

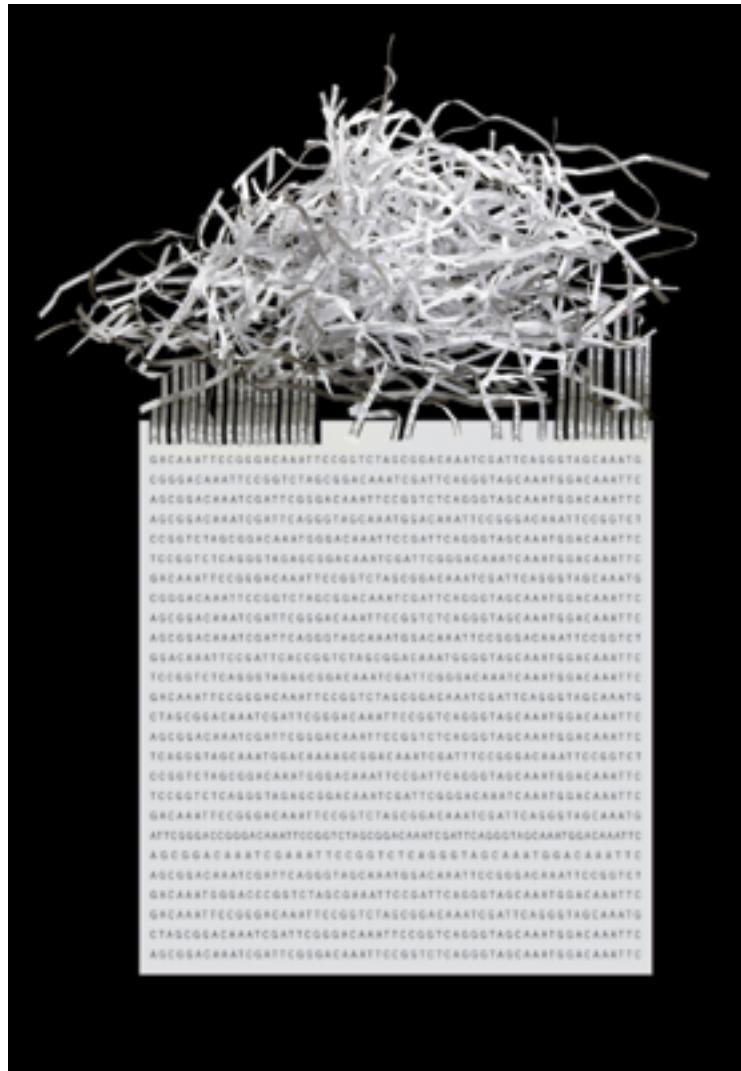
# Introduction to single-cell genome assembly

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# Outline: introduction

- Assembly basics
- Assembly metrics
- Single-cell data specific problems
- Available assemblers
- How SPAdes works
- Sample
- Today's exercise



*De novo* genome assembly: what every biologist should know **Monya Baker**  
*Nature Methods* 9, 333–337 (2012) doi:10.1038/nmeth.1935

# Assembly puzzle



# Assembly puzzle



1. Fragment DNA and sequence

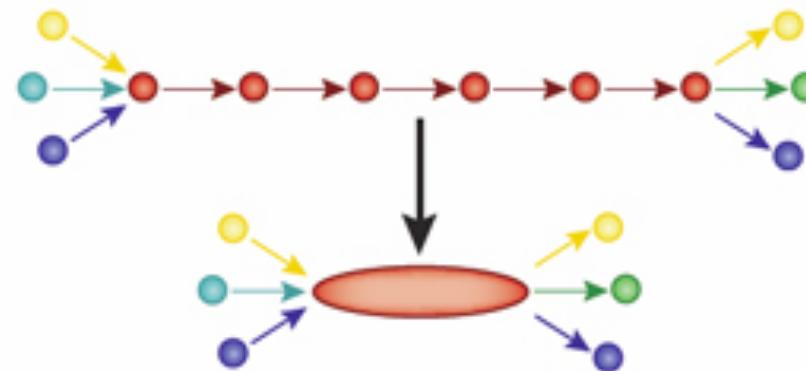


2. Find overlaps between reads

...AGCCTAGACCTACAGGATGCGCGACACGT  
GGATGCGCGACACGT CGCATATCCGGT...

3. Assemble overlaps into contigs

Contigs =  
continuous  
sequence

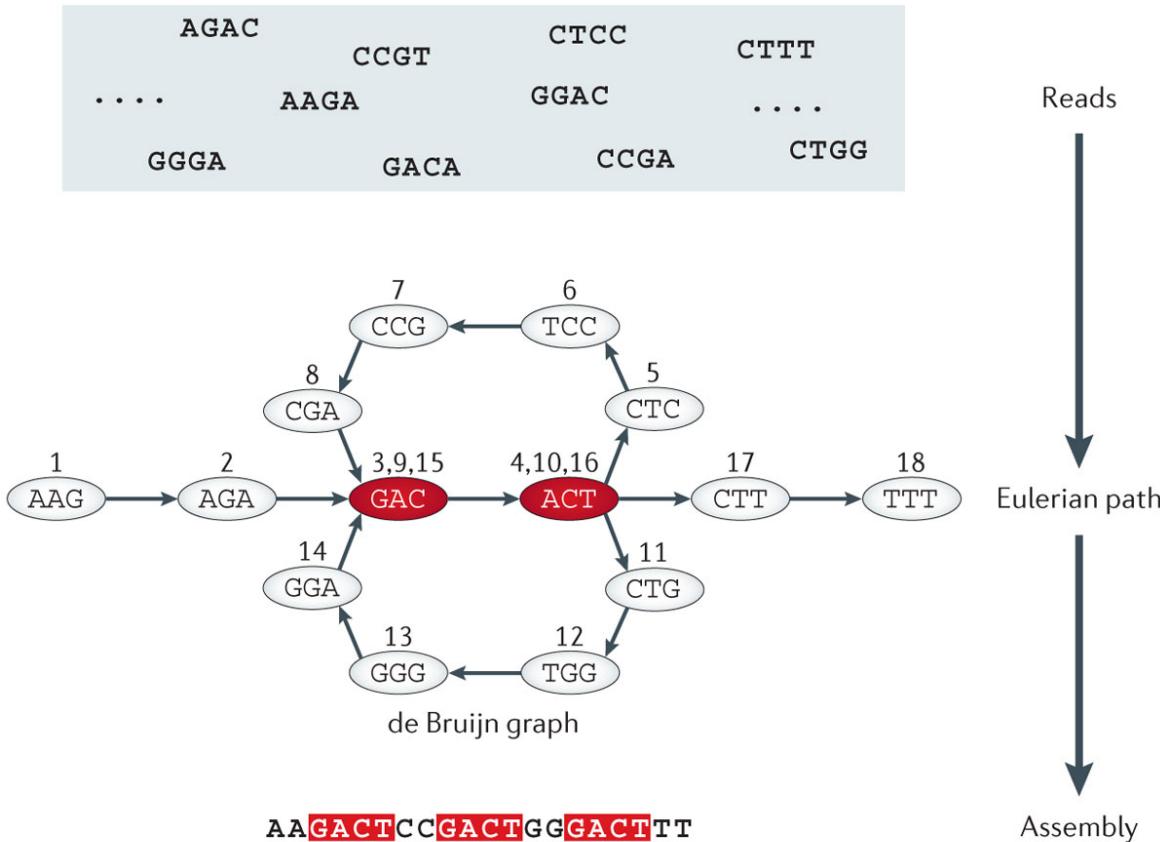


Scaffolds =  
ordered contigs  
with gaps

4. Assemble contigs into scaffolds



# de Bruijn graph assembly



Nature Reviews | Genetics

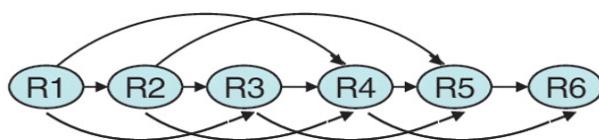
Nat Rev Genet. 2013 May;14(5):333-46. doi: 10.1038/nrg3433.  
Computational solutions for omics data. Berger B1, Peng J, Singh M.

# Overlap vs kmer graphs

**A**

ATATAT[ACTGGCGTATCGCAGTAAAC]GCGCCG

R1: ACTGGCGTAT  
R2: TGGCGTATCG  
R3: GGC GTATCGC  
R4: CGTATCGCAG  
R5: TATCGCAGTA  
R6: CGCAGTAAAC



**B**

ATATAT[ACTGGCGTATCGCAGTAAAC]GCGCCG

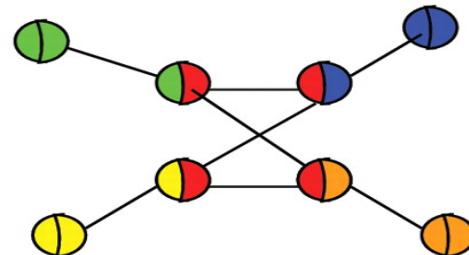
K1: ACTGG  
K2: CTGGC  
K3: TGGCG  
K..: .....  
K14:  
K15:  
K16: AGTAA  
GTAAA  
TAAAC



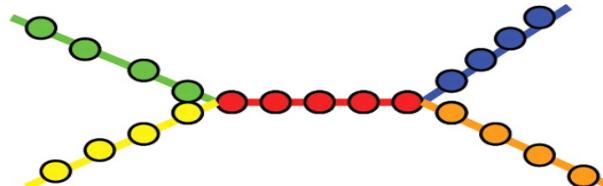
**A**



**B**

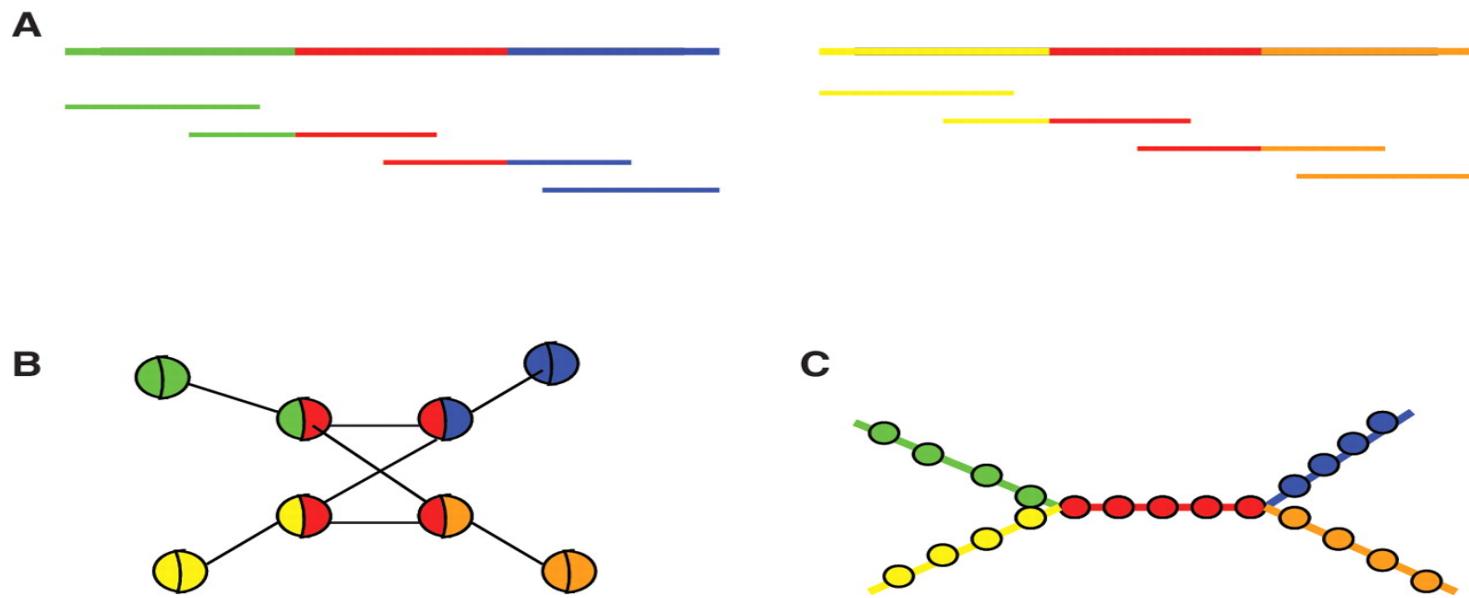


**C**



# Assembly difficulties

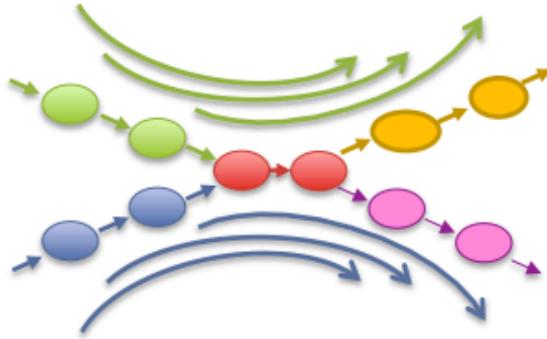
REPEATS



Slide courtesy of Francesco Vezzi, SciLife Lab

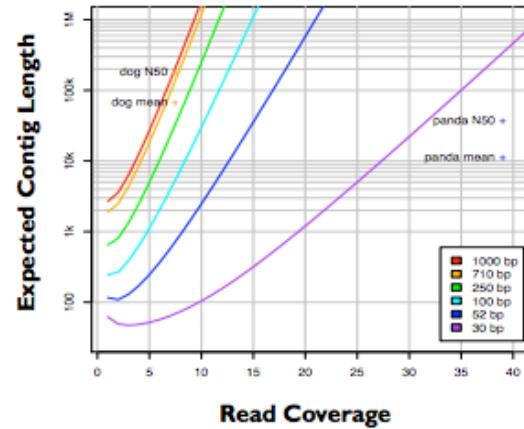
# Ingredients for a good assembly

## Read Length



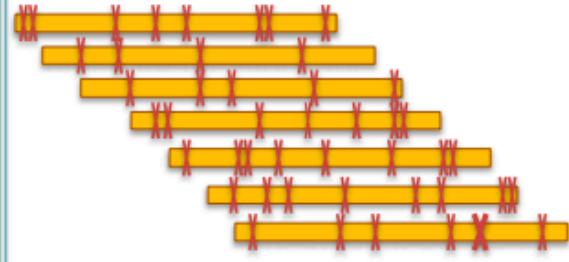
**Reads & mates must be longer than the repeats**

## Coverage

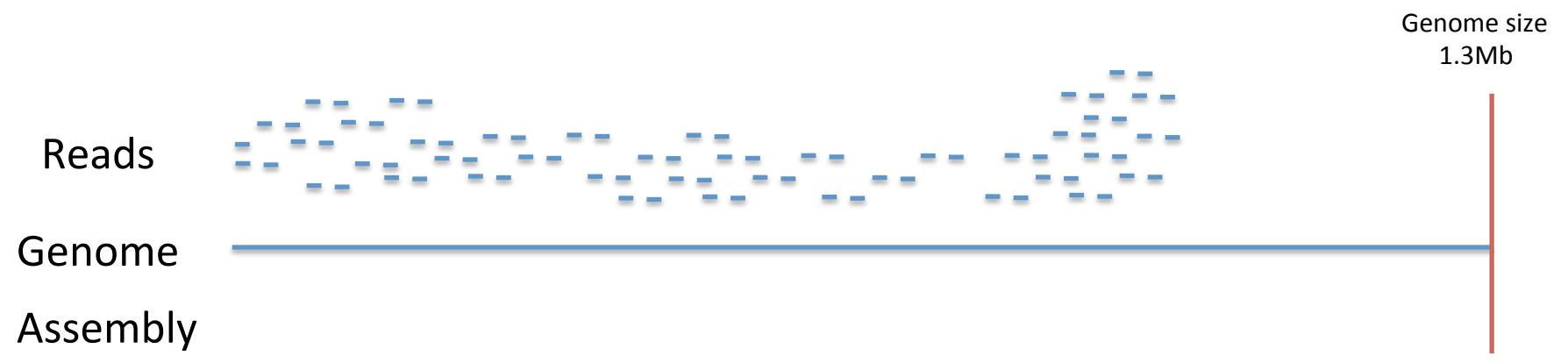


**High coverage is required**

## Quality

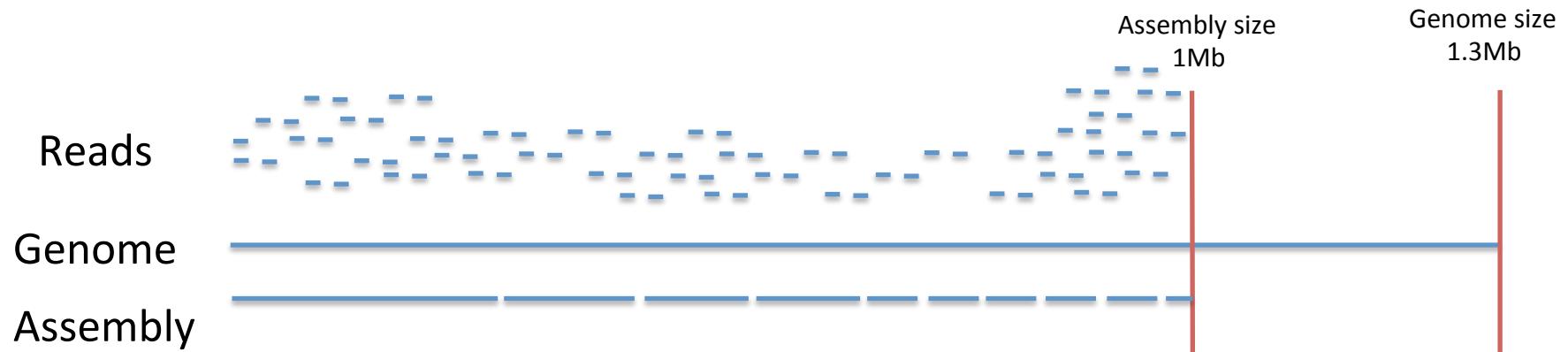


**Errors obscure overlaps**



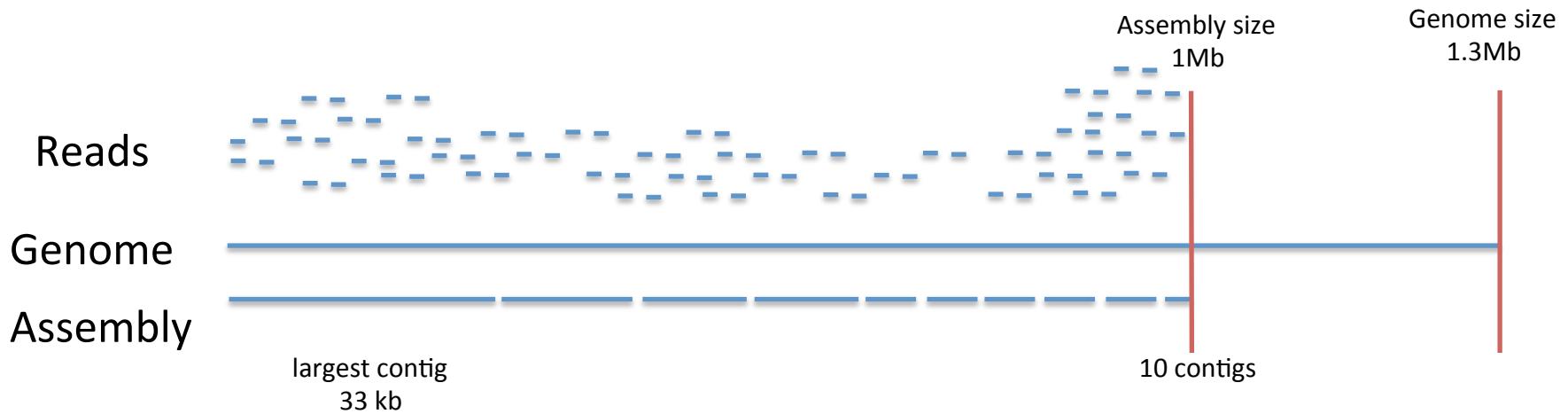
# Assembly metrics

- assembly size
- number of contigs, largest contig
- N50



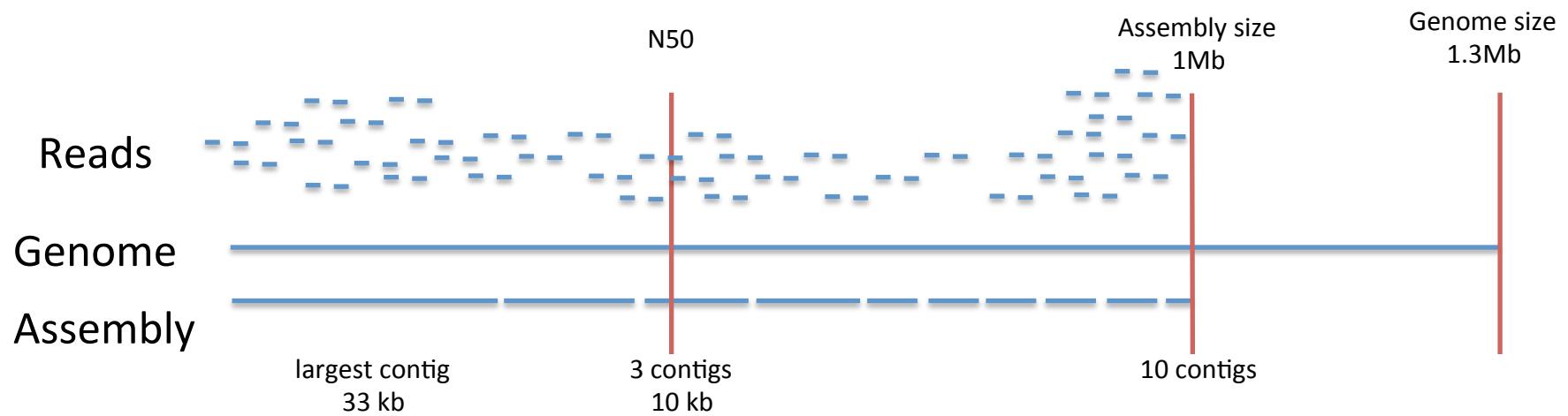
# Assembly metrics

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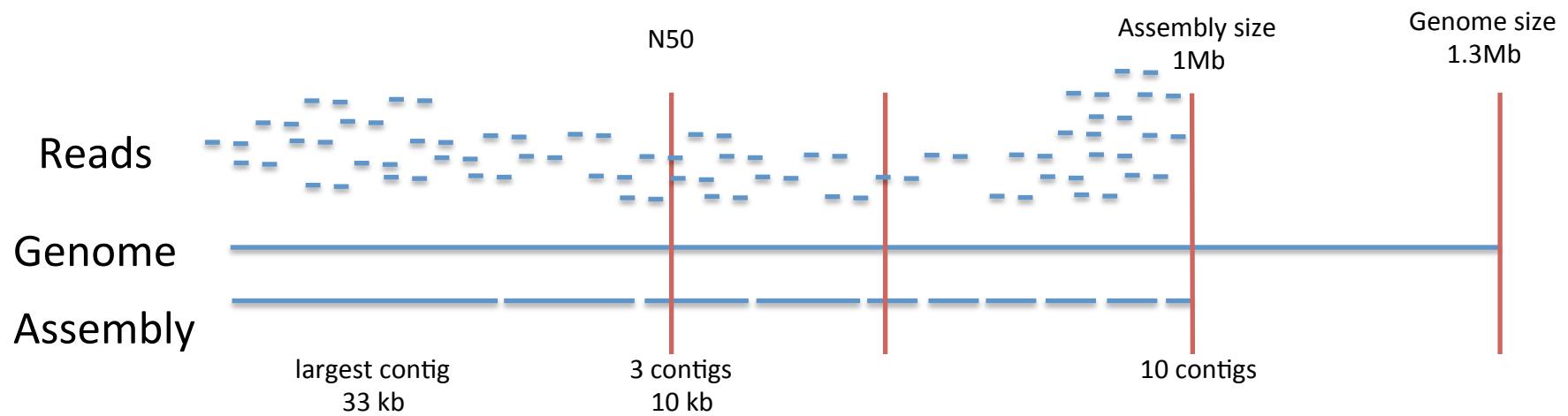
# Assembly metrics

- assembly size
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# Assembly metrics

- assembly size
- number of contigs, largest contig
- N50



# Outline: single cell assemblies

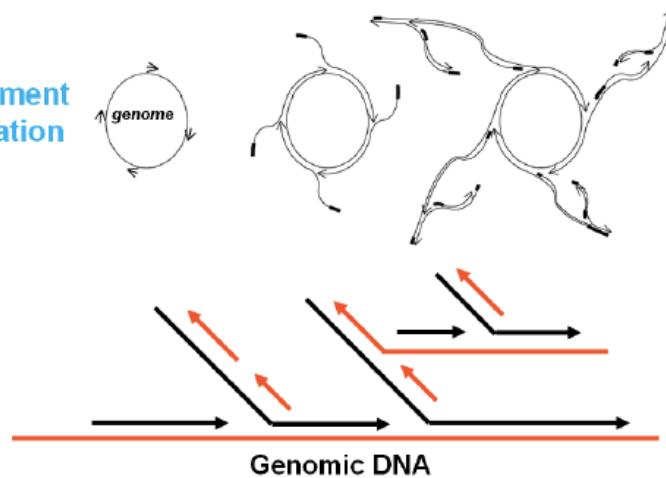
- Assembly basics
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- Single-cell data specific problems
- Available assemblers
- How SPAdes works
- Sample
- Today's exercise

# Problems with single-cell data

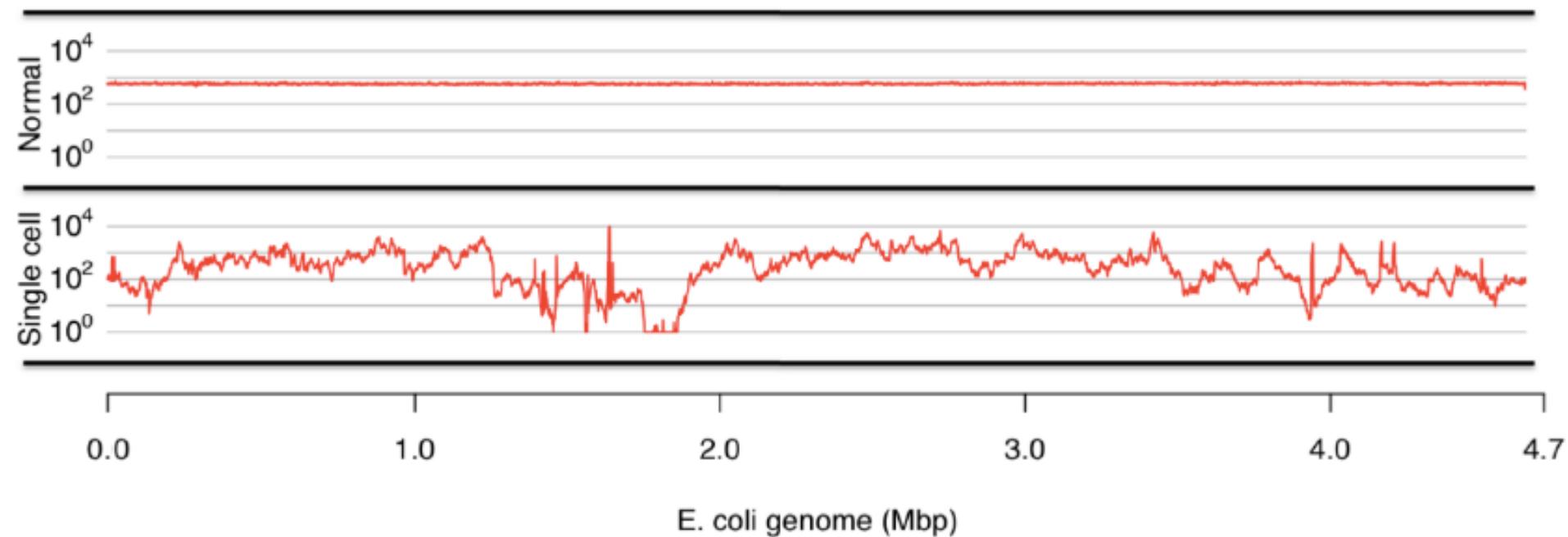
## MDA artefacts

- Chimeras
- Uneven coverage

Multiple Displacement Amplification (MDA)



Coverage



# How does this affect assembly?

- de Bruijn graph sensitive to k-mer quality
- Bad quality k-mers from low-coverage regions
  - Erroneous graph connections → misassemblies
  - Or gaps due to removal of low-coverage areas
- Specialized single-cell genome assemblers are needed

# Single-cell genome assemblers available currently

- E+V-SC (Euler+Velvet-SC) (2011)
  - Euler and Velvet modification
  - Not for pairs
  - single k-mer
- IDBA-UD (2012)
  - Error correction
  - Multiple k-mers
  - paired-end reads
- SPAdes (2012)
  - Error correction
  - Multiple k-mers
  - paired-end reads
  - Also tries to solve chimera problems

# Why use SPAdes?

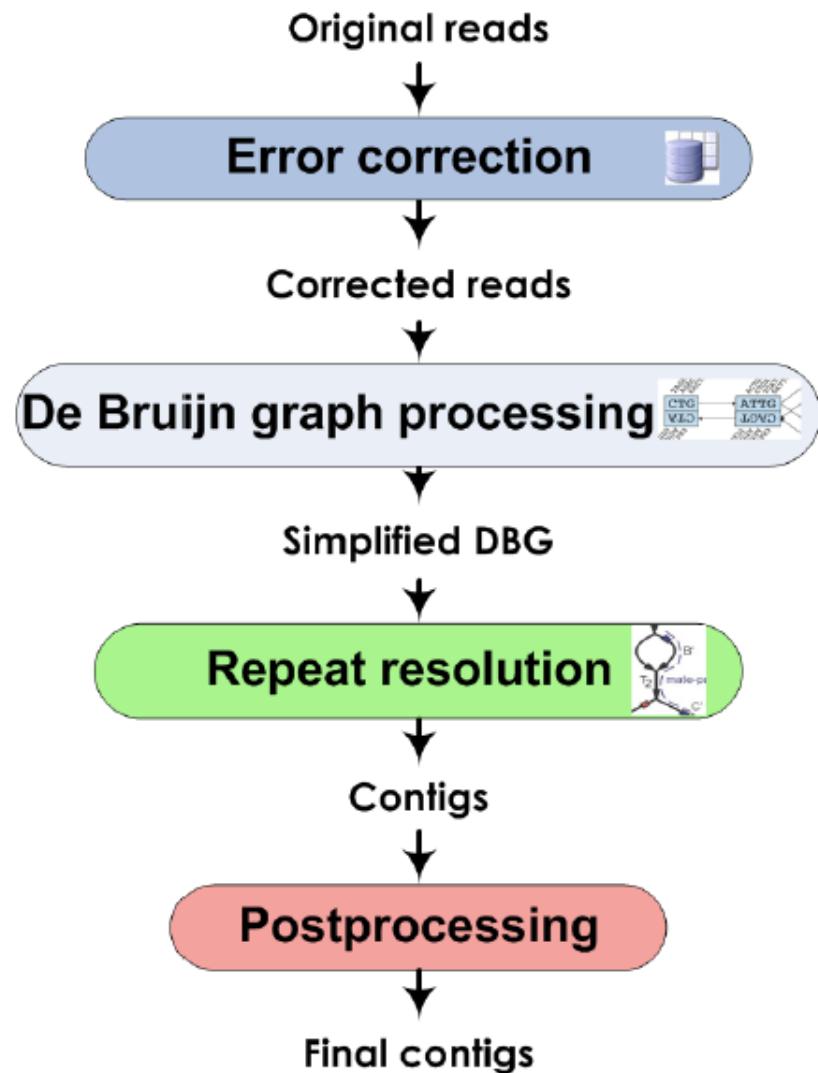
## (better assembly results)

Assembly	NG50	# of contigs	Largest contig	Total length	Misassemled contigs	mismatch (bp per 100kbp)	indels (bp per 100kbp)	Mapped genome (%)	# genes
A5	14399	745	101584	4441145	8	12.01	0.17	89.88	3444
ABySS	68534	179	178720	4345617	6	3.32	1.68	88.268	3704
CLC	32506	503	113285	4656964	2	5.53	1.42	92.291	3768
EULER-SR	26662	429	140518	4248713	17	10.87	35.67	84.898	3416
Ray	45448	361	210820	4379139	17	6.29	2.83	88.372	3636
SOAPdenovo	1540	1166	51517	2958144	1	1.87	0.11	57.672	1766
Velvet	22648	261	132865	3501984	2	2.19	1.23	73.765	3080
<b>E+V-SC</b>	<b>32051</b>	<b>344</b>	<b>132865</b>	<b>4540286</b>	<b>2</b>	<b>2.33</b>	<b>0.73</b>	<b>91.744</b>	<b>3771</b>
IDBA-UD contigs	98306	244	284464	4814043	8	5.09	0.27	95.21	4045
IDBA-UD scaffolds	109057	229	284464	4813609	8	5.14	0.77	95.199	4052
<b>SPAdes3.1 contigs</b>	<b>109059</b>	<b>238</b>	<b>268493</b>	<b>4797090</b>	<b>1</b>	<b>3.29</b>	<b>0.45</b>	<b>94.936</b>	<b>4036</b>
<b>SPAdes1.1 scaffolds</b>	<b>110081</b>	<b>233</b>	<b>268493</b>	<b>4799481</b>	<b>1</b>	<b>4.02</b>	<b>0.64</b>	<b>94.959</b>	<b>4041</b>

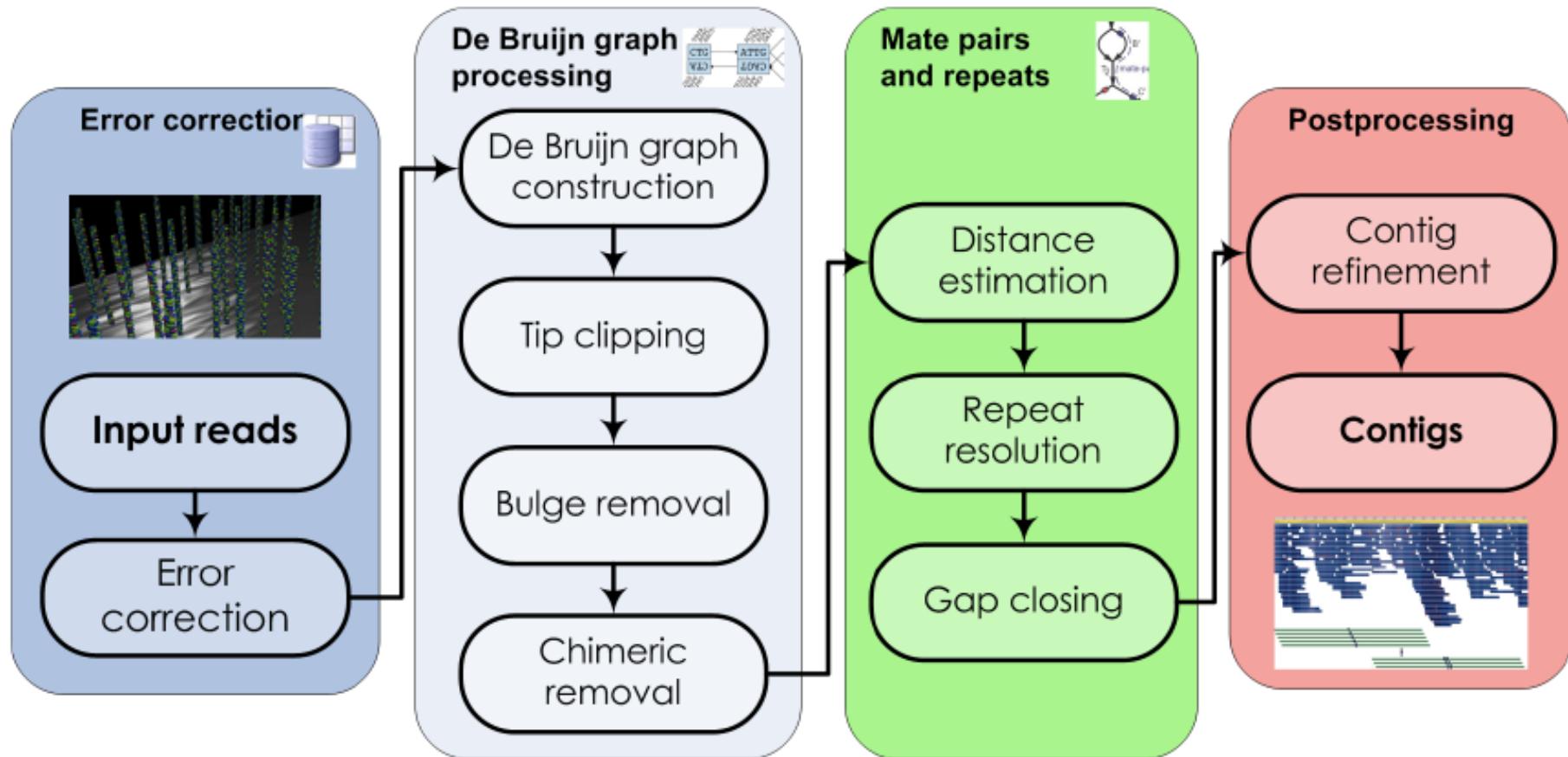
Using *E. coli* single-cell

# How does SPAdes achieve this?

- Error correction of reads before assembly
  - *Uses novel algorithm: BayesHammer*
  - *This reduces erroneous k-mers that could mess up assembly*
- Use of multiple k-mers to construct assembly graph
  - *Improved resolution of assembly graphs*
- Uses mate pairs to improve de Bruijn graph construction
  - *Paired de Bruijn graphs (“Rectangle Graphs”)*
    - *helps to resolve repeats*
    - *Helps with contig scaffolding*
- Removal of chimeric connections in graph
  - *Less mis-assemblies in the contigs*
- Final correction of errors in contigs (using bwa)
  - *Improved contig quality*
- All these steps in a single command
  - *Other tools need multiple tools to do same procedures*



# More details of each step

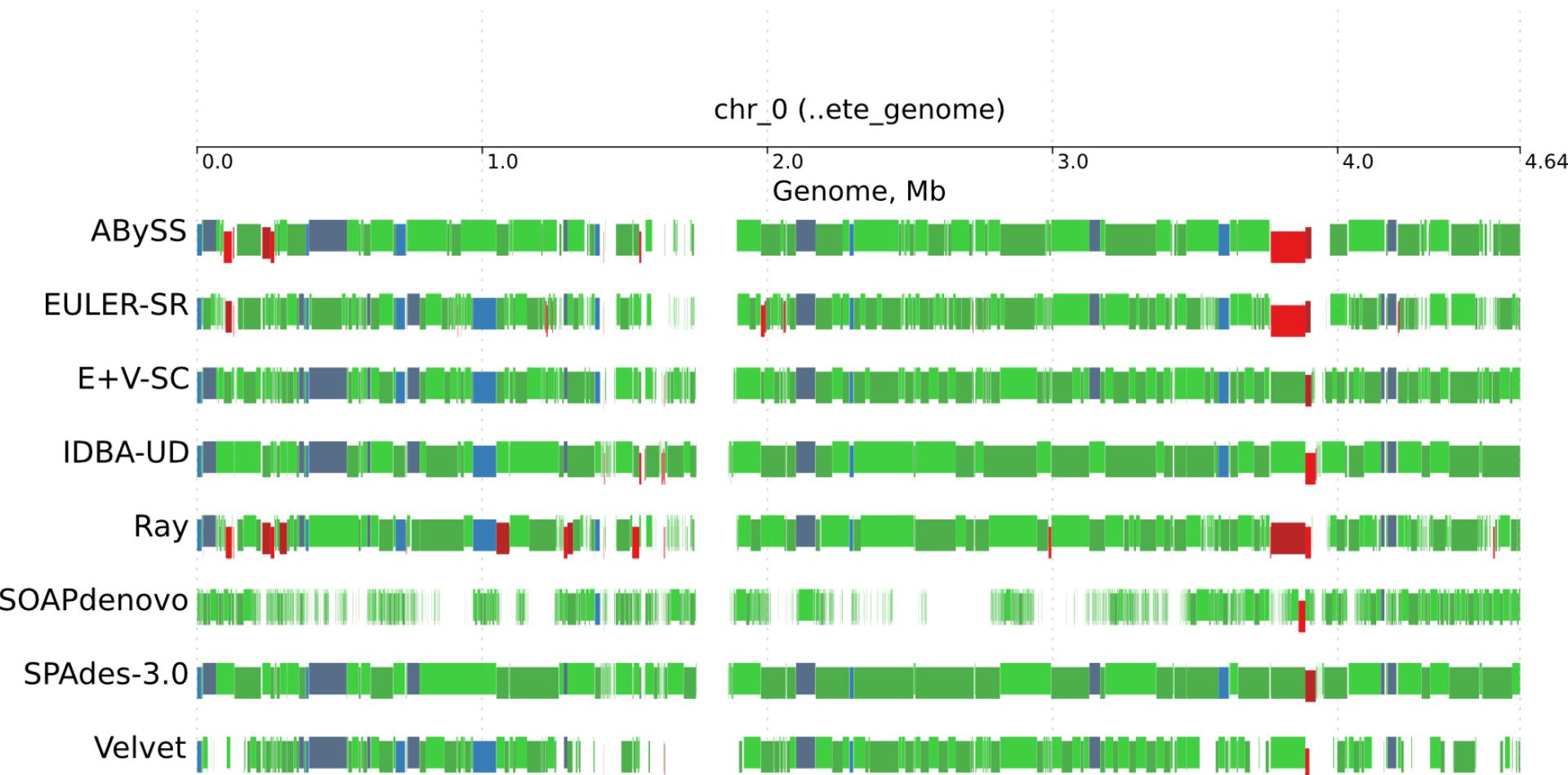


# A few things to consider when using SPAdes

- SPAdes currently only works on Illumina data
  - Other NGS data won't work
- HiSeq data
  - 100-150 bp paired end reads
    - Shorter k-mers
    - Faster assembly
- MiSeq data
  - 250-300 bp paired end reads (longer)
    - Larger k-mers
      - assembly takes longer if smaller k-mers are used
    - User may need to optimize k-mer selection to produce optimal assembly
- In general, it works better with short, high quality reads
- Can also be used for multi-cell genomic data

# Why use SPAdes?

## (better genome coverage)



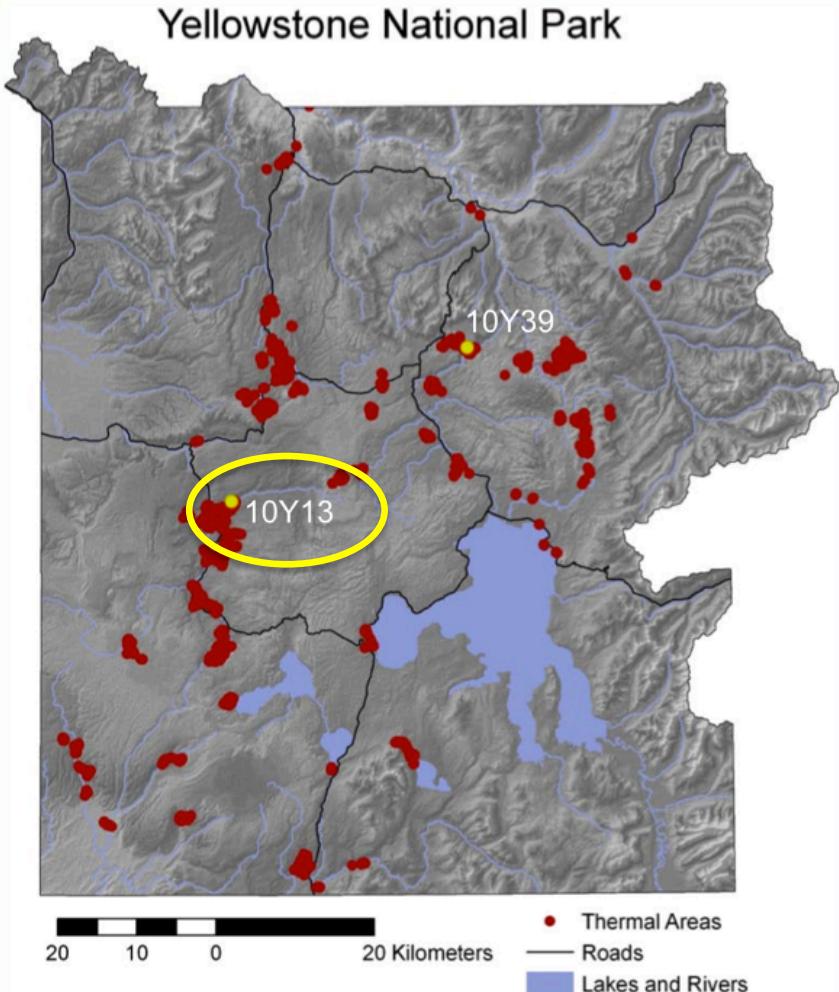
# Acknowledgements

- Jimmy Saw (single cell analysis)
- Anders Lind (Coverage/chimera checks)
- Joran Martijn (MEGAN analysis)
- Lionel Guy (Genome completeness estimates)

# Outline: practical part

- Assembly basics
- Assembly metrics
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# Sample



Culex Basin  
pH 8.6, T=68.8°C

Images on courtesy of Cristina Takacs-Vesbach and Dan Coleman

# Datasets to be used

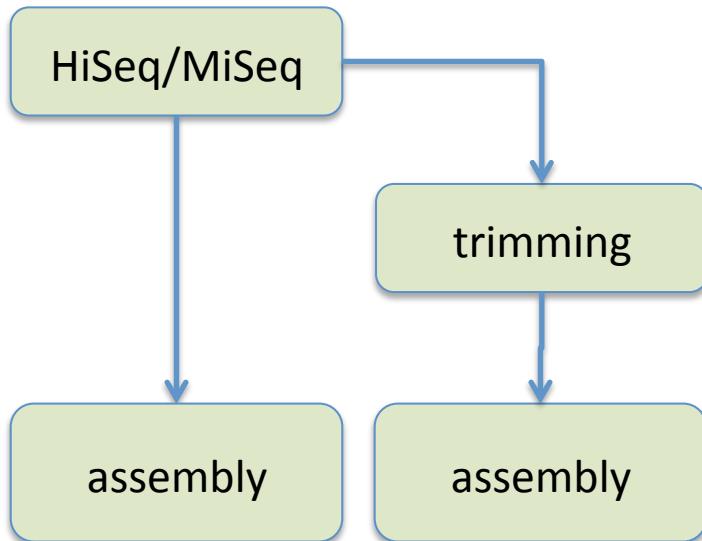
From same SAG

- **Dataset1**
  - Paired end **HiSeq** data for **G5**
  - G5\_Hiseq\_R1\_001.fastq
  - G5\_Hiseq\_R2\_001.fastq
- **Dataset2**
  - Paired end **MiSeq** data for **G5**
  - G5\_Miseq\_R1\_001.fastq
  - G5\_Miseq\_R2\_001.fastq

**12 assemblies per group**

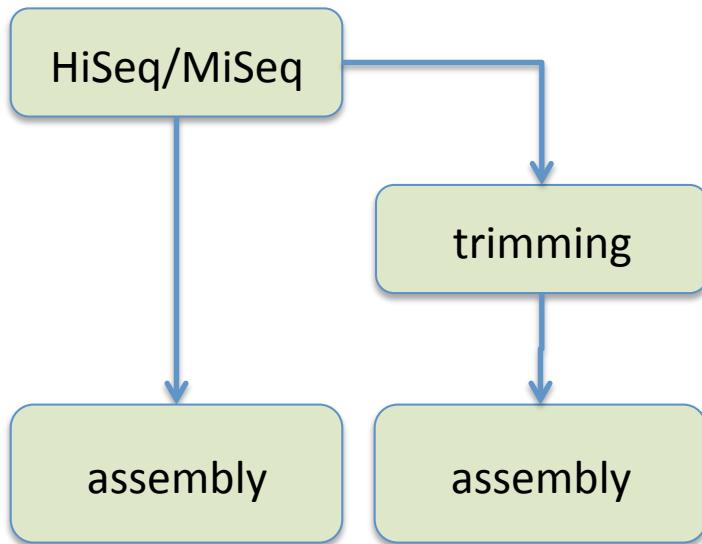
- **6 assemblies**
  - 3 assemblies with original data
  - 3 assemblies with trimmed data
- **6 assemblies**
  - 3 assemblies with original data
  - 3 assemblies with trimmed data

# Overview of exercises today



1. General instructions
2. Familiarizing with data (QC)
3. Single-cell genome assemblies using SPAdes (HiSeq data)

# Overview of exercises today



1. General instructions
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## 3 programs:

- Spades
- IDBA-UD
- Ray

# Exercise: compare assemblies

Table 1	HiSeq data (Original data)	Spades	IDBA-UD	Ray
Number of reads				
Assembly time				
Number of contigs				
Total assembly size				
Largest contig				
N50				
G+C%				
Number of ORFs				
Completeness (%)				

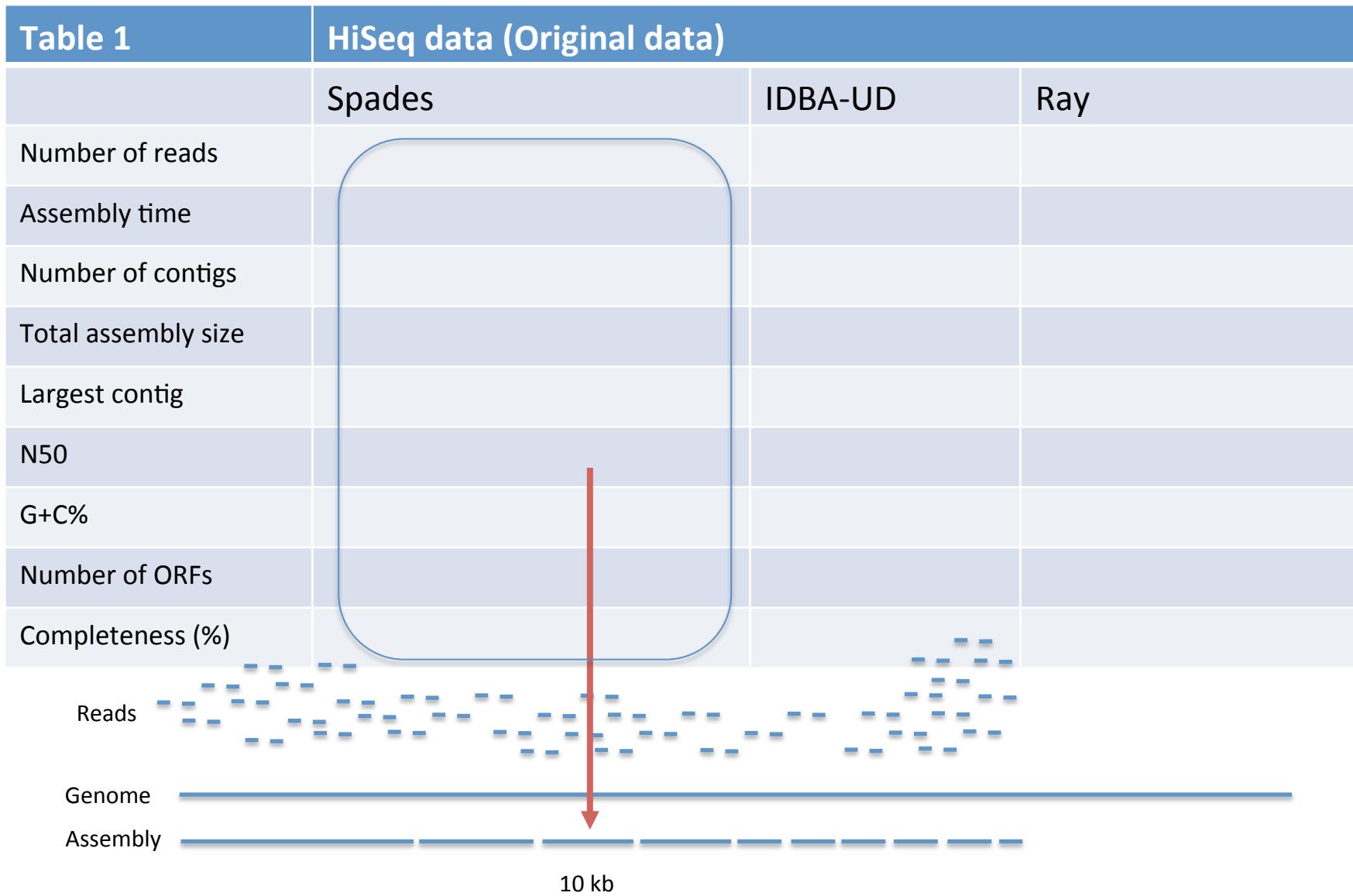
Reads

Genome

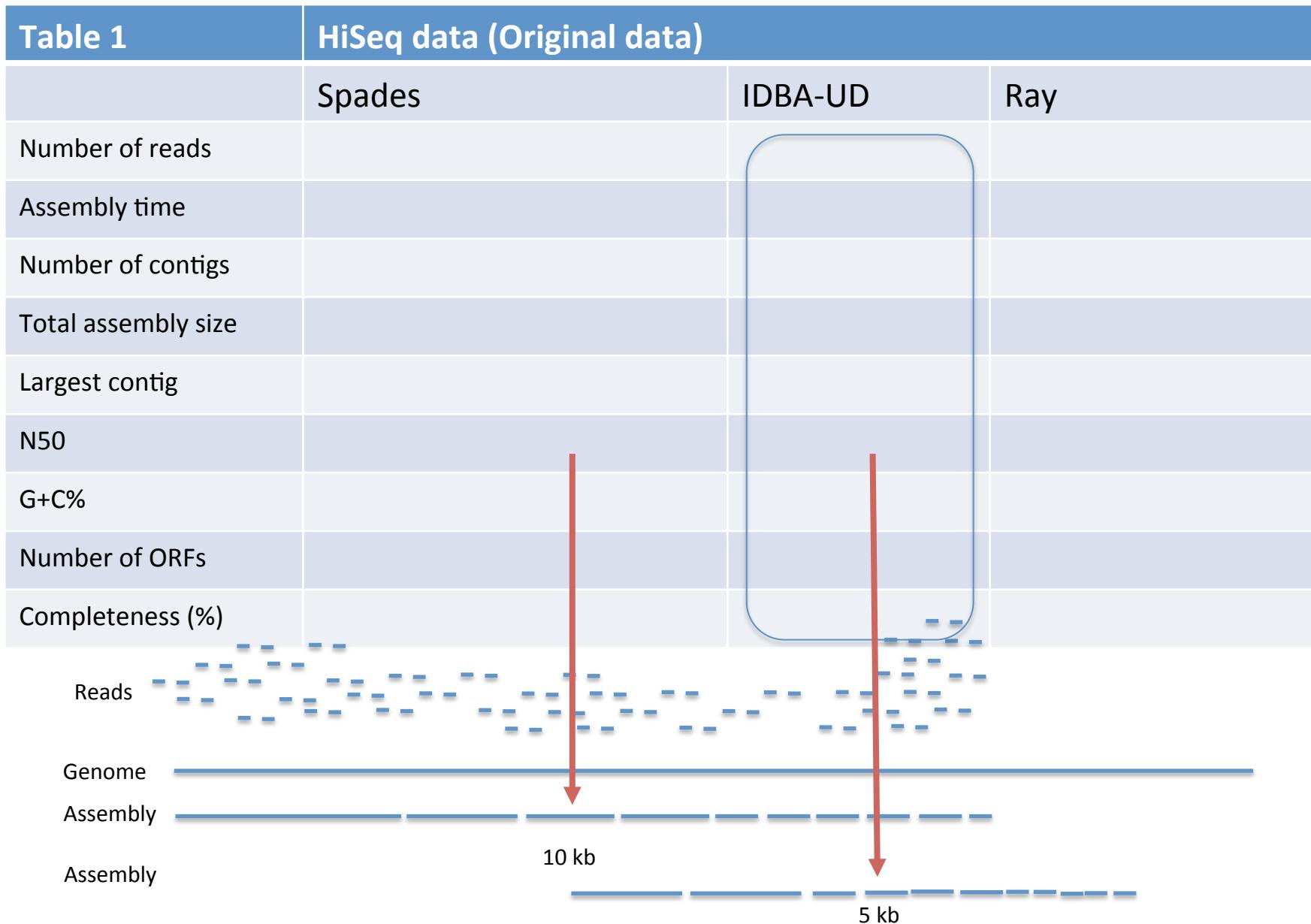
Assembly

— Read coverage — Genome — Assembly —

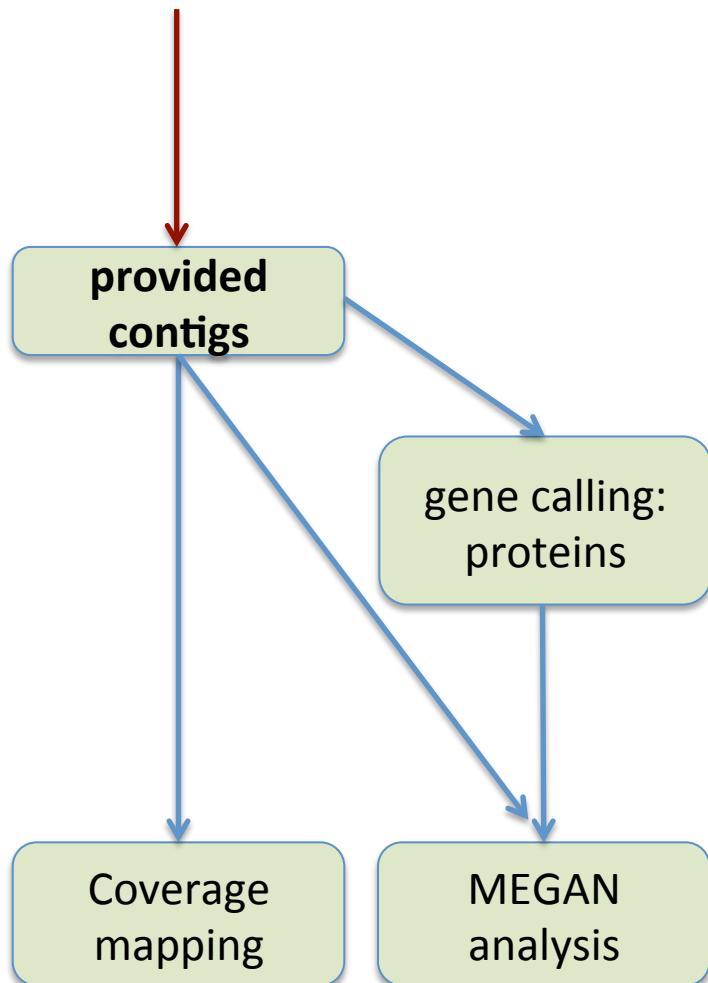
# Exercise: compare assemblies



# Exercise: compare assemblies



# Overview of exercises today



1. General instructions
2. Familiarizing with data (QC)
3. Single-cell genome assemblies using SPAdes (HiSeq data)

## PROVIDED CONTIGS

4. Assessing read coverage and chimera checking (with Artemis)
5. Checking for contaminants (with MEGAN)

# Datasets to be used

From same SAG

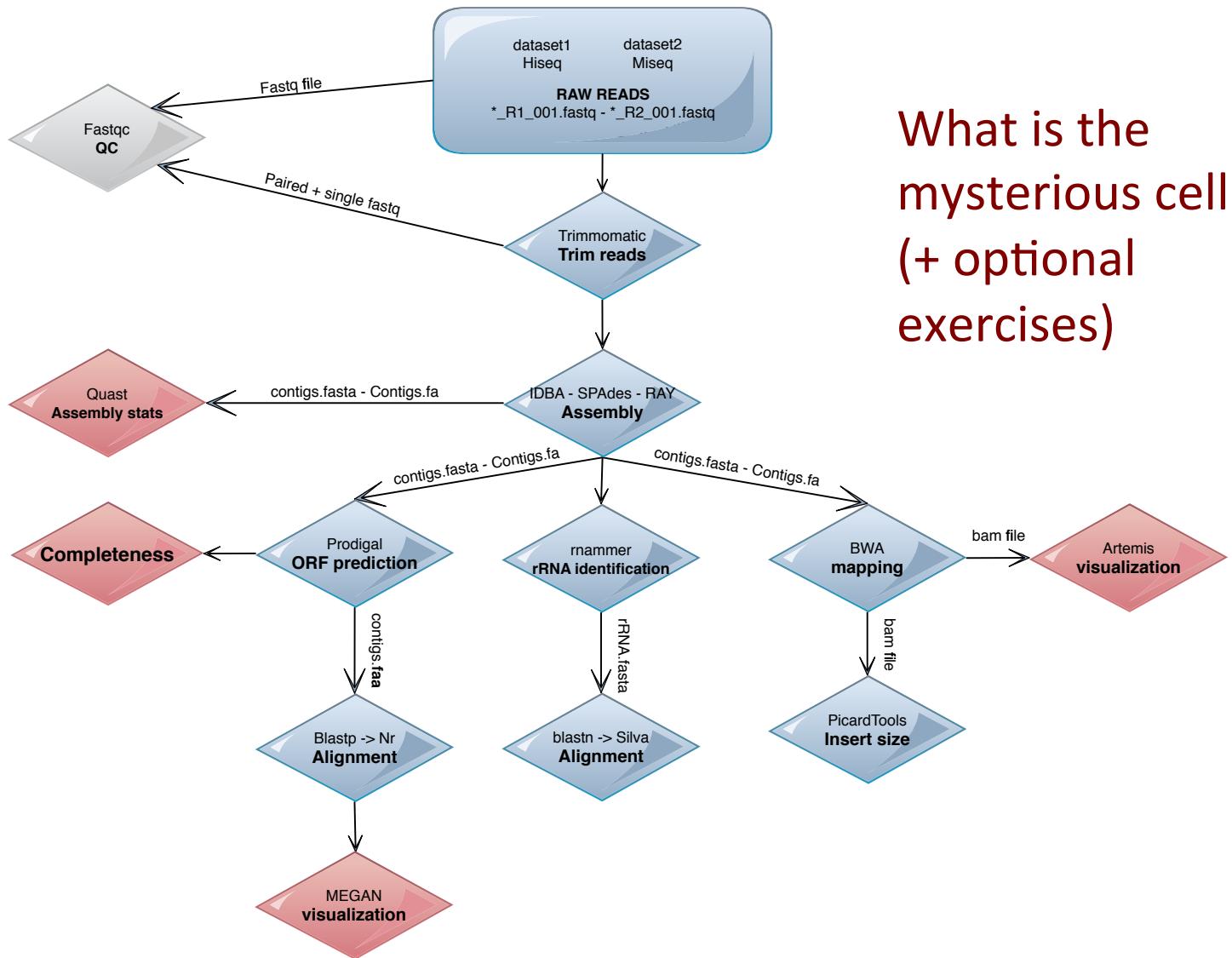
- **Dataset1**
  - Paired end **HiSeq** data for **G5**
  - G5\_Hiseq\_R1\_001.fastq
  - G5\_Hiseq\_R2\_001.fastq
- **Dataset2**
  - Paired end **MiSeq** data for **G5**
  - G5\_Miseq\_R1\_001.fastq
  - G5\_Miseq\_R2\_001.fastq
- **Dataset3**
  - Paired end **MiSeq** data for **N21**
  - N21\_Miseq\_1.fastq
  - N21\_Miseq\_2.fastq

Mysterious SAG

**12 assemblies per group**

- **6 assemblies**
  - 3 assemblies with original data
  - 3 assemblies with trimmed data
- **6 assemblies**
  - 3 assemblies with original data
  - 3 assemblies with trimmed data
- **choose assembly yourself**
  - Use same settings as before
  - Try optimizing assembly (program, kmer, flags, ... )

# Overview of exercises today



What is the  
mysterious cell?  
(+ optional  
exercises)

# Organization into groups

- Groups
  - Put your names in google doc
  - Decide who does which assembly
- Morning session
  - Playing with the data individually (familiarize)
  - Each person runs 3 assemblies (total 12 per group)
- Afternoon session (individually or in groups/pairs)
  - Coverage and chimera checking analyses
  - MEGAN analysis
  - Choose the steps to find out what the mysterious SAG is
  - Choose optional exercises if you have time