

Metassembler: Improving de novo genome assembly

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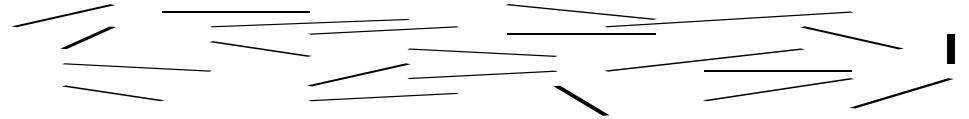
AGBT



@mike_schatz / #AGBT

Assembling a Genome

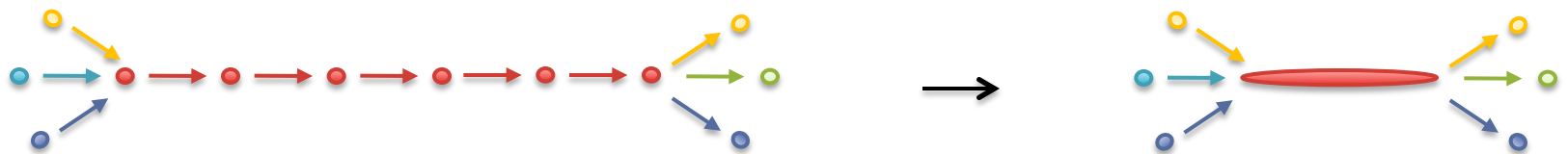
1. Shear & Sequence DNA



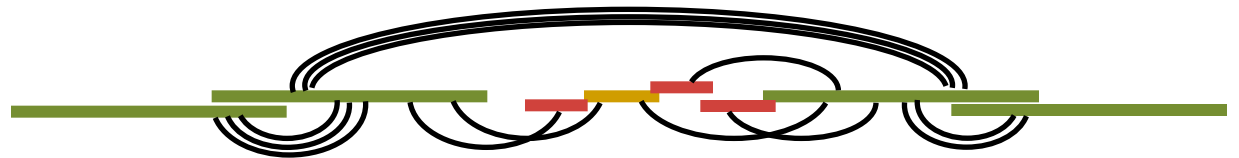
2. Construct assembly graph from overlapping reads

...AGCCTAGACCTACAGGATGCGCGACACGT
GGATGCGCGACACGTTCGCATATCCGGT...

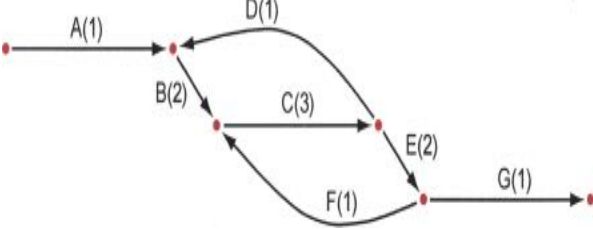
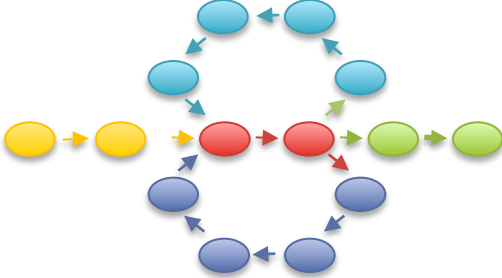
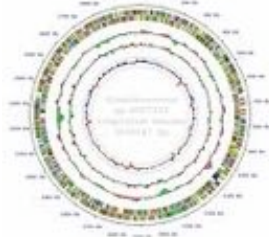
3. Simplify assembly graph



4. Detangle graph with long reads, mates, and other links



Genome Assemblers

ALLPATHS-LG	SOAPdenovo	Celera Assembler
		
<p>Broad's assembler (Gnerre et al. 2011)</p>	<p>BGI's assembler (Li et al. 2010)</p>	<p>JCVI's assembler (Miller et al. 2008)</p>
<p>Unipath graph Short + PacBio (patching)</p>	<p>De bruijn graph Short reads</p>	<p>Overlap graph Medium + Long reads</p>
<p>Easy to run if you have compatible libraries</p>	<p>Most flexible, but requires a lot of tuning</p>	<p>Supports Illumina/454/PacBio Hybrid assemblies</p>

Plus several dozens more
Each balancing the tension between connectivity and accuracy in a different way

2011: Year of the Assembly Bakeoff

THE ASSEMBLATHON



Genome Assembly Gold-Standard Evaluations

- Simulated genome distantly related to human chr13
- 17 labs, 50+ assemblies
- 4 real genomes ranging from bacteria to individual human chromosome
- Internal evaluation of 8 assemblers

Assemblathon 1: A competitive assessment of de novo short read assembly methods.

Earl, DA *et al.* (2011) *Genome Research*. In press.

GAGE: A critical evaluation of genome assemblies and assembly algorithms.

Salzberg, SL *et al.* (2011) *Genome Research*. In press.

Final Rankings

ID	Overall	CPNG50	SPNG50	Struct.	CC50	Subs.	Copy. Num.	Cov. Tot.	Cov. CDS
BGI	36	★					★	★	★
Broad	37	★	★	★	★				
WTSI-S	46		★	★	★	★			
CSHL	52	★							★
BCCGSC	53							★	★
DOEJGI	56		★	★	★	★			
RHUL	58								
WTSI-P	64							★	
EBI	64						★		
CRACS	64					★			

- SOAPdenovo and ALLPATHS came out neck-and-neck followed closely behind by SGA, Celera Assembler, and ABySS
- My recommendation for “typical” short read assembly is to use ALLPATHS

Assemblathon 2

- Real sequence data, *de novo* assembly



- Step 1: Apply best practices from Assemblathon 1
- Step 2: Add secret weapon for winning...

Images from Assemblathon

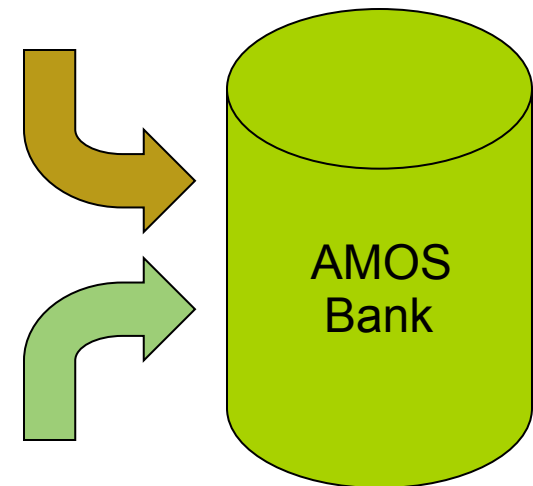
Assembly Forensics

Computationally scan an assembly for mis-assemblies.

- Data inconsistencies are indicators for mis-assembly
- Some inconsistencies are merely statistical variations

AMOSvalidate

1. Analyze Mate Pairs & Libraries
2. Analyze Depth of Coverage
3. Analyze Read Alignments
4. Analyze Read Breakpoints
5. Load Mis-assembly Signatures into Bank



Genome Assembly forensics: finding the elusive mis-assembly.

Phillippy, AM, Schatz, MC, Pop, M. (2008) *Genome Biology* 9:R55.

Hawkeye & AMOS: Visualizing and assessing the quality of genome assemblies

Schatz, MC *et al.* (2012) *Briefings in Bioinformatics*. In Press.

Mate Evaluation

- Correct: mates have expected orientation and separation



- Mis-assembled: mates have incorrect orientation and separation

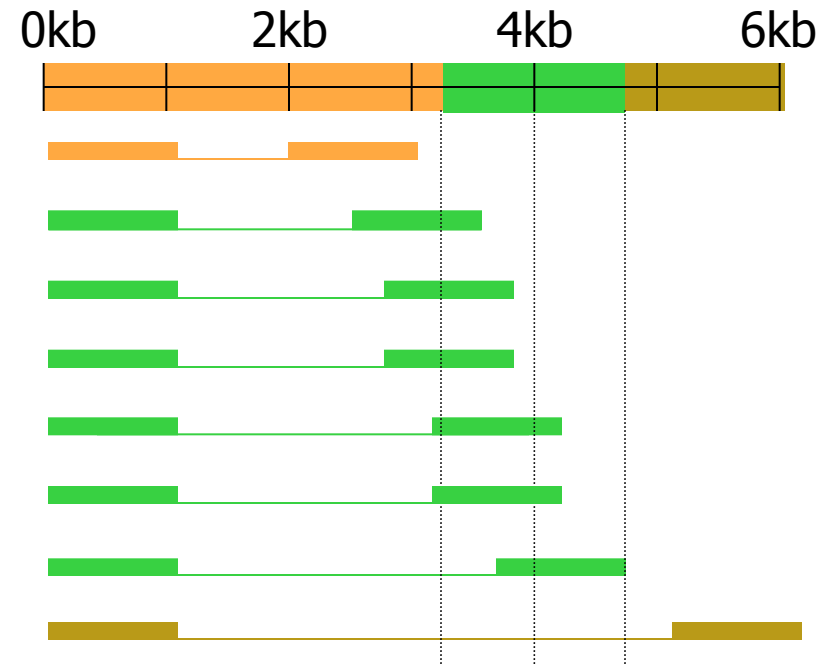
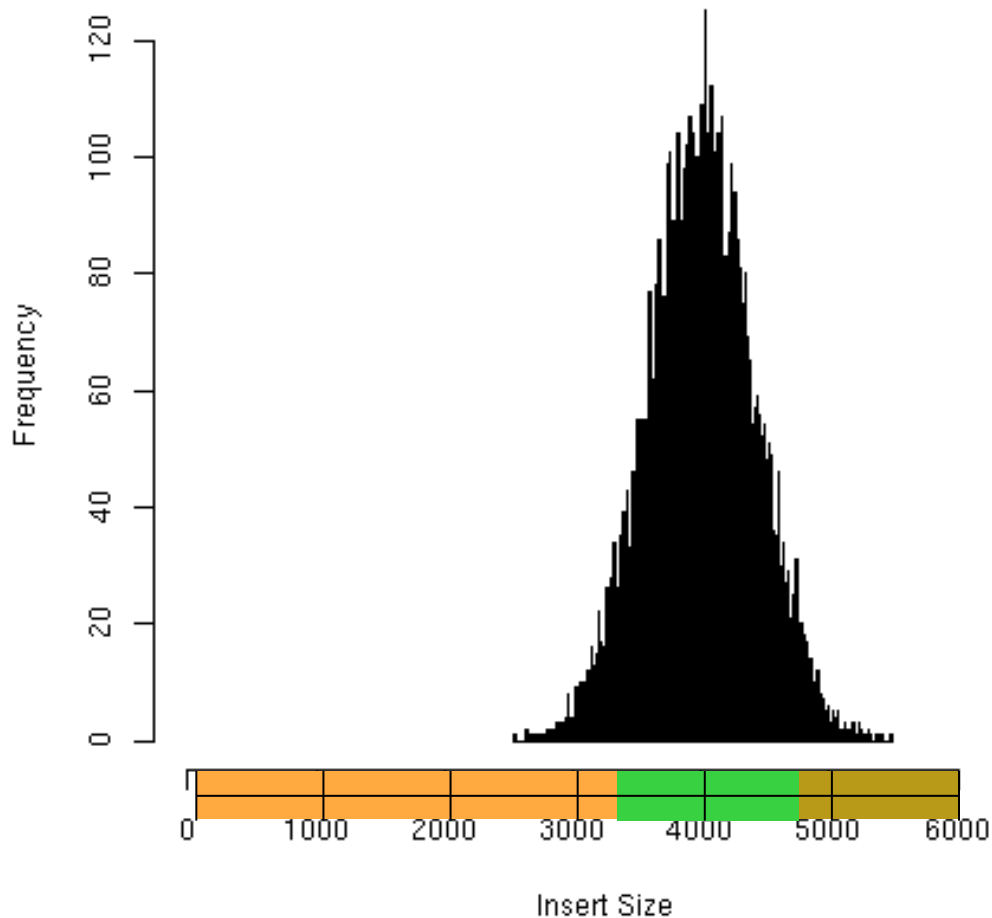


- Slightly compressed/expanded mates are expected because mates are sampled from a distribution of fragments

Compression/Expansion Statistic

Library size distribution

Mean: 4000, SD: 400



8 inserts: 3kb-6kb

Local Mean: 4048

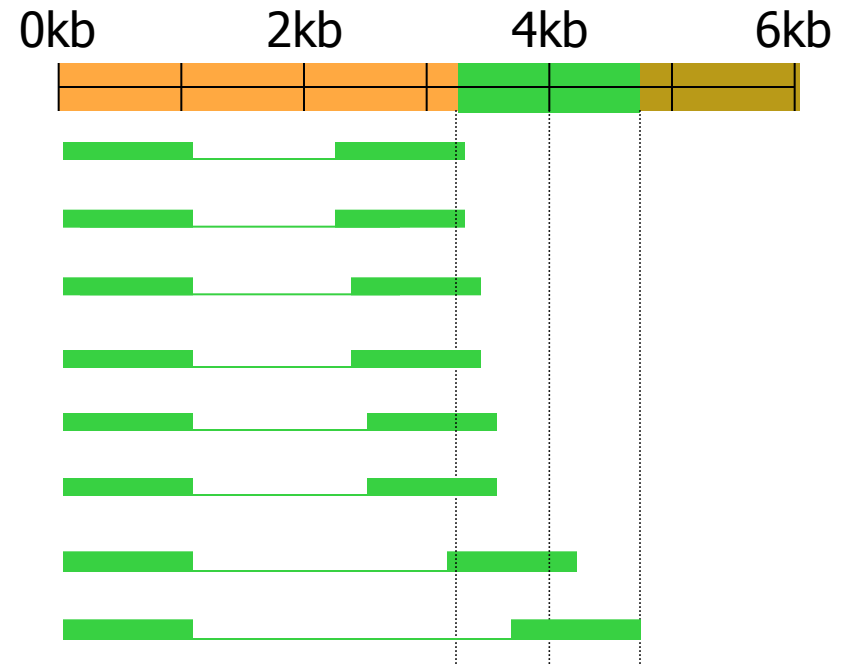
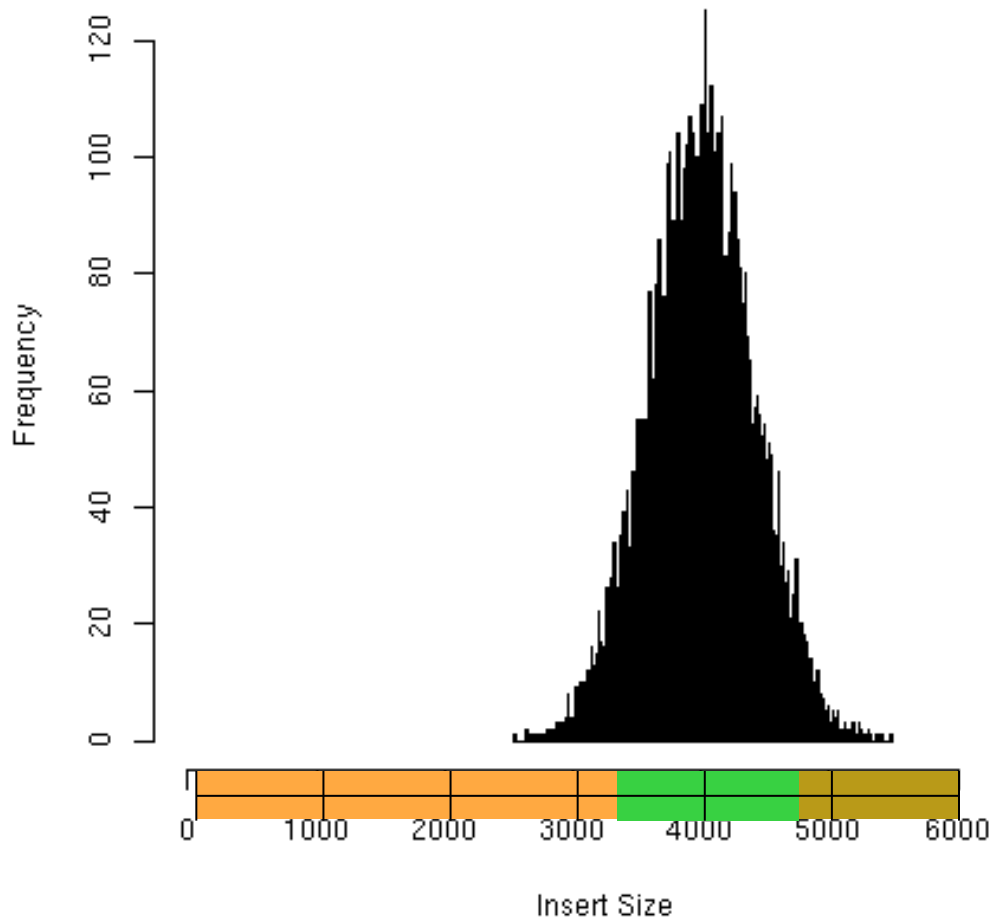
$$\text{C/E Stat: } \frac{(4048 - 4000)}{(400 / \sqrt{8})} = +0.33$$

Near 0 indicates overall happiness

Hidden Compression

Library size distribution

Mean: 4000, SD: 400



8 inserts: 3.2 kb-4.8kb

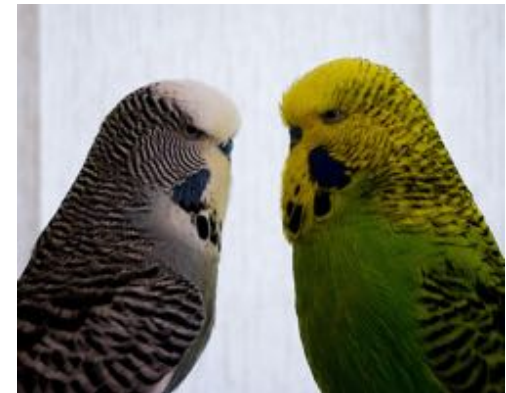
Local Mean: 3488

$$\text{C/E Stat: } \frac{(3488 - 4000)}{(400 / \sqrt{8})} = -3.62$$

C/E Stat \leq -3.0 indicates Compression

Assemblathon 2

- Real sequence data, *de novo* assembly



- Step 1: Apply best practices from Assemblathon 1
- Step 2: Add secret weapon for winning...

Images from Assemblathon

Fish Metassembly

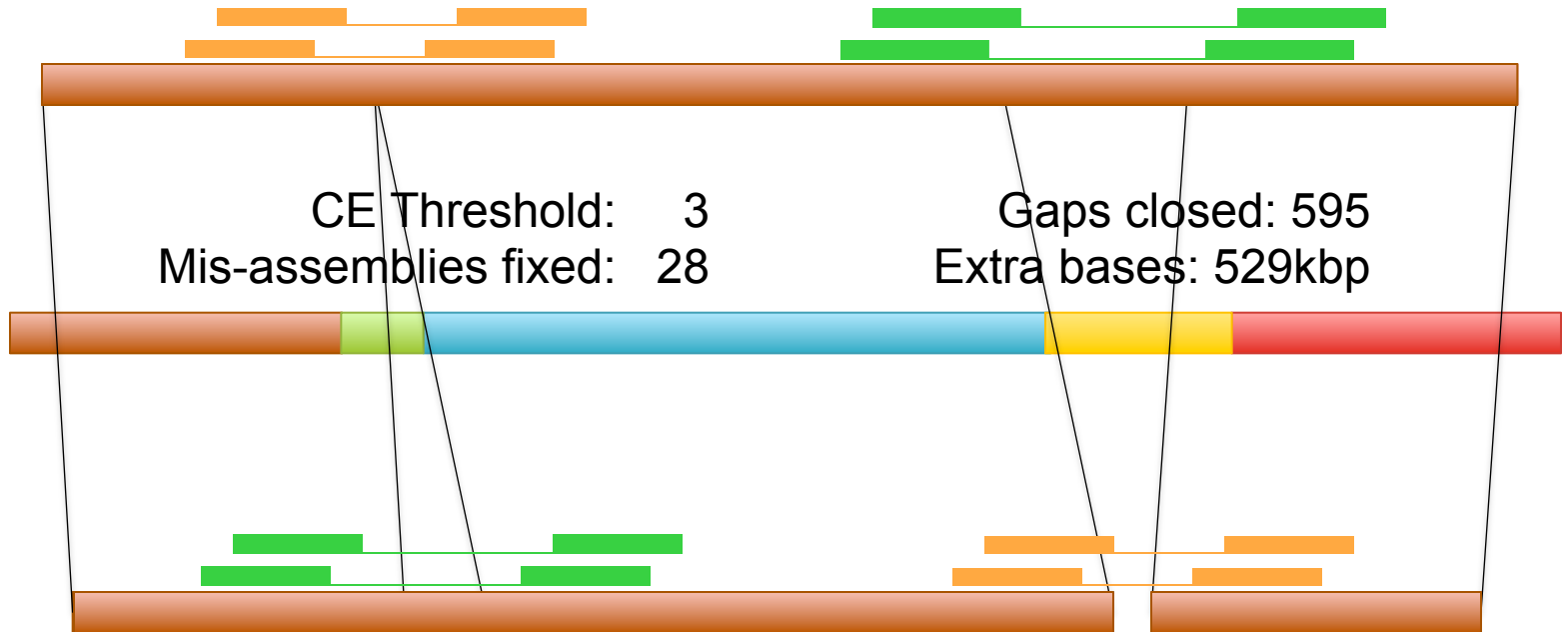
<http://metassembler.sf.net>



ALLPATHS-LG

Scaffold N50: 3,710,017
#>1000: 2,791

Contig N50: 20,183
#>1000: 68,591



SOAPdenovo
+ FLASH
+ Quake
+ AMOS

Scaffold N50: 285,413
#>1000: 29,119

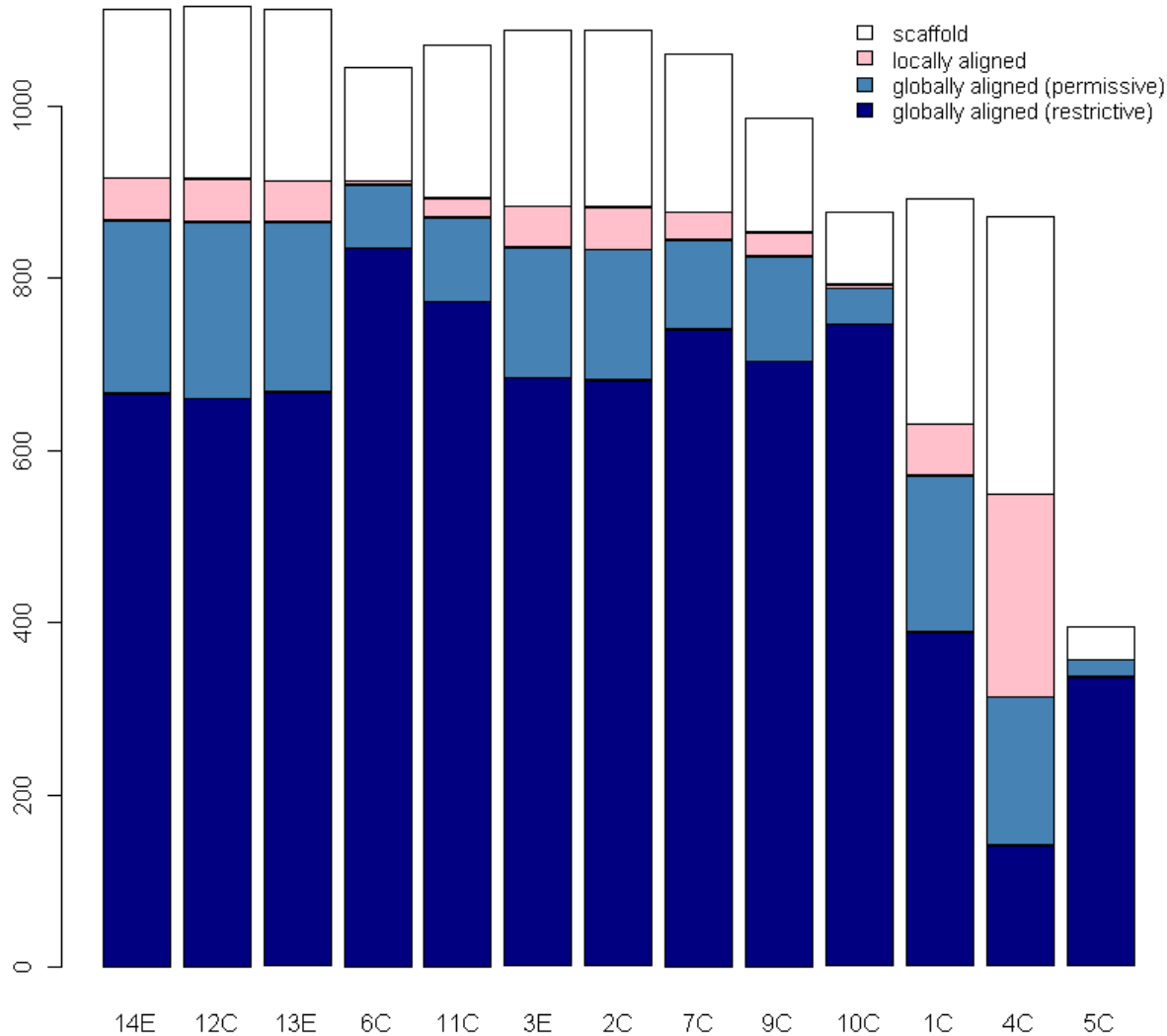
Contig N50: 1,607
#>1000: 218,643

Parrot Metassembly

<http://metassembler.sf.net>



Bird Scaffold Alignments to Optical Map



- Crowd-source individual assemblies
 - 13 submissions (including variants of same basic assembly)
- Use optical maps to evaluate long range consistency as the gold standard

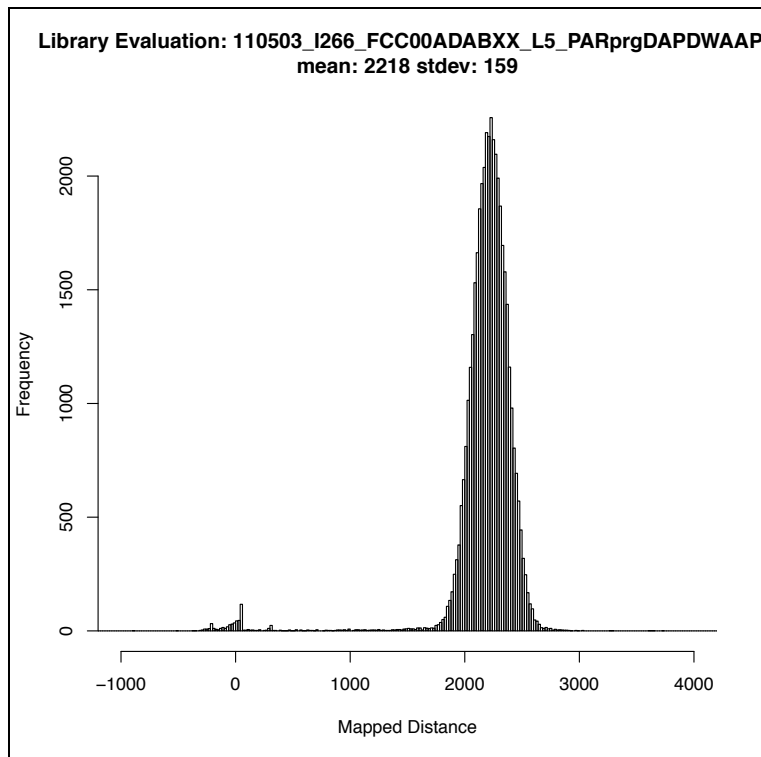
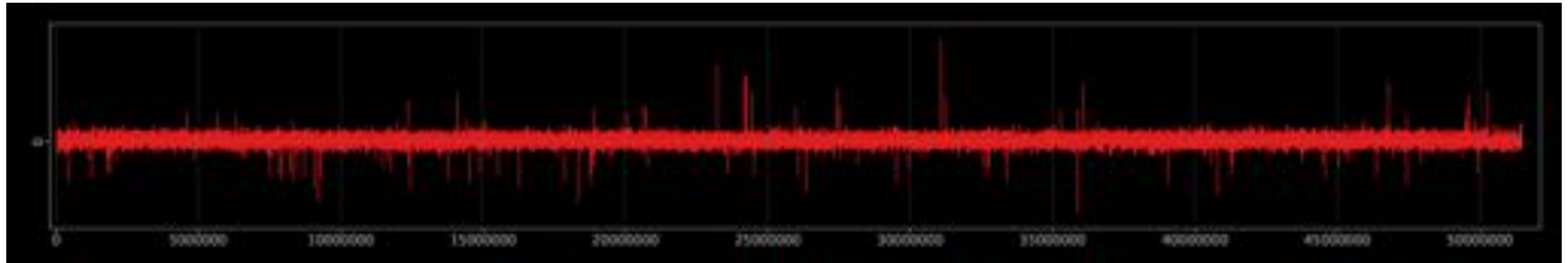
Fig. from Steve Goldstein

Parrot Metassembly



CE statistic (projected) across 51.1 Mbp scaffold

6C



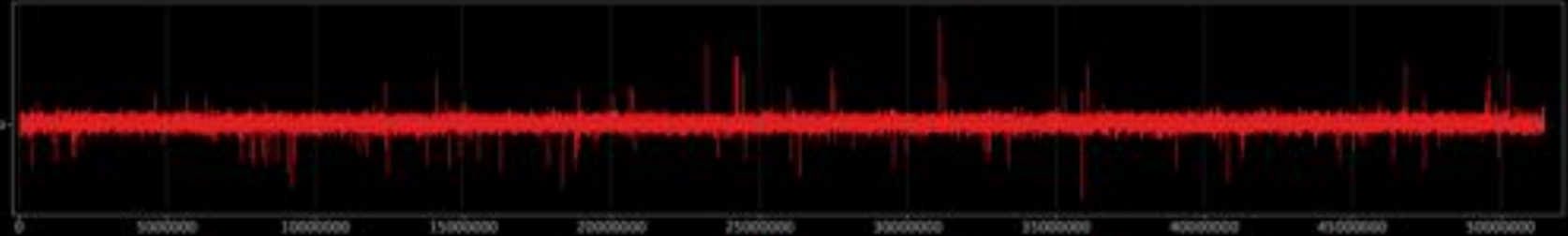
- Re-map 2kbp mates to each draft assembly, compute CE statistic at every position
- Extreme CE values are likely to be mis-assemblies
 - Can also look at coverage, mis-oriented mates, and other forensics features
 - Approximately 1.4 major events per Mbp

Parrot Metassembly

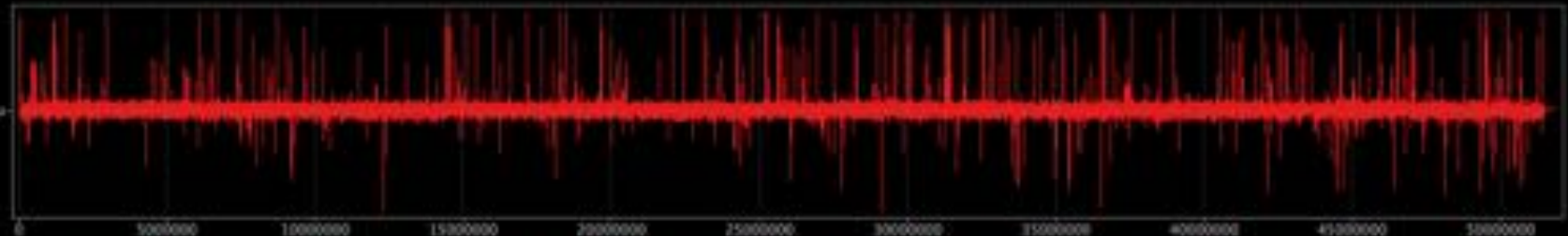
CE statistic (projected) across 51.1 Mbp scaffold



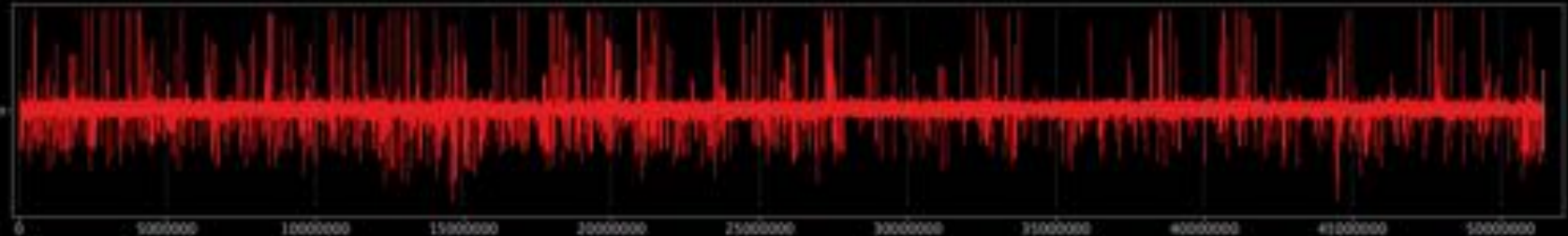
6C



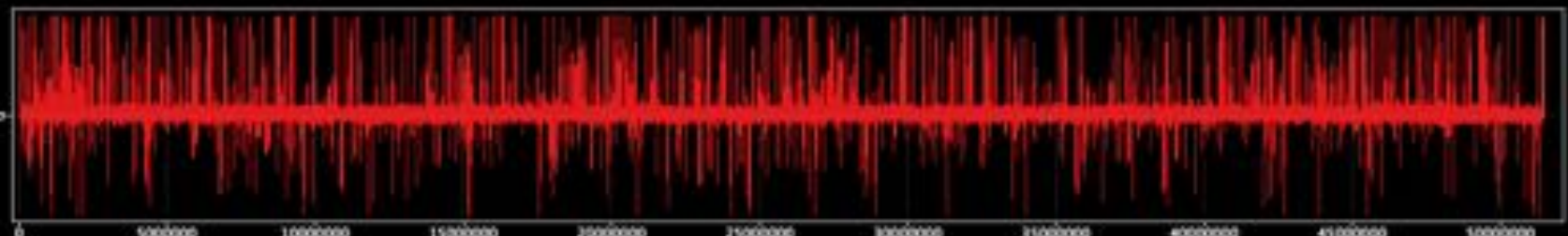
2C



13E



7C

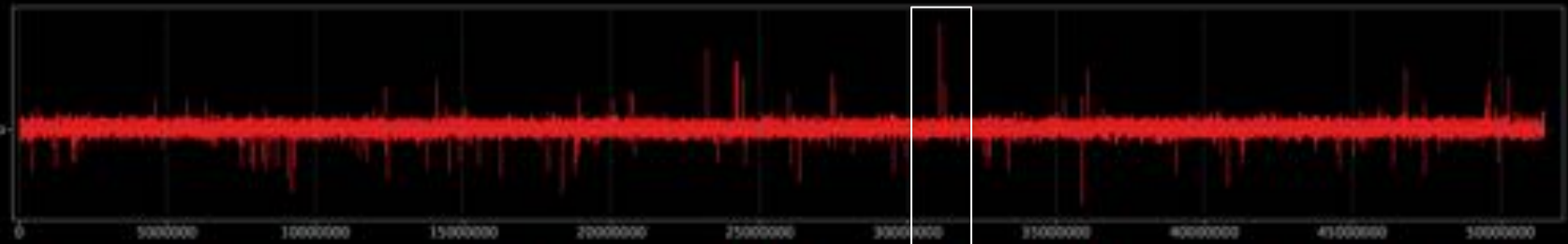


Parrot Metassembly

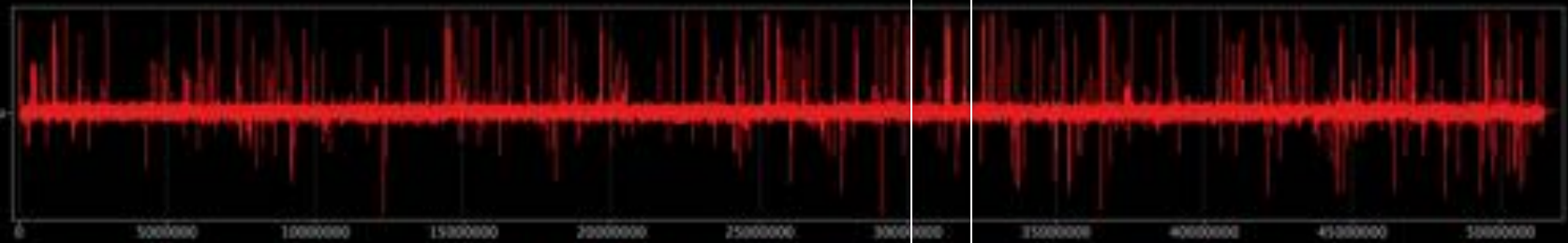


CE statistic (projected) across 51.1 Mbp scaffold

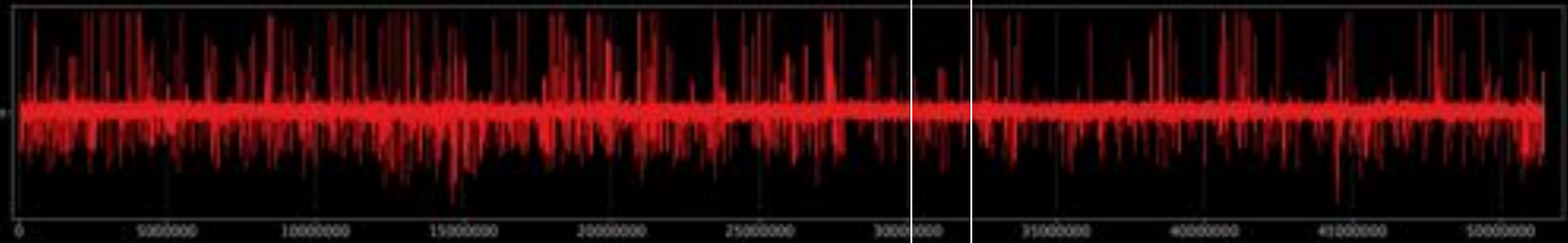
6C



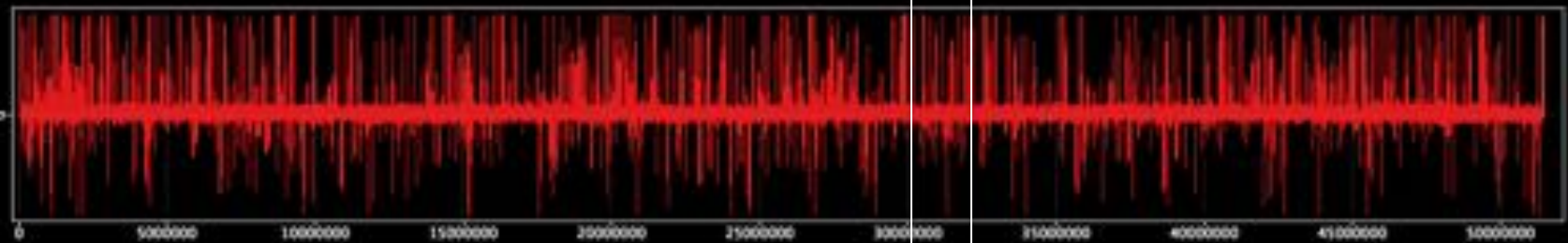
2C



13E



7C

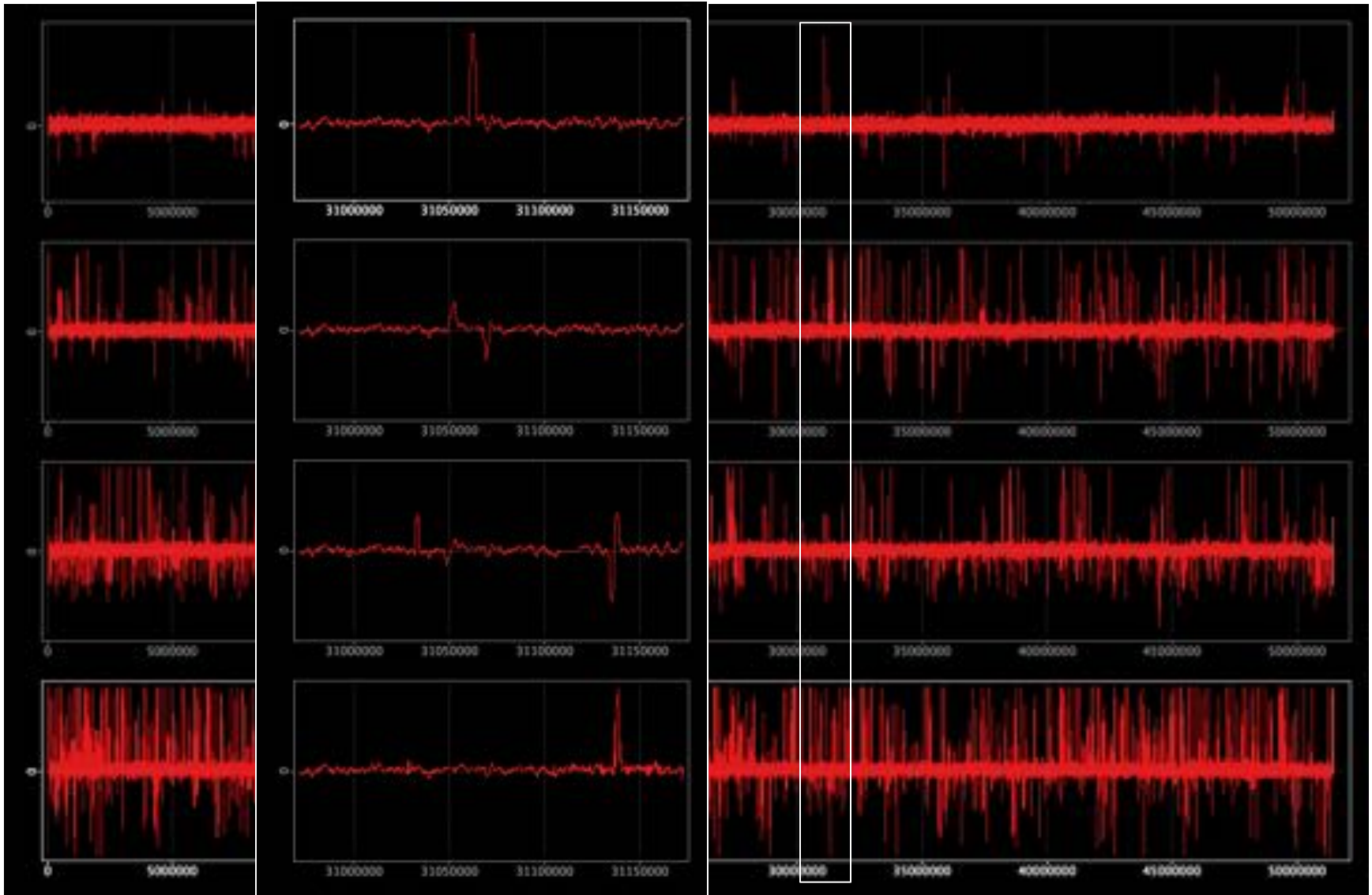


Parrot Metassembly

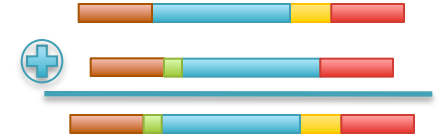
CE statistic (projected) across 51.1 Mbp scaffold



6C



Summary



- Metassembly can correct nearly every mis-assembly and small gap in the parrot genome
 - Sliding window to select best representation along the 6C backbone
- Metassembly draws on individual strengths of each submission to locally optimize the problem
 - Different sequencing technologies
 - Different algorithms
 - Different parameters
- Summary/Consensus methods extremely powerful in virtually every complex optimization computation

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David Kelley

Cole Trapnell

Duke

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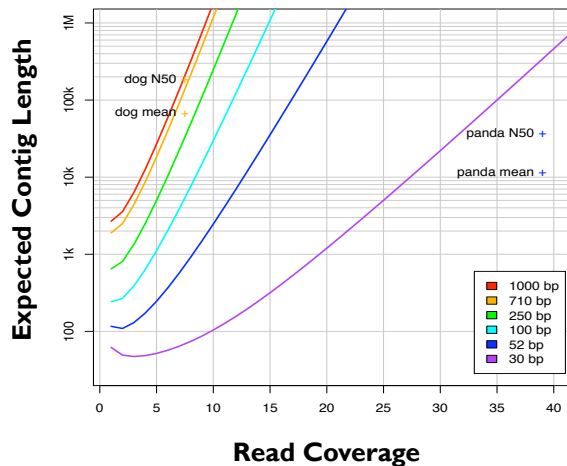
Plus all the Assemblathon Members

Thank You

<http://schatzlab.cshl.edu>
@mike_schatz / #AGBT

Ingredients for a good assembly

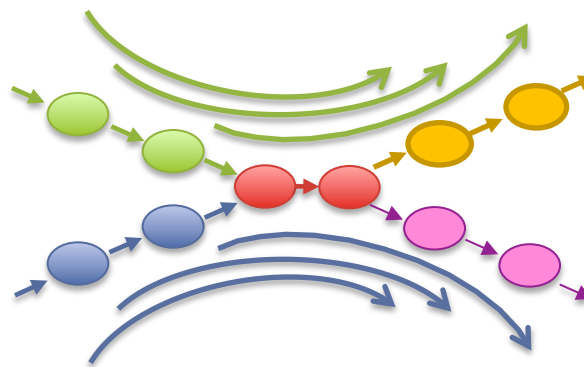
Coverage



High coverage is required

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

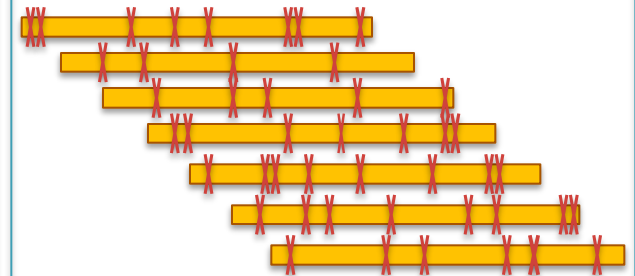
Read Length



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

Assembly of Large Genomes using Second Generation Sequencing

Schatz MC, Delcher AL, Salzberg SL (2010) *Genome Research*. 20:1165-1173.