



Cold Spring Harbor Laboratory

Current Advances in Sequencing Technology

James Gurtowski
Schatz Lab

Outline

1. Assembly Review

2. Pacbio

Technology Overview

Data Characteristics

Algorithms

Results – Assemblies

3. Oxford Nanopore

Technology Overview

Data Characteristics

Algorithms

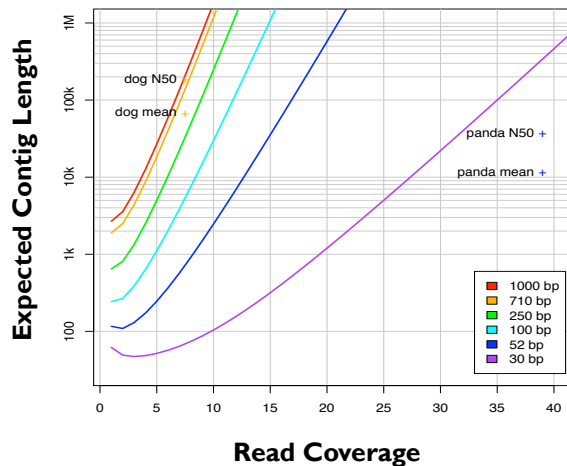
Results – Assemblies

4. Summary



Ingredients for a good assembly

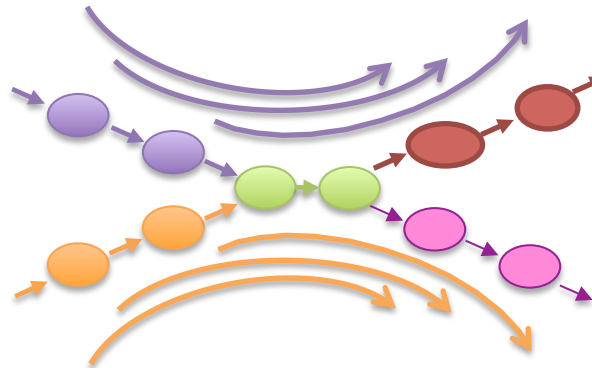
Coverage



High coverage is required

- Oversample the genome to ensure every base is sequenced with long overlaps between reads
- Biased coverage will also fragment assembly

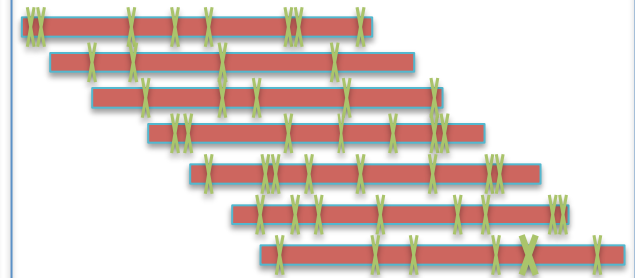
Read Length



Reads & mates must be longer than the repeats

- Short reads will have **false overlaps** forming hairball assembly graphs
- With long enough reads, assemble entire chromosomes into contigs

Quality



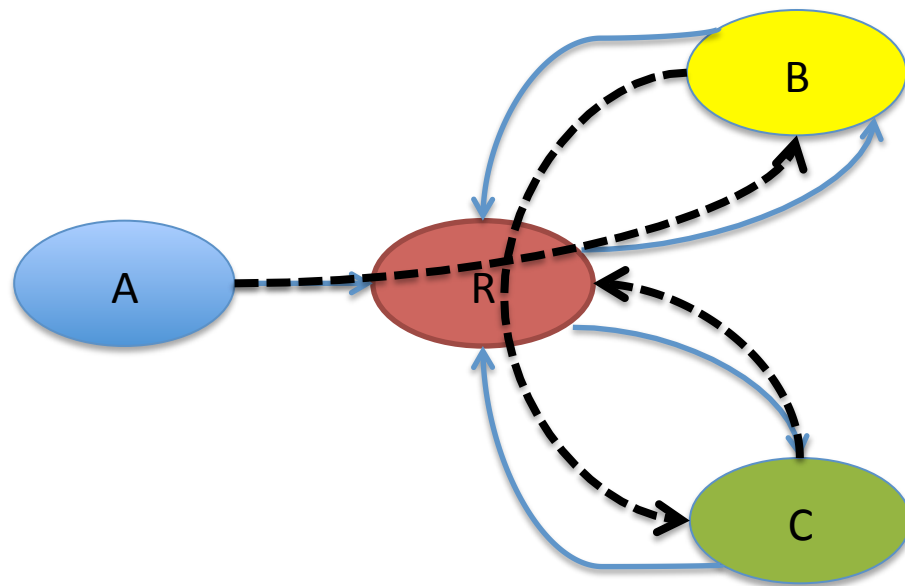
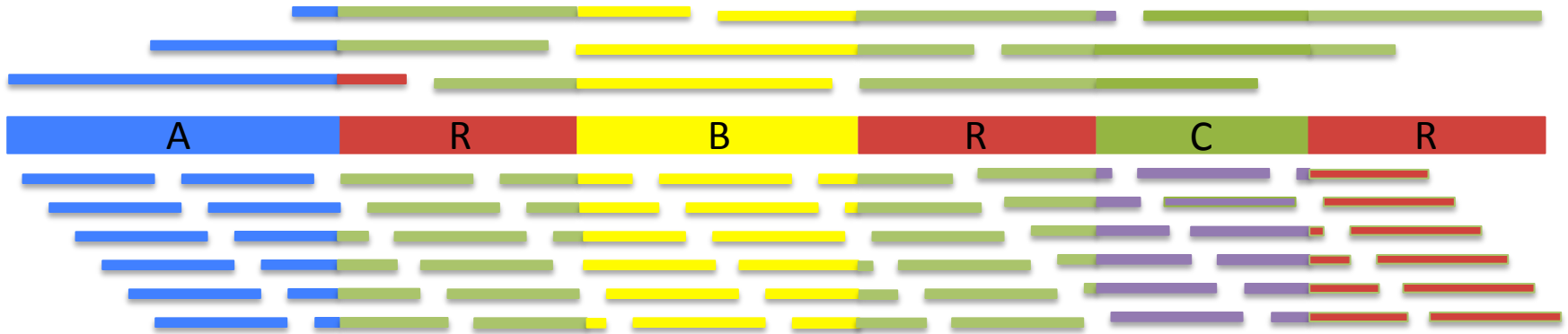
Errors obscure overlaps

- Reads are assembled by finding kmers shared in pair of reads
- High error rate requires very short seeds, increasing complexity and forming assembly hairballs

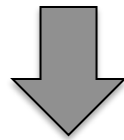
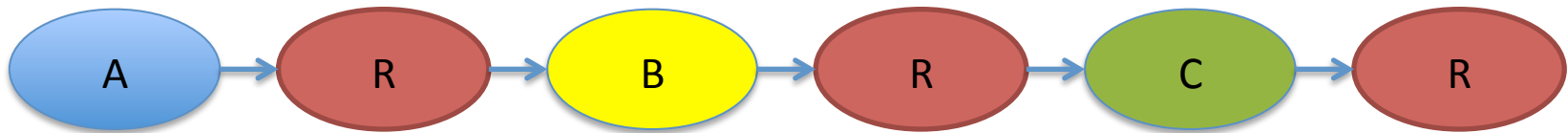
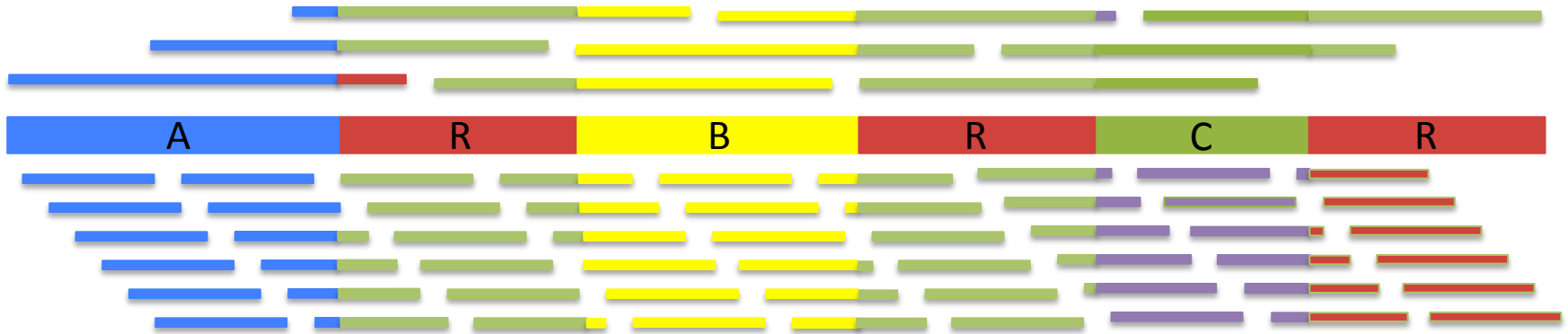
Current challenges in *de novo* plant genome sequencing and assembly

Schatz MC, Witkowski, McCombie, WVR (2012) *Genome Biology*. 12:243

Assembly Complexity



Assembly Complexity



The advantages of SMRT sequencing

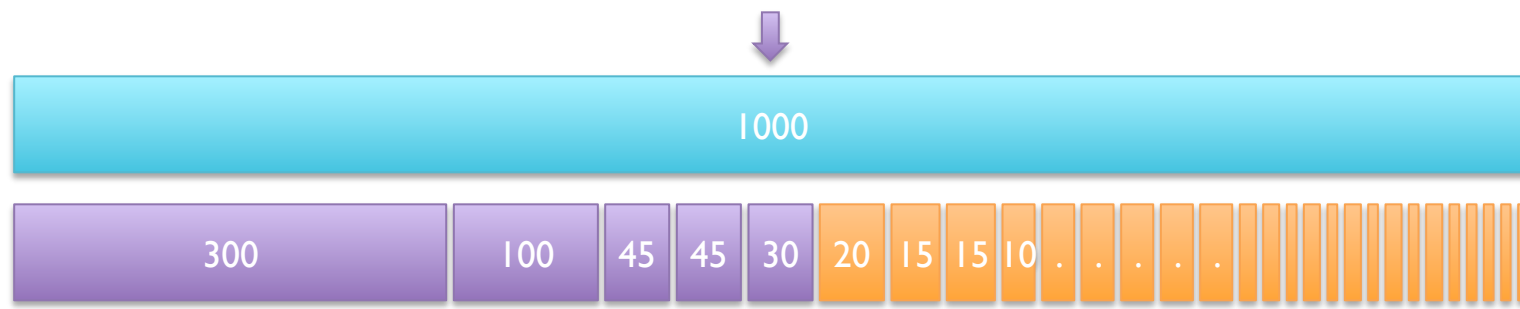
Roberts, RJ, Carneiro, MO, Schatz, MC (2013) *Genome Biology*. 14:405

N50 size

Def: 50% of the genome is in contigs as large as the N50 value

Example: 1 Mbp genome

50%



N50 size = 30 kbp

(300k+100k+45k+45k+30k = 520k \geq 500kbp)

A greater N50 is indicative of improvement 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

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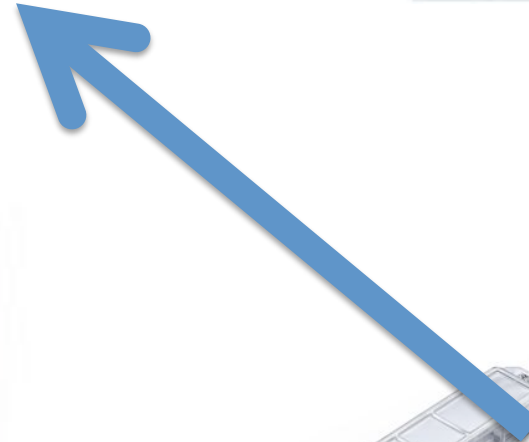
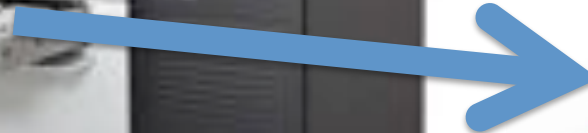
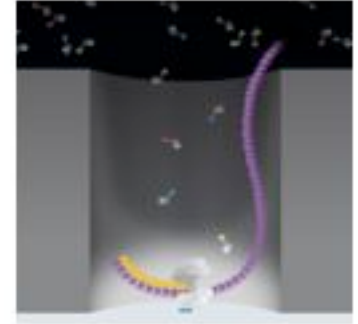
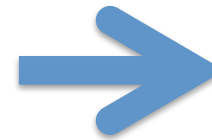
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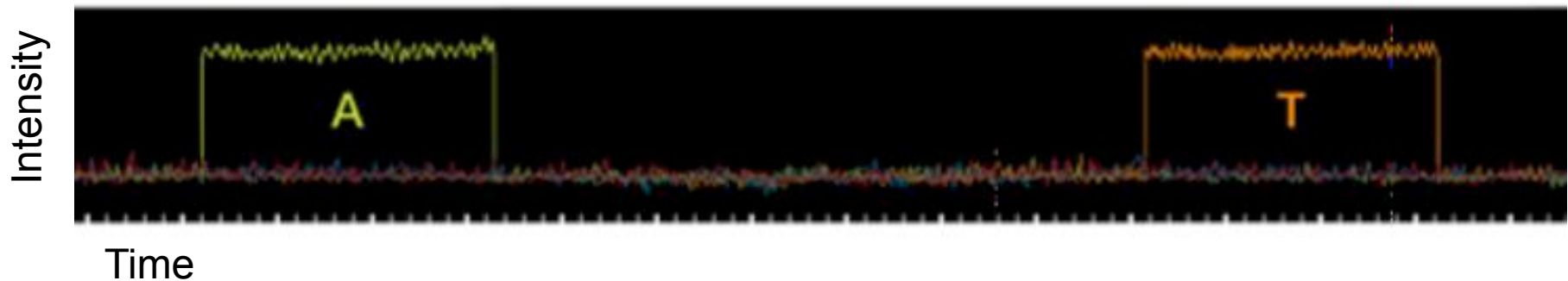
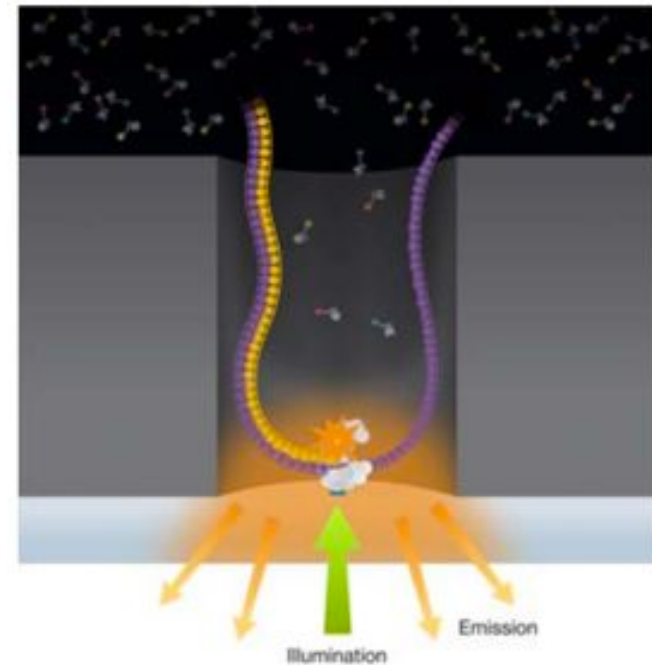
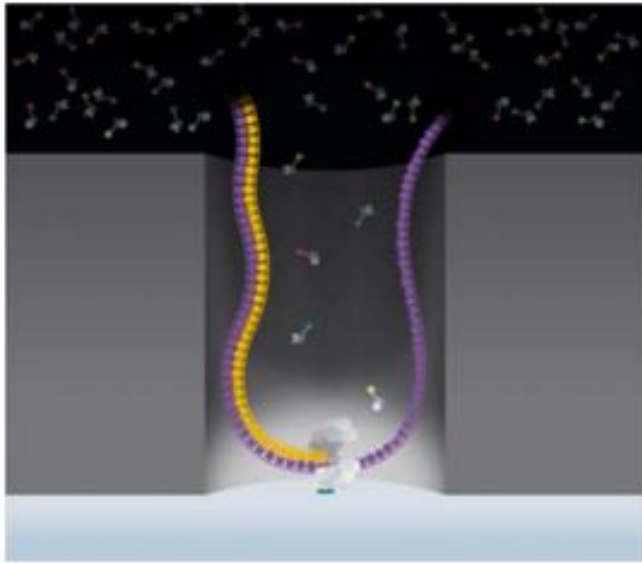
4. Summary



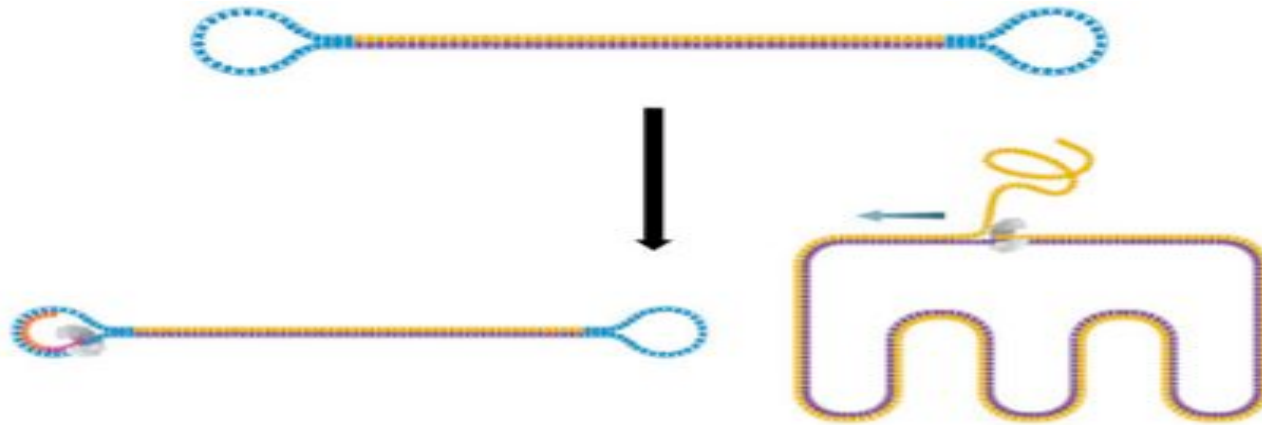


SMRT Sequencing

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



SMRT Read Types



- **Standard sequencing**

- Long inserts so that the polymerase can synthesize along a single strand

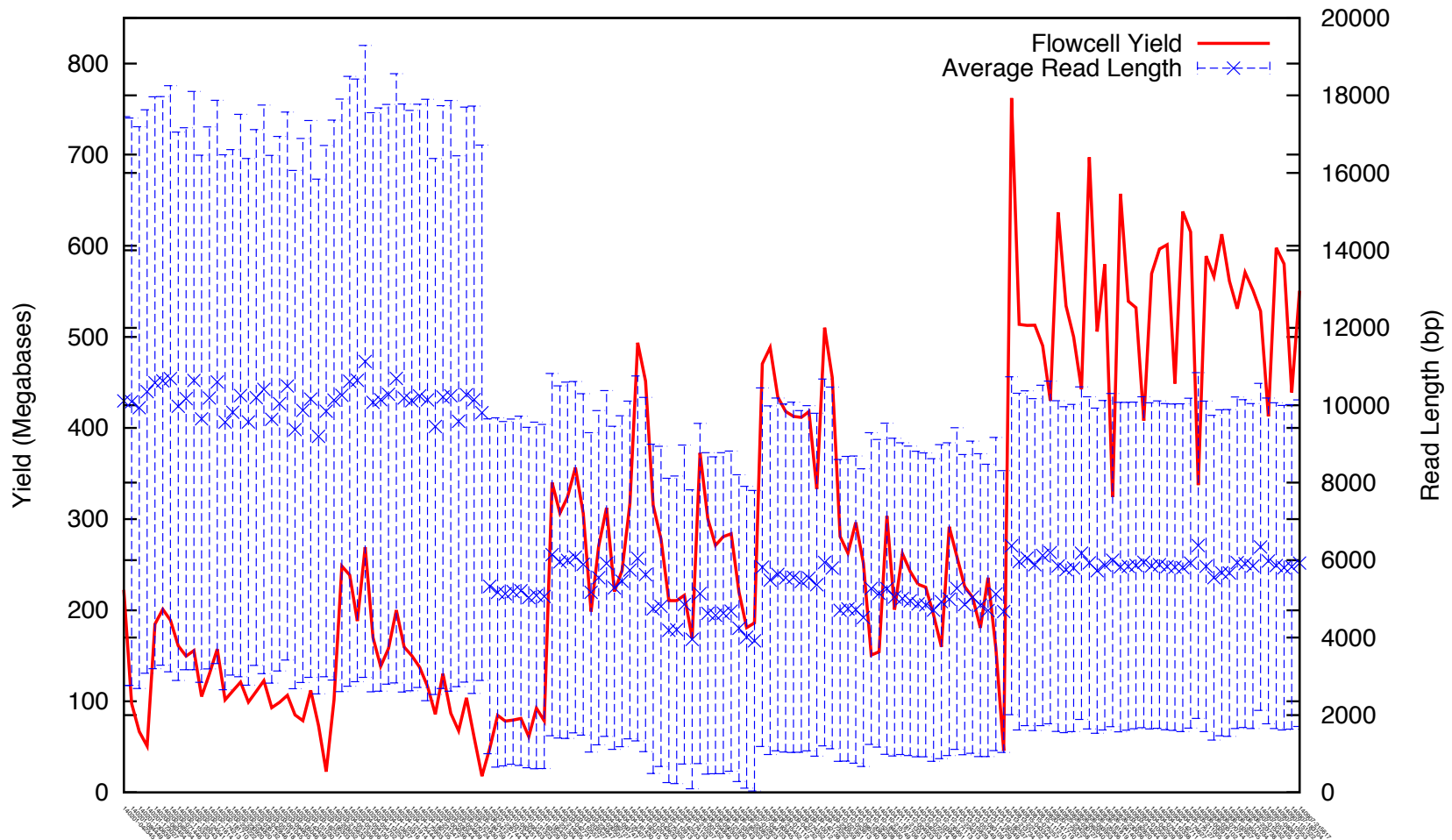
- **Circular consensus sequencing**

- Short inserts, so polymerase can continue around the entire SMRTbell multiple times and generate multiple sub-reads from the same single molecule.
- Barbell sequence: ATCTCTCTCttttcctcctcctccgttggttggttGAGAGAGAT

Example Flowcell Yields

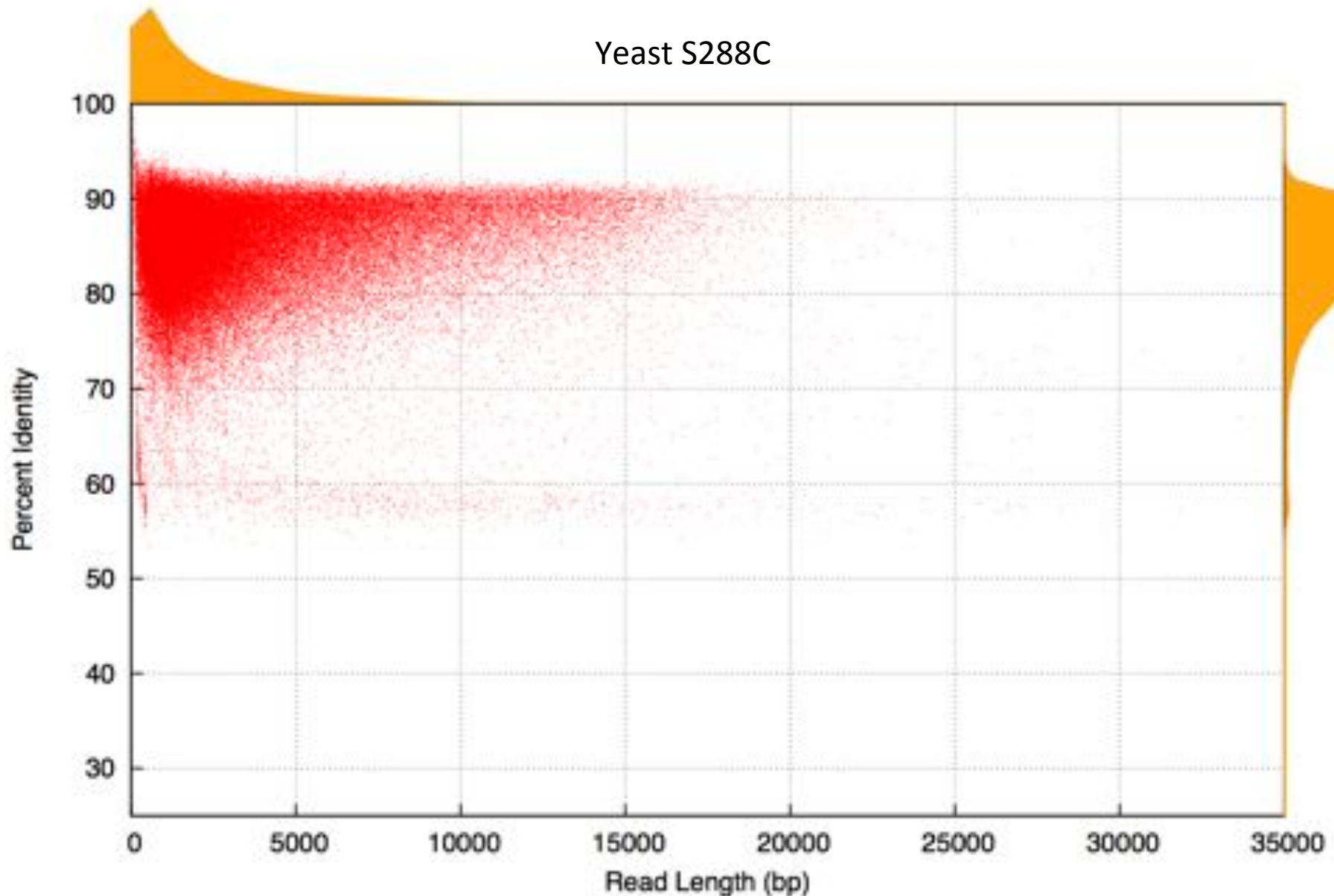
Rice Strain IR64 (Feb – Sep 2014)

Pacbio Flowcell Yields



Read Length Distribution and Identity

Yeast S288C



Average Error Rate	16%
Mismatches	30.0%
Deletions	11.5%
Insertions	58.5%

> chr7
Length=1098948

Score = 238 bits (1072), Expect = 3e-68
Identities = 612/809 (76%), Gaps = 172/809 (21%)
Strand=Plus/Minus

```

Query 28      TTCTCCTTGTTTCTTTTCAACCTATAAAAACGTTTTTTATTGACGCATAGACAACCTAAGCCA 87
Sbjct 1058792  TTCTCCTTGTTT-TTTECAA--TAT-----TTTATTGA-G-ATAGACAACCTAA--CCA 1058744

Query 88      ACAACGAAACAAACCCGGTCAAGATCGGCGCGTCTGCTACTACCATTCTTCGCTATG 147
Sbjct 1058743  A-AAAGAA--AA--CGGTCAAAGAT----GCCATTCT-CAGCA--C-TGGCTA-- 1058700

Query 148     CTGCAGGCTACTTGGCTAGCTTATTTTTCAGCGAGCTTCTACCCTCAAGGTTTCAGCTAT 207
Sbjct 1058699  CTGCA-GCTAC-TGC---GC-TATTTTTC-A-C-AGC--CT-CC--CAA-GTTTCAGCTAT 1058654

Query 208     TGGTGAAGTACGCTTTAAGTCTGGGTGTGCTAAGTGAACAACGATAGGTAGCTGTGAAG 267
Sbjct 1058653  TGGTGAAGTACGCTTTAACT-TGGGTGT-C-AA--CAACAGGAT-GGTA-CT-TGTAAG 1058602

Query 268     TCCACTCCTTCGGAACCTATTGTAAAGATATACAAGGCTTTGACTATAGCAGCTTCCAA 327
Sbjct 1058601  TCCA---CTTC-CG-AC-TATGAAACCGAATTACAA-GCTTTGAGAGCTACACTTCC-A 1058559

Query 328     CCGAGCTTAAAGTTATACTGCTGTGCTCAGATCGTAACAGCTTTTGCAAAACTTAGG 387
Sbjct 1058549  CC-GTC---AAAGTT-TAC-GC--TGCCTCAGAT-TGTAACA-C-TTGCAAAACTTA-G 1058502

Query 388     GTCTGCGCCTGTCTCTGAGGGCTGAGGACTATTTACTCATGCTTATAGTCGGTGTGTT 447
Sbjct 1058501  GTCTG-----CTCTGAA--GCTGAGG--GATTTACT-AT-CTT-TCTCGGTGTT 1058457

Query 448     GGCGCAGAGGACGACAGTGCAGTGATACGCTGCTG-AAAGGCCTTCGTTTGLAAR 506
Sbjct 1058456  GGC--CAACAGA-CGACAGT-CA-T--TACGCTGCTGAARAGGC-CC----TTTGLAAR 1058409

Query 507     CCTAATTTGGCGCTAAAATTAANAACCAATCCACTGTTAGGCTGGTTTTCTTGGTTGGGA 566
Sbjct 1058408  CCT-ATTT----GCCAAAATTA-AAAG-AATCCACTGTT--GCTGG-TTTCTTGTTGG-- 1058360

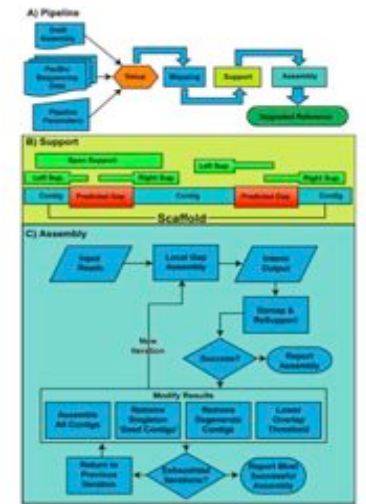
```

Error Profile
Dominated by
Insertions



Long Read Correction Algorithms

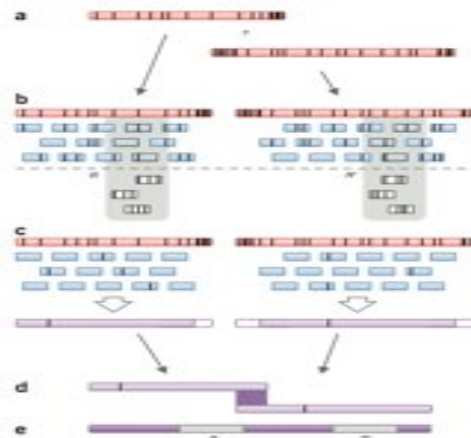
PBJelly



Gap Filling and Assembly Upgrade

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

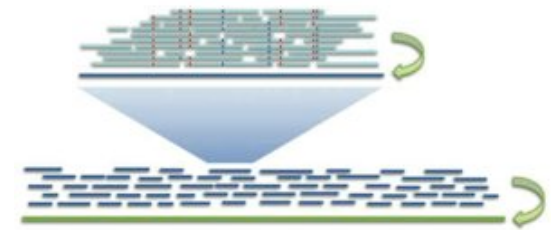
PacBioToCA & ECTools



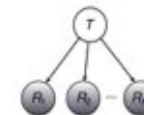
Hybrid Error Correction

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

LR-only Correction & Polishing

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

< 5x

Long Read Coverage

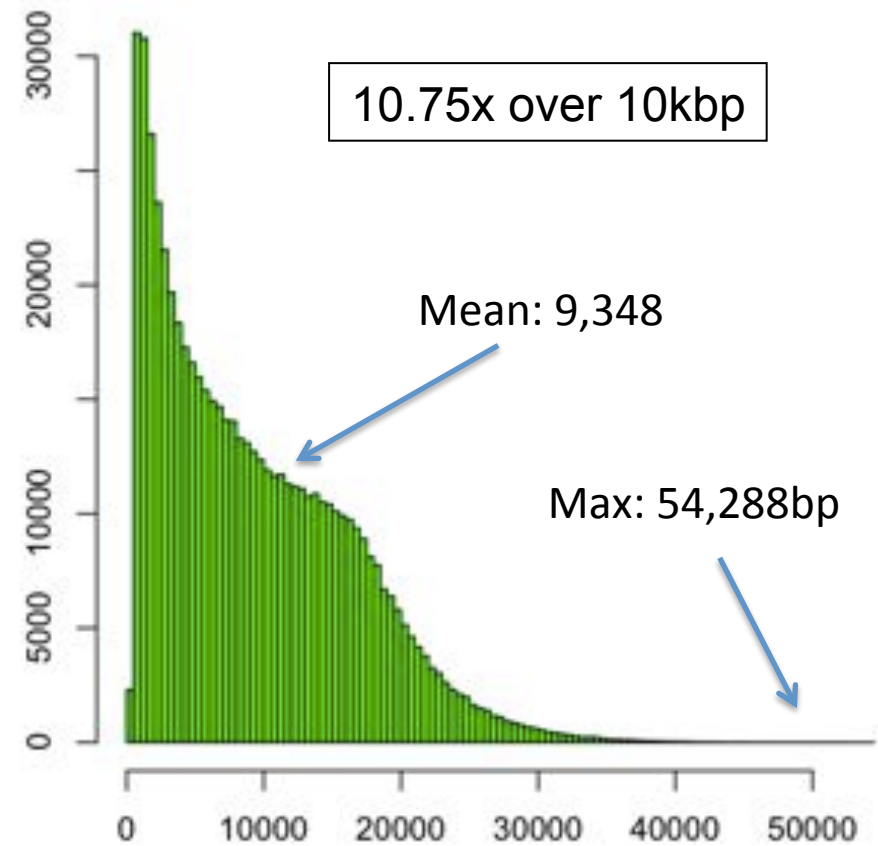
> 50x

O. sativa pv Indica (IR64)

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

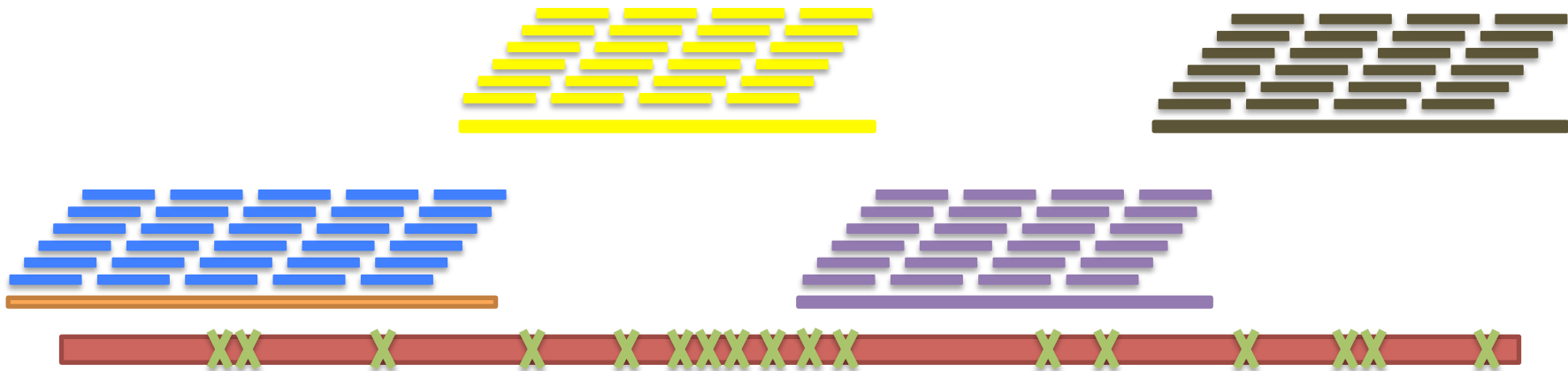


Assembly	Contig NG50
“ALLPATHS-recipe” 50x 2x100bp @ 180 36x 2x50bp @ 2100 51x 2x50bp @ 4800	18,450
MiSeq Fragments 25x 456bp (3 runs 2x300 @ 450 FLASH)	19,078
PacbioToCA – 47 SMRTCells 10.7x @ 10kbp	144,042
ECTools - 47 SMRTCells 10.7x @ 10kbp	????



ECTools: Error Correction with pre-assembled reads

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



Short Reads -> Assemble Unitigs -> Align & Select -> Error Correct

Can Help us overcome:

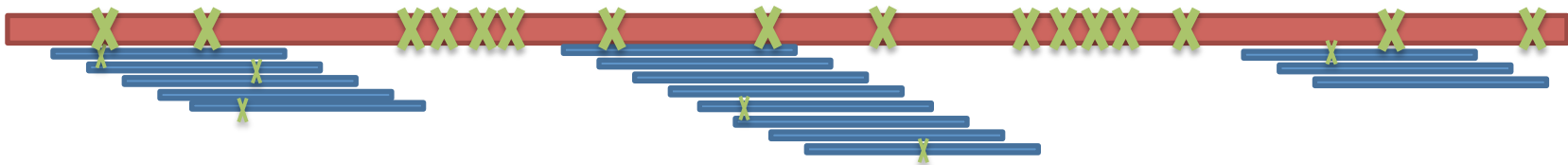
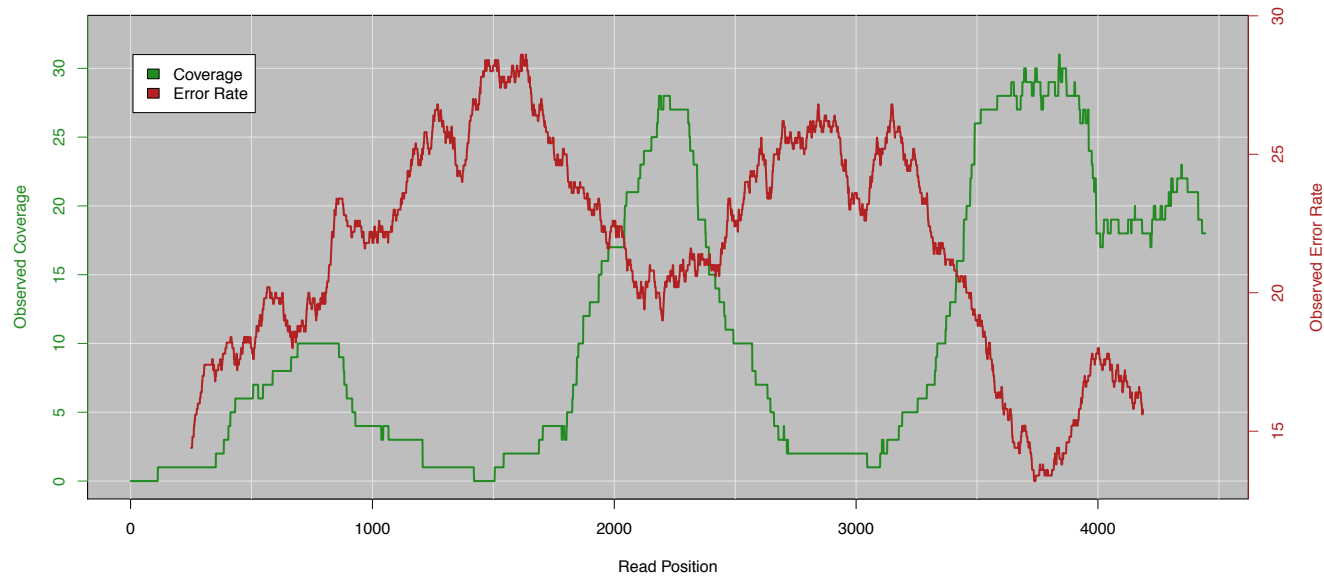
1. Error Dense Regions – Longer sequences have more seeds to match
2. Simple Repeats – Longer sequences easier to resolve

However, cannot overcome Illumina coverage gaps & other biases

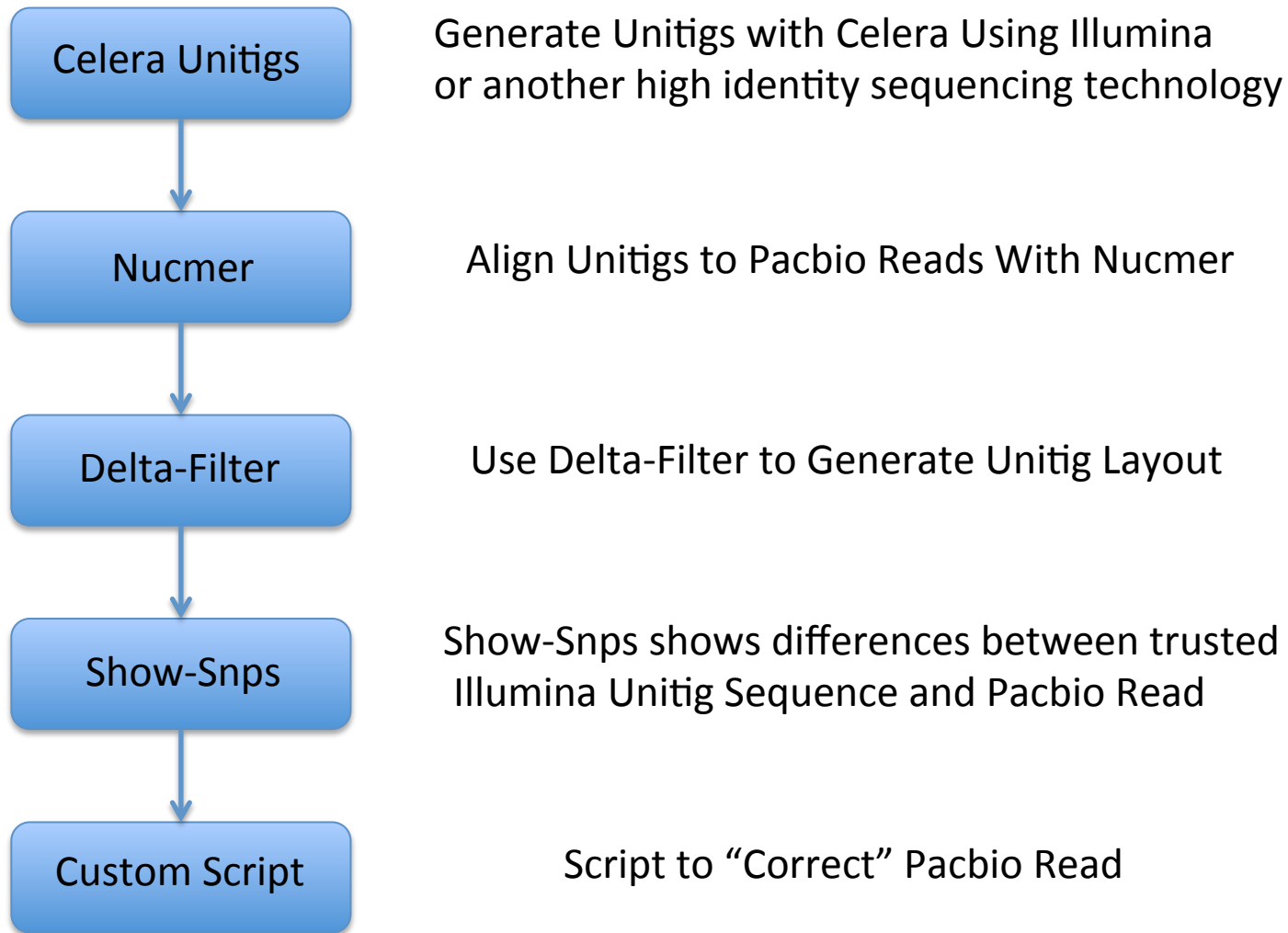
Low Coverage Regions

1. Simple Repeats – Kmer Frequency Too High to Seed Overlaps
2. GC Rich Regions – Known Illumina Bias
3. **Error Dense Regions – Difficult to compute overlaps with many errors**

Position Specific Coverage and Error Rate



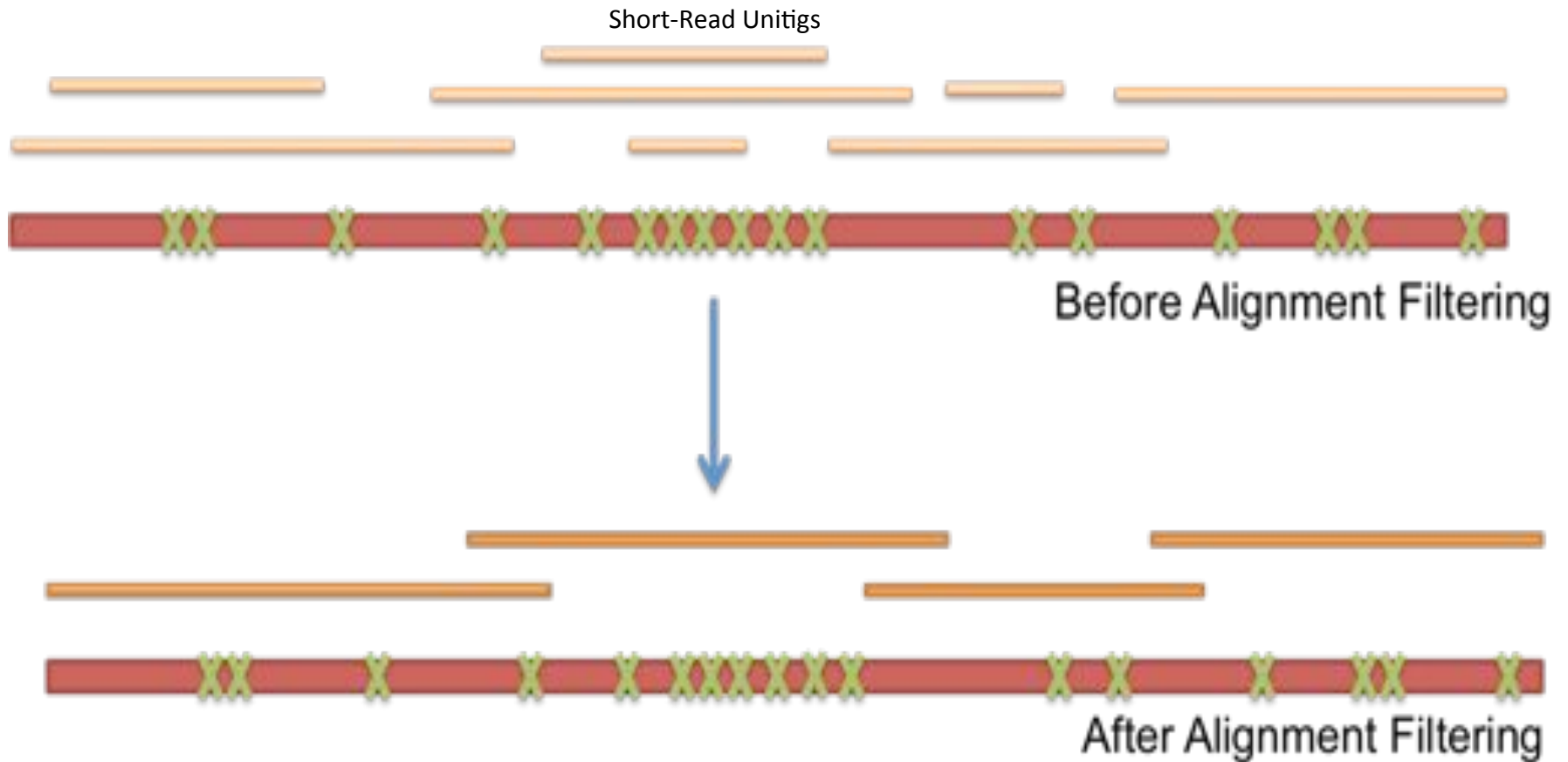
ECTools Pipeline



Note: Reads are never split or trimmed

Delta-Filter Alignment filtering

Uses Dynamic Programming (Longest Increasing Subset) to find the longest mutually consistent subset of unitigs with respect to the Pacbio Read

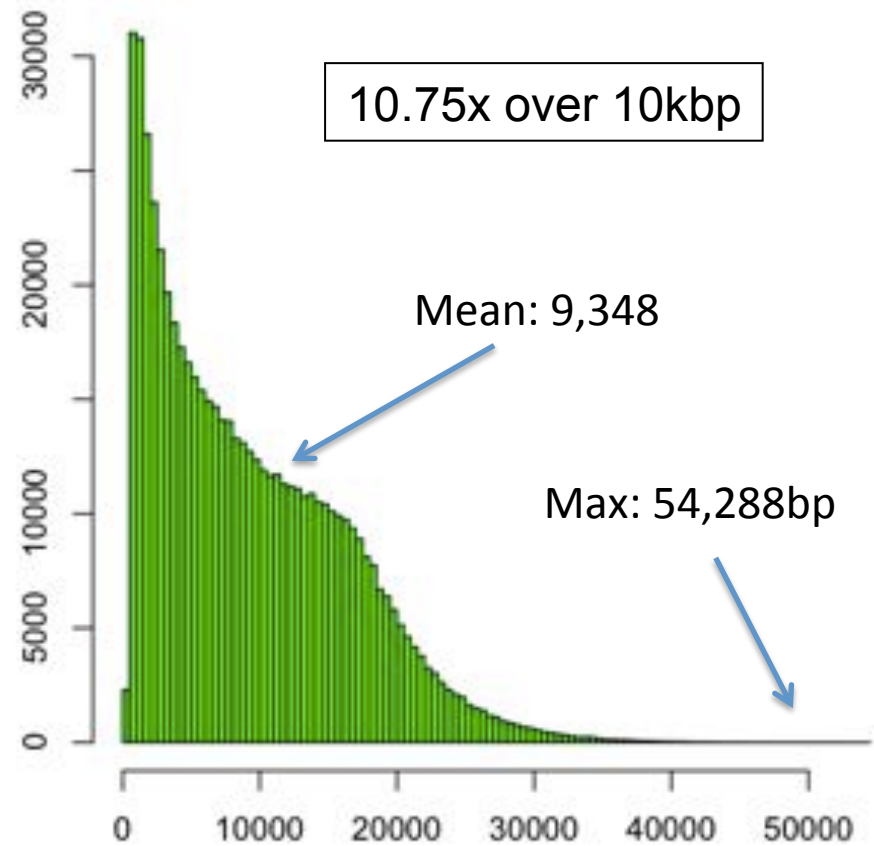


O. Sativa Indica (IR64)

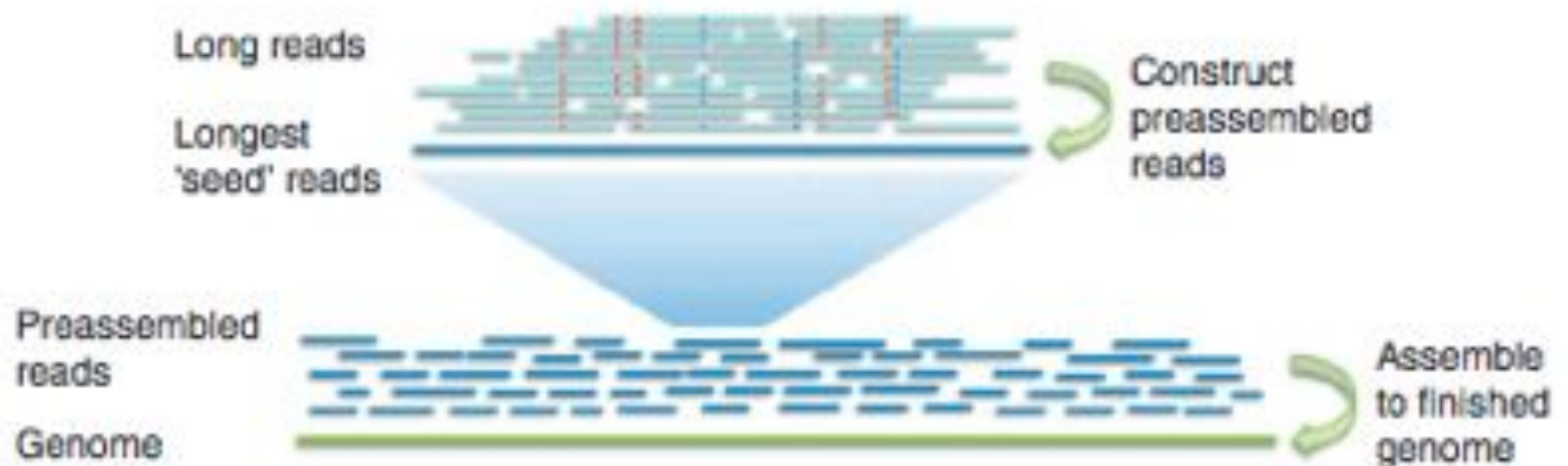
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PacbioToCA – 47 SMRTCells 10.7x @ 10kbp	144,042
ECTools - 47 SMRTCells 10.7x @ 10kbp	272,137



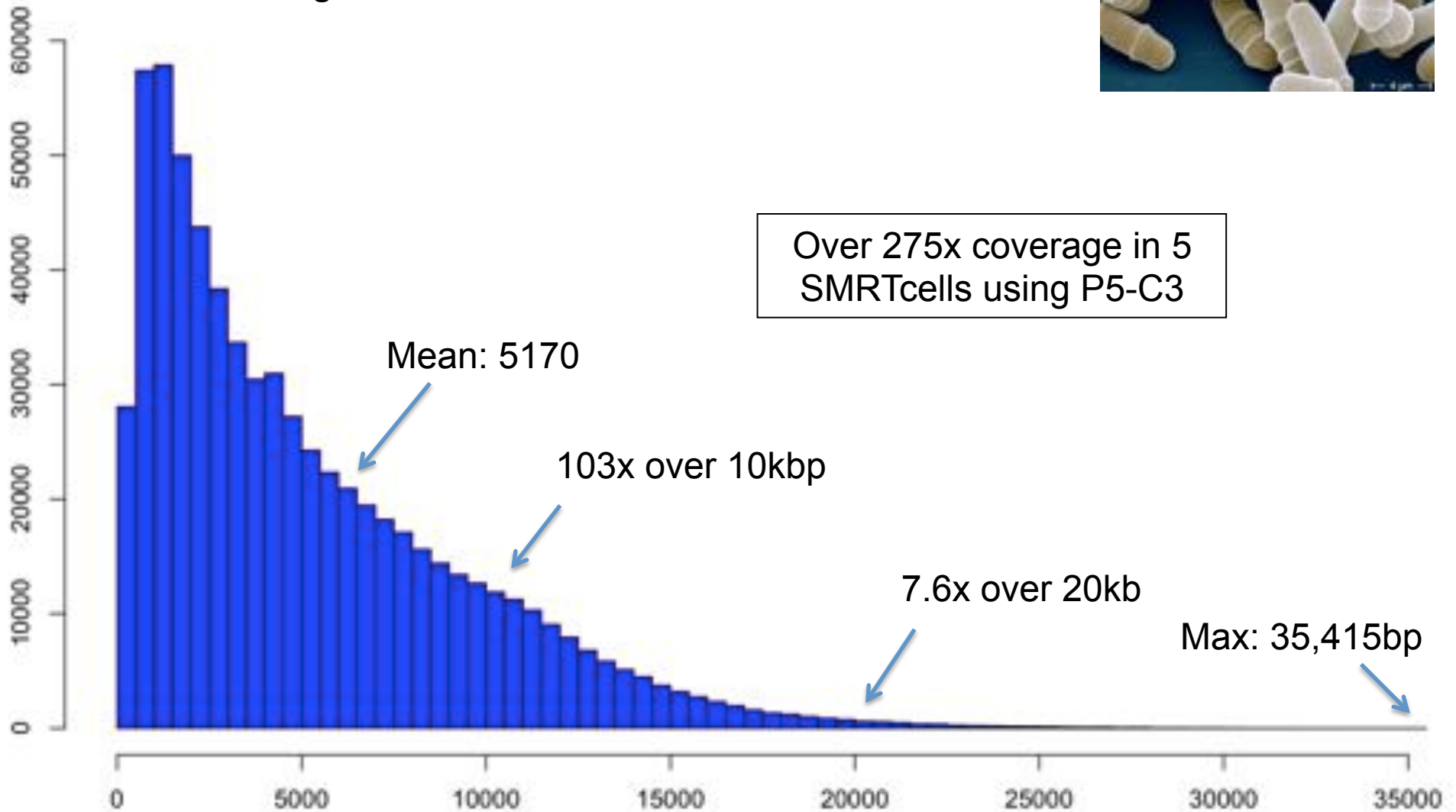
HGAP Error Correction



S. pombe dg2 I

PacBio RS II sequencing at CSHL

- Size selection using an 7 Kb elution window on a BluePippin™ device from Sage Science



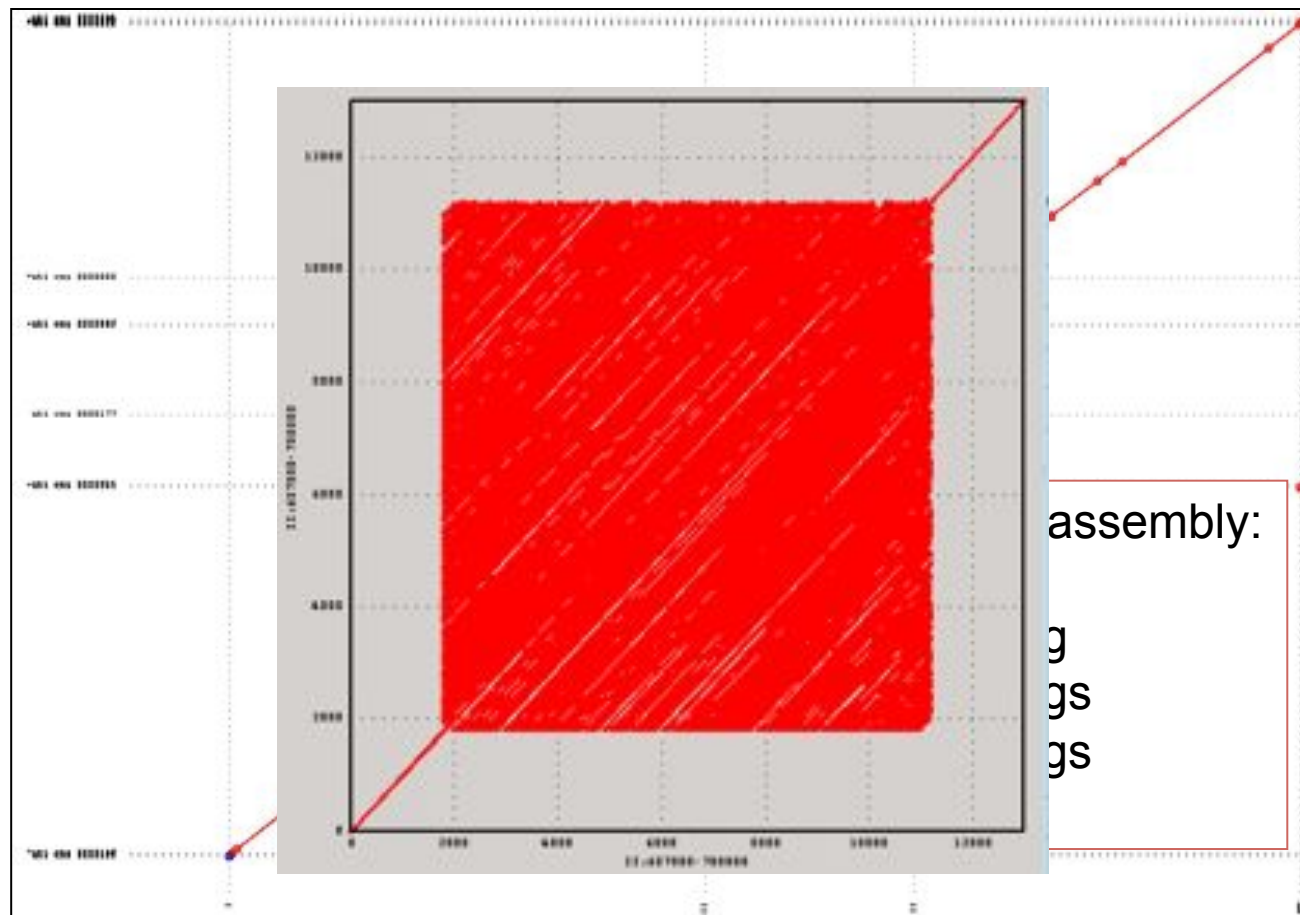
S. pombe dg21

ASM294 Reference sequence

- 12.6Mbp; 3 chromo + mitochondria; N50: 4.53Mbp

PacBio assembly using HGAP + Celera Assembler

- 12.7Mbp; 13 non-redundant contigs; N50: 3.83Mbp; >99.98% id



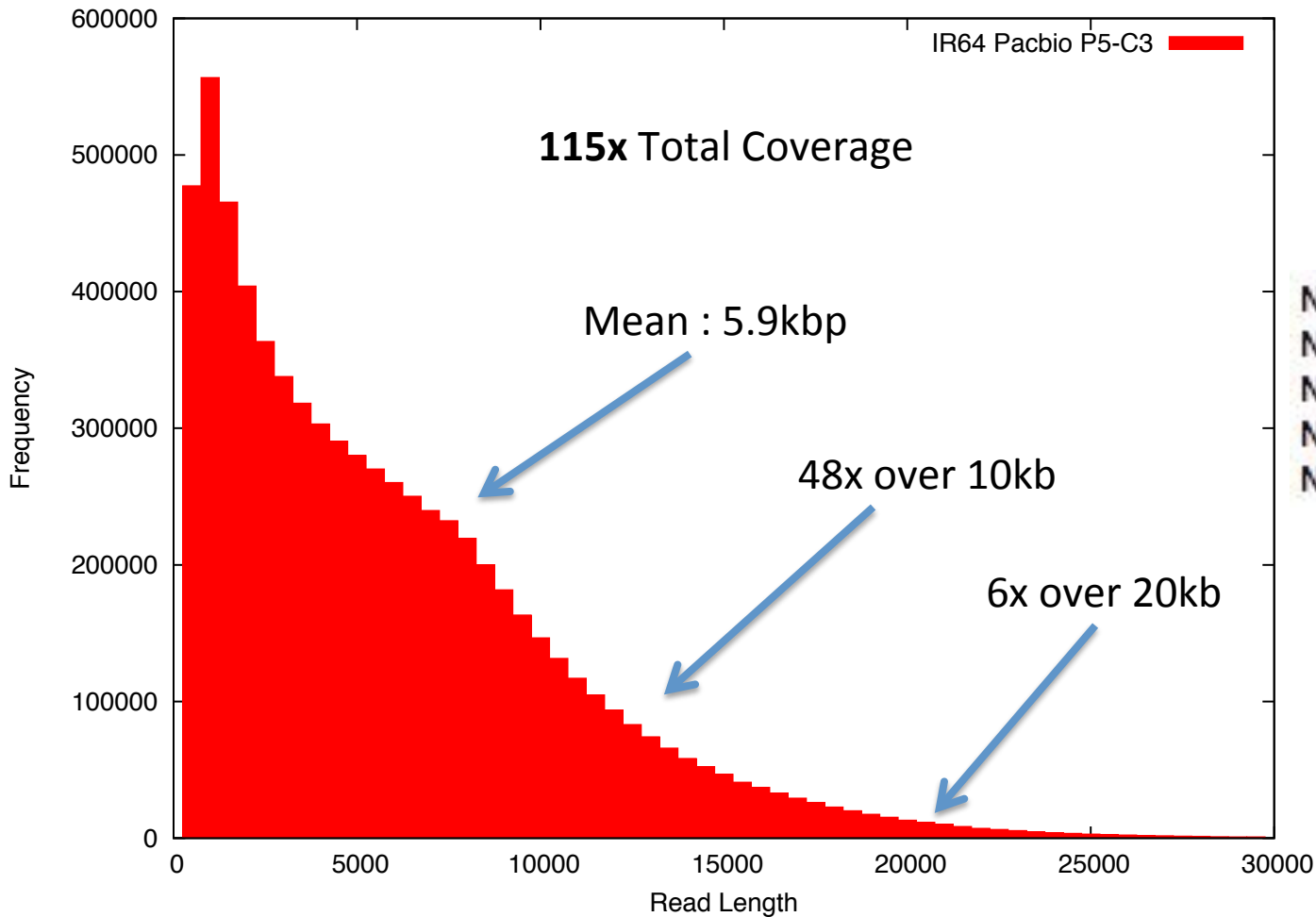
O. sativa pv Indica (IR64)

Genome size: ~370 Mb

Chromosome N50: ~29.7 Mbp



September 2014



Assembly Results

N10=10425102 N10cnt=3
N25=6571607 N25cnt=11
N50=3900937 N50cnt=29
N75=1783229 N75cnt=66
N90=859087 N90cnt=108

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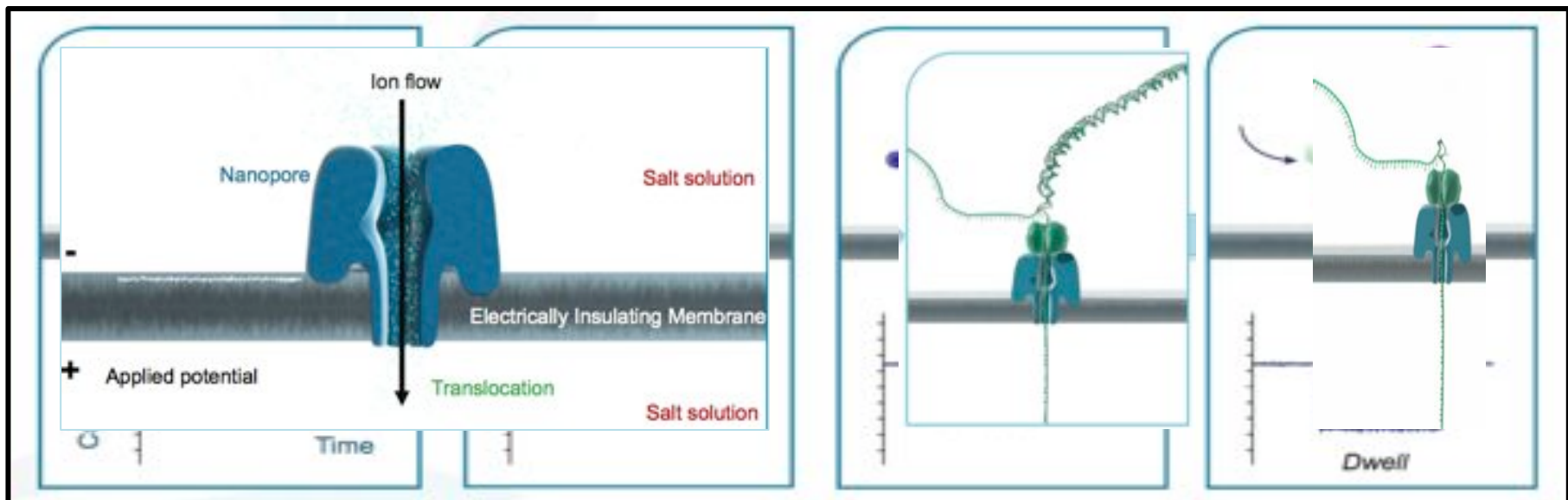
4. Summary



Oxford Nanopore MinION



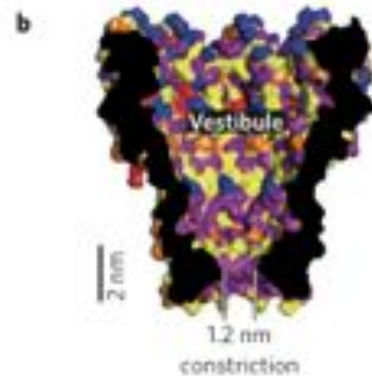
- MAP Program
- Thumb drive sized sequencer powered over USB
- Senses DNA by measuring changes to ion flow
- Reads both DNA Strands (2D)



Advantages And Challenges of Nanopore DNA Sequencing

Advantages

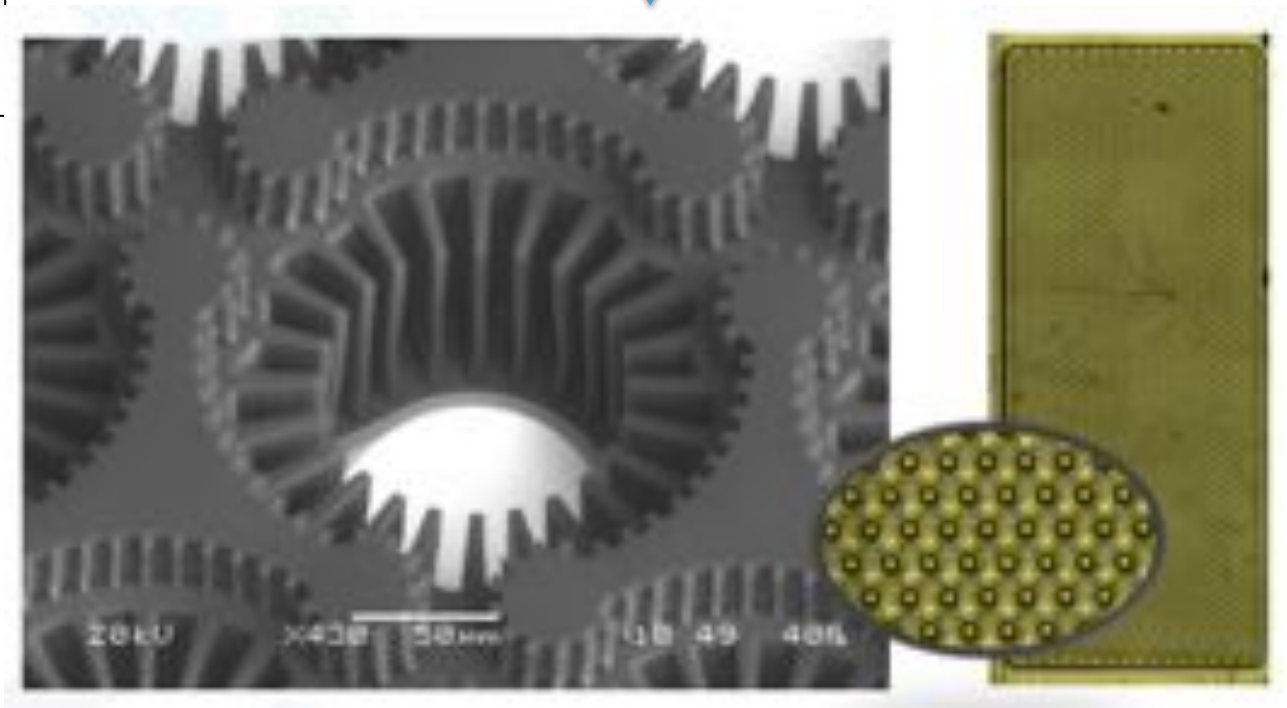
- Label-free
- Amplification-free
- Single-Molecule
- High-Throughput
- Inexpensive Instrument
- Simple/quick Sample Prep
- Produces Long Reads



Challenges

- Controlling rate at which DNA translocates through the pore (1base/microsecond too fast to accurately measure current change)
- Pore does not have single base resolution (complicates basecalling and makes it hard to deal with modified bases)
- Commercially: Biological pores are sensitive to pH, temperature, salt concentration

Under the hood of the MinION



Oxford Nanopore DNA Prep



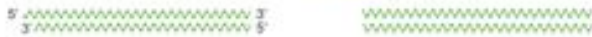
High molecular weight DNA >30 kb



Shear



Fragments
• 3' overhangs
• 5' overhangs
• Blunt ends



End-repair
dA-tail



Purify



Add Adapters and
Motor Protein



Ligate
Purify



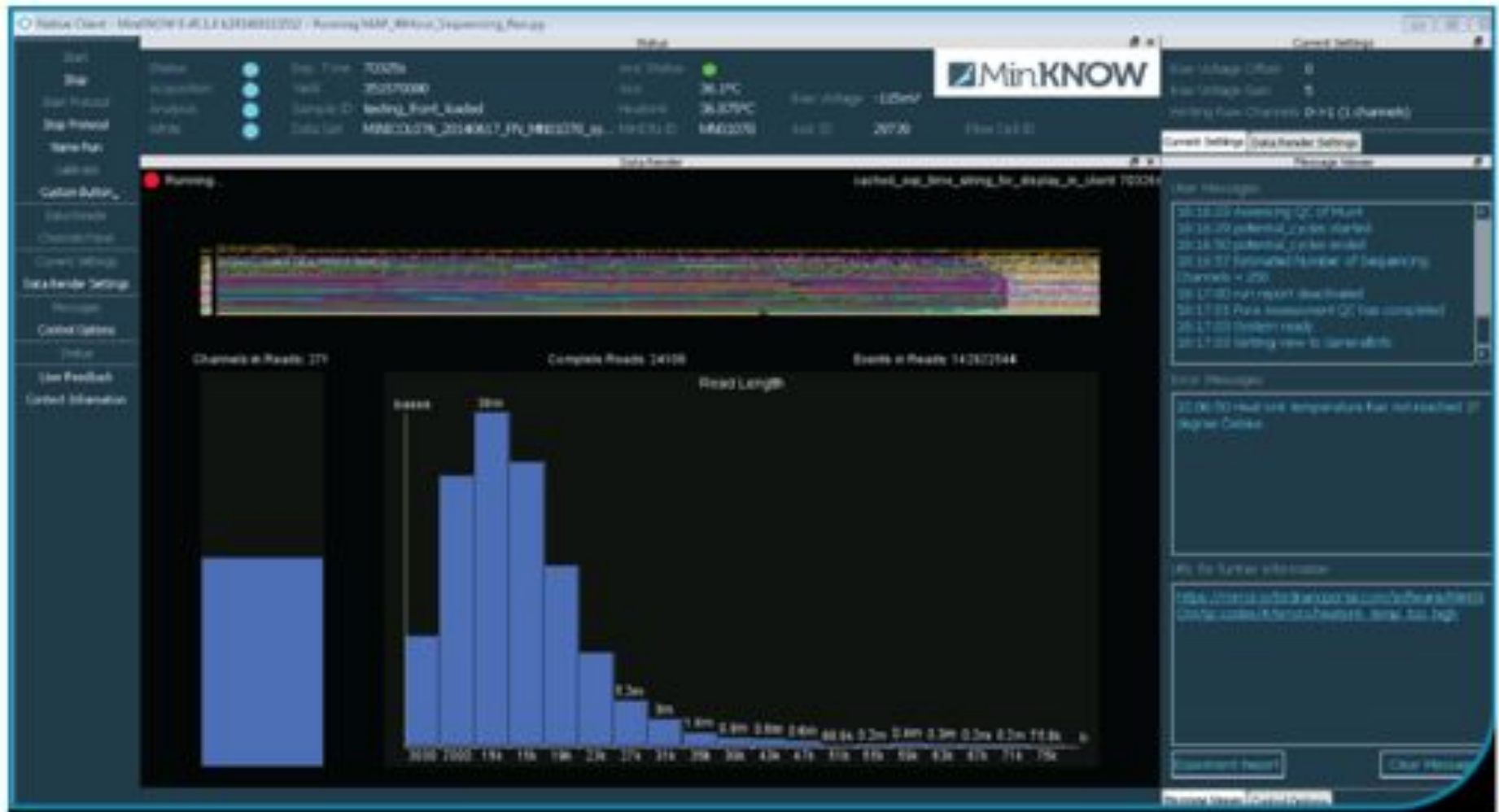
Condition the fragments
for nanopore sequencing

Simple DNA prep

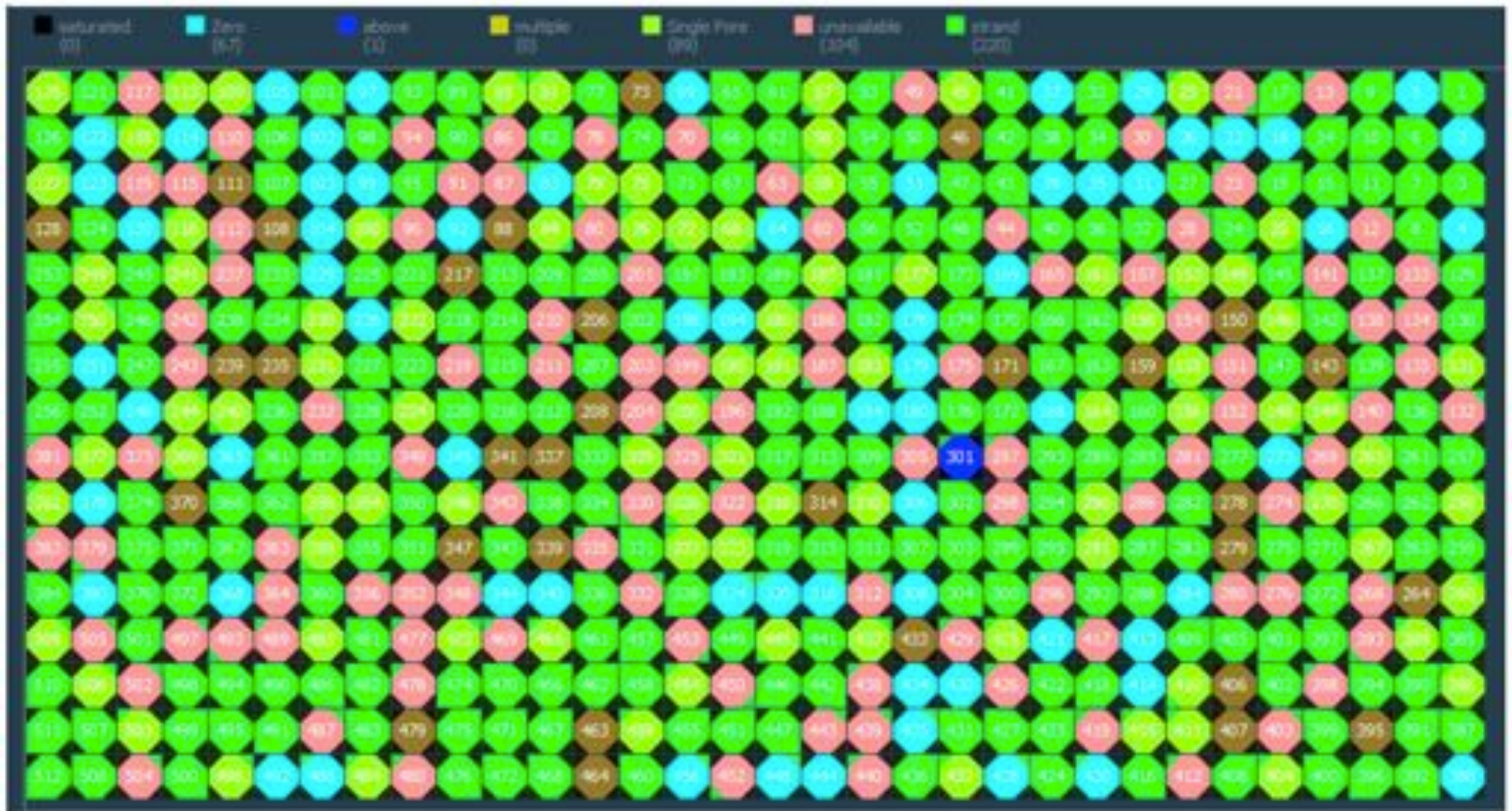
Can do it in the field?



Nanopore Desktop Software



View of Pore Activity



Base Calling

Local Software Agent facilitates data transaction with cloud basecaller

Raw Signal data contained in fast5 (hdf5) files on local machine

Name	Date modified	Type	Size
TEST12345_ch1_file1	14-04-2014 13:38	Fast5 file	257 KB
TEST12345_ch1_file2	14-04-2014 13:38	Fast5 file	200 KB
TEST12345_ch1_file3	14-04-2014 13:38	Fast5 file	276 KB
TEST12345_ch1_file4	14-04-2014 13:38	Fast5 file	242 KB
TEST12345_ch2_file1	14-04-2014 13:39	Fast5 file	204 KB
TEST12345_ch2_file2	14-04-2014 13:39	Fast5 file	243 KB
TEST12345_ch2_file3	14-04-2014 13:39	Fast5 file	246 KB
TEST12345_ch2_file4	14-04-2014 13:39	Fast5 file	181 KB
TEST12345_ch3_file1	14-04-2014 13:39	Fast5 file	369 KB
TEST12345_ch3_file2	14-04-2014 13:40	Fast5 file	399 KB
TEST12345_ch3_file3	14-04-2014 13:40	Fast5 file	363 KB
TEST12345_ch3_file4	14-04-2014 13:40	Fast5 file	311 KB
TEST12345_ch4_file1	14-04-2014 13:40	Fast5 file	464 KB
TEST12345_ch4_file2	14-04-2014 13:40	Fast5 file	245 KB
TEST12345_ch4_file3	14-04-2014 13:40	Fast5 file	238 KB
TEST12345_ch4_file4	14-04-2014 13:41	Fast5 file	249 KB
TEST12345_ch5_file1	14-04-2014 13:41	Fast5 file	248 KB
TEST12345_ch5_file2	14-04-2014 13:41	Fast5 file	275 KB
TEST12345_ch5_file3	14-04-2014 13:41	Fast5 file	276 KB
TEST12345_ch5_file4	14-04-2014 13:41	Fast5 file	181 KB

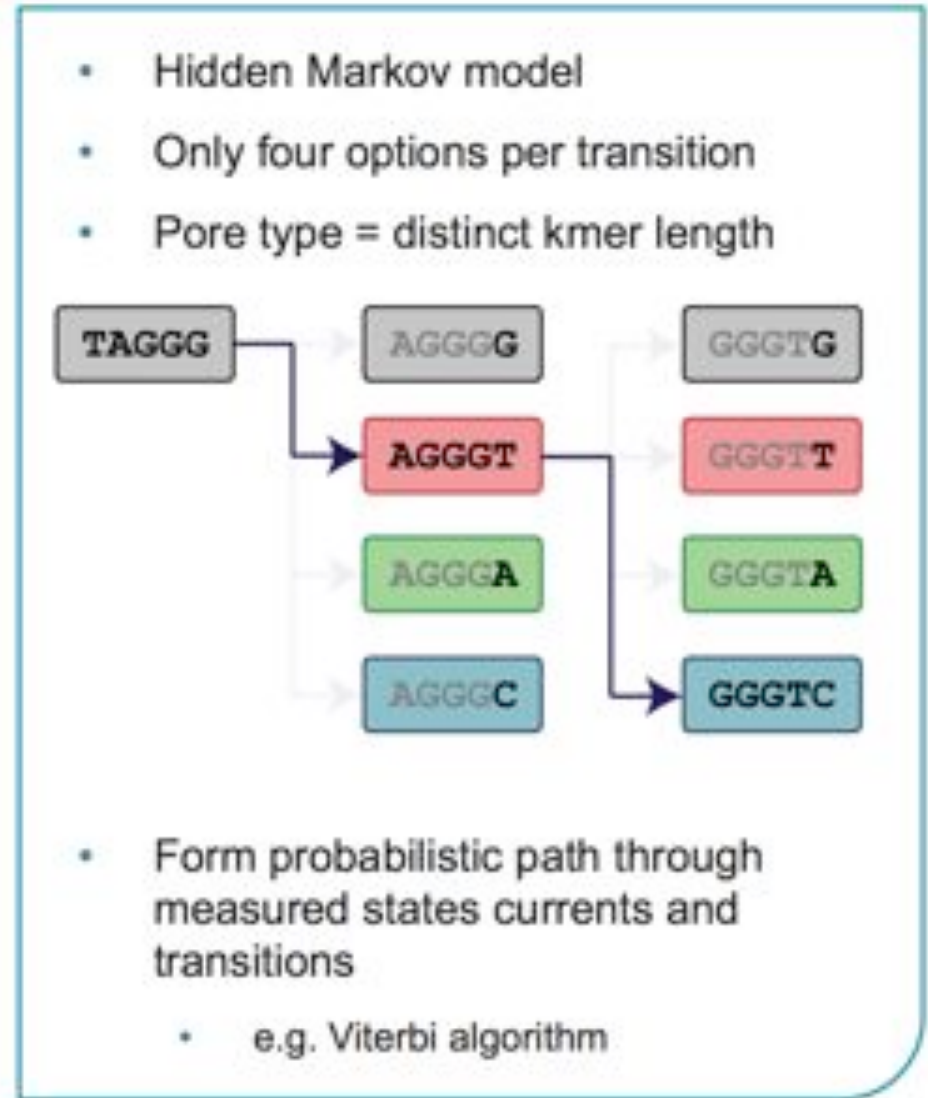
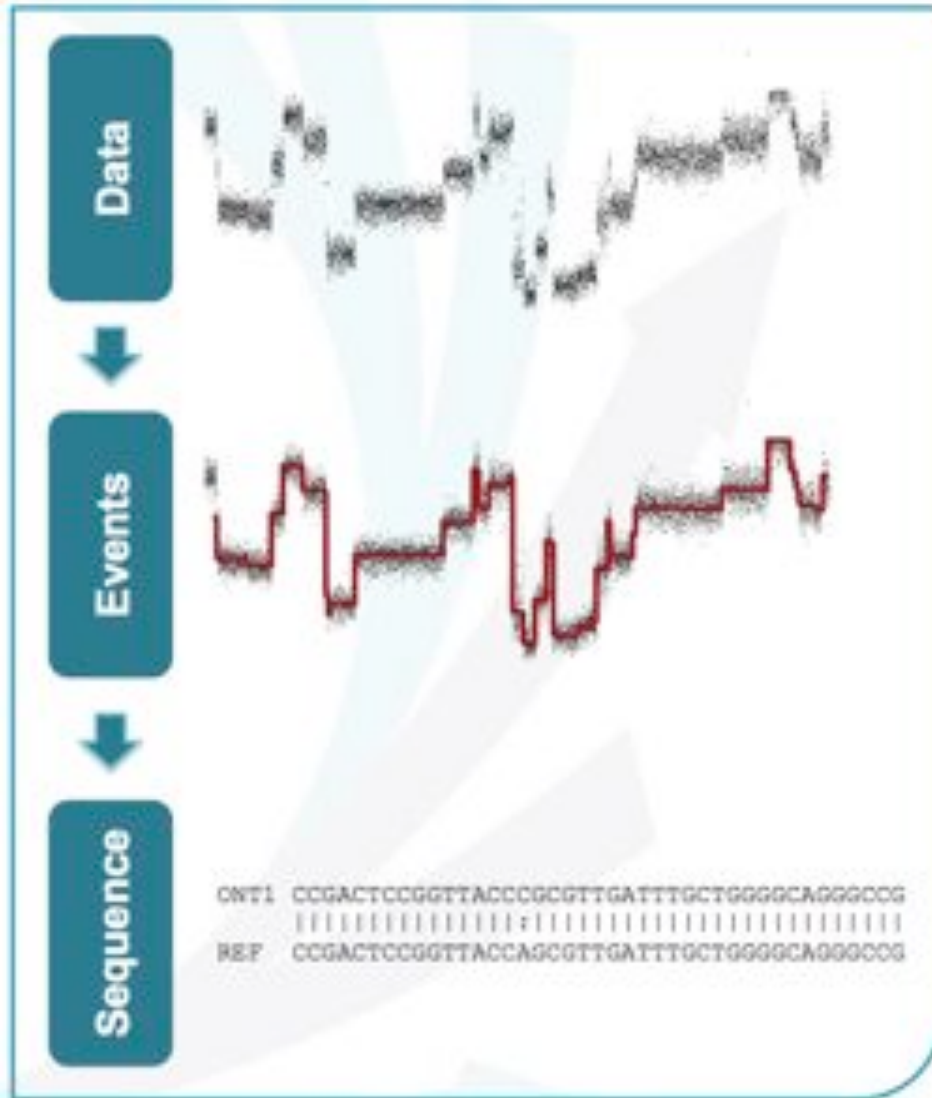


speu peoqdn



Base Calling in Cloud

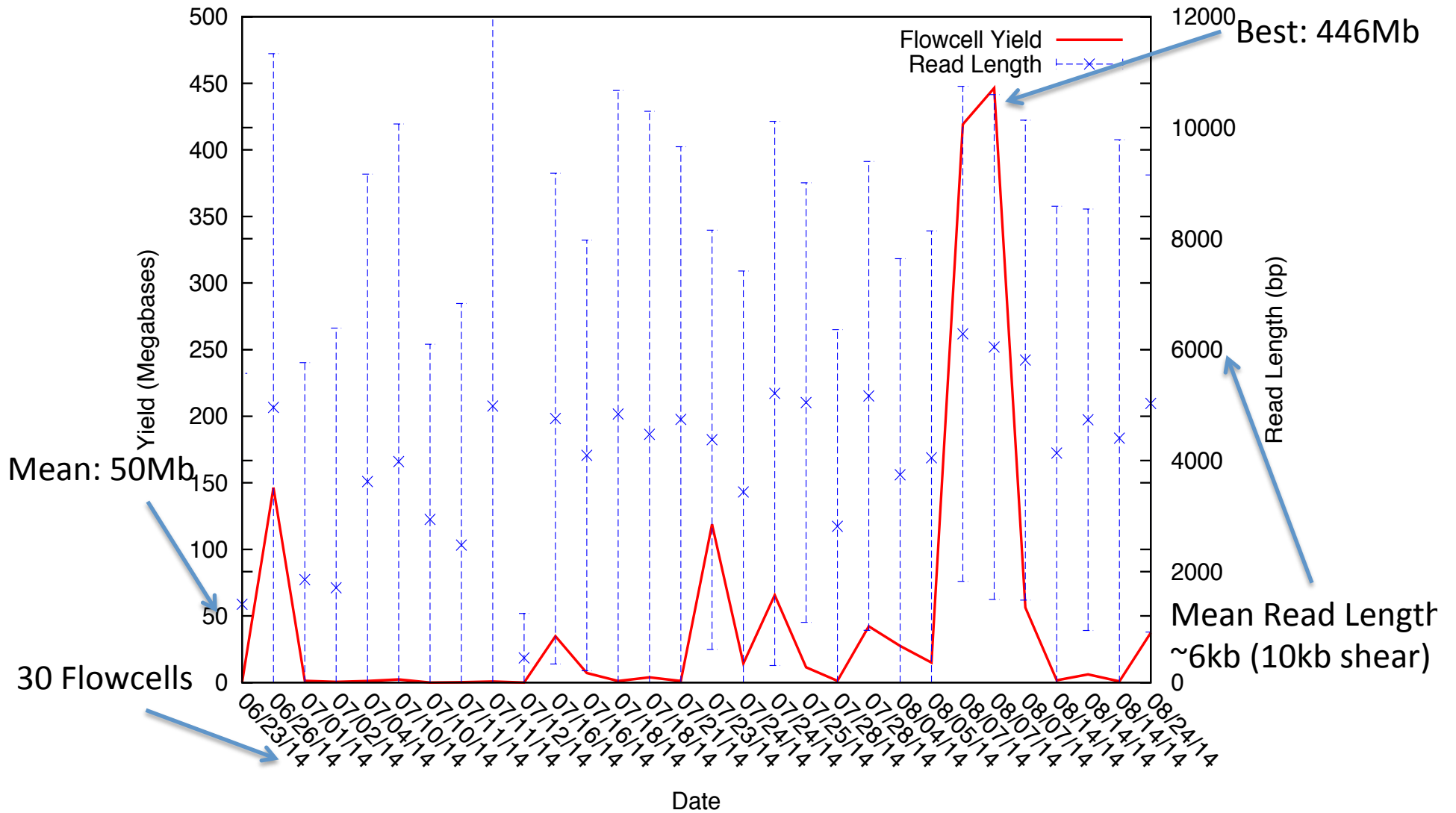
Nanopore Basecalling



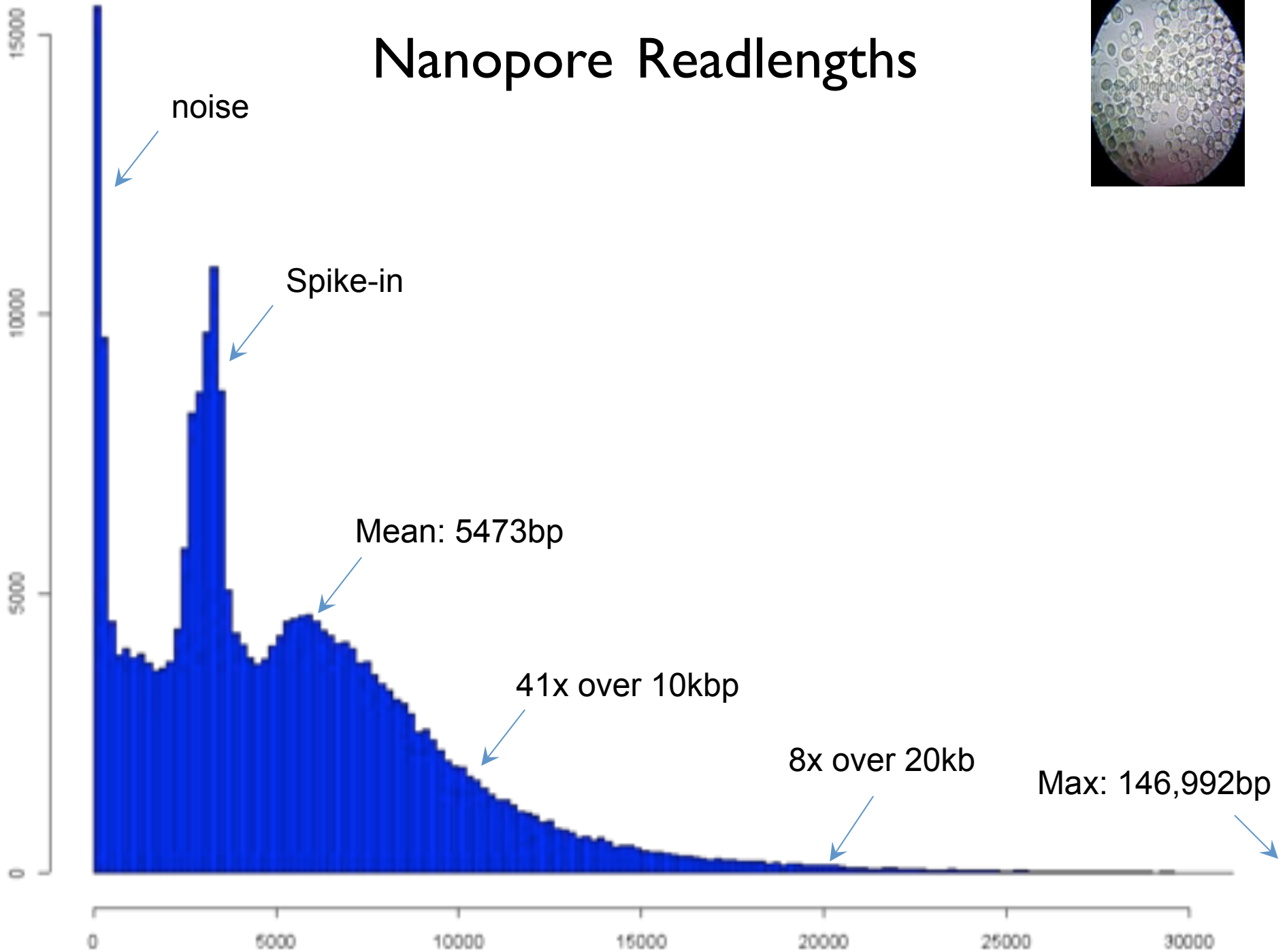
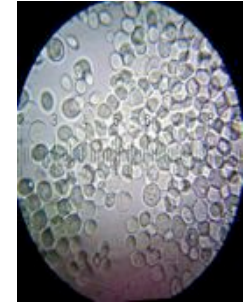
Basecalling currently performed at Amazon with frequent updates to algorithm

Our Data - Yeast W303

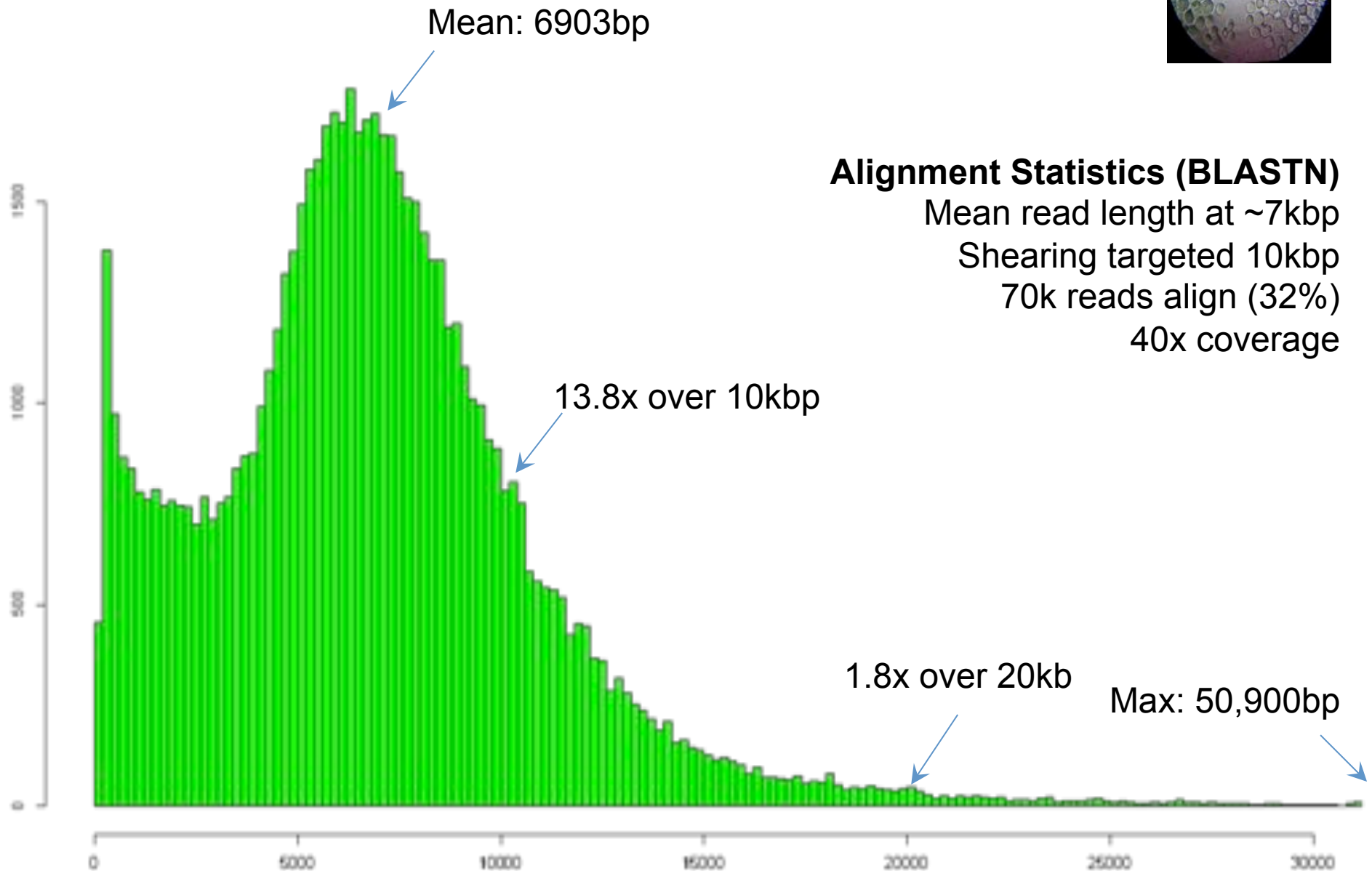
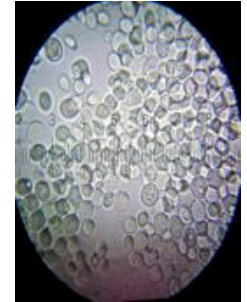
Oxford Flowcell Yields



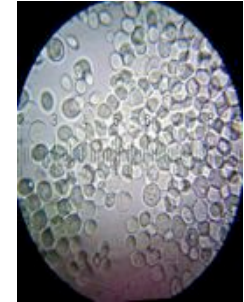
Nanopore Readlengths



Nanopore Alignments

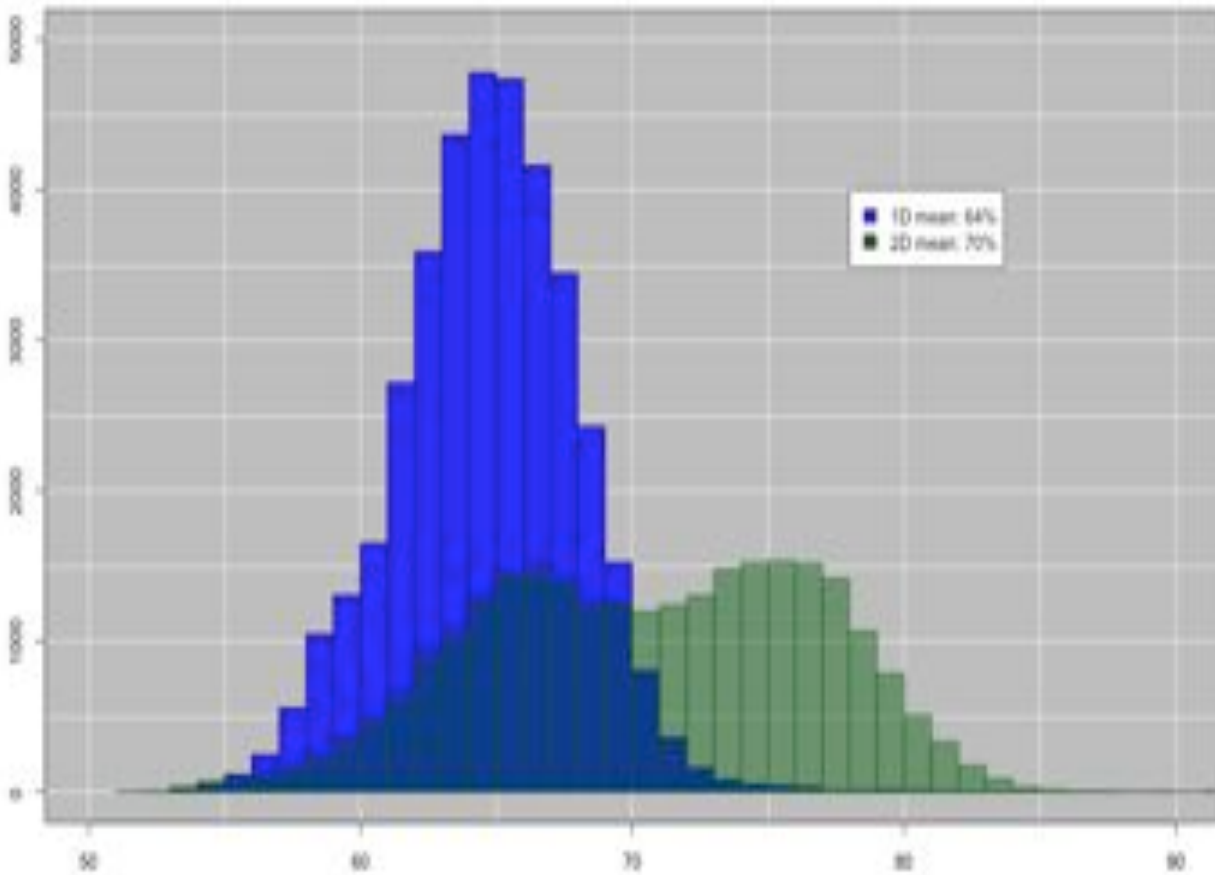


Nanopore Accuracy



Alignment Quality (BLASTN)

Of reads that align, average ~64% identity
“2D base-calling” improves to ~70% identity

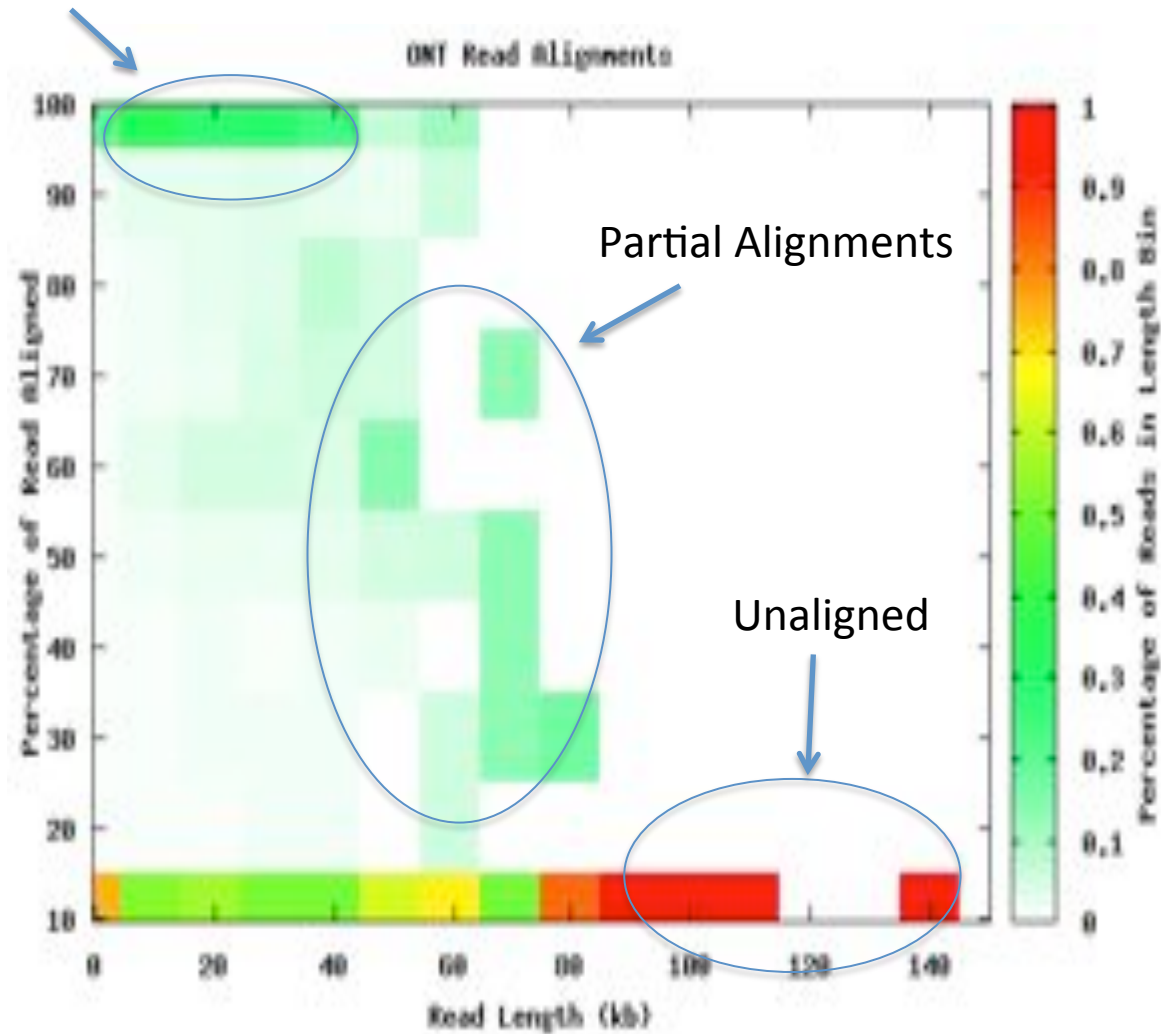


57% Mismatches
32% Deletions
11% Insertions

Nanopore Alignment Summary

32% of the data map using BLASTN

Full Length Alignments

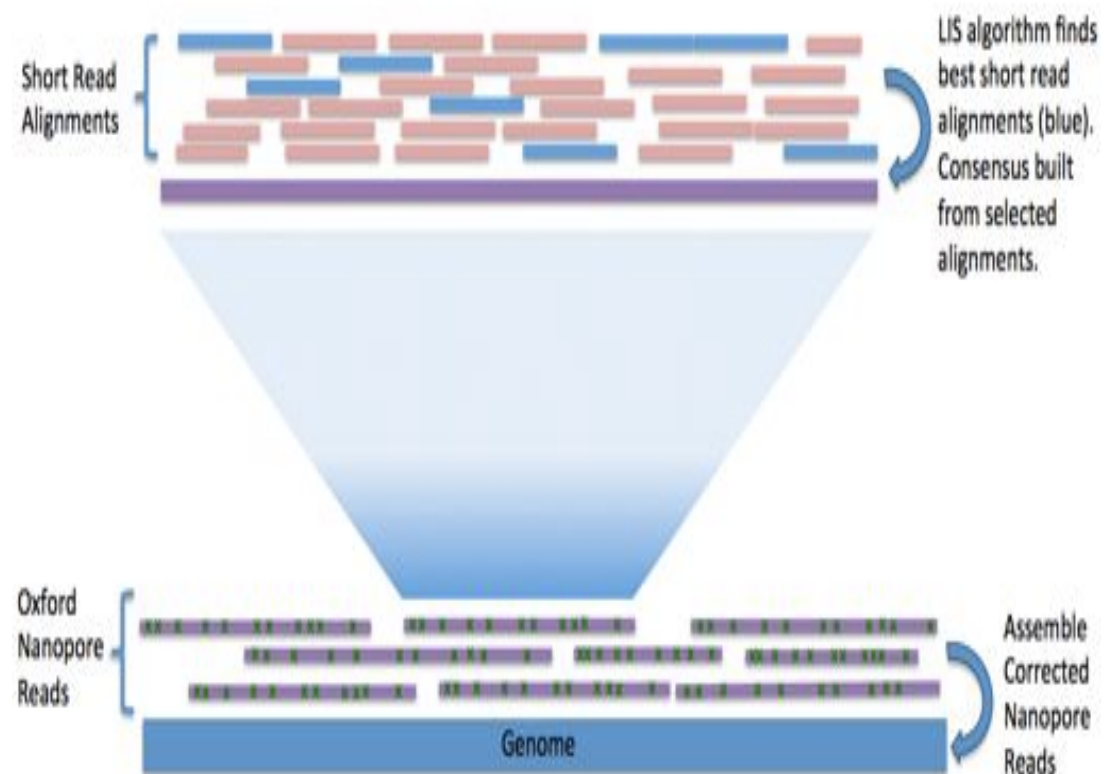


NanoCorr: Nanopore-Illumina Hybrid Error Correction

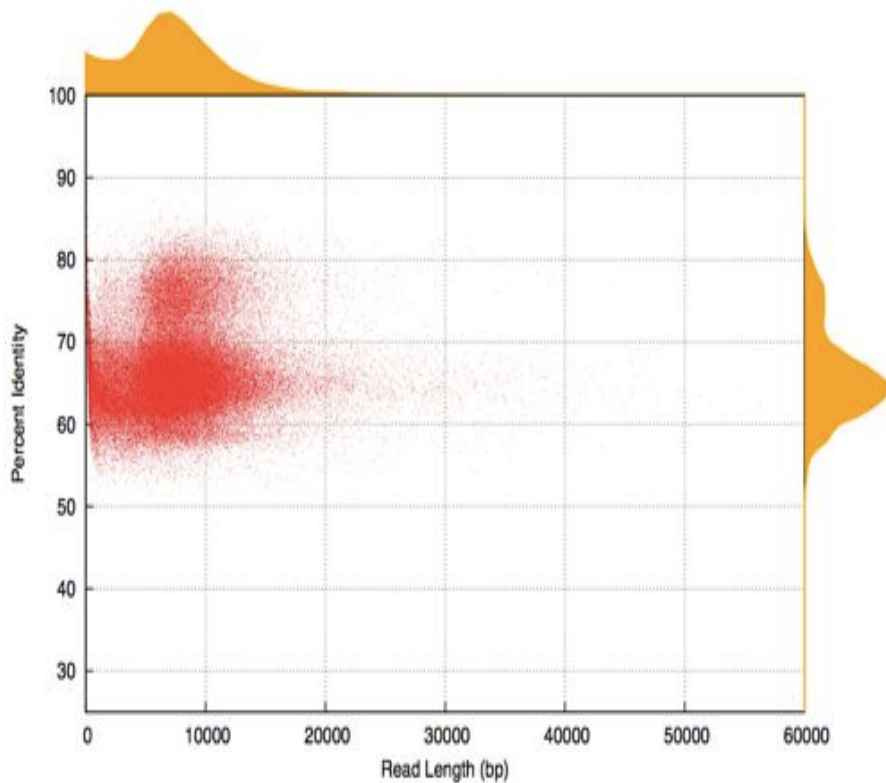


<https://github.com/jgurtowski/nanocorr>

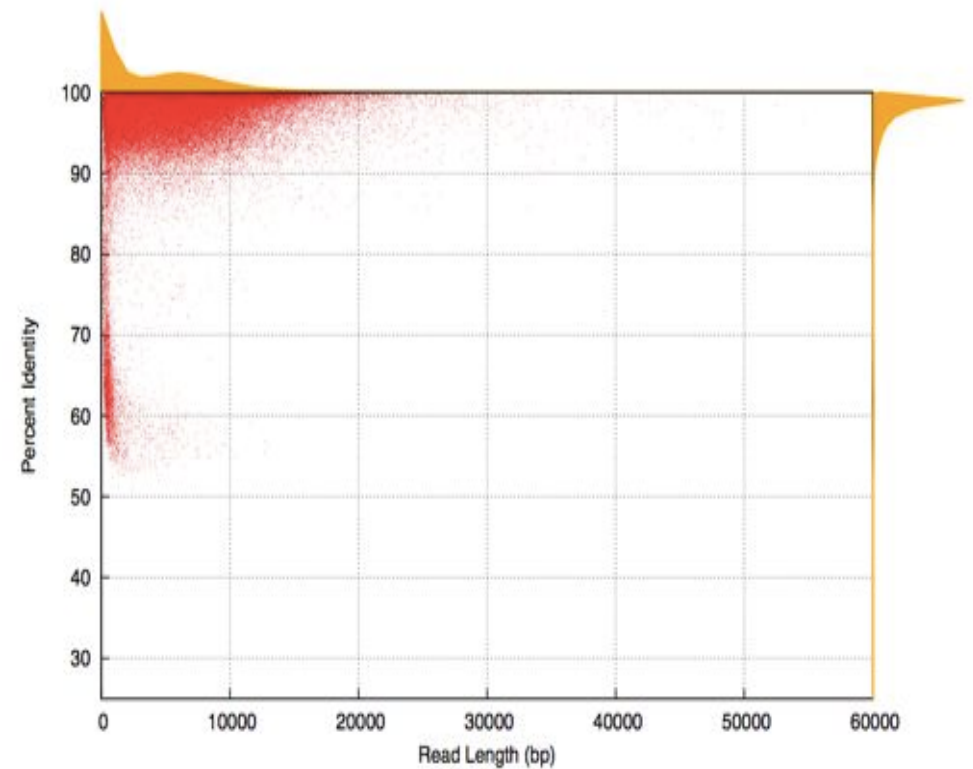
1. BLAST Miseq reads to all raw Oxford Nanopore reads
 1. First pass scans to remove “contained” alignments
 2. Second pass uses Dynamic Programming (LIS) to select set of high-identity alignments with minimal overlaps
3. Compute consensus of each Oxford Nanopore read
 1. Currently using Pacbio’s pbdagcon



Nanocorr correction pipeline significantly improves read identity



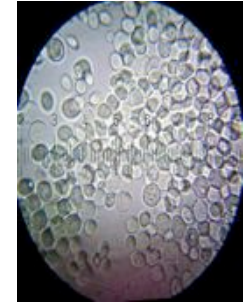
Before



After

Percent identity versus read length before and after nanocorr correction

Long Read Assembly



S288C Reference sequence

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

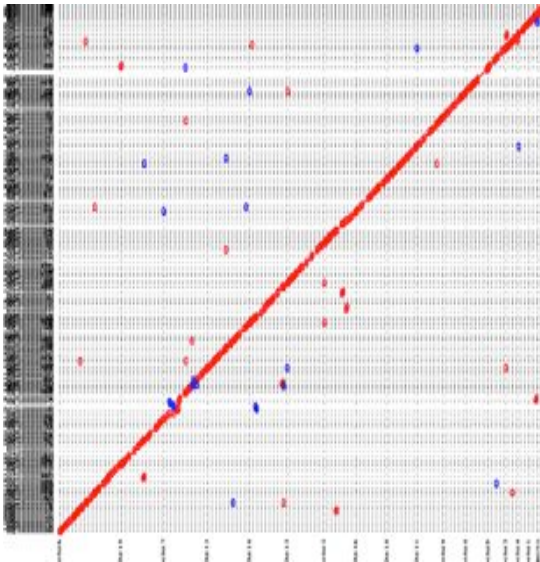
illumina MiSeq

30x, 300bp PE (Flashed)



Celera Assembler

- 6953 non-redundant contigs
- N50: 59kb >99.9% id



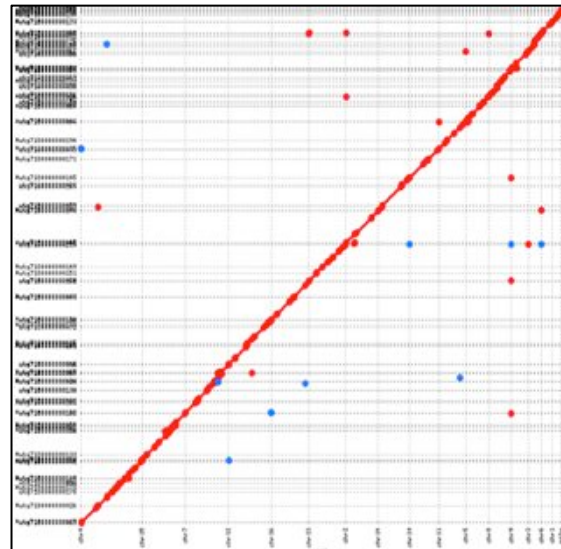
Oxford Nanopore

30x corrected reads > 6kb

NanoCorr + Celera Assembler



- 214 non-redundant contigs
- N50: 472kbp >99.78% id



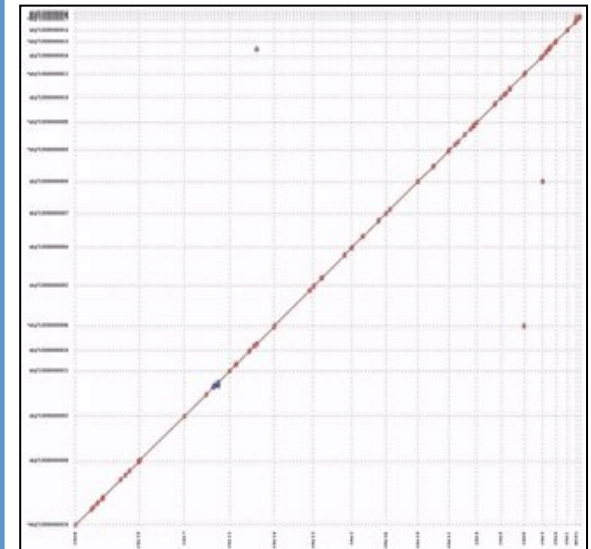
Pacific Biosciences

25x corrected reads > 10kb

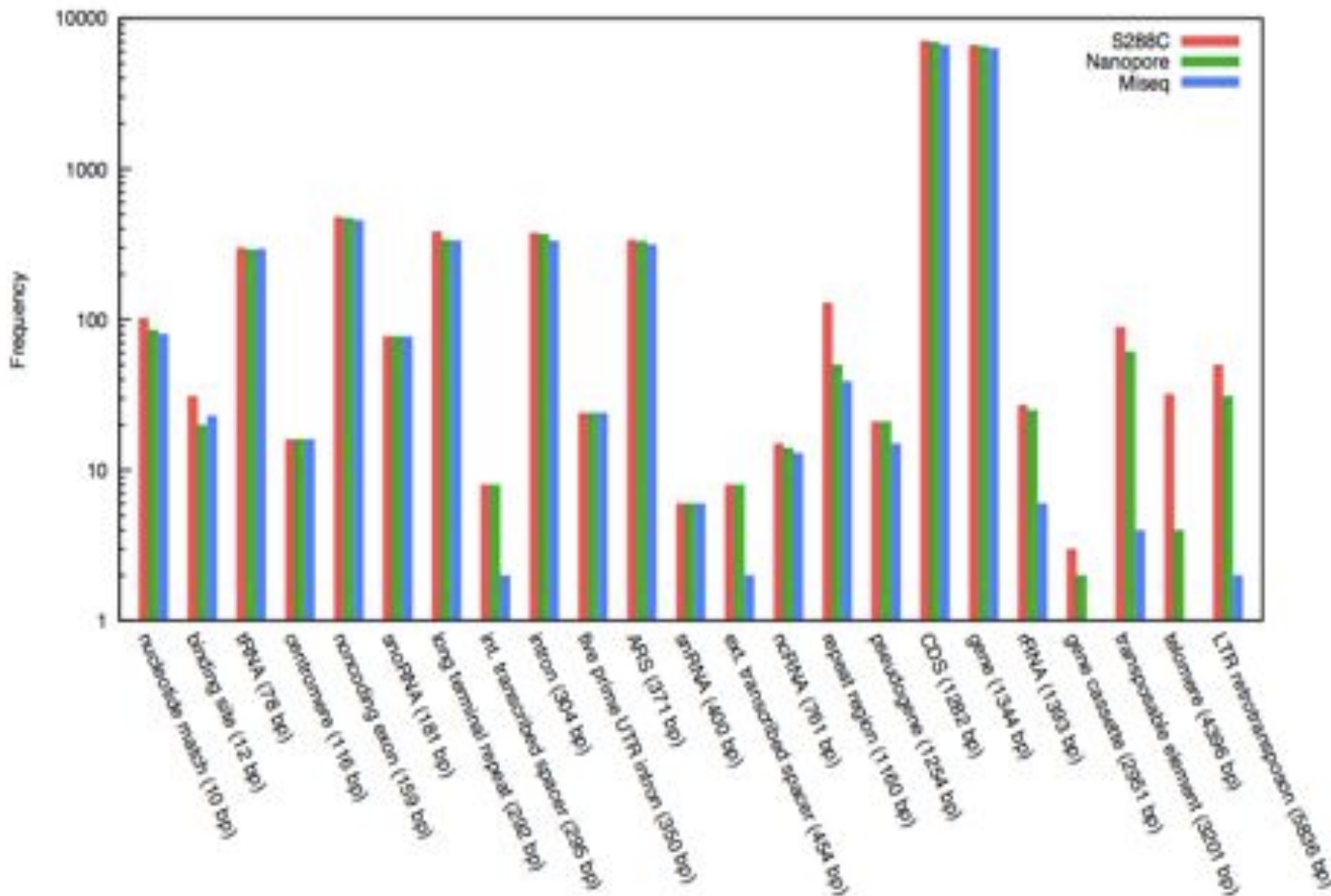
HGAP + Celera Assembler



- 21 non-redundant contigs
- N50: 811kb >99.8% id



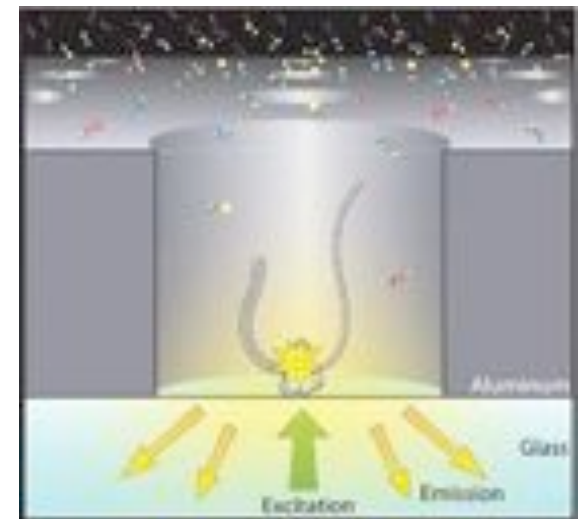
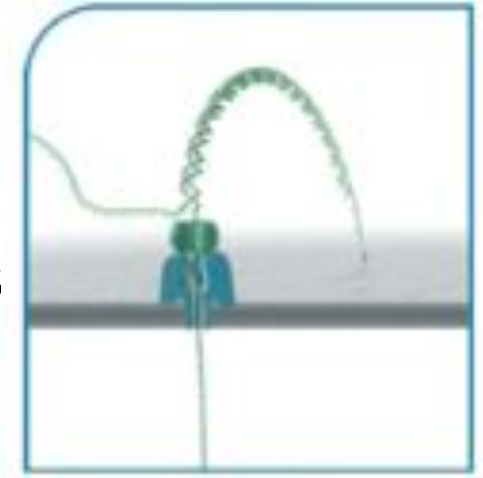
An assembly generated from Oxford Nanopore long reads is better able to identify genomic features



- S288C is an extremely high quality reference
- In virtually all cases the Oxford estimate of the frequency of a genomic feature is closer to S288C than data generated by miSeq
- In some cases (gene cassette, telomere) the miSeq is completely unable to detect features

Summary

1. New Single Molecule Sequencing Technologies
2. Produce very long reads
3. Have High Error rate -> Error Correction
4. Long reads Produce great assemblies, far better than short read technologies -> repeat resolution



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