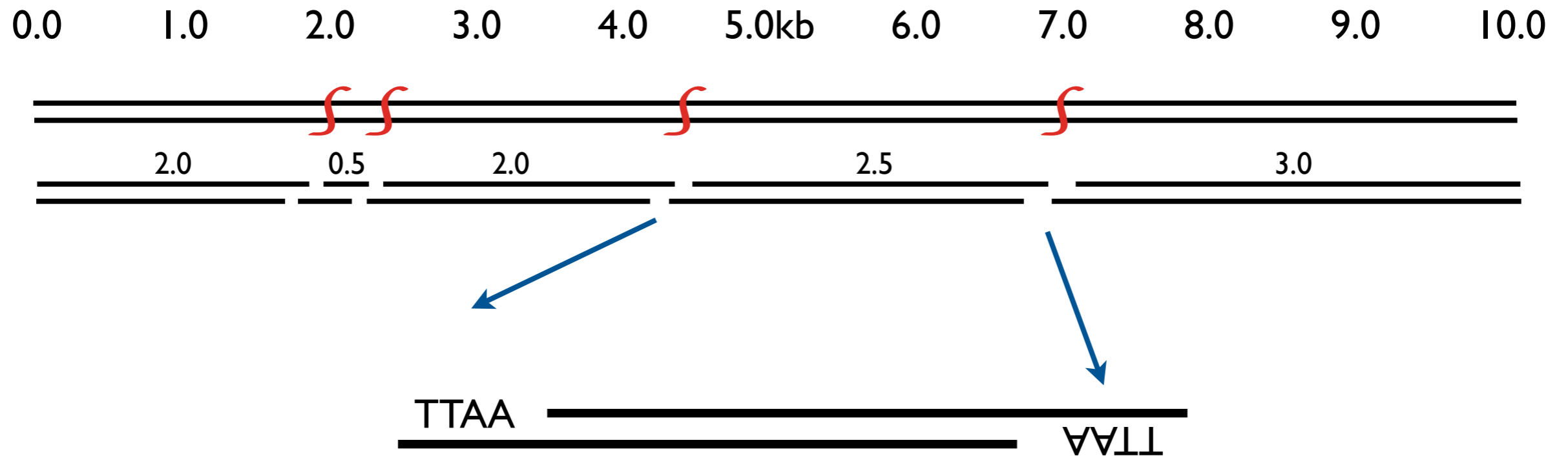
Restriction Enzyme (RE) digestion and first adaptor ligation



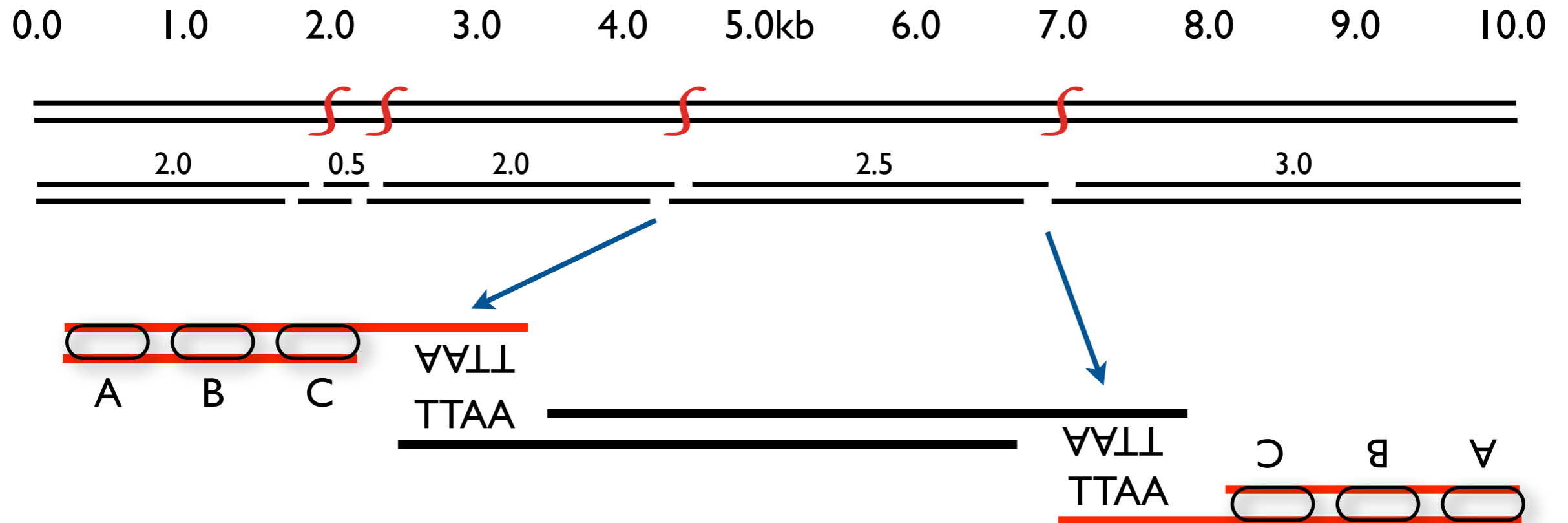
Restriction Enzyme (RE) digestion and first adaptor ligation



Restriction Enzyme (RE) digestion and first adaptor ligation

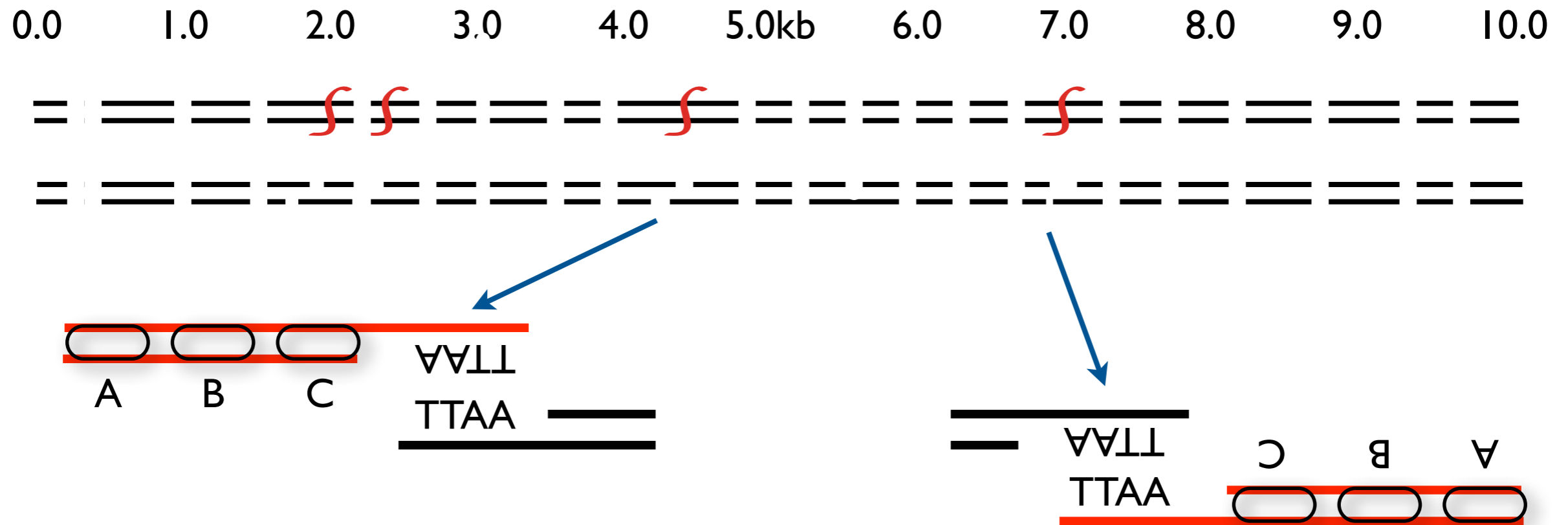


Restriction Enzyme (RE) digestion and first adaptor ligation



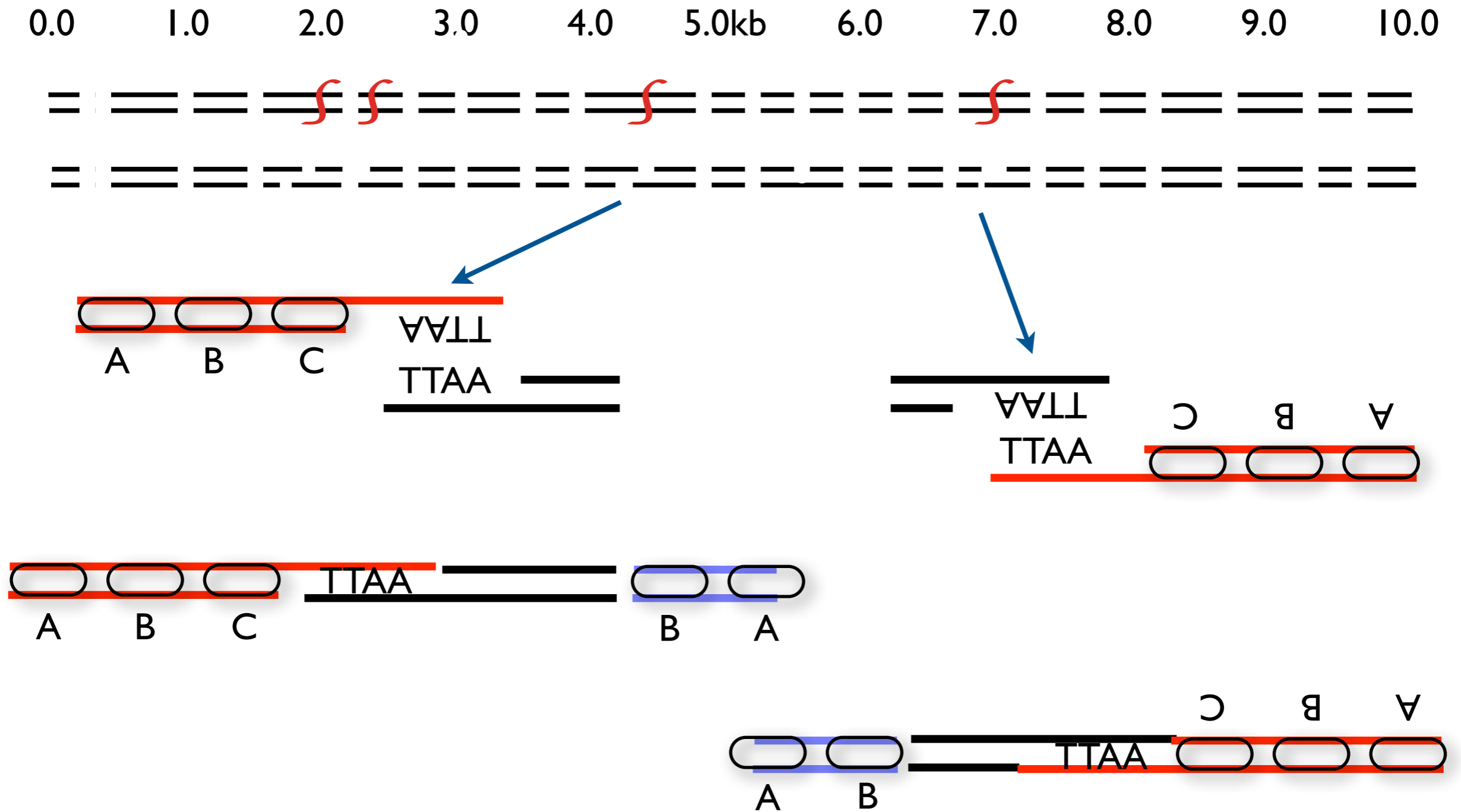
- A = Amplification primer
- B = Sequencing primer
- C = Barcode

Shearing and second adaptor ligation



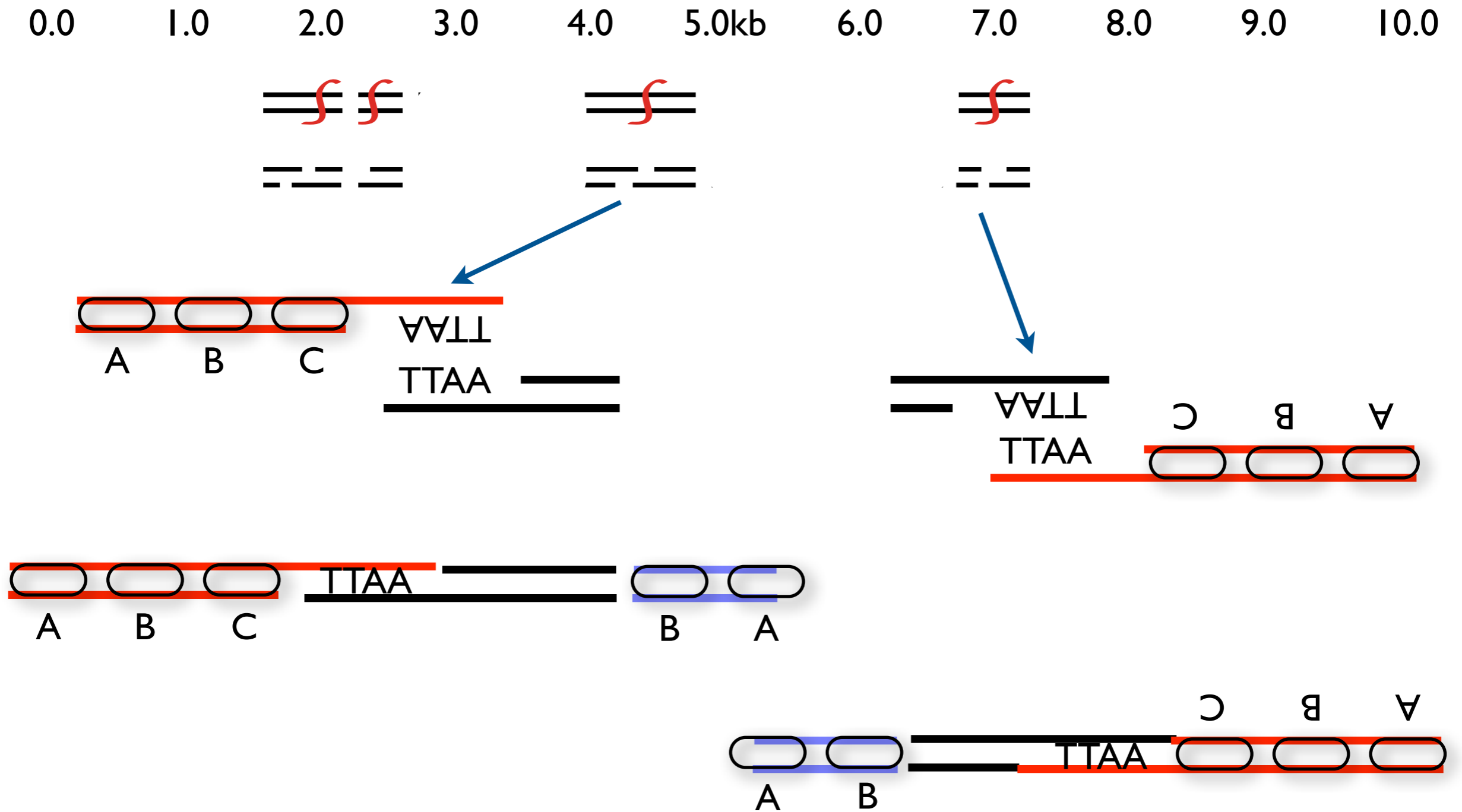
A = Amplification primer
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Shearing and second adaptor ligation



A = Amplification primer
B = Sequencing primer
C = Barcode

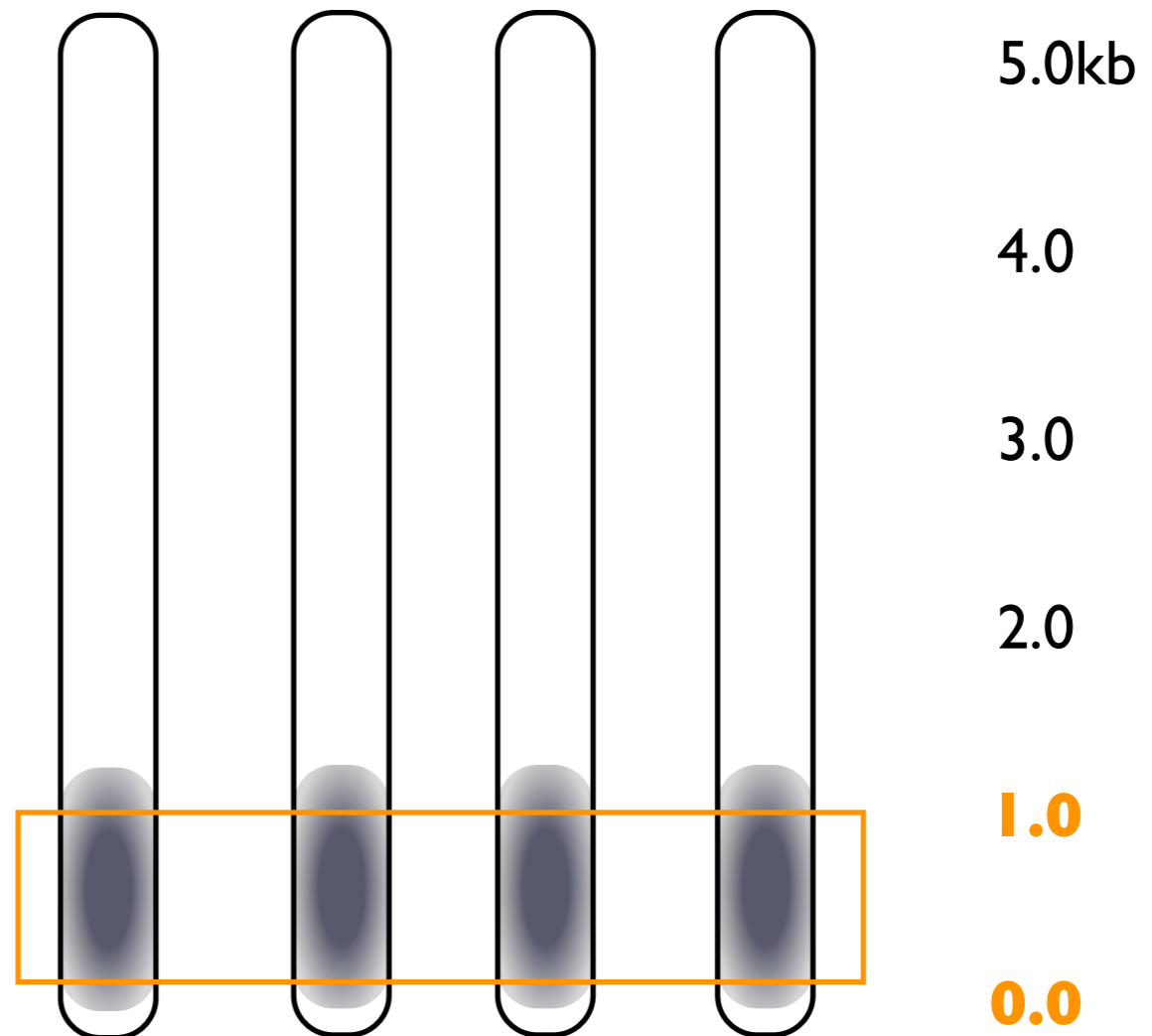
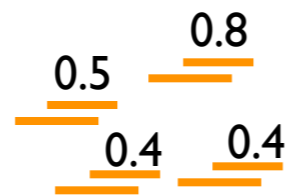
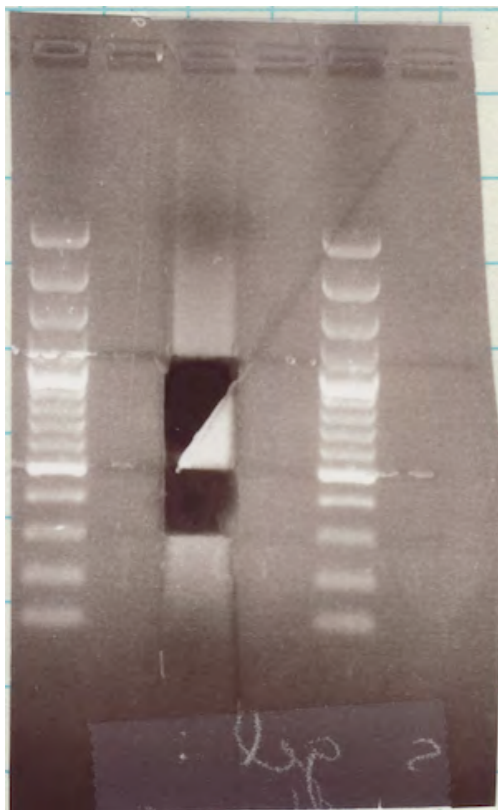
Shearing and second adaptor ligation



A = Amplification primer
B = Sequencing primer
C = Barcode

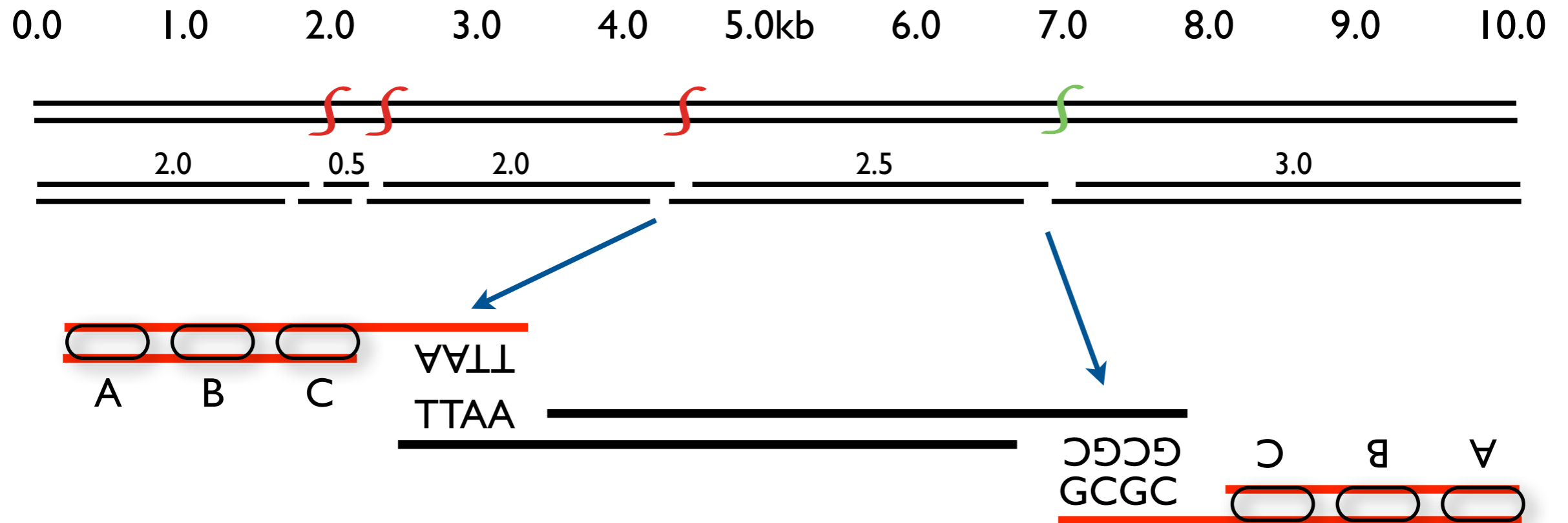
Shearing makes consistent fragments for sequencing

0.0 1.0 2.0 3.0 4.0 5.0kb 6.0 7.0 8.0 9.0 10.0



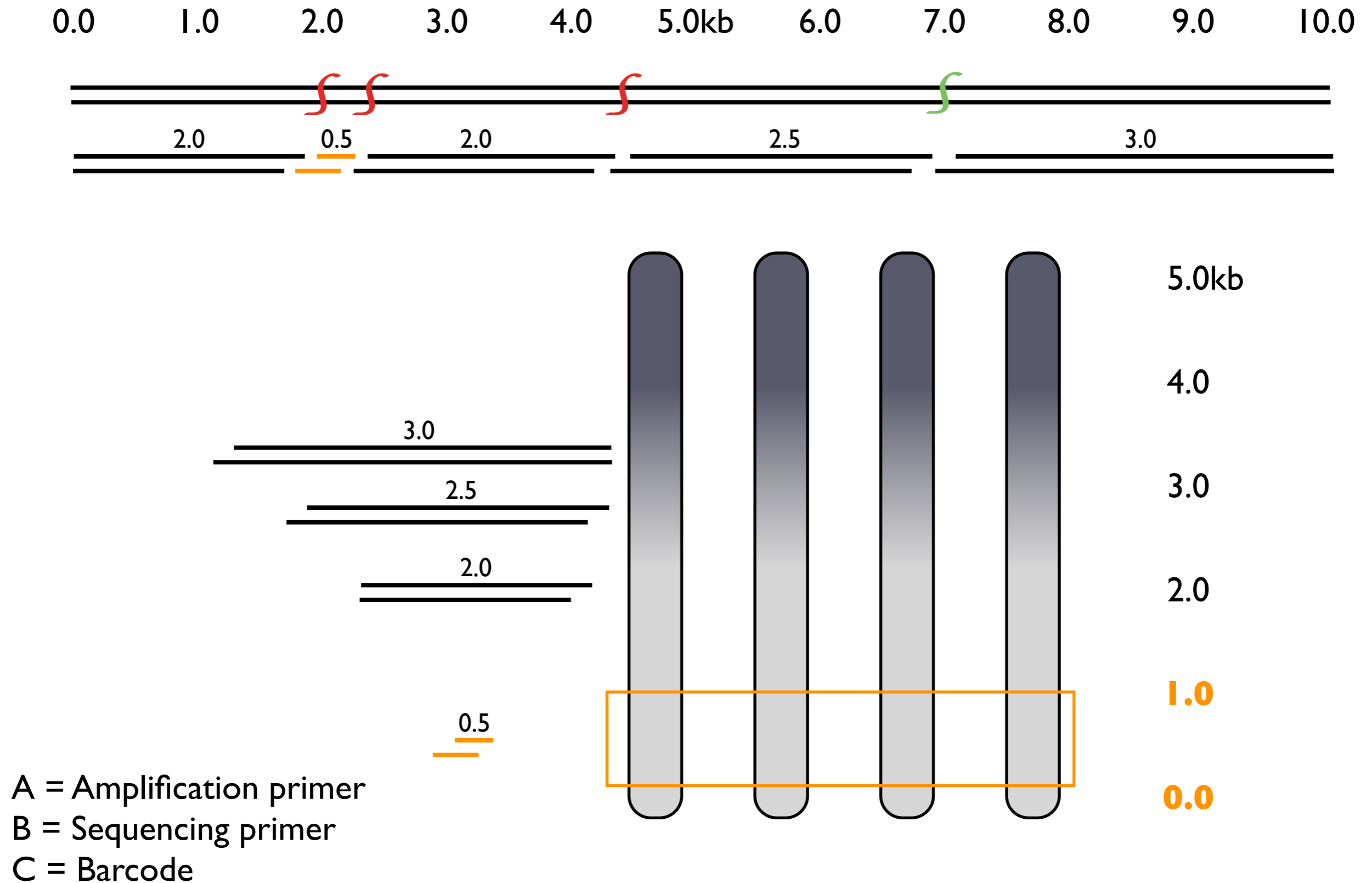
A = Amplification primer
B = Sequencing primer
C = Barcode

Single (GBS) or Double Digest RAD (ddRAD)

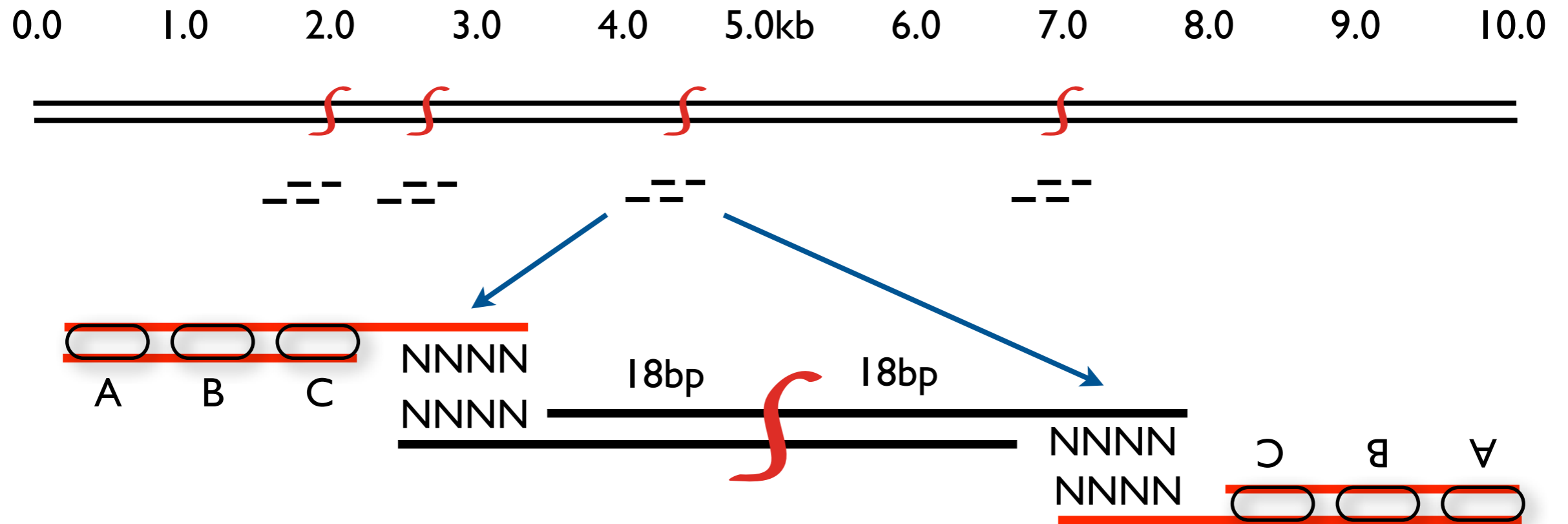


A = Amplification primer
B = Sequencing primer
C = Barcode

Size selection is more problematic without shearing



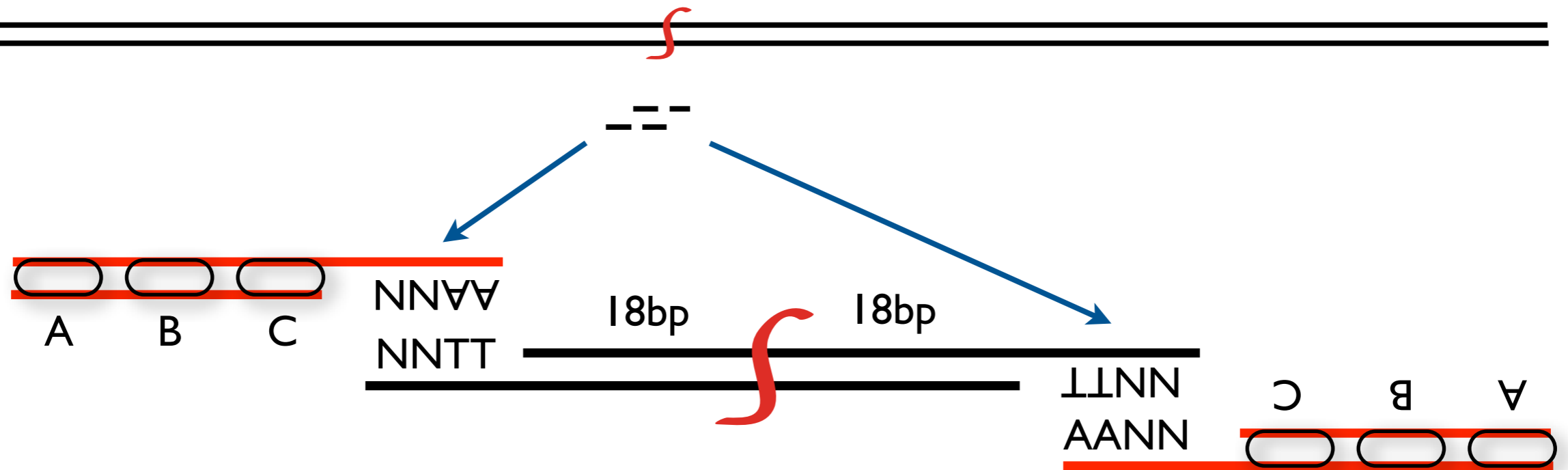
2bRAD - type 2b restriction enzyme



A = Amplification primer
B = Sequencing primer
C = Barcode

2bRAD - can scale number of markers easily

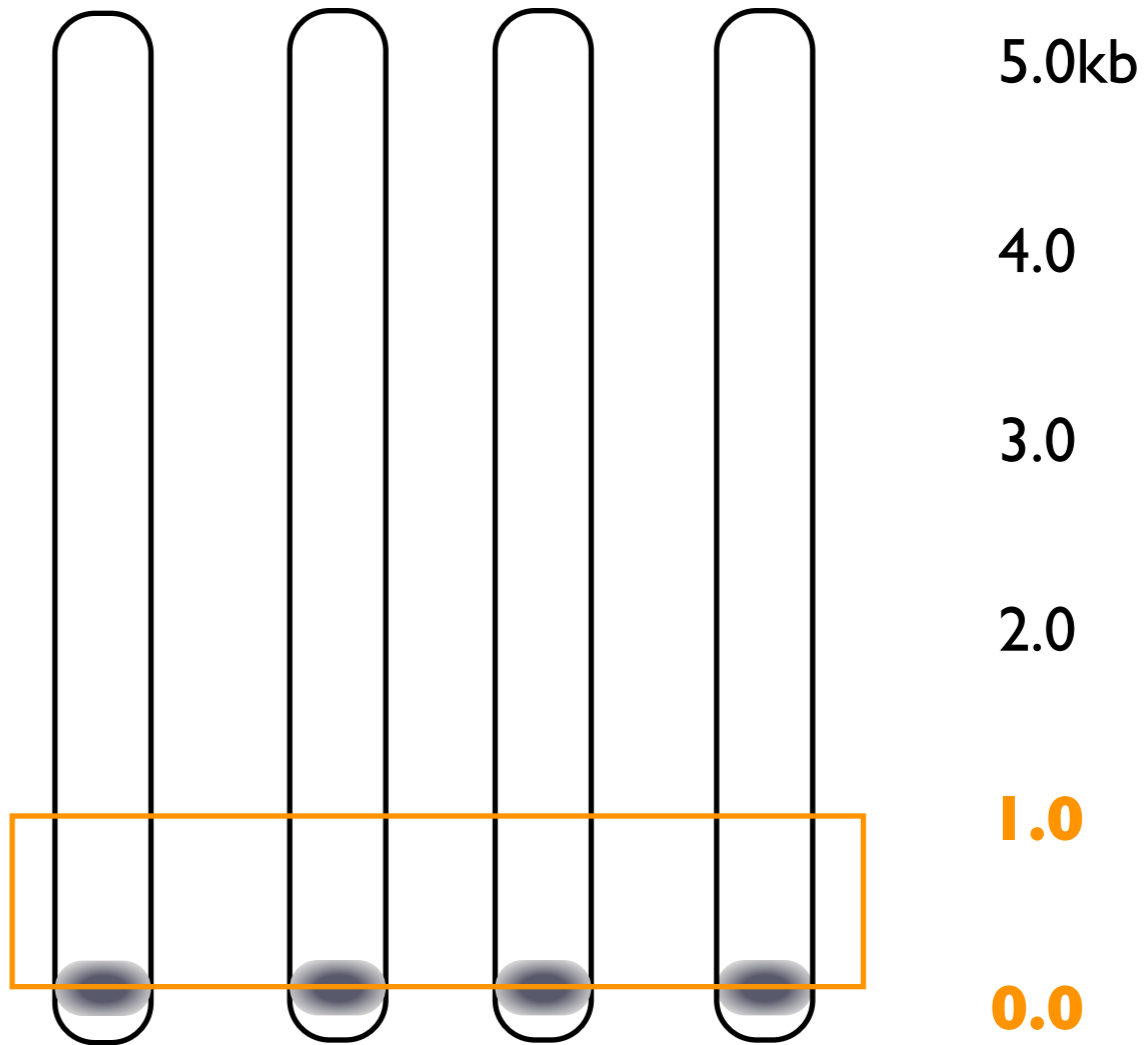
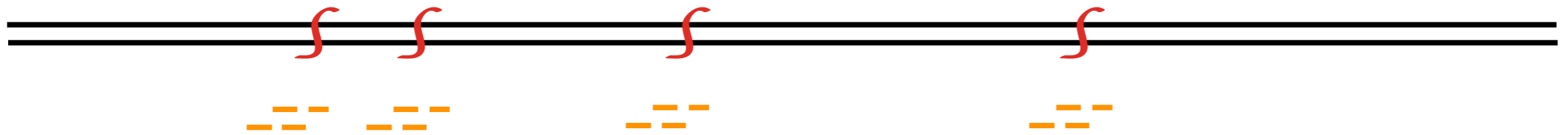
0.0 1.0 2.0 3.0 4.0 5.0kb 6.0 7.0 8.0 9.0 10.0



- A = Amplification primer
- B = Sequencing primer
- C = Barcode

2bRAD - size selection is difficult

0.0 1.0 2.0 3.0 4.0 5.0kb 6.0 7.0 8.0 9.0 10.0



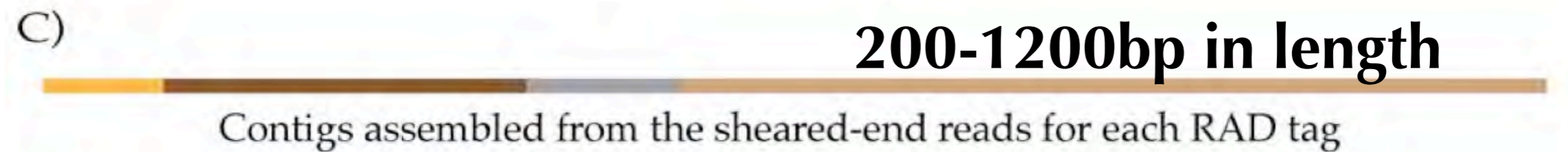
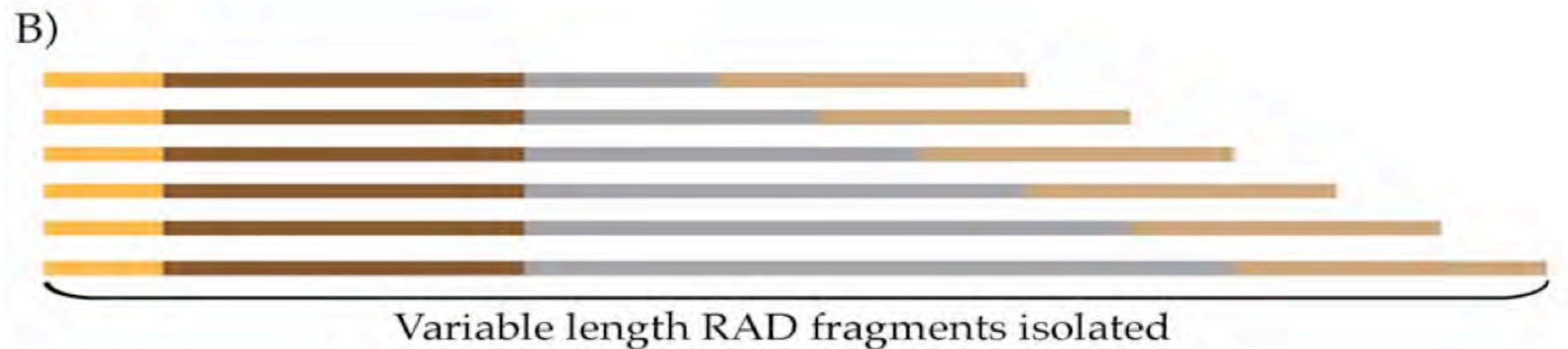
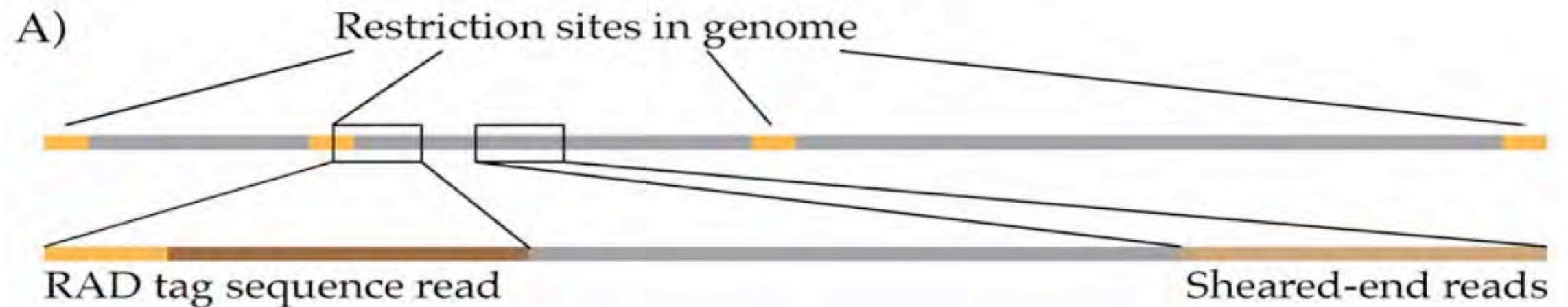
36bp

A = Amplification primer
B = Sequencing primer
C = Barcode

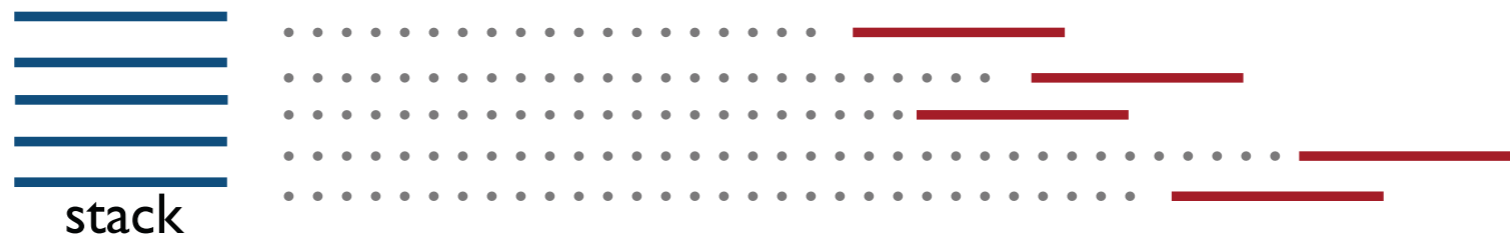
Summary of plusses and minuses of RAD family

	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- Easier marker scaling	<ul style="list-style-type: none">- Fewest steps- Easiest 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

Benefits of random shearing in RAD



Acquire
paired-end
sequence



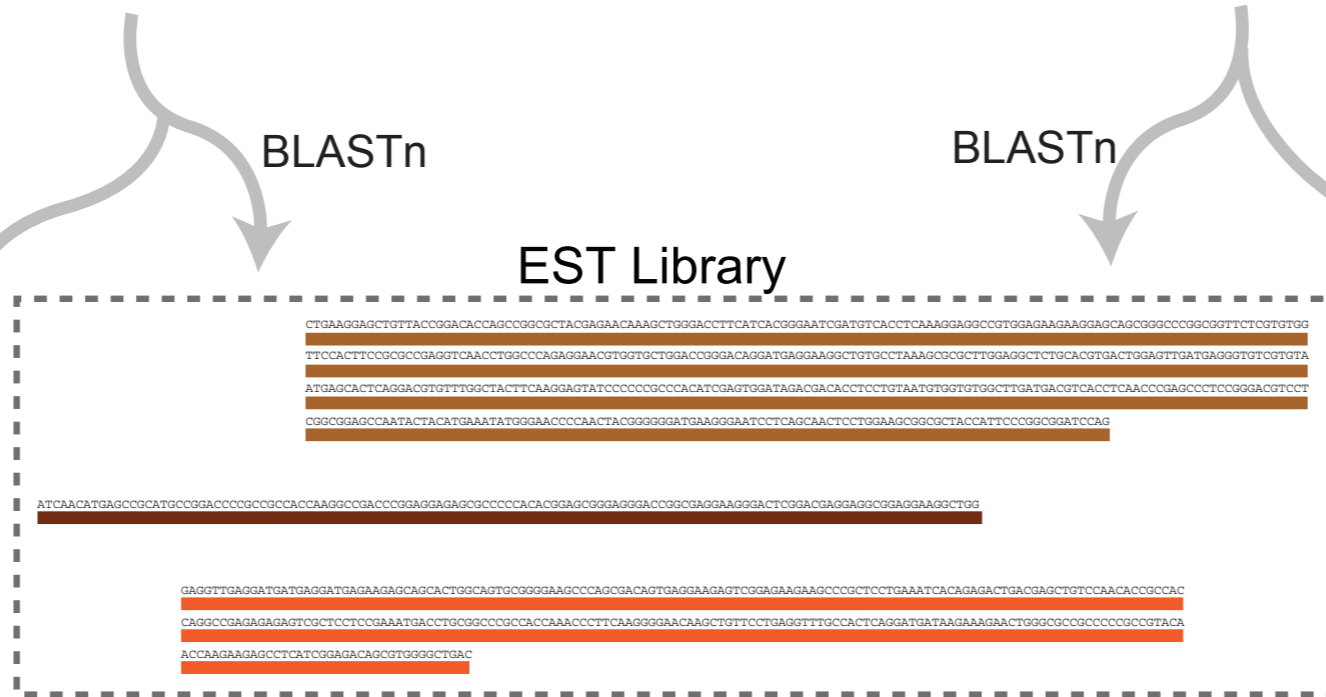
Match to marker catalog

TGCAGGGGTATTAGCATAA

Collate/Assemble PE reads

AACTAATTTTTCACTAGCCATCTTGAATGTGAGTAGCATTTTAAAGTAACTATAATTG

Associate
markers / PE
contigs with
ESTs

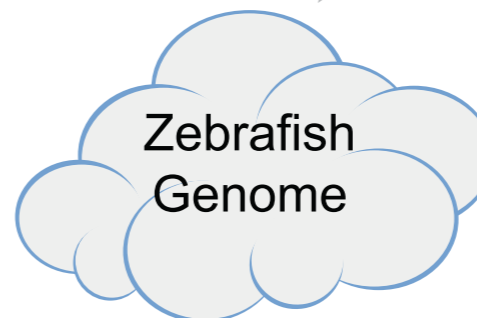
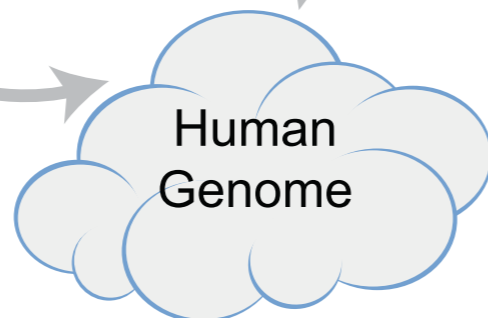


Assign
orthology to:
markers
PE contigs
ESTs

BLASTx

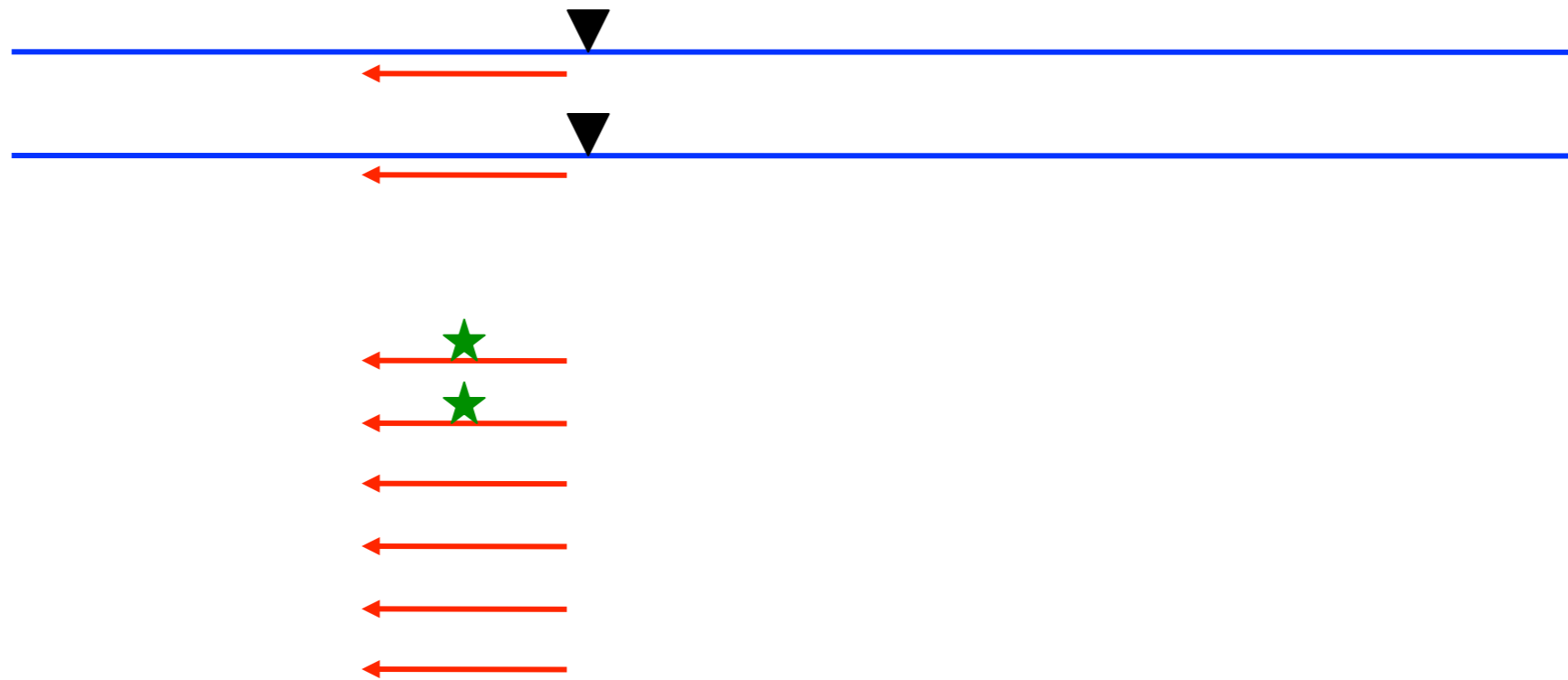
BLASTx

BLASTx



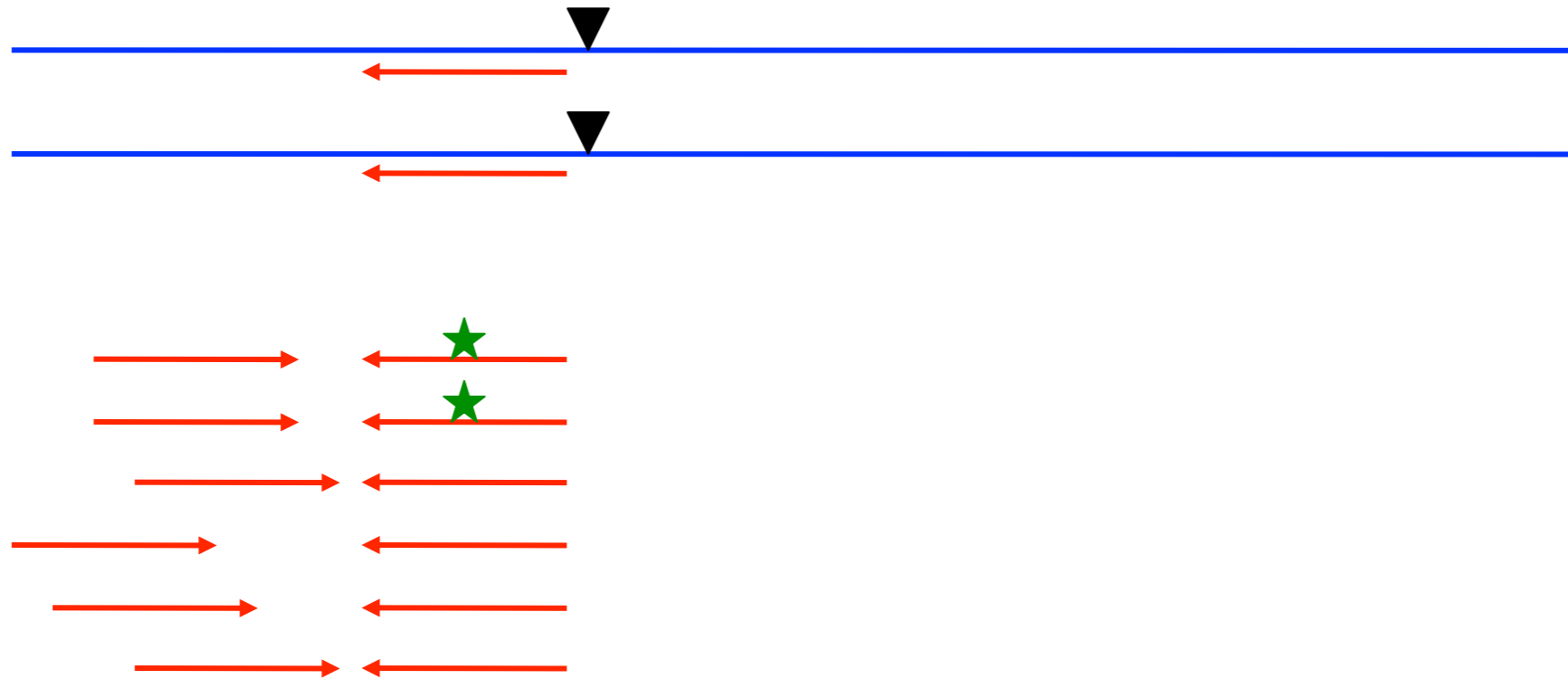
Random shearing benefits in RAD

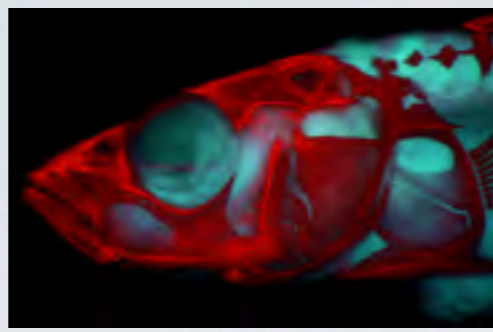
Eliminating PCR duplicates:



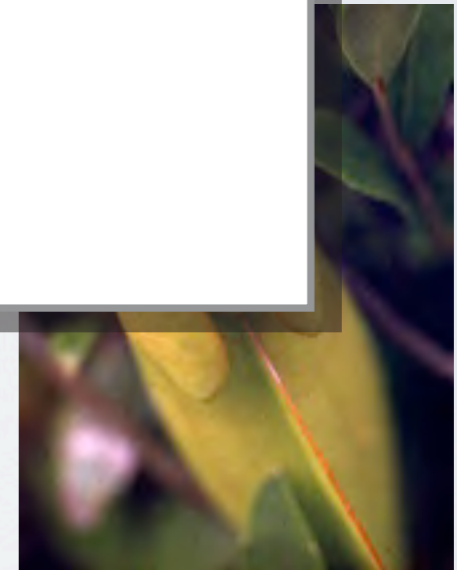
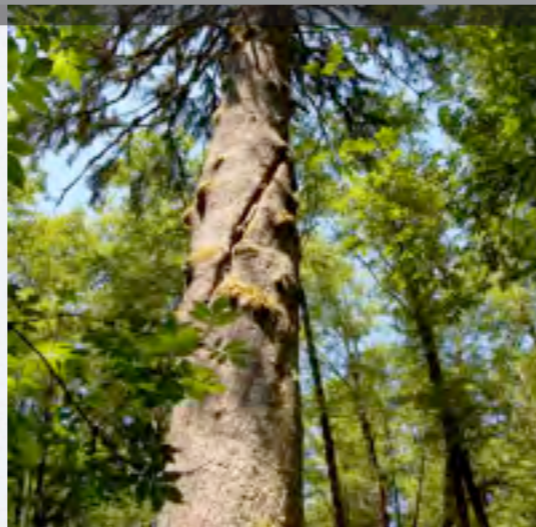
Random shearing benefits in RAD

Eliminating PCR duplicates:





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

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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

Differentiating SNPs from error

Restriction enzyme recognition site

Reference genome sequence

sequence reads

```
CTTCAGGTTGGGTGAGTTGTCATCAGTCGGAATGCGCAGGTCACCTTACCTGCAGGCAGCTCTCTGAAGCGCAGGTACTCCATCGACCGGGTGGTGACTAG
NNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNNN
ATCAGTCGGAATGCGCAGGTCACCTTACCTGCA
ATCAGTCGGAATGCGCAGGTCACCTTACCTGCA
ATCAGTCtGAATGTCAGGTCaCTTAcctGcA
ATCAGTCGGAATGCGCAGGTCACCTTAcctGcA
ATCAtTCGGAATGCGCAGGTCACCTTAcctgca
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ATCAGTCGGAATGCGCAGGTCACCTTAcctgca
ATCAGTCGGAATGCGCAGGTCACCTTAcctgca
ATCAGTCGGAATGCGCAGGTCACCTTAcctgca
ATCAGTCGGAATGCGCAGGTCACCTTAcctgca
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TGCaggCAGCTCTCTGAAGCGCAGGtACTCCA
TgCaggCAGCTCTCTGAAGCGCAGGtAcTccA
TGCaGgCAGCTCTCTGAAGCGCAGGtACTCCA
tgcaggCAGCTcTATGAAGcGCAGGtActcca
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TgCaggCAGCTCTCTGAAGCGCAGGtAcTCCA
```

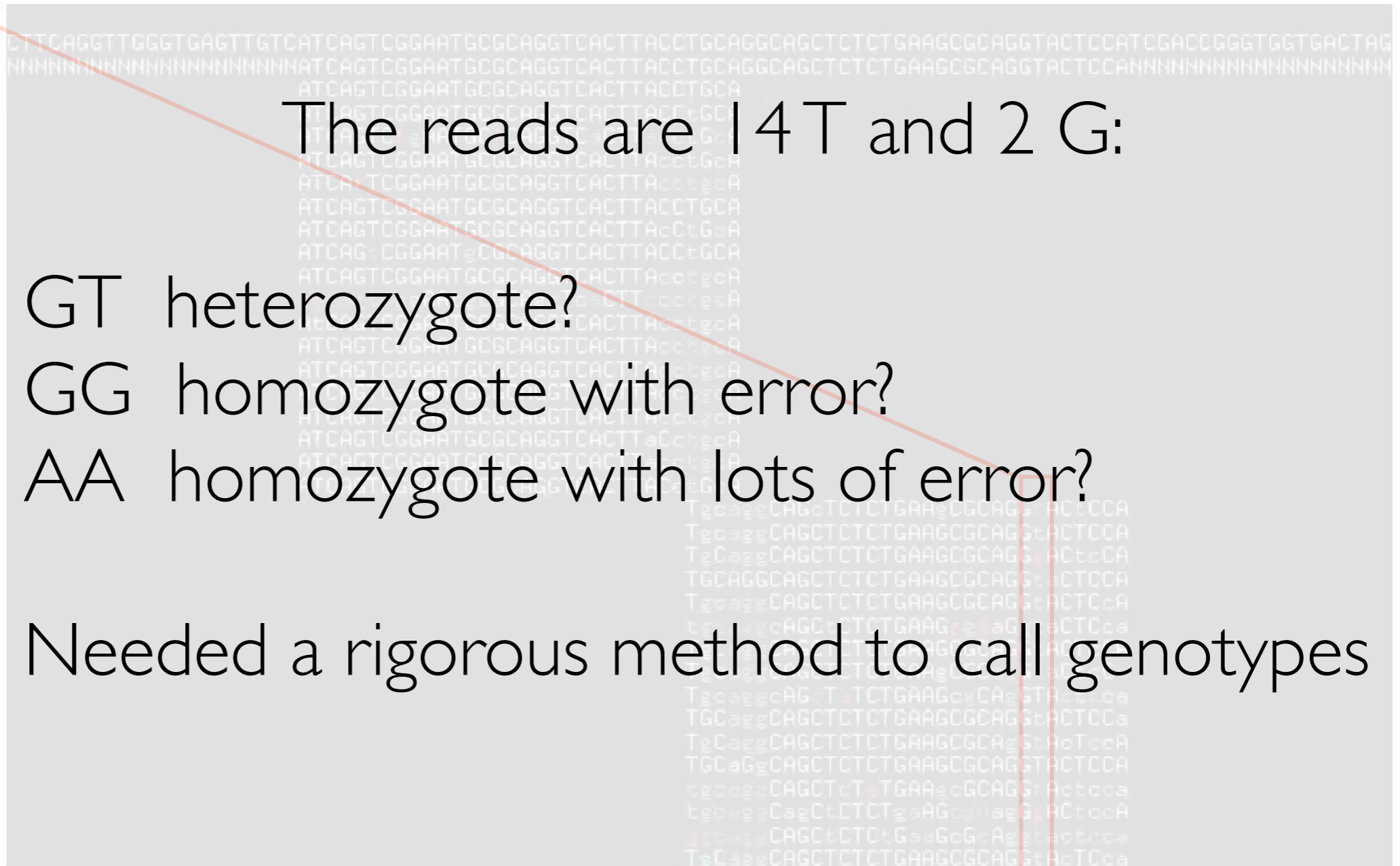

Differentiating SNPs from error

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

The reads are 14 T and 2 G:

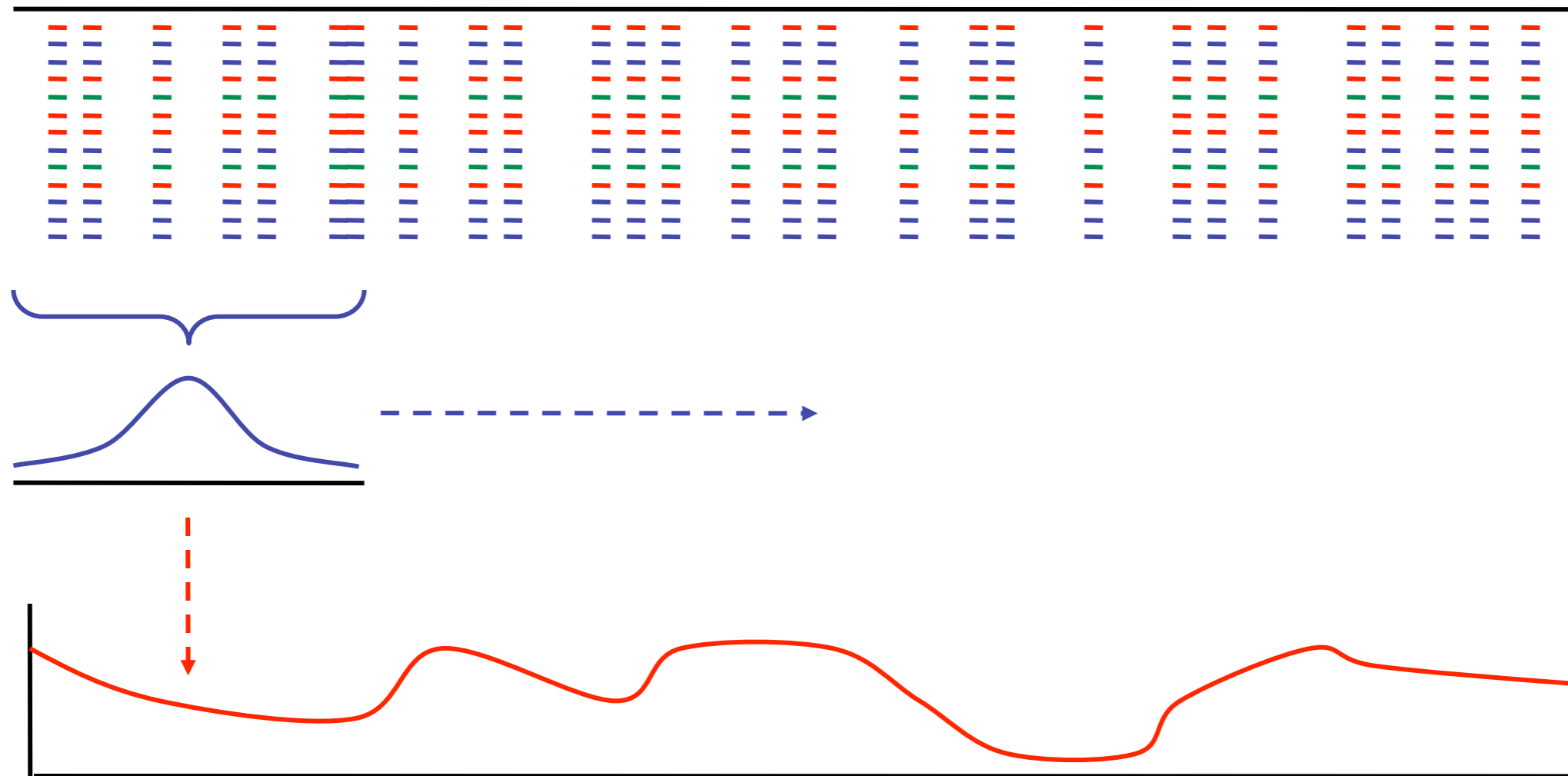
GT heterozygote?
GG homozygote with error?
AA homozygote with lots of error?

Needed a rigorous method to call genotypes



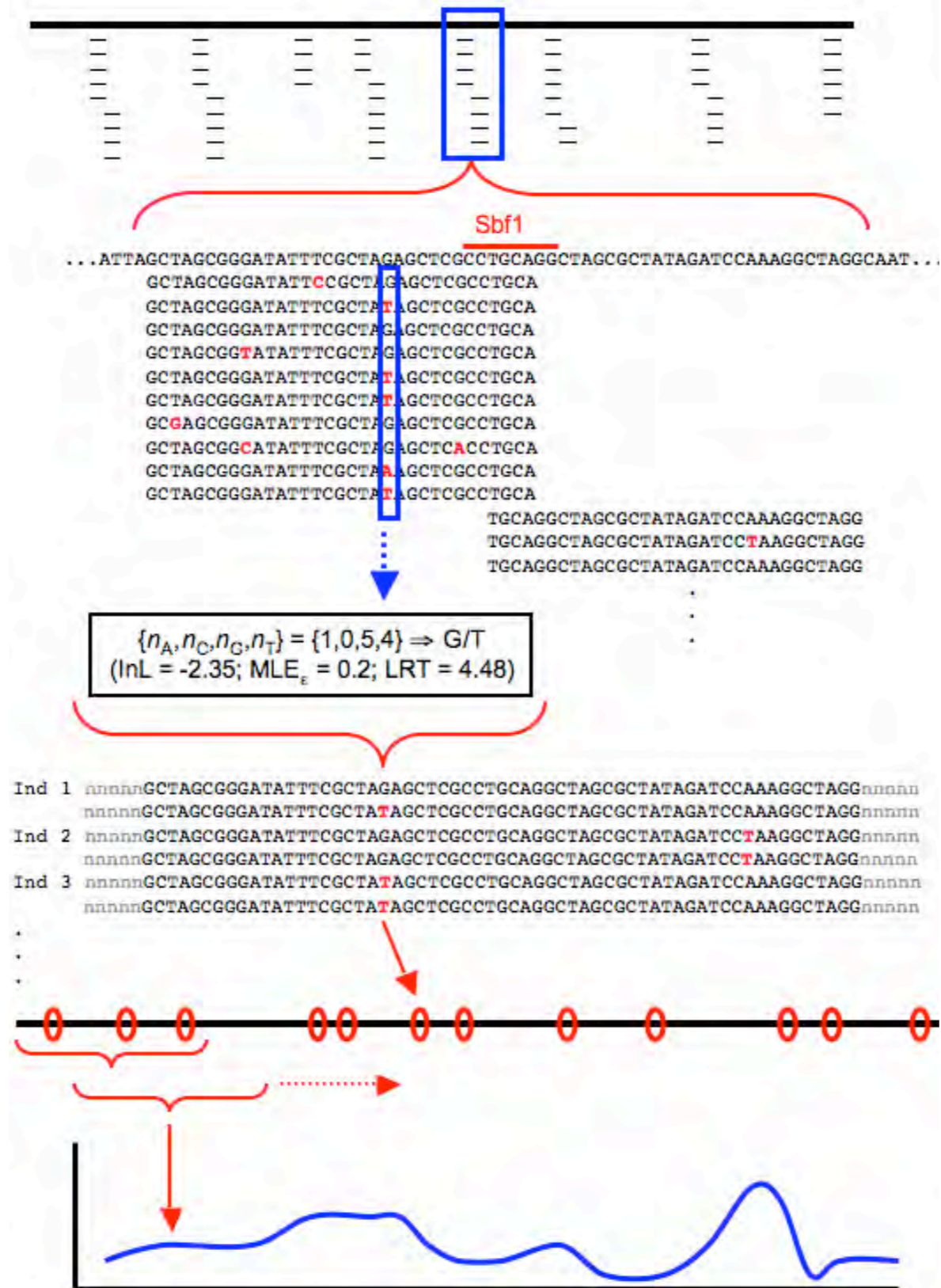
Making statistics continuous across the genome

Kernel-smoothing average of summary statistics along genome



Bootstrap re-sampling to estimate significance of moving average

Overall pipeline



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

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

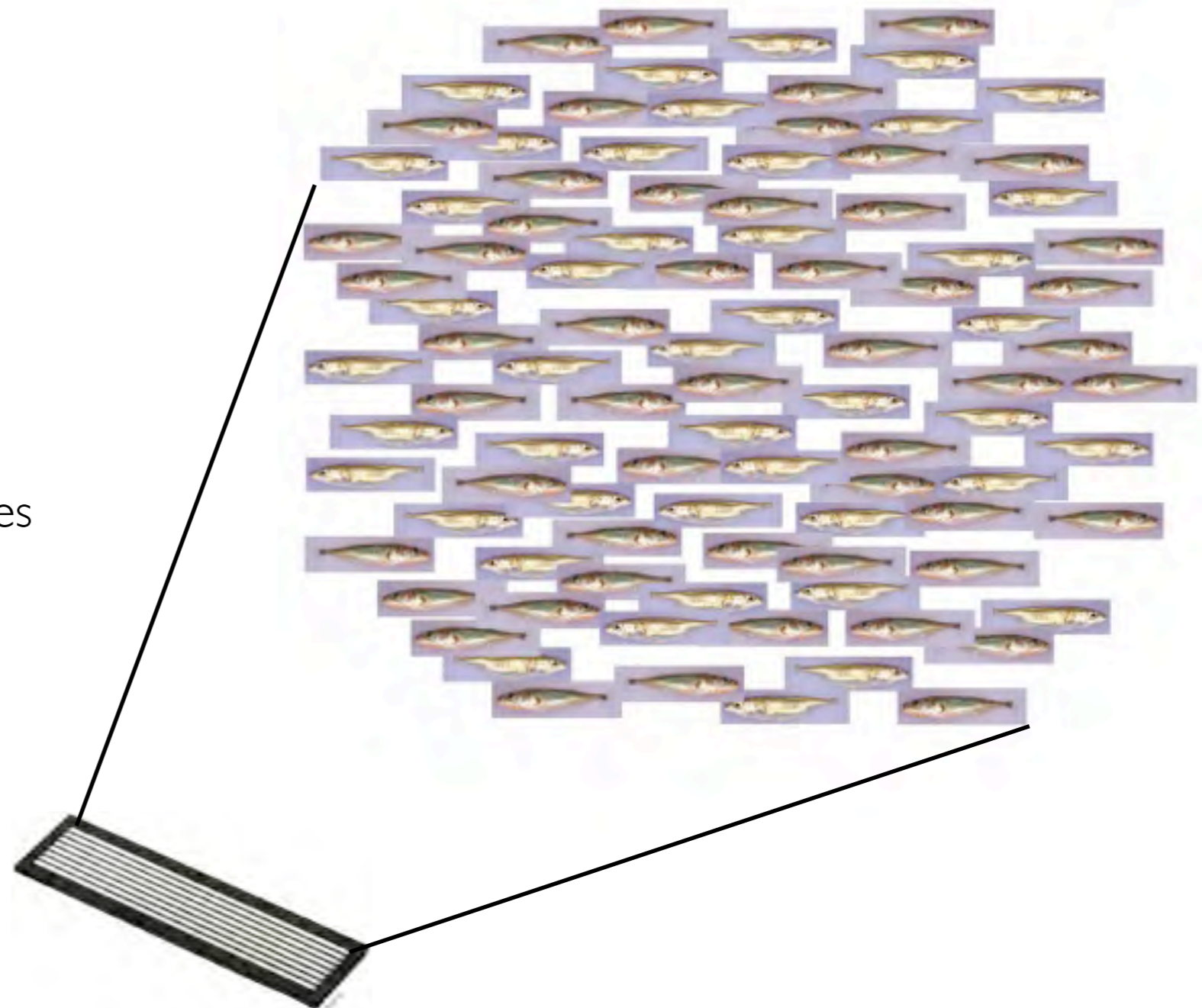
Barcoded adapters

5 to 8nt barcodes

Variable length barcodes

Combinatorial barcodes (PE)

Barcode distance - two mismatches



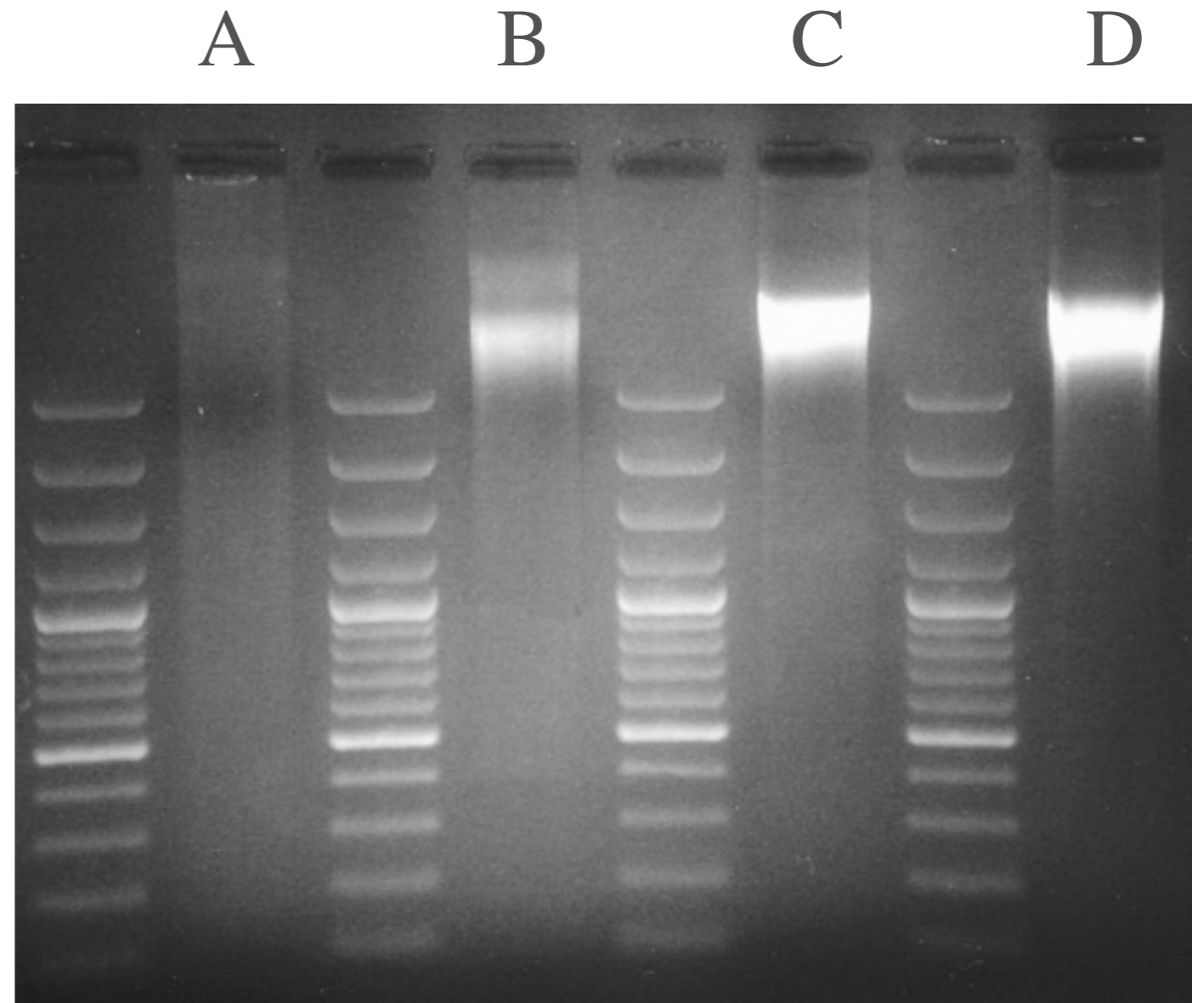
Molecular considerations in library building

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

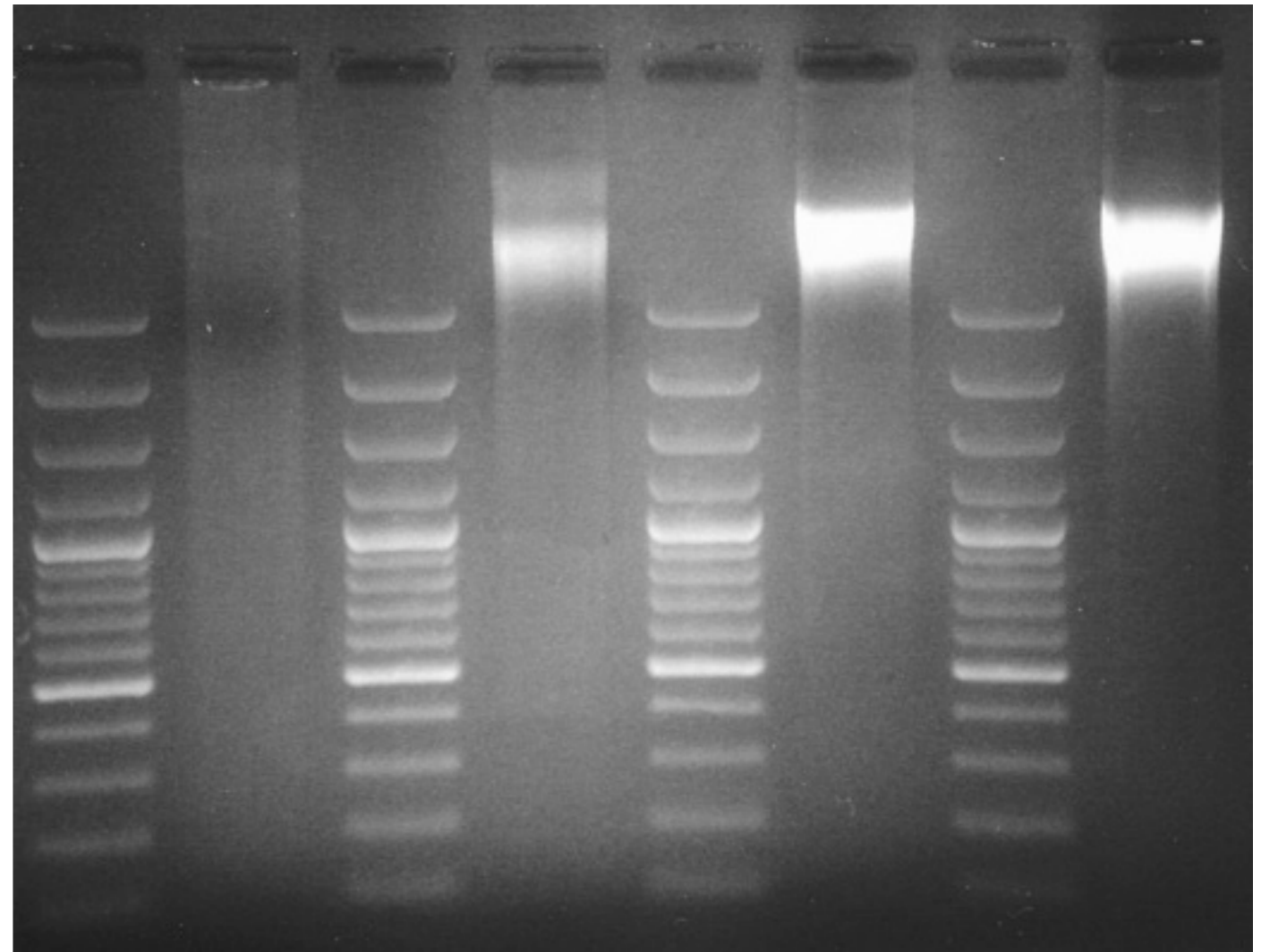
Illumina cluster calling is confused by repetition in first 4 bases - can offset barcodes

CGATA

GTACA

TAGCC

ACTGC



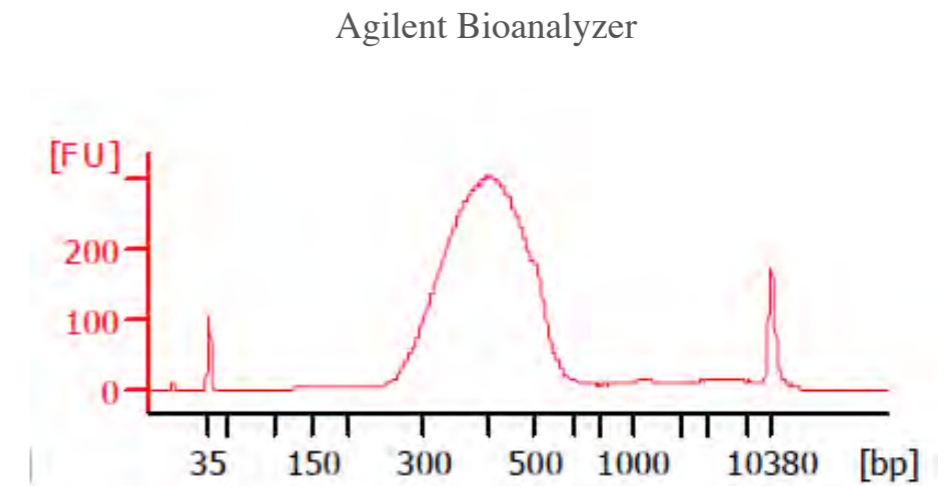
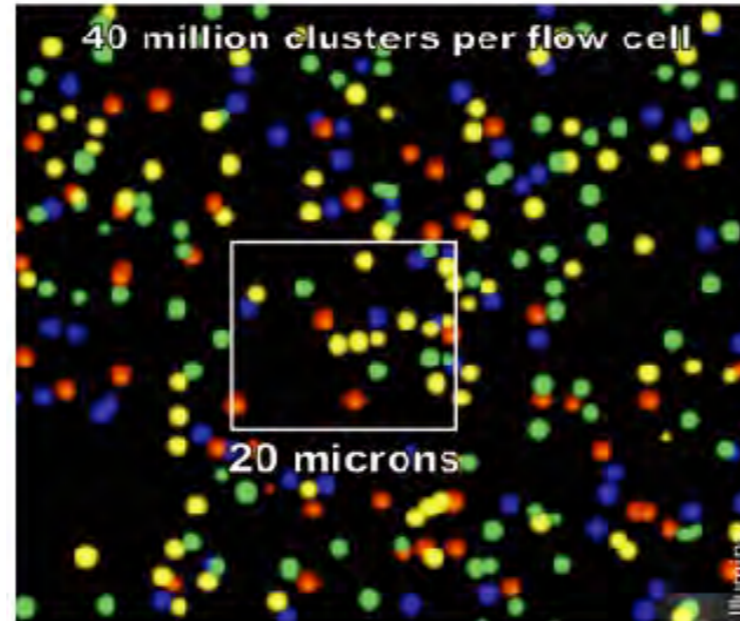
Molecular considerations in library building

How can I get the best depth of coverage?

Things to consider

Fragment size

Smaller/tighter is better



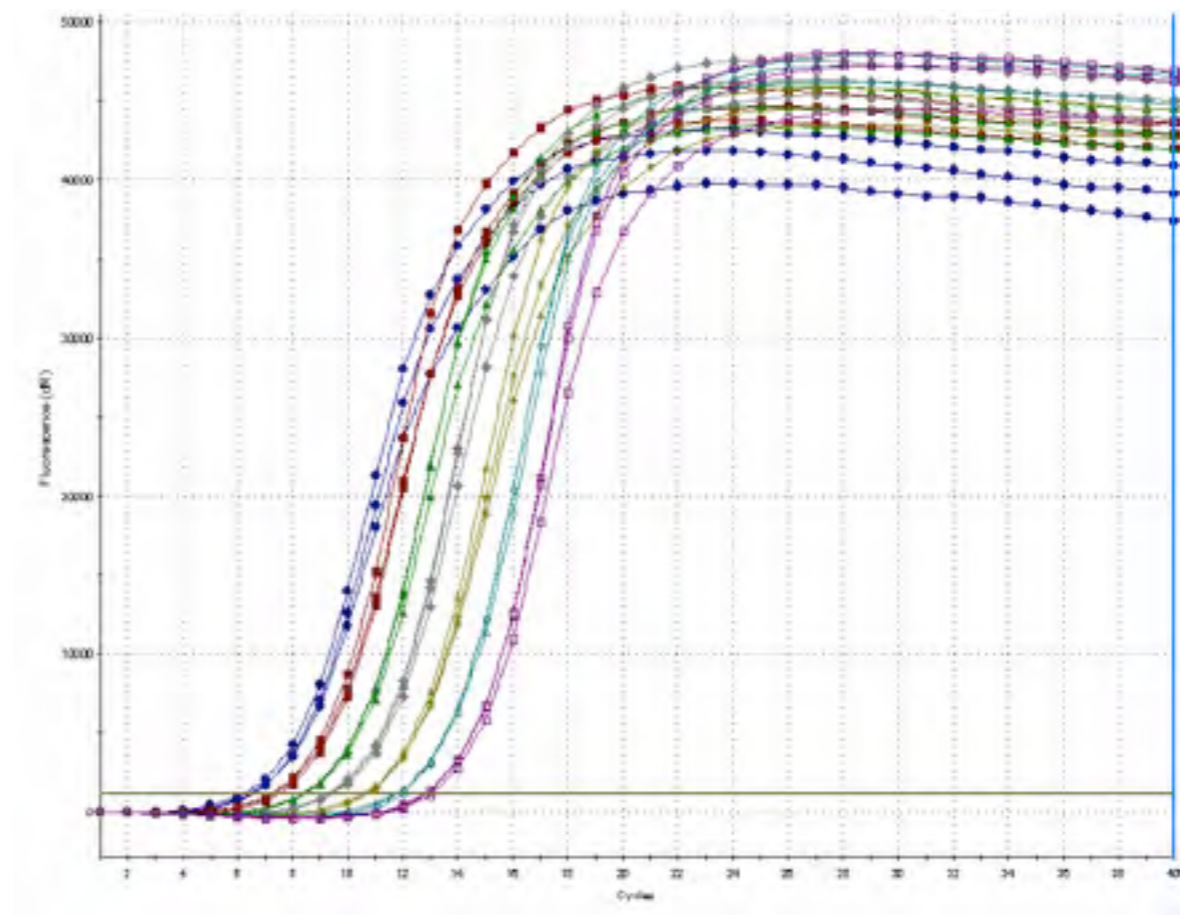
Molecular considerations in library building

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

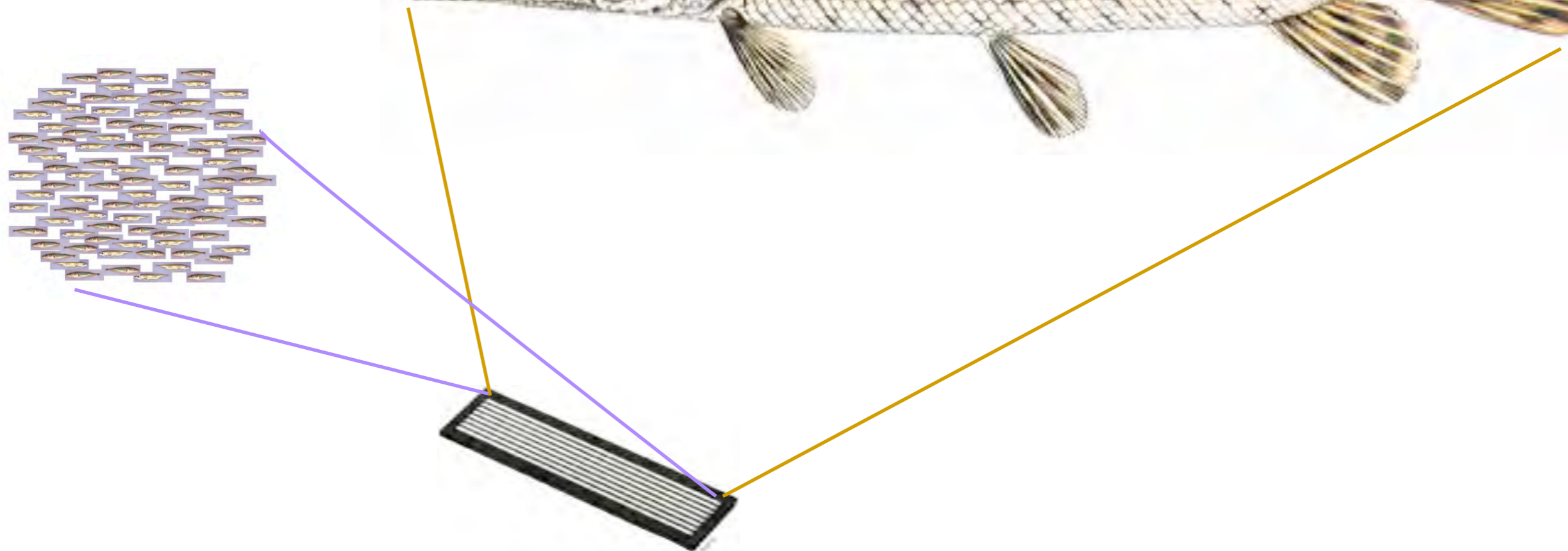
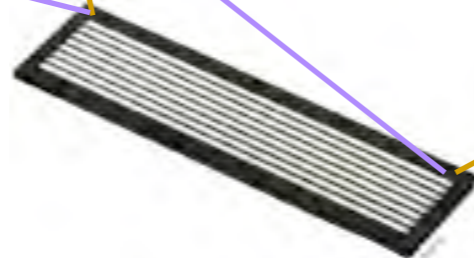
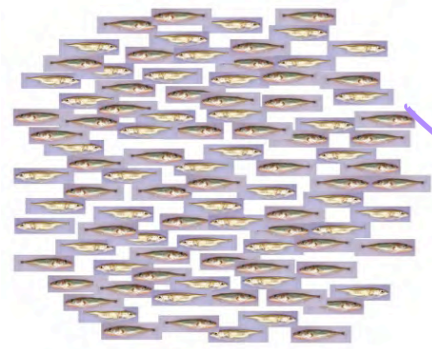
Fragment size

Library quality

qPCR

Pilot Experiment:

Spike or split a lane



A case study of using RAD for an organism with a reference genome: population genomics of threespine stickleback fish



Susie Bassham, Julian Catchen, Paul Hohenlohe
Emily Lescak and Frank von Hippel

Threespine stickleback, *Gasterosteus aculeatus*

- *Ancestral Oceanic Populations*
Marine and Anadromous
Old (> 10 million years)
Phenotypically similar

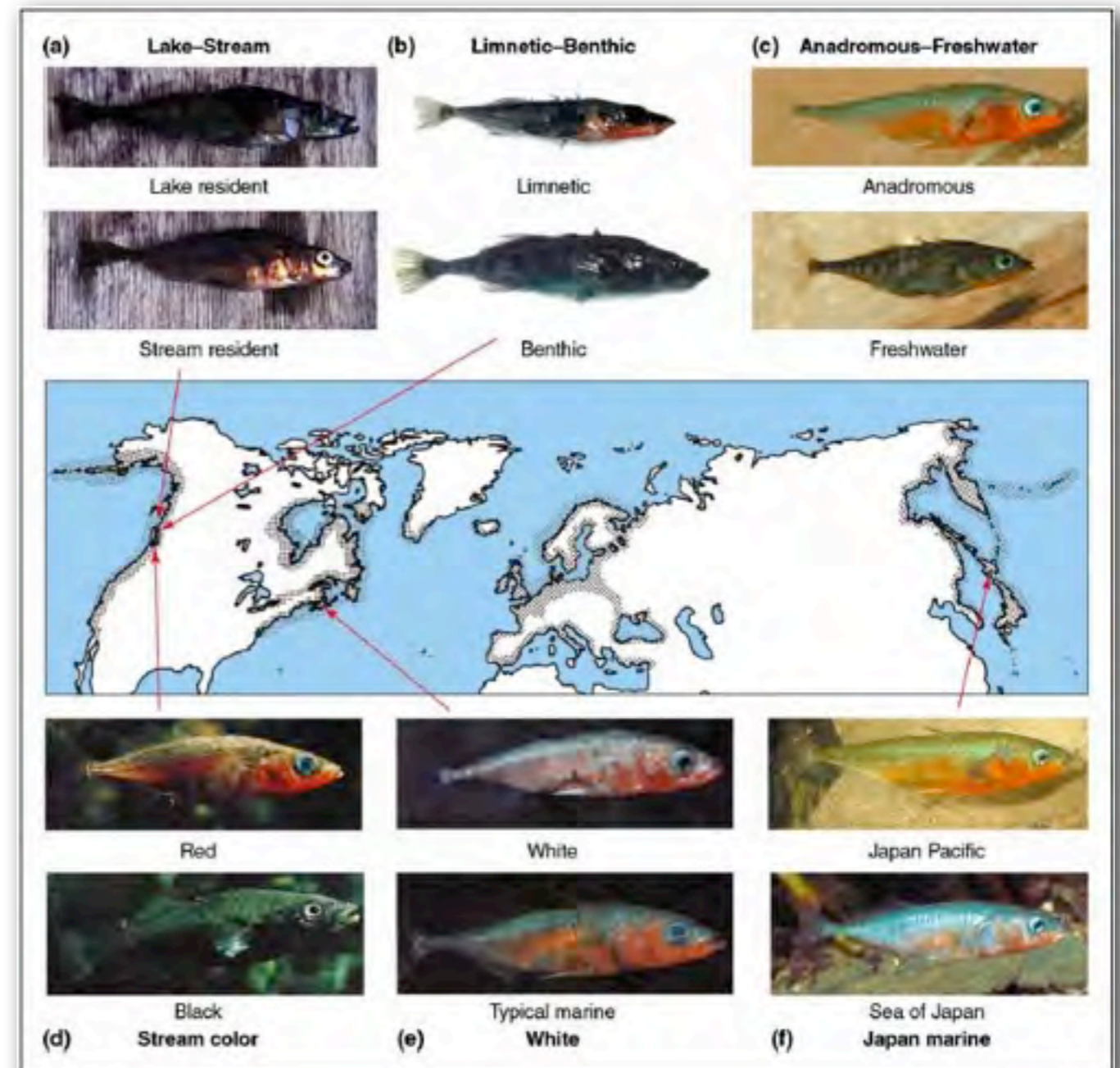
- *Derived Freshwater Populations*
Lake and stream
Young (<15,000 years)
Phenotypically diverse



Threespine stickleback, *Gasterosteus aculeatus*

- *Ancestral Oceanic Populations*
Marine and Anadromous
Old (> 10 million years)
Phenotypically similar

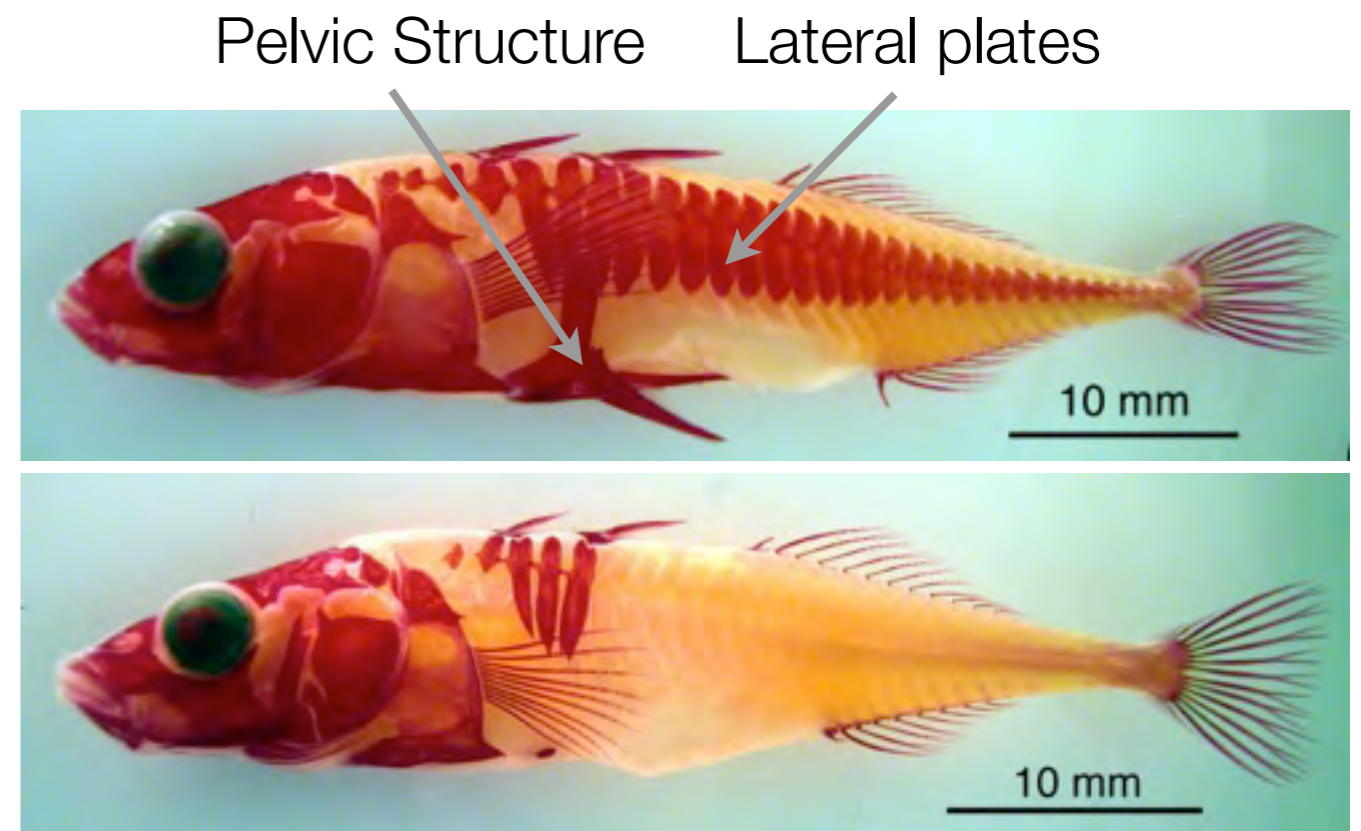
- *Derived Freshwater Populations*
Lake and stream
Young (<15,000 years)
Phenotypically diverse



Threespine stickleback, *Gasterosteus aculeatus*

- *Ancestral Oceanic Populations*
Marine and Anadromous
Old (> 10 million years)
Phenotypically similar

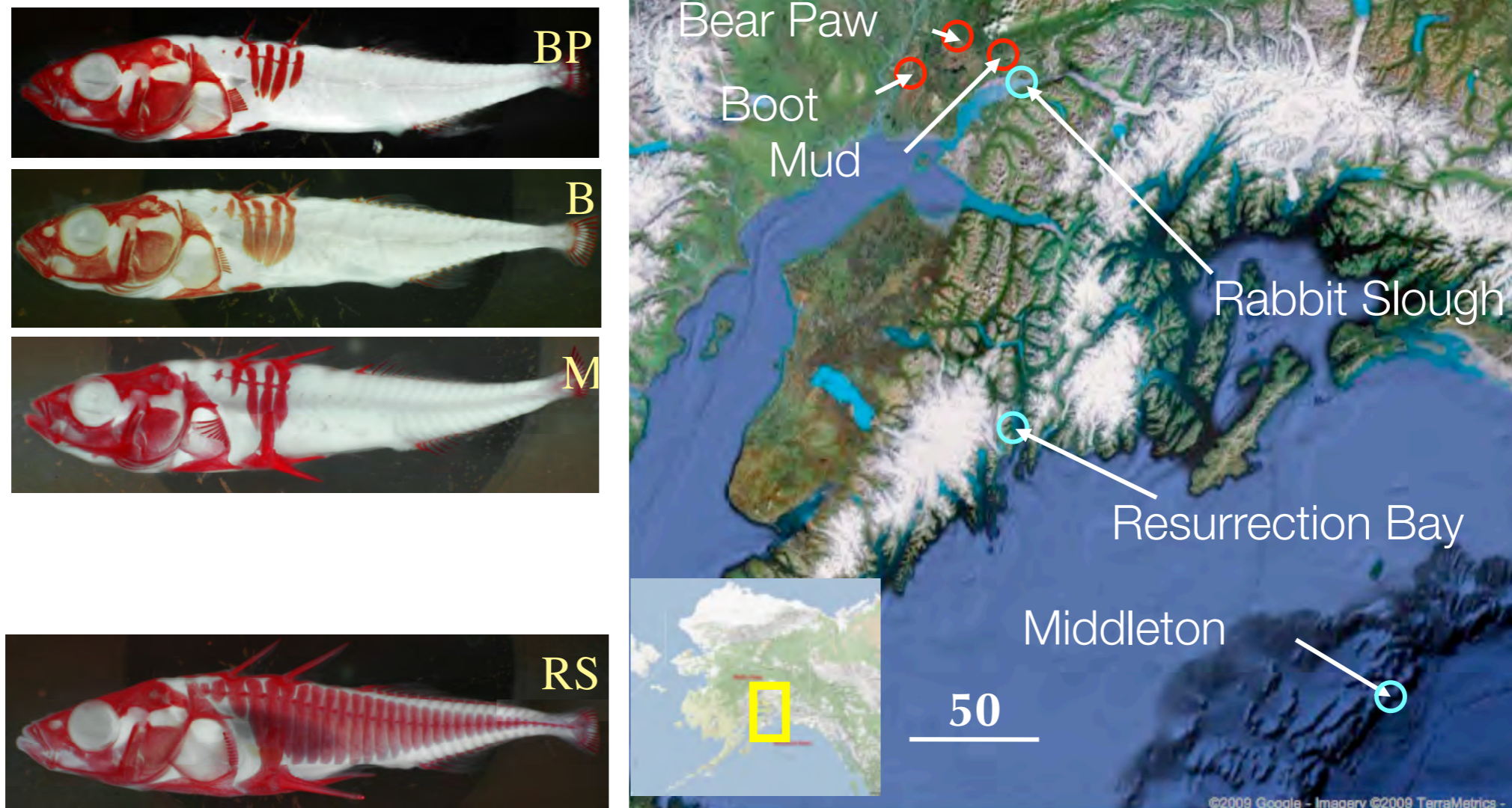
- *Derived Freshwater Populations*
Lake and stream
Young (<15,000 years)
Phenotypically diverse



QTL
mapping

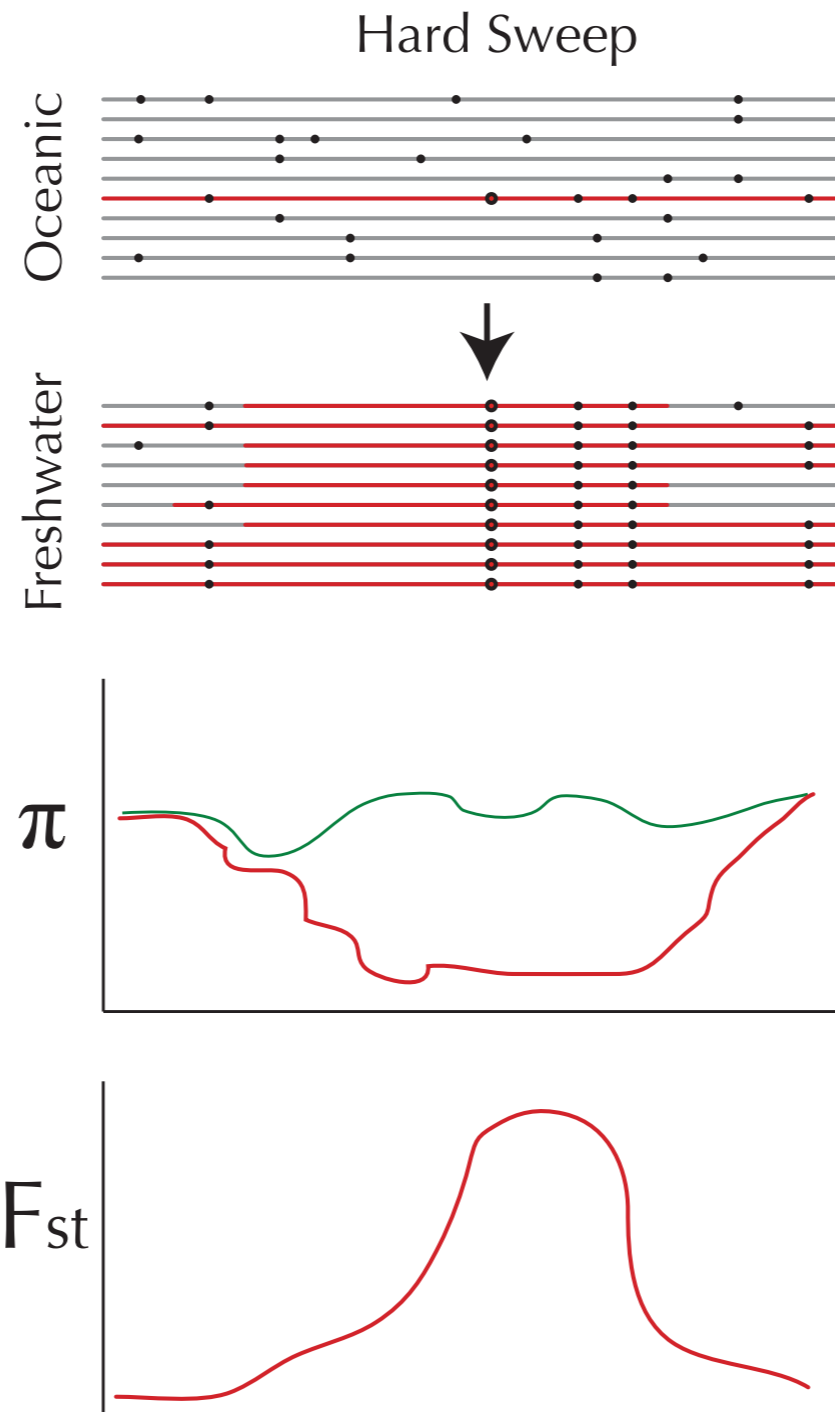
Cresko et al. 2004. PNAS
Colosimo et al. 2005. Science
Shapiro et al. 2004. Nature
Albert et al. 2008. Evolution
Miller et al. 2007. Cell
Chan et al. 2010. Nature

Signatures of natural selection across the genome

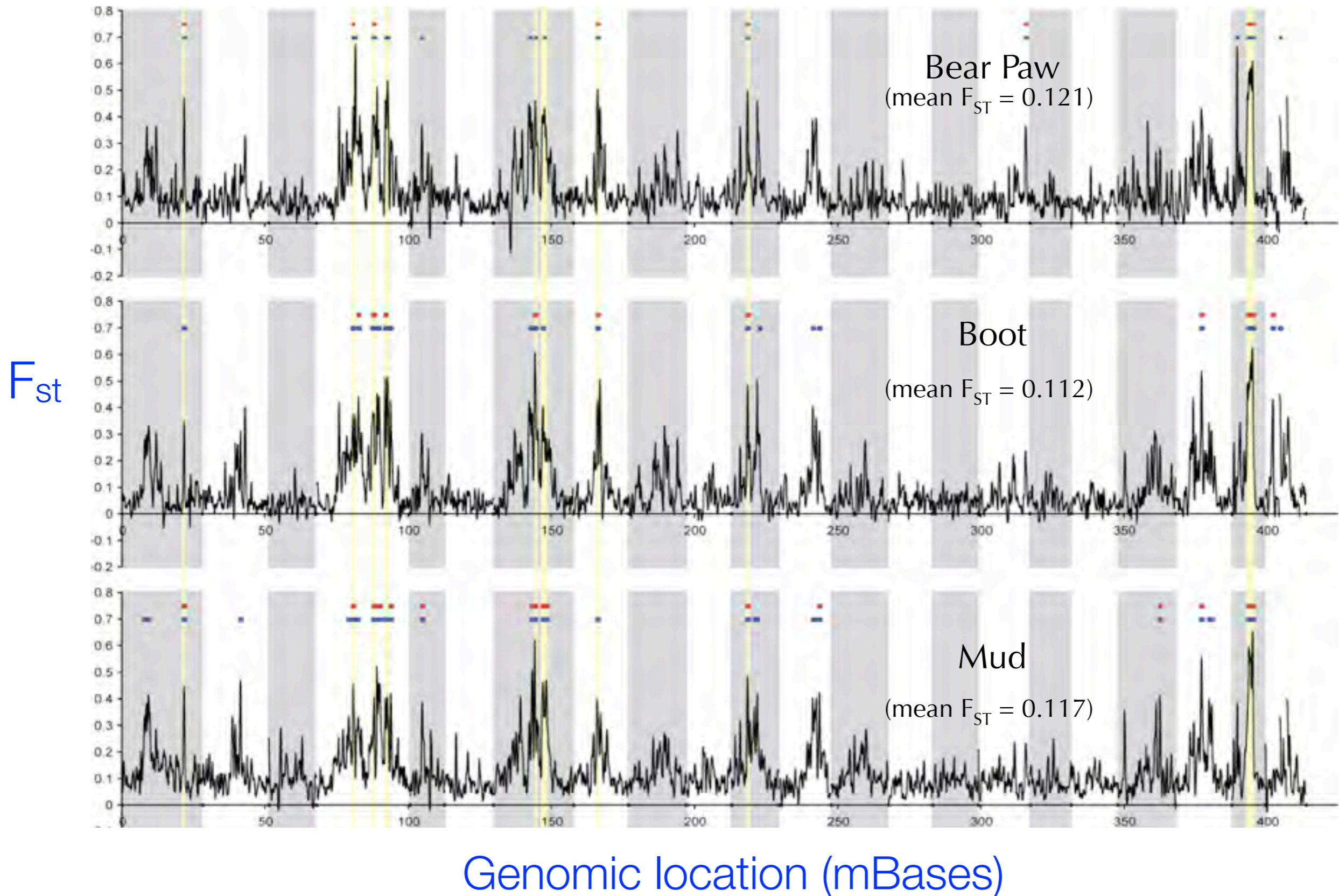


20 individuals in each of 5 popn's
2 Ocean & 3 Freshwater
45,000 SNPs in each individual

What genomic regions are subject to selection during parallel evolution?

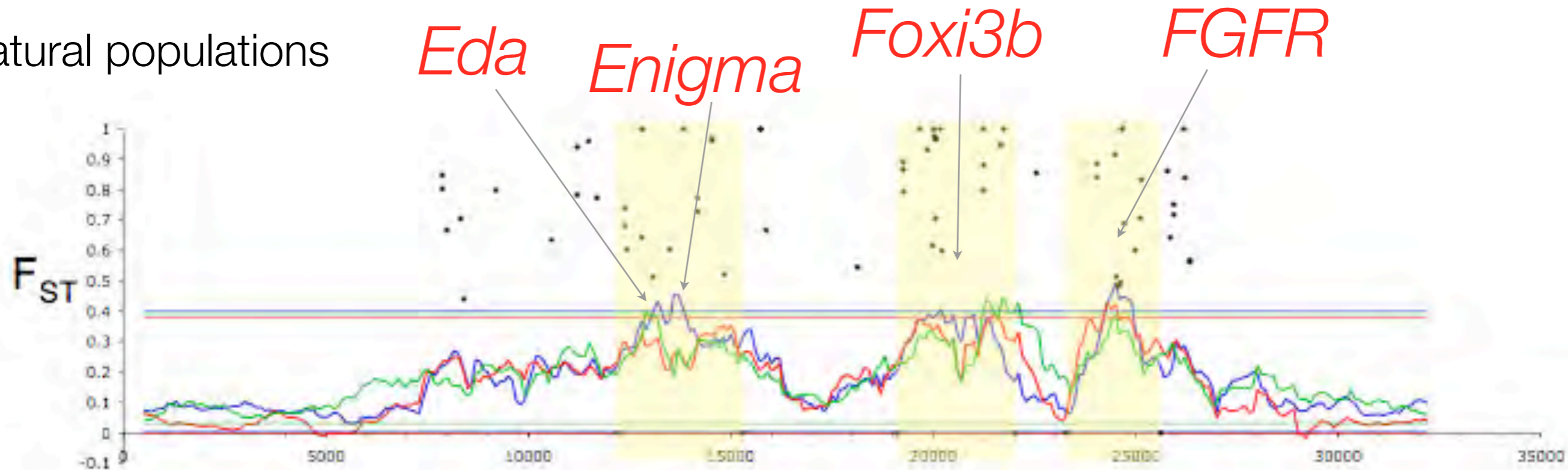


Parallel signatures of selection across the genomes



Previously identify quantitative trait loci (QTLs) are under selection

Natural populations

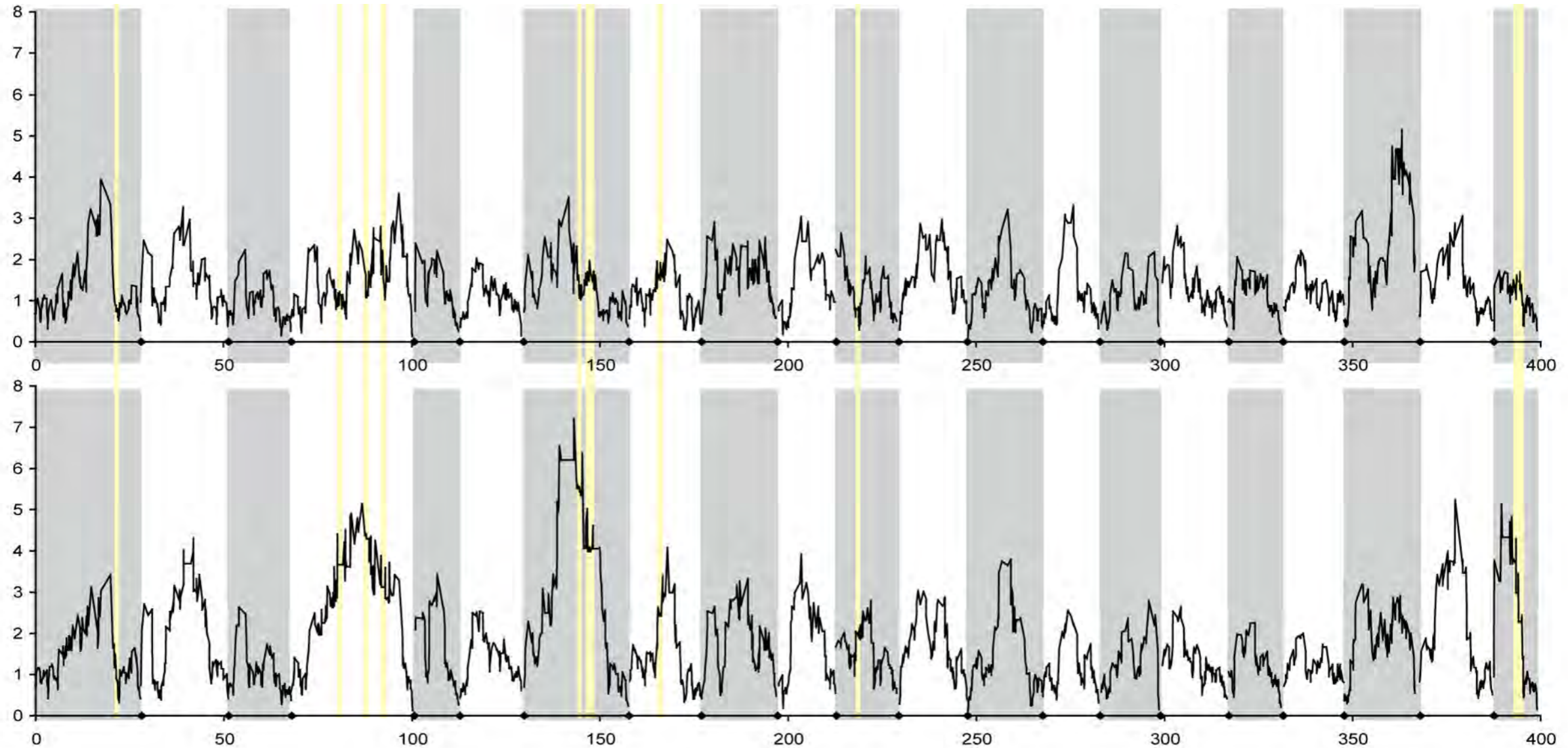


Lateral plate major locus
on LGIV (4000 SNPs)



Extensive LD across the genome

Freshwater



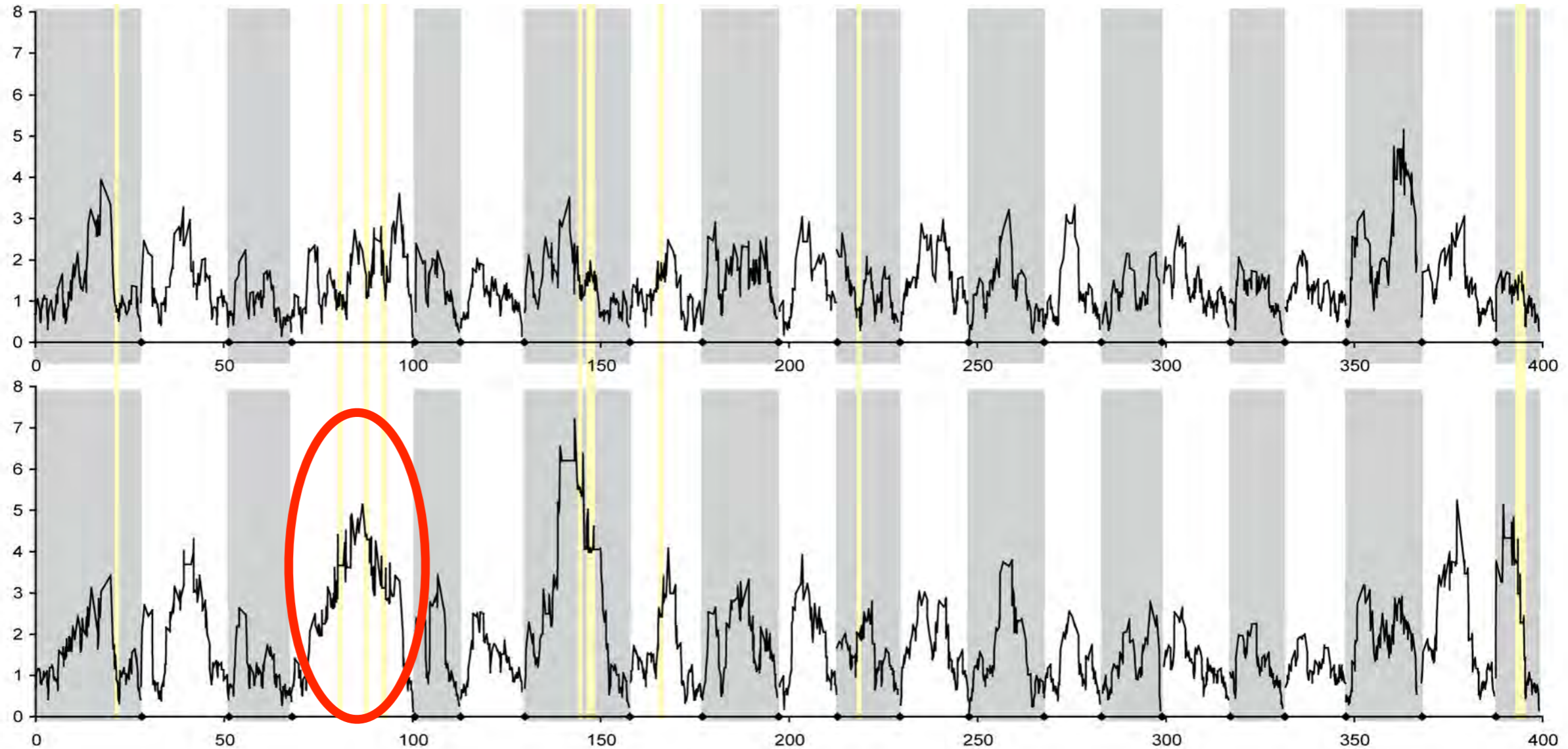
Ocean

Position (Mb)

Extensive LD across the genome

More in oceanic than in freshwater populations

Freshwater



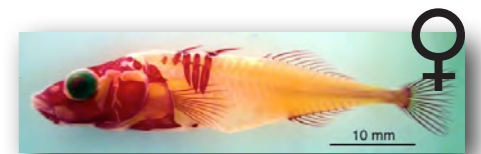
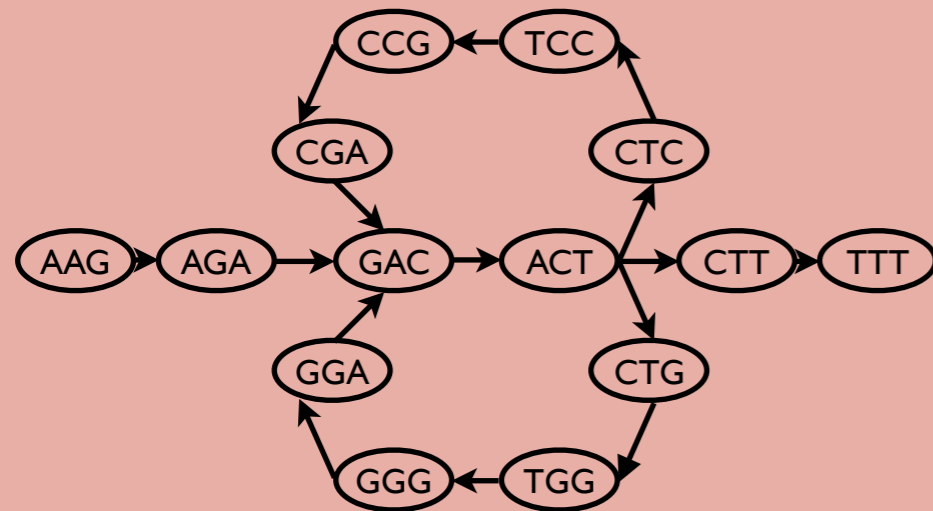
Ocean

Position (Mb)

Could genome rearrangements in the stickleback genome be affecting these patterns?

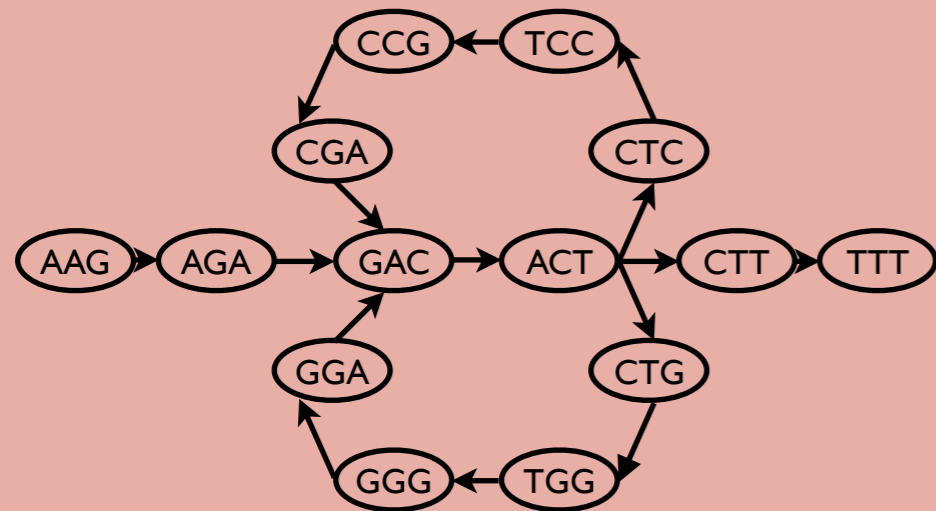
Julian Catchen, Susie Bassham and Kate Ituarte

Genome Assembly

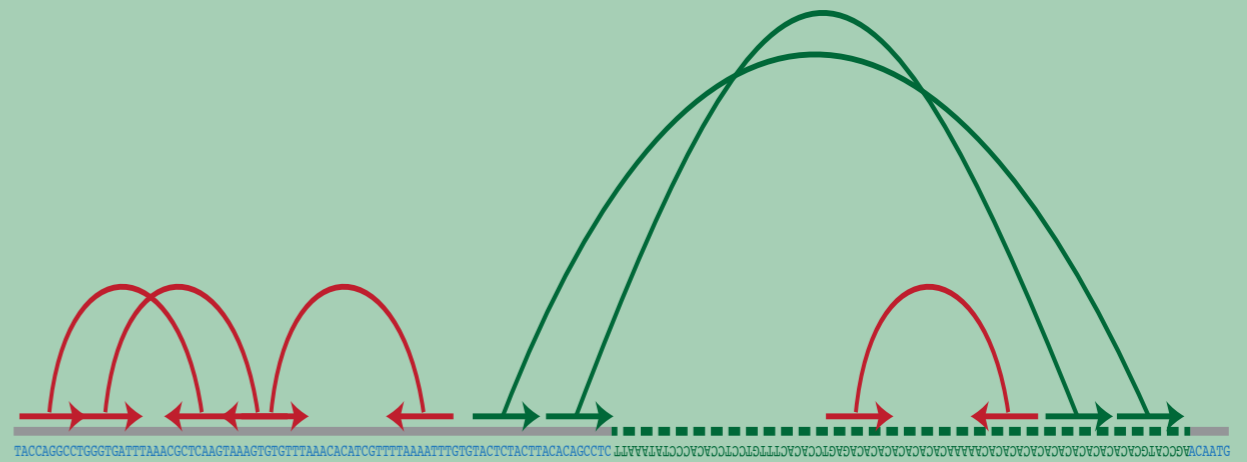


N50	17,417 bp	18,982 bp	15,555 bp	15,534 bp
Max	199,905 bp	192,283 bp	238,768 bp	254,734 bp
Total	488.8 Mb	472.5 Mb	456.4 Mb	473.4 Mb
Median Coverage	24.6x	26.5x	24.1x	25.8x

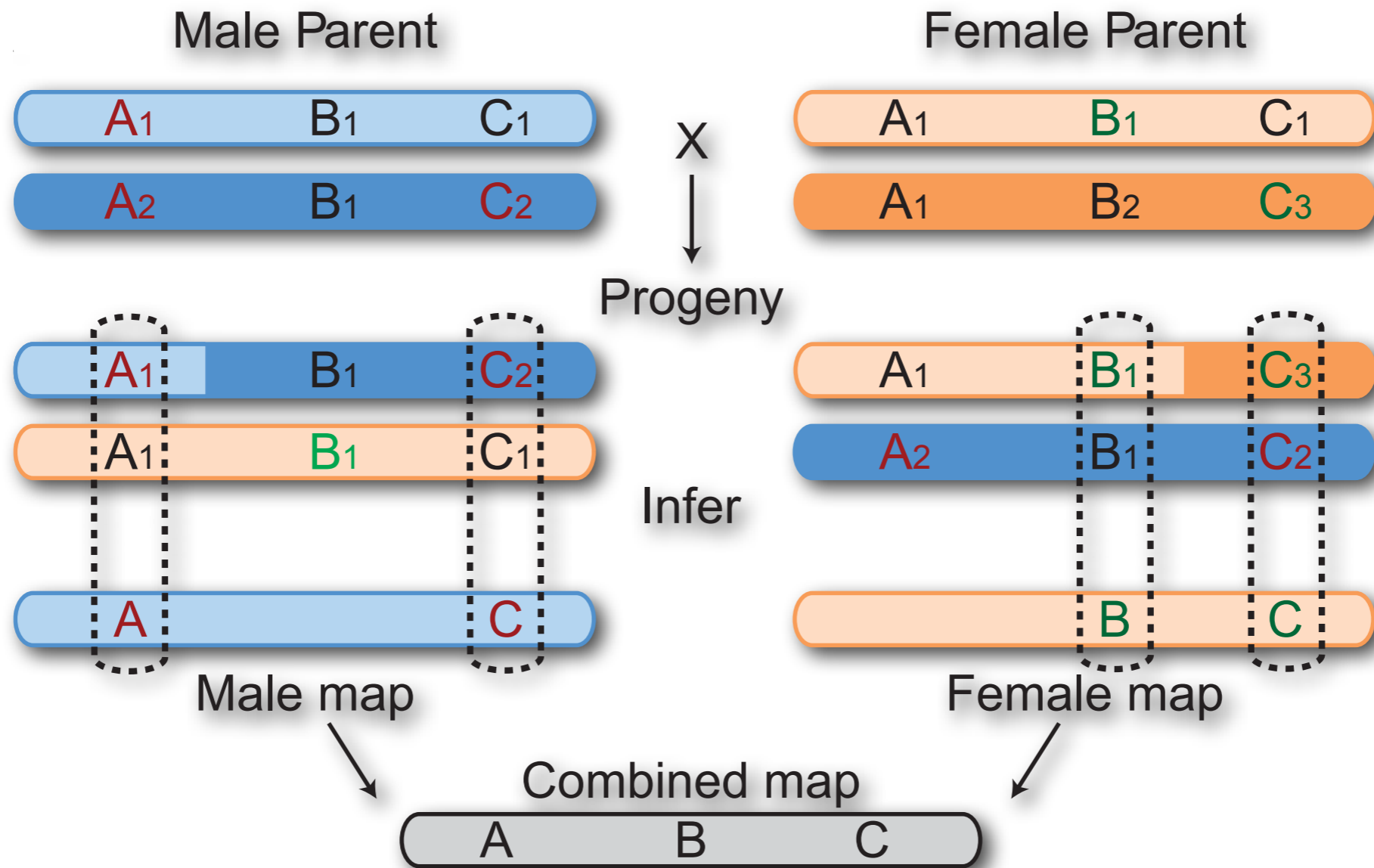
Genome Assembly



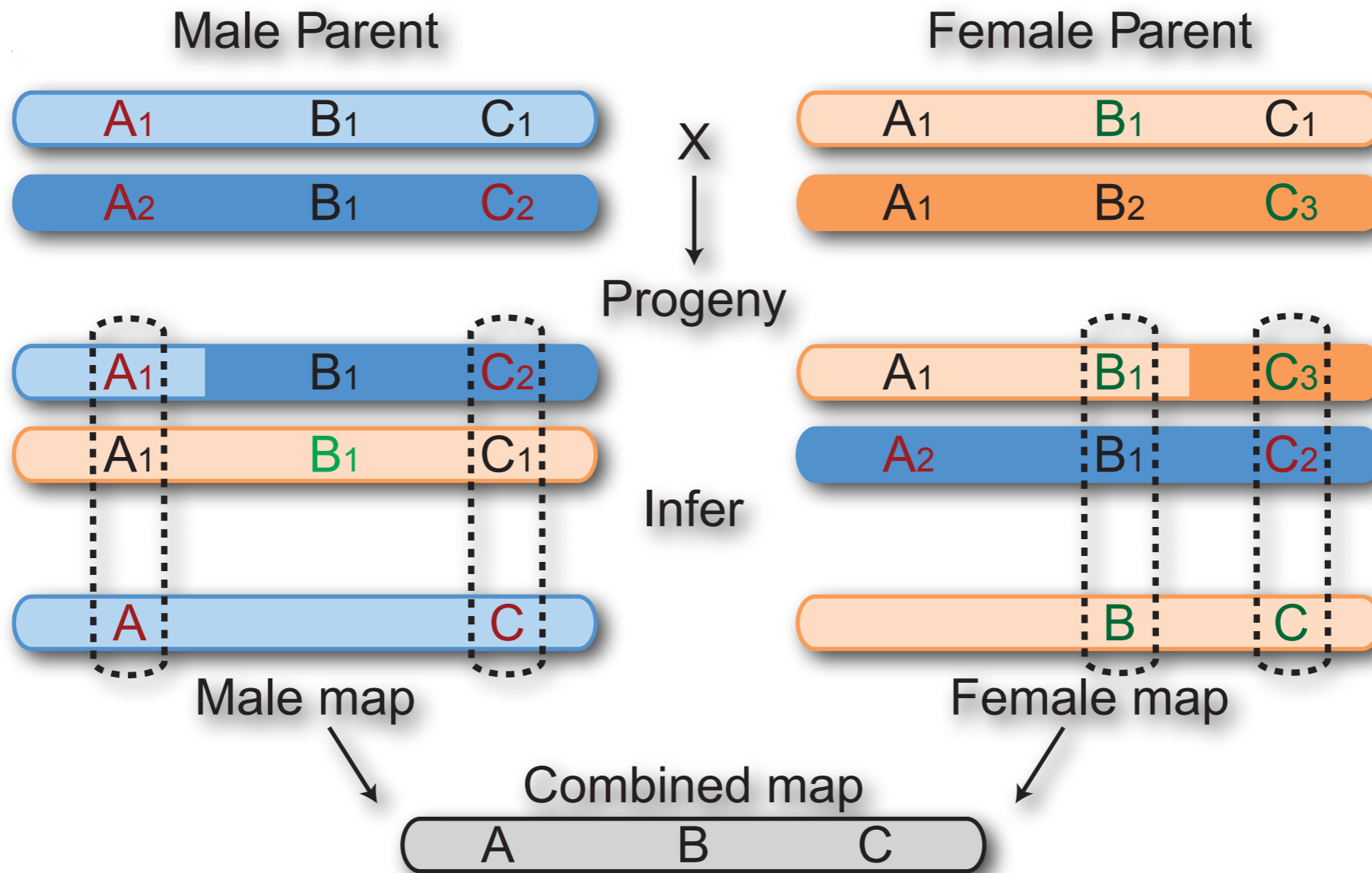
Paired-end Alignments



F1 Pseudo-testcross



F1 Pseudo-testcross



93 progeny
66,071 loci
5,351 markers

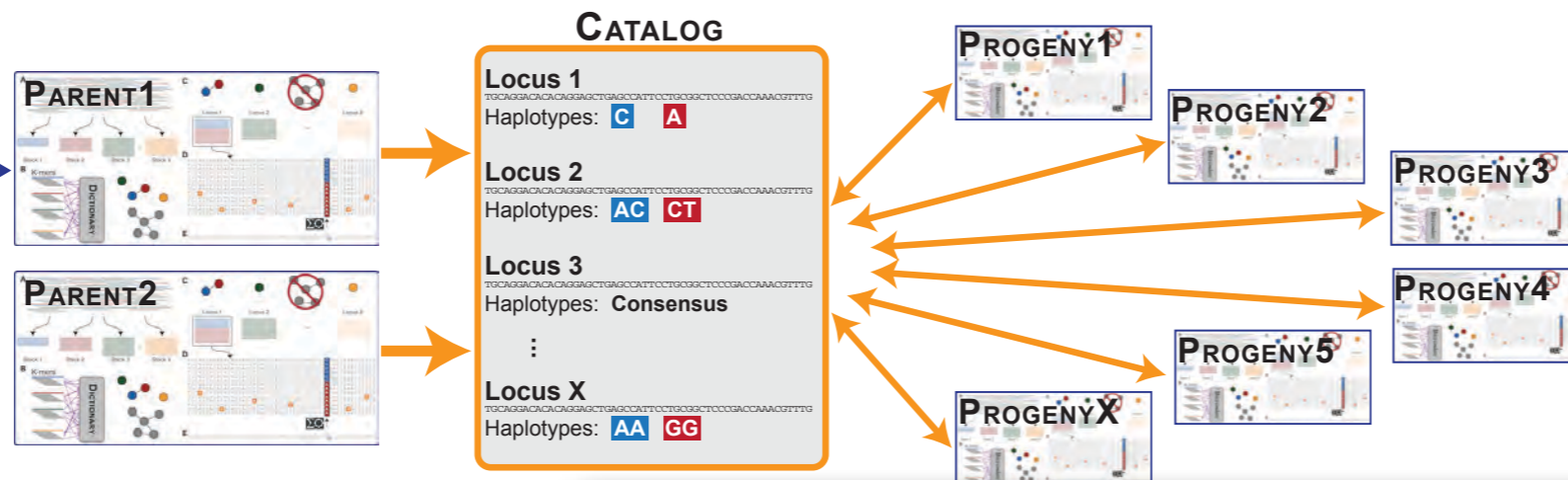
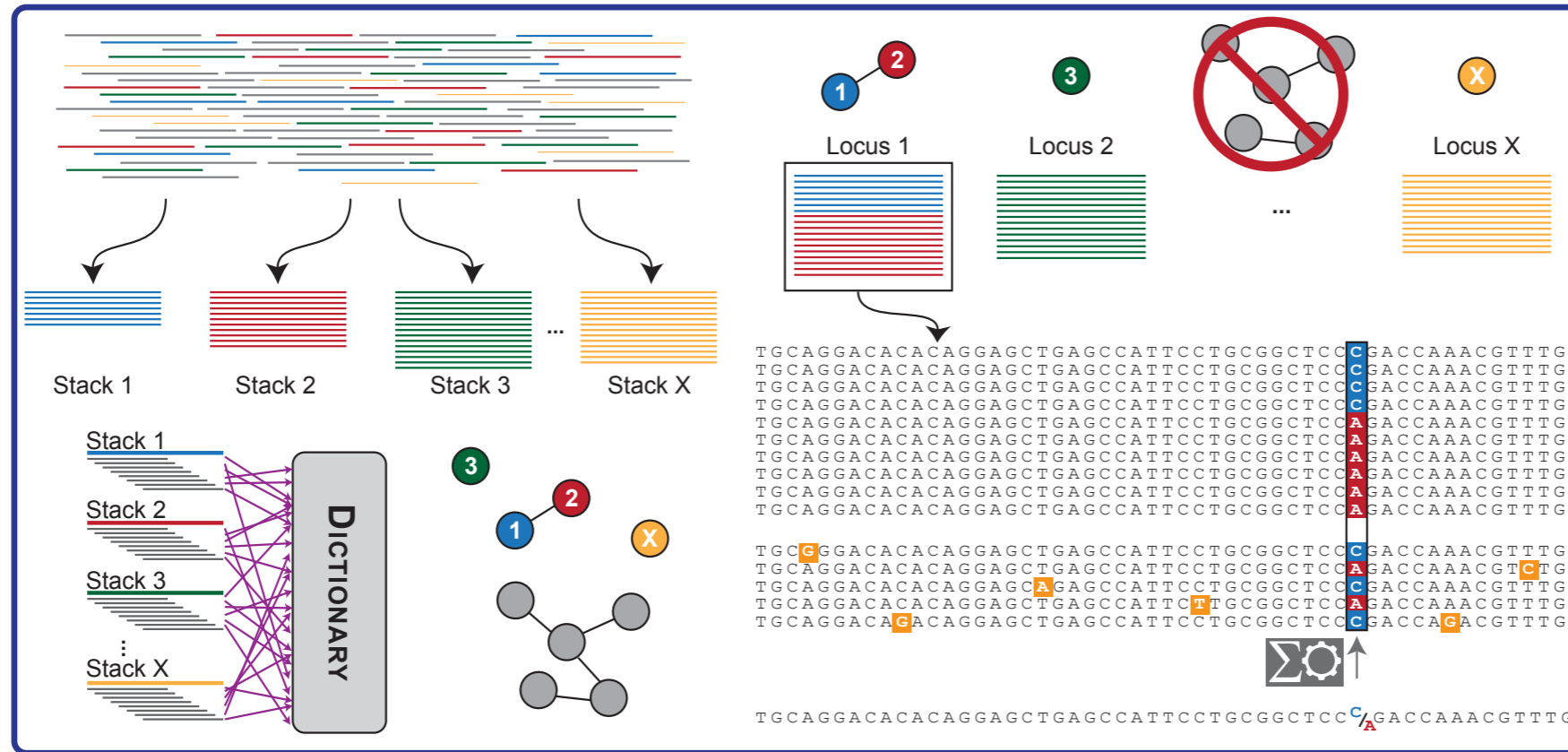
93 progeny
45,301 loci
3,927 markers

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

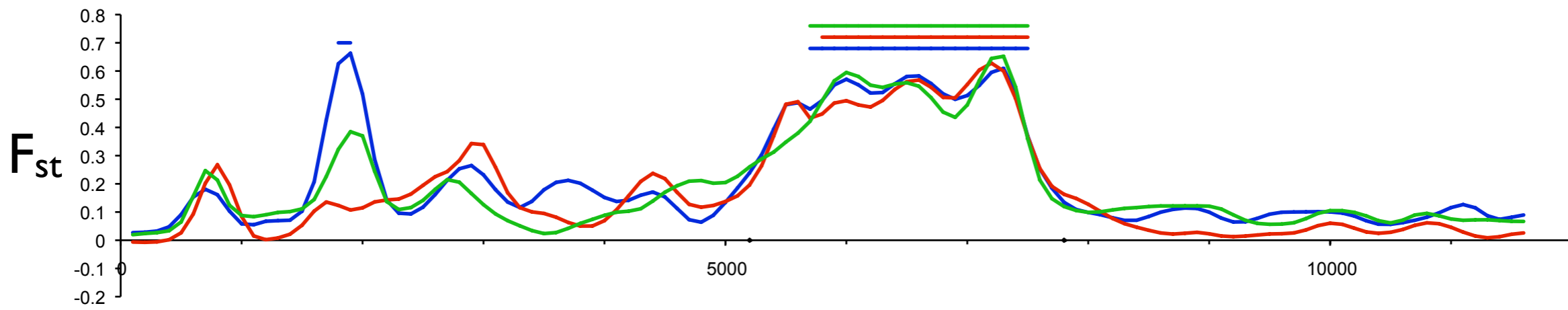
Stacks



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

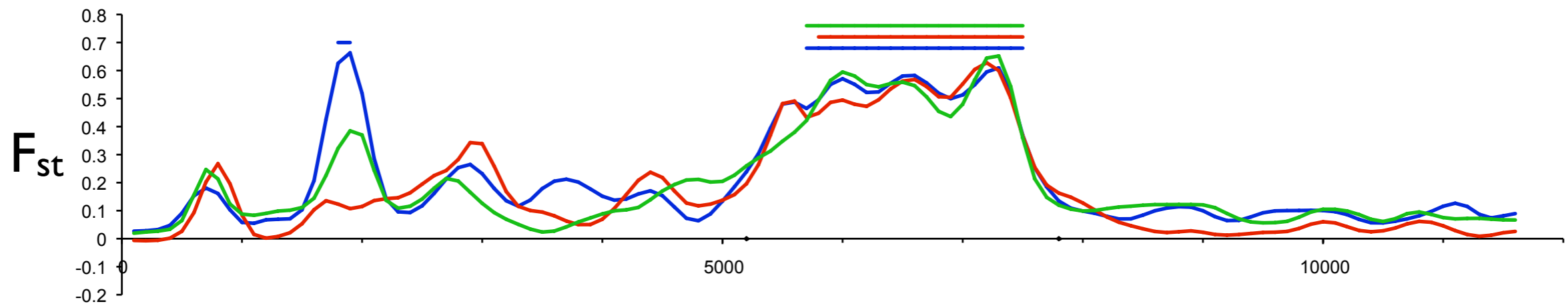
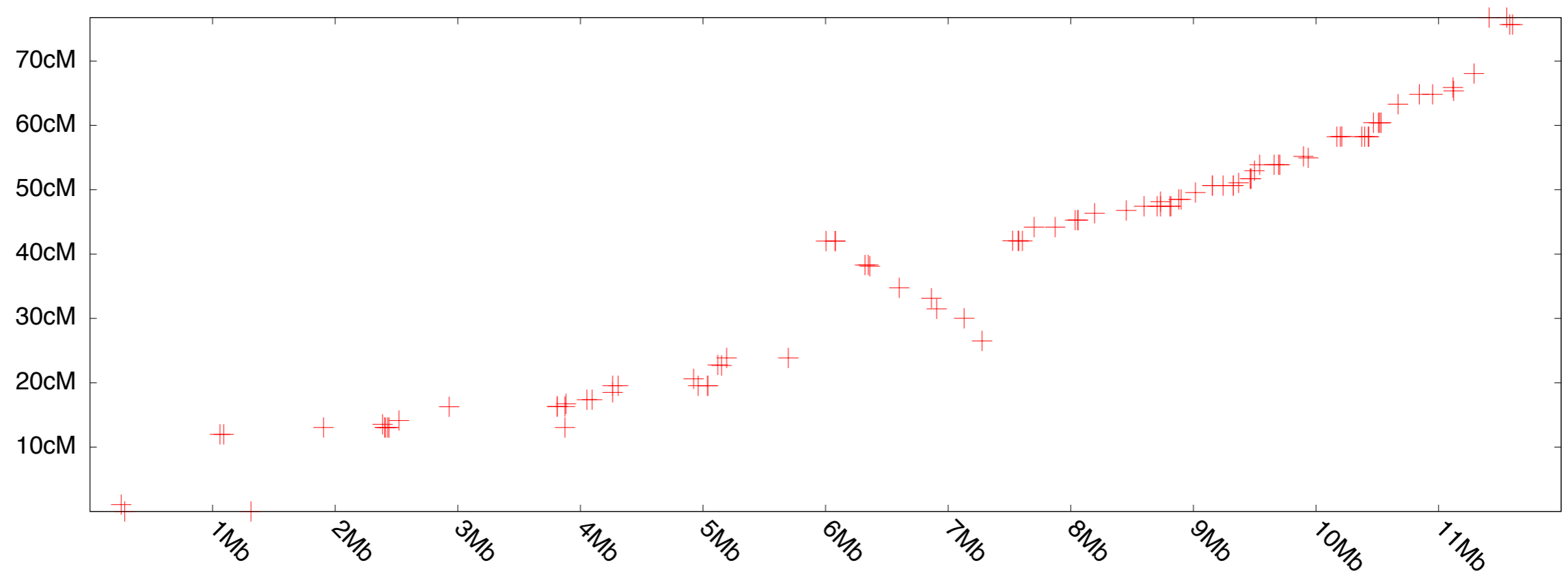
Julian M. Catchen,* Angel Amores,[†] Paul Hohenlohe,* William Cresko,* and John H. Postlethwait^{†,1}
 *Center for Ecology and Evolutionary Biology and [†]Institute of Neuroscience, University of Oregon, Eugene, Oregon 97403

LG2I



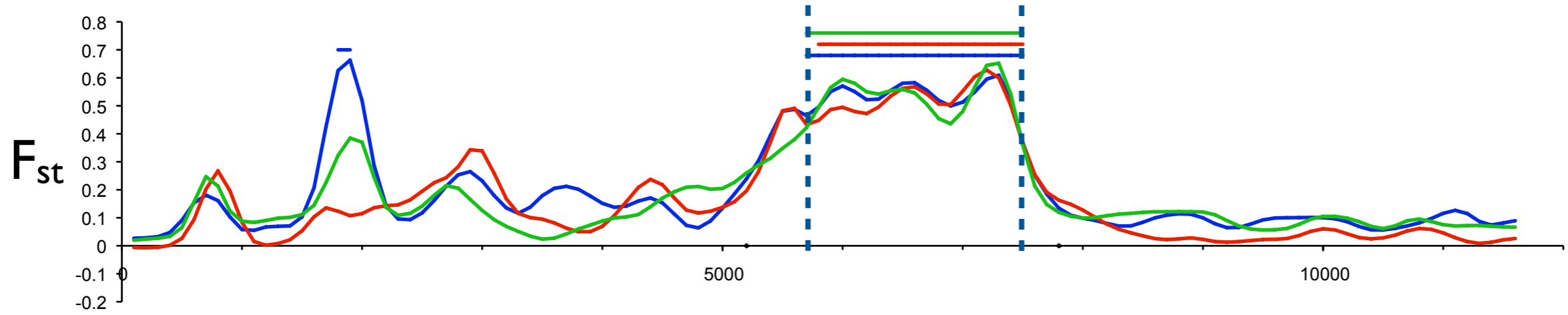
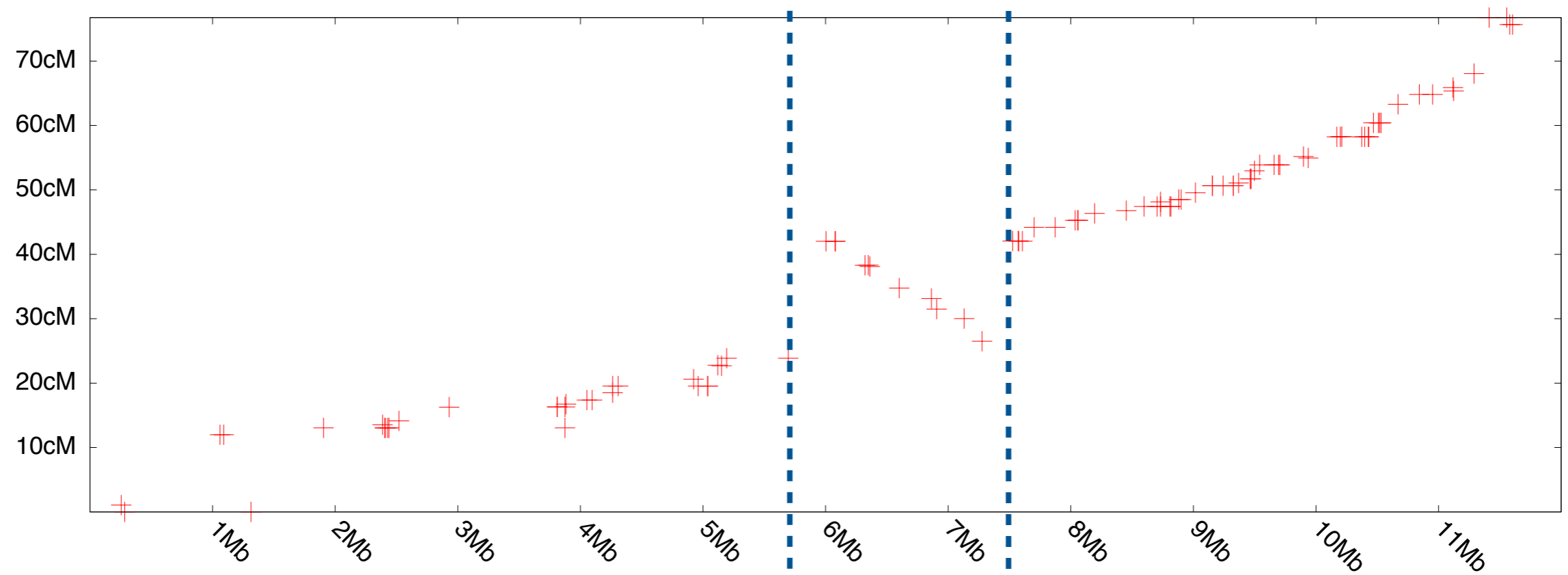


LG2I





LG2I

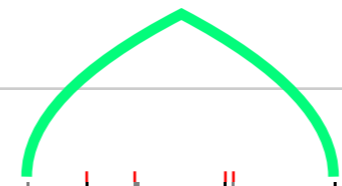




Aligned
Opposite
Inward
5.00Mb

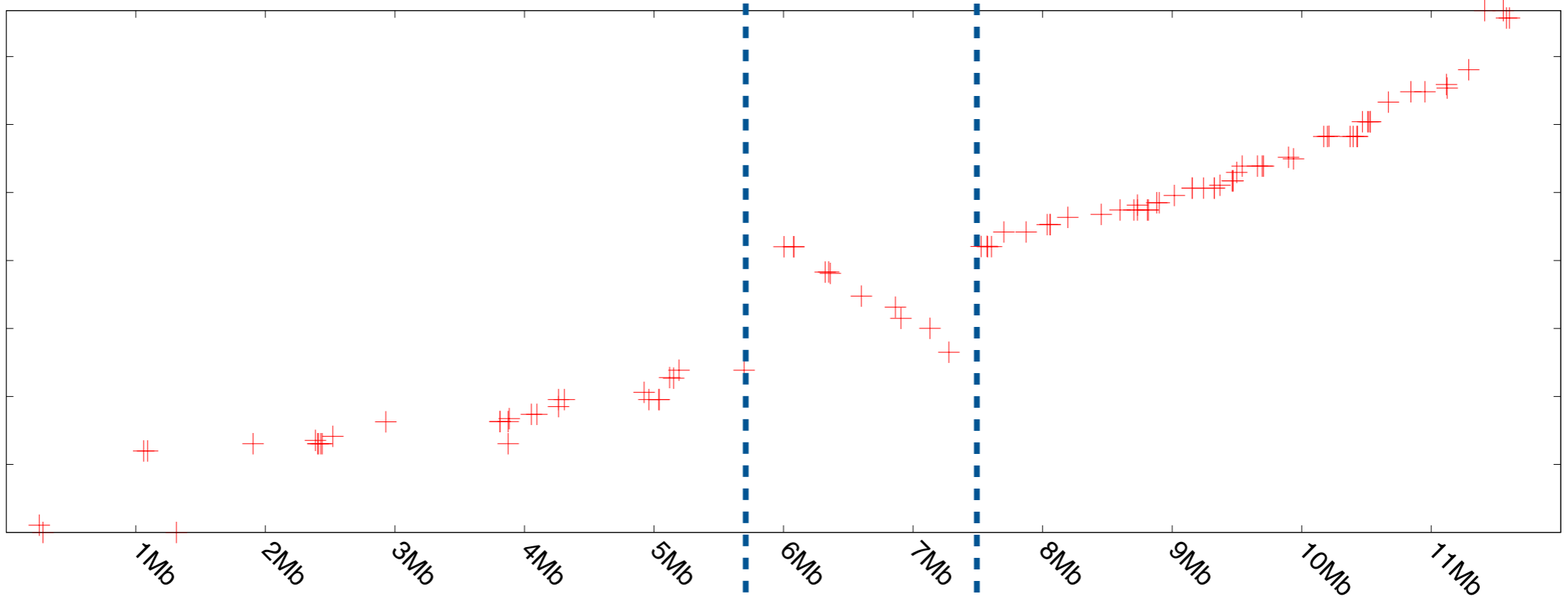


1.00Mb



LG2I

70cM
60cM
50cM
40cM
30cM
20cM
10cM

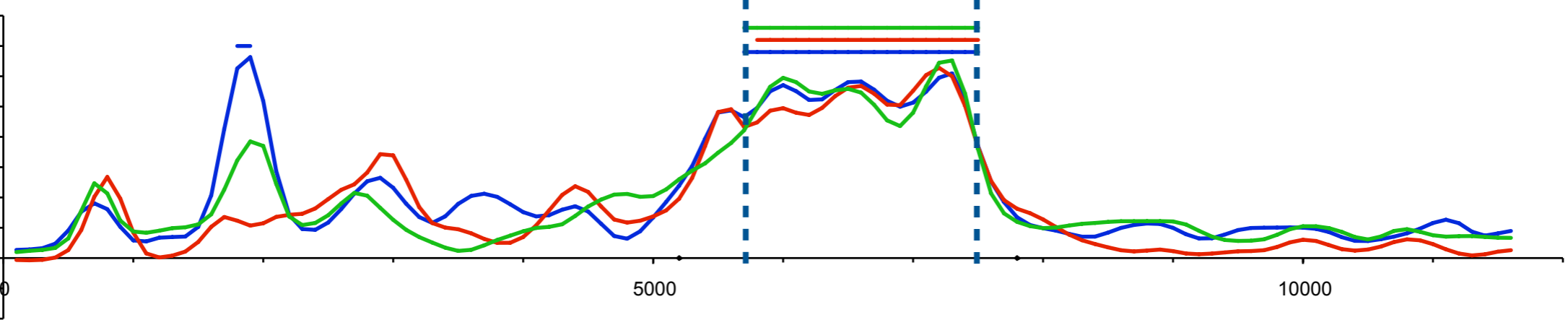


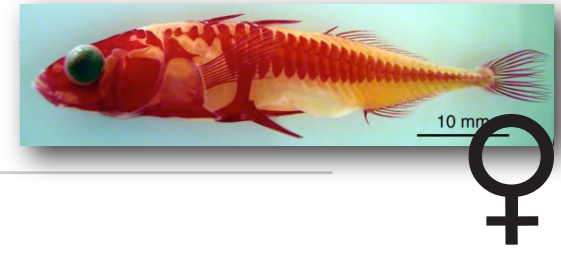
F_{st}

0.8
0.7
0.6
0.5
0.4
0.3
0.2
0.1
0
-0.1
-0.2

5000

10000





Aligned
Opposite
Inward
5.00Mb

1.00Mb

70cM

60cM

50cM

40cM

30cM

20cM

10cM

1Mb

2Mb

3Mb

4Mb

5Mb

6Mb

7Mb

8Mb

9Mb

10Mb

11Mb

F_{st}

0.8

0.7

0.6

0.5

0.4

0.3

0.2

0.1

0

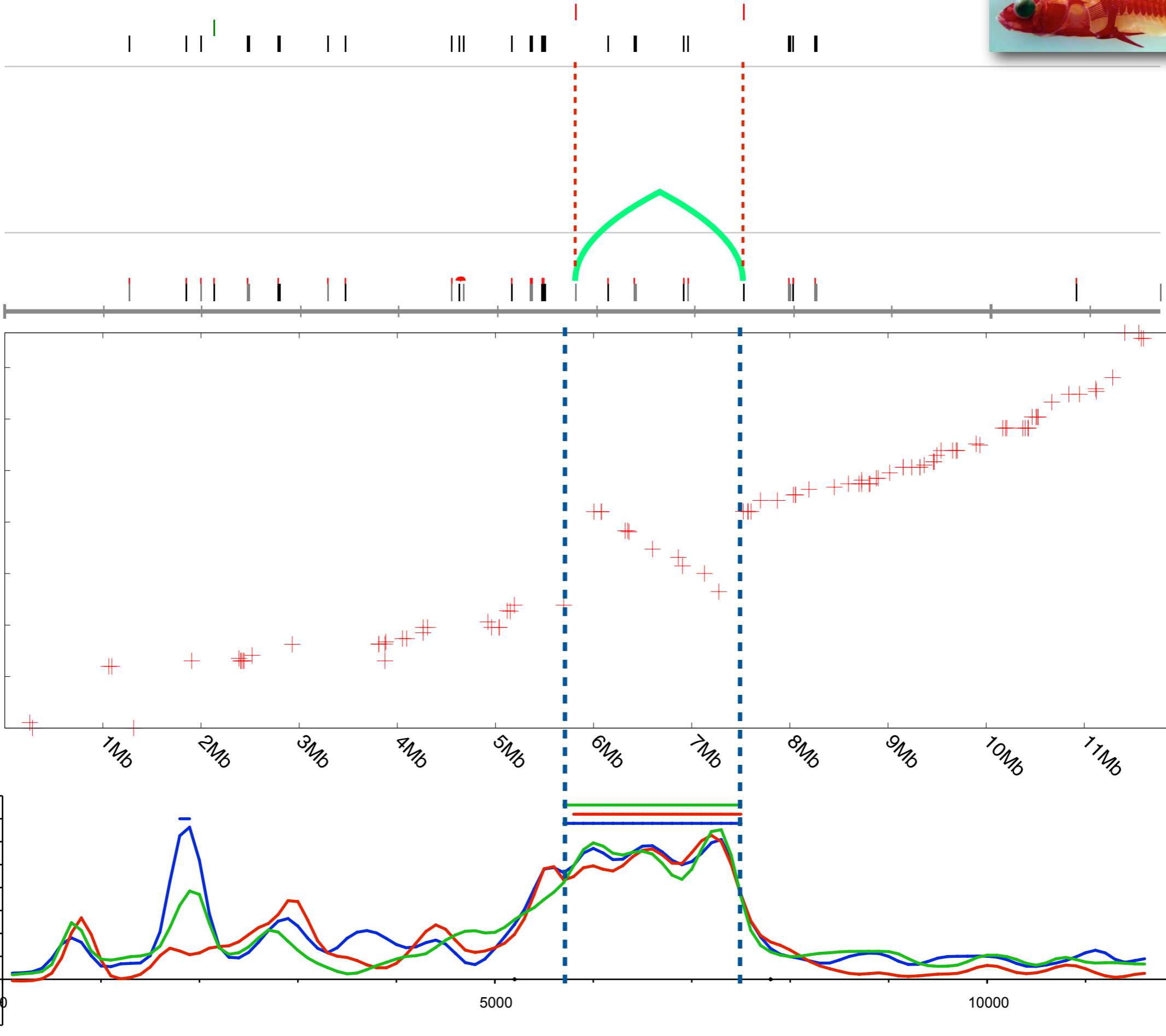
-0.1

-0.2

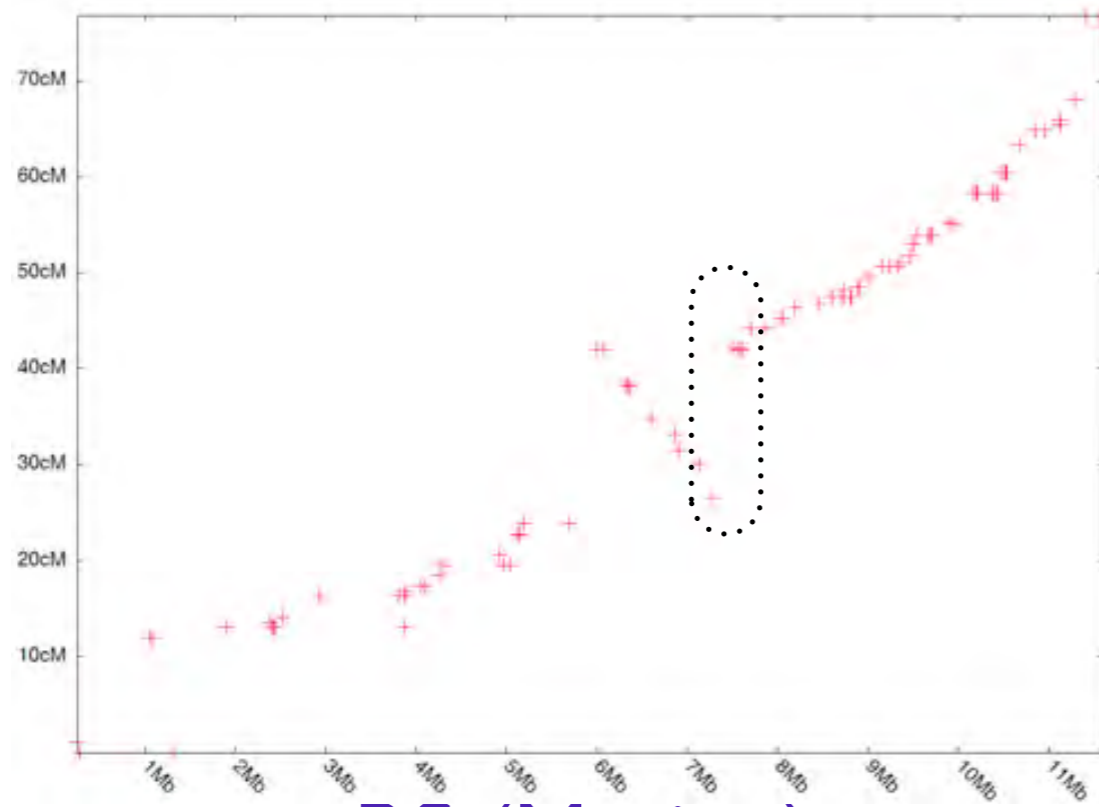
5000

10000

LG2I



Linkage Group XXI

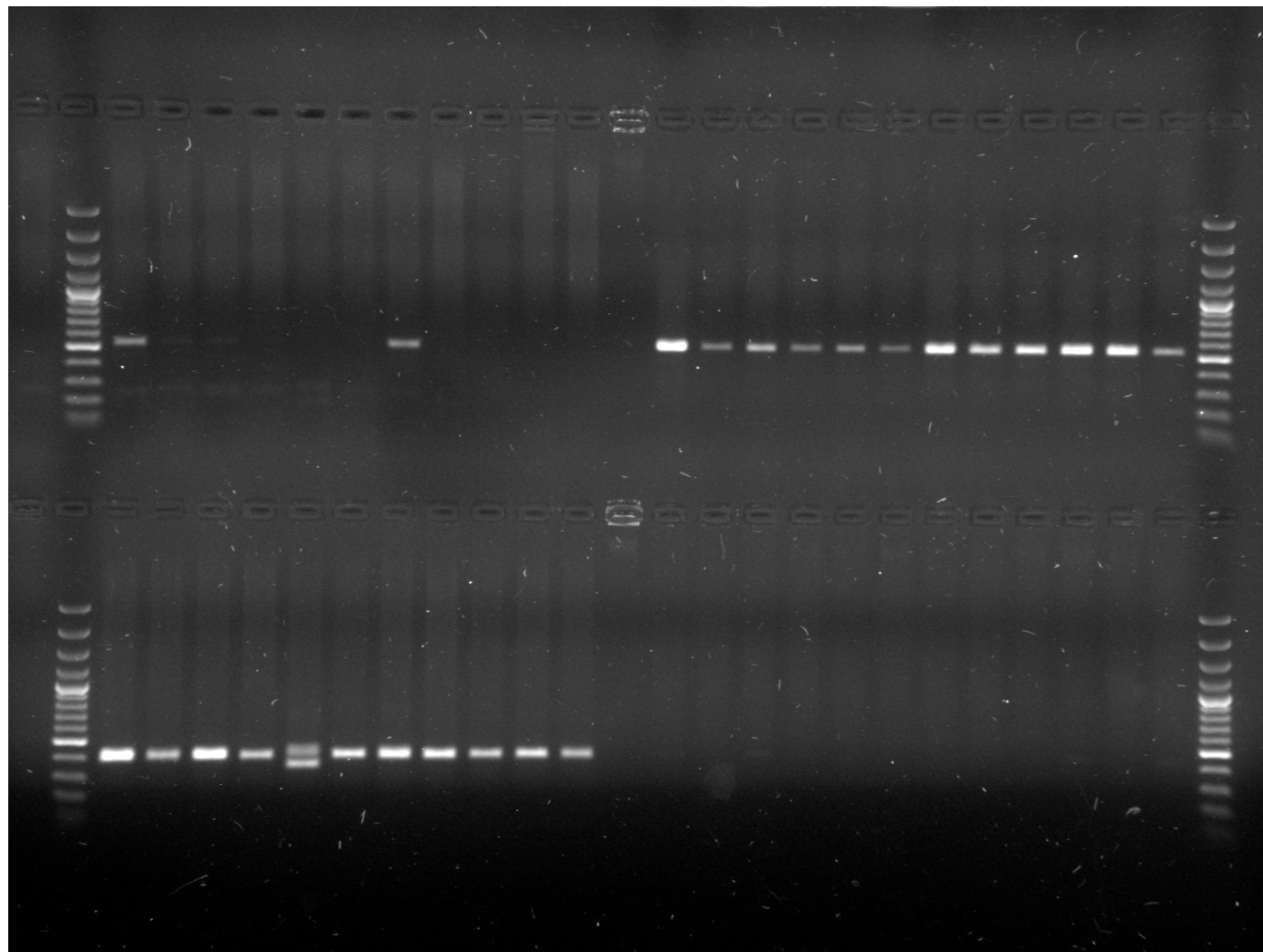


RS (Marine)

Boot

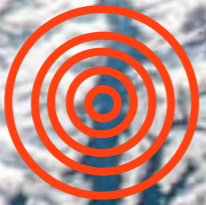
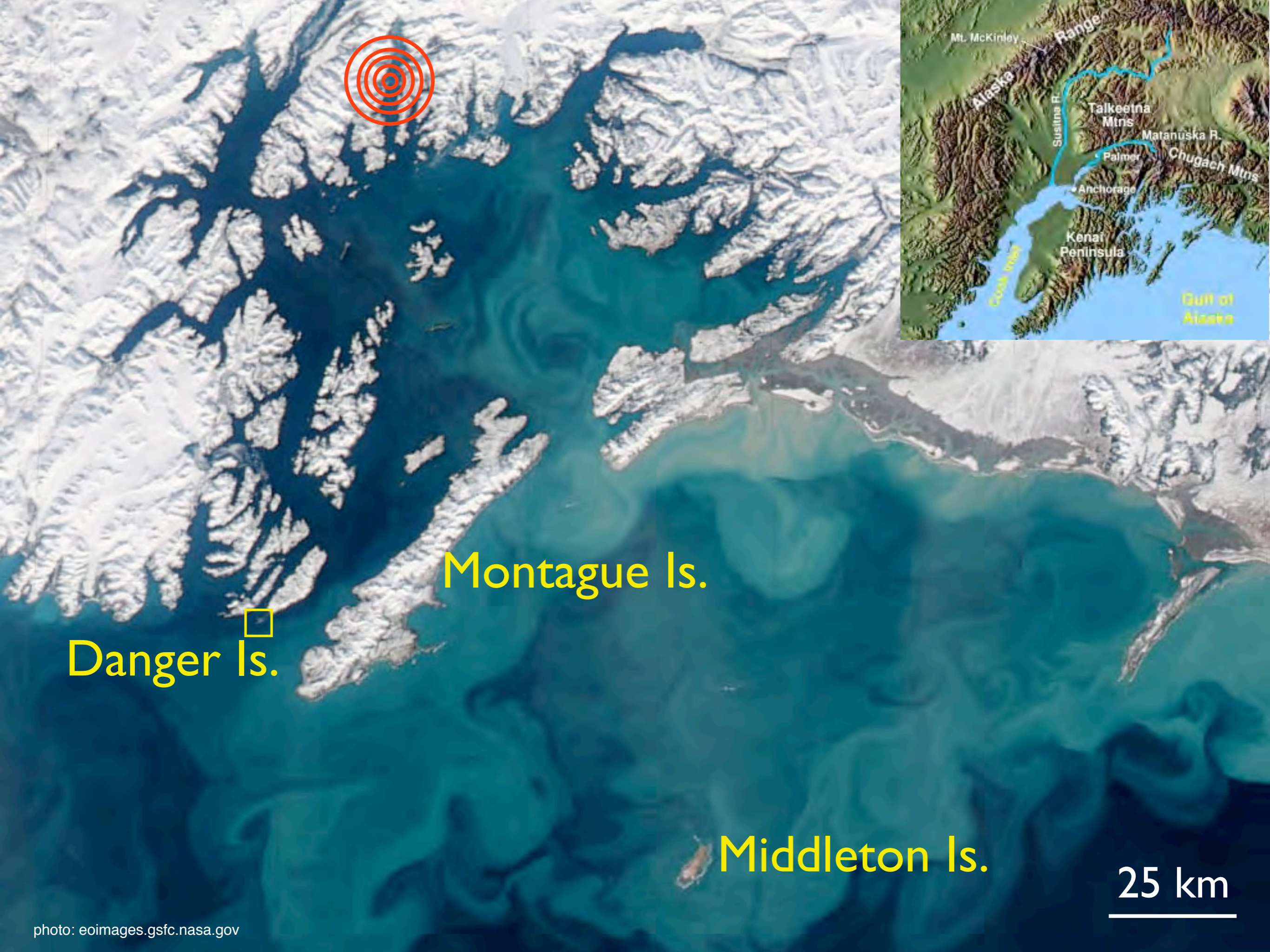
Like
Bear Paw

Inverted



How quickly does stickleback evolution occur?



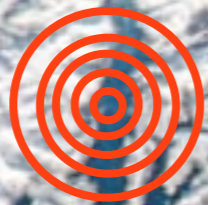


Montague Is.

Middleton Is.

Danger Is.

25 km



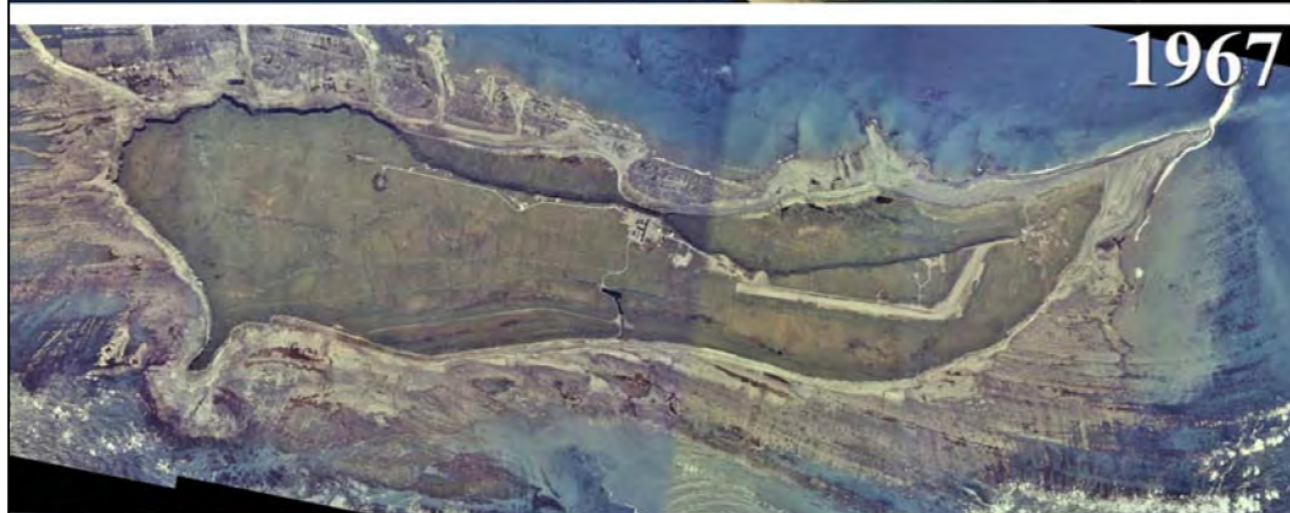
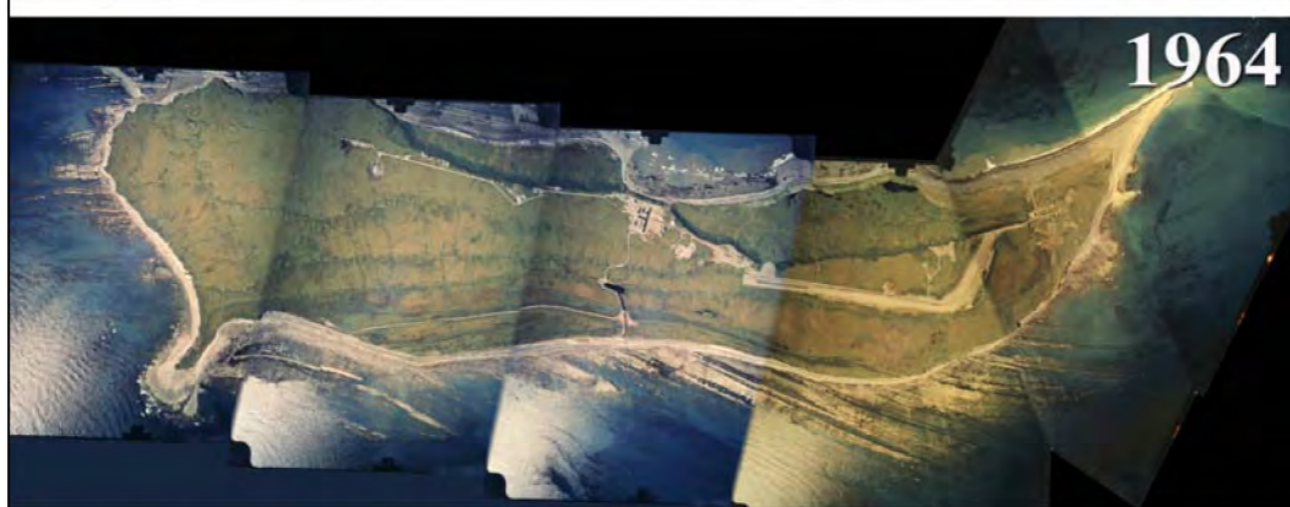
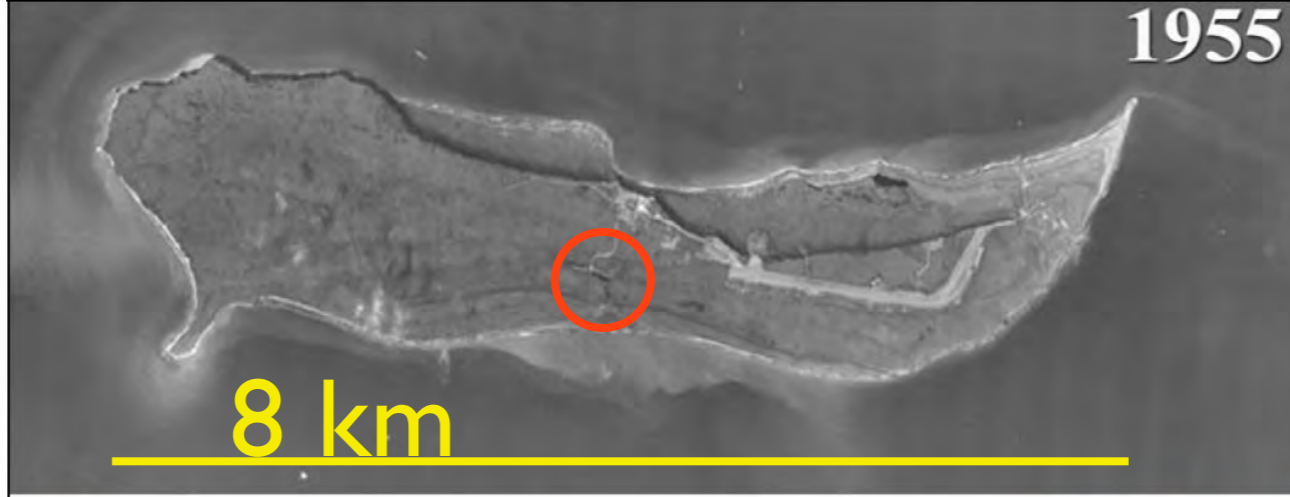
Montague Is.

Danger Is.

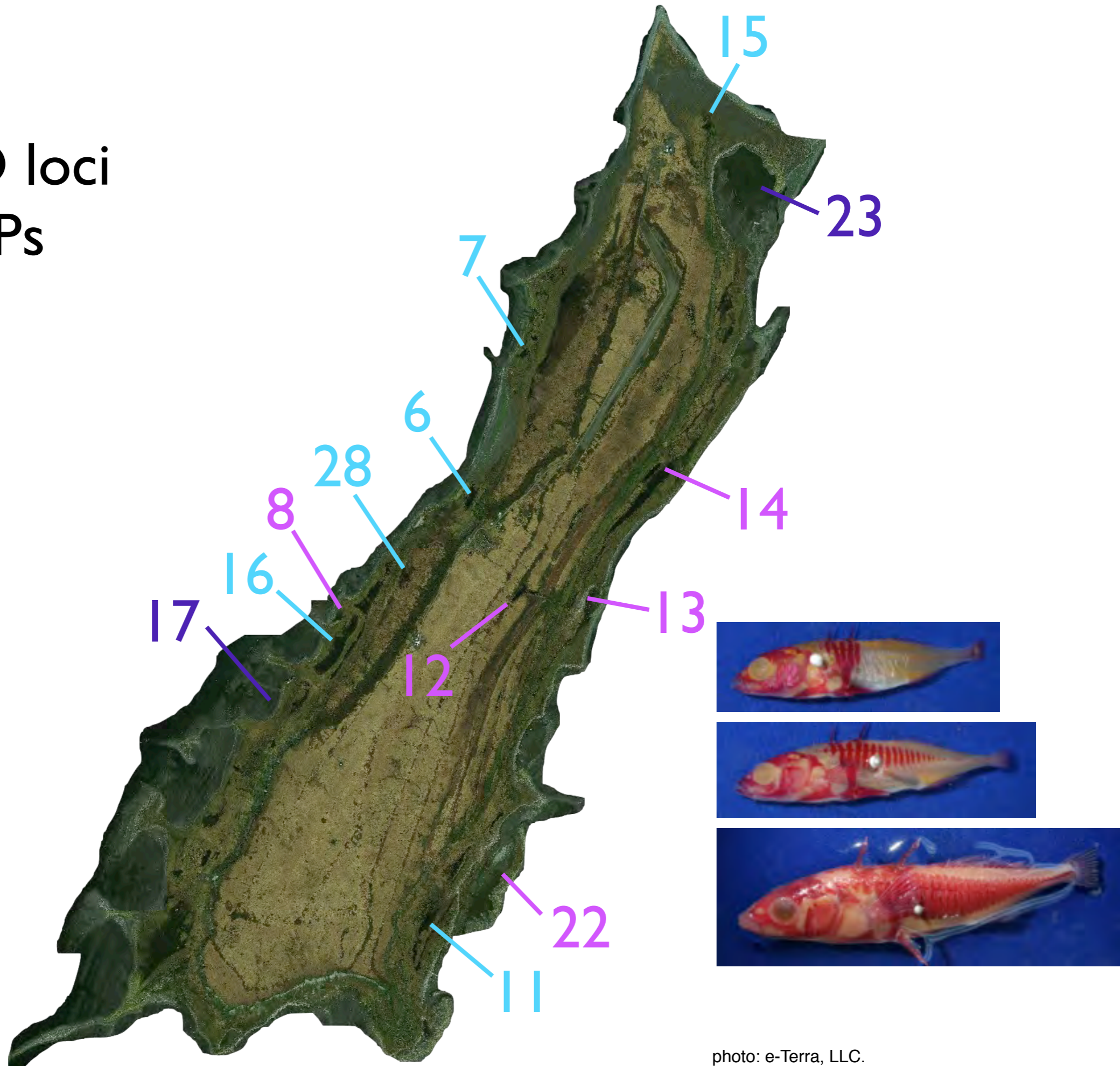
Middleton Is.

25 km

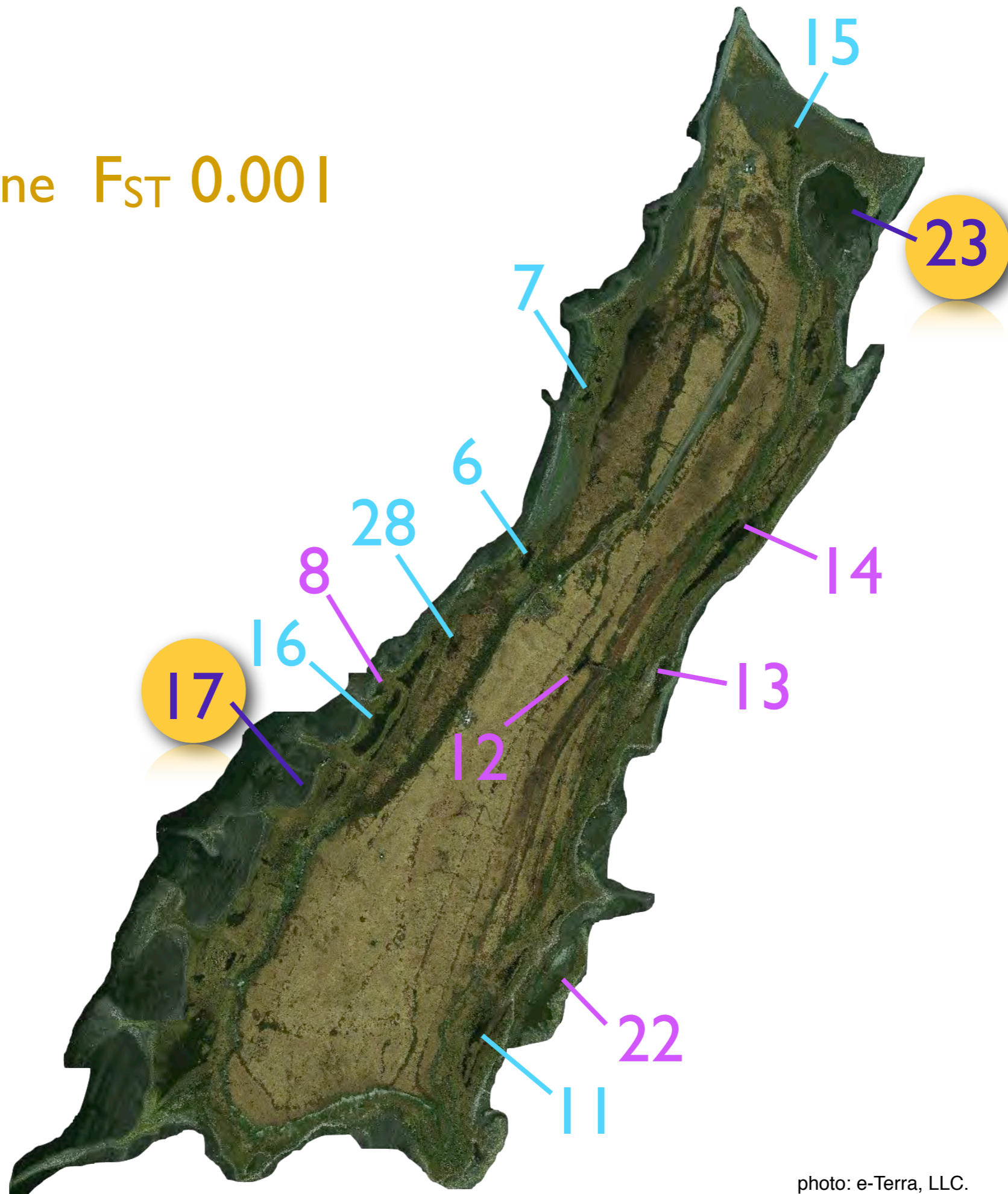




743 fish
27,878 RAD loci
110,000 SNPs



Marine x Marine F_{ST} 0.001

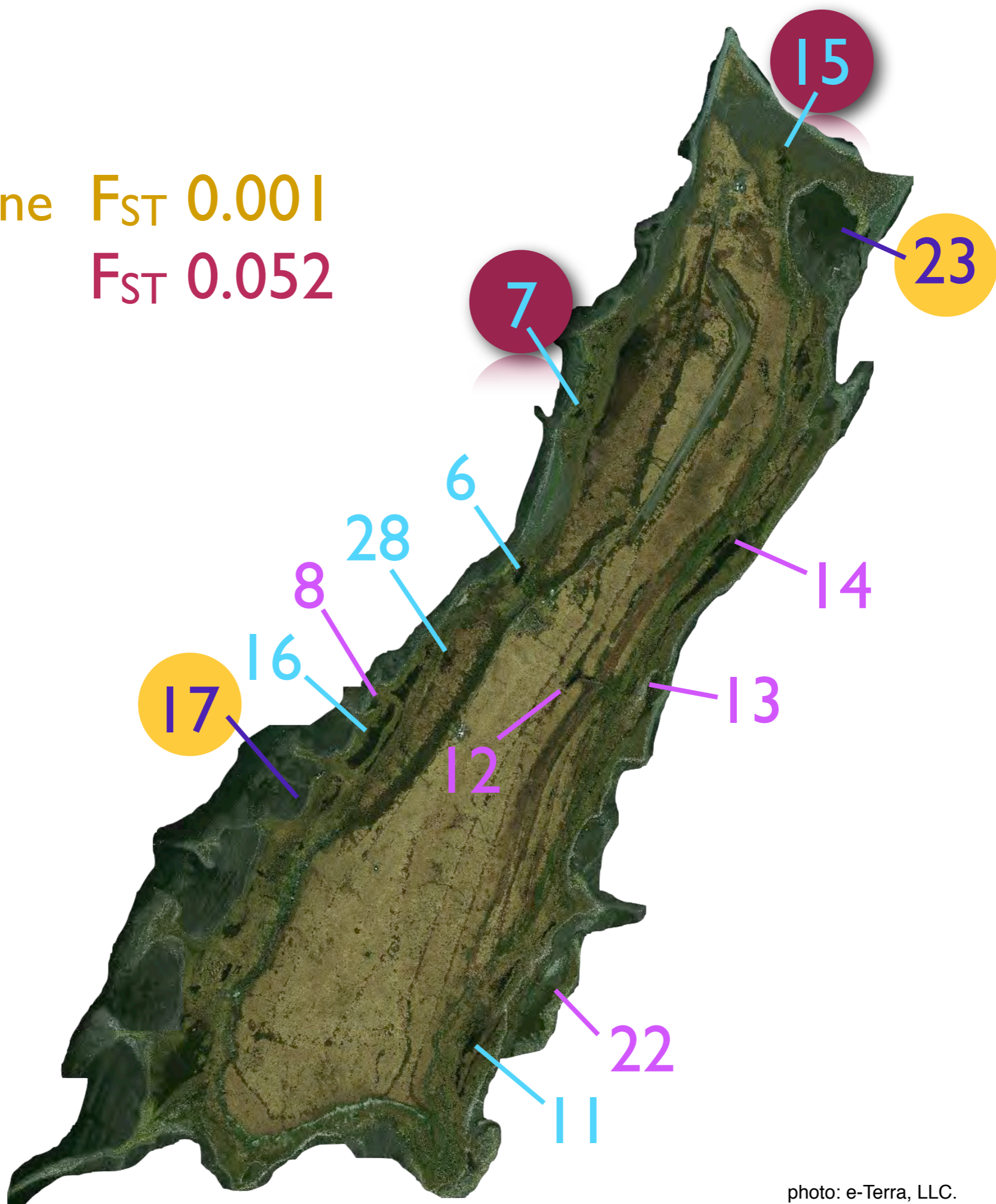


Marine x Marine

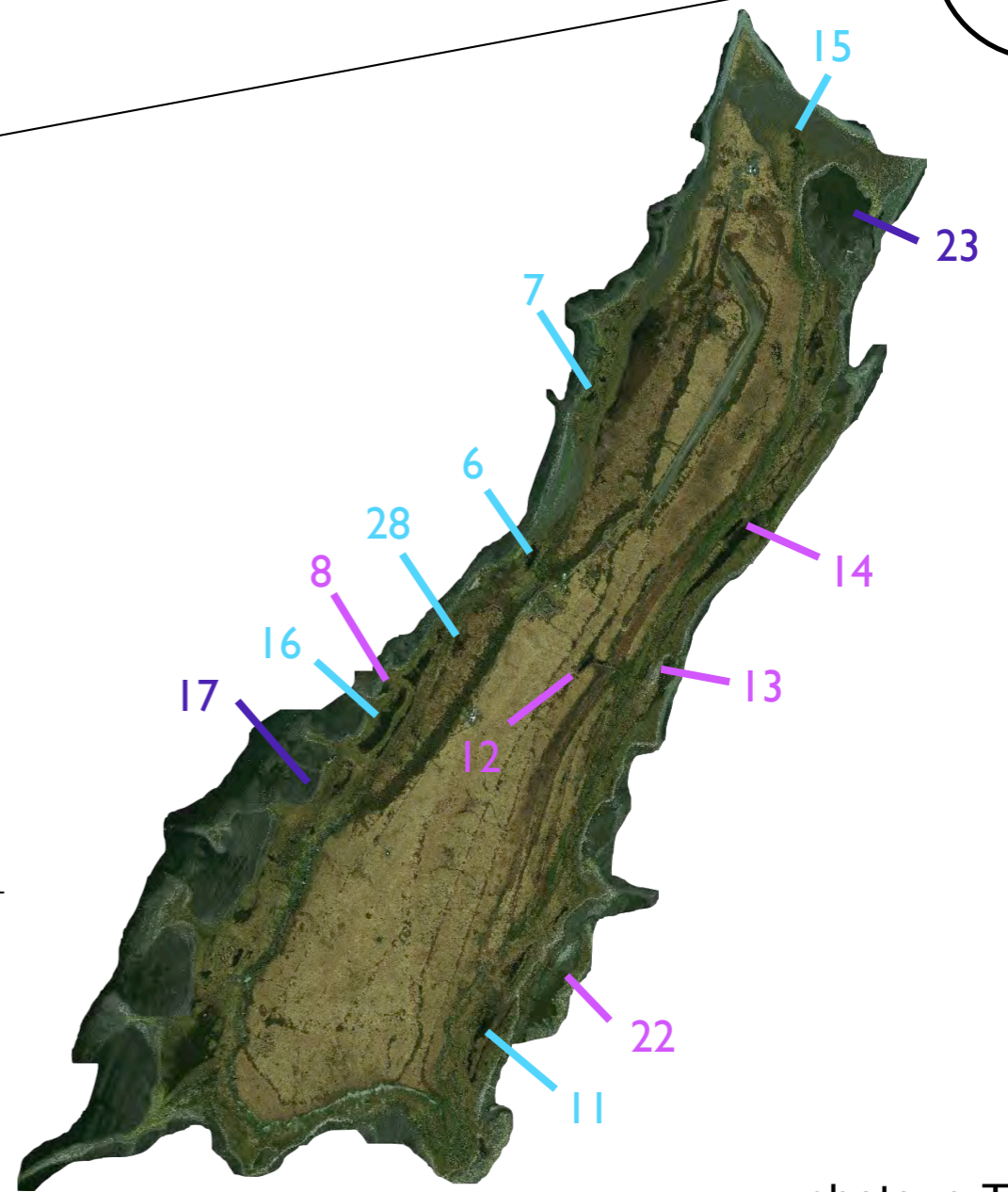
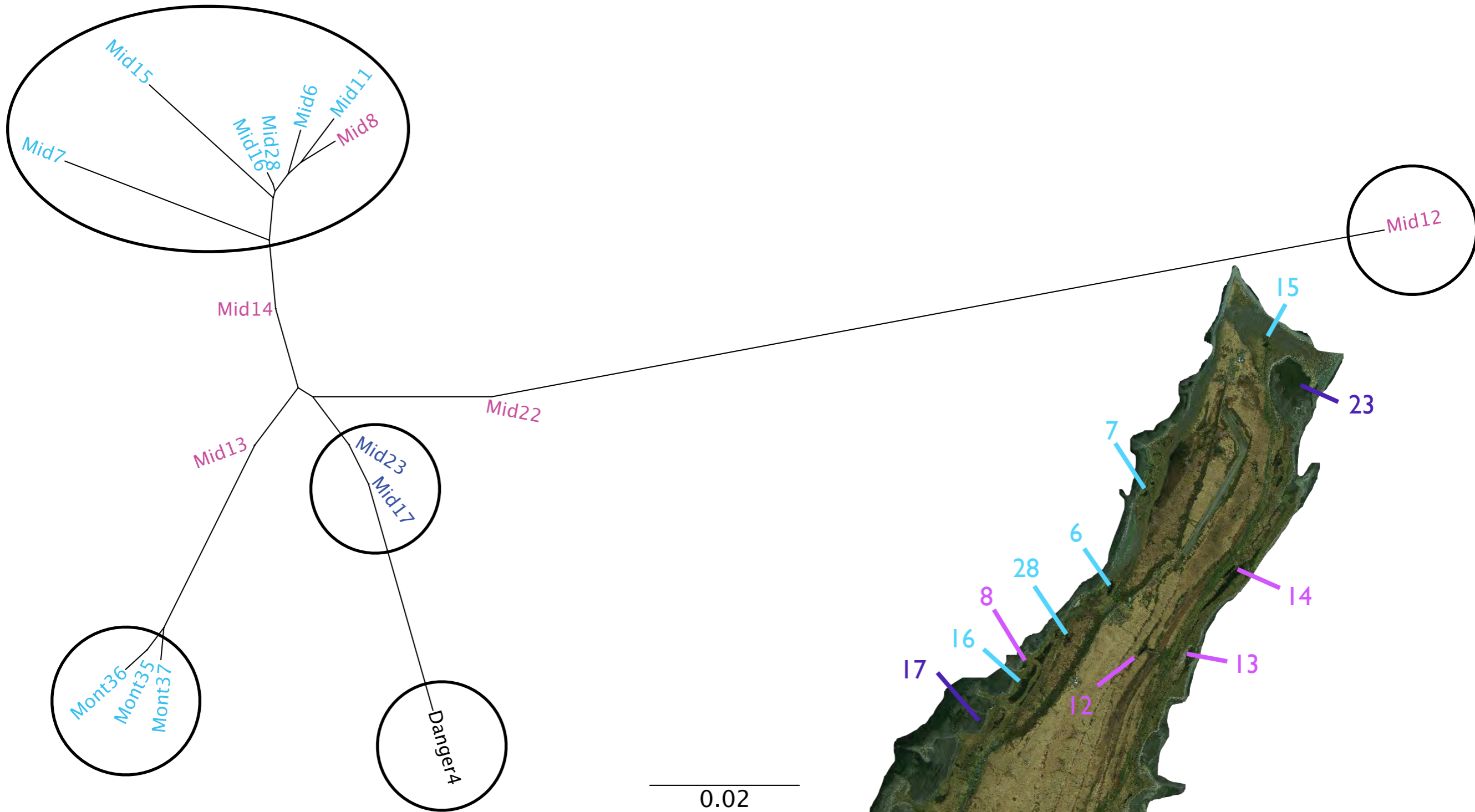
F_{ST} 0.001

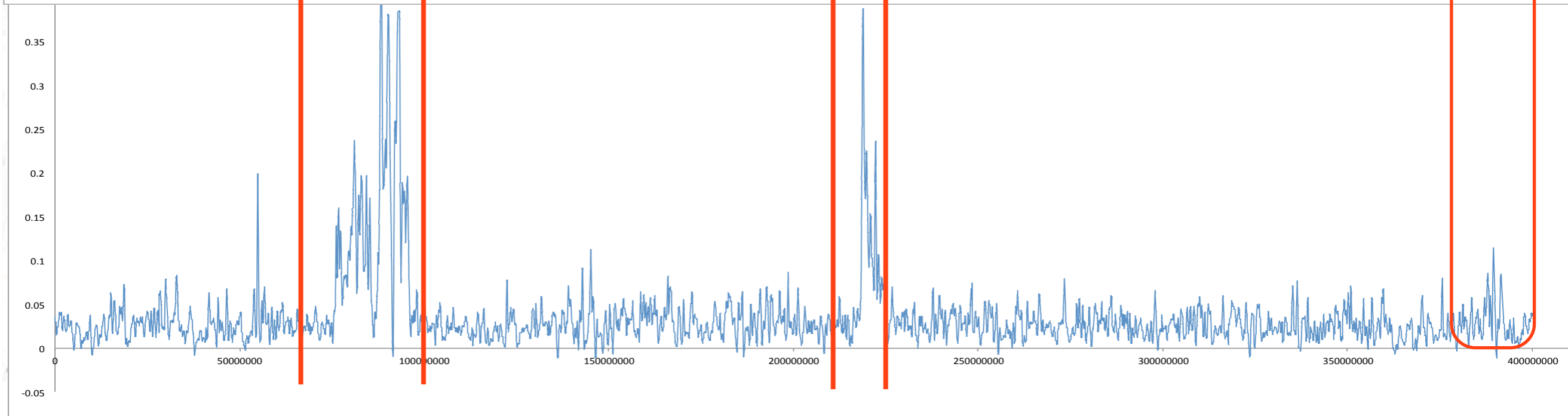
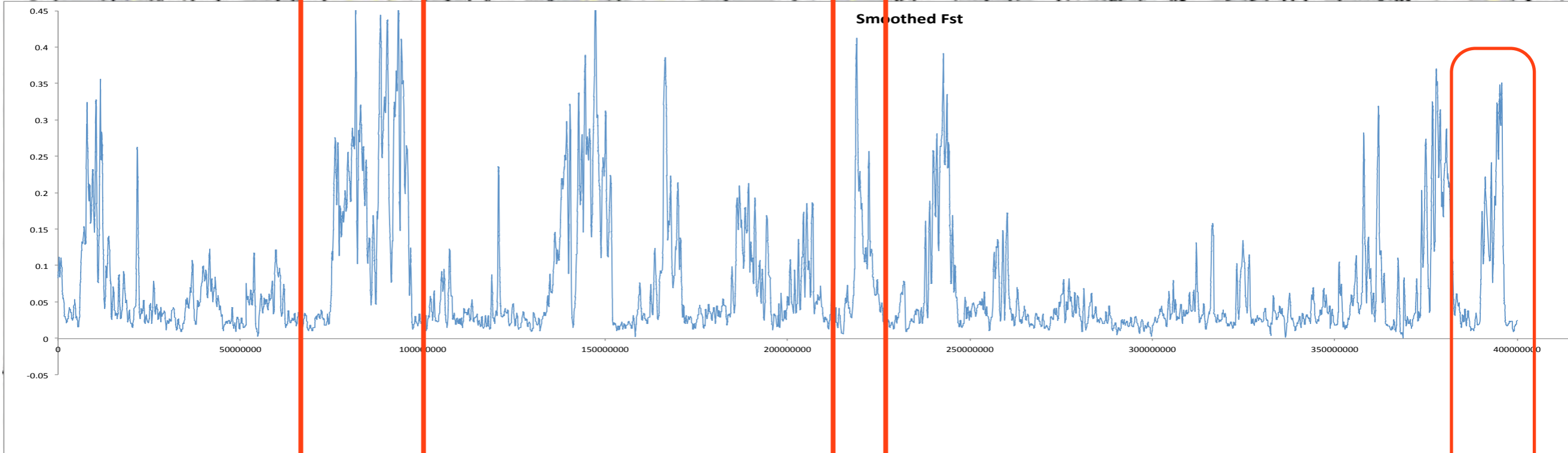
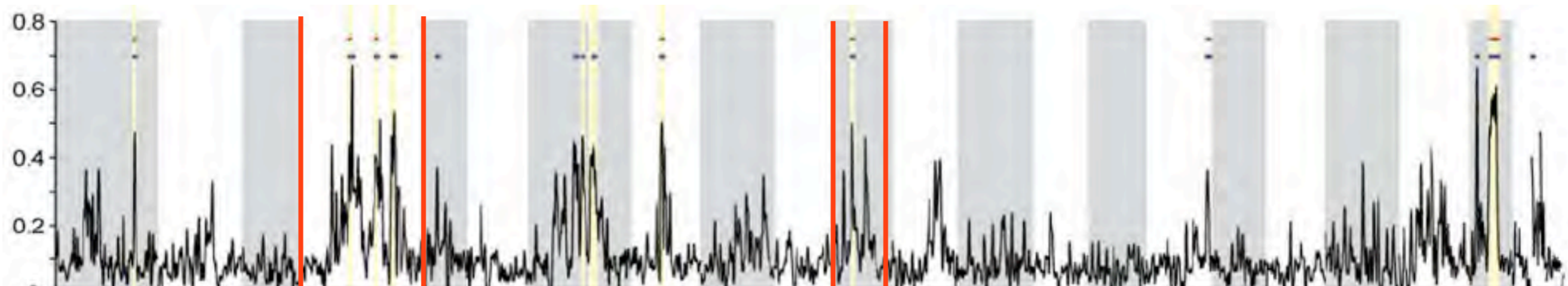
Fresh x Fresh

F_{ST} 0.052



NJ F_{ST} tree

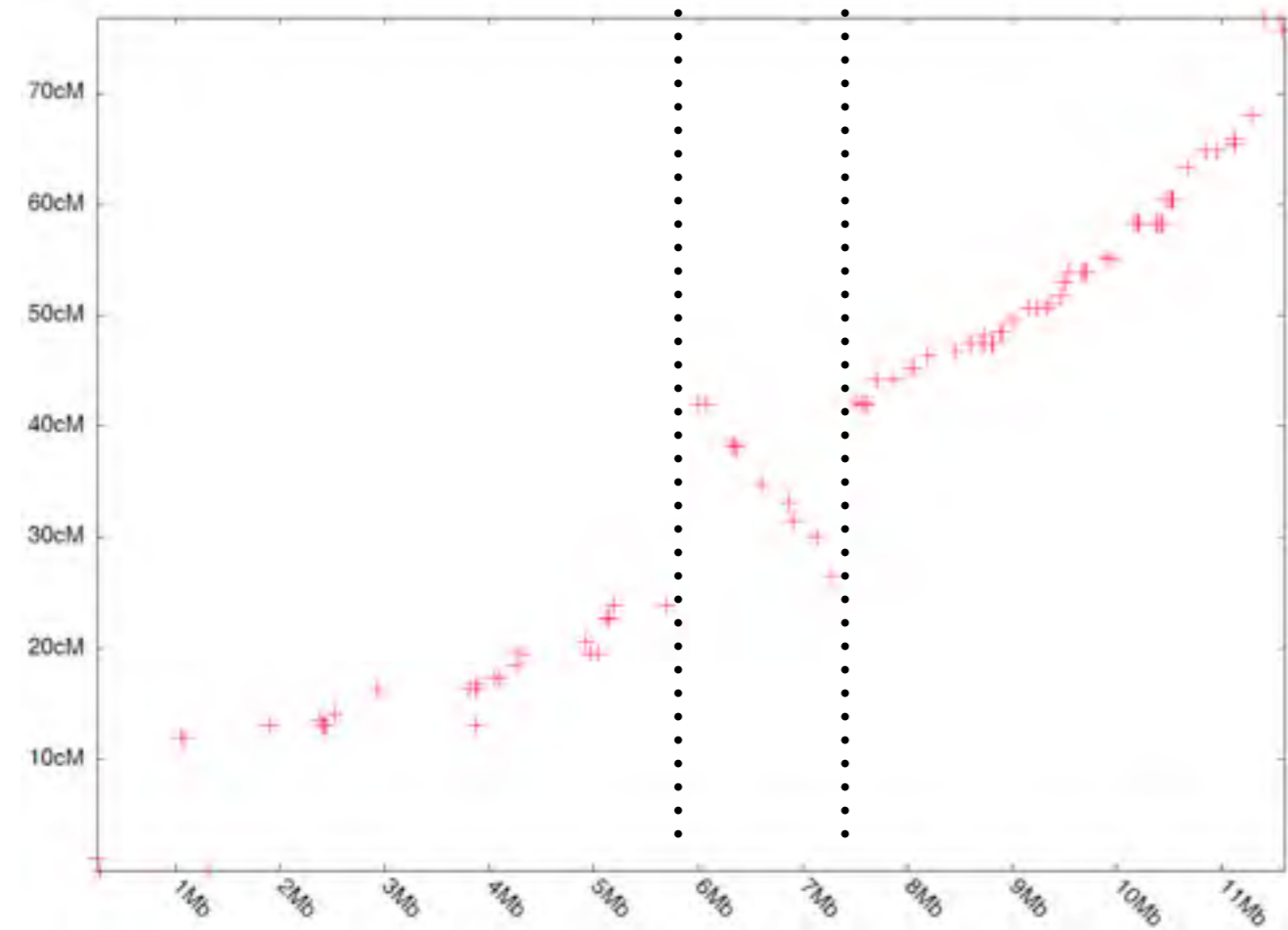
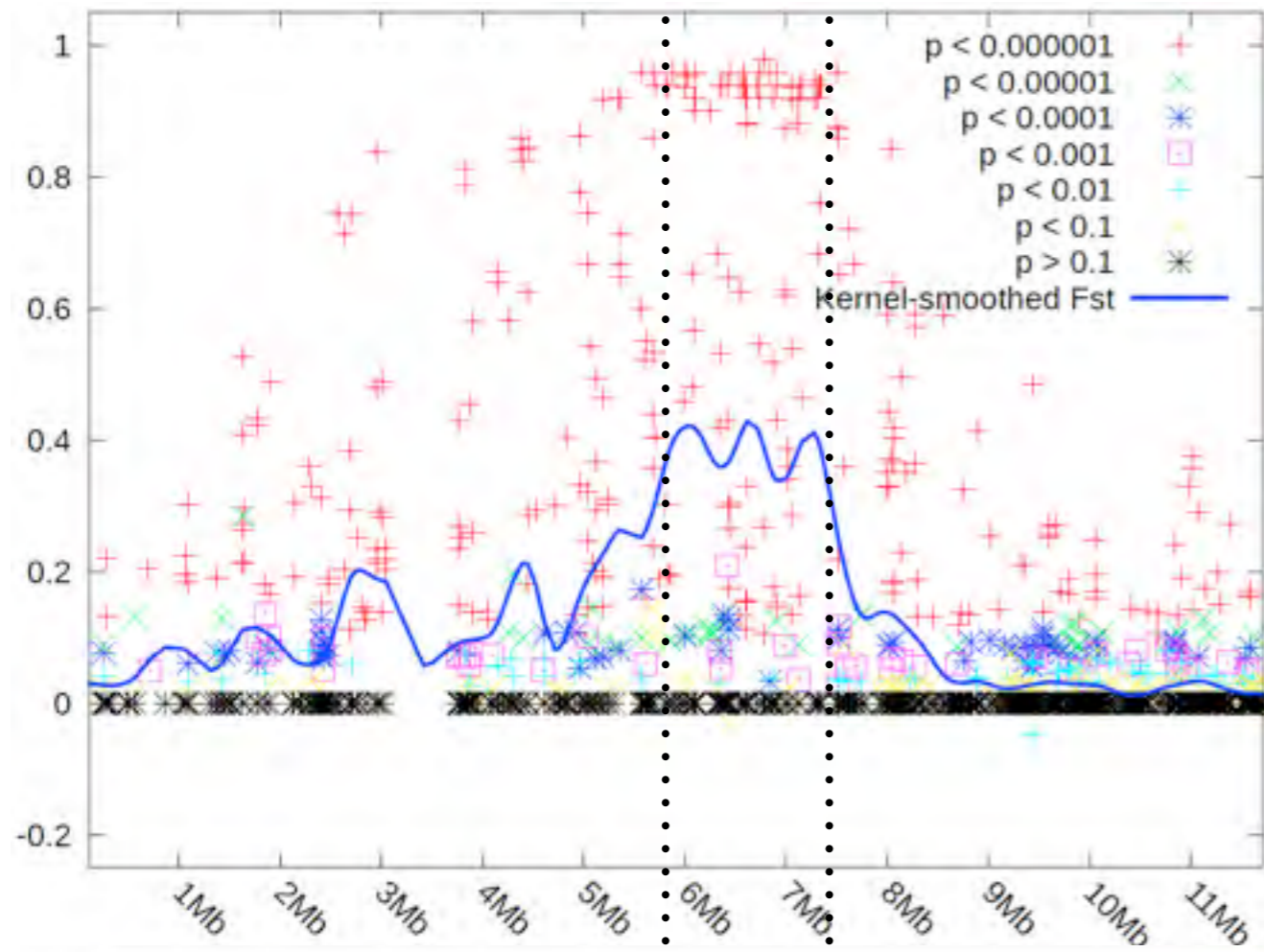




Linkage Group XXI

Fresh vs Marine 7

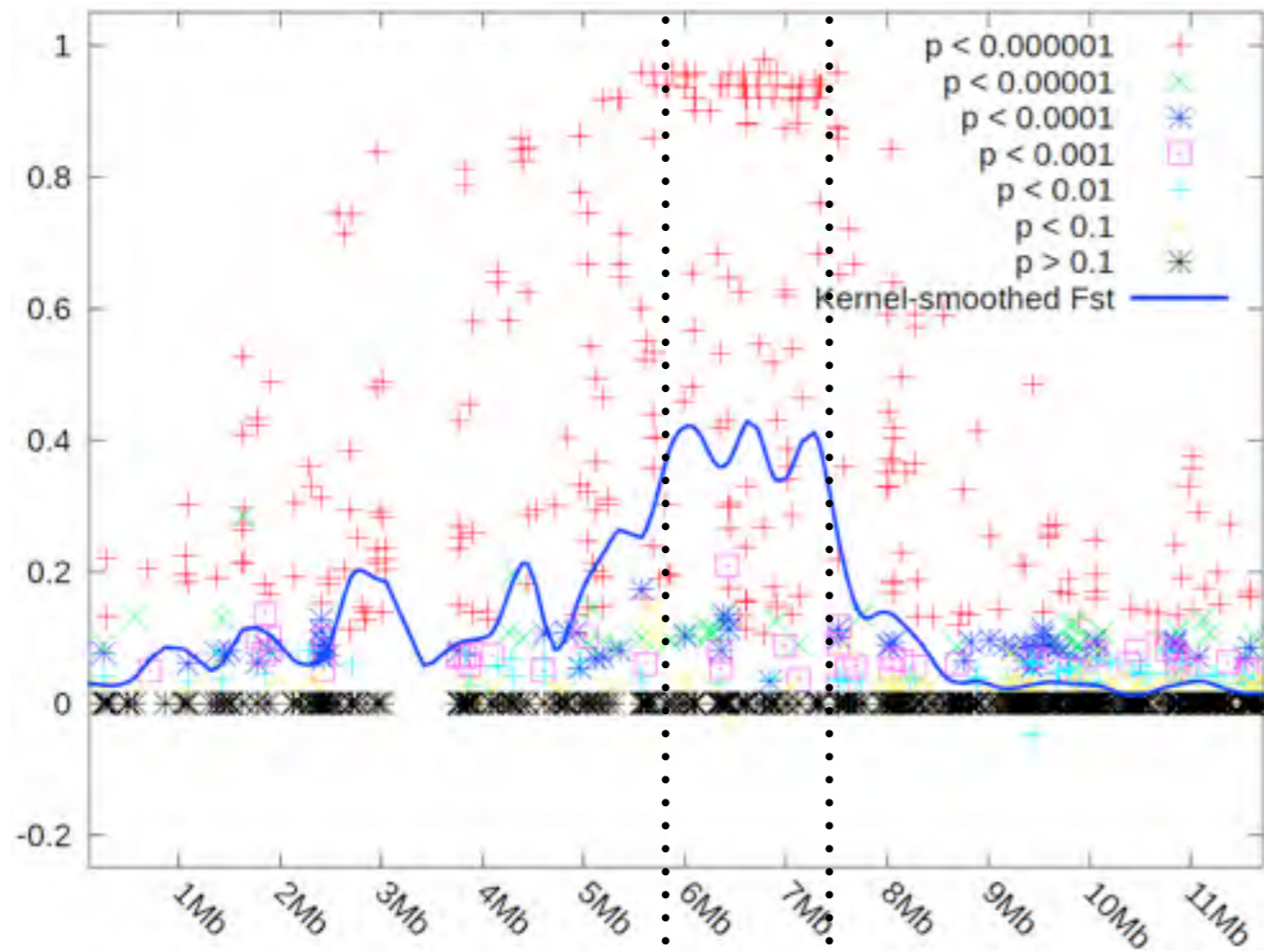
17



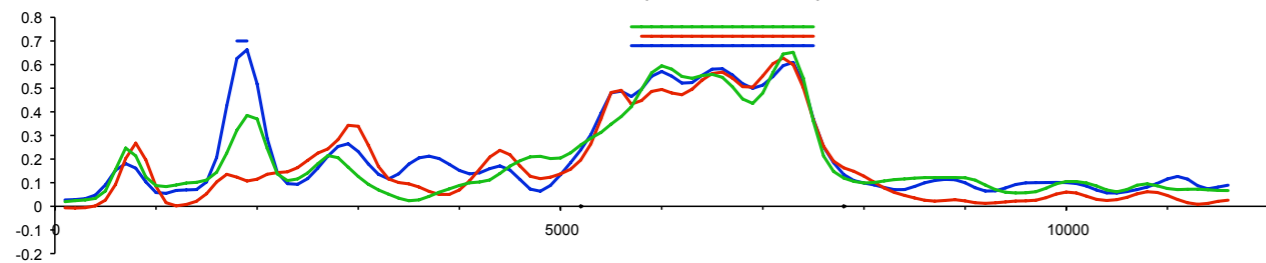
Rabbit SI x Rabbit SI



Linkage Group XXI



Fresh vs Marine 7



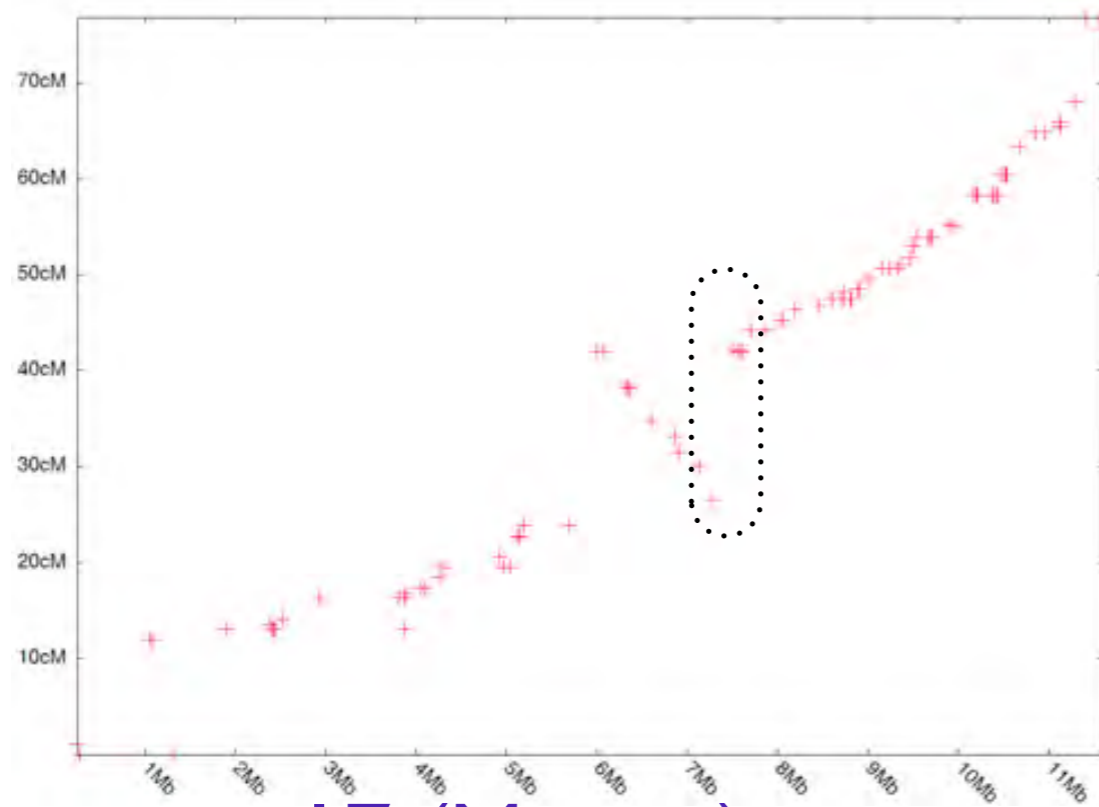
Bear Paw Lk

Boot Lk vs Marine

Mud Lk

(Hohenlohe, Bassham et al. 2010)

Linkage Group XXI



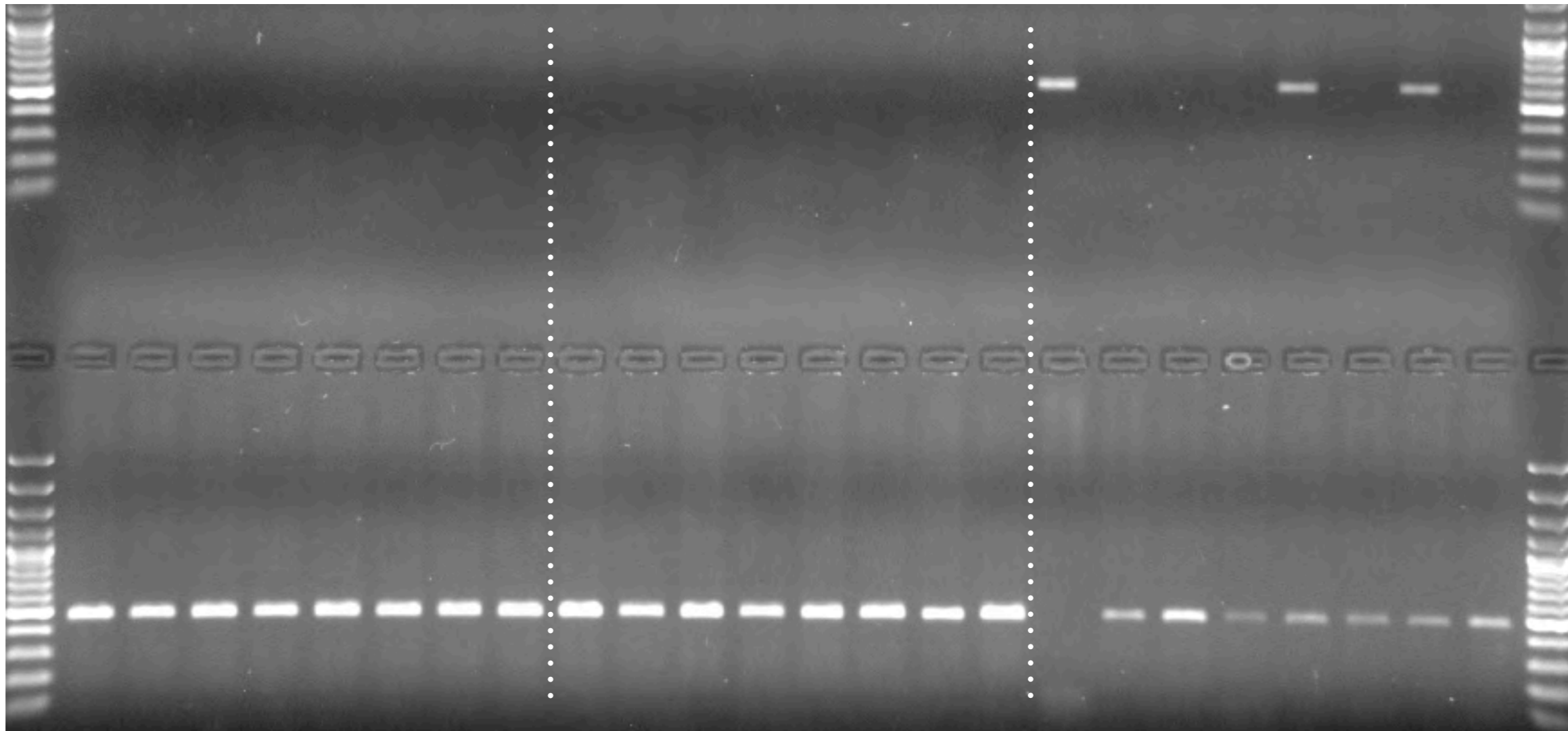
17 (Marine)

23 (Marine)

8 (Both)

Like
Bear Paw

Inverted



Other recent uses of RAD-seq

Quantitative Trait Loci (QTL) mapping

Population genomics & Genome Wide Association Studies (GWAS)

Phylogenetics and phylogeography

Genetic mapping, comparative genomics

de novo Genome assembly

Identifying signatures of selection in natural populations

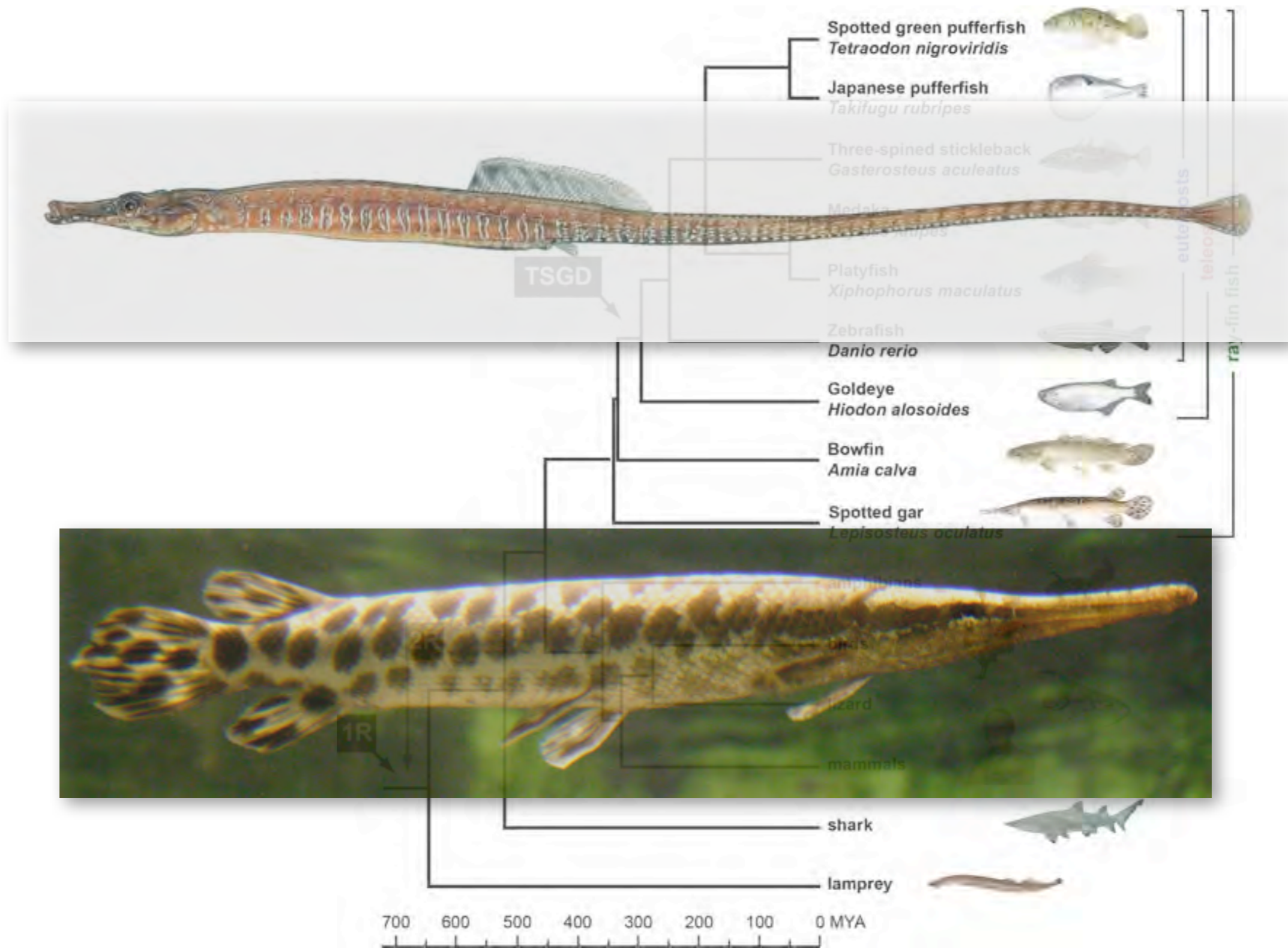
Inferring parentage and pedigrees in the wild

Quantitative genetics in outbred populations

Allele specific transcriptional profiling using RNA-seq

What if you don't have a genome sequence?

Genomically enabling very non-model organisms

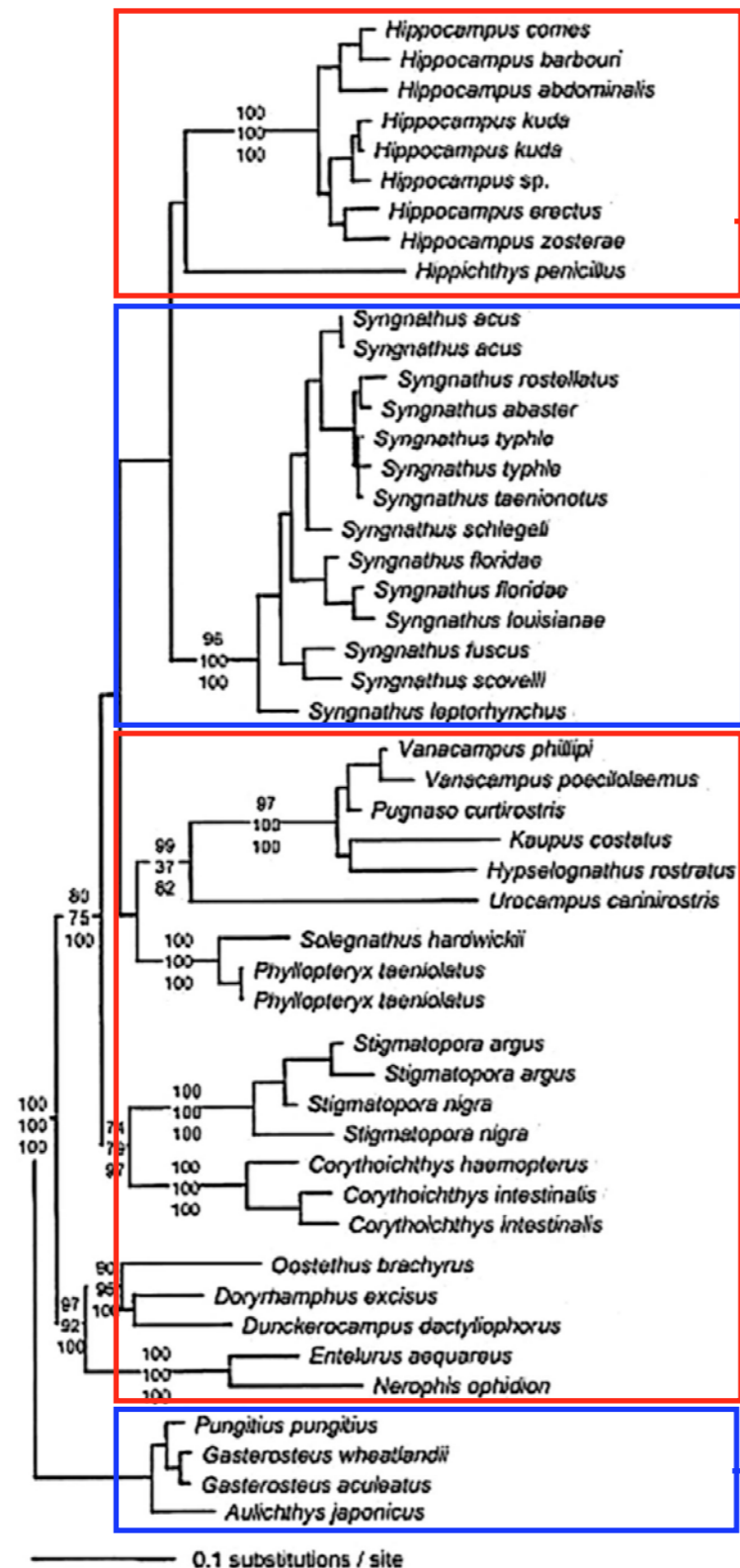


Andrew Nishida, Julian Catchen, Susie Bassham,
Clay Small and Adam Jones

Seahorses, sea dragons and pipefishes



Gasterosteidae and Syngnathidae are historically considered to be closely related



Seahorses



Pipefish



Seadragons



Stickleback



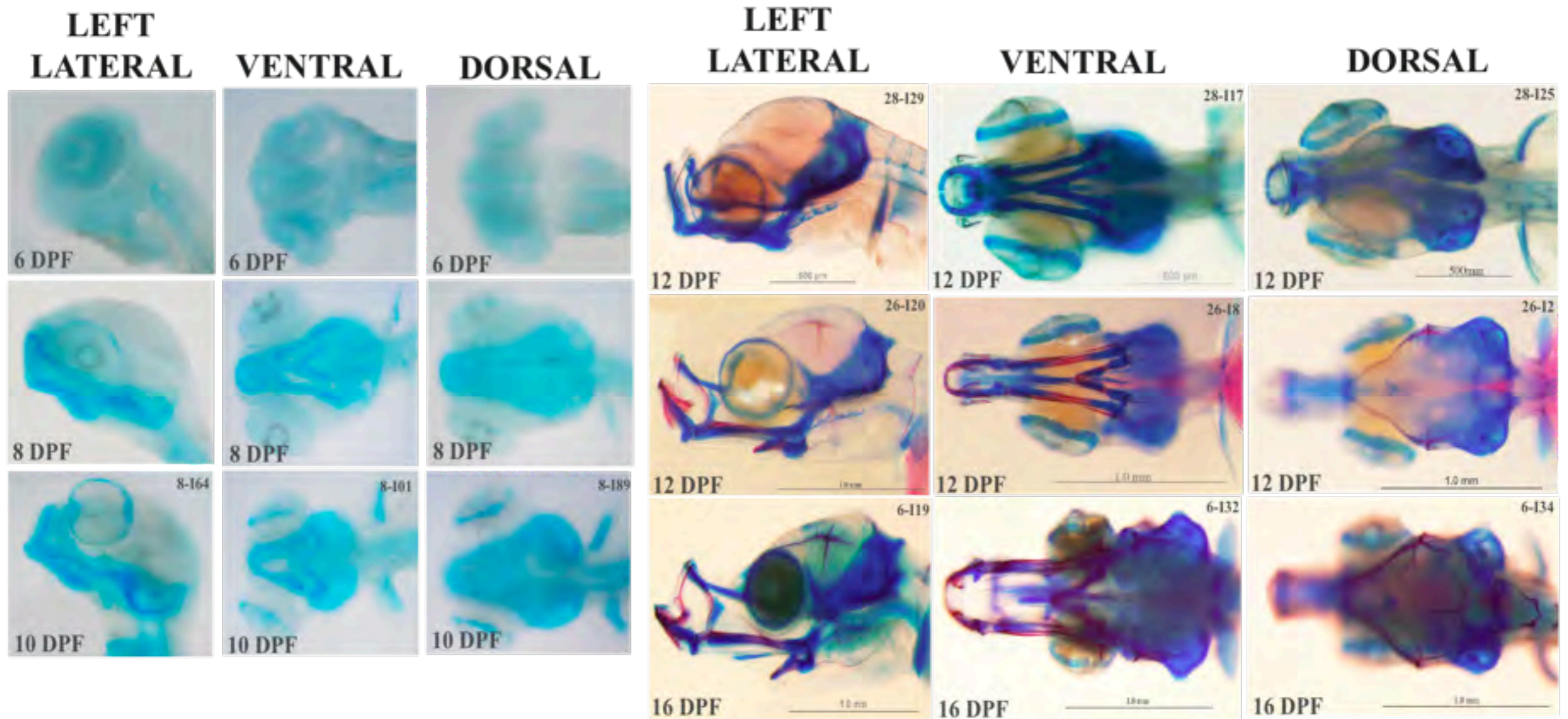


Gulf Pipefish *Syngnathus scovelli*

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



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

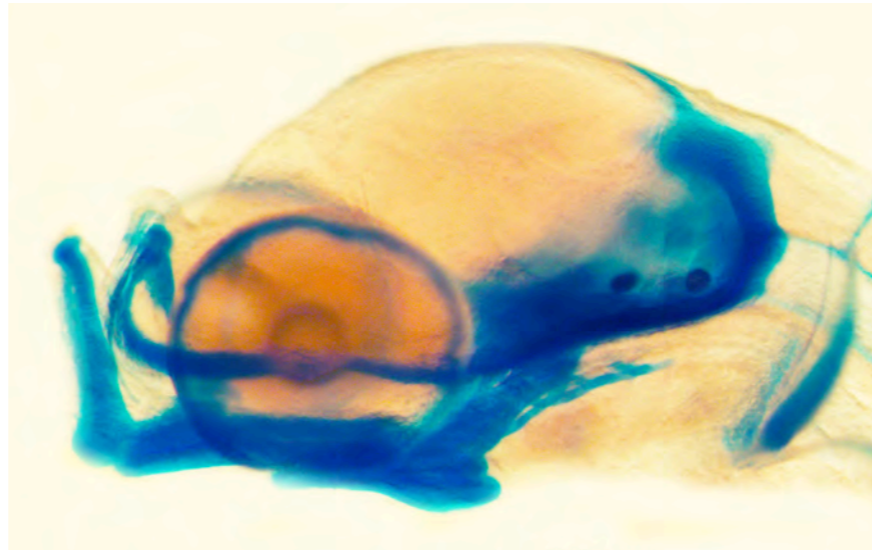
3) Deep coverage shotgun sequencing of genome

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)



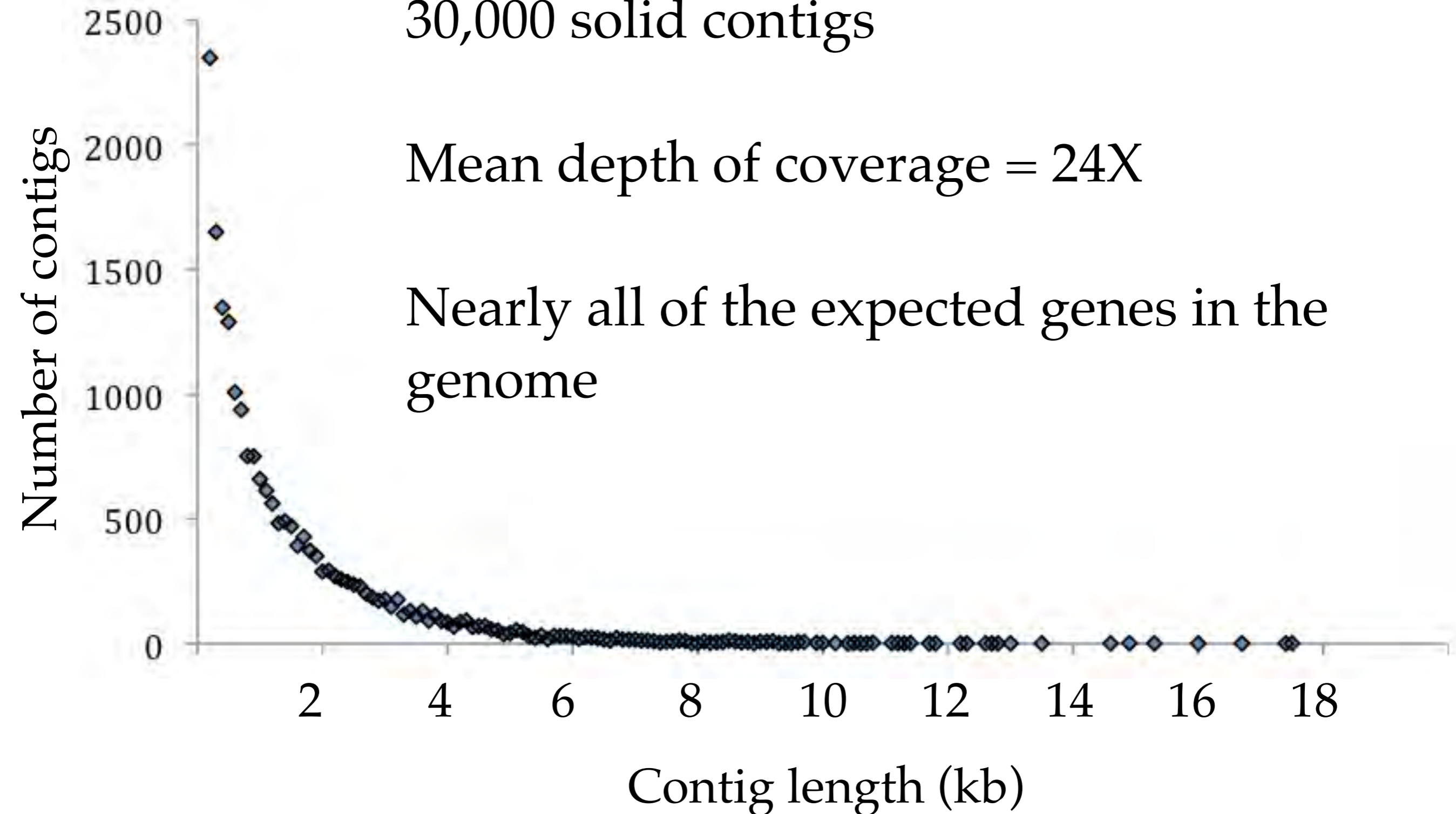
Assembly of transcripts

Transcriptome

30,000 solid contigs

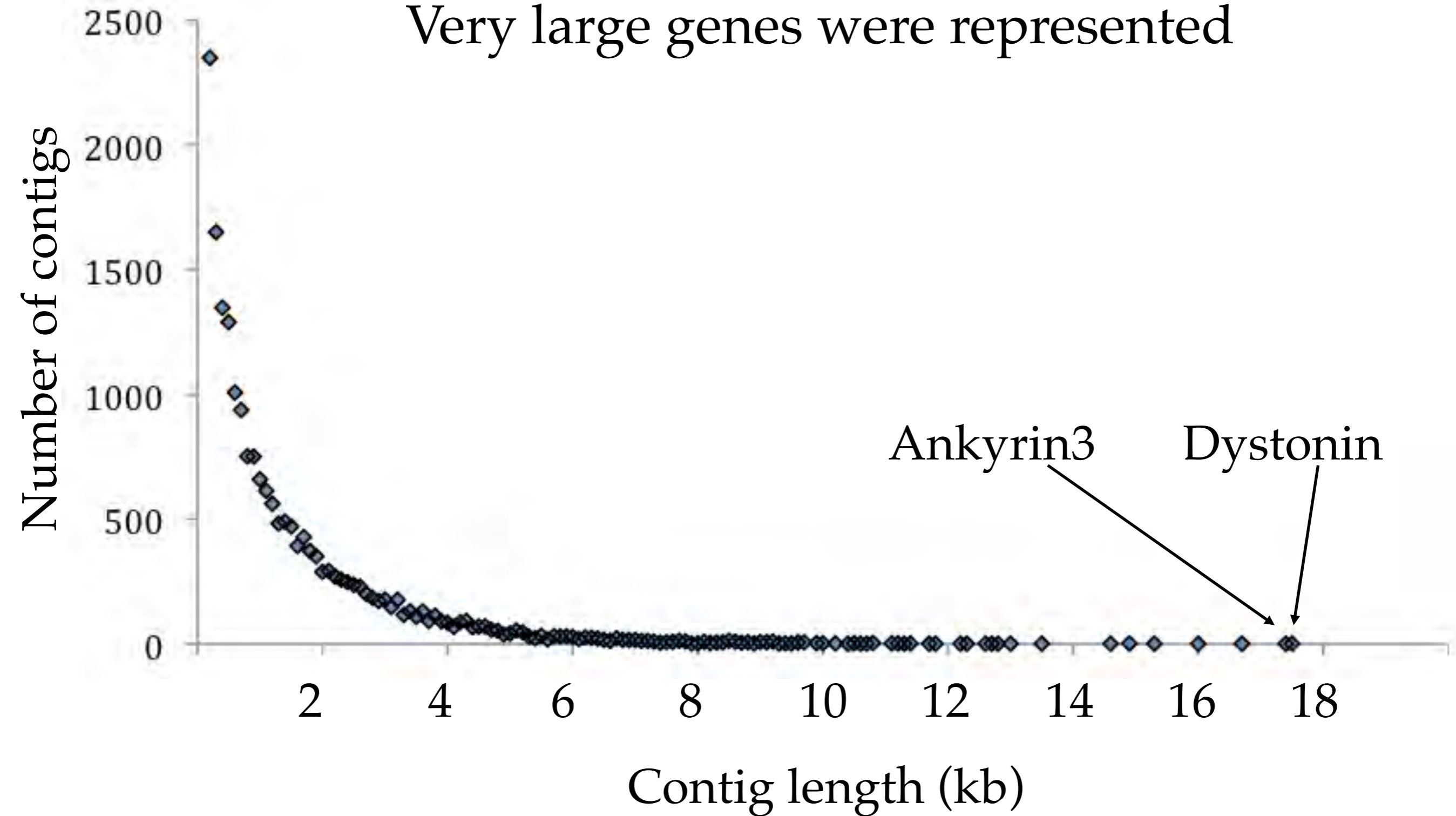
Mean depth of coverage = 24X

Nearly all of the expected genes in the genome



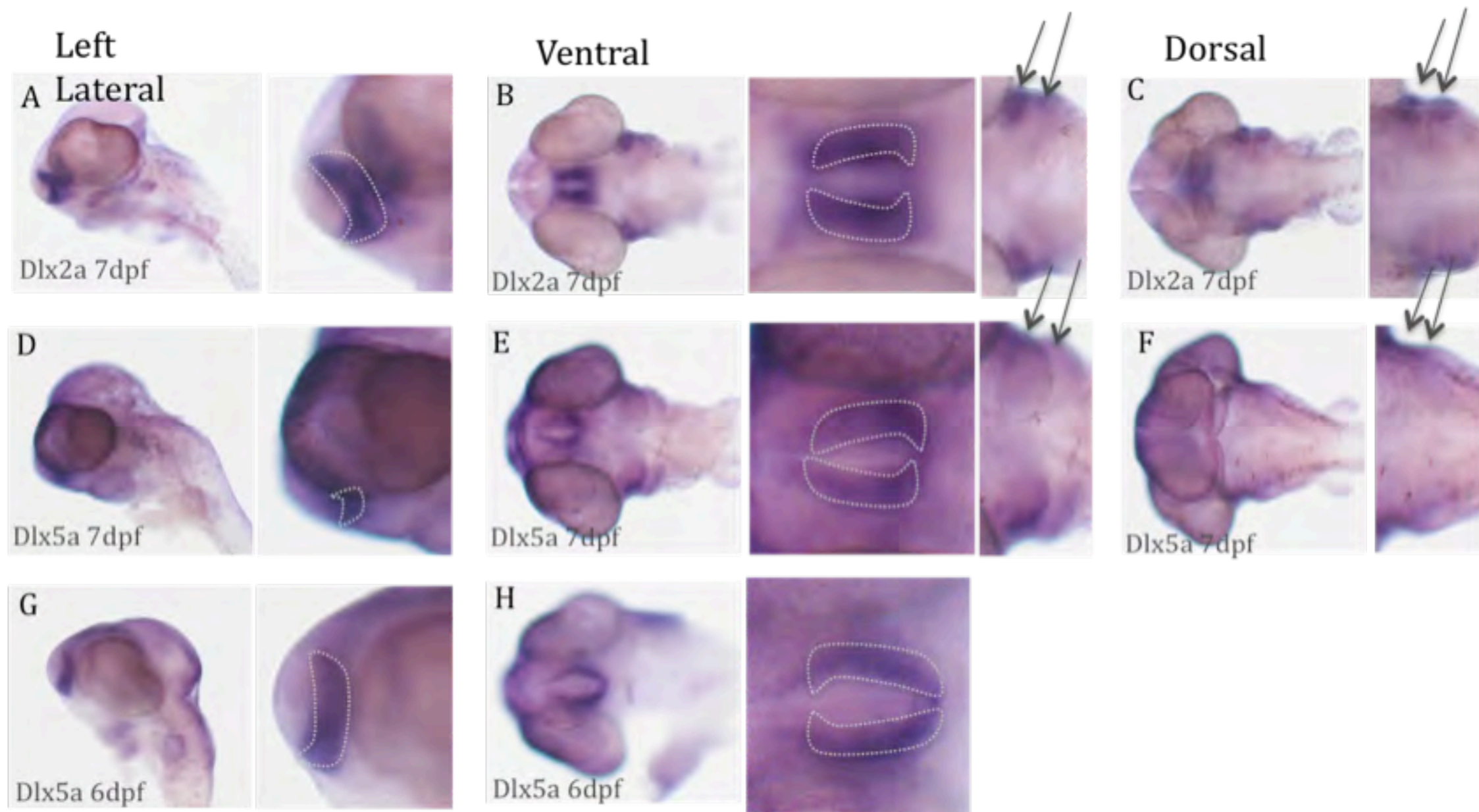
Transcriptome

Very large genes were represented



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

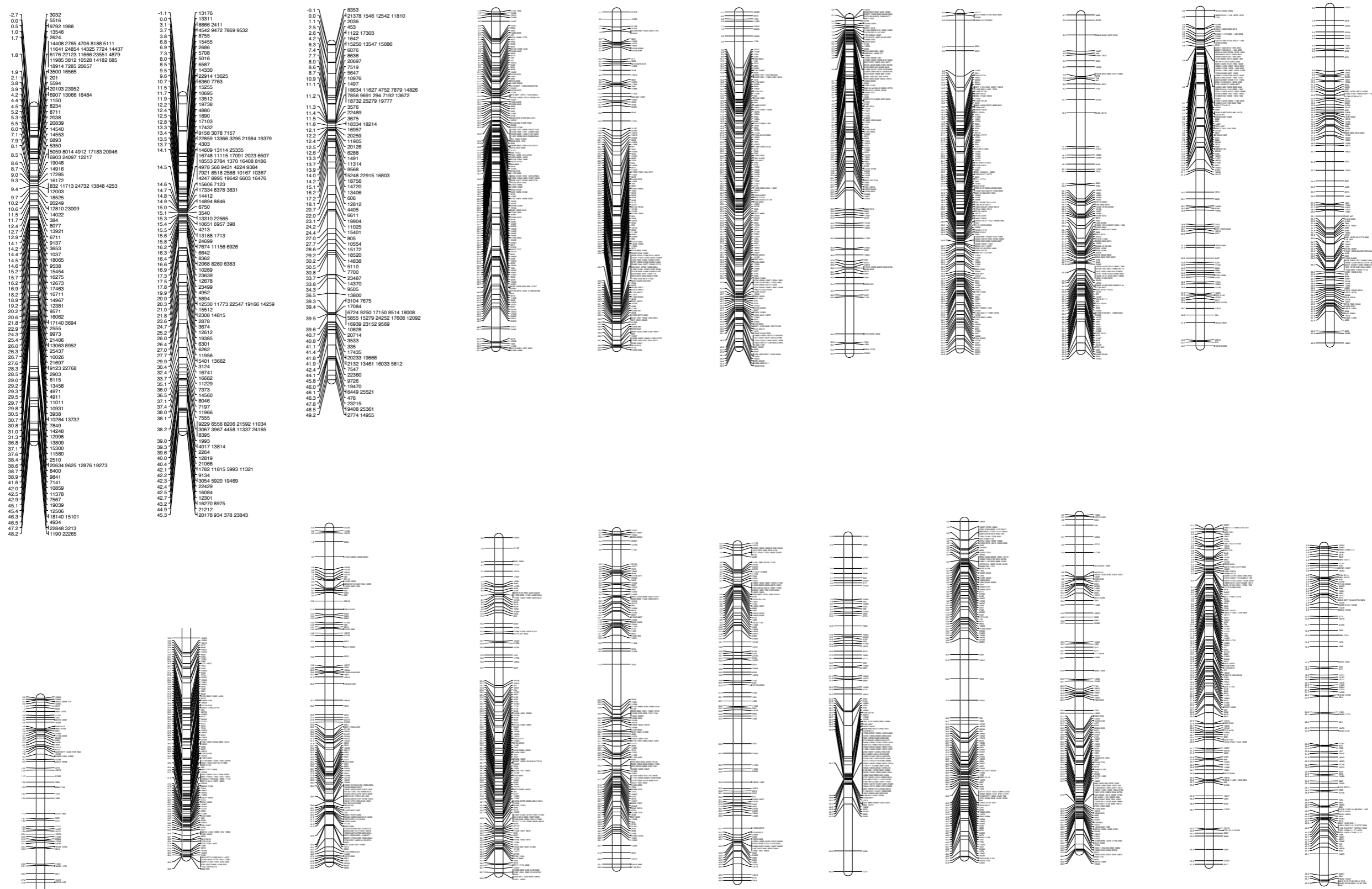
Analyzed the data using *Stacks*

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

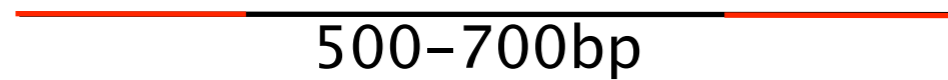
Random Illumina shotgun sequencing

Removed highly repetitive kmers

Produced *several* different genome assemblies

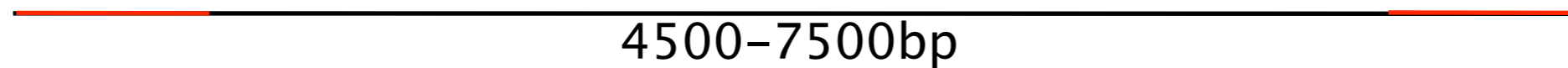
Illumina genomic libraries for pipefish genome

paired end 101bp



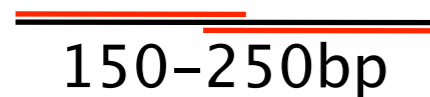
25x

mate pair



2x

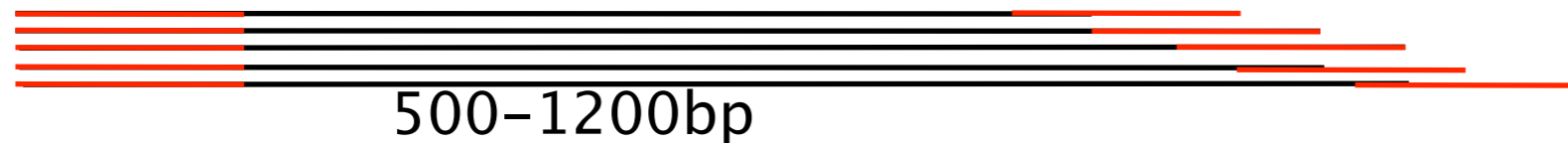
overlapping



40x

paired end RAD

ACTCTC



15-25x of
3% of the
genome

Pipefish genome assembly version 0.99

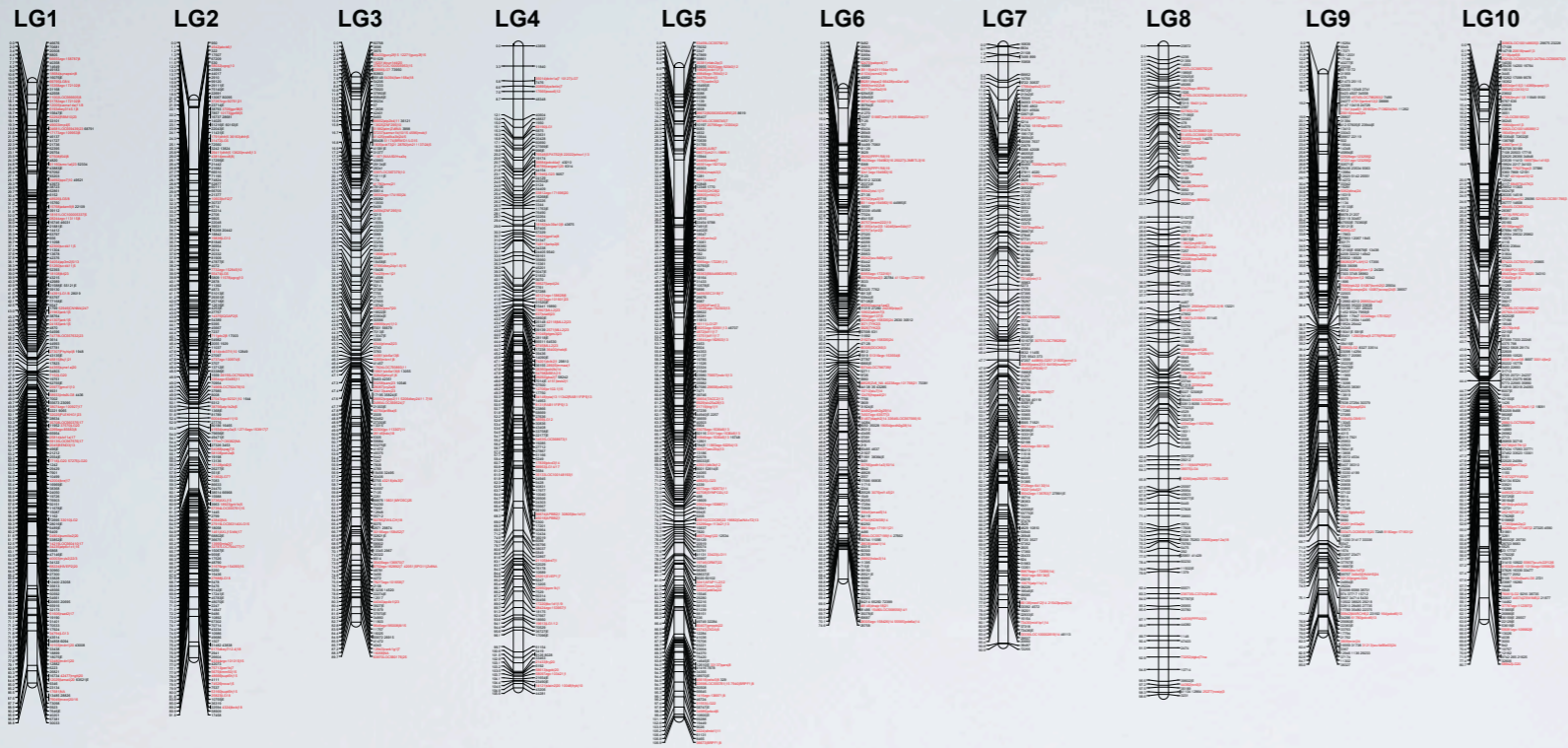
Nearly the whole genome is covered

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

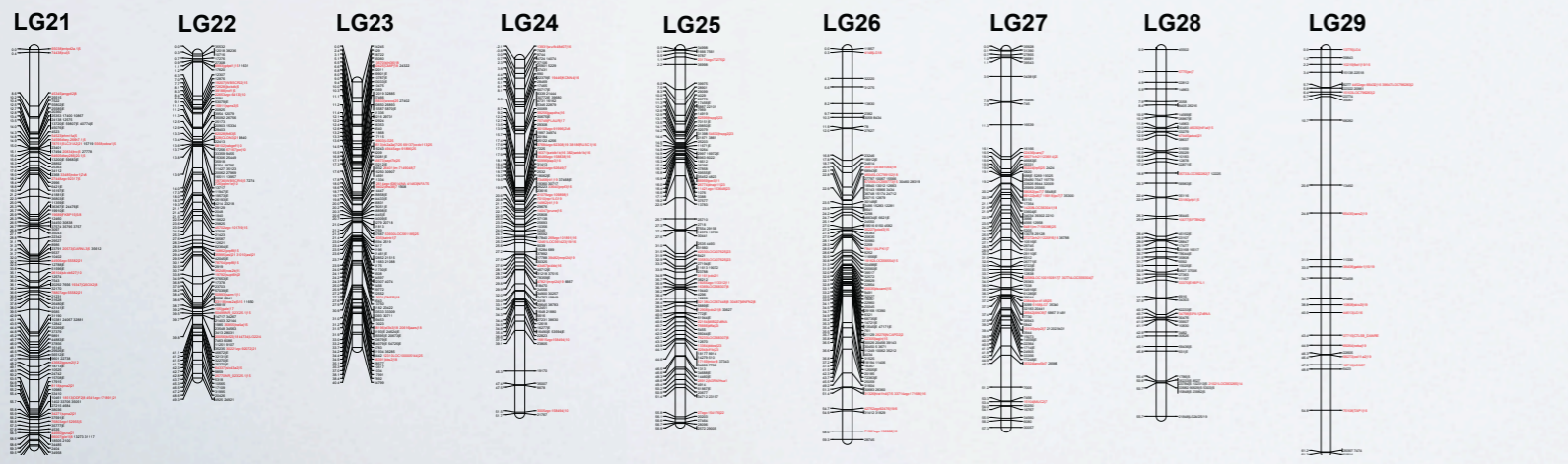
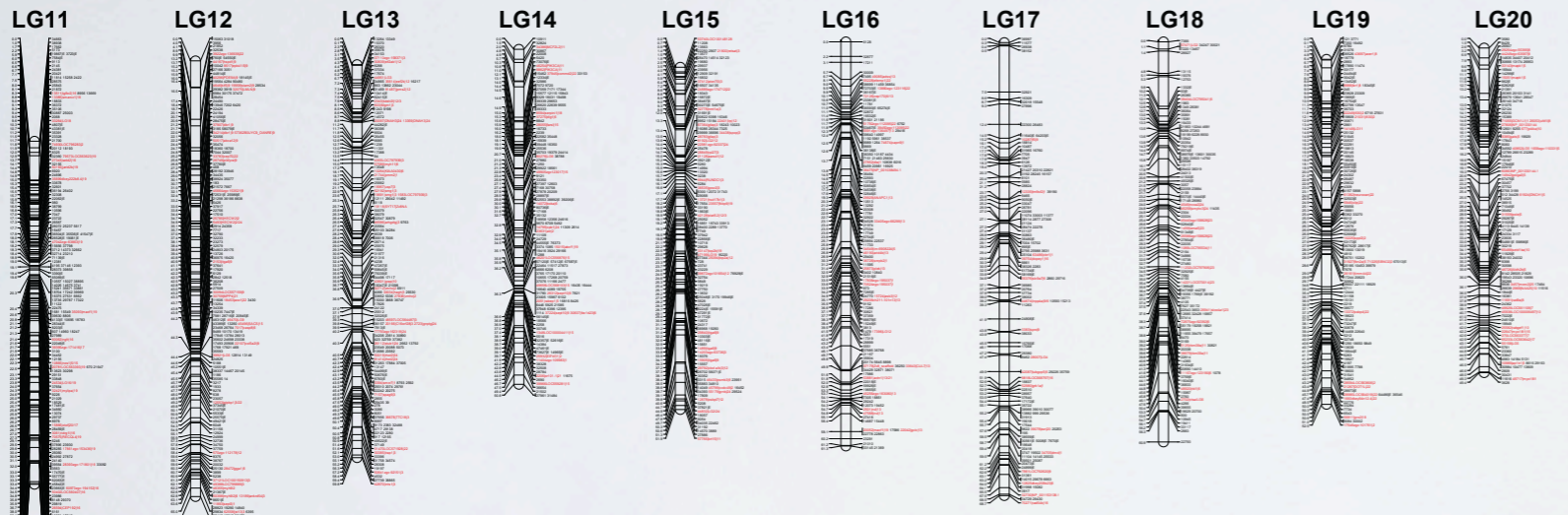
Max	Average Length	Total Length	Gap Length	%
198,155	9,916.35	336,273,415	38,303,839	(11.39%)

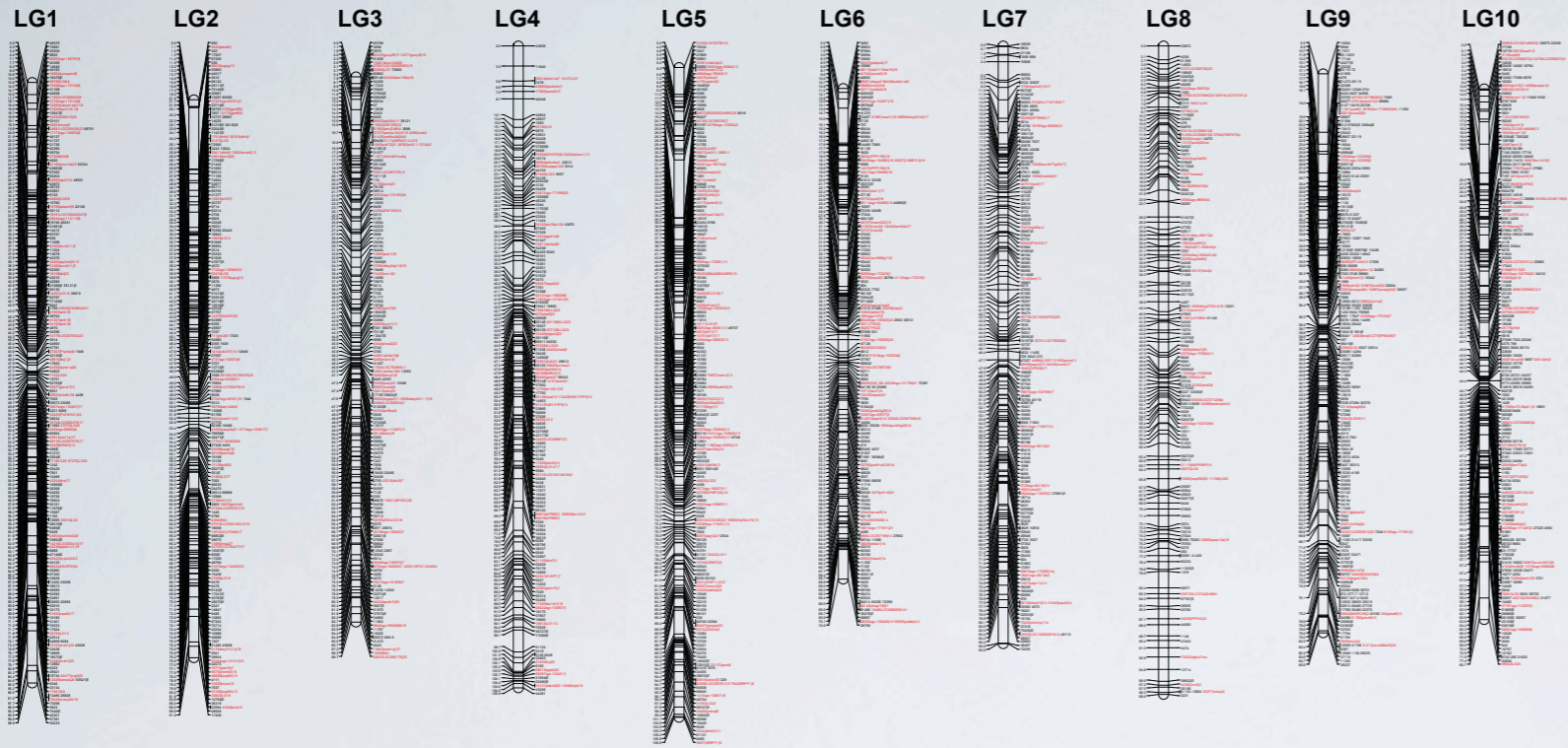
Bringing it all together; the spotted gar



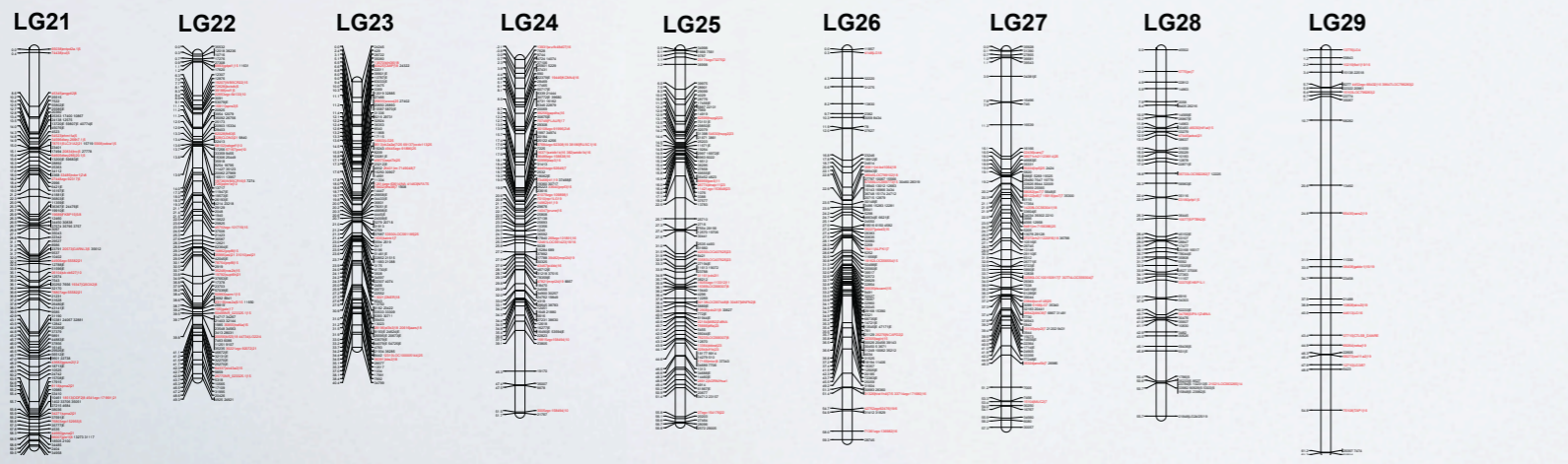
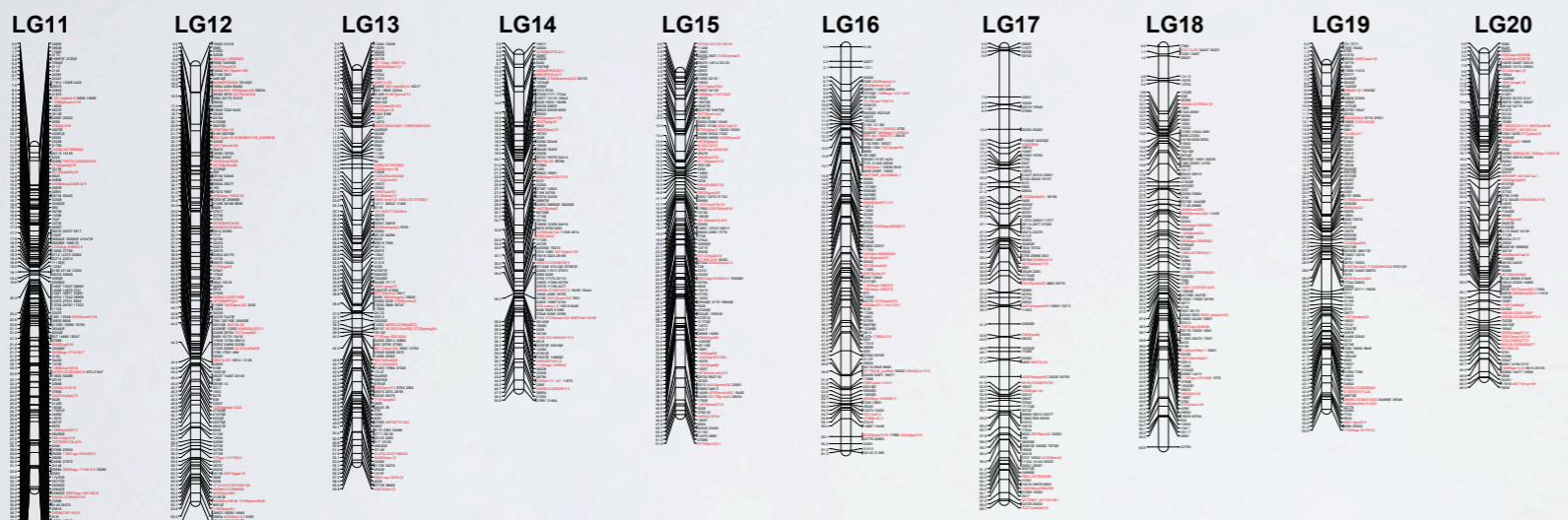


94 Individuals
 15,076 Markers
 8,046 Mapped
 974 In Genes

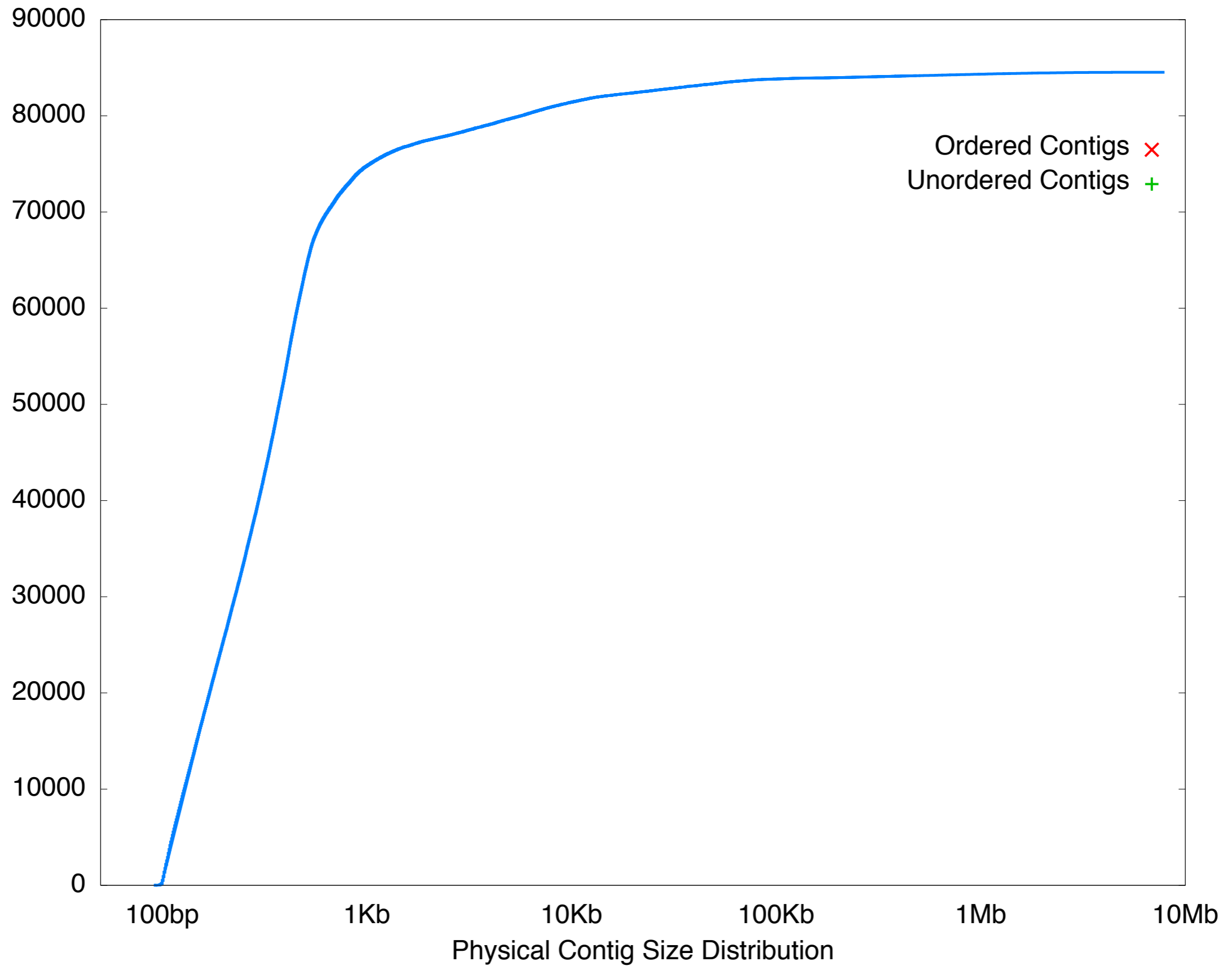


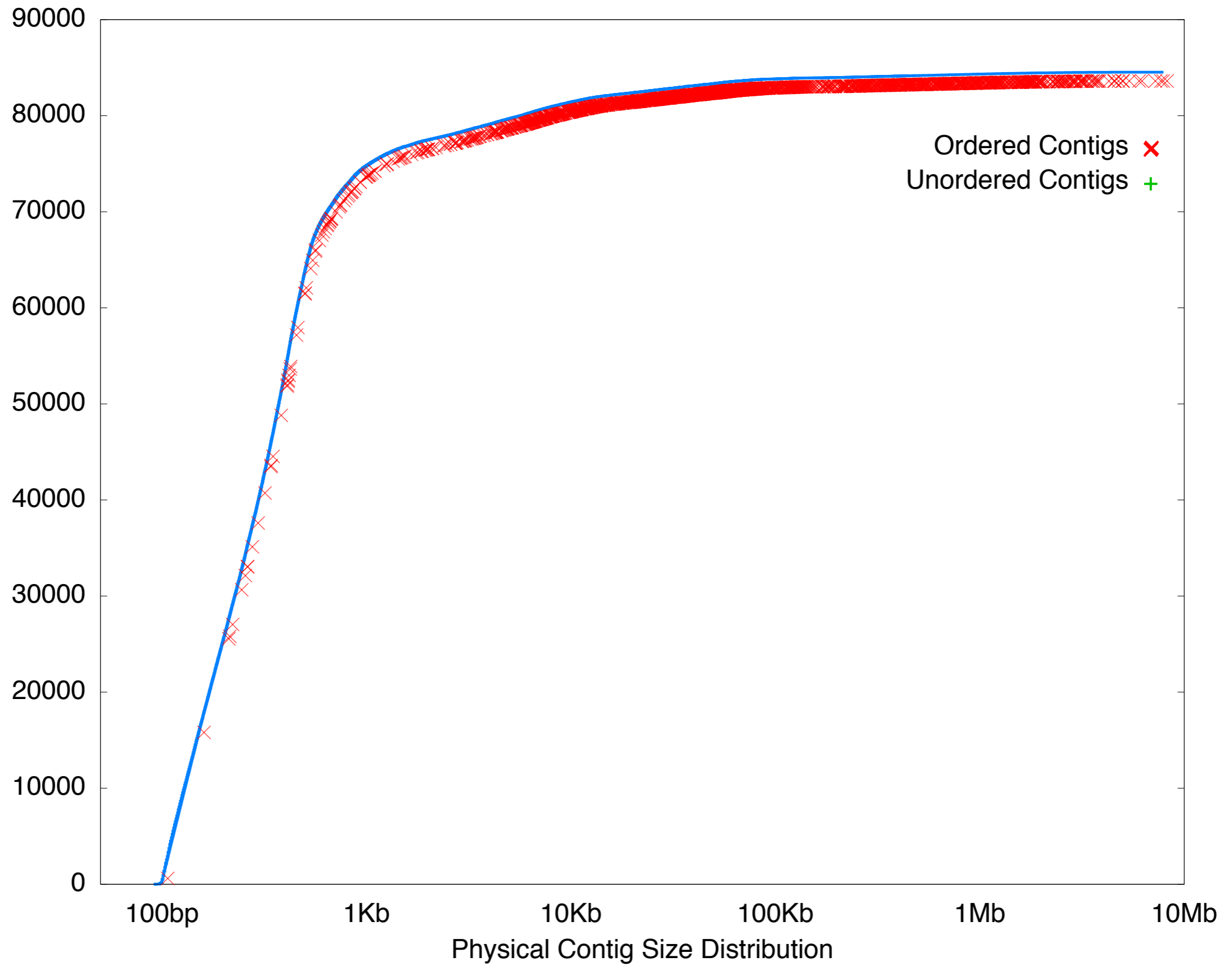


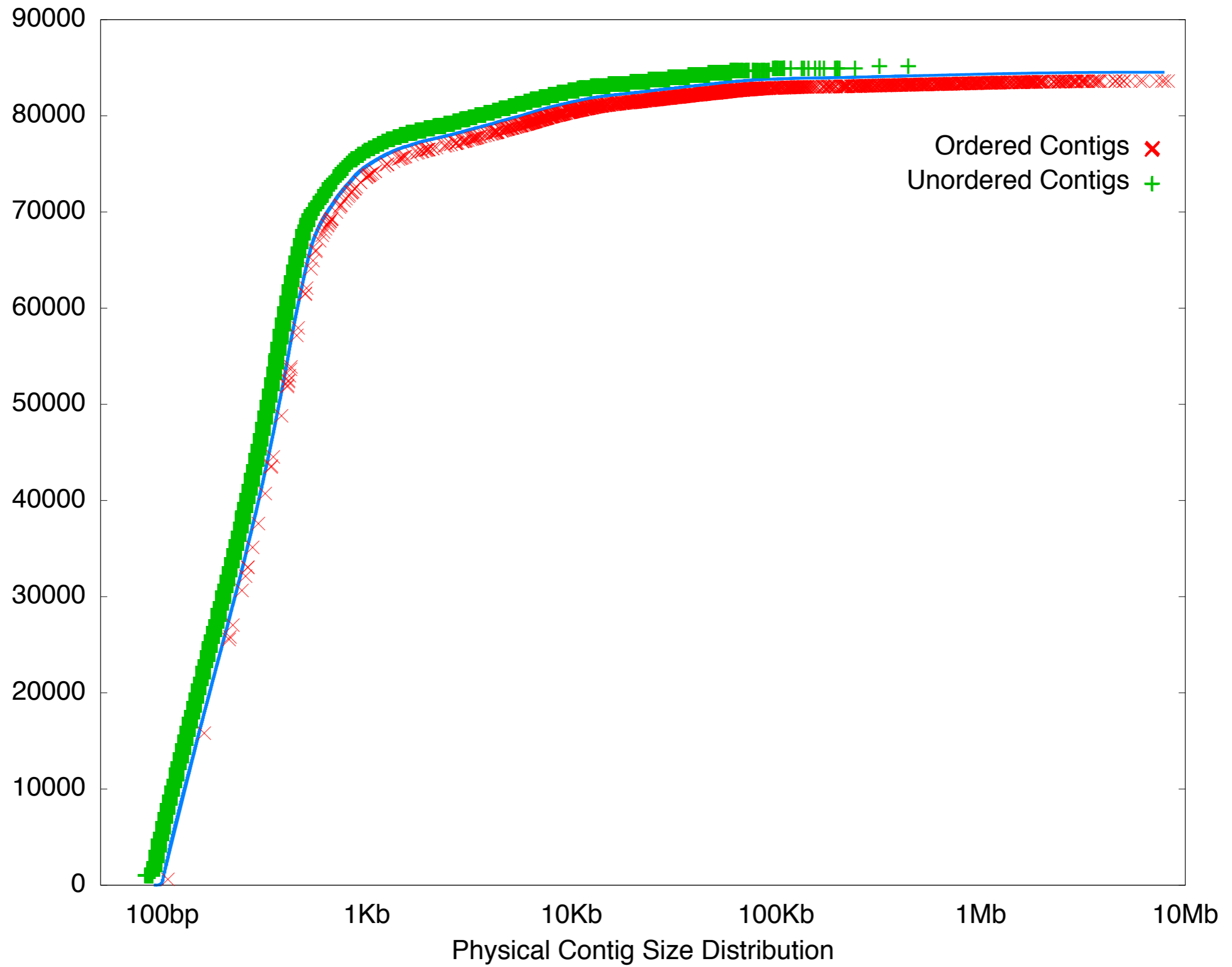
94 Individuals
 15,076 Markers
 8,046 Mapped
 974 In Genes



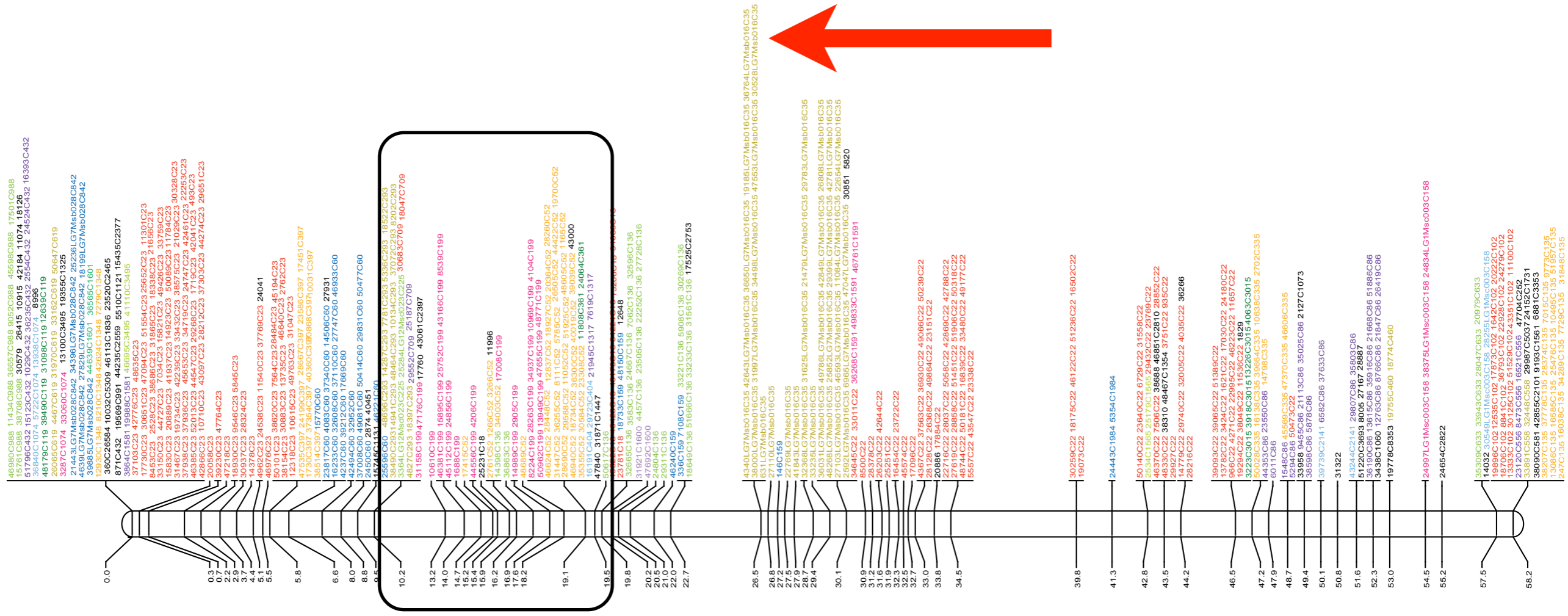
Organism	Markers
Silver carp	483
Guppy	790
Barramundi	240
Catfish	331
Sea bass	368
Cichlid	204
Platyfish	290
Halibut	604
Sea bream	204







LG7



- multiple RAD sites per segregating marker means that more of the genome can be tiled
- Mis-assemblies are easily identified

Using PE RAD
for local
assembly



Match to marker catalog

TGCAGGGTATTAGCATAA

Collate/Assemble PE reads

AACTAATTTTTCACCTAGCCATCTTGAATGTGAGTAGCATTTTAAGTAACTATAATTG

Associate
markers / PE
contigs with
ESTs & genomic
contigs

BLASTn

BLASTn

EST Library



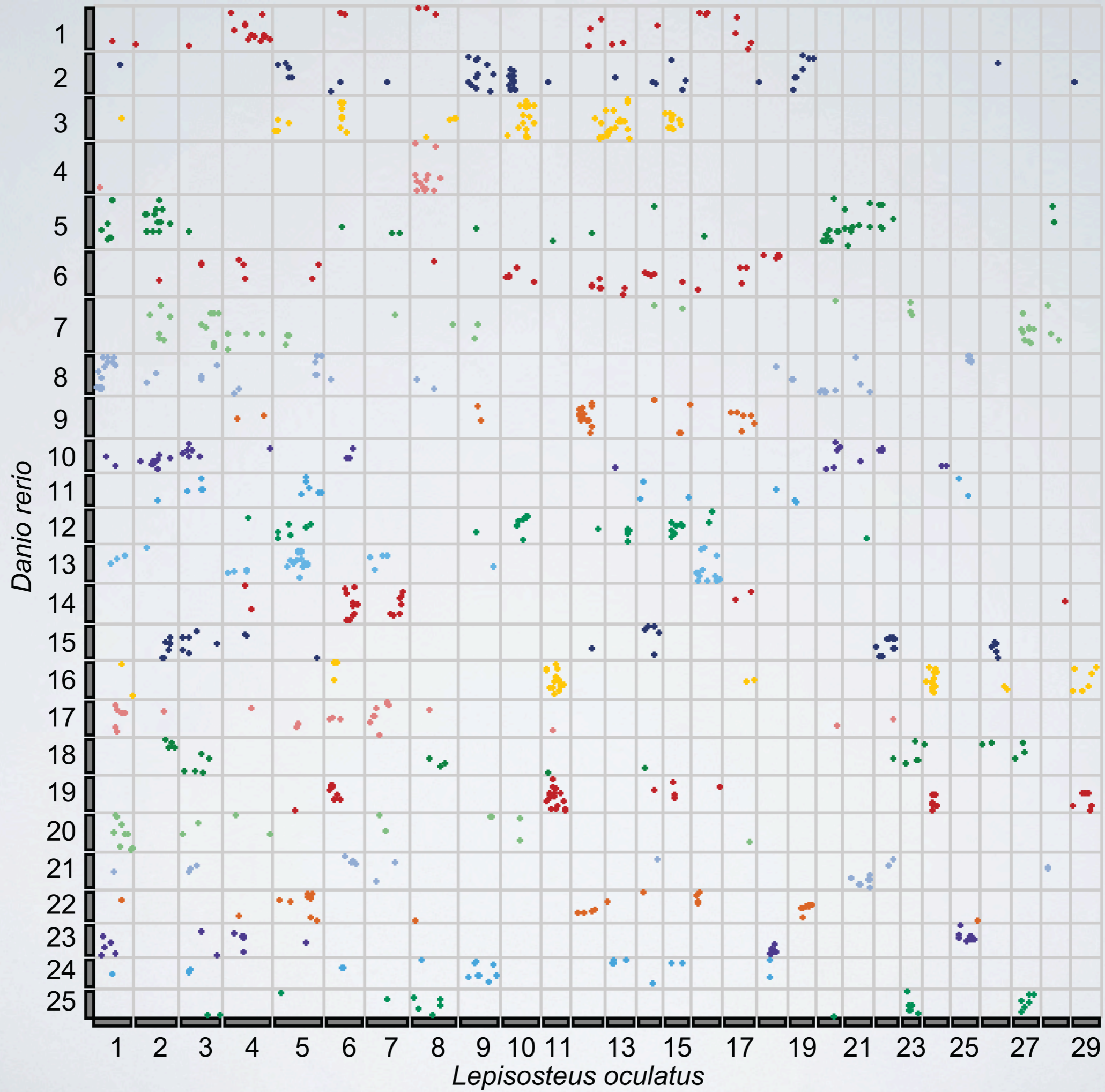
Assign
orthology to:
markers
PE contigs
ESTs

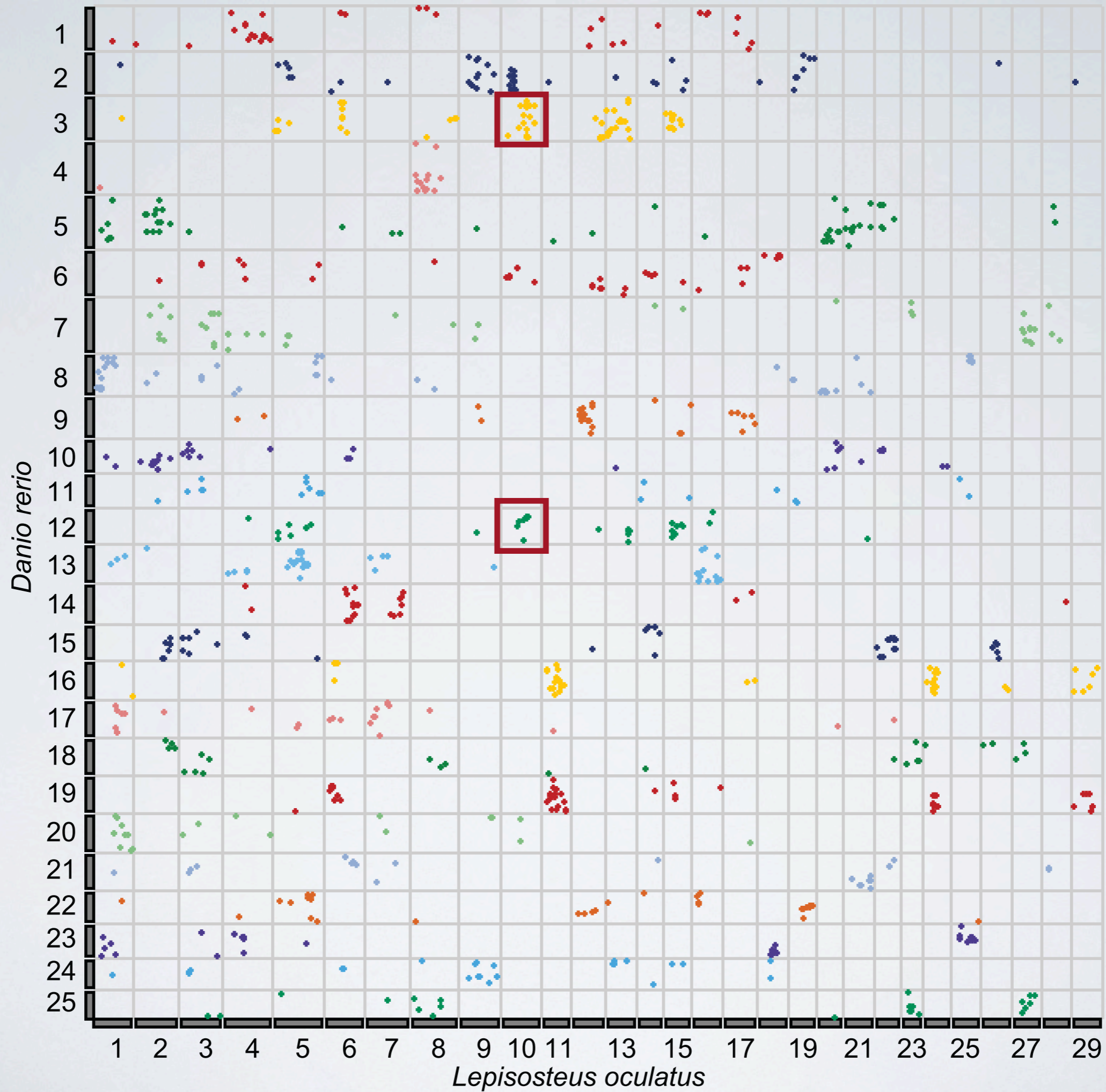
BLASTx

BLASTx

BLASTx







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

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 Python is essential**

TUTORIAL - USING STACKS



G3: Genes, Genomes, Genetics

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

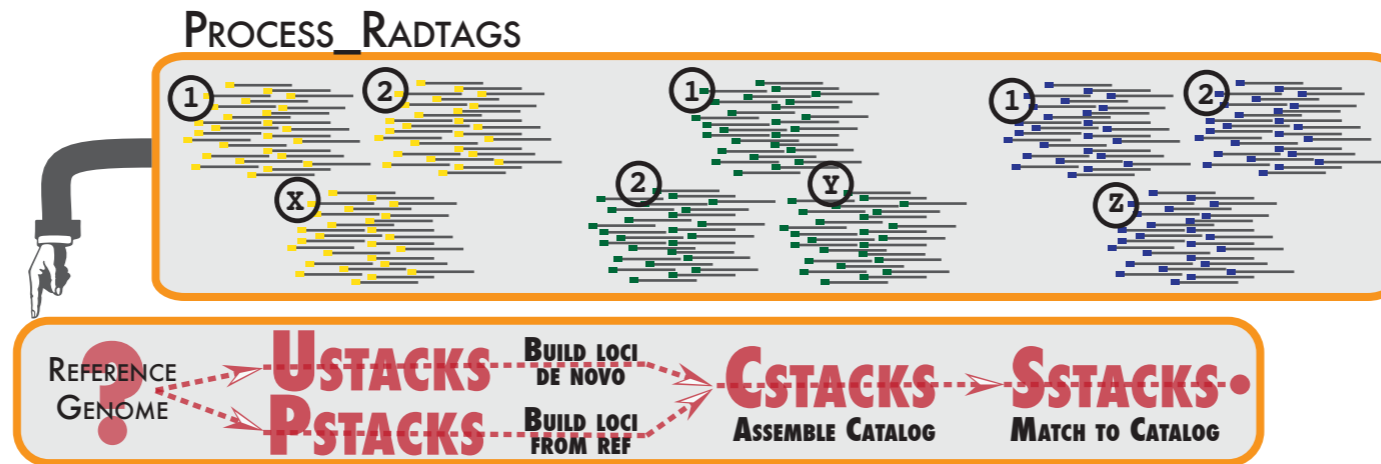
Julian M. Catchen,* Angel Amores,[†] Paul Hohenlohe,* William Cresko,* and John H. Postlethwait^{†,1}
*Center for Ecology and Evolutionary Biology and [†]Institute of Neuroscience, University of Oregon, Eugene, Oregon 97403

Stacks workflow

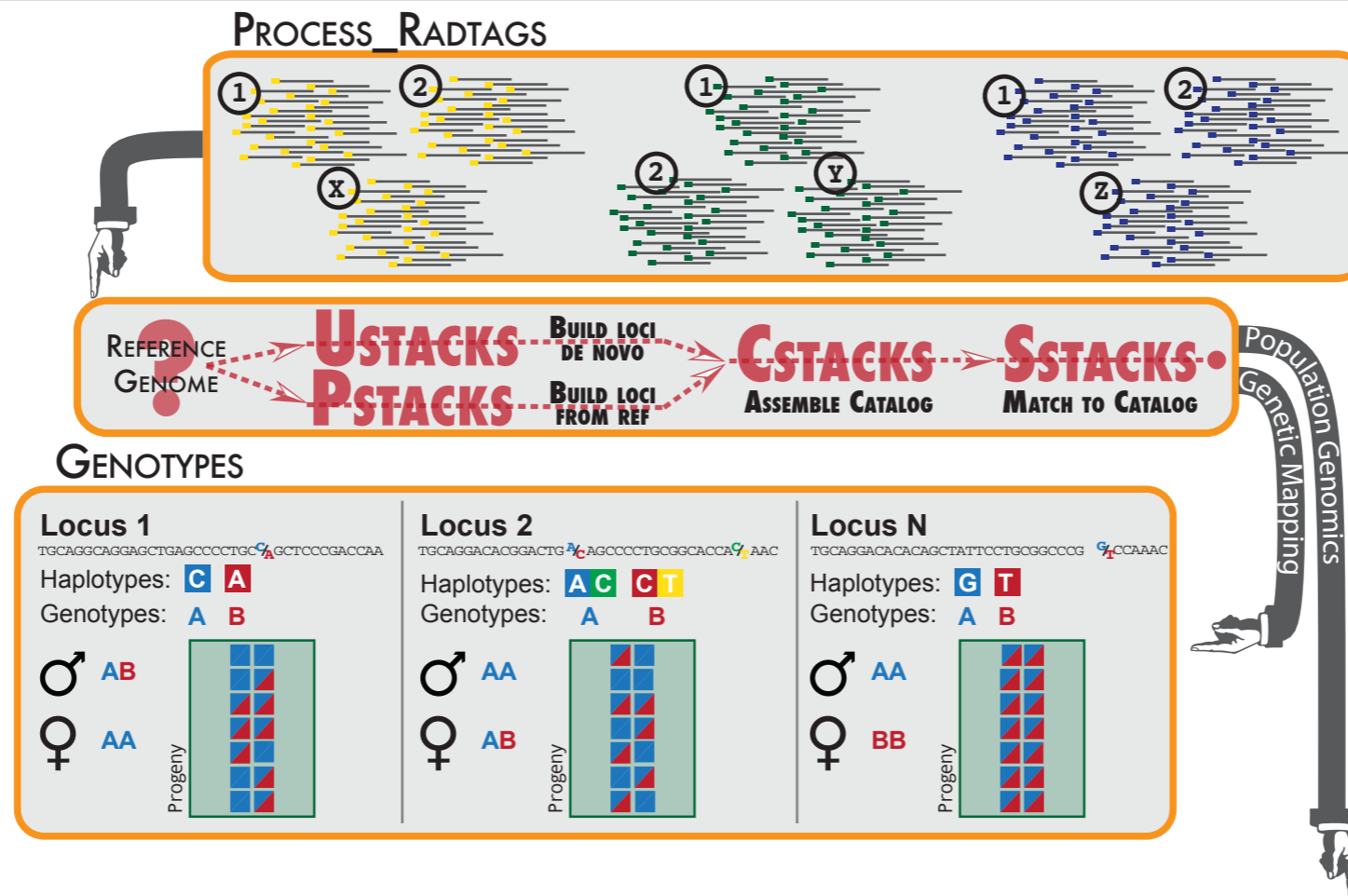
PROCESS_RADTAGS



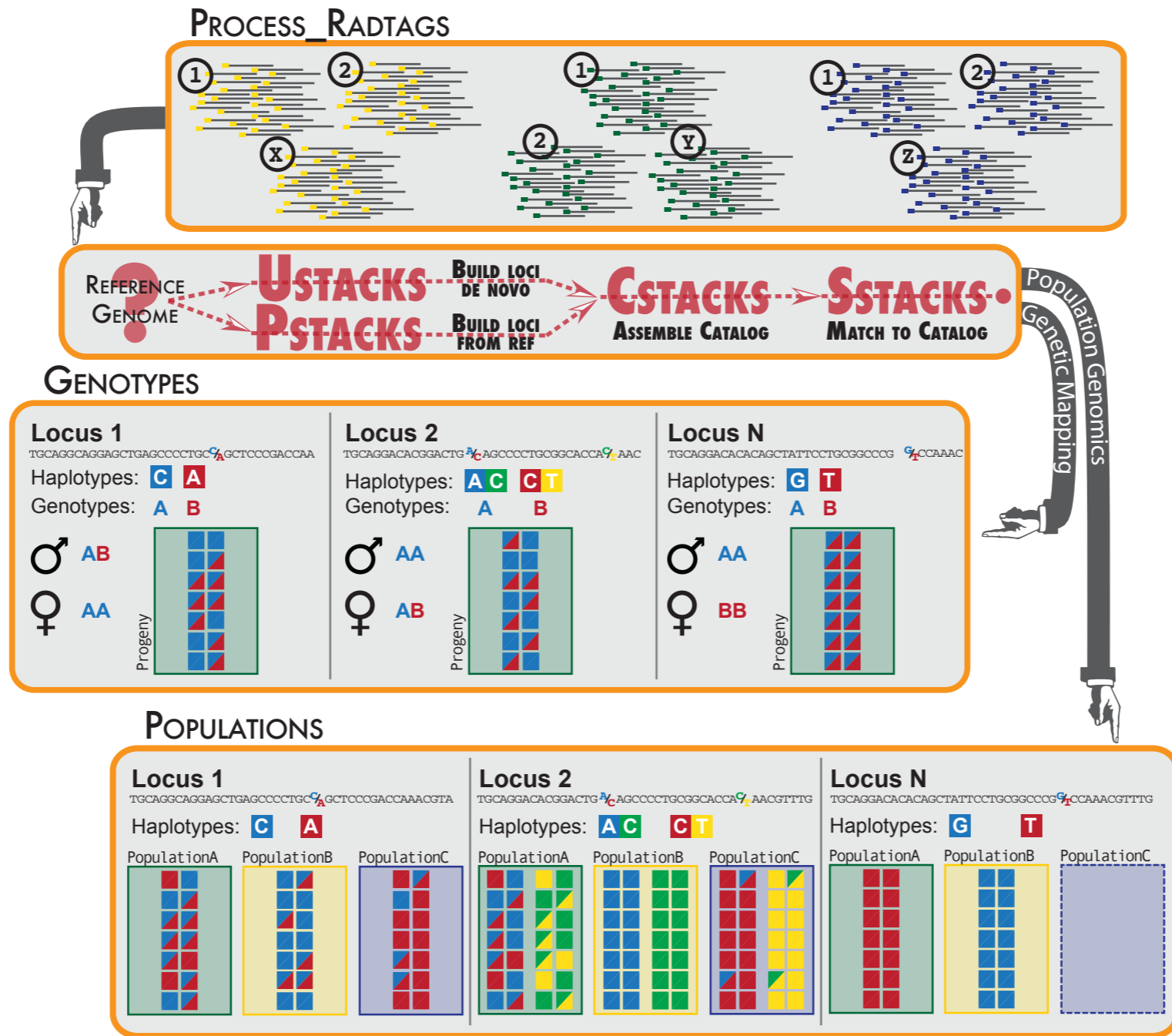
Stacks workflow



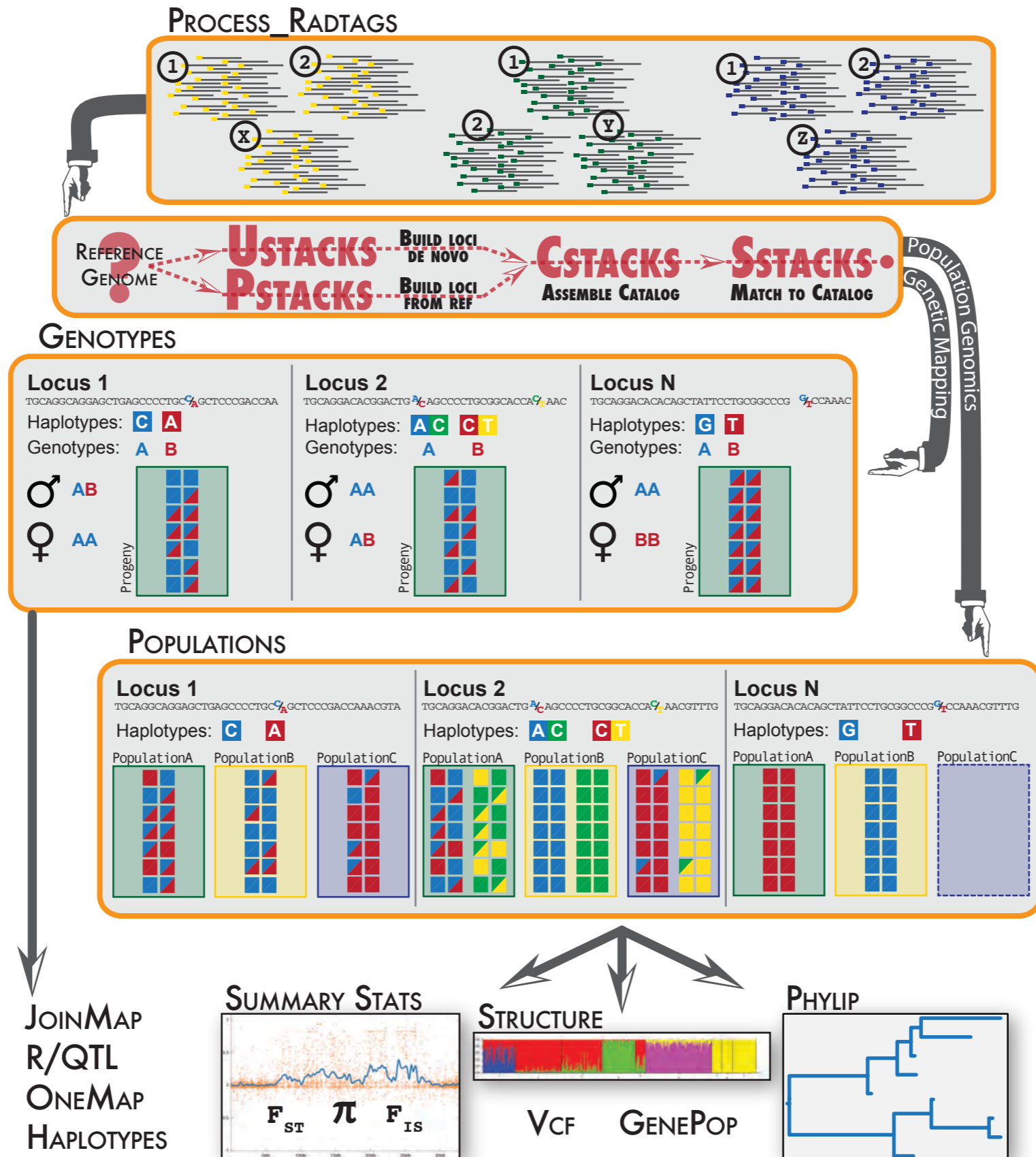
Stacks workflow



Stacks workflow



Stacks workflow



1 (1 tags)

tags per page 10

Id	SNP	Consensus	Matching Parents	Progeny	Marker	Ratio	Genotypes
~ 103 annotate	Yes [2nuc]	TGCAGGAGCCCTCCCACTCGCTGATGCCCACTCCATTCAGTGGACCGAGAGCCCAAAGCAACACTTCACATTC	2	92 / 91	ab/ac	aa: 25 (27.5%) ab: 24 (26.4%) ac: 18 (19.8%) bc: 24 (26.4%)	91

SNPs

Column: 52; G/A
Column: 70; T/G

Alleles

a : 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
Progeny 9	Progeny 10	Progeny 11	Progeny 12	Progeny 13	Progeny 14	Progeny 15	Progeny 16	Progeny 17	Progeny 18
GT	GT	GG / GT	GT / AG	GG / AG	GT / AG	GT / GG	GG / GT	GG / AG	GT
Progeny 19	Progeny 20	Progeny 21	Progeny 22	Progeny 23	Progeny 24	Progeny 25	Progeny 26	Progeny 27	Progeny 28
GT / AG	AG / GG	GT / AG	AG / GT	GG / AG	GG / AG	GT	GG / GT	GG / AG	GG / GT
Progeny 29	Progeny 31	Progeny 32	Progeny 33	Progeny 34	Progeny 35	Progeny 36	Progeny 37	Progeny 38	Progeny 39
GT / GG	GT	GT	GT	GT	GT / GG	GT	GT / AG	GT	AG / GT
Progeny 40	Progeny 41	Progeny 42	Progeny 43	Progeny 44	Progeny 45	Progeny 46	Progeny 47	Progeny 48	Progeny 49
GT	GT	GT	GT / GG	GG / GT	GT	GG / GT	GG / AG	GT	GT / GG
Progeny 50	Progeny 51	Progeny 52	Progeny 53	Progeny 54	Progeny 55	Progeny 56	Progeny 57	Progeny 58	Progeny 59
GT	GT	GT / AG	GG / GT	GT / GG	AG / GG	GT	AG / GT	GT / AG	GG / GT
Progeny 60	Progeny 61	Progeny 62	Progeny 63	Progeny 64	Progeny 65	Progeny 66	Progeny 67	Progeny 68	Progeny 70
GT / GG	GT / GG	GT / AG	GG / AG	GG / GT	GT	GT	GG / GT	GT	GG / AG
Progeny 71	Progeny 72	Progeny 73	Progeny 74	Progeny 75	Progeny 76	Progeny 77	Progeny 78	Progeny 79	Progeny 80
GG / AG	AG / GG	GT	GG / AG	GT / GG	GT	GG / AG	GG / AG	GT / GG	GT
Progeny 81	Progeny 82	Progeny 83	Progeny 84	Progeny 85	Progeny 86	Progeny 87	Progeny 88	Progeny 89	Progeny 90
GT / AG	GT / AG	GG / AG	GT	GT / GG	GT / GG	GT	GG / AG	GT	GG / AG
Progeny 91	Progeny 92	Progeny 93	Progeny 94						
AG / GG	GT / AG	AG / GG	GG / AG						

1 (1 tags)

tags per page 10

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tags per page 10

Id	SNP	Consensus	Matching Parents	Progeny	Marker	Ratio	Genotypes
~ 103 annotate	Yes [2nuc]	TGCAGGAGCCCTCCCCTCGCTGATGGCCACTCCATTTCAGTGGACCGAGAGCGCAAAGCAACACTTCACAATCC	2	92 / 91	ab/ac	aa: 25 (27.5%) ab: 24 (26.4%) ac: 18 (19.8%) bc: 24 (26.4%)	91

SNPs

Column: 52; G/A

Column: 70; T/G

Alleles

a : 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 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	Progeny 10	Progeny 11	Progeny 12	Progeny 13	Progeny 14	Progeny 15	Progeny 16	Progeny 17	Progeny 18
GT 25	GT 23	GG / GT 32 / 14	GT / AG 22 / 7	GG / AG 7 / 8	GT / AG 7 / 8	GT / GG 2 / 3	GG / GT 19 / 14	GG / AG 9 / 4	GT 15
Progeny 19	Progeny 20	Progeny 21	Progeny 22	Progeny 23	Progeny 24	Progeny 25	Progeny 26	Progeny 27	Progeny 28
GT / AG 6 / 3	AG / GG 6 / 9	GT / AG 18 / 9	AG / GT 4 / 5	GG / AG 7 / 6	GG / AG 8 / 10	GT 7	GG / GT 10 / 16	GG / AG 3 / 3	GG / GT 4 / 5
Progeny 29	Progeny 31	Progeny 32	Progeny 33	Progeny 34	Progeny 35	Progeny 36	Progeny 37	Progeny 38	Progeny 39
GT / GG 8 / 5	GT 11	GT 10	GT 17	GT 20	GT / GG 7 / 3	GT 8	GT / AG 12 / 4	GT 9	AG / GT 12 / 7
Progeny 40	Progeny 41	Progeny 42	Progeny 43	Progeny 44	Progeny 45	Progeny 46	Progeny 47	Progeny 48	Progeny 49
GT 9	GT 5	GT 9	GT / GG 9 / 12	GG / GT 3 / 6	GT 6	GG / GT 4 / 11	GG / AG 3 / 7	GT 18	GT / GG 5 / 6
Progeny 50	Progeny 51	Progeny 52	Progeny 53	Progeny 54	Progeny 55	Progeny 56	Progeny 57	Progeny 58	Progeny 59
GT 18	GT 9	GT / AG 8 / 5	GG / GT 10 / 8	GT / GG 5 / 6	AG / GG 8 / 10	GT 22	AG / GT 17 / 16	GT / AG 23 / 24	GG / GT 25 / 13
Progeny 60	Progeny 61	Progeny 62	Progeny 63	Progeny 64	Progeny 65	Progeny 66	Progeny 67	Progeny 68	Progeny 70
GT / GG 12 / 18	GT / GG 22 / 29	GT / AG 7 / 23	GG / AG 15 / 11	GG / GT 13 / 20	GT 44	GT 27	GG / GT 23 / 17	GT 30	GG / AG 14 / 13
Progeny 71	Progeny 72	Progeny 73	Progeny 74	Progeny 75	Progeny 76	Progeny 77	Progeny 78	Progeny 79	Progeny 80
GG / AG 15 / 7	AG / GG 9 / 6	GT 42	GG / AG 31 / 29	GT / GG 15 / 22	GT 41	GG / AG 14 / 17	GG / AG 25 / 17	GT / GG 29 / 14	GT 34
Progeny 81	Progeny 82	Progeny 83	Progeny 84	Progeny 85	Progeny 86	Progeny 87	Progeny 88	Progeny 89	Progeny 90
GT / AG 17 / 29	GT / AG 29 / 24	GG / AG 16 / 25	GT 41	GT / GG 14 / 24	GT / GG 6 / 4	GT 15	GG / AG 5 / 11	GT 18	GG / AG 5 / 17
Progeny 91	Progeny 92	Progeny 93	Progeny 94						
AG / GG 14 / 13	GT / AG 12 / 6	AG / GG 7 / 7	GG / AG 3 / 2						

1 (1 tags)

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SNPs

 Column: 52; G/A
 Column: 70; T/G

Alleles

 a: GT
 b: GG
 c: AG

Matching Samples

View: Haplotypes Allele Depths Genotypes

Mare	Female	Progeny 1	Progeny 2	Progeny 3	Progeny 4	Progeny 5	Progeny 6	Progeny 7	Progeny 8
GT / GG 34 / 13	AG / GT 12 / 14	GT 7 aa	AG / GG 8 / 16 bc	GG / AG 26 / 14 bc	GG / GT 15 / 11 ab	GG / AG 14 / 8 bc	AG 29 AC	GT / GG 22 / 11 ab	AG / GT 12 / 5 ac
Progeny 9 GT 25 aa	Progeny 10 GT 23 aa	Progeny 11 GG / GT 32 / 14 ab	Progeny 12 GT / AG 22 / 7 ac	Progeny 13 GG / AG 7 / 8 bc	Progeny 14 GT / AG 7 / 8 ac	Progeny 15 GT / GG 2 / 3 ab	Progeny 16 GG / GT 19 / 14 ab	Progeny 17 GG / AG 9 / 4 bc	Progeny 18 GT 15 aa
Progeny 19 GT / AG 6 / 3 ac	Progeny 20 AG / GG 6 / 9 bc	Progeny 21 GT / AG 18 / 9 ac	Progeny 22 AG / GT 4 / 5 ac	Progeny 23 GG / AG 7 / 6 bc	Progeny 24 GG / AG 8 / 10 bc	Progeny 25 GT 7 AC	Progeny 26 GG / GT 10 / 16 ab	Progeny 27 GG / AG 3 / 3 bc	Progeny 28 GG / GT 4 / 5 ab
Progeny 29 GT / GG 8 / 5 ab	Progeny 31 GT 11 aa	Progeny 32 GT 10 aa	Progeny 33 GT 17 aa	Progeny 34 GT 20 aa	Progeny 35 GT / GG 7 / 3 ab	Progeny 36 GT 8 aa	Progeny 37 GT / AG 12 / 4 ac	Progeny 38 GT 9 aa	Progeny 39 AG / GT 12 / 7 ac
Progeny 40 GT 9 aa	Progeny 41 GT 5 aa	Progeny 42 GT 9 aa	Progeny 43 GT / GG 9 / 12 ab	Progeny 44 GG / GT 3 / 6 ab	Progeny 45 GT 6 AC	Progeny 46 GG / GT 4 / 11 ab	Progeny 47 GG / AG 3 / 7 bc	Progeny 48 GT 18 aa	Progeny 49 GT / GG 5 / 6 ab
Progeny 50 GT 18 aa	Progeny 51 GT 9 aa	Progeny 52 GT / AG 8 / 5 ac	Progeny 53 GG / GT 10 / 8 ab	Progeny 54 GT / GG 5 / 6 ab	Progeny 55 AG / GG 8 / 10 bc	Progeny 56 GT 22 aa	Progeny 57 AG / GT 17 / 16 ac	Progeny 58 GT / AG 23 / 24 ac	Progeny 59 GG / GT 25 / 13 ab
Progeny 60 GT / GG 12 / 18 ab	Progeny 61 GT / GG 22 / 29 ab	Progeny 62 GT / AG 7 / 23 ac	Progeny 63 GG / AG 15 / 11 bc	Progeny 64 GG / GT 13 / 20 ab	Progeny 65 GT 44 aa	Progeny 66 GT 27 aa	Progeny 67 GG / GT 23 / 17 ab	Progeny 68 GT 30 aa	Progeny 70 GG / AG 14 / 13 bc
Progeny 71 GG / AG 15 / 7 bc	Progeny 72 AG / GG 9 / 6 bc	Progeny 73 GT 42 aa	Progeny 74 GG / AG 31 / 29 bc	Progeny 75 GT / GG 15 / 22 ab	Progeny 76 GT 41 aa	Progeny 77 GG / AG 14 / 17 bc	Progeny 78 GG / AG 25 / 17 bc	Progeny 79 GT / GG 29 / 14 ab	Progeny 80 GT 34 aa
Progeny 81 GT / AG 17 / 20	Progeny 82 GT / AG 20 / 24	Progeny 83 GG / AG 16 / 25	Progeny 84 GT 41	Progeny 85 GT / GG 14 / 24	Progeny 86 GT / GG 6 / 4	Progeny 87 GT 15	Progeny 88 GG / AG 5 / 11	Progeny 89 GT 18	Progeny 90 GG / AG 5 / 17

Stacks

version 0.998

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 Column: 70	G/A T/G	AG GT	46.15% 53.85%	False False False

Relationship	Seq ID	Sequence
consensus model		TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
1 primary	CAGTC_2_0018_768_1365_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
2 primary	CAGTC_2_0029_1628_1751_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
3 primary	CAGTC_2_0053_1692_1388_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
4 primary	CAGTC_2_0058_1588_1038_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
5 primary	CAGTC_2_0059_1524_1186_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
6 primary	CAGTC_2_0094_1356_1854_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
7 primary	CAGTC_2_0096_1791_1246_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
8 primary	CAGTC_2_0021_877_296_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
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10 primary	CAGTC_2_0025_108_810_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
11 primary	CAGTC_2_0039_1252_1764_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
12 primary	CAGTC_2_0061_596_159_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
13 primary	CAGTC_2_0068_1310_997_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
14 primary	CAGTC_2_0070_644_2040_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
15 primary	CAGTC_2_0074_328_659_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
16 primary	CAGTC_2_0075_1668_1862_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
17 primary	CAGTC_2_0079_1481_505_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
18 primary	CAGTC_2_0084_805_1974_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
19 primary	CAGTC_2_0100_481_1043_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
20 secondary	CAGTC_2_0014_728_1008_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
21 secondary	CAGTC_2_0016_86_1022_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
22 secondary	CAGTC_2_0042_426_1001_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
23 secondary	CAGTC_2_0052_867_1387_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
24 secondary	CAGTC_2_0012_221_1043_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
25 secondary	CAGTC_2_0095_120_1067_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC
26 secondary	CAGTC_2_0077_1003_356_1[35245]	TGCAGGAGCCCTCCCACTCGCTGATGGCCACTCCATTCAGTGGACCGAGAGCGCAAAGCAACACTTCACATGCC