# **SMRT-assembly**

Error correction and de novo assembly of complex genomes using single molecule, real-time sequencing

#### Michael Schatz

Jan 17, 2012 PAG-XX: PacBio Workshop



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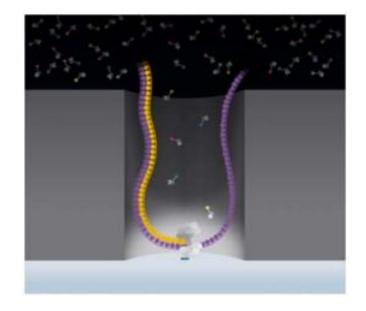


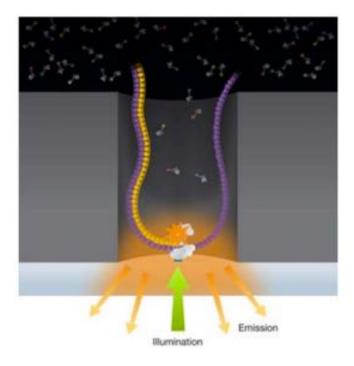
### Outline

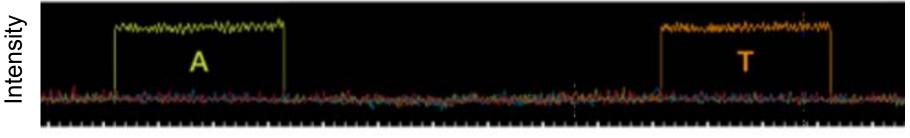
- I. SMRT-sequencing
  - I. Coverage, read length, and accuracy
- 2. SMRT-assembly approaches
  - I. SMRT-de novo: SMRT-only assembly
  - 2. SMRT-scaffolding: Long reads as links
  - 3. SMRT-hybrid: Short and long together
- 3. Review and best practices

### **SMRT** Sequencing

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

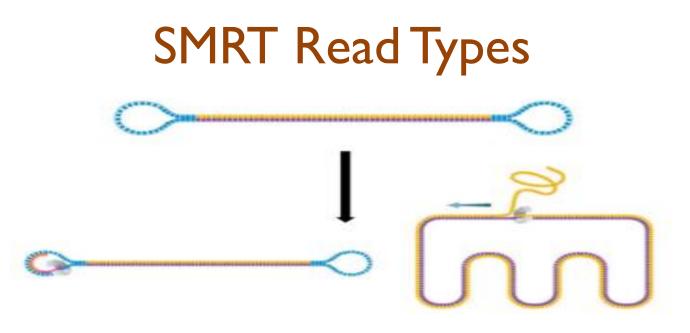






Time

http://www.pacificbiosciences.com/assets/files/pacbio\_technology\_backgrounder.pdf



#### 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.

#### • Strobe sequencing

 Very long inserts, alternate the lasers in the instrument between on and off. On periods generate strobe sub-reads and the off periods determine the length of the spacing between, known as strobe advance

http://www.pacificbiosciences.com/assets/files/pacbio\_technology\_backgrounder.pdf

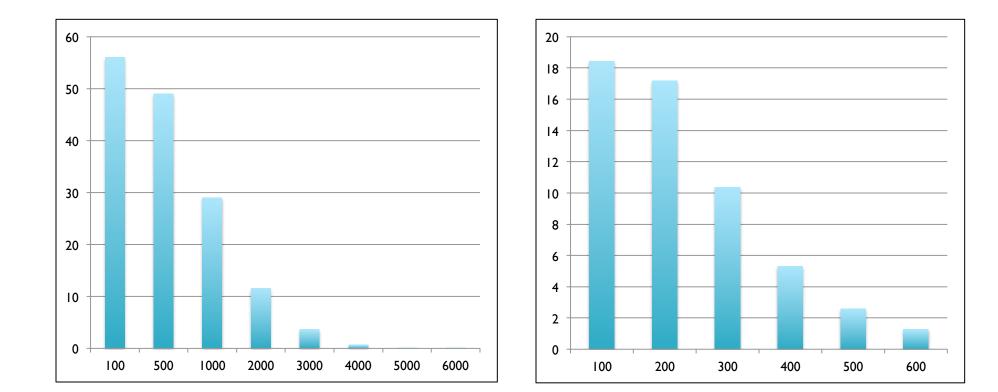
#### **SMRT Sequencing Runs**

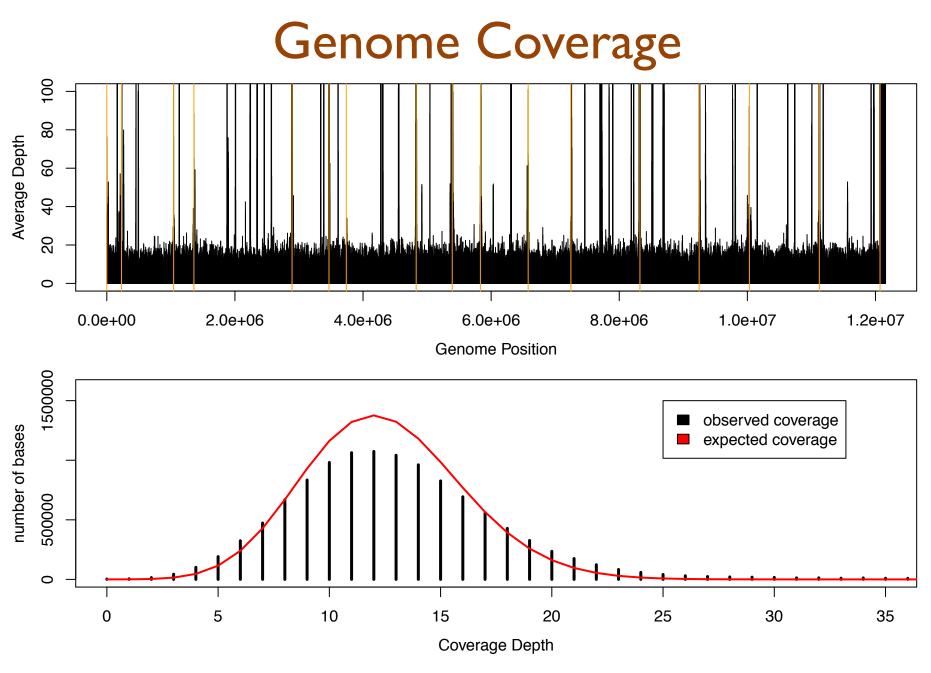
#### Yeast – Long reads

969,445 reads after filtering Mean: 710 +/- 663 Median: 558 Max: 8,495

#### Yeast – CCS reads

731,638 reads after filtering Mean: 306 +/- 115 Median: 279 Max: 1,425

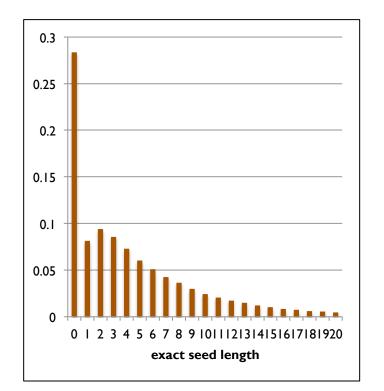




Coverage plots of long reads along yeast genome computed by BLASR

### Alignment Quality

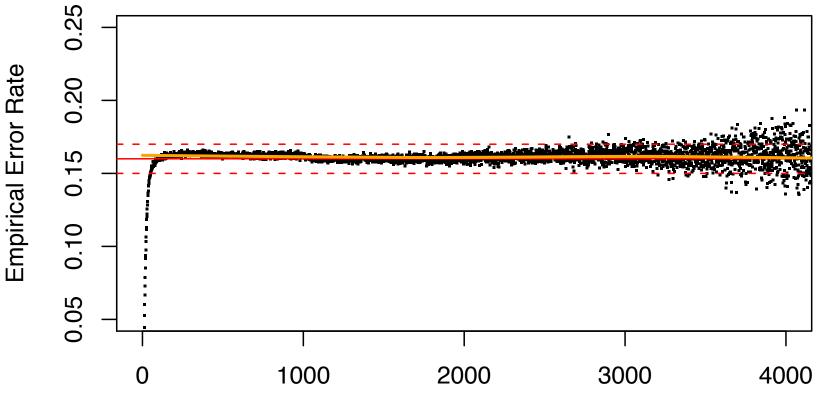
Match	83.7%
Mismatch	1.4%
Insertions	11.5%
Deletions	3.4%



4	TTGTAAGCAGTTGAAAACTATGTGT <mark>G</mark> GATTTAG <mark>A</mark> ATAAAGAACATG <mark>A</mark> AAG
539752	TTGTAAGCAGTTGAAAACTATGTGT-GATTTAG-ATAAAGAACATG <mark>G</mark> AAG
54	ATTATAAA-CAGTTGATCCATT-AGAAGA-AAACGCAAAAGGCGGCTAGG
539800	A-TATAAATCAGTTGATCCATTAAGAA-AGAAACGC-AAAGGC-GCTAGG
101	CAACCTTGAATGTAATCGCACTTGAAGAACAAGATTTTATTCCGCGCCCG
539846	C-ACCTTG-ATGT-ATCACTTGAAGAACAAGATTTTATTCCGCGCCCG
151	TAACGAATCAAGATTCTGAAAACACAT-ATAACAACCTCCAAAA-CACAA
539891	T-ACGAATC-AGATTCTGAAAACA-ATGATACCTCCAAAAGCACAA
199	-AGGAGG <mark>GGAAAGGGGGG</mark> GAATATCT-AT <mark>A</mark> AAAGATTACAAATT <mark>A</mark> GA-TGA
539934	GAGGAGGAAGAATATCTGAT-AAAGATTACAAATT-GAGTGA
246	ACT-AATTCACAATA-AATAACACTTTTA-ACAGAATTGAT-GGAA-GTT
539974	ACTAAATTCACAA–ATAATAACACTTTTAGACAAAATTGATGGGAAGGTT
291	TCGGAGAGATCCAAAACAATGGGC-ATCGCCTTTGA-GTTAC-AATCAAA
540023	TC-GAGAGATCC-AAACAAT-GGC <mark>G</mark> ATCG-CTTTGA <mark>C</mark> GTTAC <mark>A</mark> AATCAAA
338	ATCCAGT <mark>G</mark> GAAAATATAATTTATGCAATCCA <mark>G</mark> GAACTTATTCACAATTAG
540069	ATCCAGT-GAAAATATATTATGC-ATCCA-GAACTTATTCACAATTAG

Sample of 100k reads aligned with BLASR requiring >100bp alignment





**Read Position** 

#### Consistent quality across the entire read

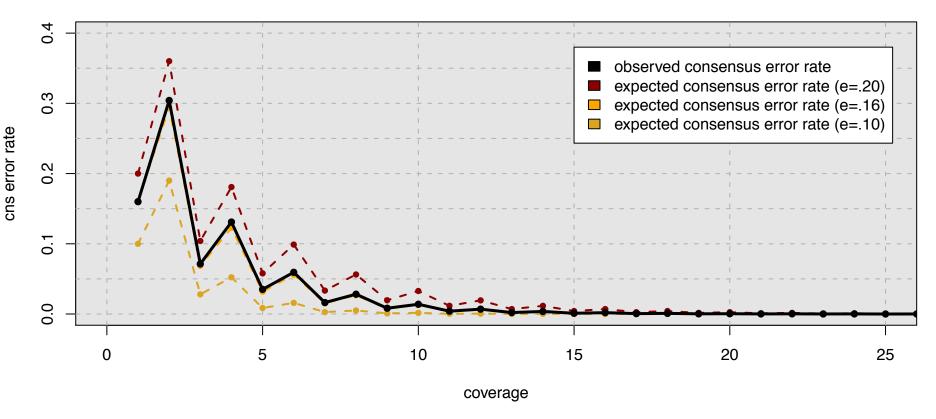
- Uniform error rate, no apparent biases for GC/motifs
- Sampling artifacts at beginning and ends of alignments

#### Consensus Quality: Probability Review

Roll *n* dice => What is the probability that at least half are 6's

n	Min to Win	Winning Events	P(Win)
I		1/6	16.7%
2		P(lof 2) + P(2 of 2)	30.5%
3		P(2 of 3) + P(3 of 3)	7.4%
4		P(2 of 4) + P(3 of 4) + P(4 of 4)	13.2%
5		P(3 of 5) + P(4 of 5) + P(5 of 5)	3.5%
n	ceil(n/2)	$\sum_{i=\lceil n/2\rceil}^{n} P(i \text{ of } n) = \sum_{i=\lceil n/2\rceil}^{n} \binom{n}{i} (p)^{i} (1-p)^{n-i}$	

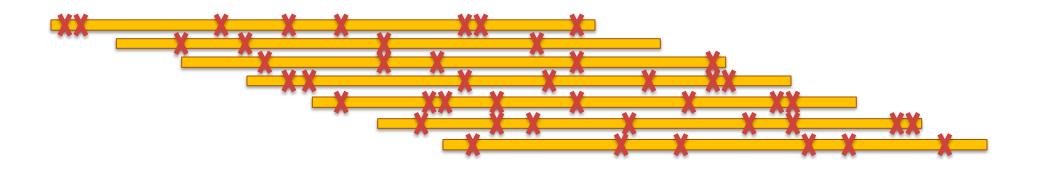
## **Consensus Accuracy and Coverage**



#### Coverage can overcome random errors

- Dashed lines accuracy model of binomial sampling
- Solid line observed consensus error rate
- For same reason, CCS is extremely accurate when using 5+ subreads

## Approach I: SMRT-de novo

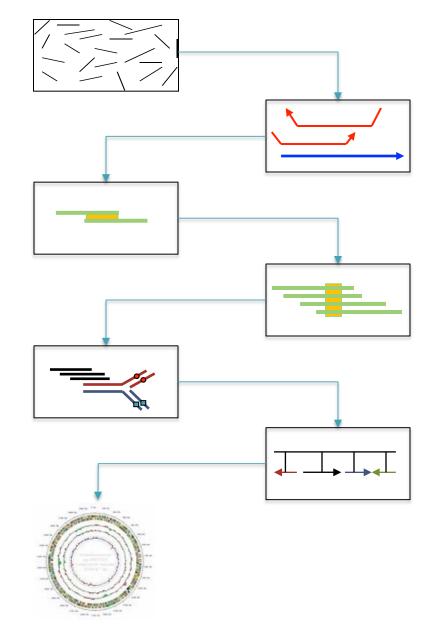


- De novo assembly of SMRT-reads
  - Rapid sequencing and assembly
  - Long reads to span repeats
- Challenges
  - 15% error rate per read equates to ~30% error rate per overlap
  - CCS reads as shorter, but higher quality reads

#### Celera Assembler

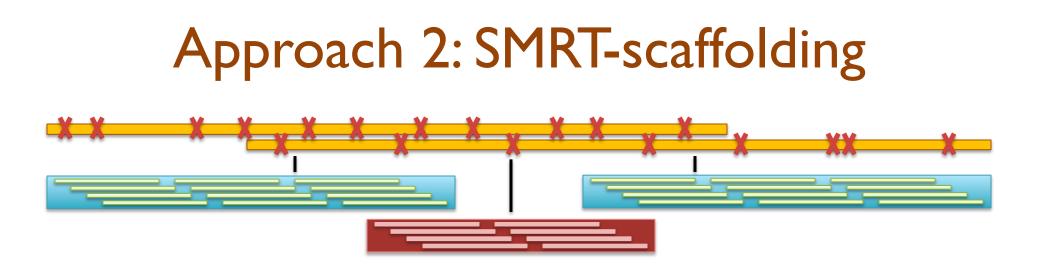
http://wgs-assembler.sf.net

- I. Pre-overlap
  - Consistency checks
- 2. Trimming
  - Quality trimming & partial overlaps
- 3. Compute Overlaps
  - Find high quality overlaps
- 4. Error Correction
  - Evaluate difference in context of overlapping reads
- 5. Unitigging
  - Merge consistent reads
- 6. Scaffolding
  - Bundle mates, Order & Orient
- 7. Finalize Data
  - Build final consensus sequences

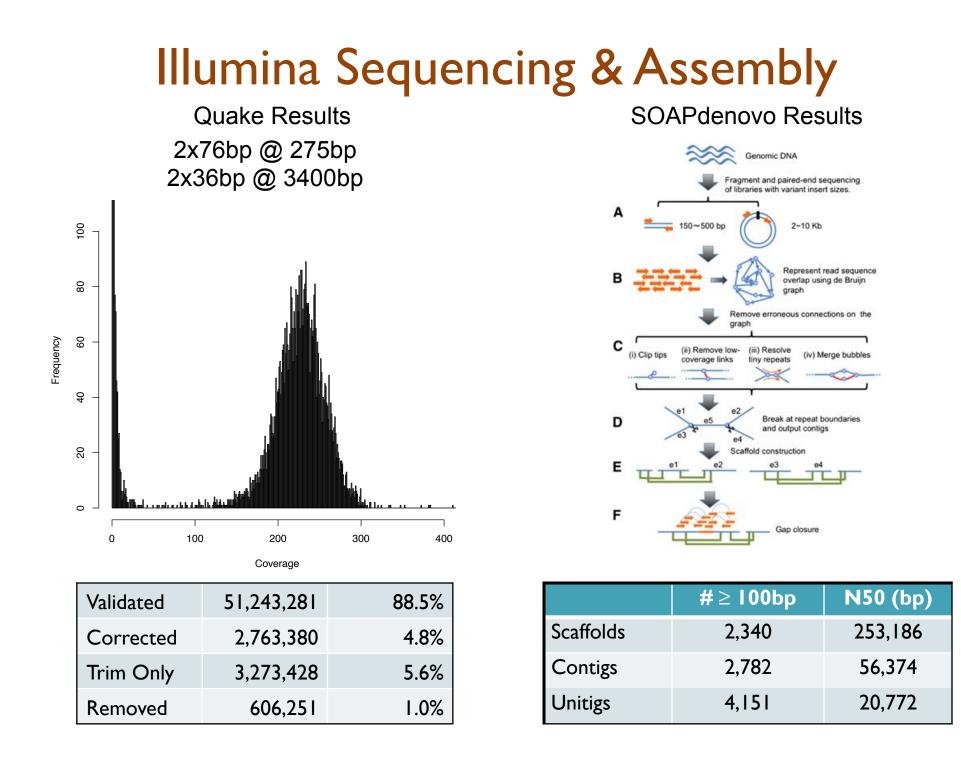


### SMRT-de novo Results

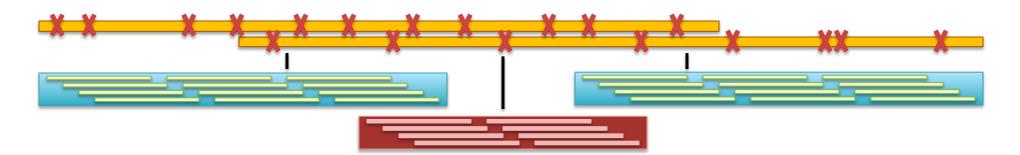
- De novo assembly of long reads
  - Experiments in progress
  - Very challenging to find good overlaps with very high error rate
- De novo assembly of CCS reads
  Contig N50: 24,582bp
- De novo assembly of ref-corrected CCS – Contig N50: 65,119bp



- Use long reads (or strobe reads) to link high quality contigs from short reads
  - Long reads (orange) span repetitive short-read contig (red)
  - Doesn't need very high coverage nor accuracy of long reads
- Challenges
  - Creating good short read assembly
  - Properly aligning reads to contigs
  - Untangling complex repeats



#### SMRT-scaffolding results



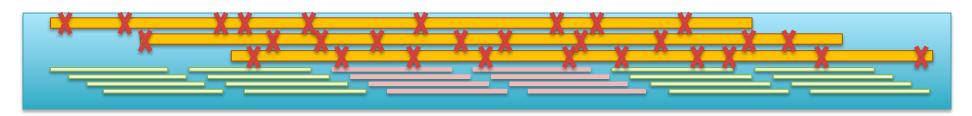
SMRTpipe hybrid scaffold of SOAPdenovo assembly + >2kbp long reads Scaffold N50: 310,246bp (+22% improvement) Scaffold cnt: 2246 (4% reduction)

SMRTpipe hybrid scaffold of ref-CCS assembly + >2kbp long reads

Scaffold N50: 97,414bp (+50% improvement)

Scaffold cnt: 6,610 (3% reduction)

## Approach 3: SMRT-hybrid



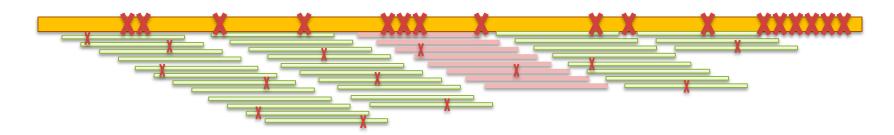
- Co-assemble long reads and short reads
  - Long reads (orange) natively span repeats (red)
  - Guards against mis-assemblies in draft assembly
  - Use all available data at once
- Challenges
  - Long reads have too high of an error rate to assemble directly
  - Assembler must supports a wide mix of read lengths

#### PacBio Error Correction http://wgs-assembler.sf.net

- I. Correction Pipeline
  - I. Map short reads (SR) to long reads (LR)
  - 2. Trim LRs at coverage gaps
  - 3. Compute consensus for each LR

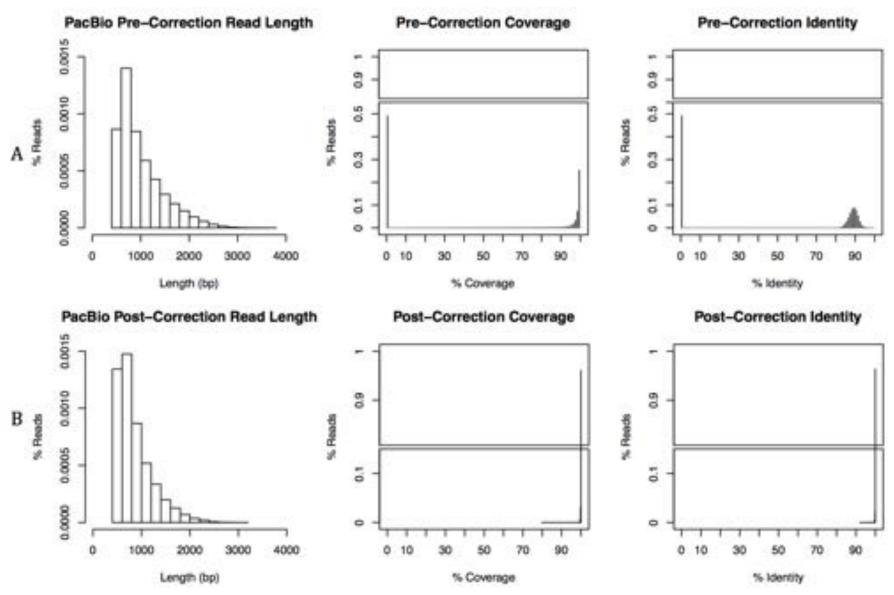


2. Error corrected reads can be easily assembled, aligned



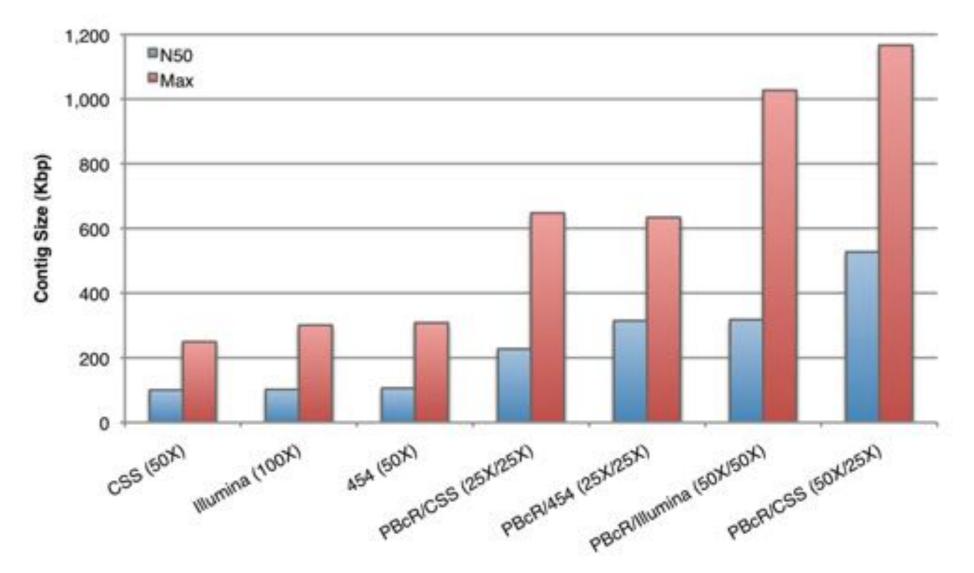
**Hybrid error correction and de novo assembly of single-molecule sequencing reads.** Koren, S, Schatz, MC, Walenz, BP, Martin, J, Howard, J, Ganapathy, G, Wang, Z, Rasko, DA, McCombie, WR, Jarvis, ED, Phillippy, AM. (2012) *Under Review* 

#### **Error Correction Results**



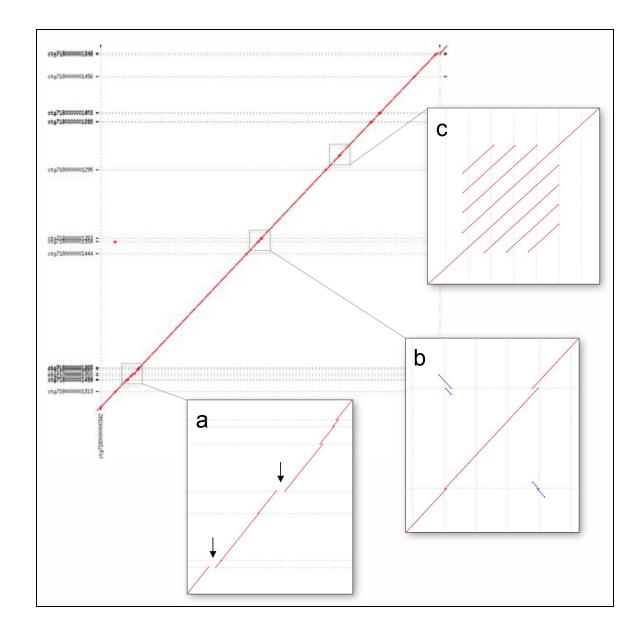
Correction results of 20x PacBio coverage of E. coli K12 corrected using 50x Illumina

#### **Assembly Results**



SMRT-hybrid assembly results of 50x PacBio corrected coverage of E. coli K12 Long reads lead to **contigs** over 1Mbp

#### PacBio Advantages

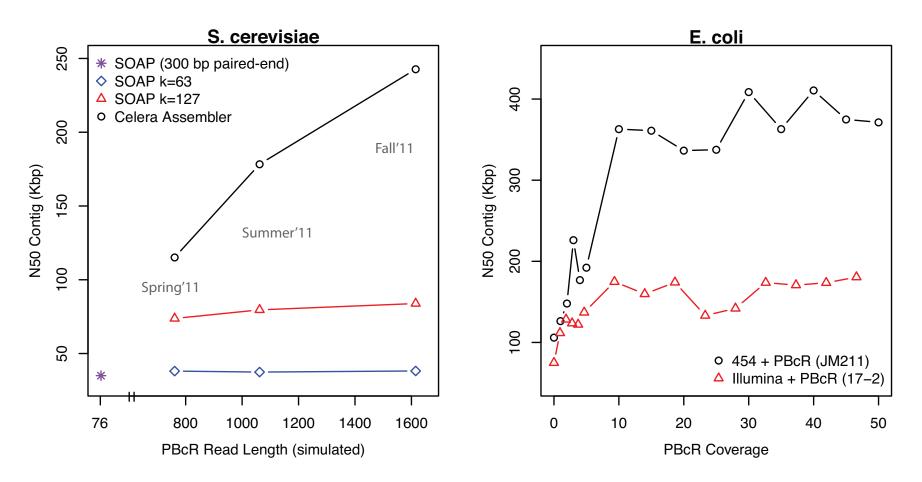


(a) Long reads close sequencing gaps

(b) Long readsassemble acrosslong repeats

(c) Long reads span complex microsatellites

#### **Assembly Results**



- The longer the read, the greater the improvement to assembly quality
- Best assemblies come from PacBio + 454 reads or PacBio + CCS
- Best assemblies need ~10x coverage PacBio long reads

## Hybrid Assembly Results

Reference bp

Assembly bp # Contigs

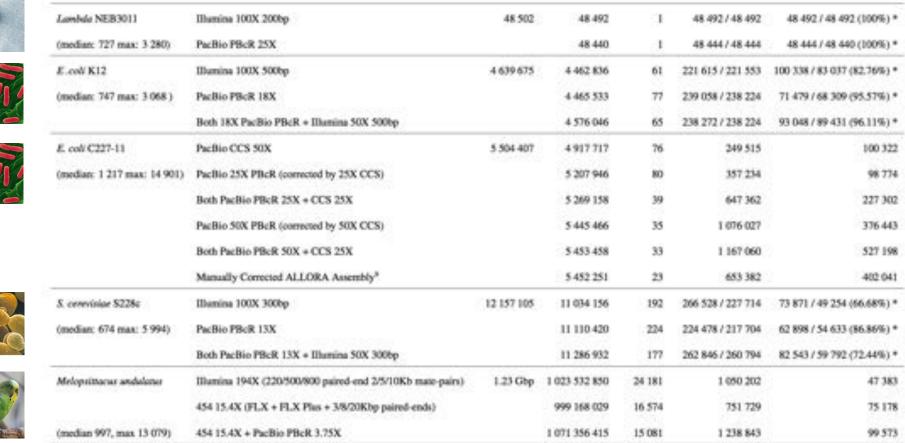
Max Contig Length

N50

Technology

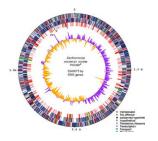


Organism



Hybrid assembly results using error corrected PacBio reads Meets or beats Illumina-only or 454-only assembly in every case

### Assembly Summary



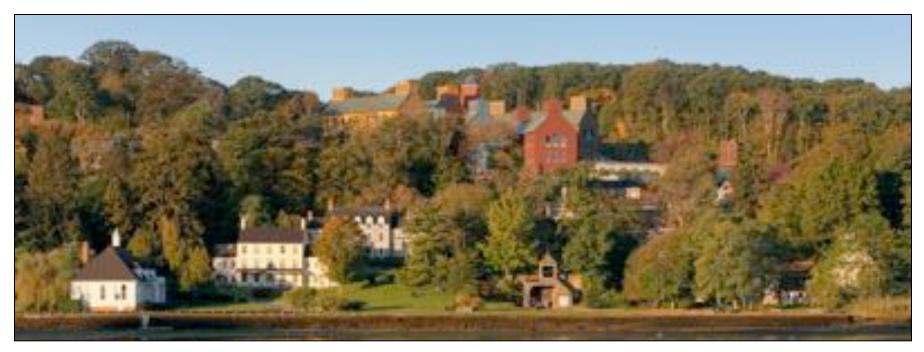
Assembly quality depends on

- I. Coverage: low coverage is mathematically hopeless
- 2. Read length: longer reads resolve repeats and complex regions
- 3. Read Quality: need clean libraries, clean reads
- PacBio RS has capabilities not found in any other technology
  - Substantially longer reads -> span repeats
  - Unbiased sequence coverage -> close sequencing gaps
  - Single molecule sequencing -> haplotype phasing, alternative splicing
- PacBio enables highest quality de novo assembly
  - Longer reads have fundamentally more information than shorter reads
  - Because the errors are random we can compensate for them informatically
  - Software available open source at http://wgs-assembler.sf.net

### Acknowledgements

Schatzlab Giuseppe Narzisi Mitch Bekritsky Matt Titmus Hayan Lee James Gurtowski Rohith Menon Goutham Bhat <u>CSHL</u> Dick McCombie Melissa Kramer Eric Antonio Mike Wigler Zach Lippman Doreen Ware Ivan Iossifov <u>NBACC</u> Adam Phillipy Sergey Koren

<u>JHU</u> Steven Salzberg Ben Langmead Jeff Leek Univ. of Maryland Mihai Pop Art Delcher Jimmy Lin David Kelley Dan Sommer Cole Trapnell



# Thank You

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