

De novo assembly of complex genomes

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UIUC



Outline

1. Genome assembly by analogy
2. Hybrid error correction and assembly
3. De novo mutations in autism



Outline

1. **Genome assembly by analogy**
2. Hybrid error correction and assembly
3. De novo mutations in autism



Shredded Book Reconstruction

- Dickens accidentally shreds the first printing of A Tale of Two Cities
 - Text printed on 5 long spools

It was	the	best	of	times,	it	was	the	worst	of	times,	it	was	the	age	of	wisdom,	it	was	the	age	of	foolishness,	...	
It was	the	best	of	times,	it	was	the	worst	of	times,	it	was	the	age	of	wisdom,	it	was	the	age	of	foolishness,	...	
It was	the	best	of	times,	it	was	the	worst	of	times,	it	was	the	age	of	wisdom,	it	was	the	age	of	foolishness,	...	
It was	the	best	of	times,	it	was	the	worst	of	times,	it	was	the	age	of	wisdom,	it	was	the	age	of	foolishness,	...	
It	was	the	best	of	times,	it	was	the	worst	of	times,	it	was	the	age	of	wisdom,	it	was	the	age	of	foolishness,	...

- How can he reconstruct the text?
 - 5 copies x 138,656 words / 5 words per fragment = 138k fragments
 - The short fragments from every copy are mixed together
 - Some fragments are identical

Greedy Reconstruction

It was the best of
age of wisdom, it was
best of times, it was
it was the age of
it was the age of
it was the worst of
of times, it was the
of times, it was the
of wisdom, it was the
the age of wisdom, it
the best of times, it
the worst of times, it
times, it was the age
times, it was the worst
was the age of wisdom,
was the age of foolishness,
was the best of times,
was the worst of times,
wisdom, it was the age
worst of times, it was

It was the best of
was the best of times,
the best of times, it
best of times, it was
of times, it was the
of times, it was the
times, it was the worst
times, it was the age

The repeated sequence make the correct reconstruction ambiguous

- It was the best of times, it was the [worst/age]

Model sequence reconstruction as a graph problem.

de Bruijn Graph Construction

- $G_k = (V, E)$
 - $V =$ All length- k subfragments ($k < l$)
 - $E =$ Directed edges between consecutive subfragments
 - Nodes overlap by $k-1$ words

Original Fragment

It was the best of

Directed Edge

It was the best → was the best of

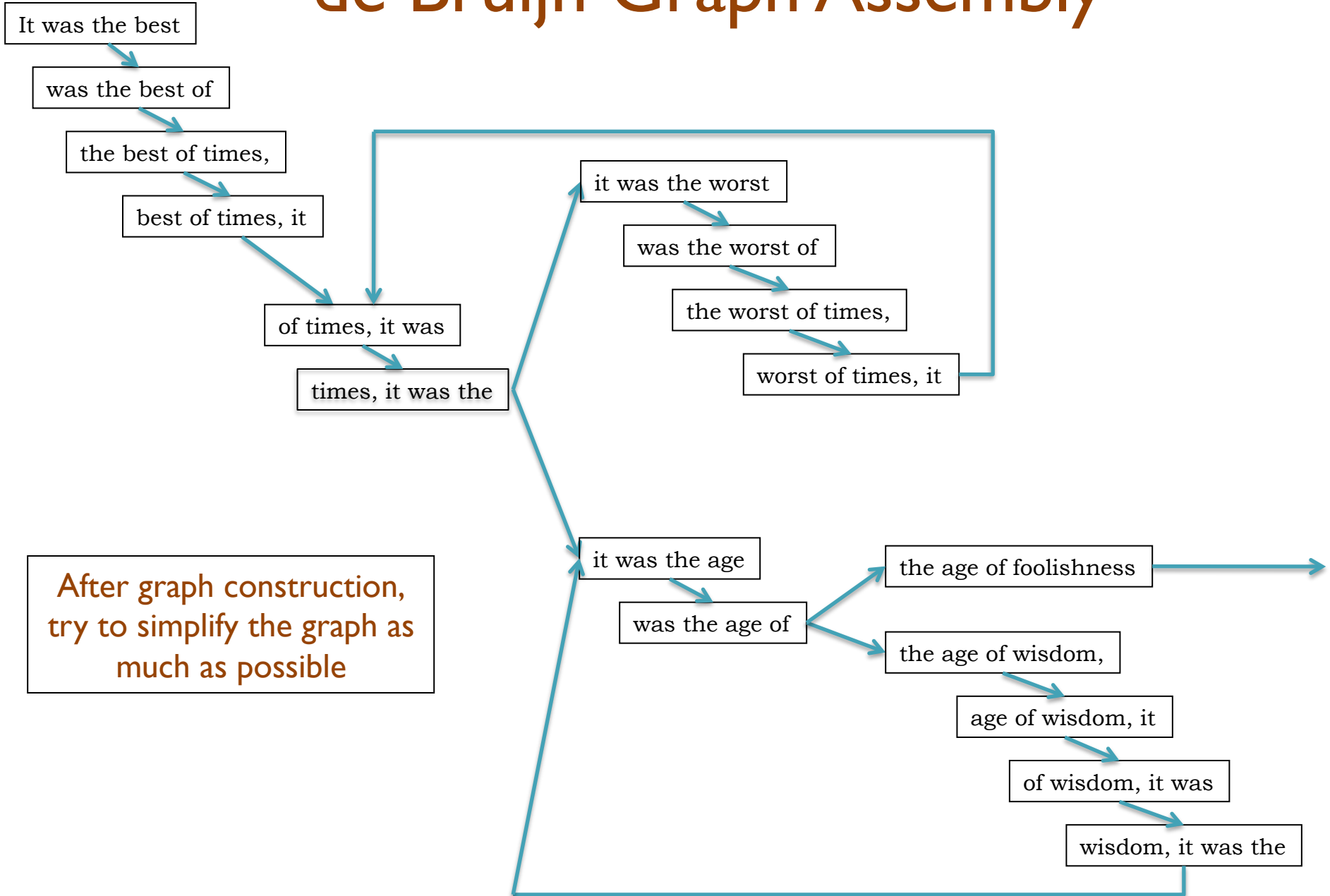
- Locally constructed graph reveals the global sequence structure
 - Overlaps between sequences implicitly computed

de Bruijn, 1946

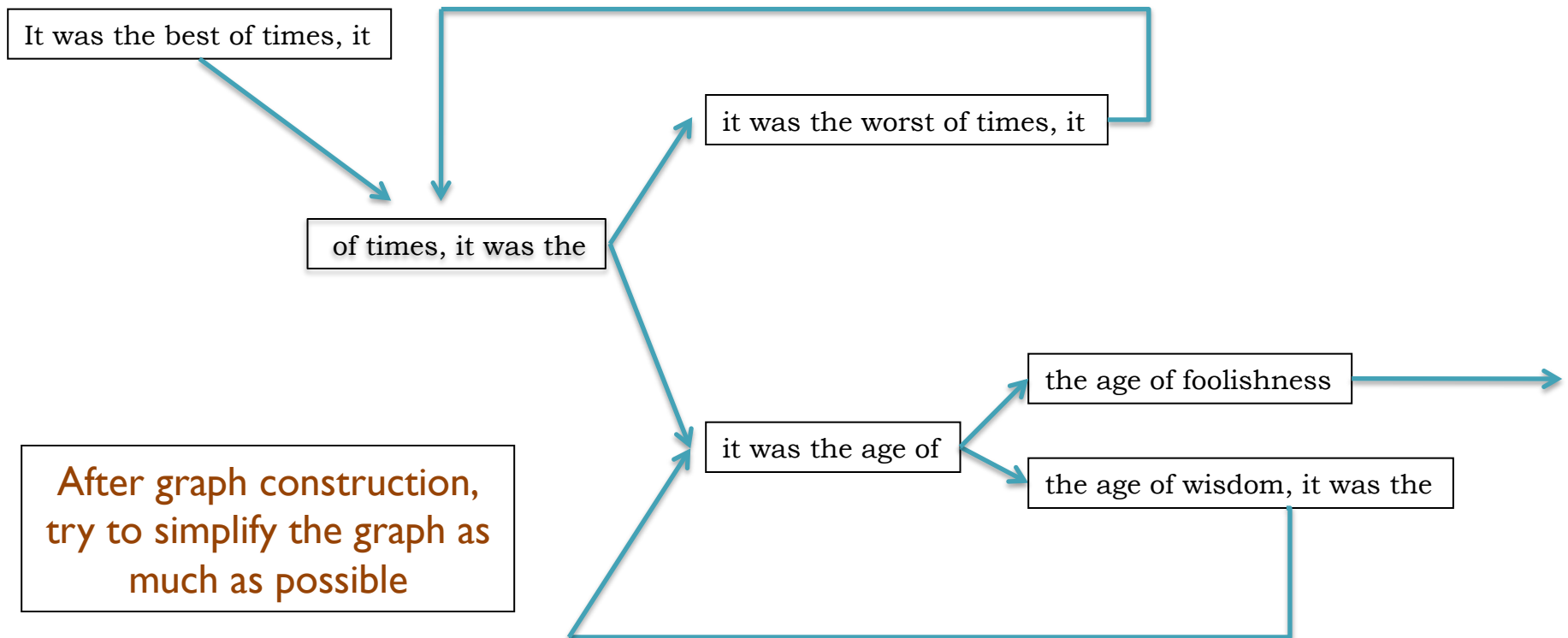
Idury and Waterman, 1995

Pevzner, Tang, Waterman, 2001

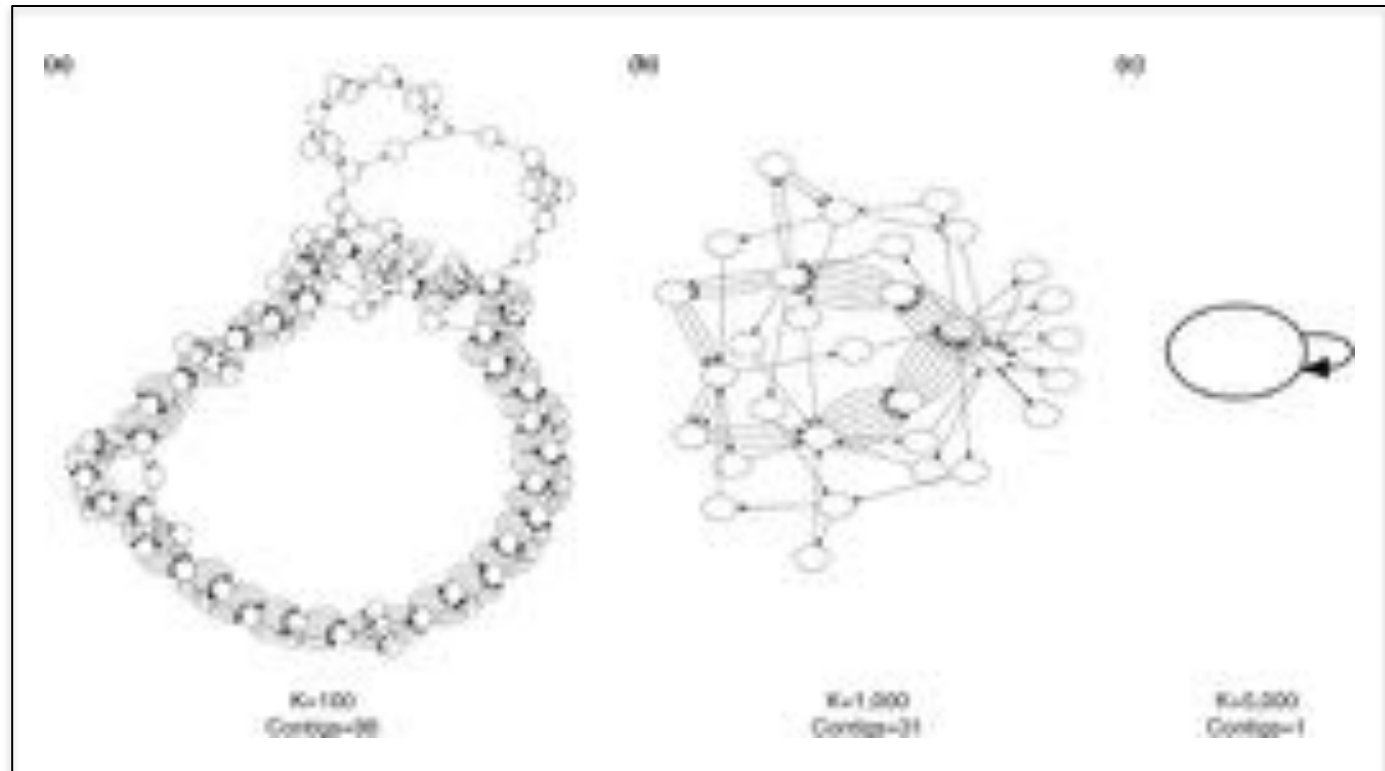
de Bruijn Graph Assembly



de Bruijn Graph Assembly



Reducing Complexity



Longer reads span more repeats, simplifying the assembly problem

- Idealized assembly of *B. anthracis* reduces to a single contig with 5kb reads
- Exact improvement depends on the specific genome

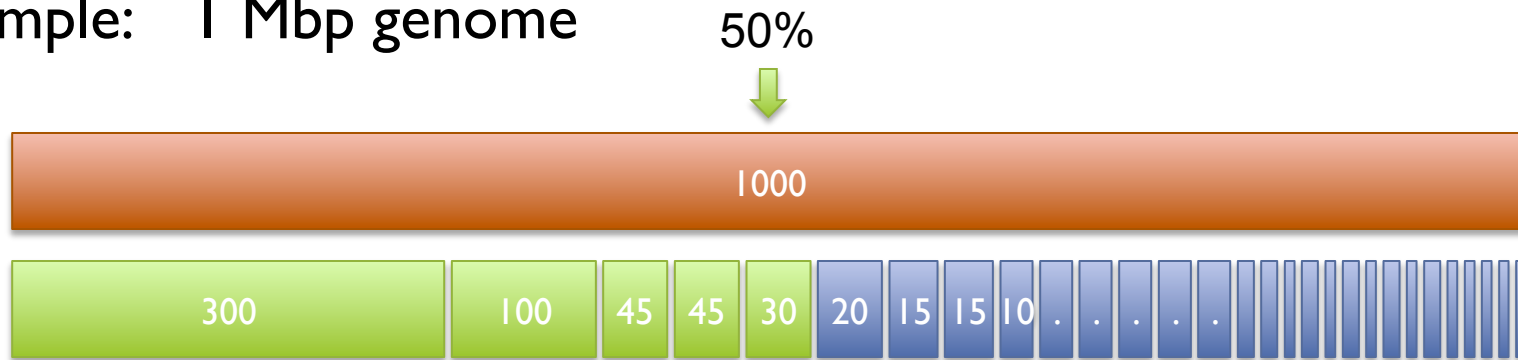
The advantages of SMRT sequencing

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



N50 size = 30 kbp

$(300\text{k} + 100\text{k} + 45\text{k} + 45\text{k} + 30\text{k} = 520\text{k} \geq 500\text{kbp})$

Note:

A “good” N50 size is a moving target relative to other recent publications. 10-20kbp contig N50 is currently a typical value for most “simple” genomes.

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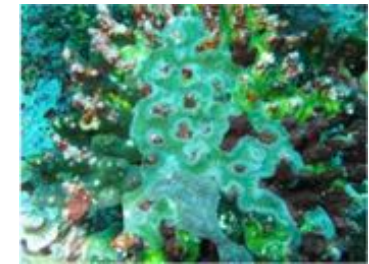


Assembly Applications

Novel genomes

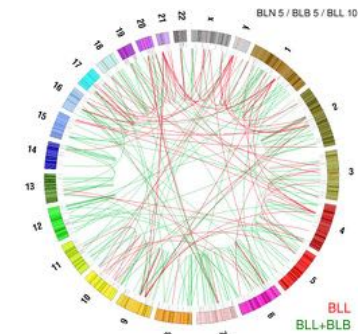
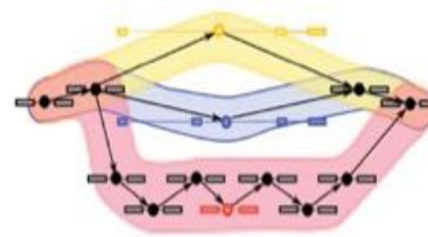


Metagenomes



Sequencing assays

- Transcript assembly
- Structural variations
- Haplotype analysis
- ...



Why are genomes hard to assemble?

1. Biological:

- (Very) High ploidy, heterozygosity, repeat content

2. Sequencing:

- (Very) large genomes, imperfect sequencing

3. Computational:

- (Very) Large genomes, complex structure

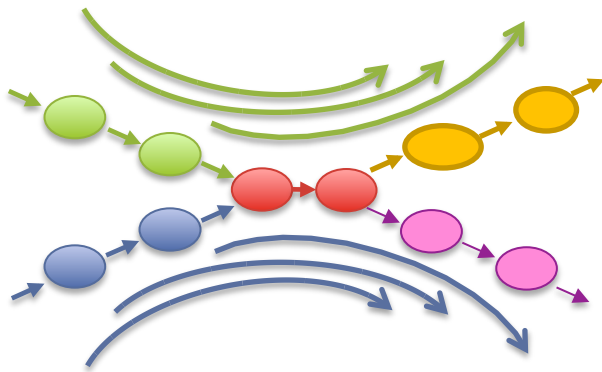
4. Accuracy:

- (Very) Hard to assess correctness



Ingredients for a good assembly

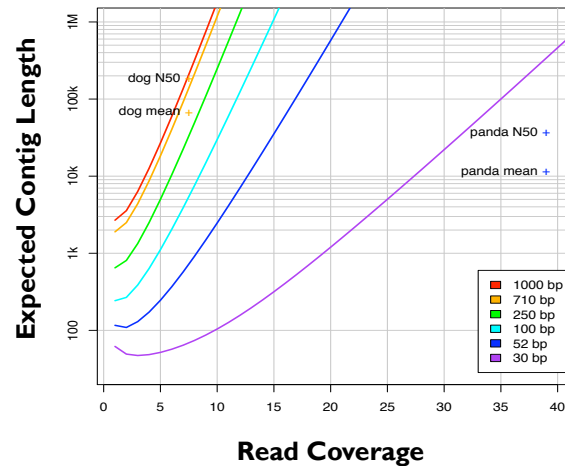
Read Length



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

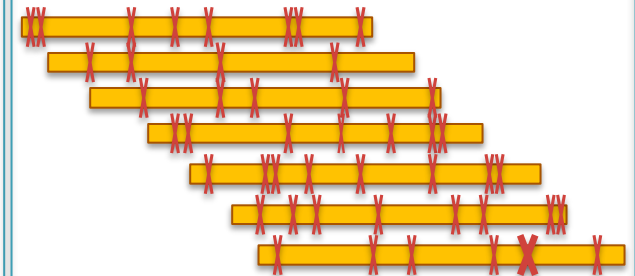
Coverage



High coverage is required

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

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

Hybrid Sequencing



Illumina

Sequencing by Synthesis

High throughput (60Gbp/day)

High accuracy (~99%)

Short reads (~100bp)



Pacific Biosciences

SMRT Sequencing

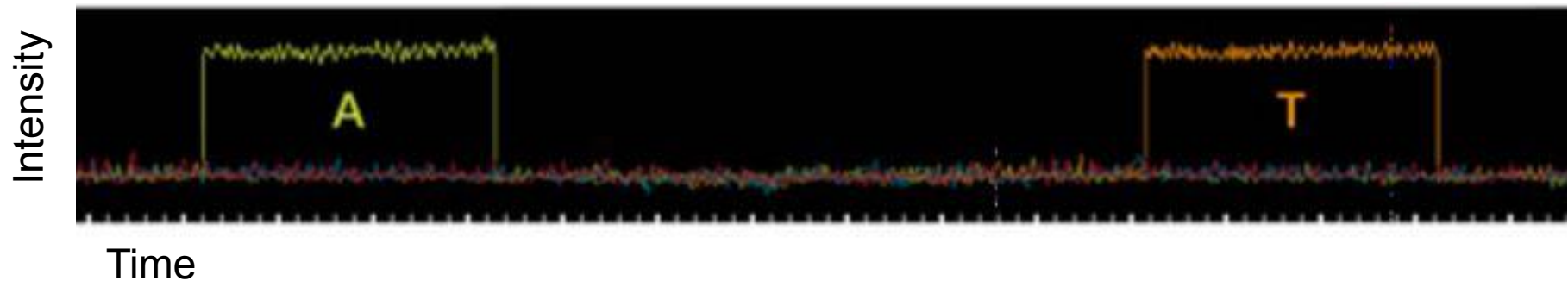
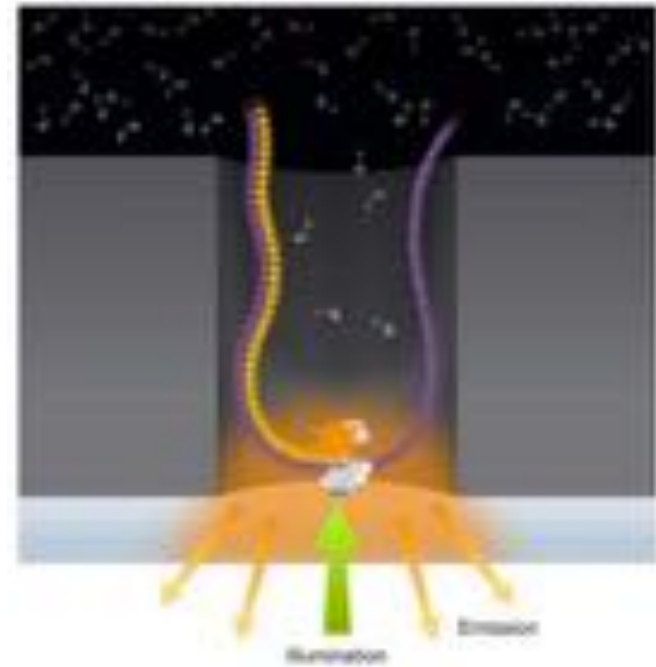
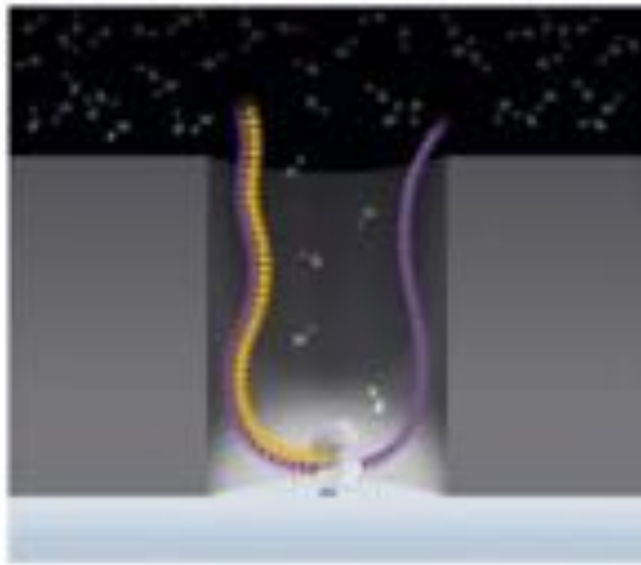
Lower throughput (1Gbp/day)

Lower accuracy (~85%)

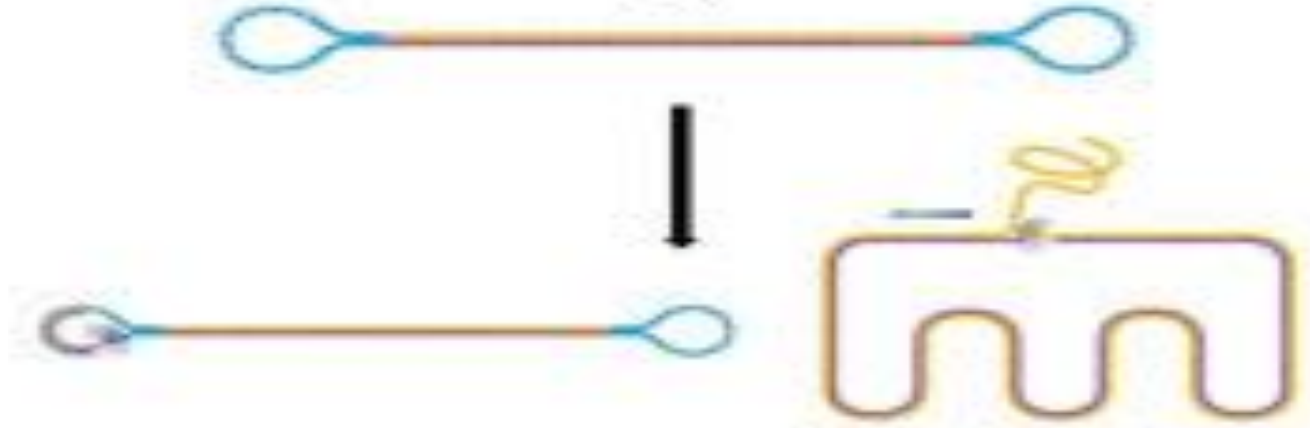
Long reads (5kbp+)

SMRT Sequencing

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

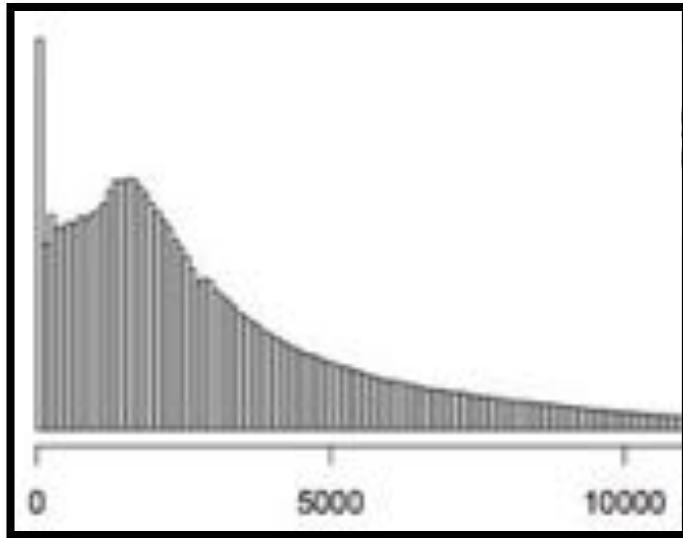


SMRT Read Types



- **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

SMRT Sequencing Data



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

TTGTAAGCAGTTGAAAACATATGTGTGGATTTAGAATAAAGAACATGAAAG
 |||
 TTGTAAGCAGTTGAAAACATATGTGT-GATTTAG-ATAAAGAACATGGAAG

ATTATAAA-CAGTTGATCCATT-AGAAGA-AAACGCAAAGGCCGGCTAGG
 |
 A-TATAAATCAGTTGATCCATTAGAA-AGAAACGC-AAAGGC-GCTAGG

CAACCTTGAATGTAATCGCACTTGAAGAACAAGATTTTATTCCGCGCCCG
 |
 C-ACCTTG-ATGT-AT--CACTTGAAGAACAAGATTTTATTCCGCGCCCG

TACGAATCAAGATTCTGAAAACACAT-ATAACAACCTCCAAAA-CACAA
 |
 T-ACGAATC-AGATTCTGAAAACA-ATGAT----ACCTCCAAAAGCACAA

-AGGAGGGGAAAAGGGGGGAATATCT-ATAAAAGATTACAAATTAGA-TGA
 |||
 GAGGAGG---AA-----GAATATCTGAT-AAAGATTACAAATT-GAGTGA

ACT-AATTCACAATA-AATAACACTTTTA-ACAGAATTGAT-GGAA-GTT
 |||
 ACTAAATTCACAA-ATAATAACACTTTTAGACAAATTGATGGGAAGGTT

TCGGAGAGATCCAAAACAATGGGC-ATCGCCTTTGA-GTTAC-AATCAA
 |||
 TC-GAGAGATCC-AAACAAT-GGCGATCG-CCTTGCAGTTACAAATCAA

ATCCAGTGAAAAATATAATTTATGCAATCCAGGAACCTTATTCACAATTAG
 |||
 ATCCAGT-GAAAAATATA--TTATGC-ATCCA-GAACCTTATTCACAATTAG

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

PacBio Error Correction

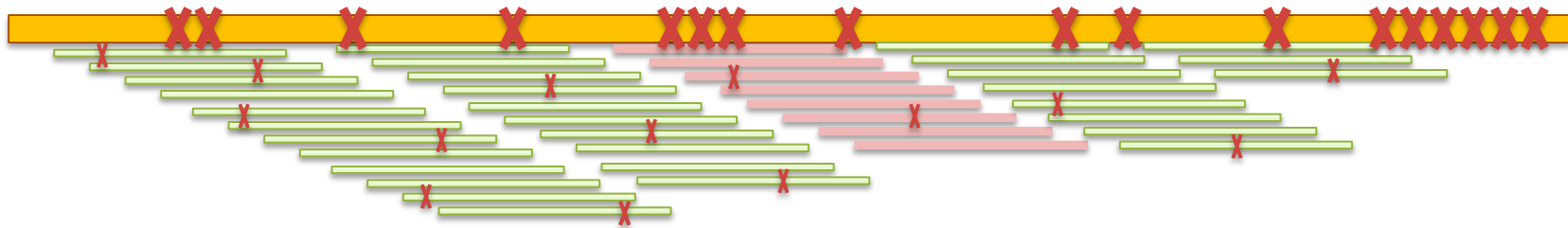
<http://wgs-assembler.sf.net>



I. Correction Pipeline

1. Map short reads to long reads
2. Trim long reads at coverage gaps
3. Compute consensus for each long read

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, et al. (2012) *Nature Biotechnology*. doi:10.1038/nbt.2280

Enhanced PacBio Error Correction

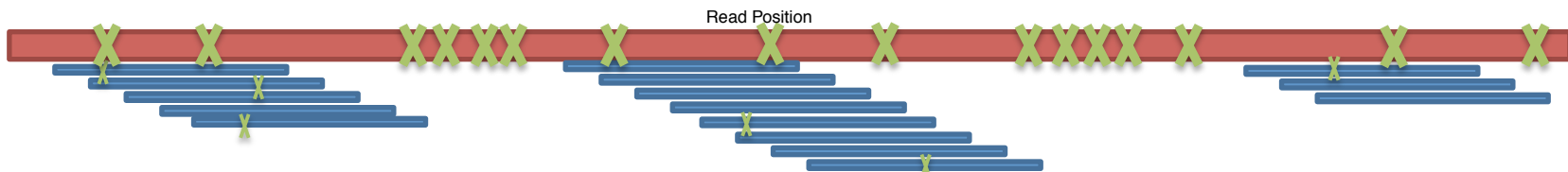
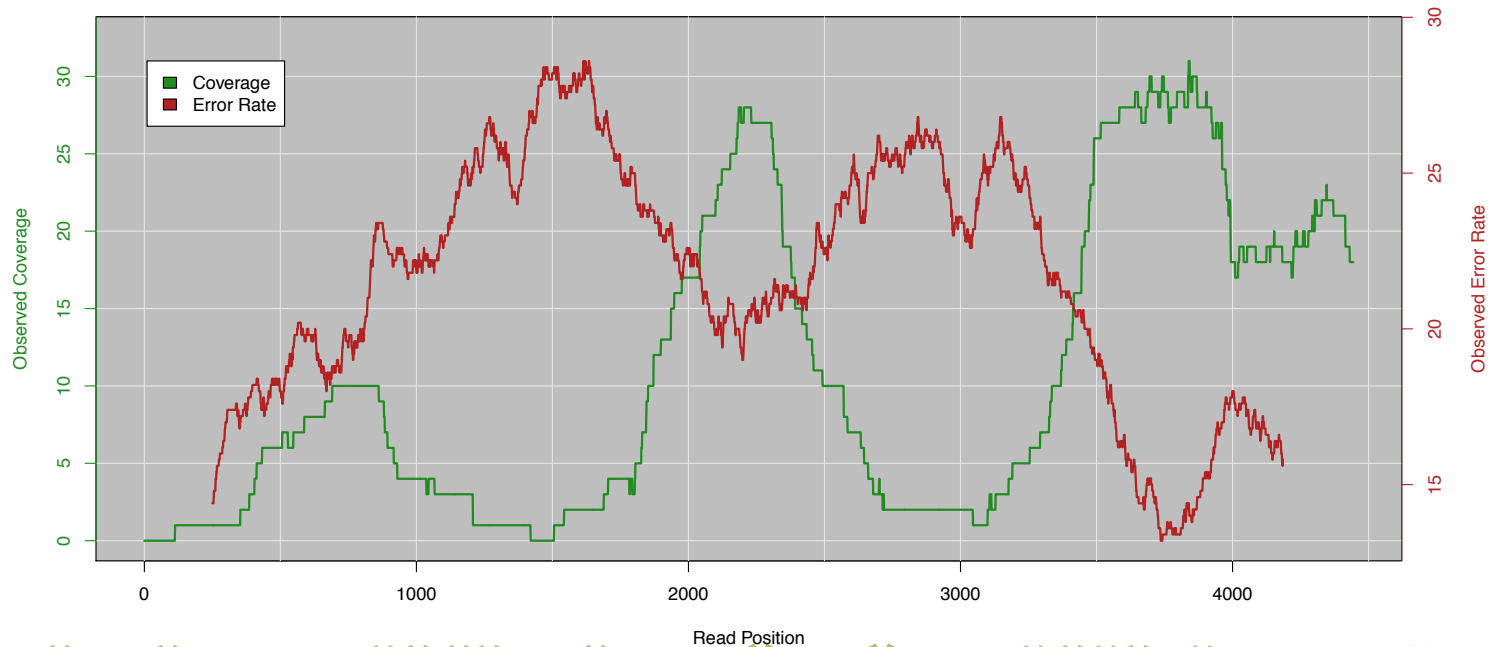
<https://github.com/jgurtowski/pbtools>

PacBioToCA fails in complex regions

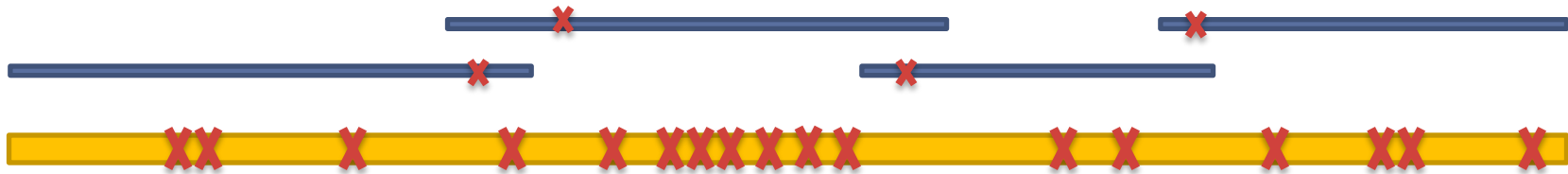
1. Simple Repeats – Kmer Frequency Too High to Seed Overlaps
2. GC Rich Regions – Known Illumina Bias
3. Error Dense Regions – Difficult to compute overlaps with many errors



Position Specific Coverage and Error Rate



Correction with Unitigs



Unitigs:

High quality contigs formed from unambiguous, unique overlaps of reads

Illumina reads ->

Illumina unitigs ->

Map and error correct PacBio reads ->

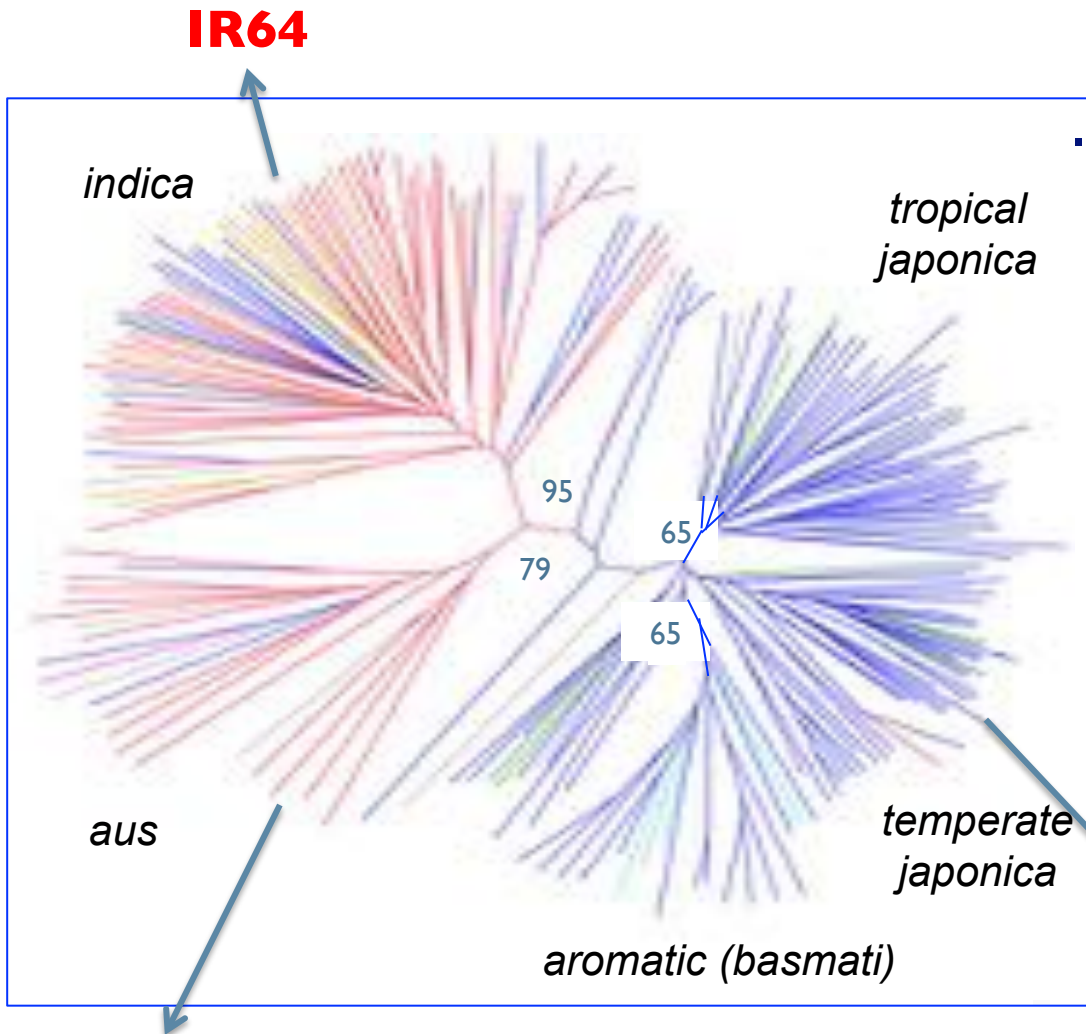
Assemble PacBio reads

Can Help us overcome:

- 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**

Population structure in *Oryza sativa*

3 varieties selected for *de novo* sequencing



High quality BAC-by-BAC reference

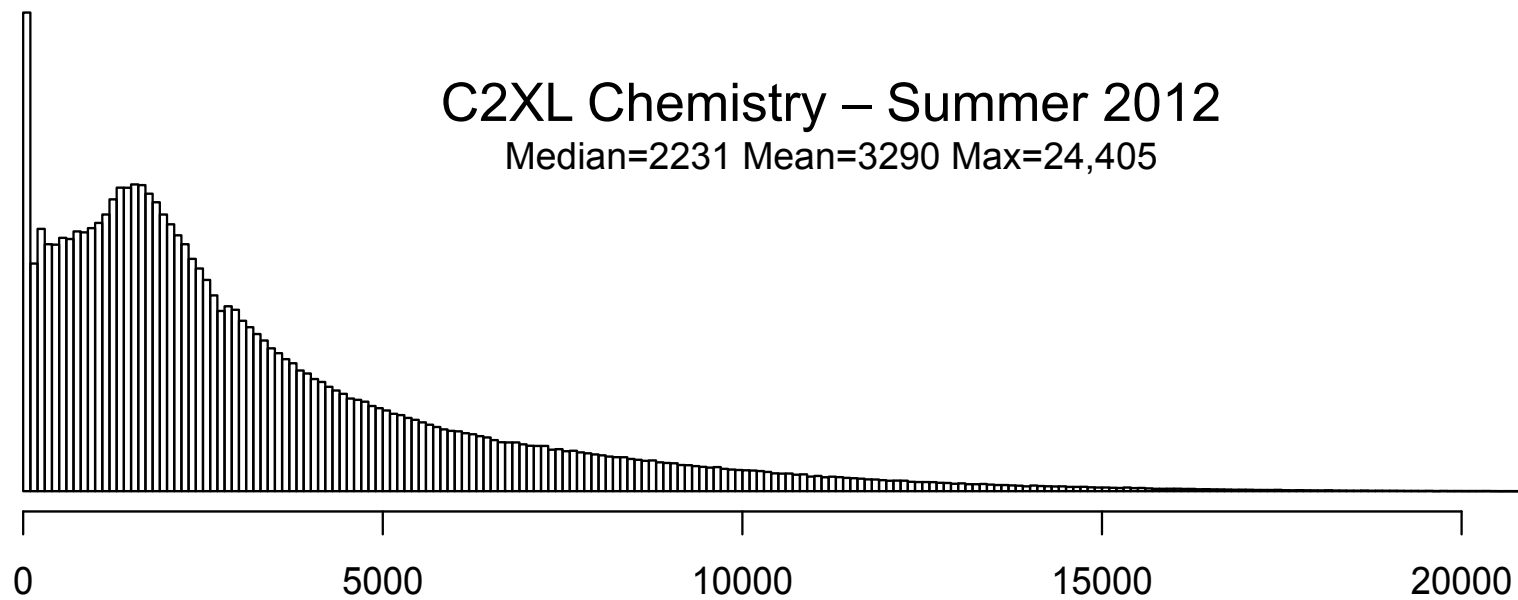
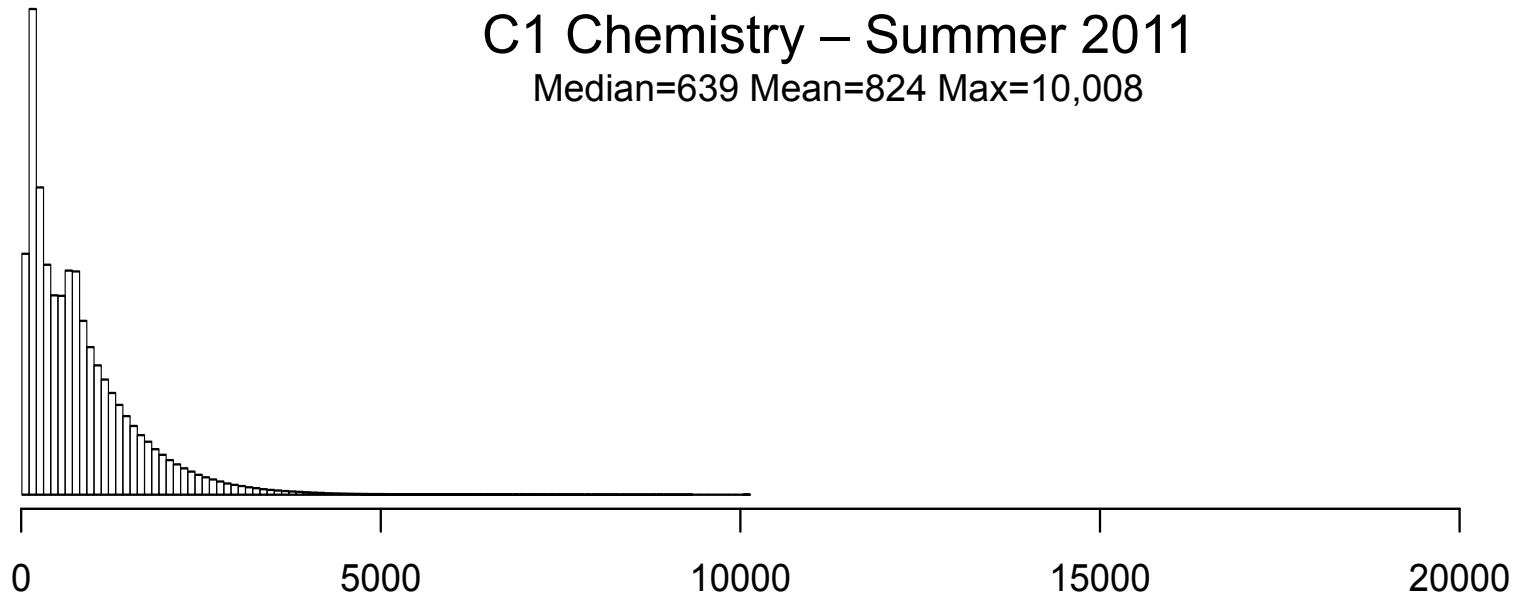
- ~370 Mbp genome in 12 chromosomes
- About 40% repeats:
 - Many 4-8kbp repeats
 - 300kbp max high identity repeat (99.99%)
- Useful model for other cereal genomes

DJI23

Garris et al. (2005)
Genetics 169: 1631–1638

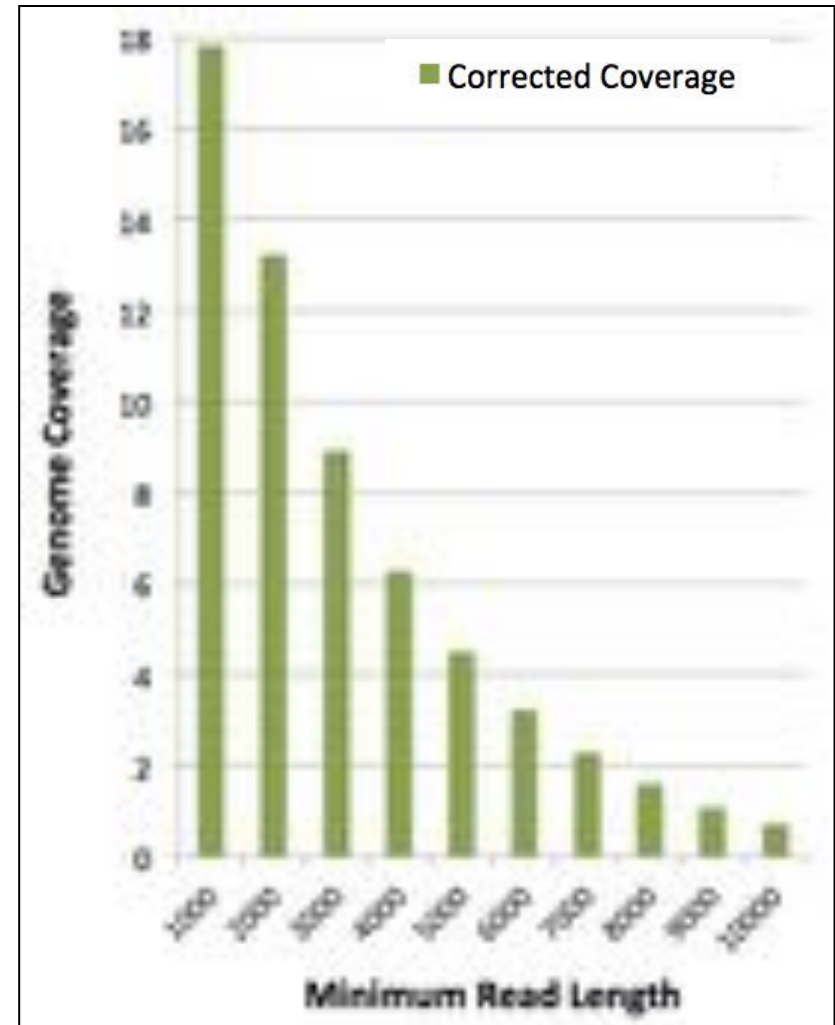
Nipponbare

PacBio Long Read Rice Sequencing



Preliminary Rice Assemblies

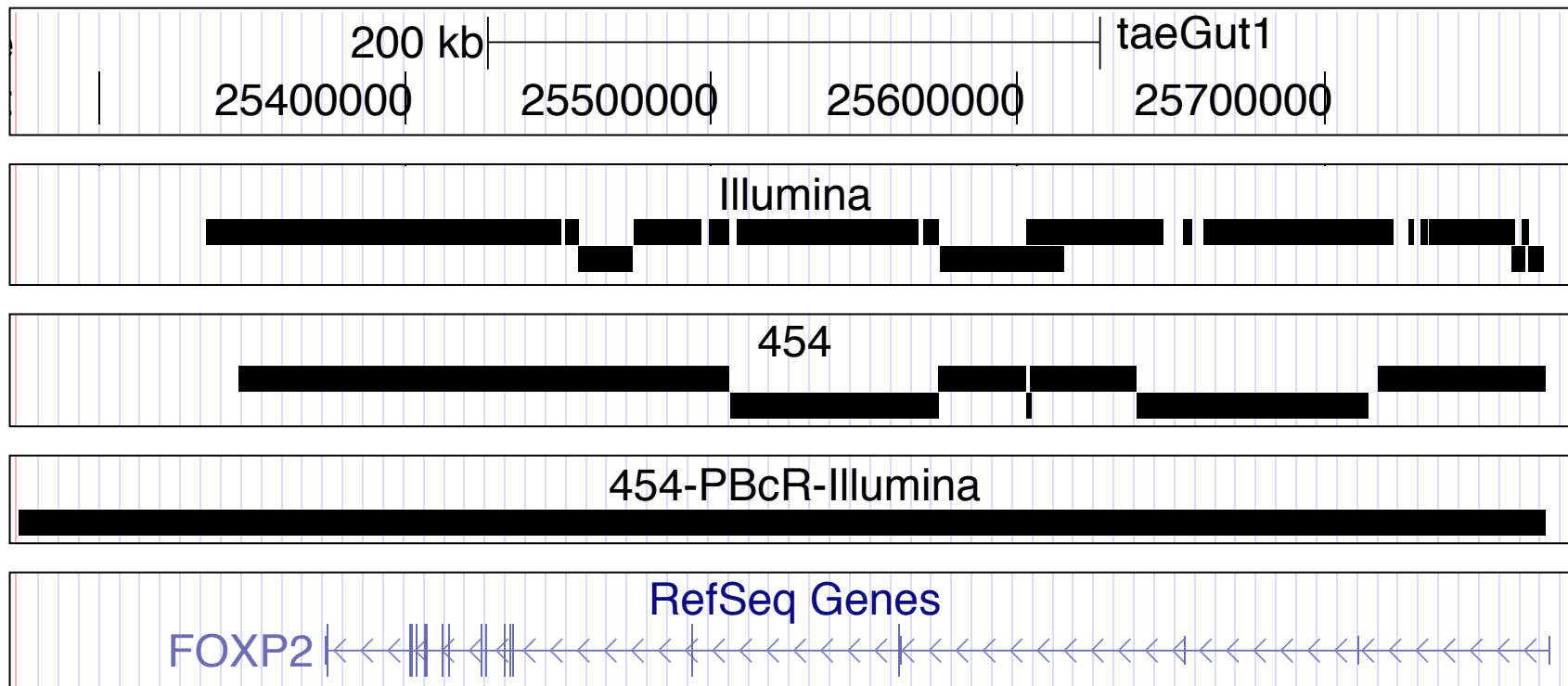
Assembly	Contig NG50
HiSeq Fragments 50x 2x100bp @ 180	3,925
MiSeq Fragments 23x 459bp 8x 2x251bp @ 450	6,332
“ALLPATHS-recipe” 50x 2x100bp @ 180 36x 2x50bp @ 2100 51x 2x50bp @ 4800	18,248



In collaboration with McCombie & Ware labs @ CSHL

Improved Gene Reconstruction

FOXP2 assembled in a single contig in the PacBio parrot assembly

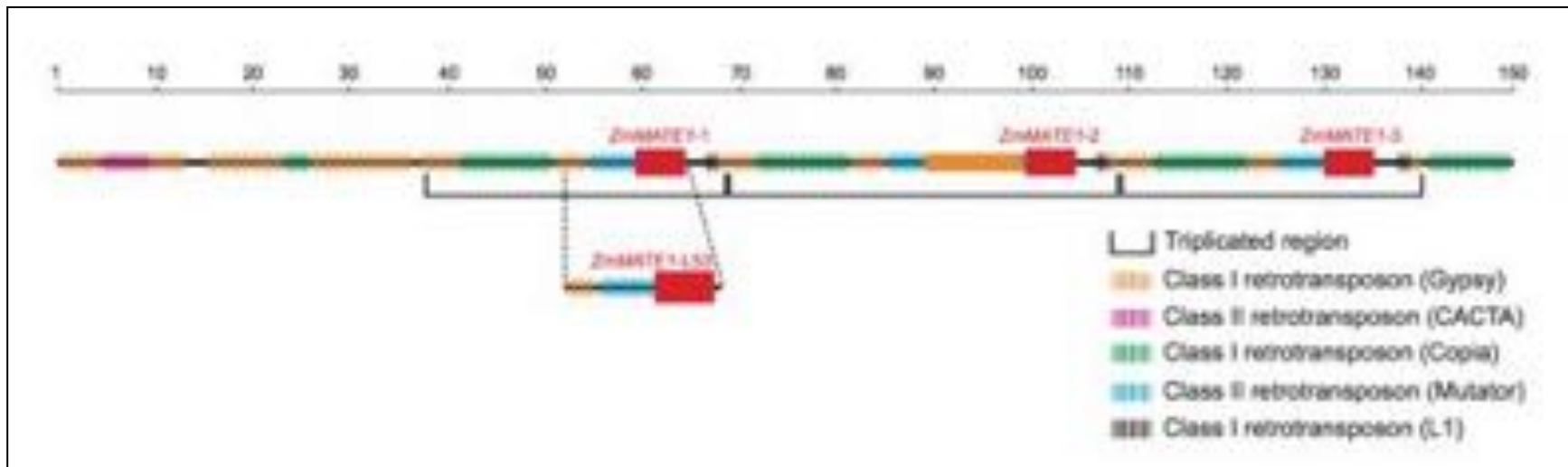


Hybrid error correction and de novo assembly of single-molecule sequencing reads.
Koren, S, Schatz, MC, et al. (2012) *Nature Biotechnology*. doi:10.1038/nbt.2280

Long Read CNV Analysis

Aluminum tolerance in maize is important for drought resistance and protecting against nutrient deficiencies

- Segregating population localized a QTL on a BAC, but unable to genotype with Illumina sequencing because of high repeat content and GC skew
- Long read PacBio sequencing corrected by CCS reads revealed a triplication of the ZnMATE1 membrane transporter



A rare gene copy-number variant that contributes to maize aluminum tolerance and adaptation to acid soils

Maron, LG *et al.* (2012) *PNAS*. doi: 10.1073/pnas.1220766110

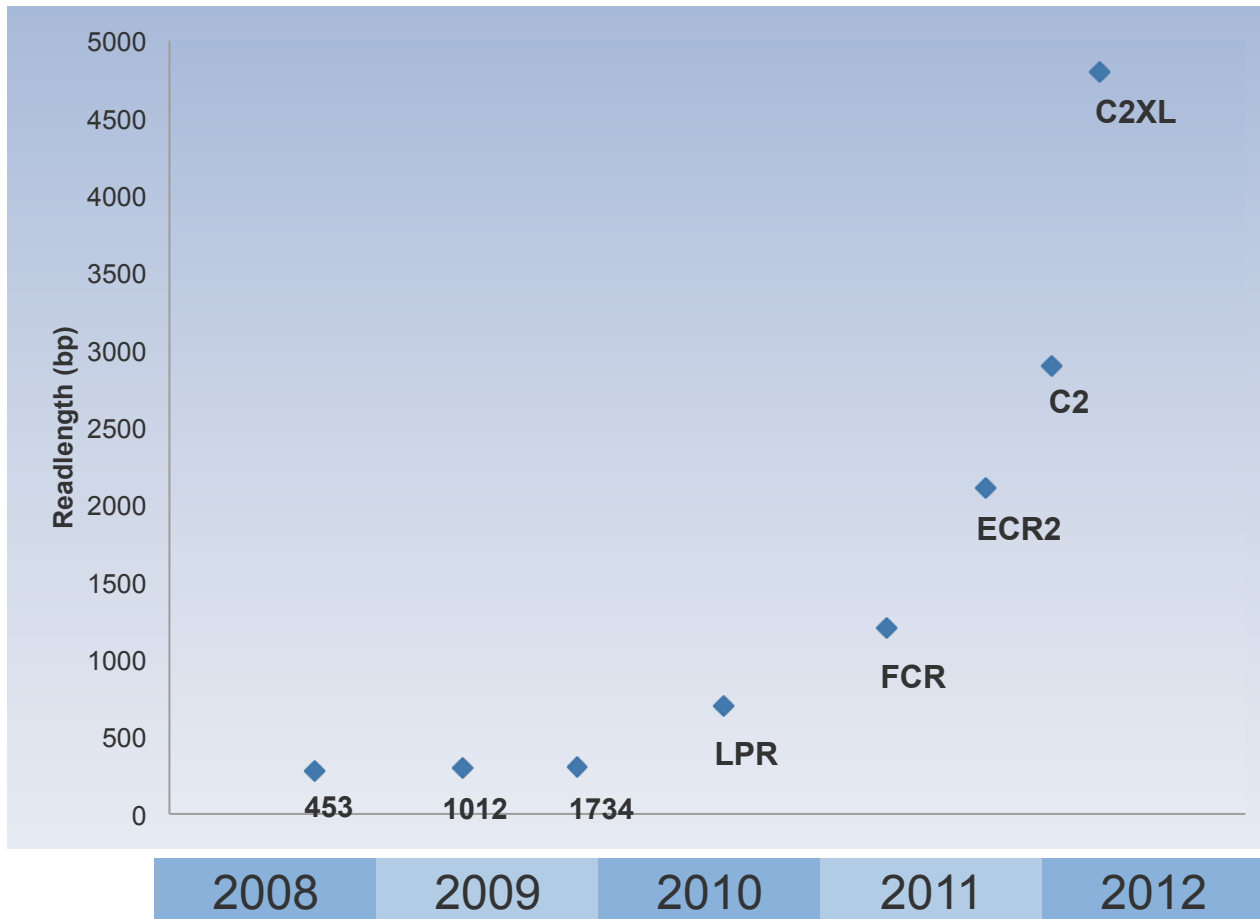
Transcript Alignment



- Long-read single-molecule sequencing has potential to directly sequence full length transcripts
 - Raw reads and raw alignments (red) have many spurious indels inducing false frameshifts and other artifacts
 - Error corrected reads almost perfectly match the genome, pinpointing splice sites, identifying alternative splicing
- New collaboration with Gingeras Lab looking at splicing in human

Hybrid error correction and de novo assembly of single-molecule sequencing reads.
Koren, S, Schatz, MC, et al. (2012) *Nature Biotechnology*. doi:10.1038/nbt.2280

PacBio Technology Roadmap

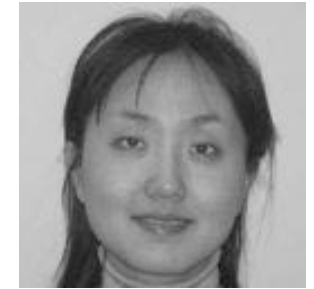
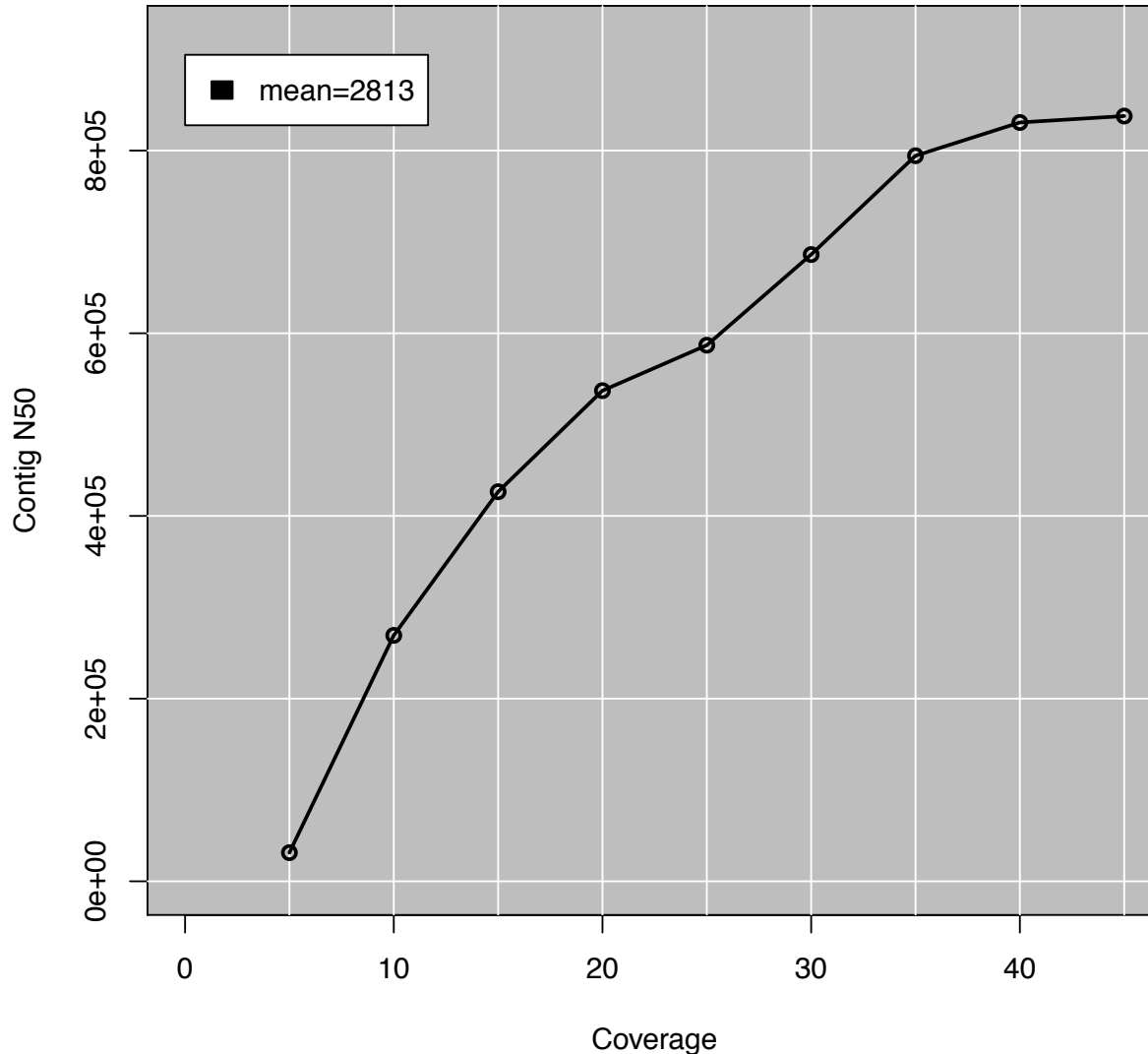


Internal Roadmap has made steady progress towards improving read length and throughput

Very recent improvements:

1. Improved enzyme:
Maintains reactions longer
2. “Hot Start” technology:
Maximize subreads
3. MagBead loading:
Load longest fragments

Assembly Coverage Model



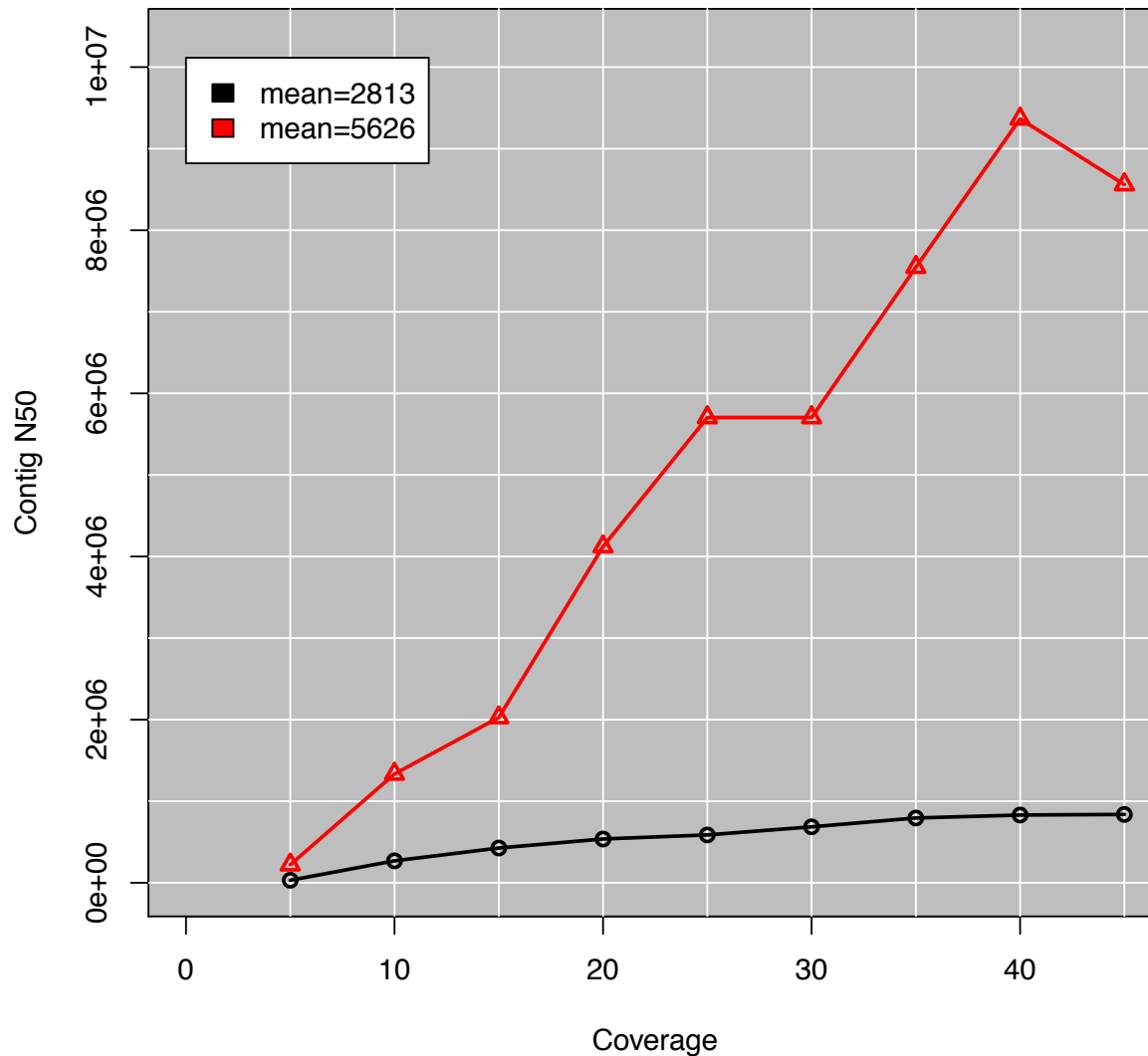
Simulate PacBio-like reads to predict how the assembly will improve as we add additional coverage

Only 8x coverage is needed to sequence every base in the genome, but 40x improves the chances repeats will be spanned by the longest reads

Assembly complexity of long read sequencing

Lee, H, Gurtowski, J, Marcus, S., Schatz MC et al. (2013) *In preparation*

Speculation for 2013



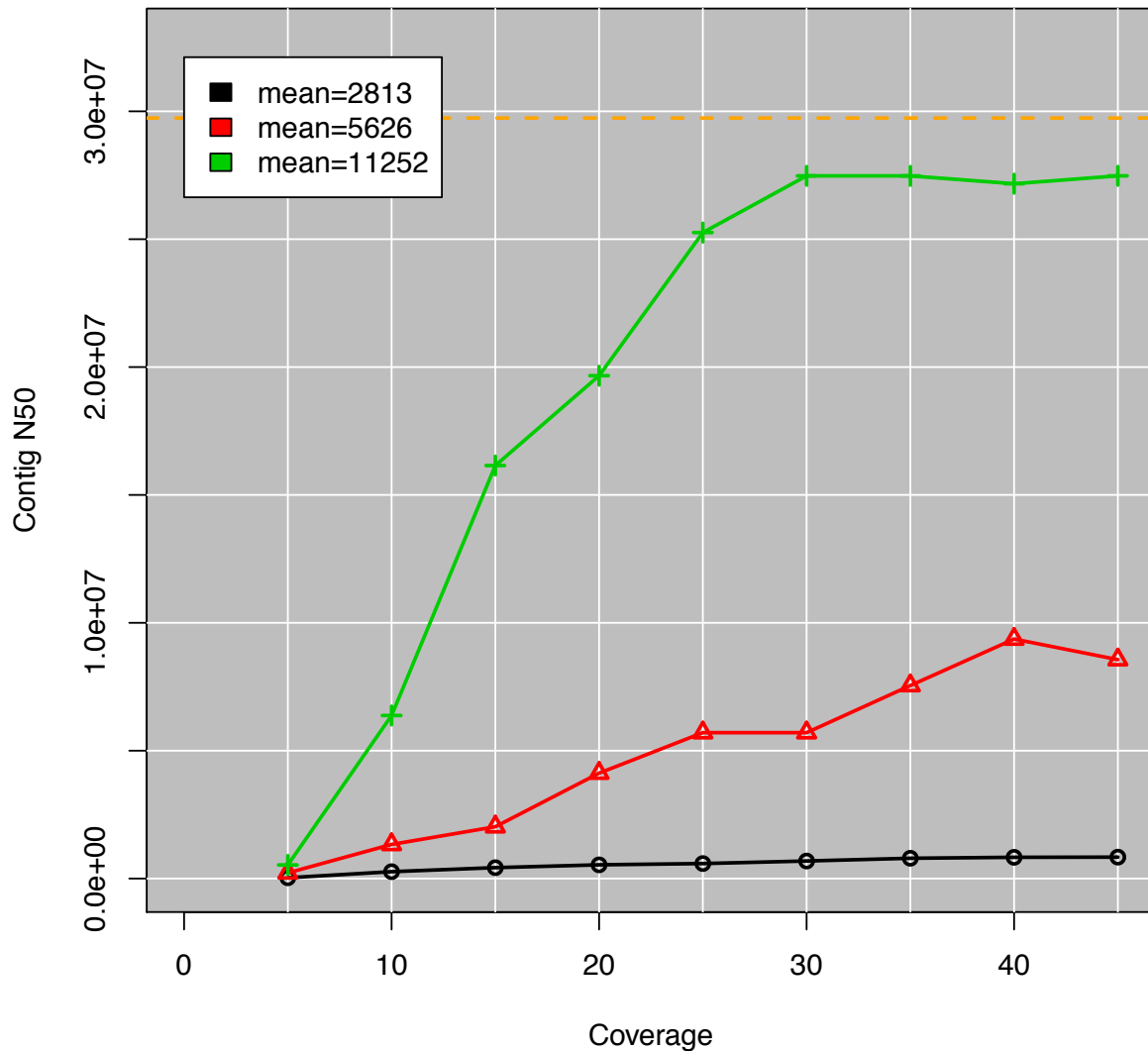
Doubling the average read length dramatically improves the assembly quality

- Able to span a larger repeats and lock contigs together

Expect to see contig N50 values over 1Mbp very soon, even in very complicated plant and animal species

- Megabase contig N50 already routine in microbial assembly with PacBio sequencing

Speculation for 2013

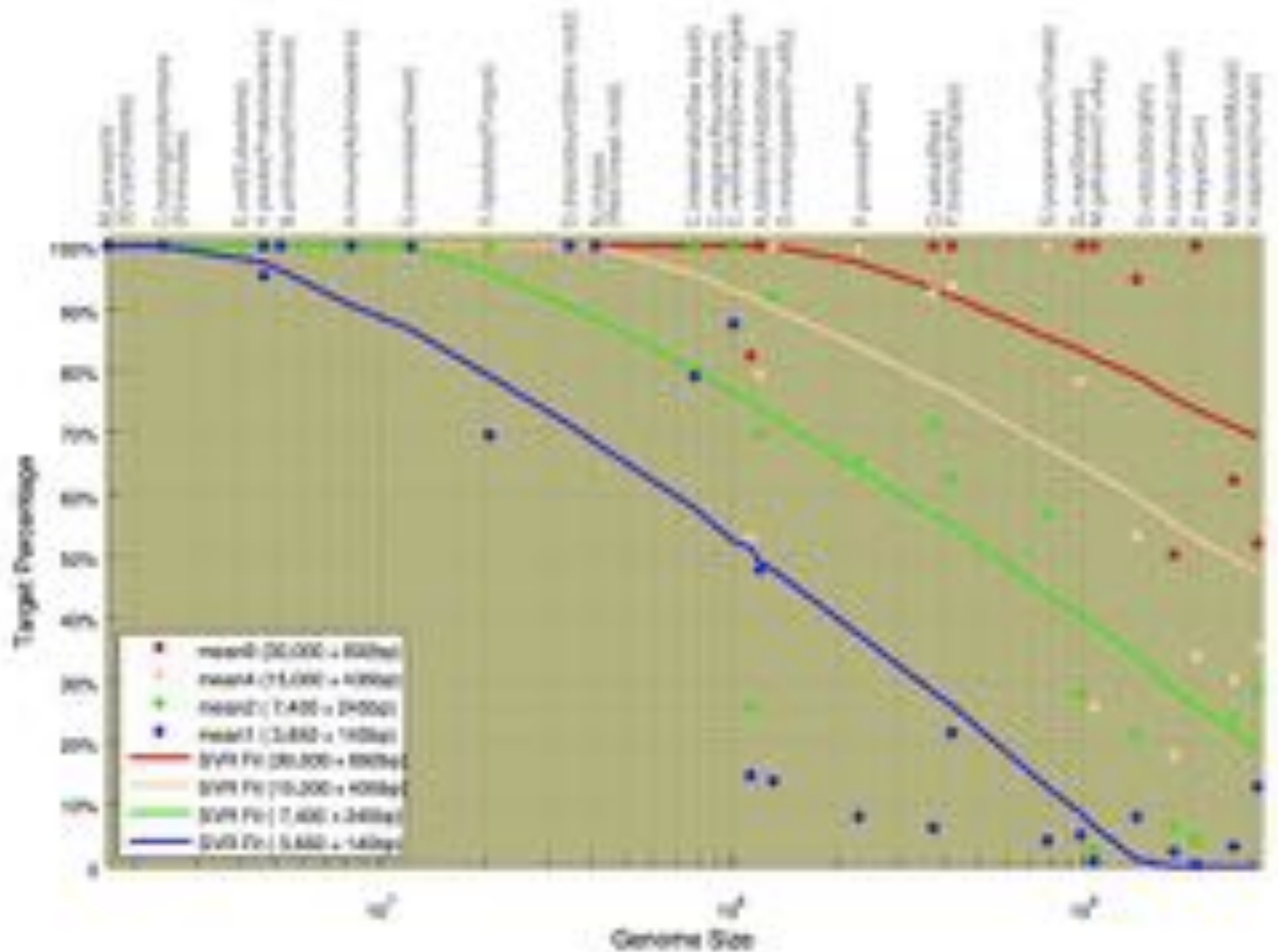


With PacBio-like reads averaging 11.2kbp (4x current), we should be able to assemble almost every chromosome arm of rice into single contigs

- The 300kbp near perfect repeat is the only exception

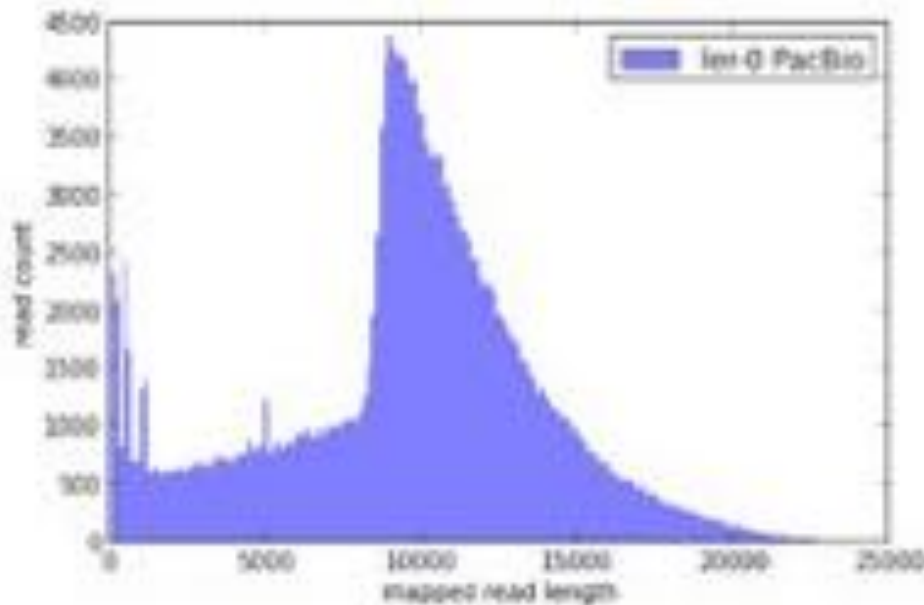
Even with the current assembly, we are seeing new genes and other sequences missing in the “high quality” BAC-by-BAC reference genome

Assembly Complexity of Long Reads



De novo assembly of Arabidopsis

<http://blog.pacificbiosciences.com/2013/08/new-data-release-arabidopsis-assembly.html>



A. thaliana Ler-0 sequenced at PacBio

- Sequenced using the latest P4 enzyme and C2 chemistry
- Size selection using an 8 Kb to 50 Kb elution window on a BluePippin™ device from Sage Science
- Total coverage >100x

Genome size: 124.6 Mb
GC content: 33.92%
Raw data: 11 Gb
Assembly coverage: 15x over 9kbp

Sum of Contig Lengths: 149.5Mb
Number of Contigs: 1788
Max Contig Length: 12.4 Mb
N50 Contig Length: 8.4 Mb

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Scalpel: Haplotype Microassembly

G. Narzisi, J. O’Rawe, I. Iossifov, Y. Lee, Z. Wang, G. Lyon, M. Wigler, and M. C. Schatz



DNA sequence **micro-assembly** pipeline for accurate detection and validation of *de novo* mutations (SNPs, indels) within exome-capture data.

Features

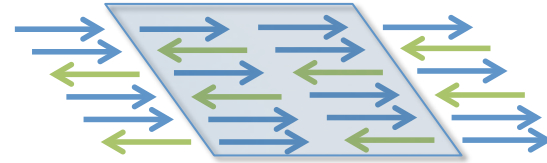
1. Combine **mapping** and **assembly**
2. Exhaustive search of **haplotypes**
3. **De novo mutations**



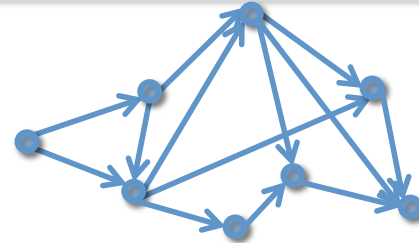
NRXN1 *de novo* SNP
(auSSC12501 chr2:50724605)

Scalpel Pipeline

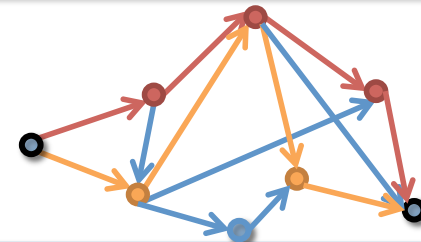
Extract reads mapping within the exon including (1) well-mapped reads, (2) soft-clipped reads, and (3) anchored pairs



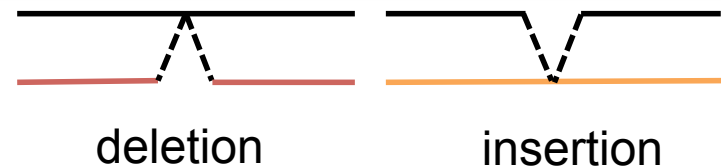
Decompose reads into overlapping k -mers and construct de Bruijn graph from the reads



Find end-to-end haplotype paths spanning the region

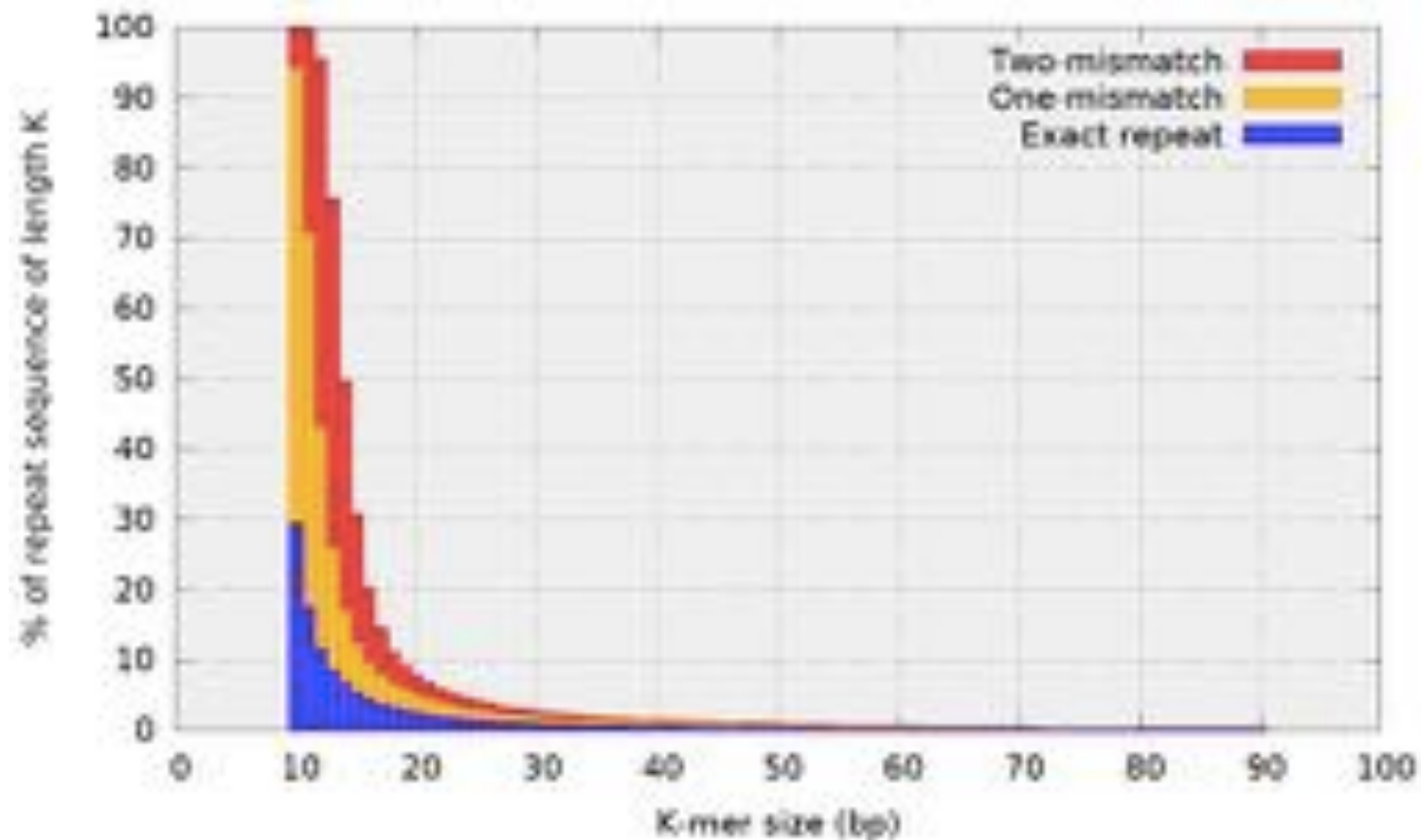


Align assembled sequences to reference to detect mutations



Repeats in the Genome

Specificity Challenge: 30% of exons have a perfect 10bp or larger repeat



Reference Exon: Localized repeat sequence

TTAGCAAGGTTGAAGAA C GGCTAAAAGCTTTGCCACATTTTGTAGGGTTTCTCTCCAGT A TGAATTCCTTATG TTAGCAAGGTTGAAGAA T GGCTAAAAGCTTTGCCACATTTTGTAGGGTTTCTCTCCAGT G TGTATCCTCTT
 AAGGTTGAAGAA C GGCTAAAAGCTTTGCCACATTTTGTAGGGTTTCTCTCCAGT G TGTATCCTCTT

Variant Read: Large deletion or critical snp?

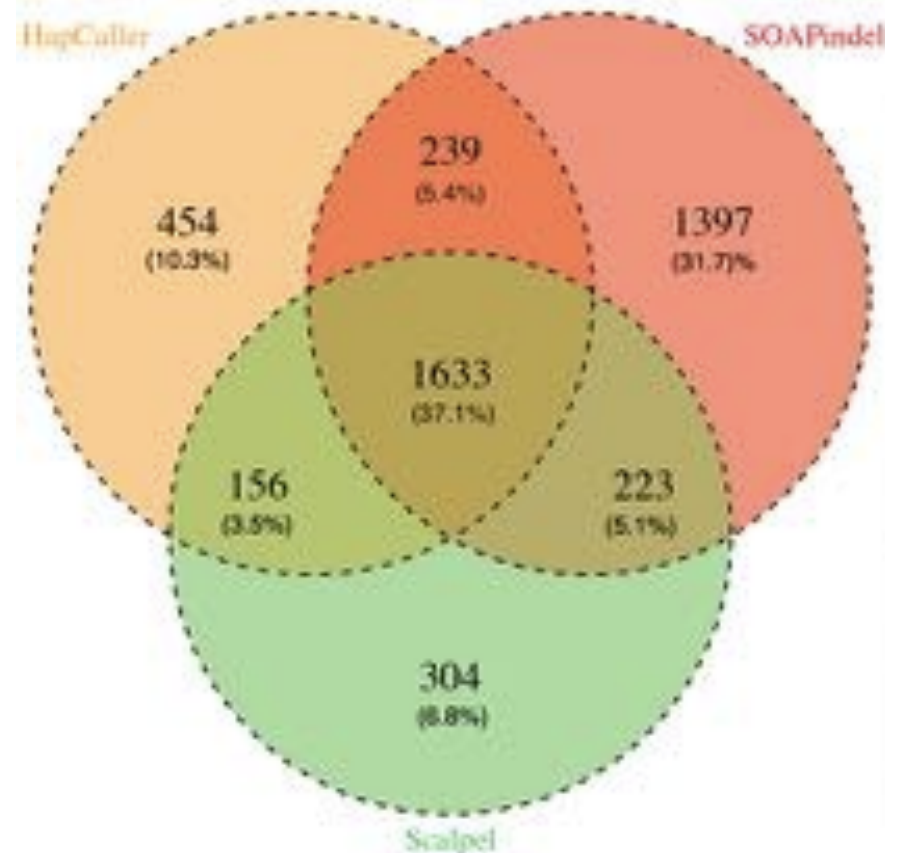
Experimental Analysis & Validation

Selected one deep coverage exome for deep analysis

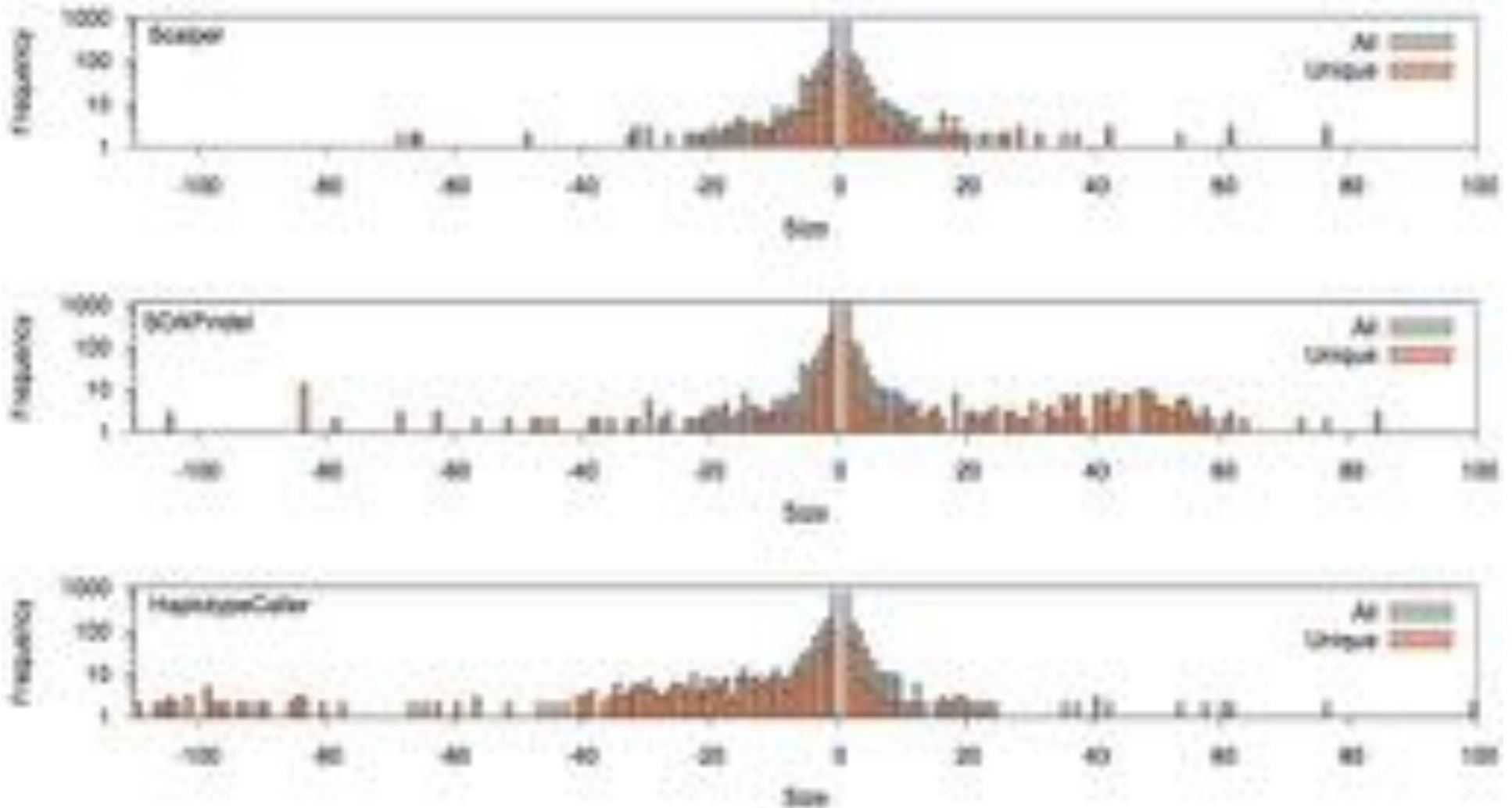
- 80% of the target at >20x coverage
- Evaluated with Scalpel, SOAPindel, and GATK Haplotype Caller

1000 indels selected for validation

- 200 Scalpel
- 200 GATK Haplotype Caller
- 200 SOAPindel
- 200 within the intersection
- 200 long indels (>30bp)

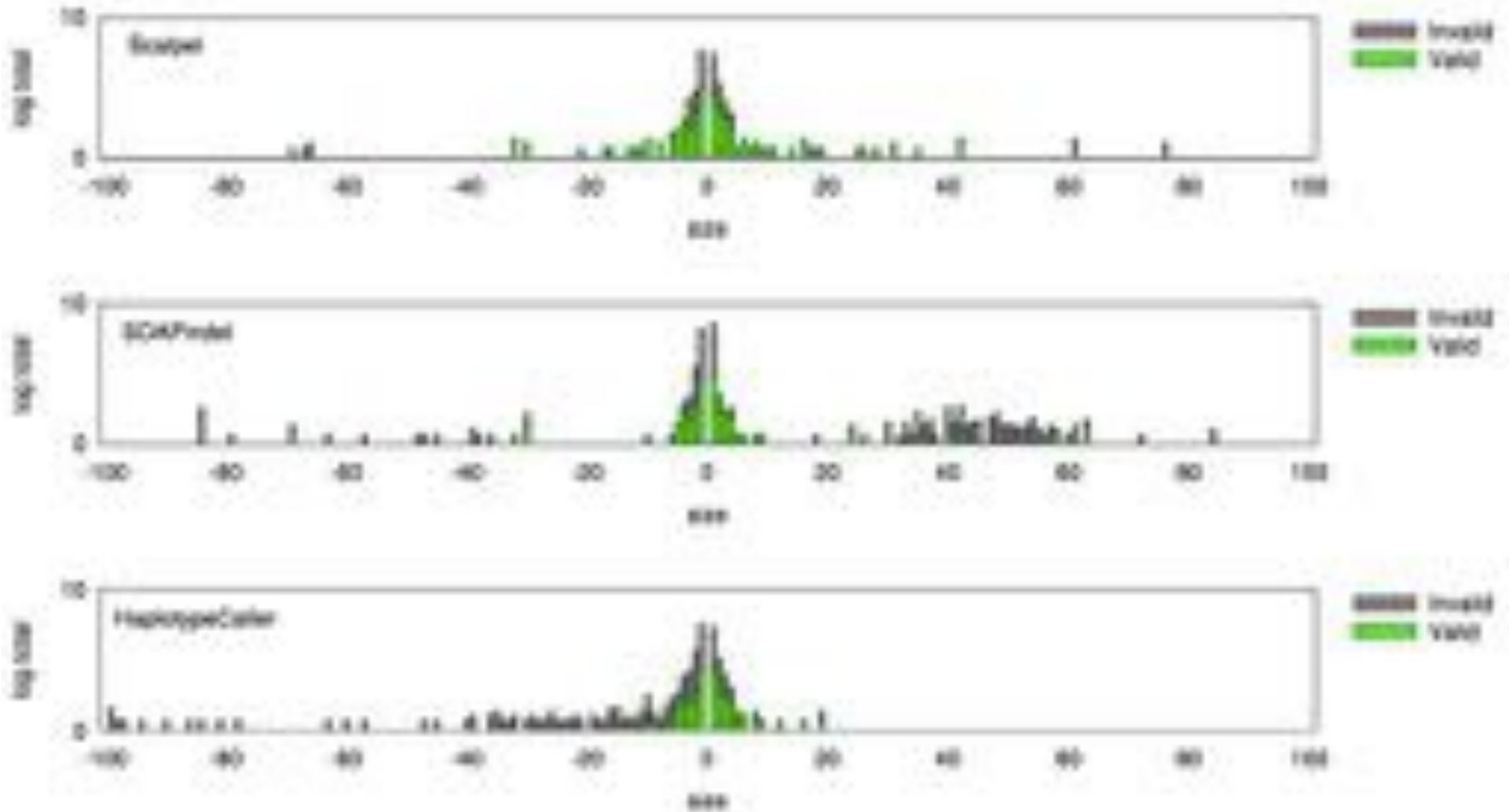


Scalpel Indel Discovery



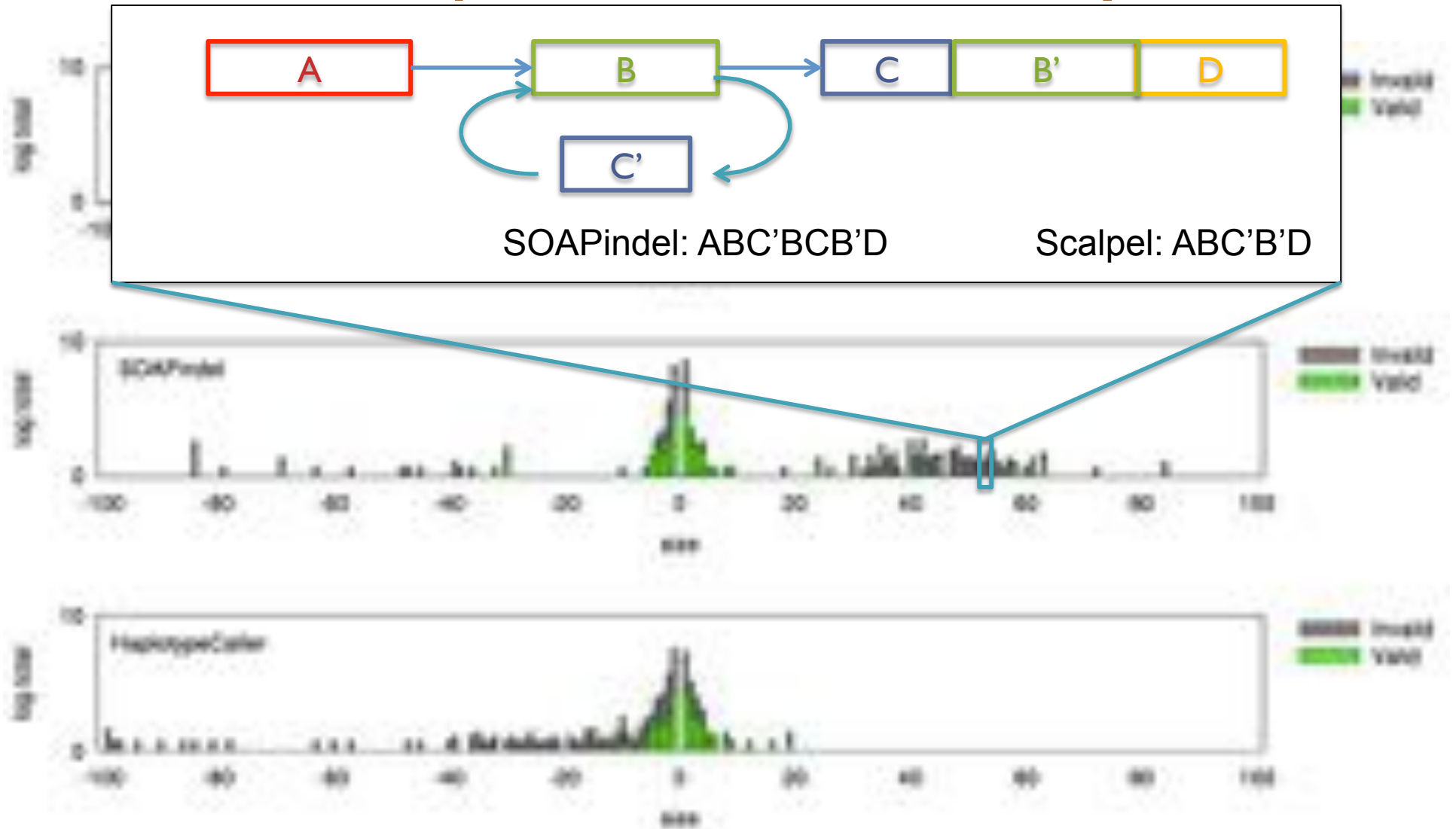
Detection of de novo mutations in exome-capture data using micro-assembly
Narzisi *et al.* (2013) *In preparation*

Scalpel Indel Discovery



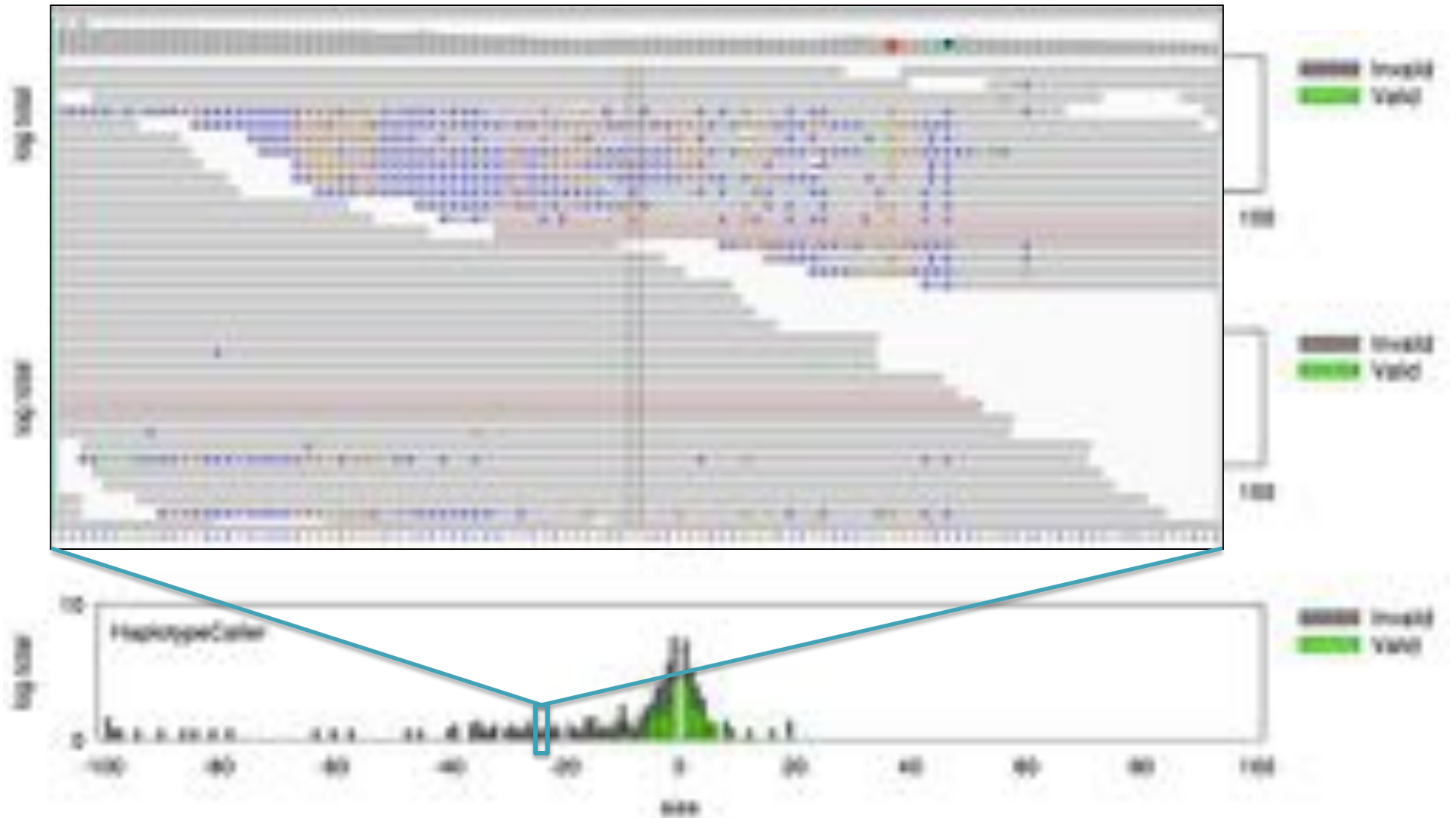
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Scalpel Indel Discovery



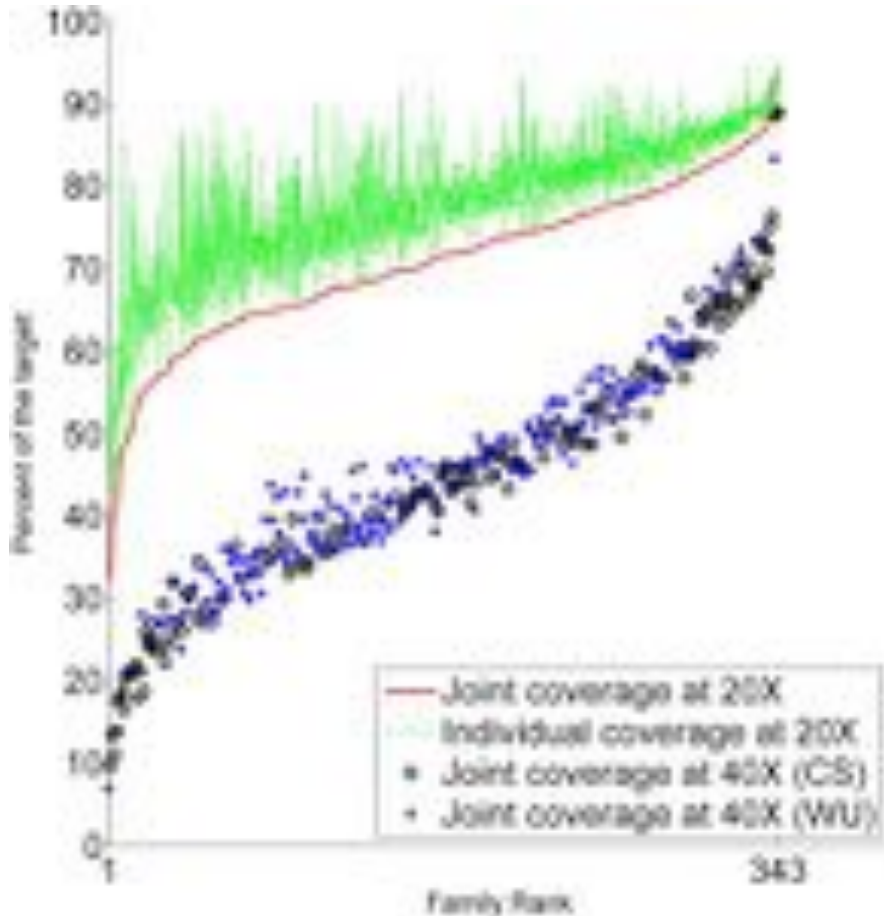
Detection of de novo mutations in exome-capture data using micro-assembly
Narzisi et al. (2013) *In preparation*

Scalpel Indel Discovery



Detection of de novo mutations in exome-capture data using micro-assembly
Narzisi et al. (2013) *In preparation*

Exome sequencing of the SSC



Sequencing of 343 families from the Simons Simplex Collection

- Parents plus one child with autism and one non-autistic sibling
- Enriched for higher-functioning individuals

Families prepared and captured together to minimize batch effects

- Exome-capture performed with NimbleGen SeqCap EZ Exome v2.0 targeting 36 Mb of the genome.
- ~80% of the target at >20x coverage with ~93bp reads

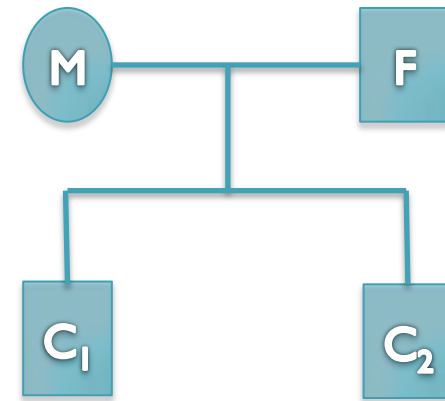
De novo gene disruptions in children on the autism spectrum

Lossifov *et al.* (2012) *Neuron*. 74:2 285-299

De novo mutation discovery and validation

Concept: Identify mutations not present in parents.

Challenge: Sequencing errors in the child or low coverage in parents lead to false positive de novos



Ref: ...TCAGAACAGCTGGATGAGATCTTAGCCAACCTACCAGGAGATTGTCTTTGCCCGGA...

Father: ...TCAGAACAGCTGGATGAGATCTTAGCCAACCTACCAGGAGATTGTCTTTGCCCGGA...

Mother: ...TCAGAACAGCTGGATGAGATCTTAGCCAACCTACCAGGAGATTGTCTTTGCCCGGA...

Sib: ...TCAGAACAGCTGGATGAGATCTTAGCCAACCTACCAGGAGATTGTCTTTGCCCGGA...

Aut(1): ...TCAGAACAGCTGGATGAGATCTTAGCCAACCTACCAGGAGATTGTCTTTGCCCGGA...

Aut(2): ...TCAGAACAGCTGGATGAGATCTTACC-----CCGGGAGATTGTCTTTGCCCGGA...

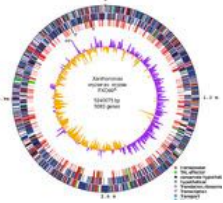
6bp heterozygous deletion at chr13:25280526 ATP12A

De novo Genetics of Autism

- In 343 family quads so far, we see significant enrichment in de novo **likely gene killers** in the autistic kids
 - Overall rate basically 1:1 (432:396)
 - 2:1 enrichment in nonsense mutations
 - 2:1 enrichment in frameshift indels
 - 4:1 enrichment in splice-site mutations
 - Most de novo originate in the paternal line in an age-dependent manner (56:18 of the mutations that we could determine)
- Observe strong overlap with the 842 genes known to be associated with fragile X protein FMR1
 - Related to neuron development and synaptic plasticity
 - Also strong overlap with chromatin remodelers

De novo gene disruptions in children on the autism spectrum

Iossifov *et al.* (2012) *Neuron*. 74:2 285-299



Summary



- Hybrid assembly let us combine the best characteristics of 2nd and 3rd gen sequencing
 - Long reads and good coverage are the keys to a good de novo assembly
 - Single contig de novo assemblies of entire microbial chromosomes are now routine; Single contig de novo assemblies of entire plant and animal chromosomes on the horizon
- Assembly is the missing link towards high accuracy indel mutation discovery
 - Allows the algorithm to break free from the expectations of the reference
 - Pinpointing de novo mutations require both high sensitivity and specificity
- We are starting to apply these technologies to discover significant biology that is otherwise impossible to measure

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Thank You!

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