

# The Resurgence of Reference Quality Genomes

Michael Schatz

April 9, 2015

UMN-MSI: Advances in Genome Assembly



# Outline

1. Assembly Fundamentals
2. PacBio Sequencing of Rice
3. Oxford Nanopore Sequencing of Yeast



# Outline

1. ~~Assembly Fundamentals~~

Thanks Jason!

2. PacBio Sequencing of Rice  
and Human Cancer

3. Oxford Nanopore Sequencing of Yeast



## ARTICLES

# The map-based sequence of the rice genome

International Rice Genome Sequencing Project\*

Rice, one of the world's most important food plants, has important genetic relationships with the other cereal species and is a model plant for studying the evolution of the grasses. The rice genome is 389 Mb in size, including 389 Mb of sequenced DNA and 389 Mb of transposable-element DNA. The rice genome is similar in size to the *Arabidopsis* genome. In a reciprocal genome comparison, we identified 20,000 genes in the rice proteome. Twenty-nine classes of transposable elements were identified in the rice genome. The maize and sorghum genomes are similar in size to the rice genome. The nuclear chromosomes of rice are similar in size to the rice genome. The additional sequences in the rice genome are likely to be involved in traits. The additional sequences in the rice genome are likely to accelerate improvement of rice.

**Table 2 | Size of each chromosome based on sequence data and estimated gaps**

Chr	Sequenced bases (bp)	Gaps on arm regions No.	Length (Mb)	Telomeric gaps* (Mb)	Centromeric gaps† (Mb)	rDNA‡ (Mb)	Total (Mb)	Coverage§ (%)
1	43,260,640	5	0.33	0.06	1.40		45.05	99.1
2	35,954,074	3	0.10	0.01	0.72		36.78	99.7
3	36,189,985	4	0.96	0.04	0.18		37.37	97.3
4	35,489,479	3	0.46	0.20			36.15	98.7
5	29,733,216	6	0.22	0.05			30.00	99.3
6	30,731,386	1	0.02	0.03	0.82		31.60	99.8
7	29,643,843	1	0.31	0.01	0.32		30.28	98.9
8	28,434,680	1	0.09	0.05			28.57	99.7
9	22,692,709	4	0.13	0.14	0.62	6.95	30.53	98.8
10	22,683,701	4	0.68	0.13	0.47		23.96	96.6
11	28,357,783	4	0.21	0.04	1.90	0.25	30.76	99.1
12	27,561,960	0	0.00	0.05	0.16		27.77	99.8
All	370,733,456	36	3.51	0.81	6.59	7.20	388.82	98.9

Contig N50: 5.1Mbp  
Total projects costs: >\$100M

## Initial Assembly Attempts with early Illumina sequencers circa 2007-2008

(older Illumina PE76 library with small insert size ~150bp)

Assembler	Data set	N50 contig size	Max contig size	Total assembly size
Velvet	25X Nipponbare	1049bp	21833bp	325.8 Mbp
Velvet	50X Nipponbare	411bp	23095bp	401.6 Mbp
Abyss	25X Nipponbare	1853bp	12688bp	288.4 Mbp
Abyss	50X Nipponbare	2847bp	34893bp	317.4 Mbp

Total costs: ~\$10k  
>1,000x times cheaper, but at what cost scientifically?

W.R. McCombie

# Genomics Arsenal in the year 2015

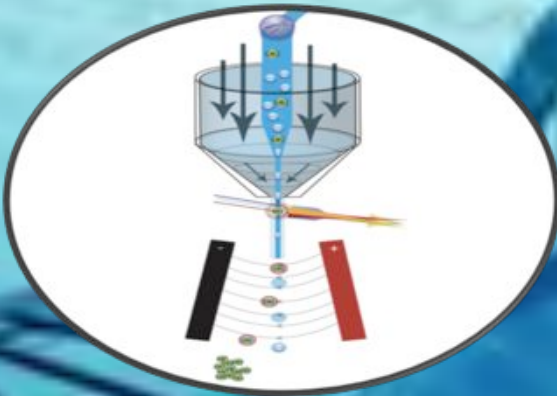
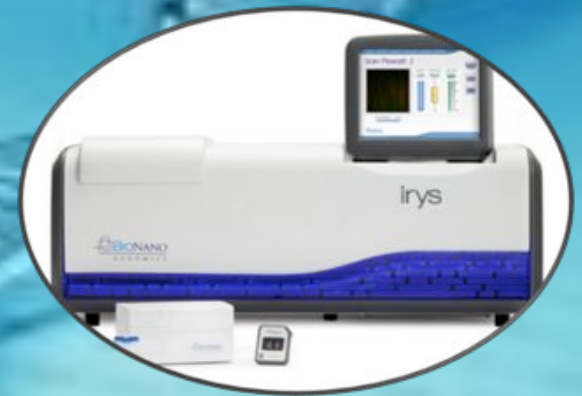
## Sample Preparation



## Sequencing

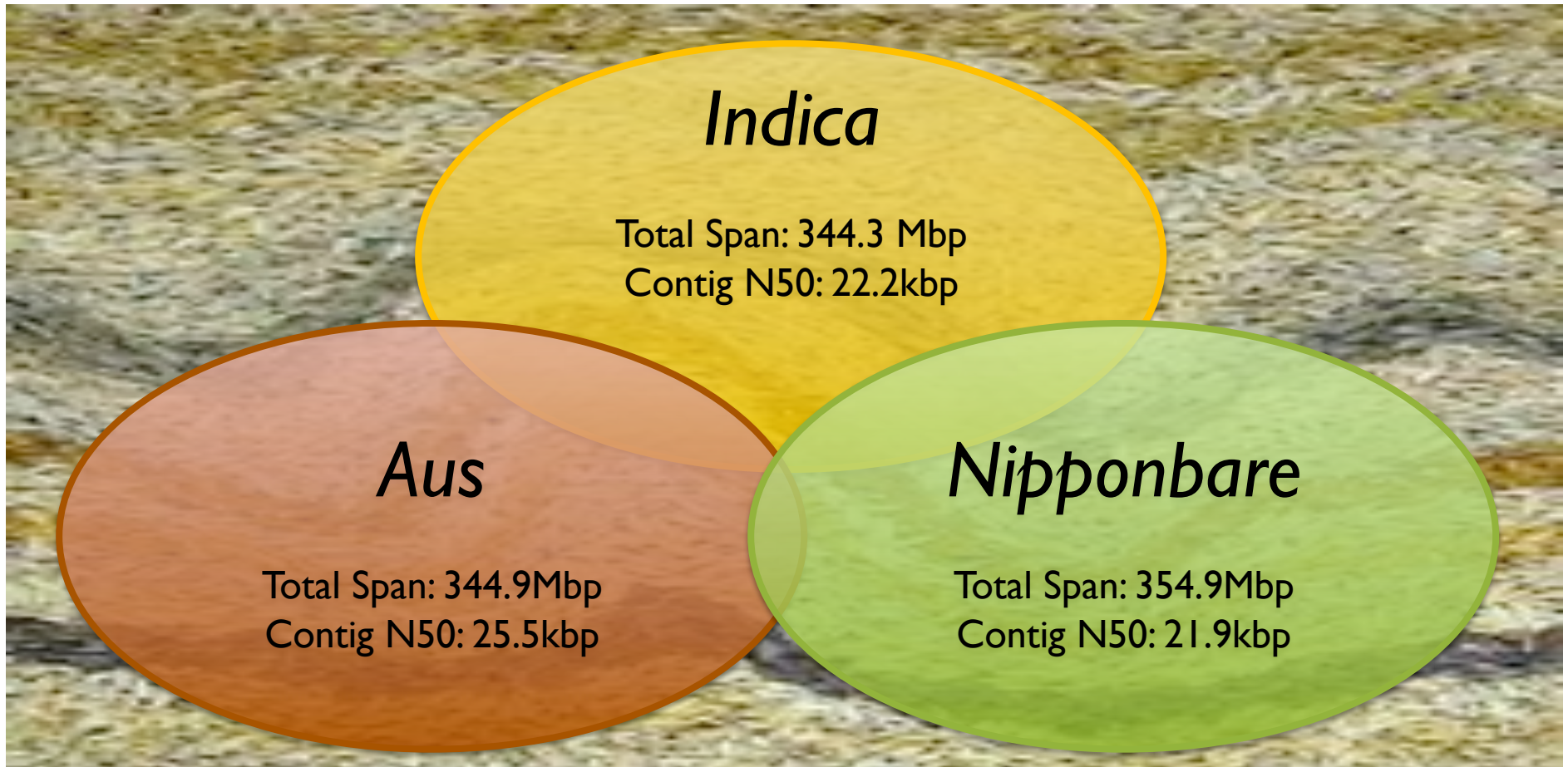


## Chromosome Mapping





# Population structure of *Oryza sativa*

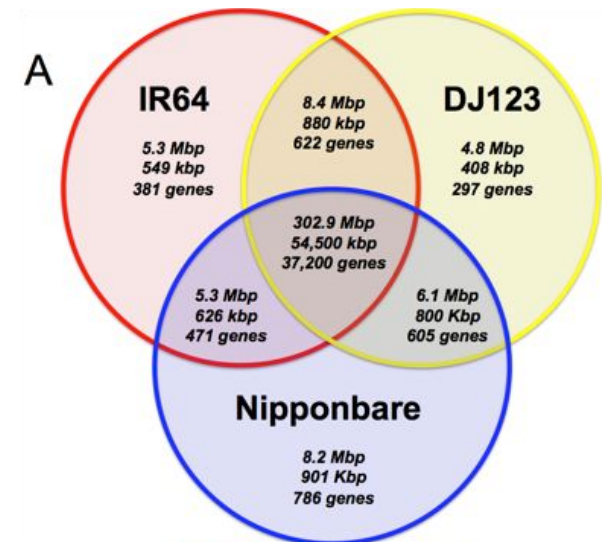


**Whole genome de novo assemblies of three divergent strains of rice (*O. sativa*) documents novel gene space of *aus* and *indica***

Schatz, Maron, Stein et al (2014) *Genome Biology*. 15:506 doi:10.1186/s13059-014-0506-z

# *Oryza sativa* Gene Diversity

- Very high quality representation of the “gene-space”
  - Overall identity ~99.9%
  - Less than 1% of exonic bases missing
- Genome-specific genes enriched for disease resistance
  - Reflects their geographic and environmental diversity
- Assemblies fragmented at (high copy) repeats
  - Difficult to identify full length gene models and regulatory features



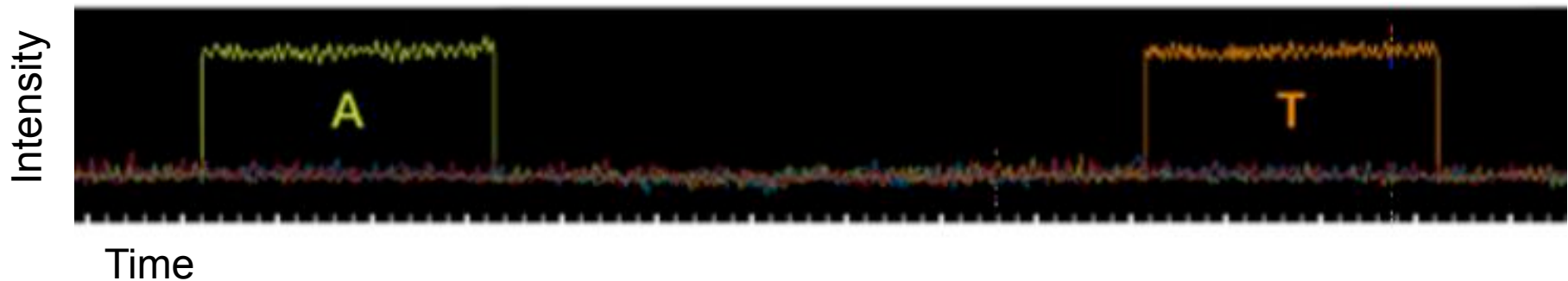
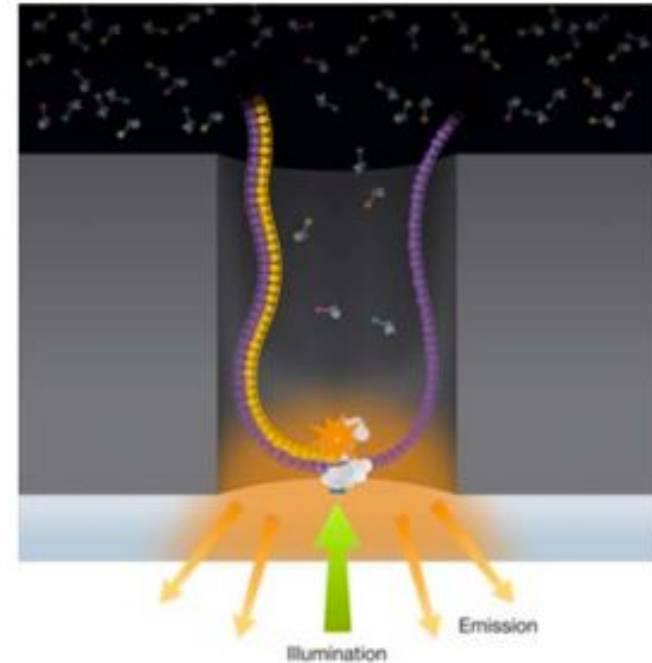
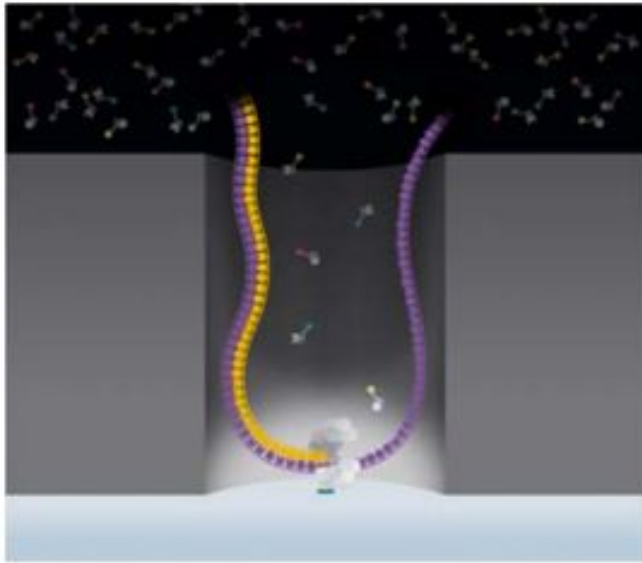
## Overall sequence content

In each sector, the top number is the total number of base pairs, the middle number is the number of exonic bases, and the bottom is the gene count. If a gene is partially shared, it is assigned to the sector with the most exonic bases.

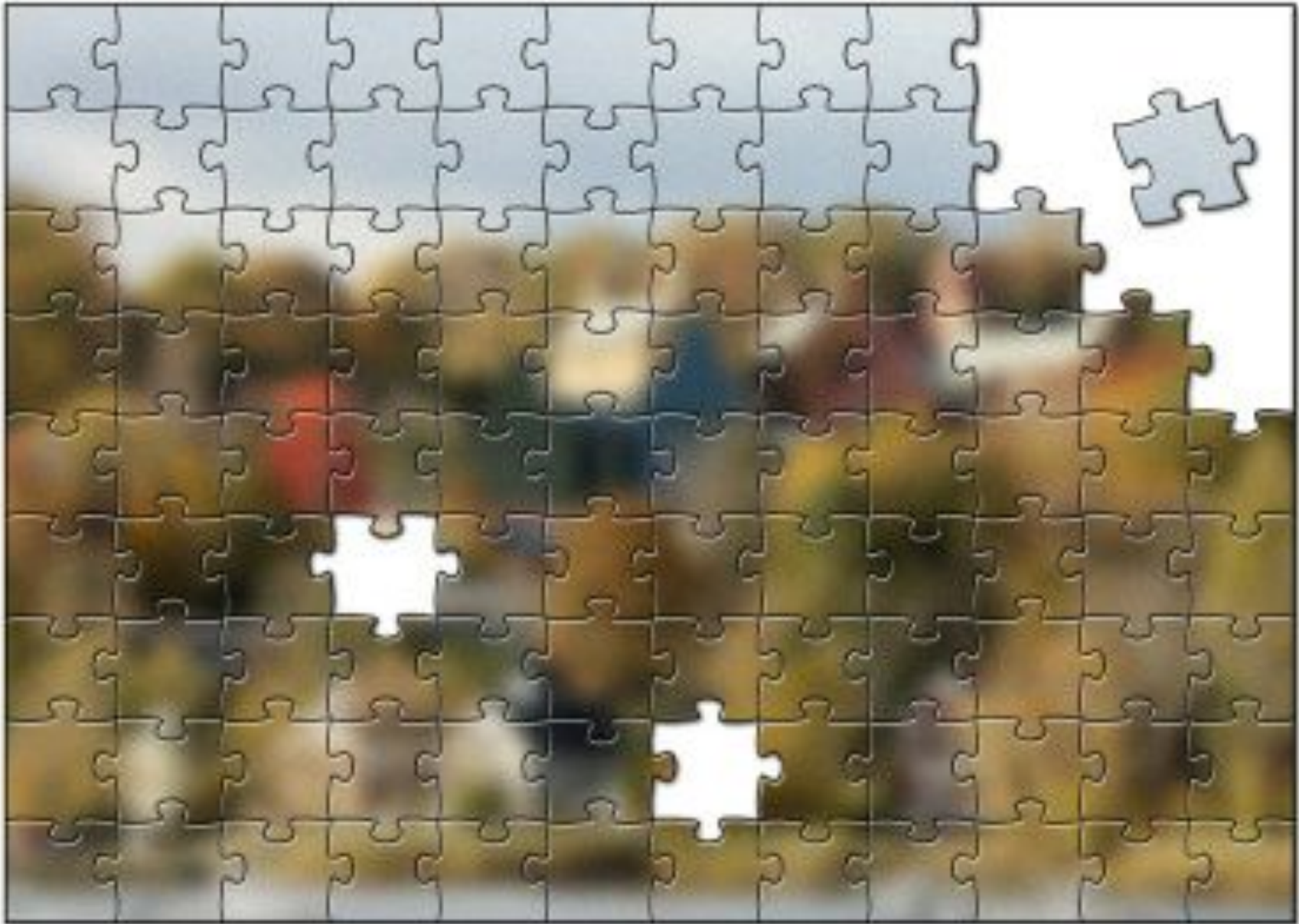


# PacBio SMRT Sequencing

Imaging of fluorescently phospholinked labeled nucleotides as they are incorporated by a polymerase anchored to a Zero-Mode Waveguide (ZMW).

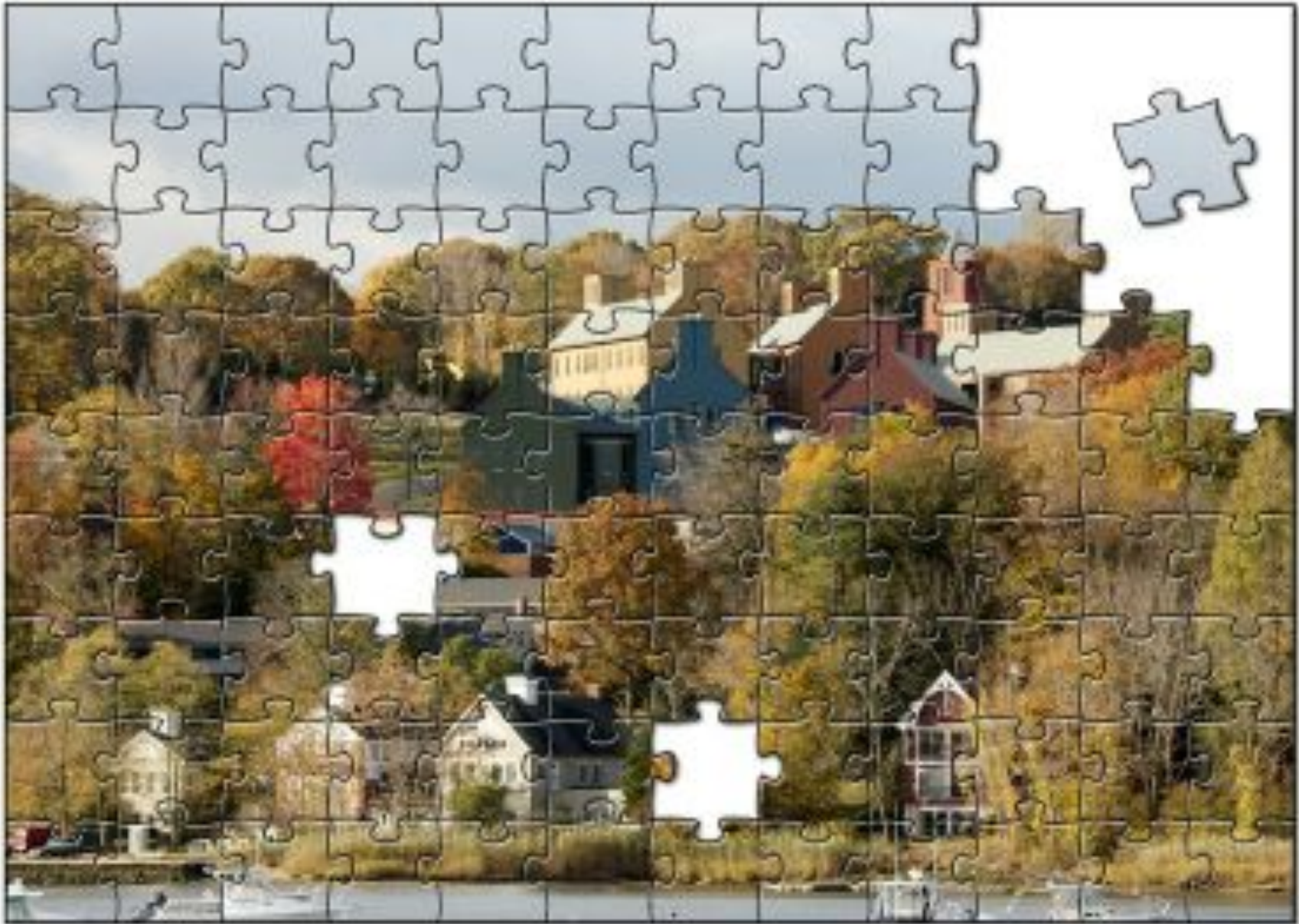


# Single Molecule Sequences

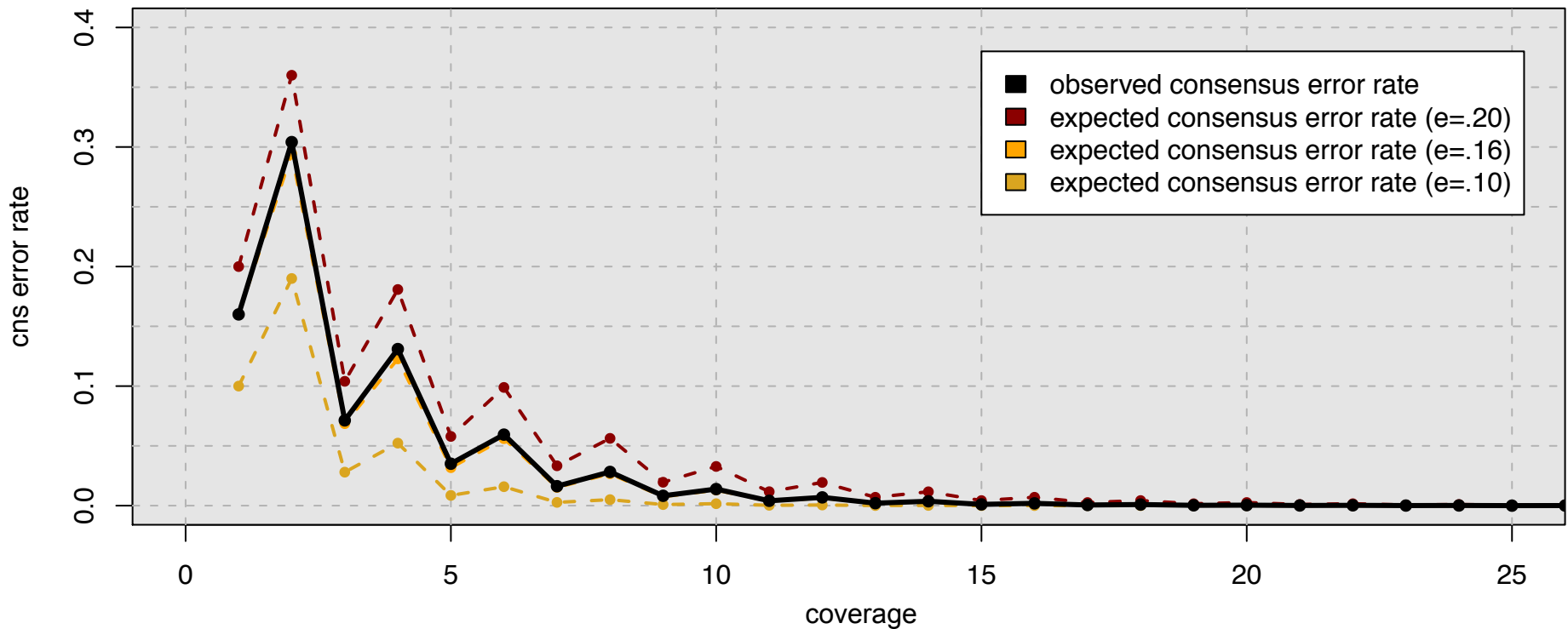




# “Corrective Lens” for Sequencing



# Consensus Accuracy and Coverage



## Coverage can overcome random errors

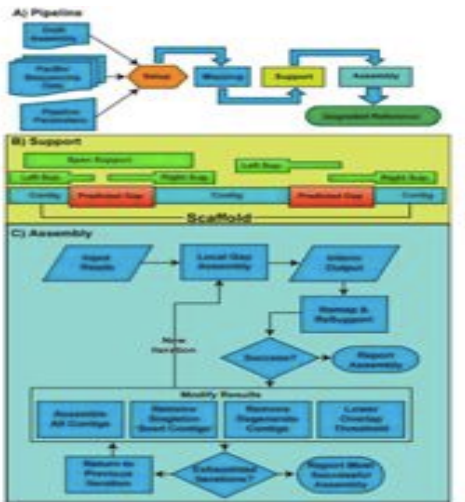
- Dashed: error model from binomial sampling
- Solid: observed accuracy

Koren, Schatz, et al (2012)  
*Nature Biotechnology*. 30:693–700

$$CNS\ Error = \sum_{i=\lfloor c/2 \rfloor}^c \binom{c}{i} (e)^i (1-e)^{n-i}$$

# PacBio Assembly Algorithms

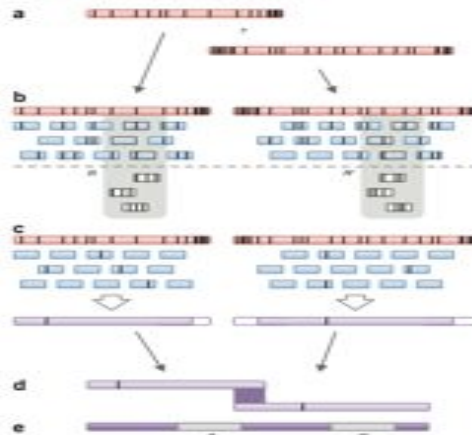
## PBJelly



**Gap Filling  
and Assembly Upgrade**

English *et al* (2012)  
*PLOS One*. 7(11): e47768

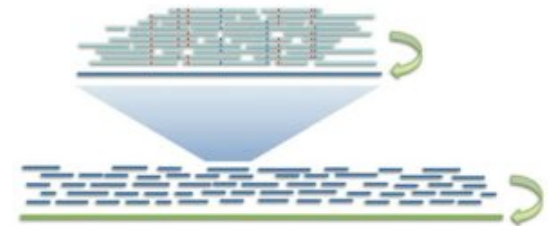
## PacBioToCA & ECTools



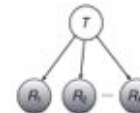
**Hybrid/PB-only Error  
Correction**

Koren, Schatz, *et al* (2012)  
*Nature Biotechnology*. 30:693–700

## HGAP & Quiver



$$\Pr(\mathbf{R} | T) = \prod_k \Pr(R_k | T)$$



Quiver Performance Results Comparison to Reference Genome ( <i>M. ruber</i> ; 3.1 MB; SMRT® Cells)		
	Initial Assembly	Quiver Consensus
QV	43.4	54.5
Accuracy	99.99540%	99.99964%
Differences	141	11

**PB-only Correction &  
Polishing**

Chin *et al* (2013)  
*Nature Methods*. 10:563–569

< 5x

PacBio Coverage

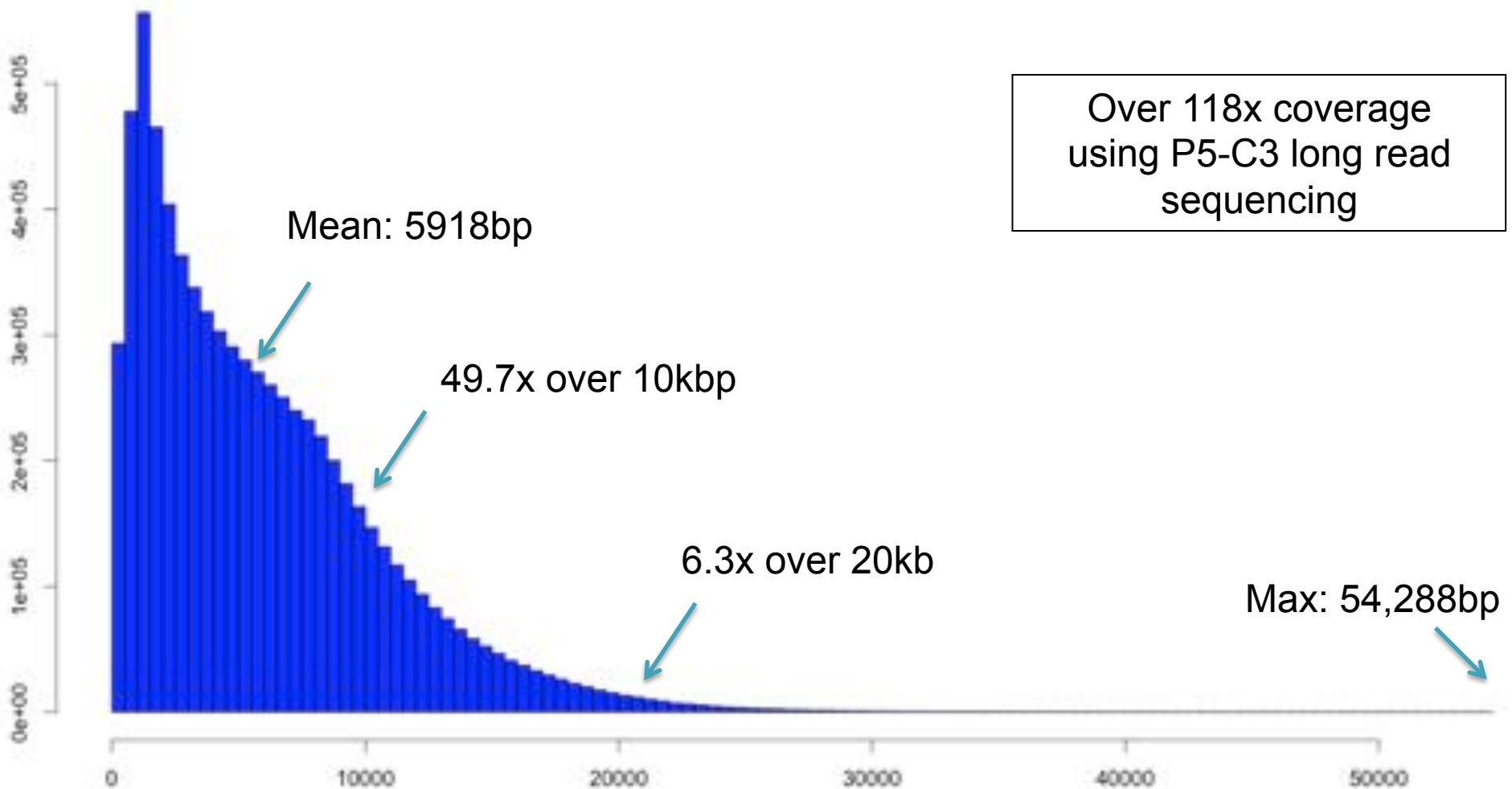
> 50x



# O. sativa pv Indica (IR64)

PacBio RS II sequencing at PacBio

- Size selection using an 10 Kb elution window on a BluePippin™ device from Sage Science

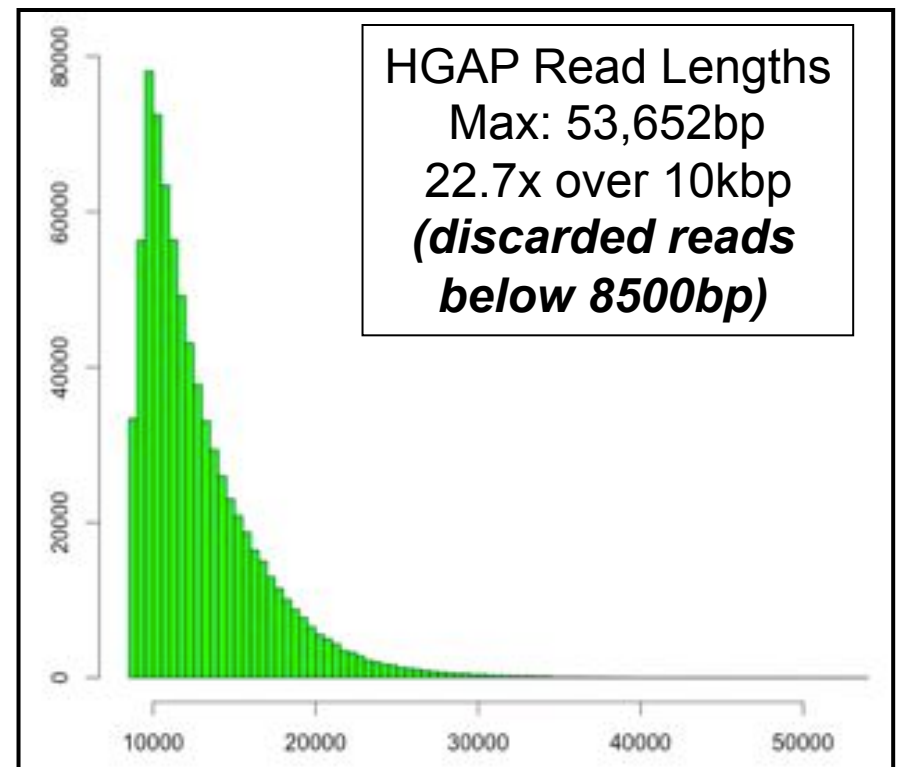


# O. sativa pv Indica (IR64)

Genome size: ~370 Mb  
Chromosome N50: ~29.7 Mbp



Assembly	Contig NG50
MiSeq Fragments 25x 456bp (3 runs 2x300 @ 450 FLASH)	19 kbp
“ALLPATHS-recipe” 50x 2x100bp @ 180 36x 2x50bp @ 2100 51x 2x50bp @ 4800	18 kbp
HGAP + CA 22.7x @ 10kbp	4.0 Mbp
Nipponbare BAC-by-BAC Assembly	5.1 Mbp



# S5 Hybrid Sterility Locus



Sanger	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...
Illumina	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...
PacBio	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...

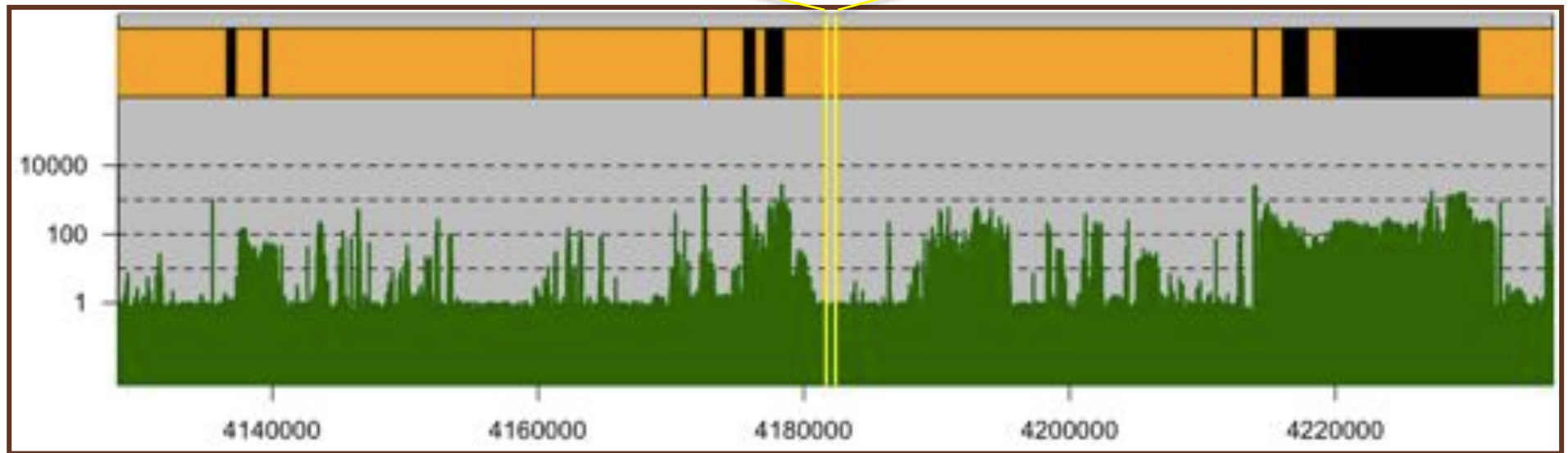
## ***S5 is a major locus for hybrid sterility in rice that affects embryo sac fertility.***

- Genetic analysis of the S5 locus documented three alleles: an indica (S5-i), a japonica (S5-j), and a neutral allele (S5-n)
- Hybrids of genotype S5-i/S5-j are mostly sterile, whereas hybrids of genotypes consisting of S5-n with either S5-i or S5-j are mostly fertile.
- Contains three tightly linked genes that work together in a 'killer-protector'-type system: ORF3, ORF4, ORF5
- The ORF5 indica (ORF5+) and japonica (ORF5-) alleles differ by only **two nucleotides**

# S5 Hybrid Sterility Locus



Sanger	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...
Illumina	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...
PacBio	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...

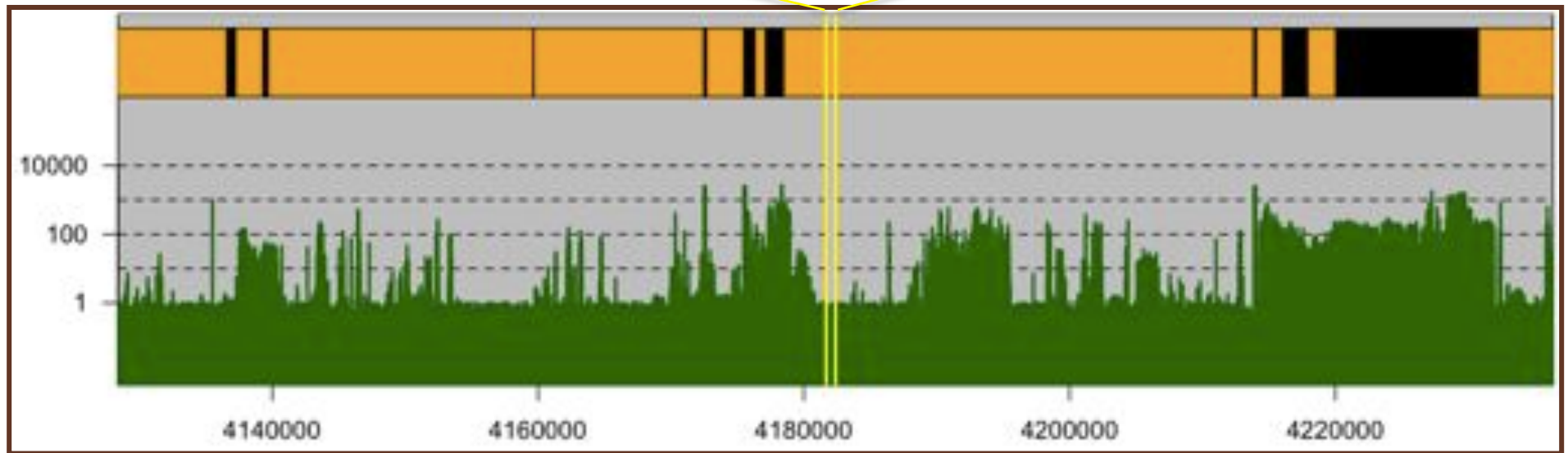


100kbp

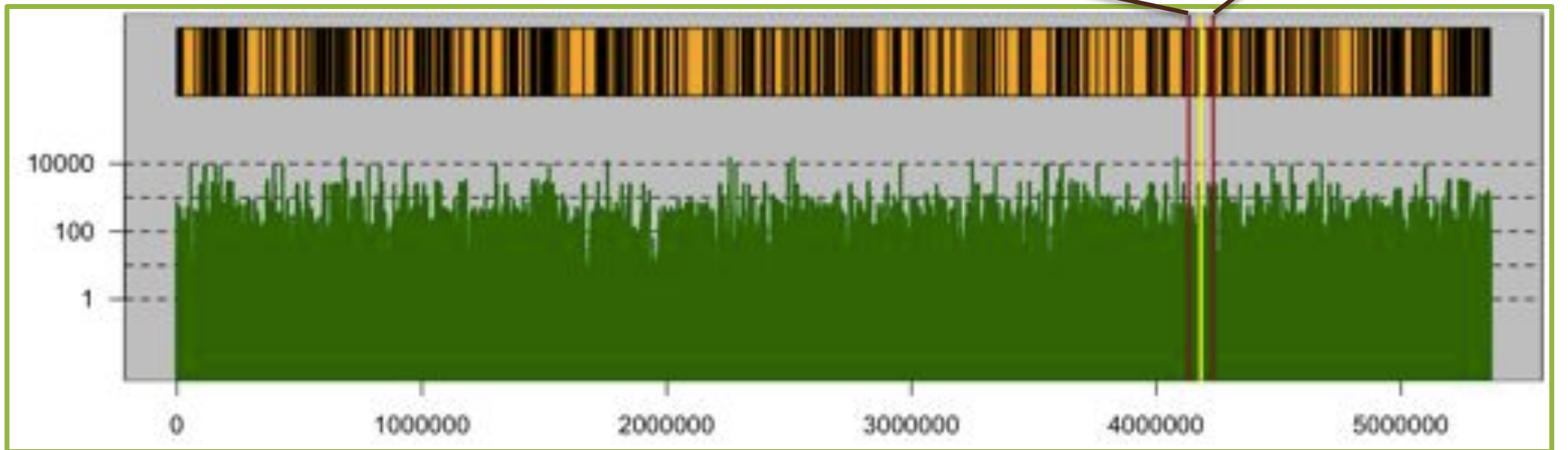
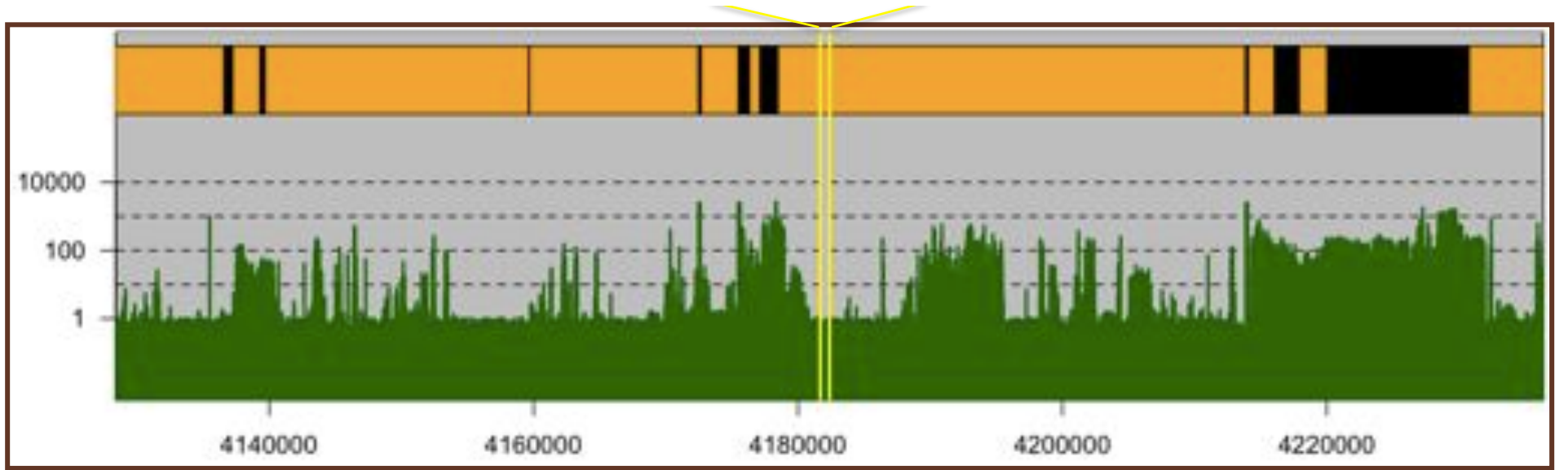
# S5 Hybrid Sterility Locus

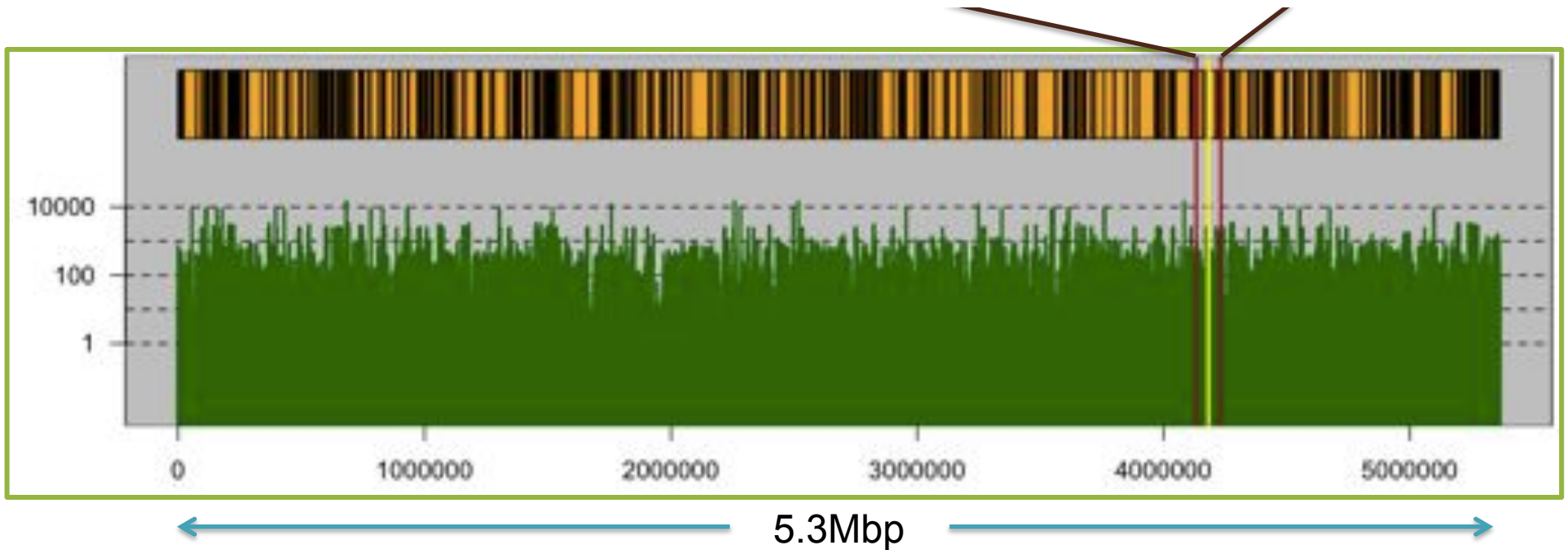


Sanger	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...
Illumina	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...
PacBio	...ACCCTGATATTCTGAGTTACAAGGCATT <b>C</b> AGCTACTGCTTGCCCACTGACGAGACC...





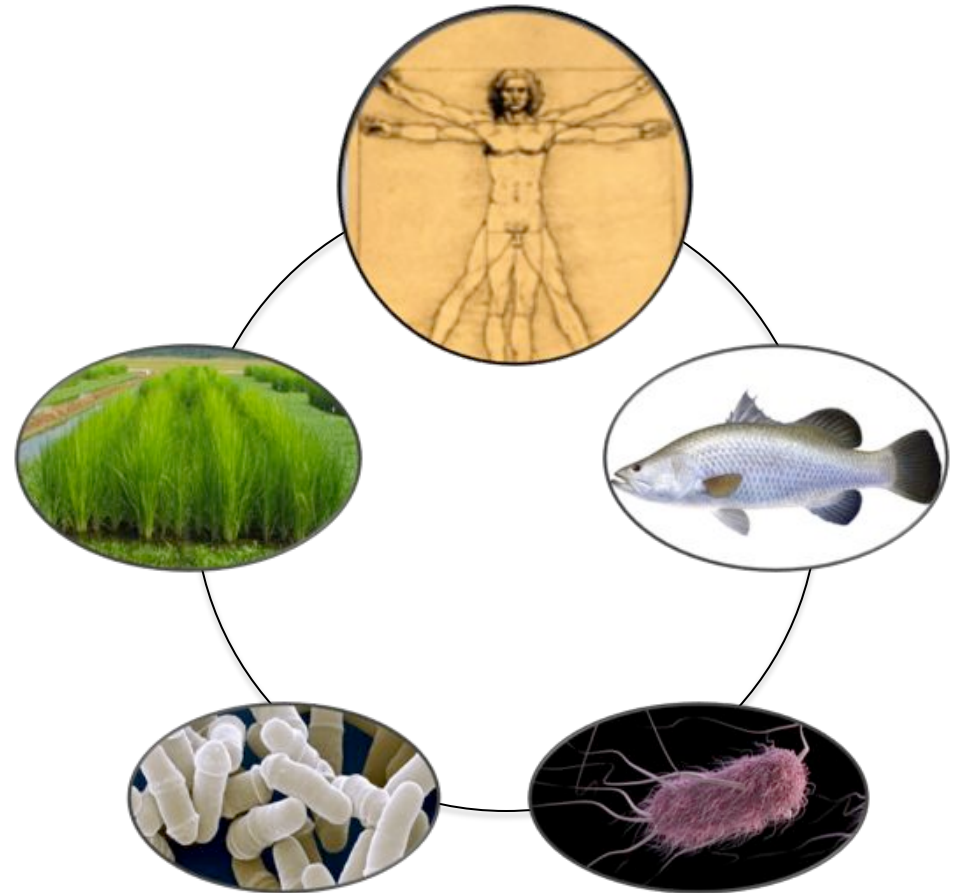
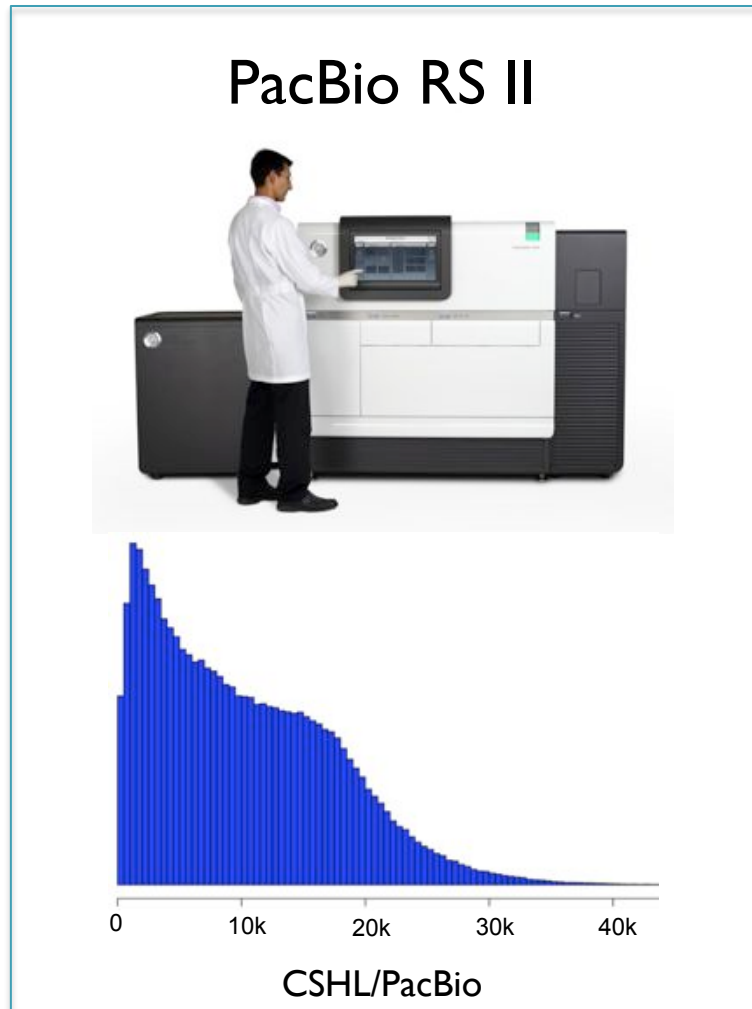




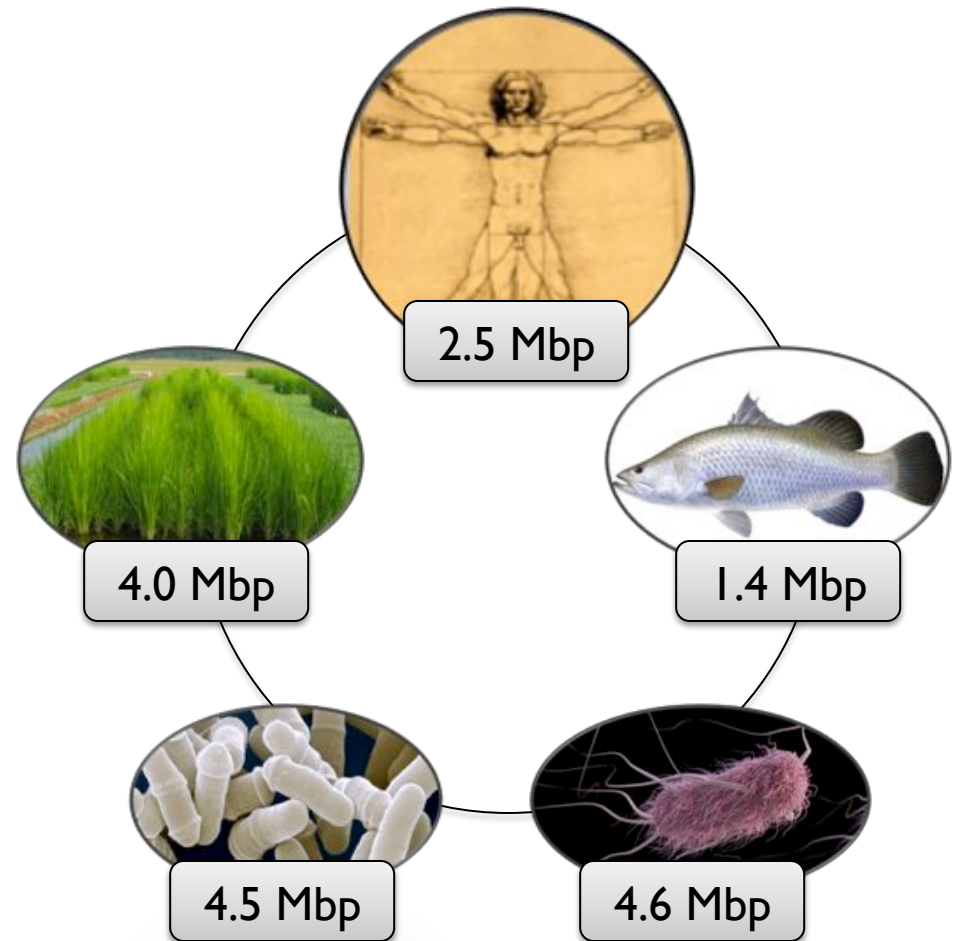
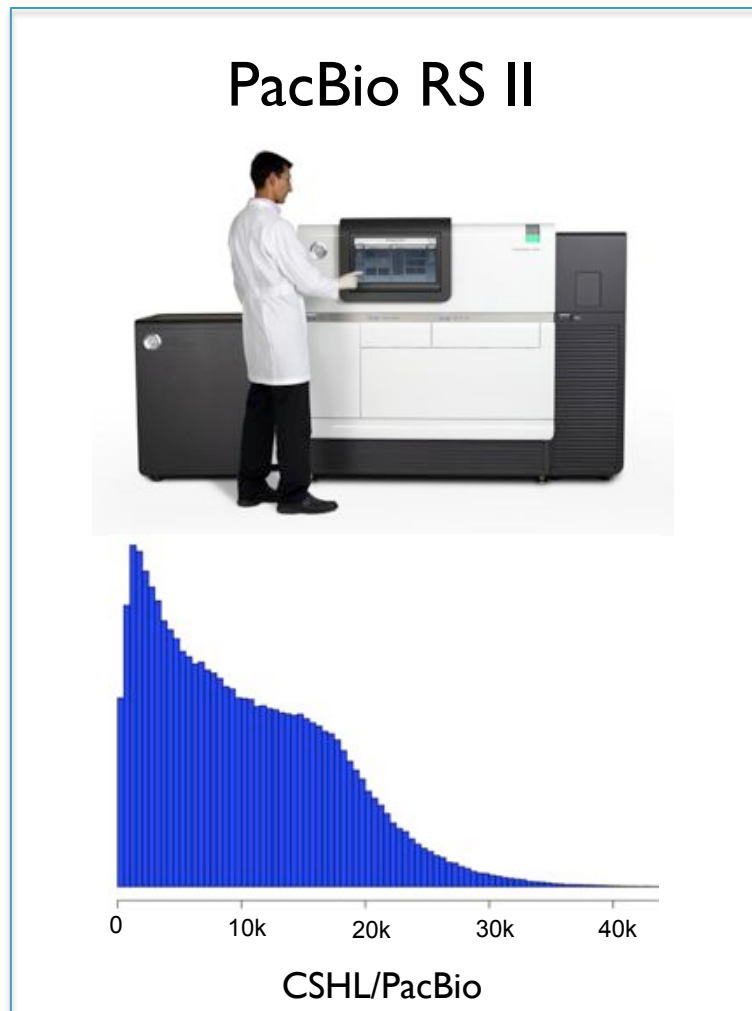
### ***Improvements from 20kbp to 4Mbp contig N50:***

- Over 20 Megabases of additional sequence
  - Extremely high sequence identity (>99.9%)
  - Thousands of gaps filled, hundreds of mis-assemblies corrected
- Complete gene models, promoter regions for nearly every gene
  - True representation of transposons and other complex features
- Opportunities for studying large scale chromosome evolution
  - Largest contigs approach complete chromosome arms

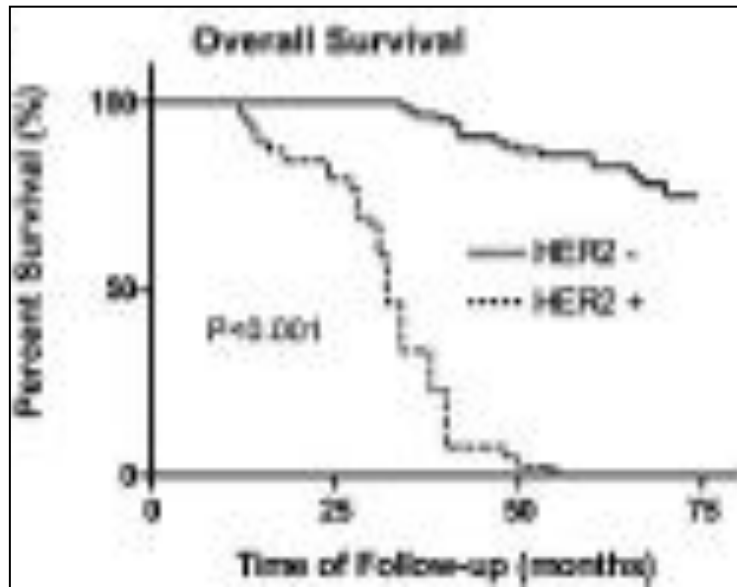
# Current Collaborations



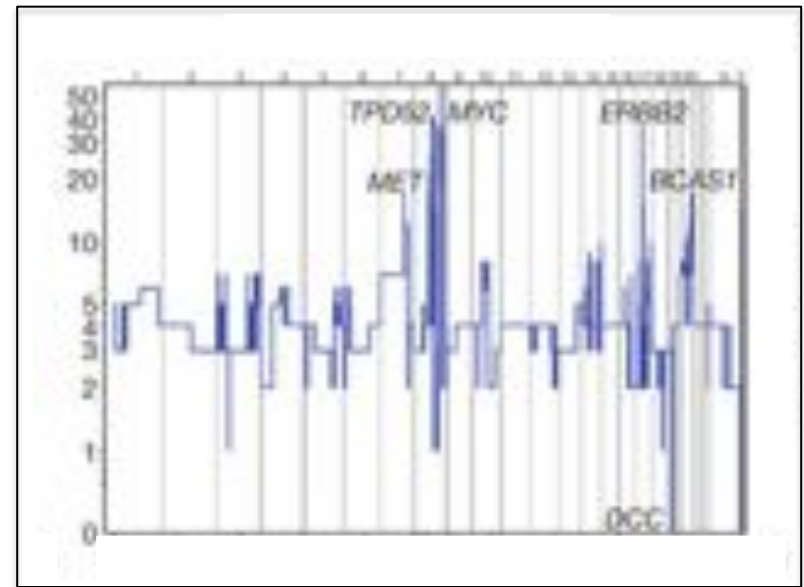
# Current Collaborations



# Long Read Sequencing of SK-BR-3



(Wen-Sheng et al, 2009)



(Navin et al, 2011)

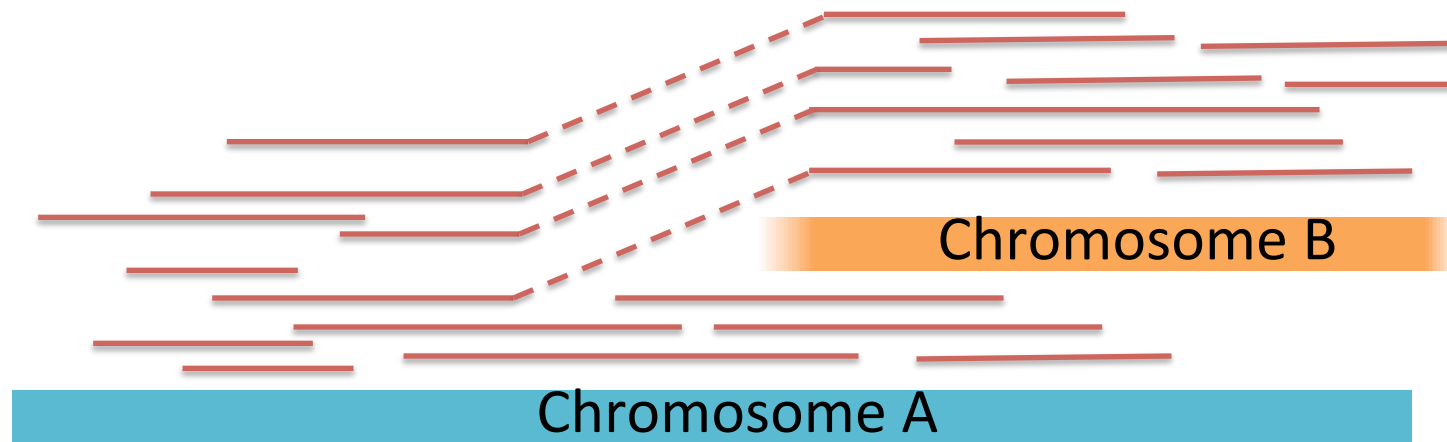
## Long read PacBio sequencing of SK-BR-3 breast cancer cell line

- Her2+ breast cancer is one of the most deadly forms of the disease
- SK-BR-3 is one of the most important models, known to have widespread CNVs
- Currently have 72x coverage with long read PacBio sequencing (mean: ~10kbp)
- Analyzing breakpoints in an attempt to infer the mutation history, especially around HER2

In collaboration with McCombie (CSHL) and McPherson (OICR) labs



# Structural variant discovery with long reads



## 1. Alignment-based split read analysis: Efficient capture of most events

BWA-MEM + Lumpy

## 2. Local assembly of regions of interest: In-depth analysis with *base-pair precision*

Localized HGAP + Celera Assembler + MUMmer

## 3. Whole genome assembly: In-depth analysis including *novel sequences*

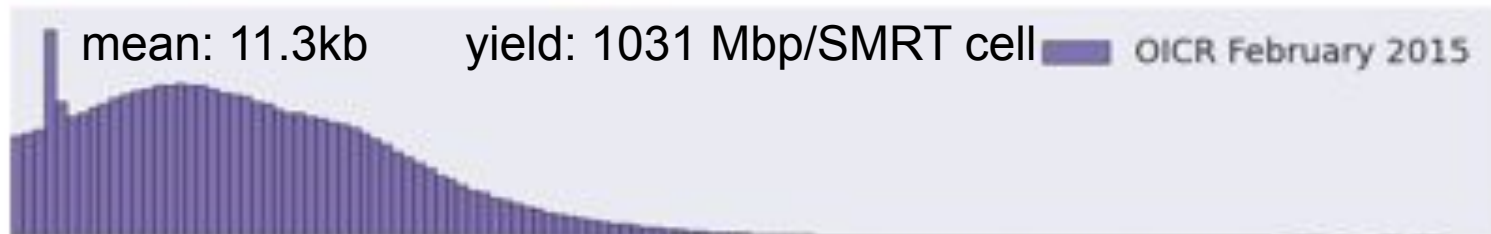
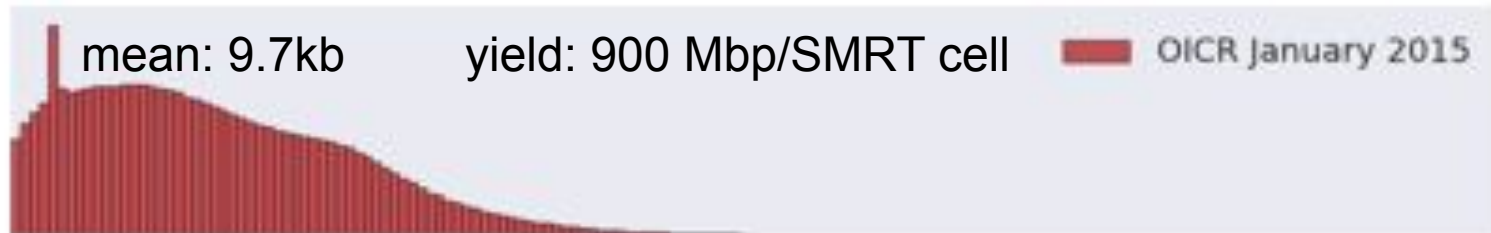
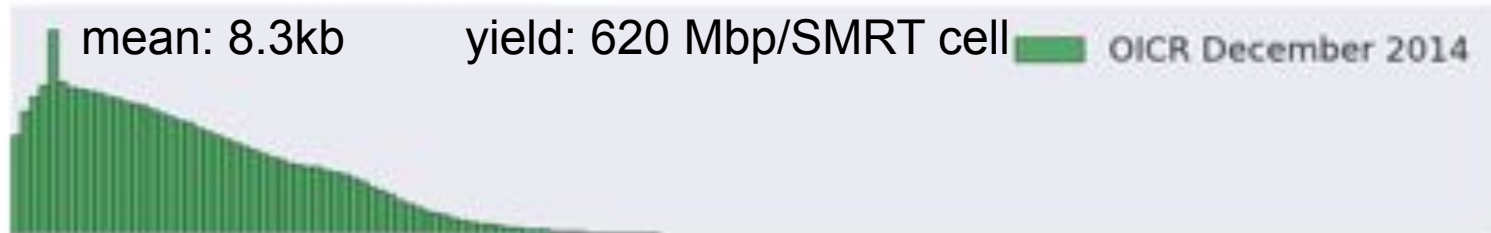
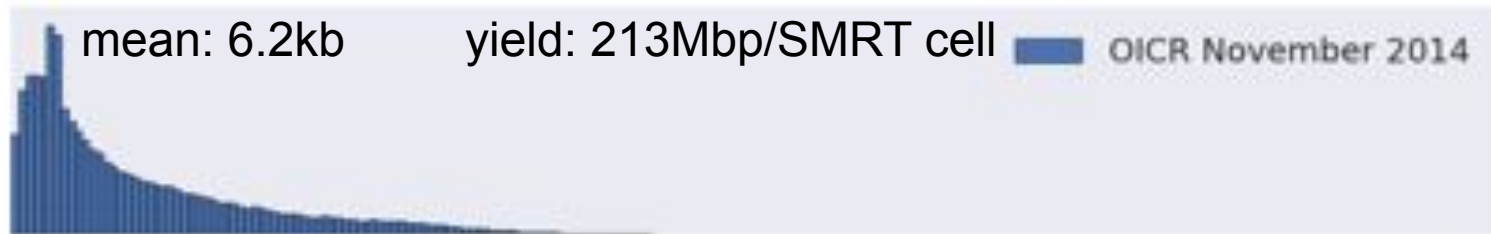
DNAnexus-enabled version of Falcon

**Total Assembly: 2.64Gbp**

**Contig N50: 2.56 Mbp**

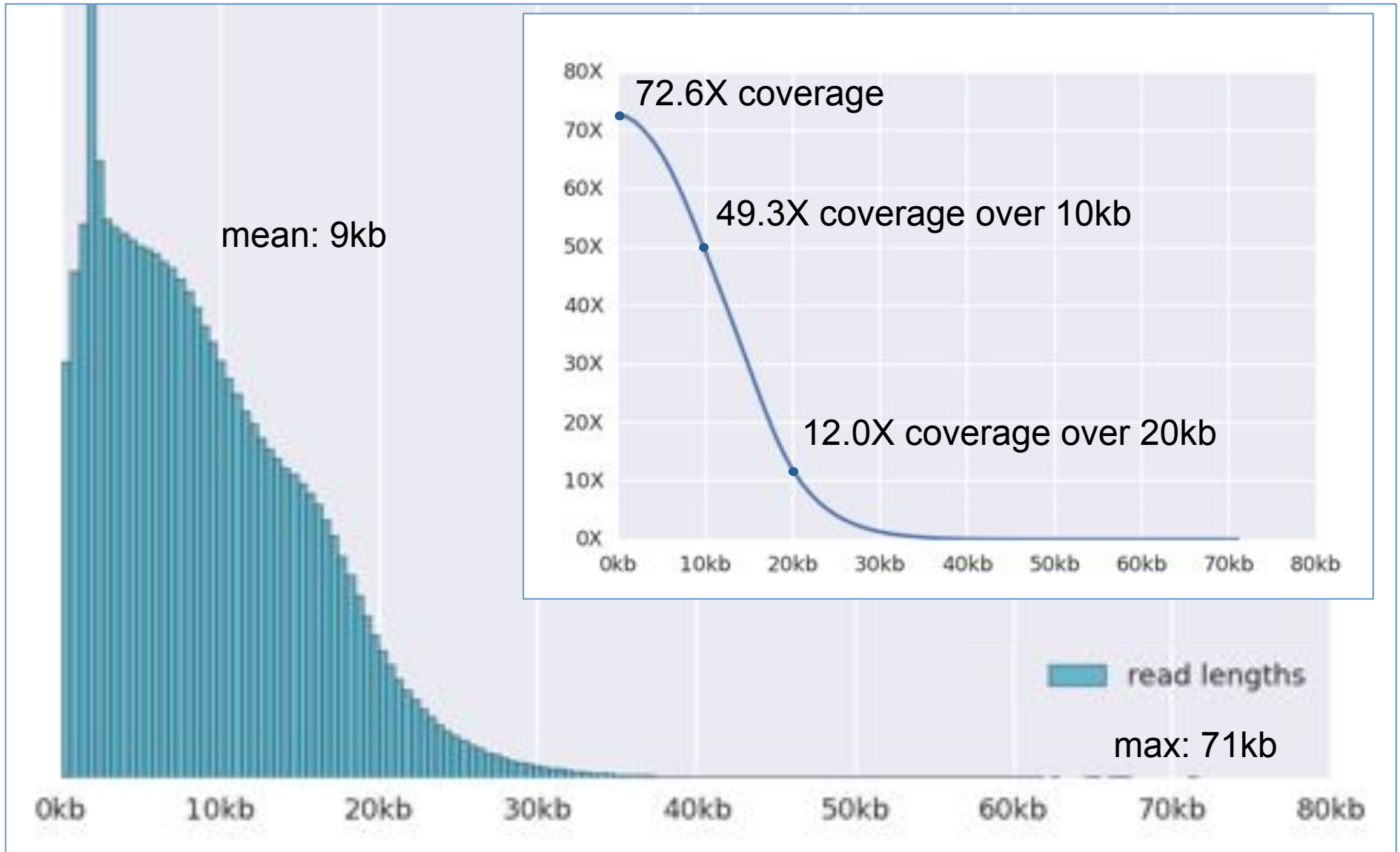
**Max Contig: 23.5Mbp**

# Improving SMRTcell Performance

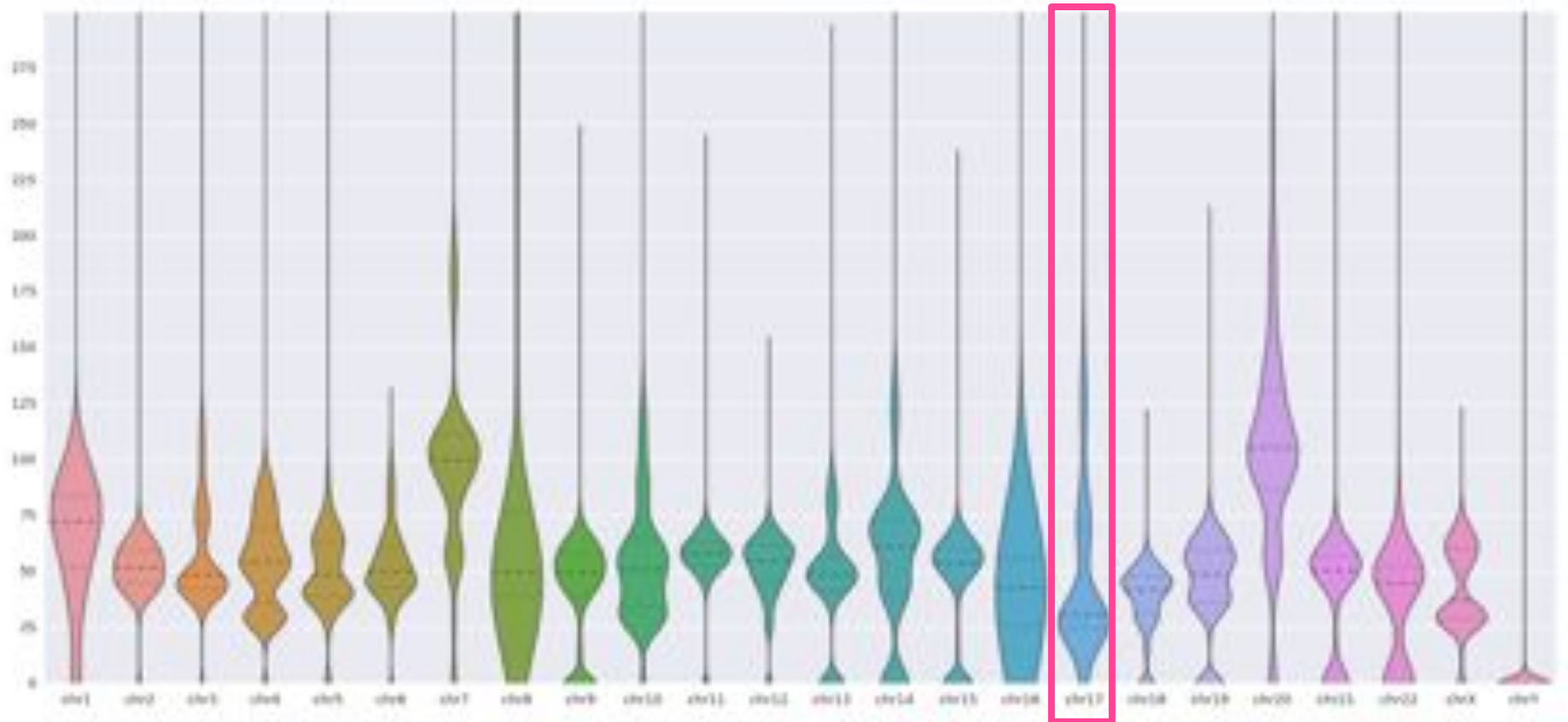


0kb      10kb      20kb      30kb      40kb      50kb      60kb      70kb

# PacBio read length distribution



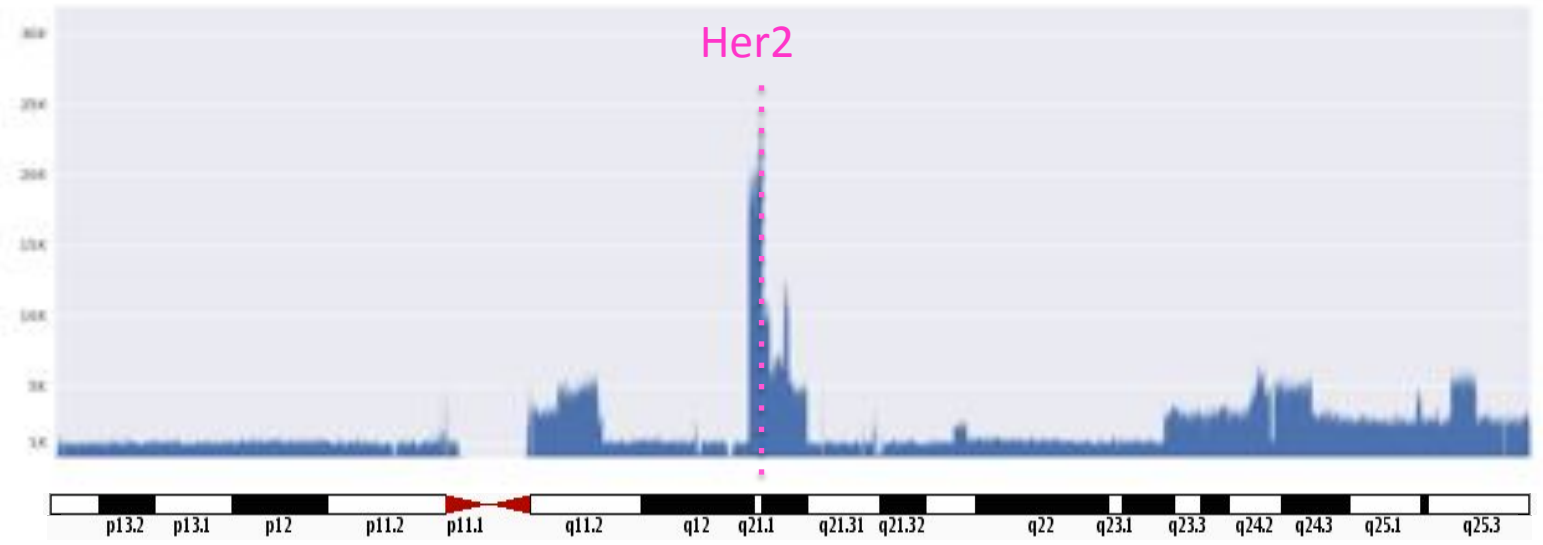
# Genome-wide alignment coverage



Genome-wide coverage averages around 54X

Coverage per chromosome varies greatly as expected from previous karyotyping results

PacBio

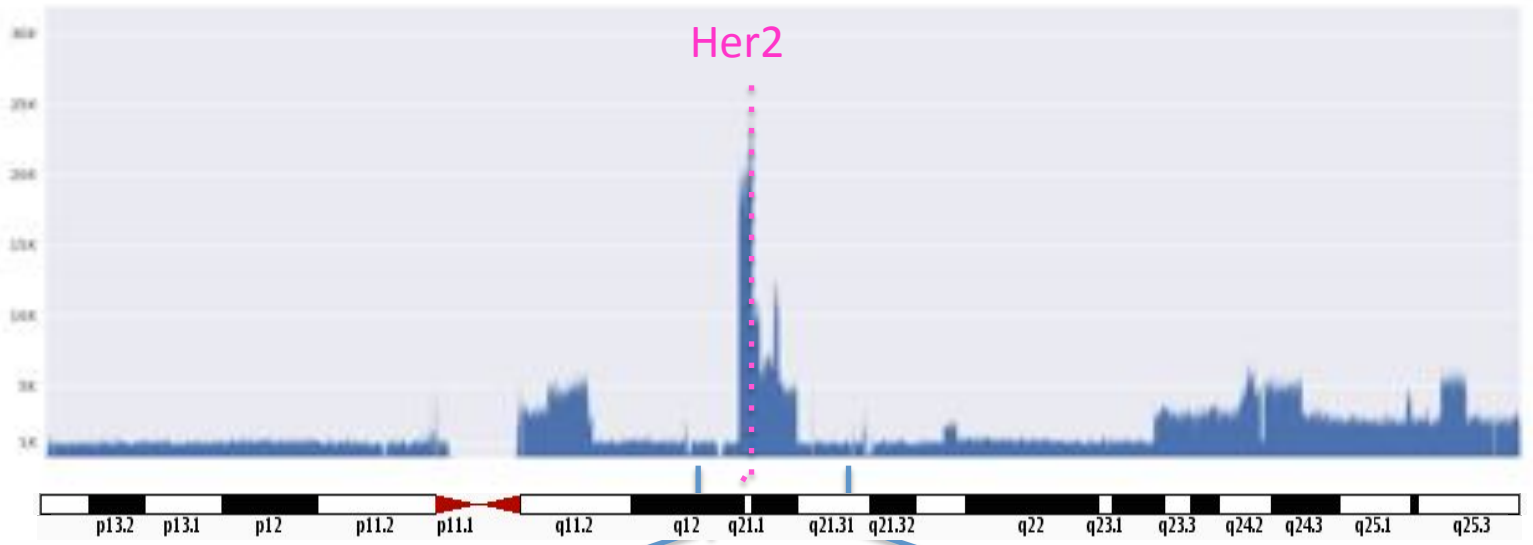


Chr 17: 83 Mb

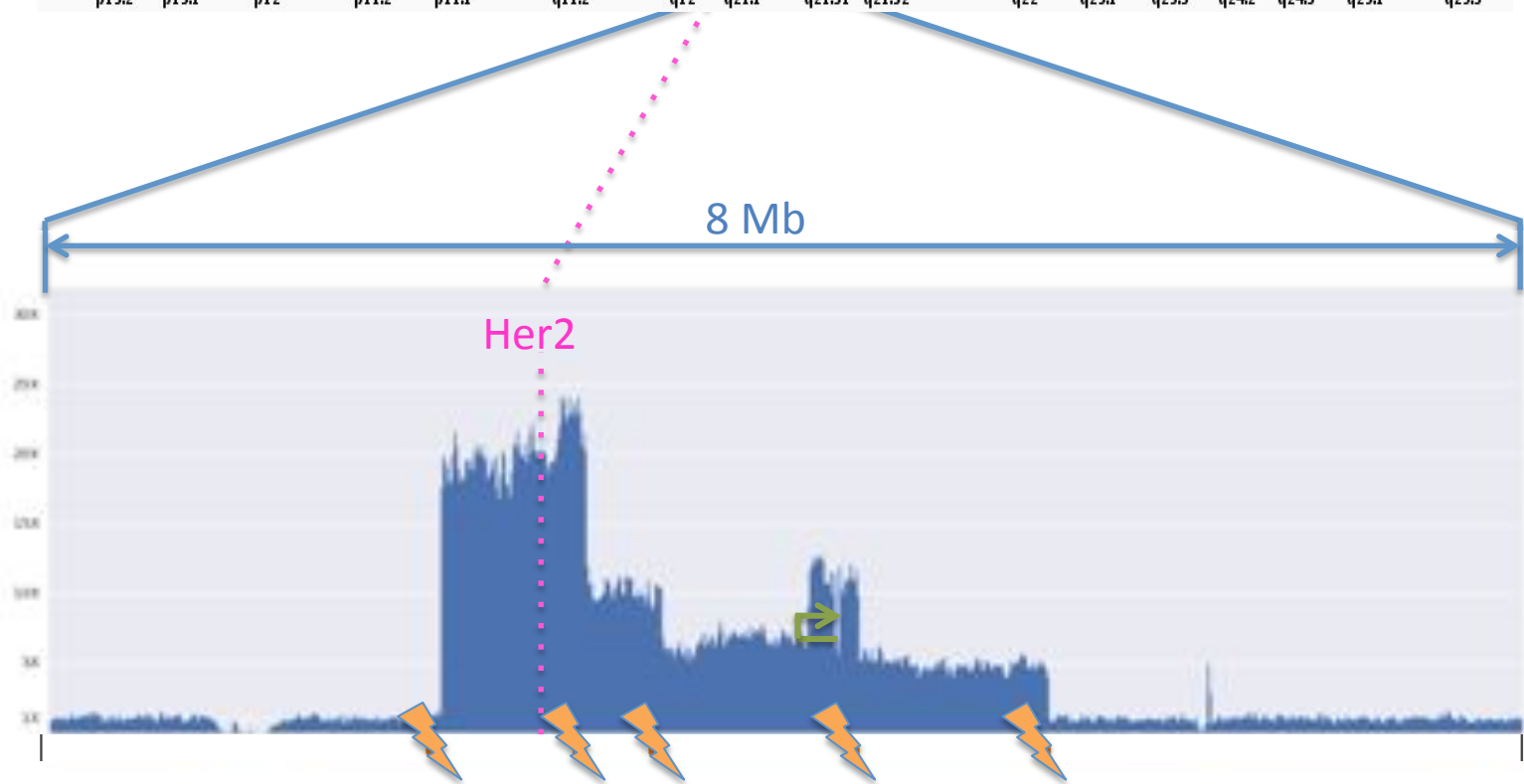




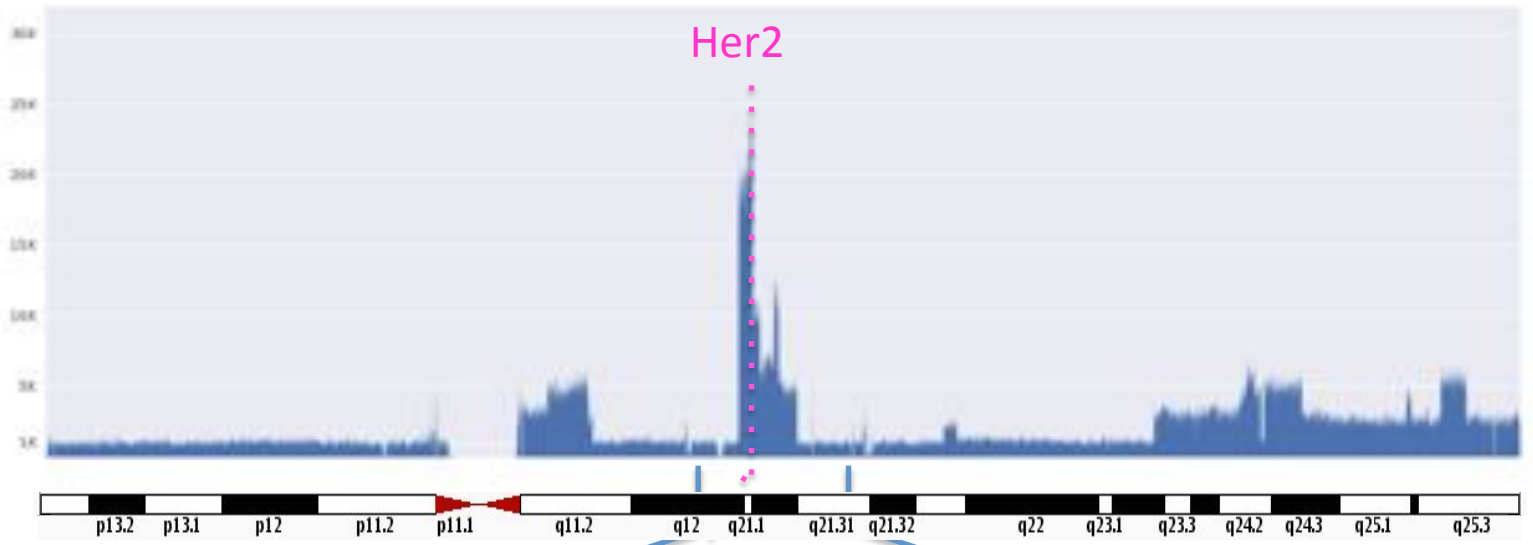
PacBio



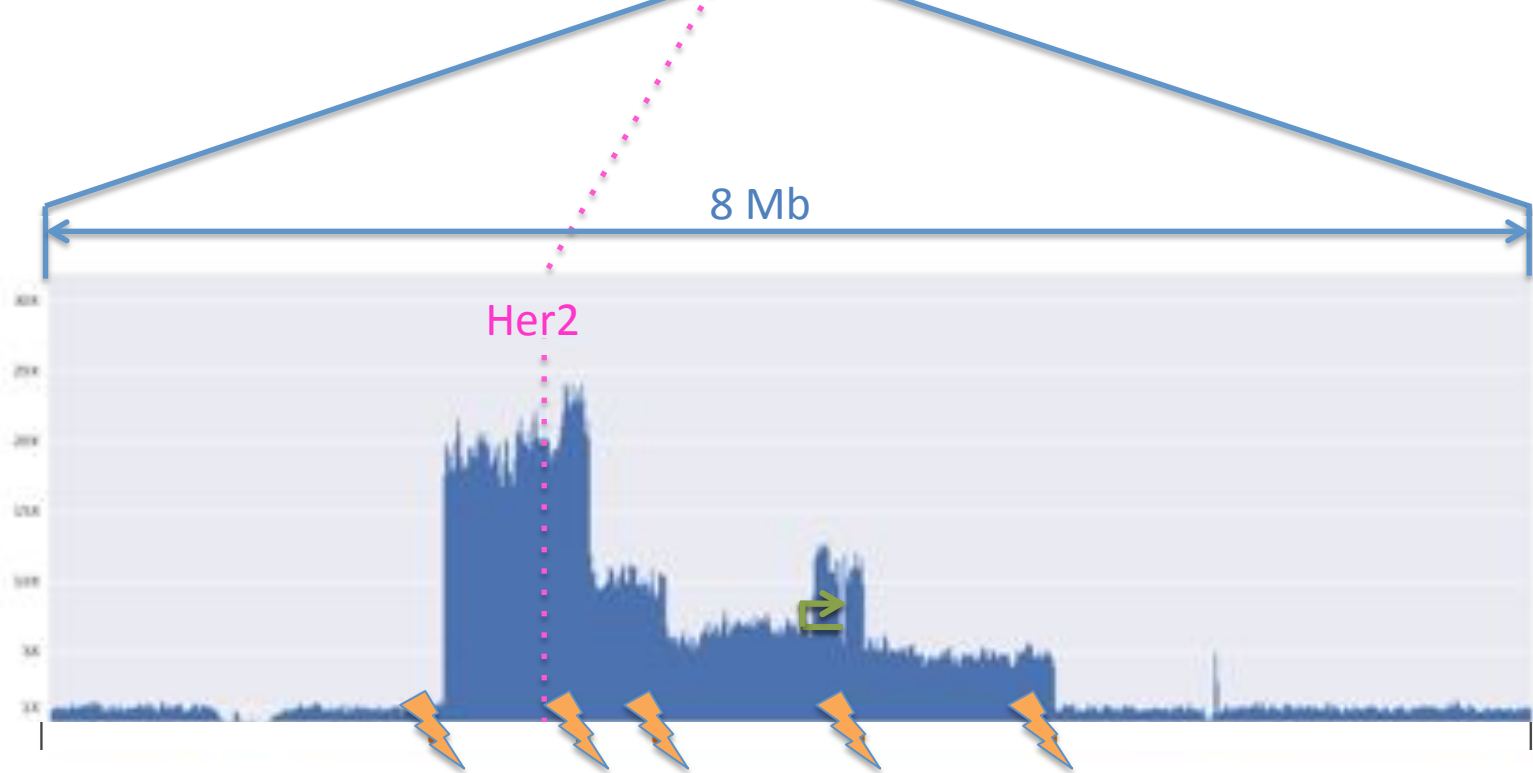
PacBio  
chr17

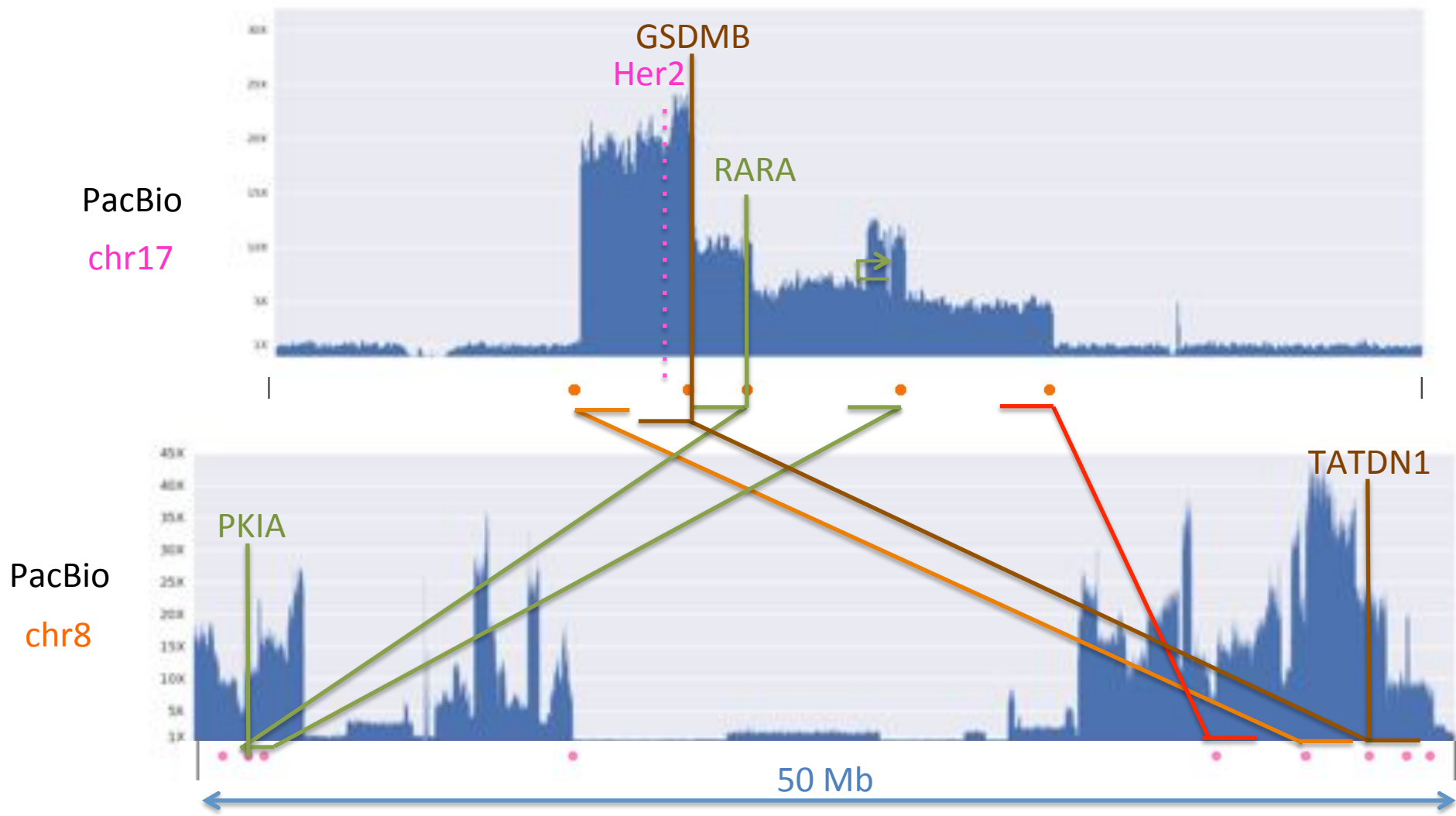


PacBio

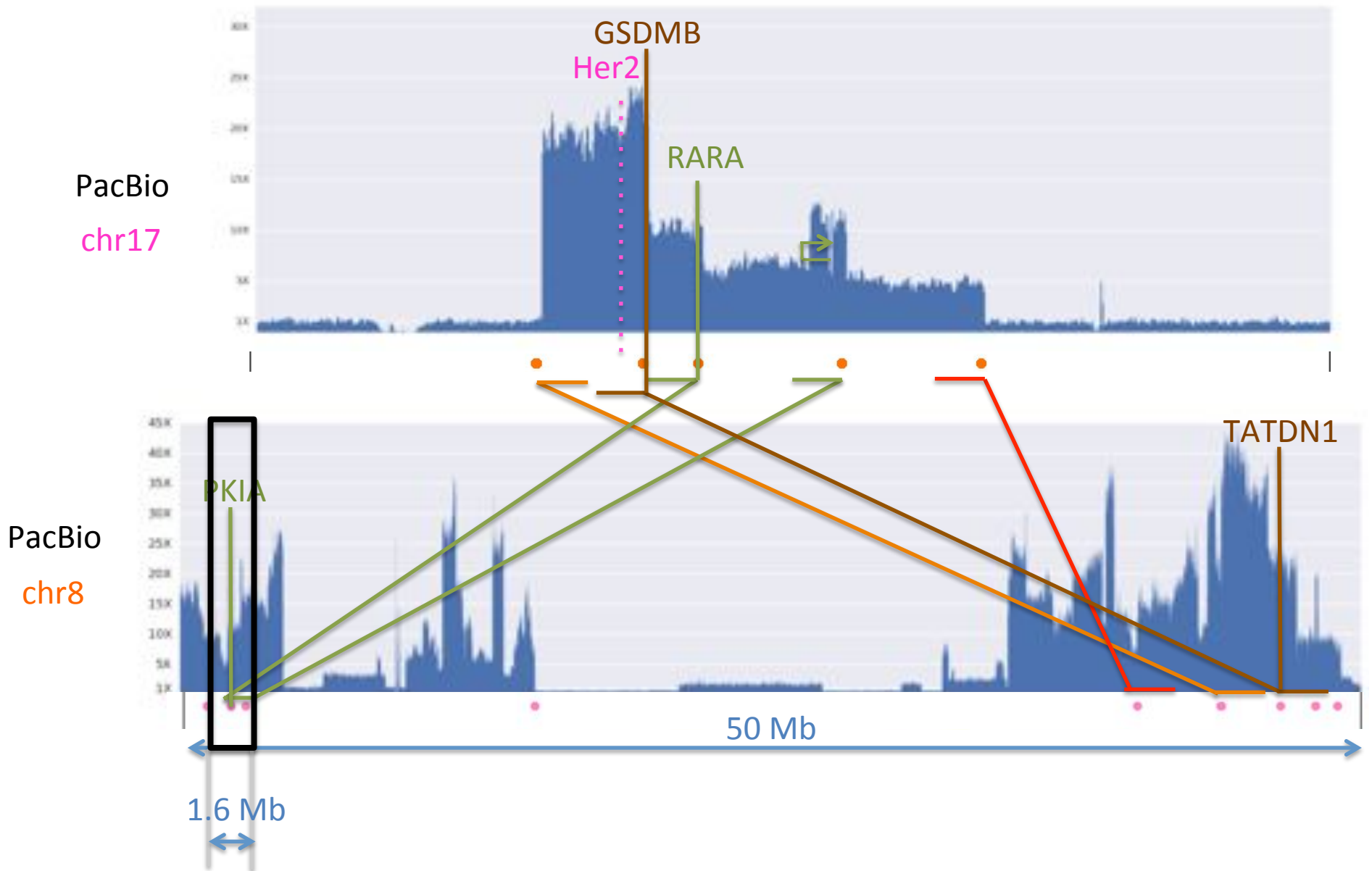


PacBio  
chr17

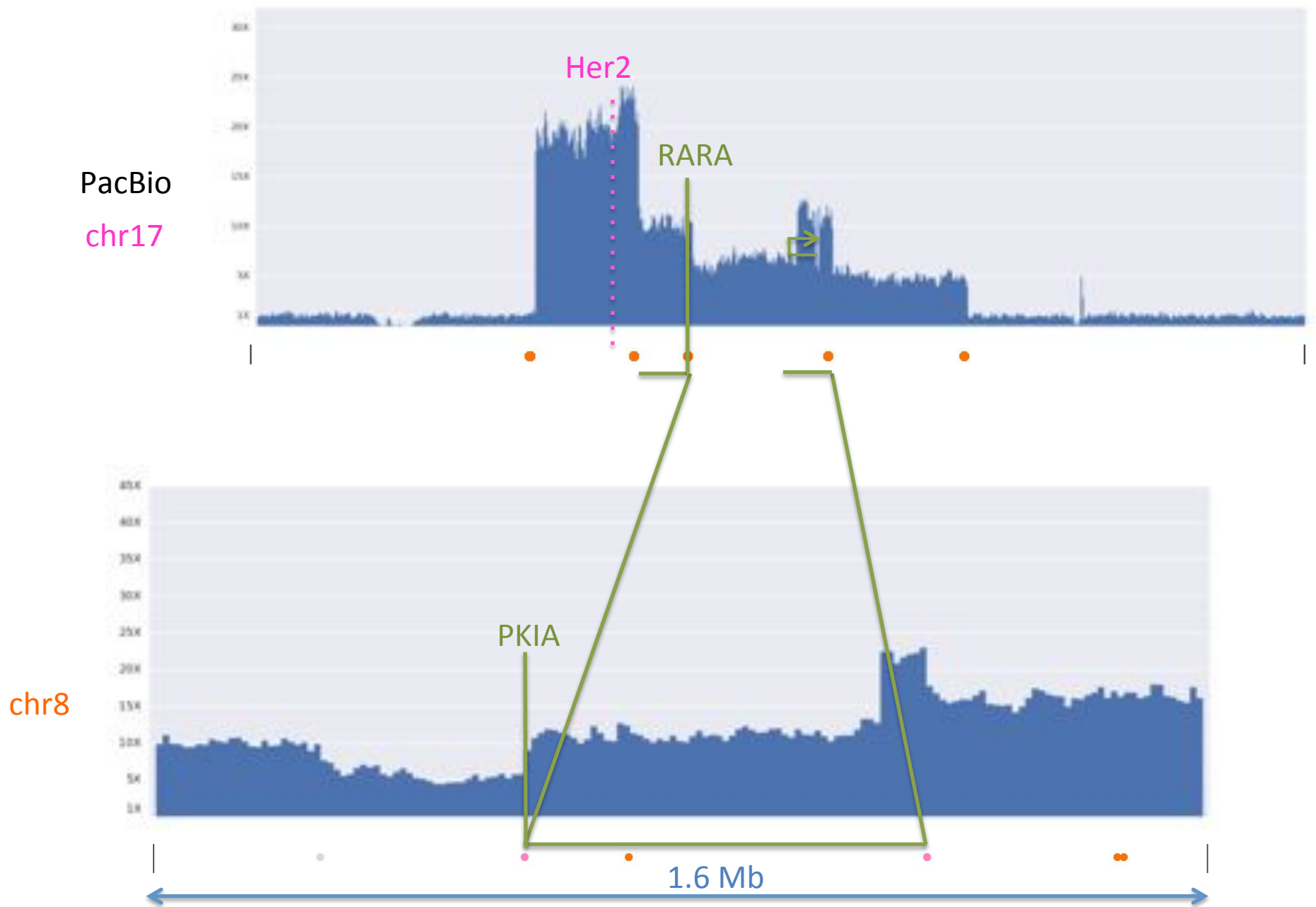




Confirmed both known gene fusions in this region



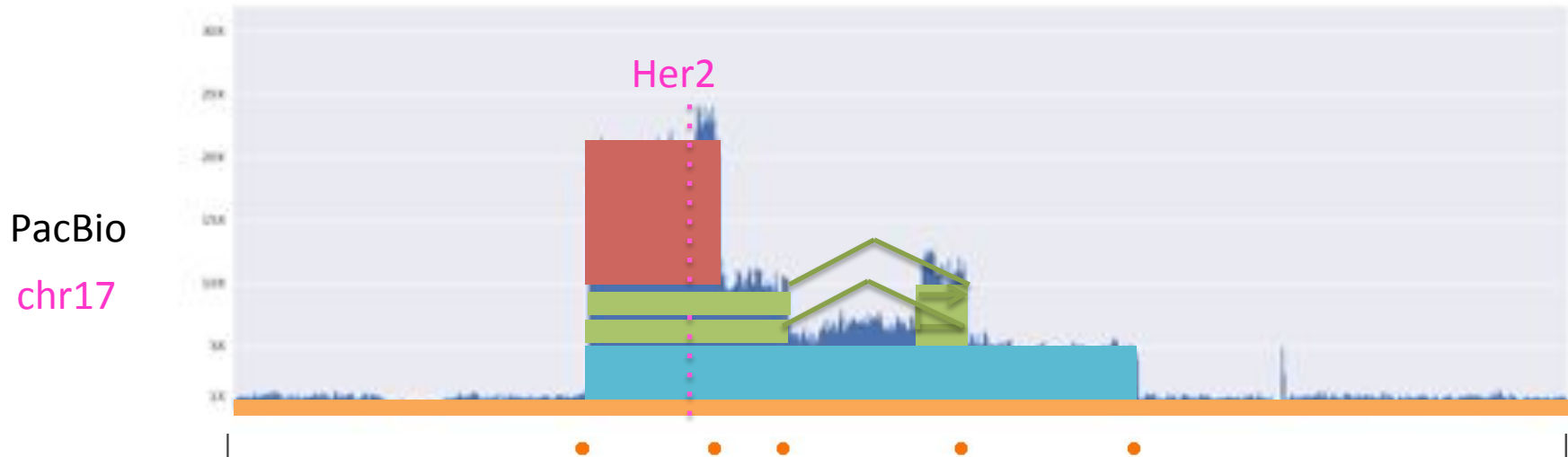
Confirmed both known gene fusions in this region



Joint coverage and breakpoint analysis to discover underlying events



# Cancer lesion Reconstruction



By comparing the proportion of reads that are spanning or split at breakpoints we can begin to infer the history of the genetic lesions.

1. Healthy diploid genome
2. Original translocation into chromosome 8
3. Duplication, inversion, and inverted duplication within chromosome 8
4. Final duplication from within chromosome 8

# Cancer lesion Reconstruction

## **Available *today* under the Toronto Agreement:**

- Fastq & BAM files of aligned reads
- Interactive Coverage Analysis with BAM.IOBIO
- Whole genome assembly

## **Available soon**

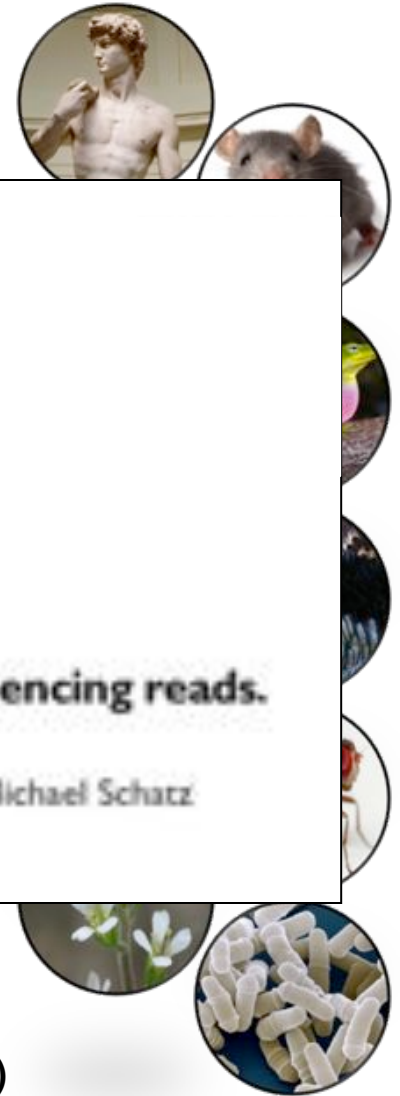
- Whole genome methylation analysis
- Full length cDNA transcriptome analysis
- Comparison to single cell analysis of >100 individual cells

<http://schatzlab.cshl.edu/skbr3>

4. Final duplication from within chromosome 8

# What should we expect from an assembly?

## *The resurgence of reference quality genomes*



**bioRxiv**  
beta  
THE PREPRINT SERVER FOR BIOLOGY

New Results

**Error correction and assembly complexity of single molecule sequencing reads.**

Hayan Lee , James Gurtowski , Shinjae Yoo , Shoshana Marcus , W. Richard McCombie , Michael Schatz  
doi: <http://dx.doi.org/10.1101/006395>

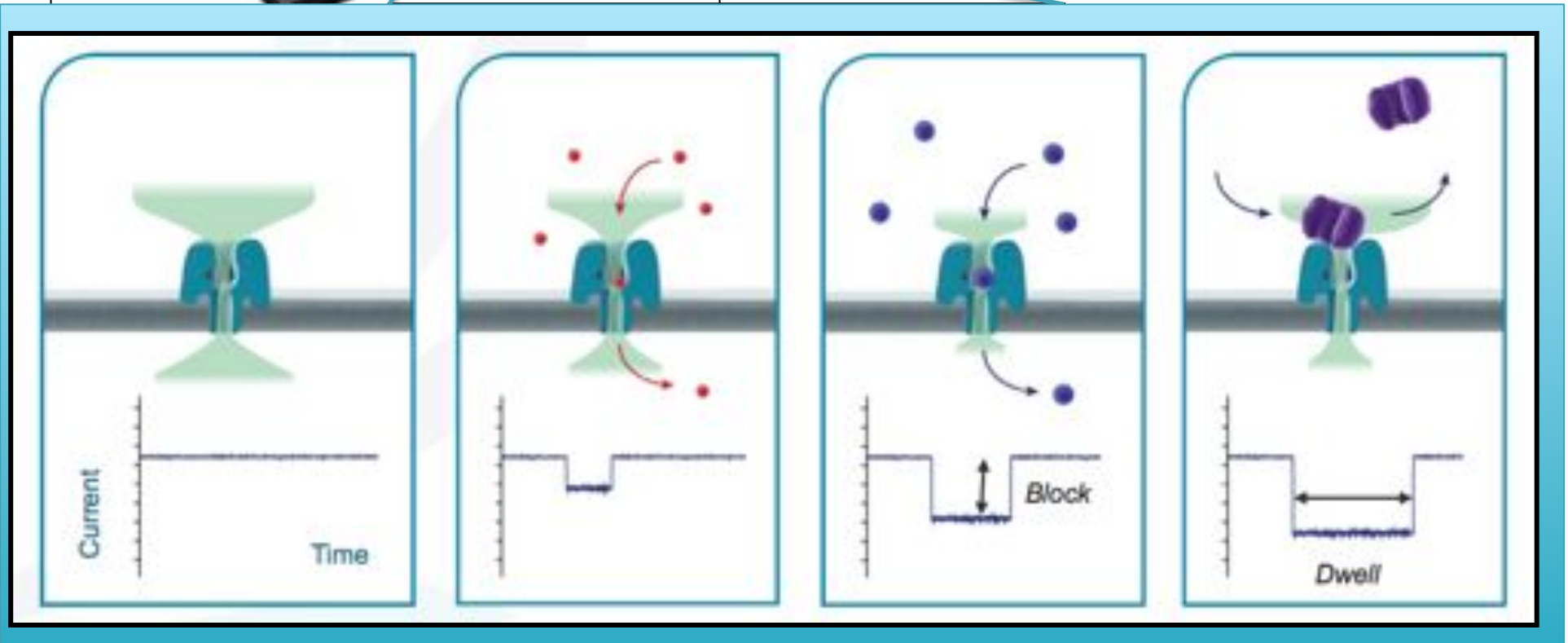
## **Caveats**

Model only as good as the available references (esp. haploid sequences)  
Technologies are quickly improving, exciting new scaffolding technologies

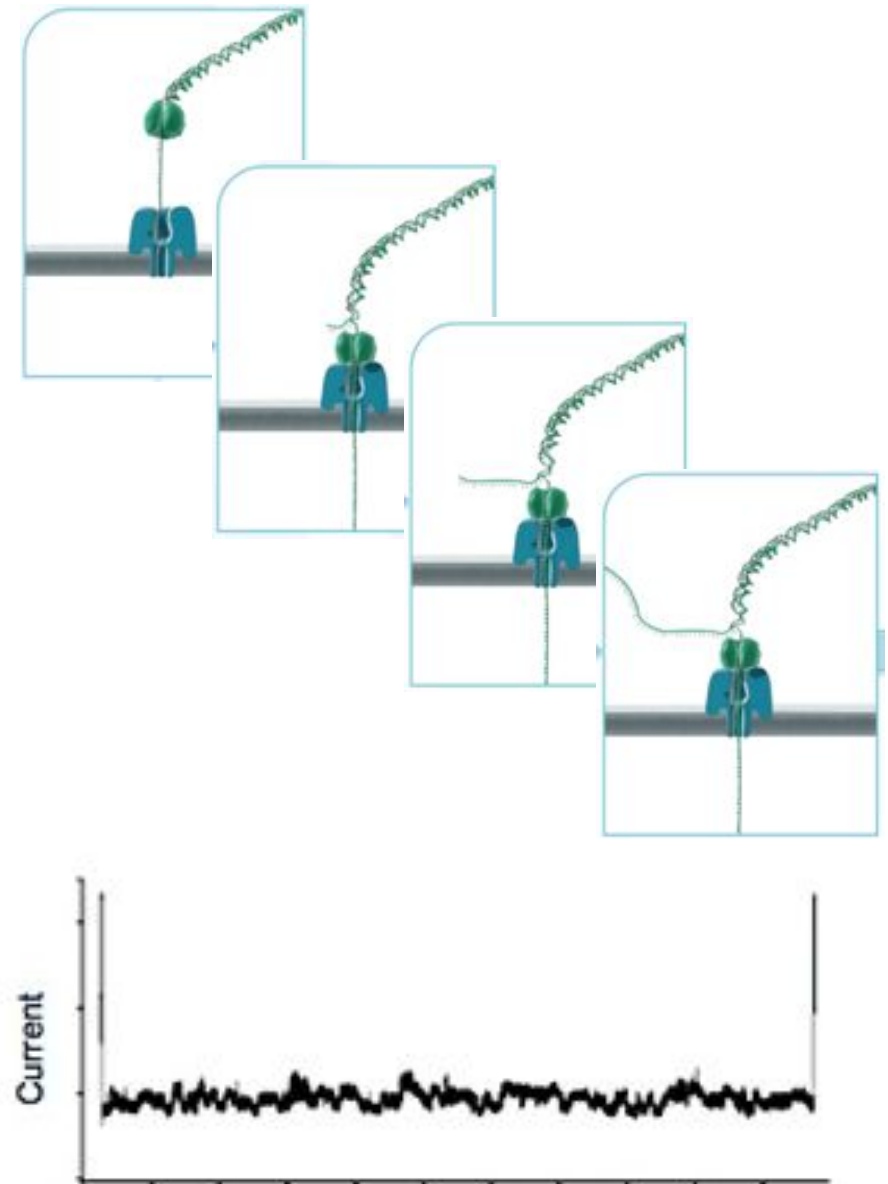
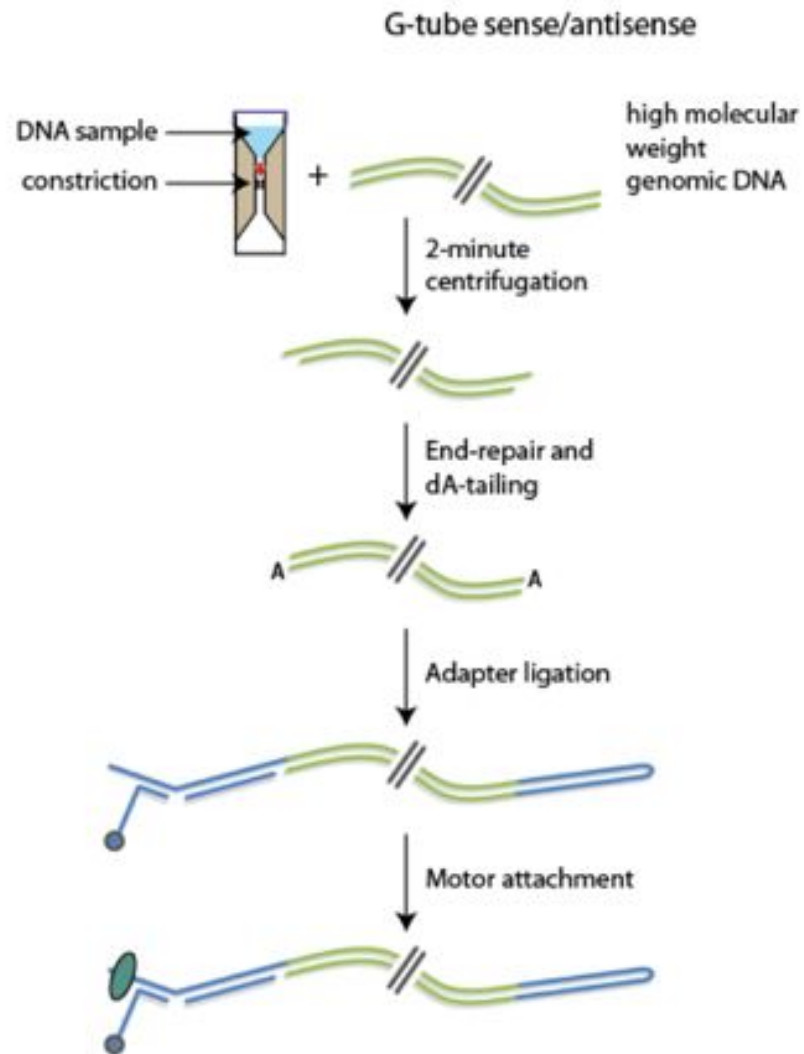
# Oxford Nanopore MinION



- Thumb drive sized sequencer powered over USB
- Capacity for 512 reads at once
- Senses DNA by measuring changes to ion flow

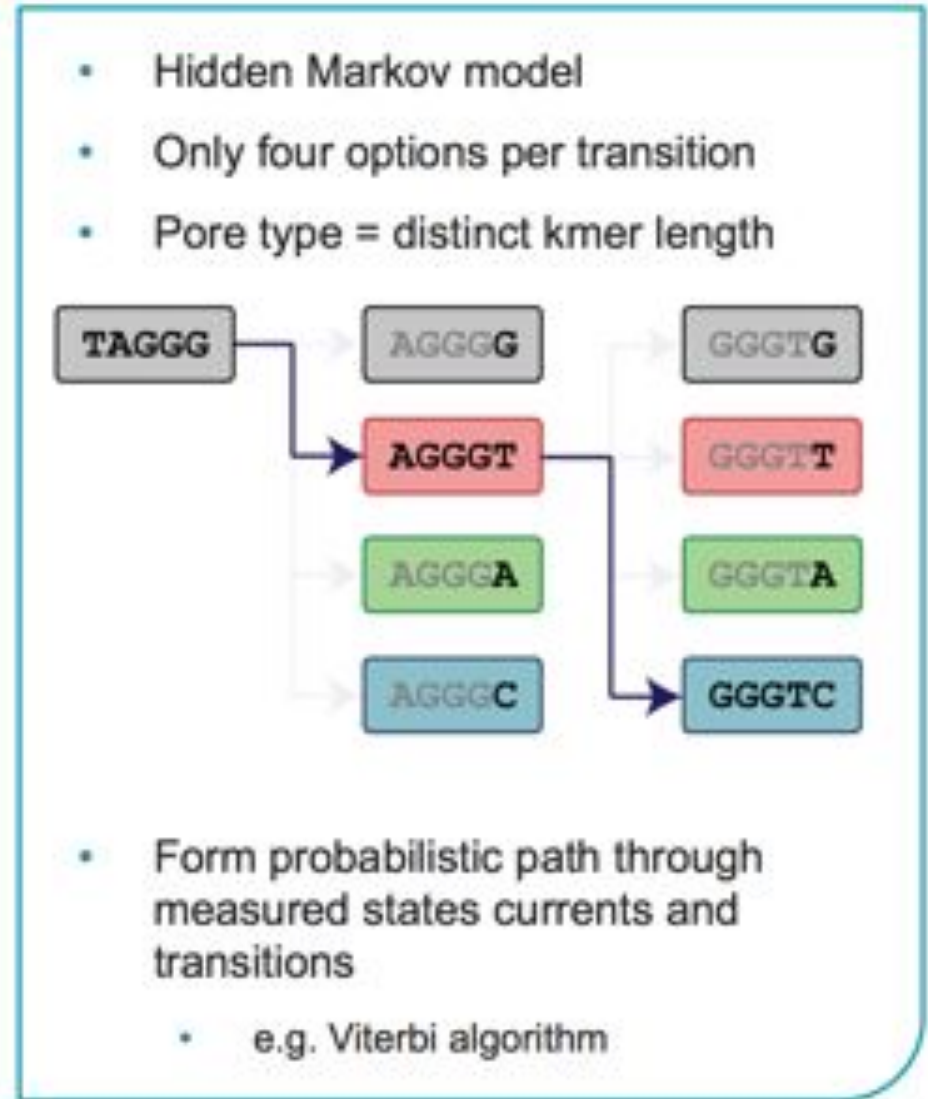
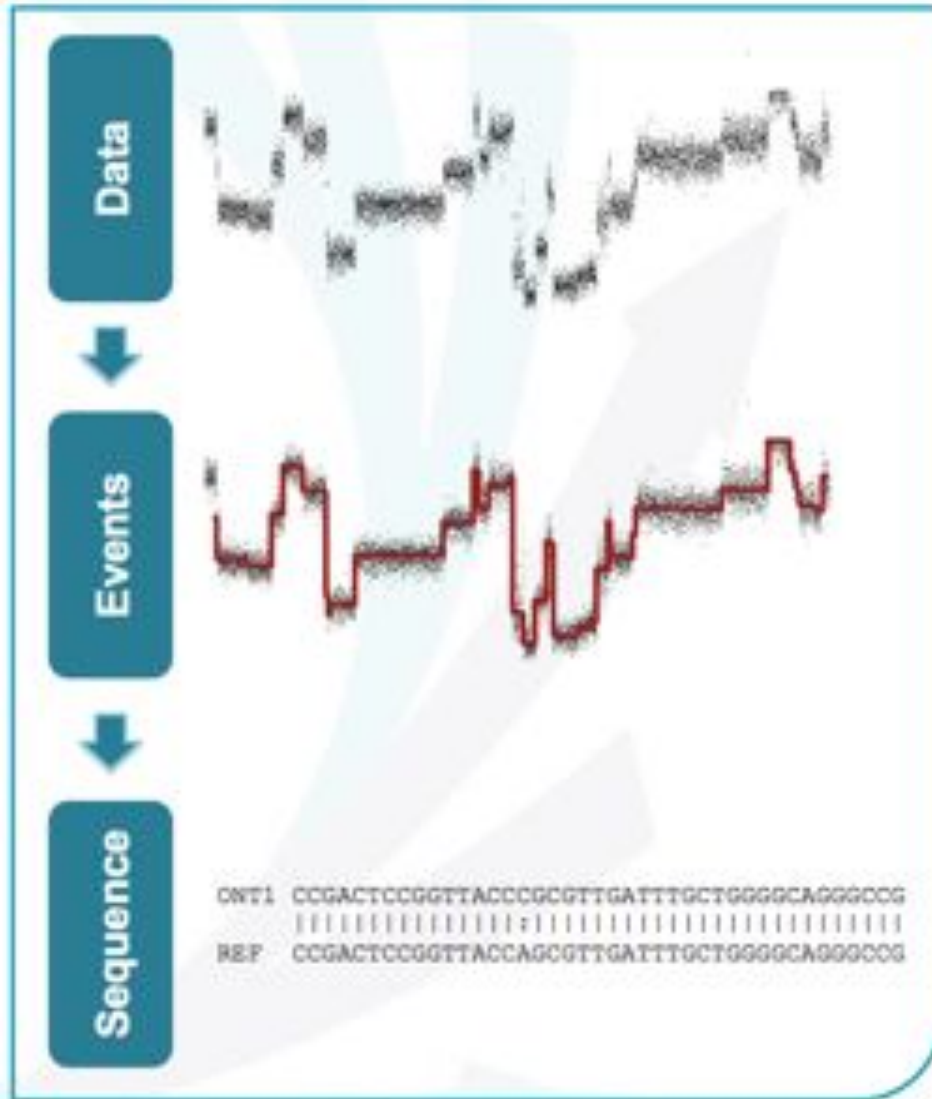


# Nanopore Sequencing



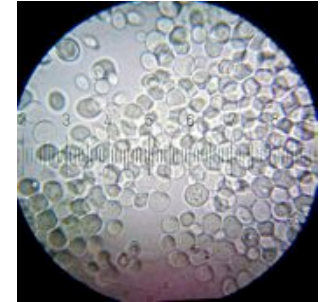


# Nanopore Sequencing



Basecalling currently performed at Amazon with frequent updates to algorithm

# Nanopore Readlengths



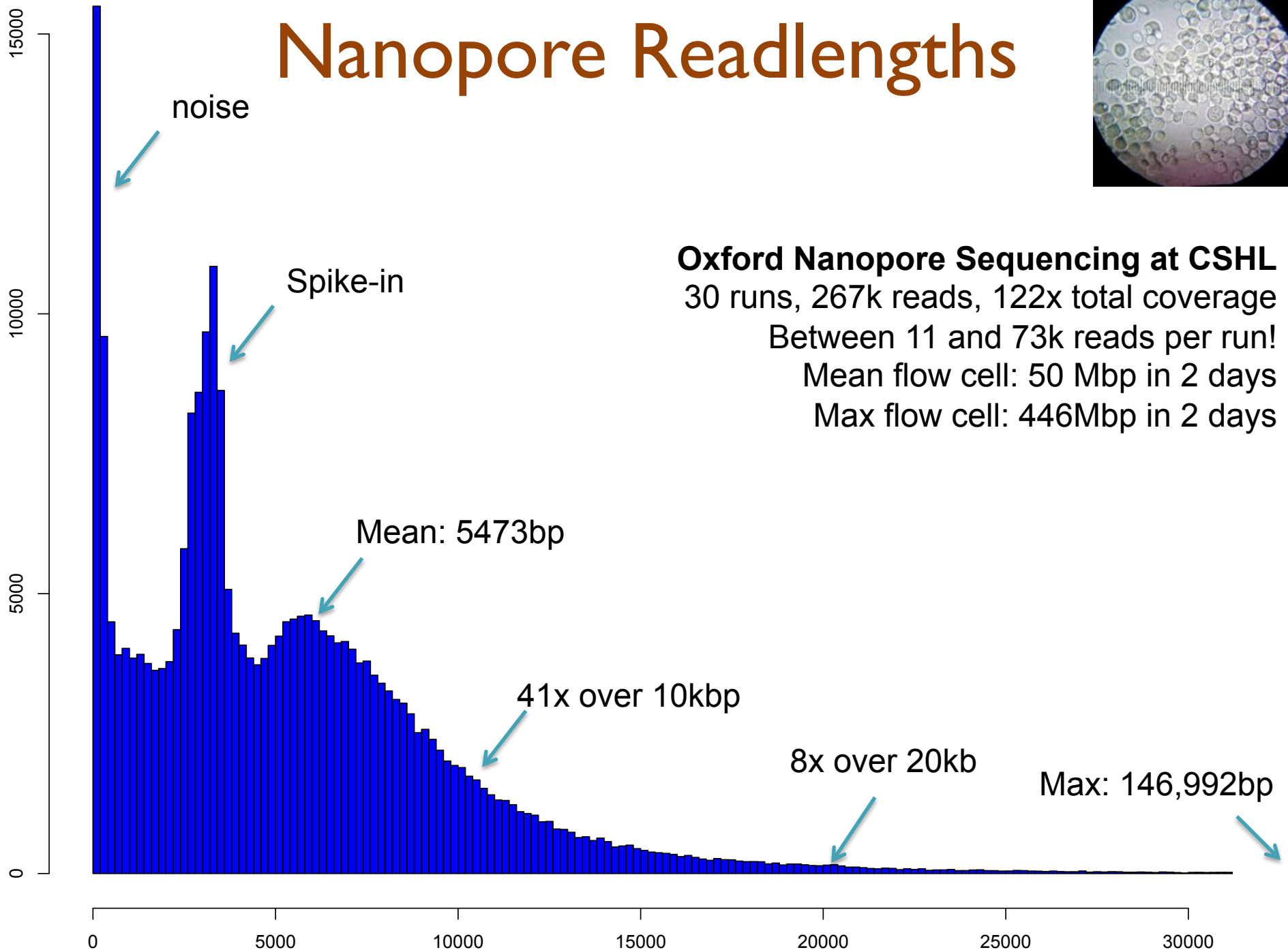
## Oxford Nanopore Sequencing at CSHL

30 runs, 267k reads, 122x total coverage

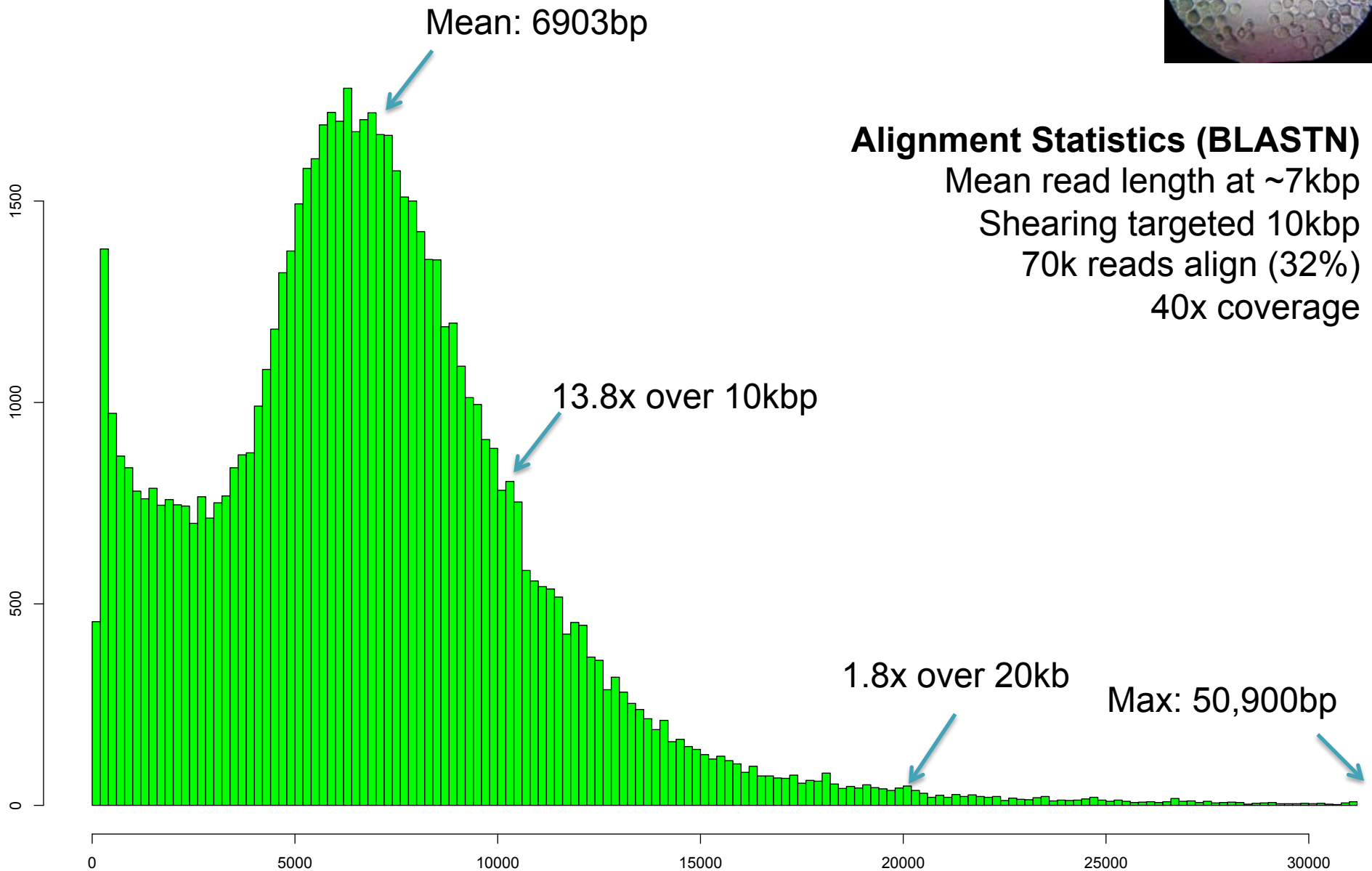
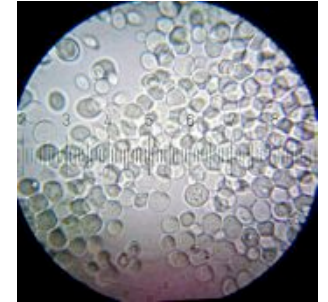
Between 11 and 73k reads per run!

Mean flow cell: 50 Mbp in 2 days

Max flow cell: 446Mbp in 2 days



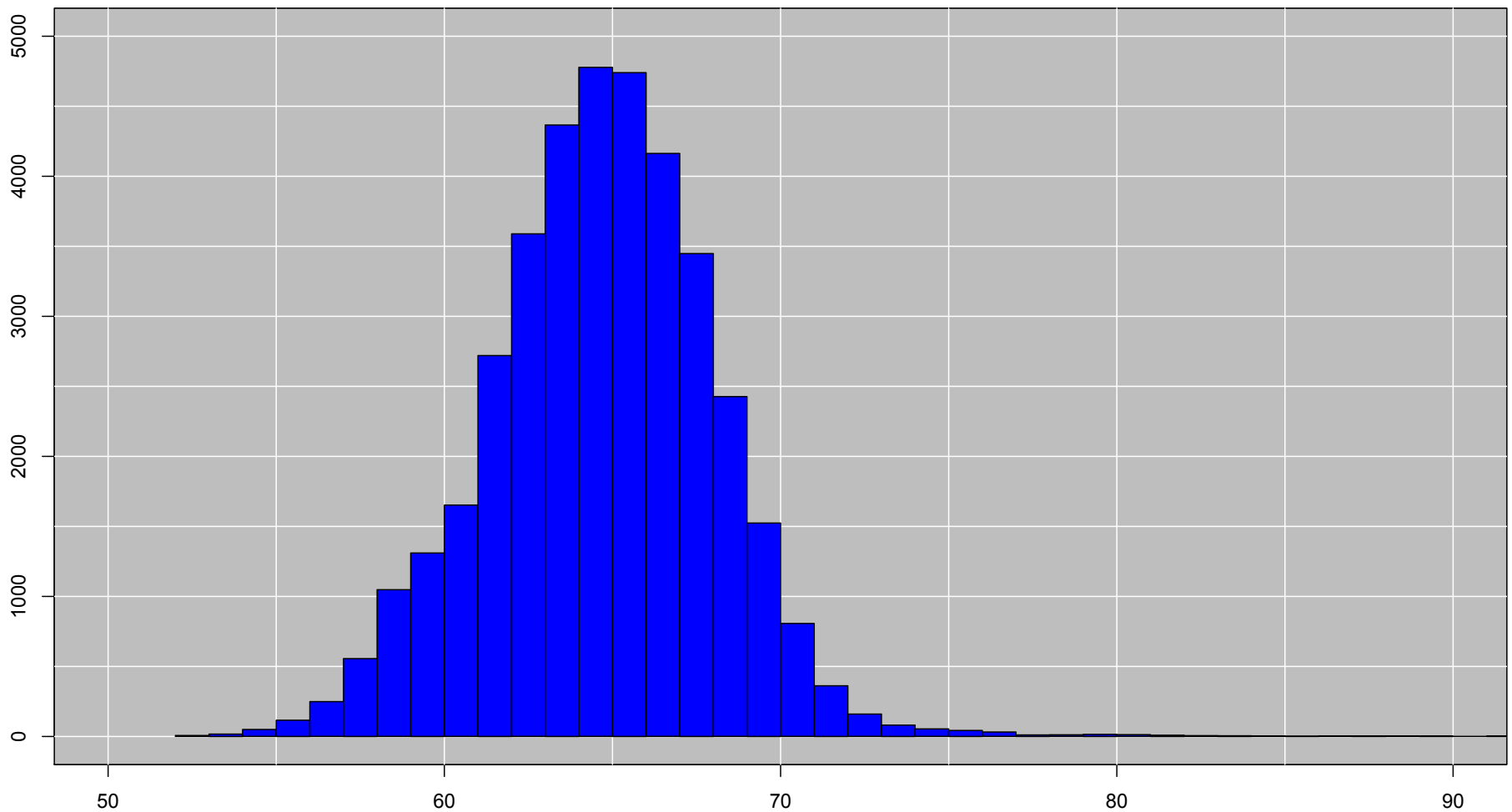
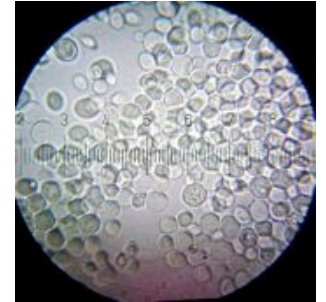
# Nanopore Alignments



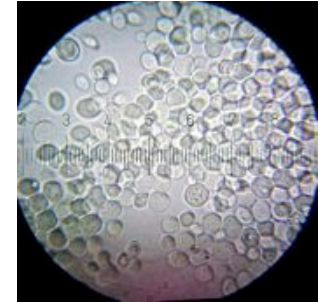
# Nanopore Accuracy

## Alignment Quality (BLASTN)

Of reads that align, average ~64% identity



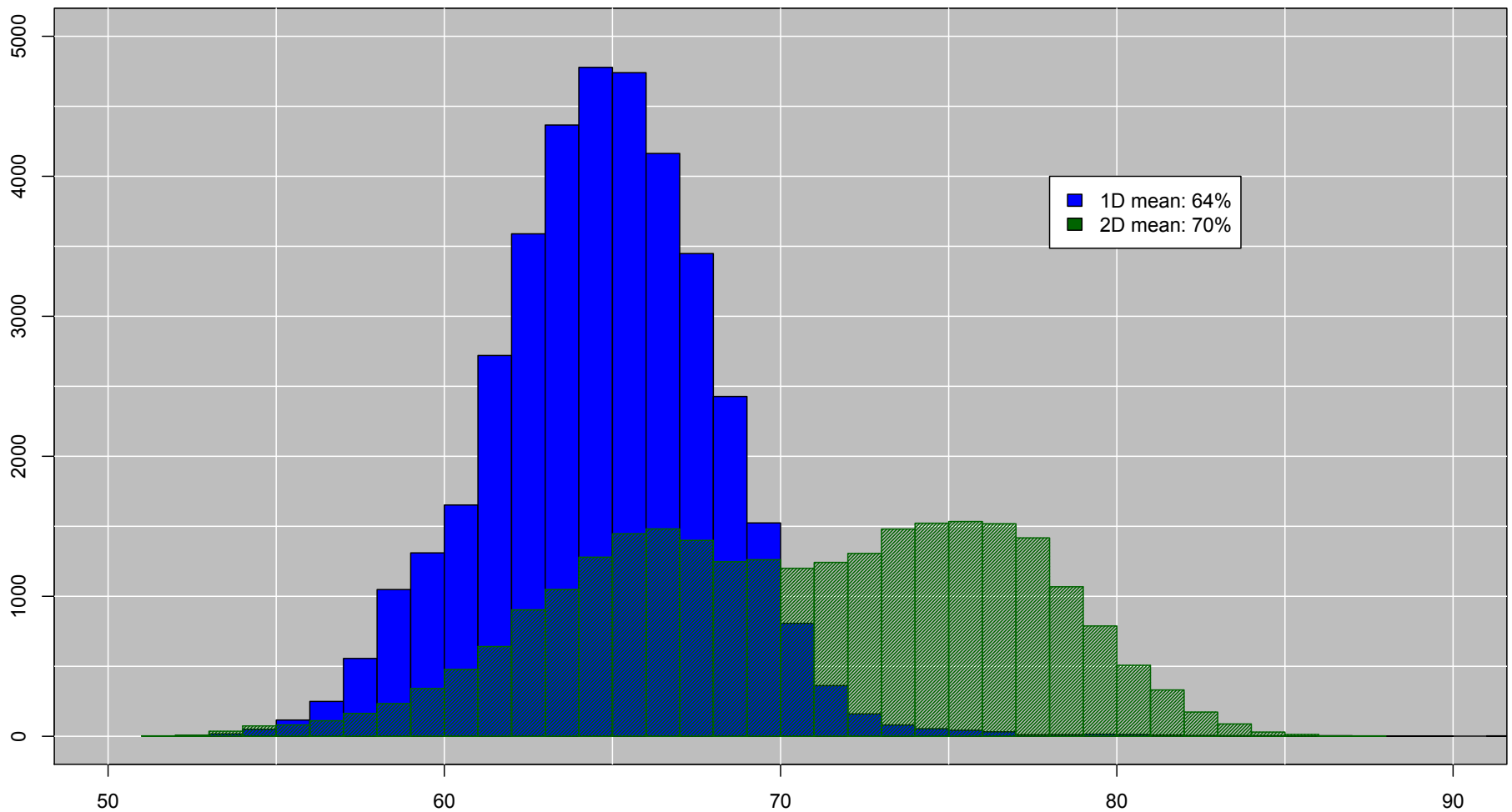
# Nanopore Accuracy



## Alignment Quality (BLASTN)

Of reads that align, average ~64% identity

“2D base-calling” improves to ~70% identity

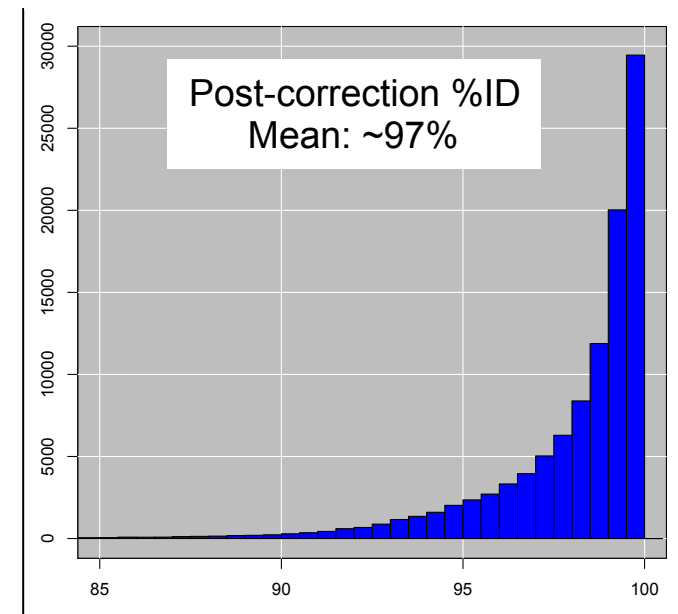
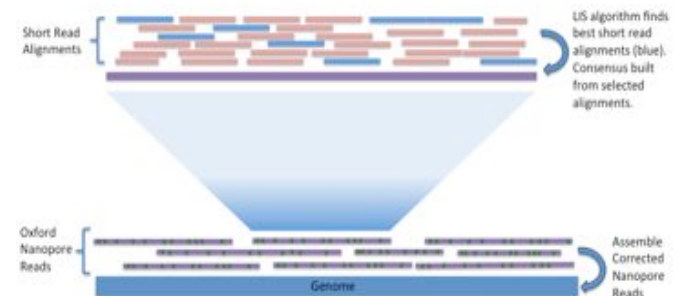


# NanoCorr: Nanopore-Illumina Hybrid Error Correction



<https://github.com/jgurtowski/nanocorr>

1. BLAST Miseq reads to all raw Oxford Nanopore reads
  - First pass scans to remove “contained” alignments
  - Second pass uses Dynamic Programming (LIS) to select set of high-identity alignments with minimal overlaps
3. Compute consensus of each Oxford Nanopore read
  - State machine of most commonly observed base at each position in read

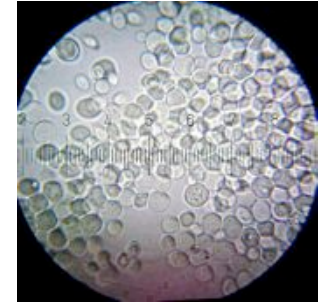


## Oxford Nanopore Sequencing and de novo Assembly of a Eukaryotic Genome

Goodwin, S, Gurtowski, J *et al.* (2015) bioRxiv doi: <http://dx.doi.org/10.1101/013490>

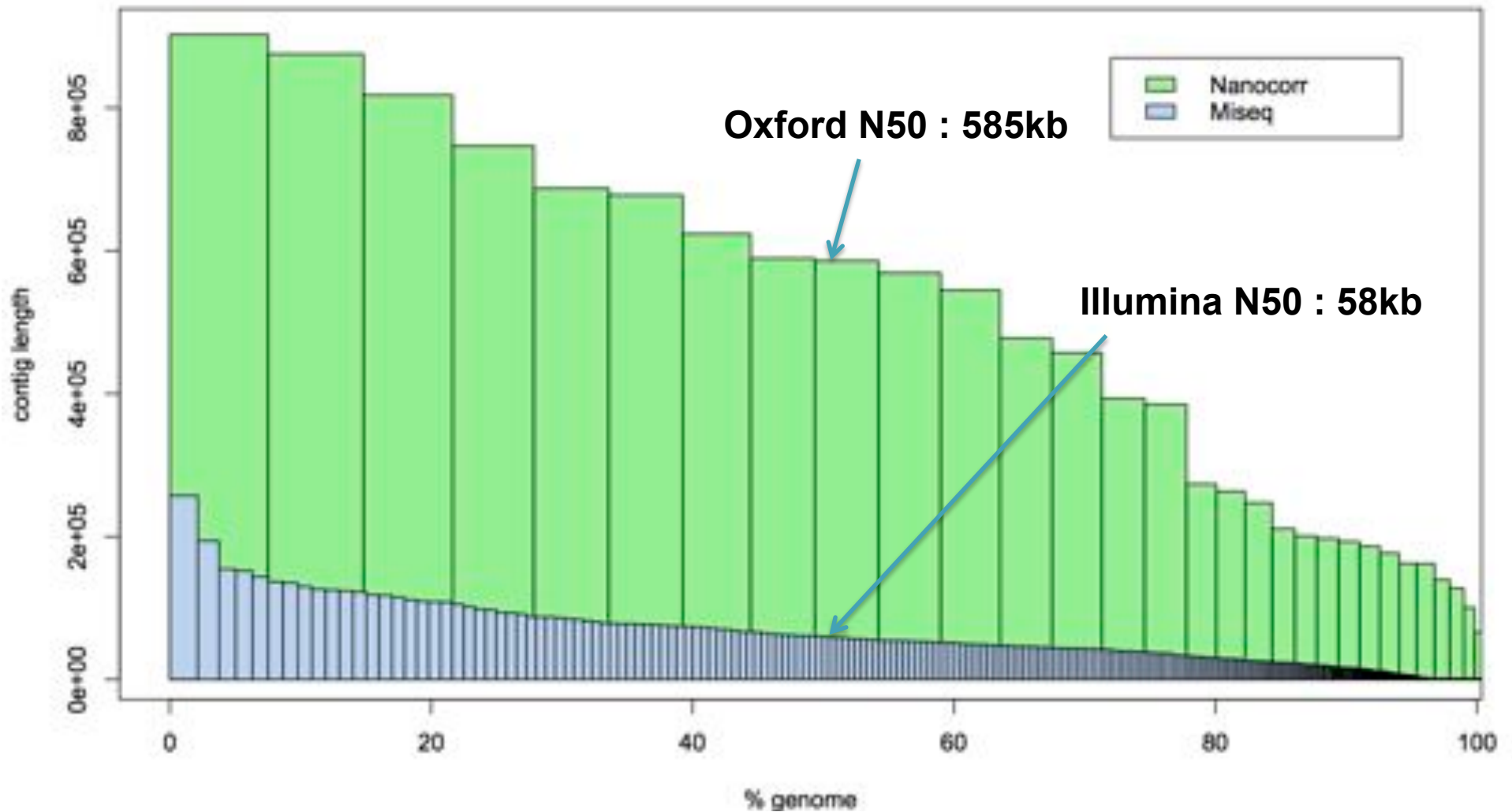


# NanoCorr Yeast Assembly



S288C Reference sequence

- 12.1Mbp; 16 chromo + mitochondria; N50: 924kbp



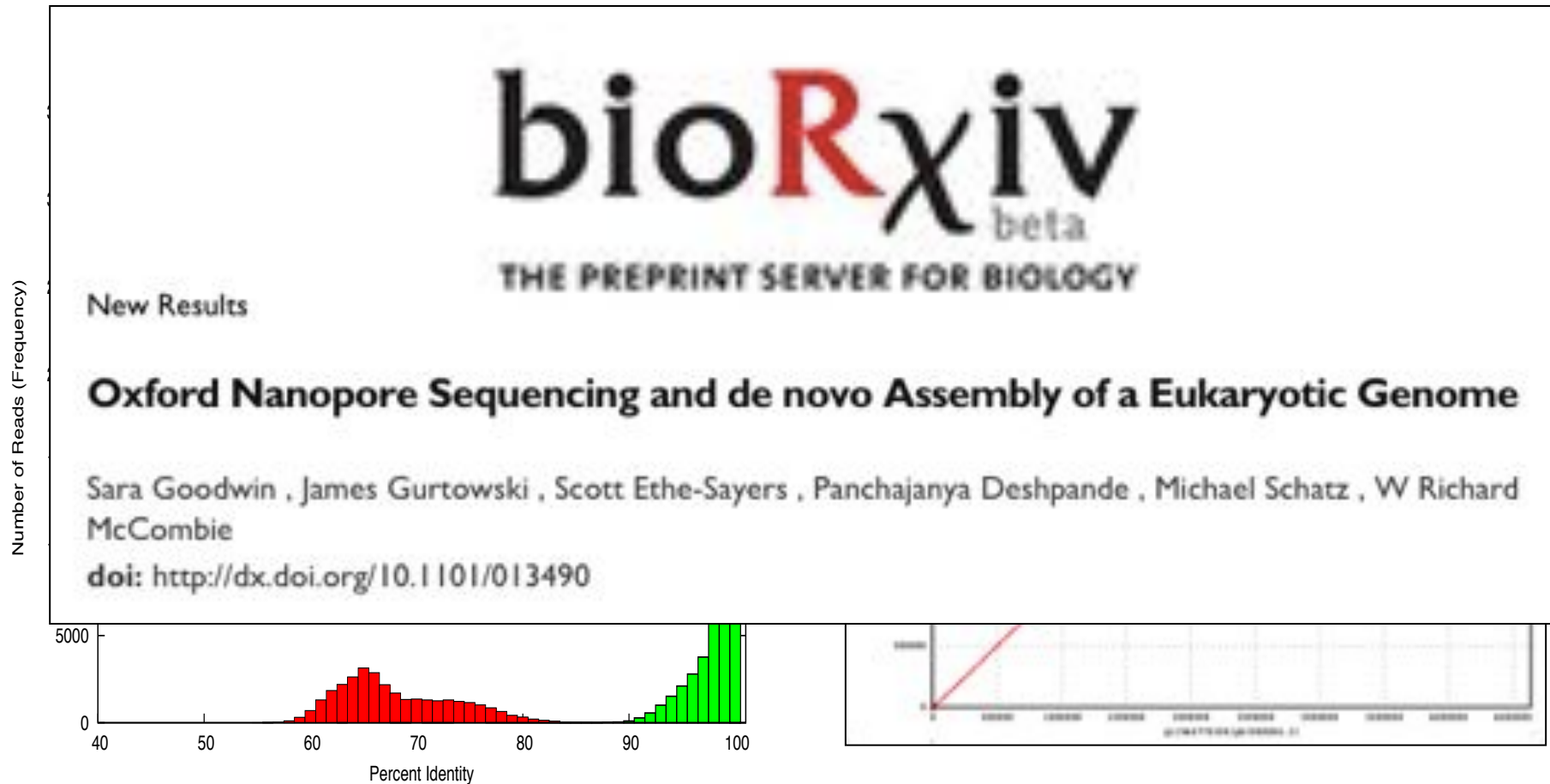
# NanoCorr E. coli K12 Assembly

## Nanocor Correction Results

145x Oxford Nanopore X 35x MiSeq

## Single Contig Assembly

99.99% Identity (Pilon polishing)



Sequencing Data From:

**A reference bacterial genome dataset generated on the MinION™ portable single-molecule nanopore sequencer**  
Joshua Quick, Aaron R Quinlan and Nicholas J Loman

# Genomic Futures?



Zamin Iqbal and 5 others retweeted



**GenomeWeb InSequence** @InSequence · Oct 20

Oxford Nanopore shows off **PromethION** at **ASHG**. #ASHG14 #nanopore



# Genomic Futures?



# iGenomics: Mobile Sequence Analysis

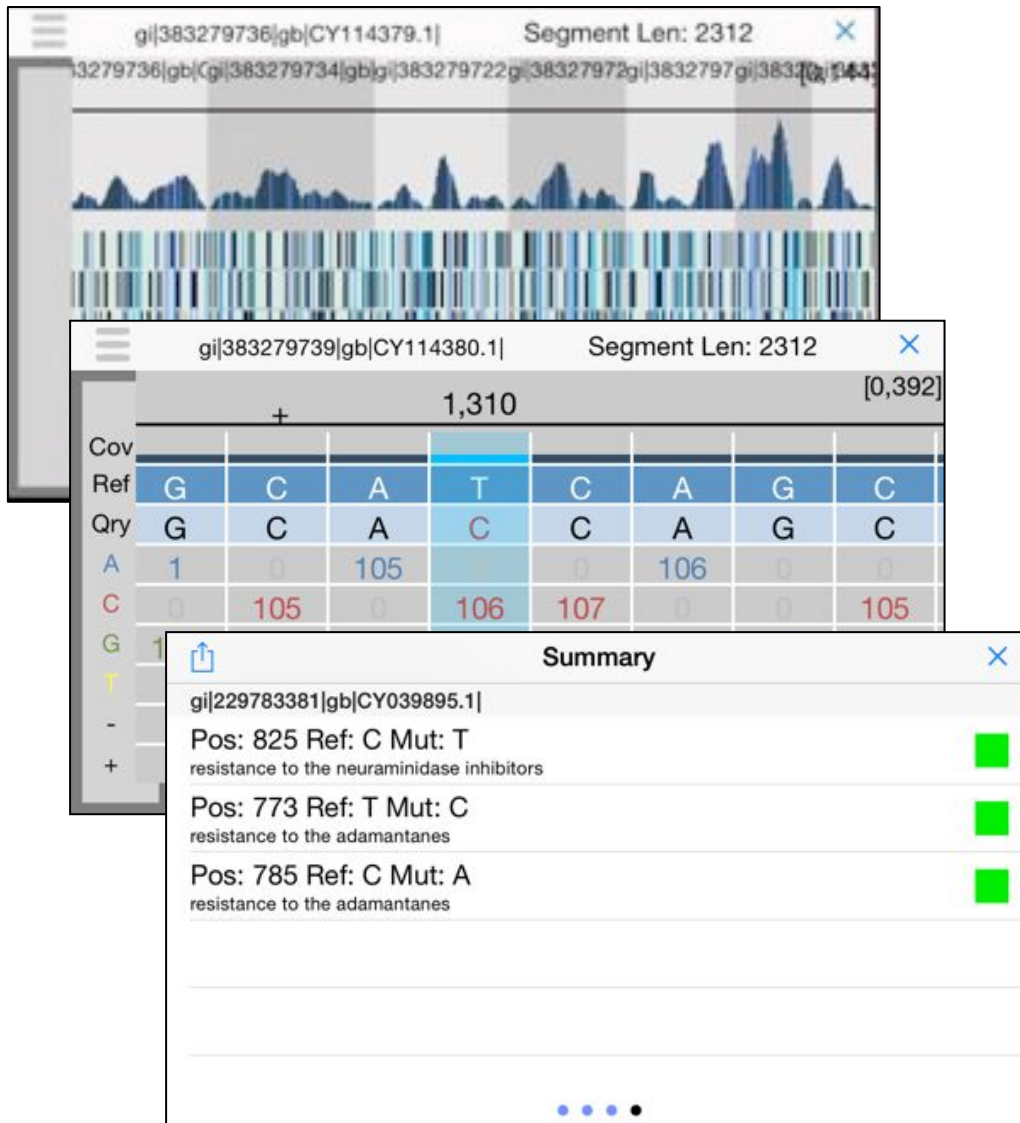
Aspyn Palatnick, Elodie Ghedin, Michael Schatz

*The worlds first genomics analysis app for iOS devices*

*BWT + Dynamic Programming + UI*

First application:

- Handheld diagnostics and therapeutic recommendations for influenza infections
- In the iOS AppStore now!



## **Future applications**

- Pathogen detection
- Food safety
- Biomarkers
- etc..



# Summary & Recommendations

## ***Reference quality genome assembly is here***

- Use the longest possible reads for the analysis
- Don't fear the error rate, coverage and algorithmics conquer most problems

## ***Megabase N50 improves the analysis in every dimension***

- Better resolution of genes and flanking regulatory regions
- Better resolution of transposons and other complex sequences
- Better resolution of chromosome organization
- Better sequence for all downstream analysis

***The year 2015 will mark the return to reference quality genome sequence***



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Lyon Lab  
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# Thank you

<http://schatzlab.cshl.edu>

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