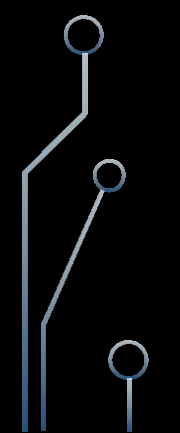
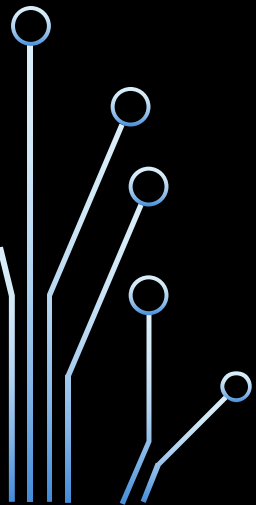


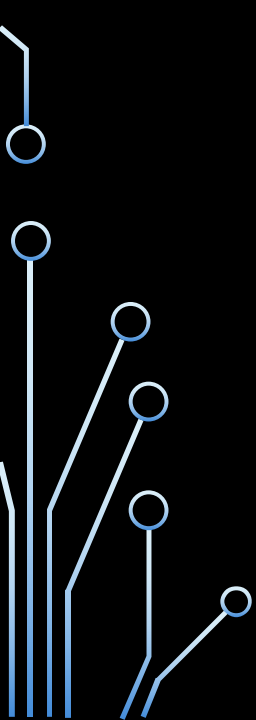
DATA PREPROCESSING

Next Generation Sequencing Analysis
Shyam Gopalakrishnan





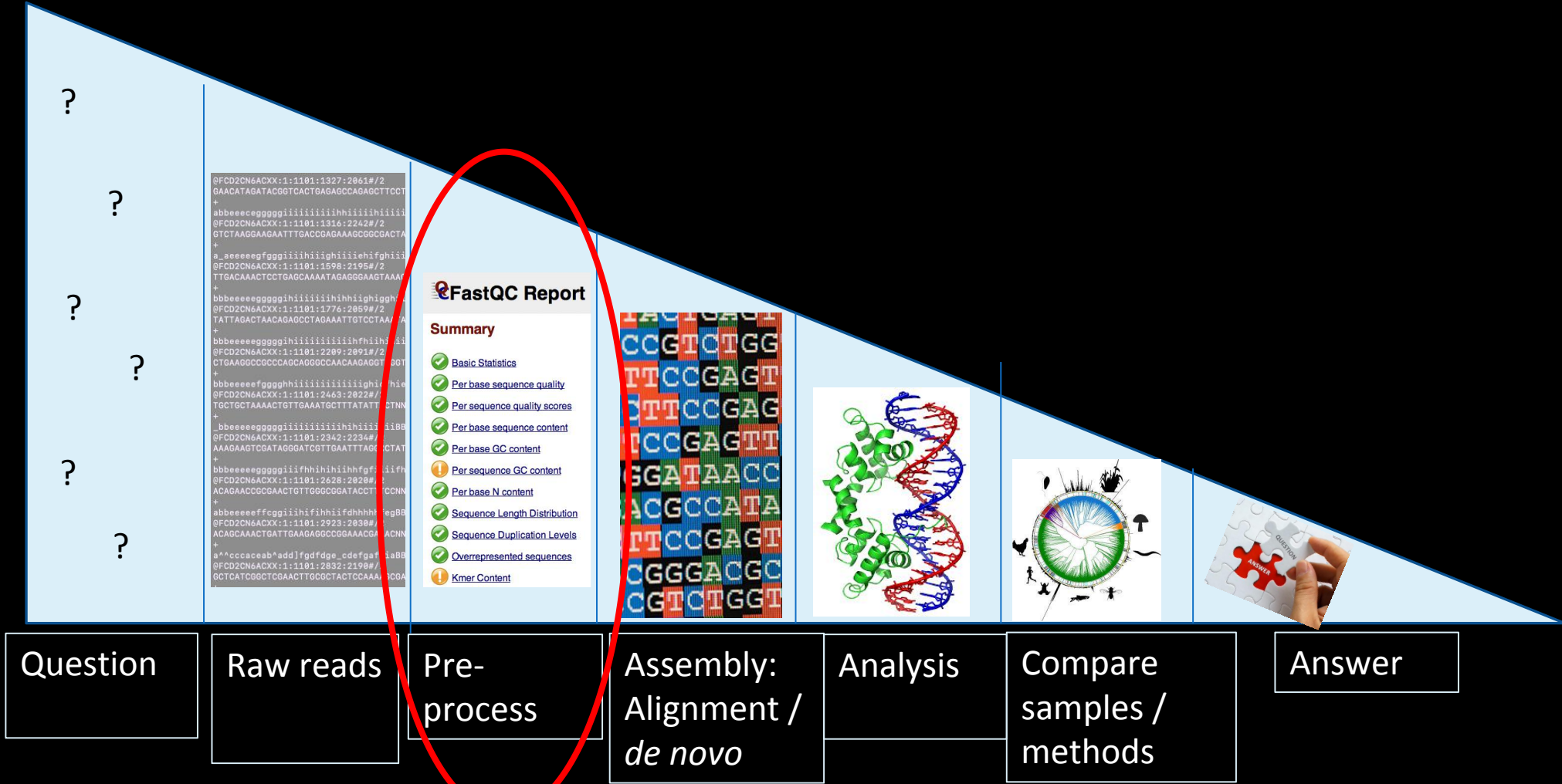
OUTLINE

- The main steps in NGS analysis
 - Why is preprocessing important?
 - Preprocessing
 - Fastqc reports
 - Adapters
 - K-mers
 - Depth of coverage vs Breadth of coverage
 - Merge paired end reads
 - Ion Torrent data
 - Exercises
- 



MAIN STEPS IN NGS ANALYSIS

DATA
SIZE



WHY IS PREPROCESSING IMPORTANT?

Quality?

Every base in a read have a quality score
Note: bases are not always correct!

Adapters?

Adapters/primers are non-biological sequences that can be a part of the raw data.



Do we trust our data?

Errors?

Different sequencing technologies has different error profiles.

Sequencing depth?

How deep is the sample sequenced. How many times that your data covers the genome.

FASTQC REPORTS

- Report basic statistics on your data
- Identify issues with your data

✓ Basic Statistics

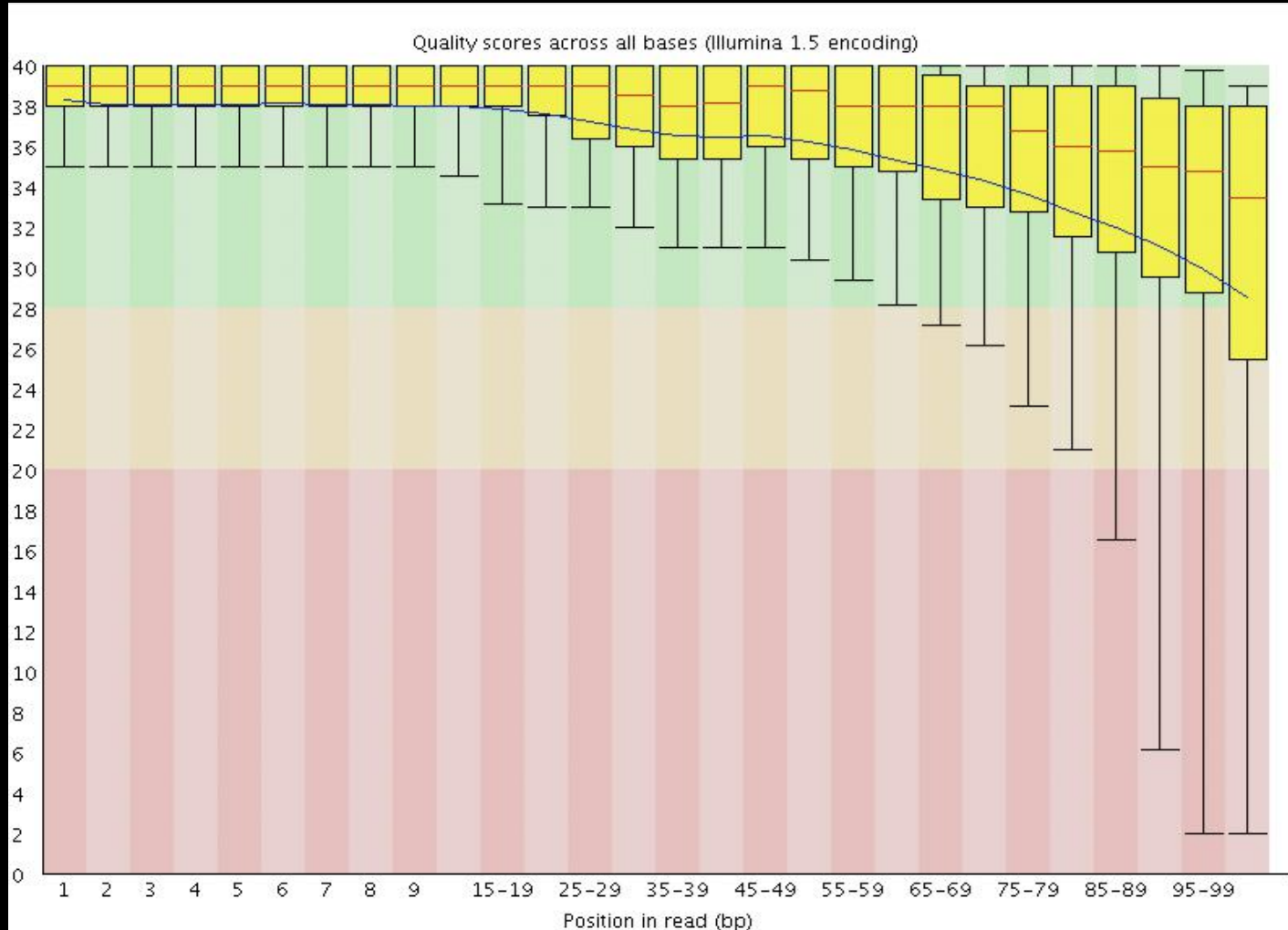
Measure	Value
Filename	tmp.fastq
File type	Conventional base calls
Encoding	Illumina 1.5
Total Sequences	250000
Filtered Sequences	0
Sequence length	101
%GC	51

FastQC Report

Summary

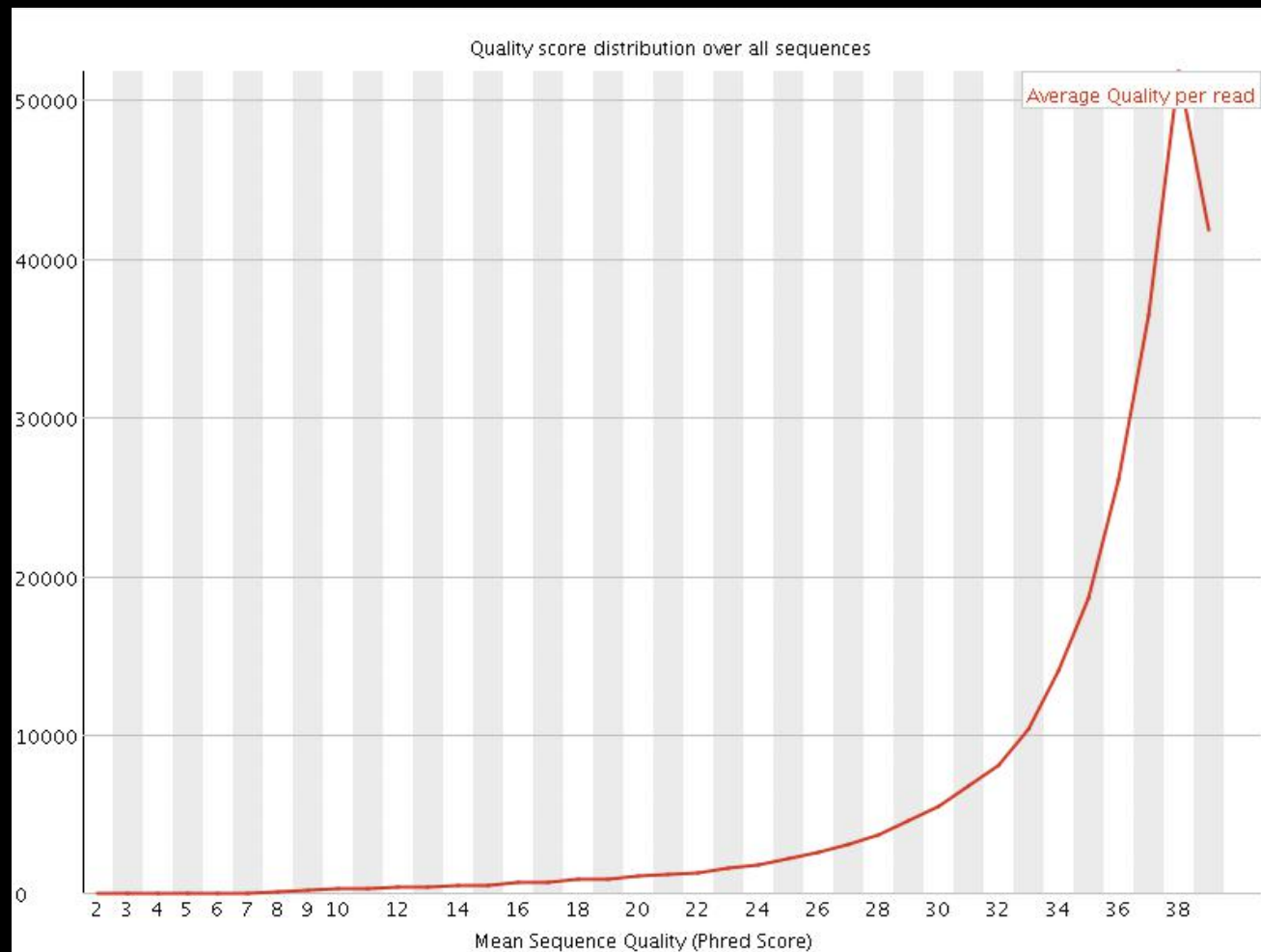
- ✓ [Basic Statistics](#)
- ✓ [Per base sequence quality](#)
- ✓ [Per sequence quality scores](#)
- ✓ [Per base sequence content](#)
- ✓ [Per base GC content](#)
- ! [Per sequence GC content](#)
- ✓ [Per base N content](#)
- ✓ [Sequence Length Distribution](#)
- ✓ [Sequence Duplication Levels](#)
- ✓ [Overrepresented sequences](#)
- ! [Kmer Content](#)

PER BASE SEQUENCE QUALITY



Quality often decreases over the read.

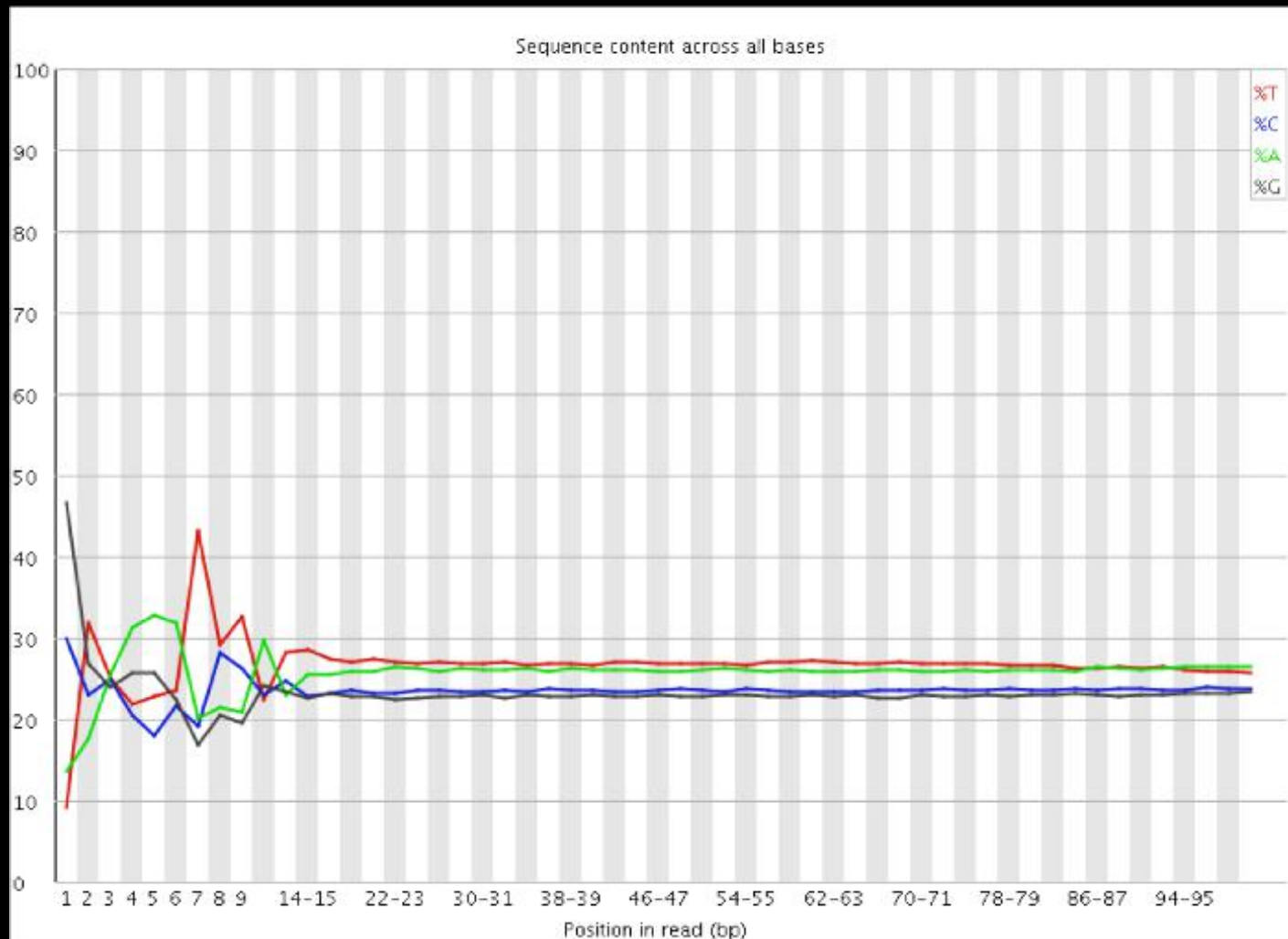
AVERAGE QUALITY



Remove reads with a quality below 20.

Remove reads with 'N' base calls.

TRIM FROM 5'



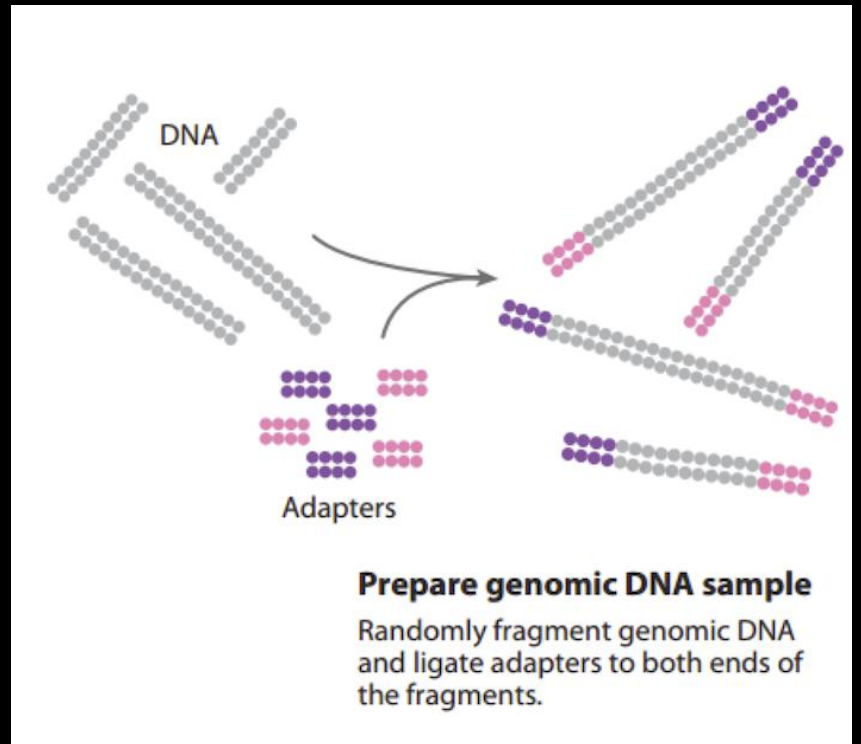
Sometimes something is fishy in the beginning of the read.

It is recommended to remove the first number of bases from the 5'.

How many bases would you remove in this case?

ADAPTERS

- Sometimes adapters / primers are also part of the read
- Adapter / primers are non-biological sequences
- The artificial repeats will disturb alignments and *de novo* assembly
- The sequence is often known, if not, FastQC may find them



ADAPTERS

Overrepresented sequences

Sequence	Count	Percentage	Possible Source
GATCGGAAGAGCACACGTCTGAACTCCAGTCACATCACGATATCGTATGC	1547768	38.192098035156306	TruSeq Adapter, Index 1 (98% over 50bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACATCACGATCTCGTATGC	146635	3.61830603513262	TruSeq Adapter, Index 1 (100% over 50bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACATCAAGATATCGTATGC	6639	0.16382128255358863	TruSeq Adapter, Index 1 (97% over 41bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACATCACGATTTTCGTATGC	6462	0.15945370204267054	TruSeq Adapter, Index 1 (98% over 50bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACATTACGATATCGTATGC	5433	0.1340625136486891	TruSeq Adapter, Index 1 (97% over 41bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACATAACGATATCGTATGC	5147	0.1270052931621209	TruSeq Adapter, Index 1 (97% over 41bp)
GATCGGAAGAGCACACGTCTGAACTCCAGTCACACCACGATATCGTATGC	4703	0.11604932849066535	TruSeq Adapter, Index 1 (97% over 41bp)

We will use “Cutadapt” and “AdapterRemoval”, but other programs can also do the job.

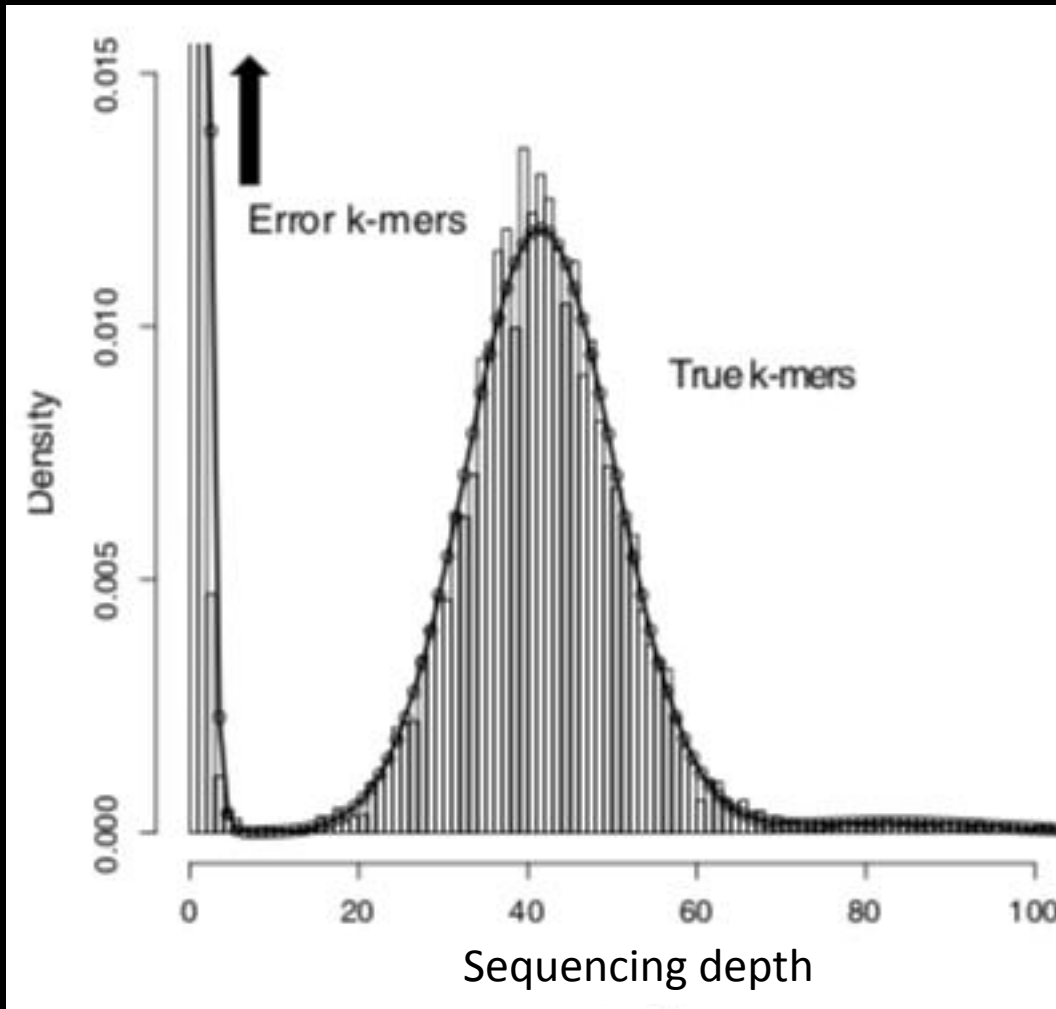
K-MER CORRECTION

- Create a sliding window of size k , move it over all your reads and count occurrence of k -mers
- We can use this to correct sequencing errors!

$k=4$ \longrightarrow
DNA: ACGTGTAACGTGACGTTGGA

ACGT
CGTG
GTGT
TGTA

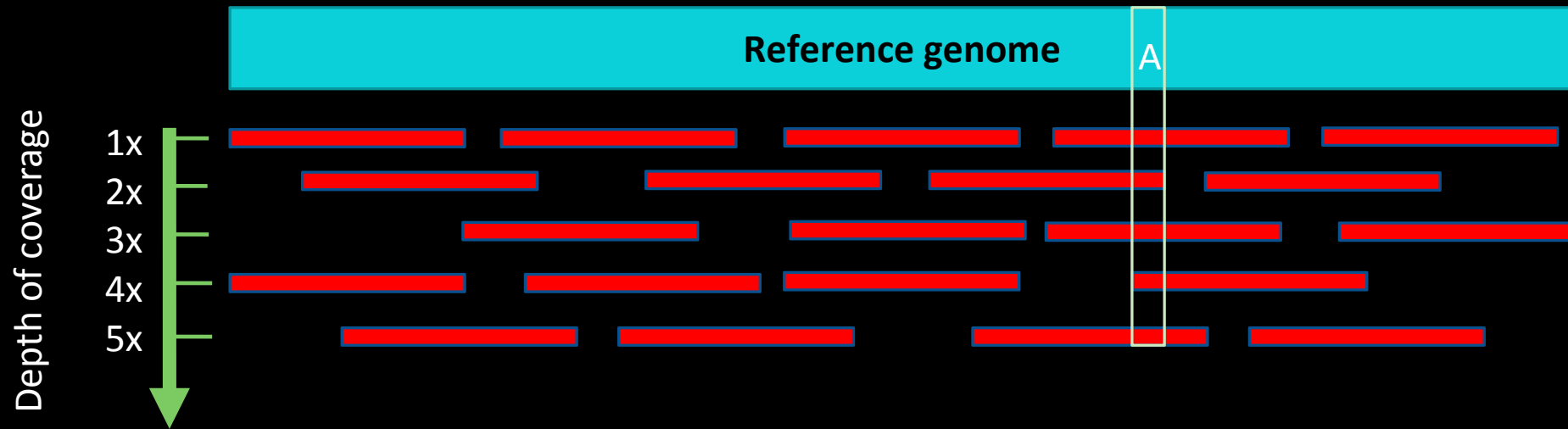
K-MER CORRECTION



Concept: rare k -mers are sequencing errors.
In general we need a $> 15x$ sequencing depth

```
ACGTGGTTGCCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA  
ACGTGGTTACCCTTAAA
```

SEQUENCING DEPTH



How many times that your data covers the genome (average).

SEQUENCING DEPTH

$$C = N \times \frac{L}{G}$$

N: Number of reads

L: Read length

G: Genome size

C: Sequencing depth

Example:

N = 5 mill

L: 100 bases

G: 5 mill bases

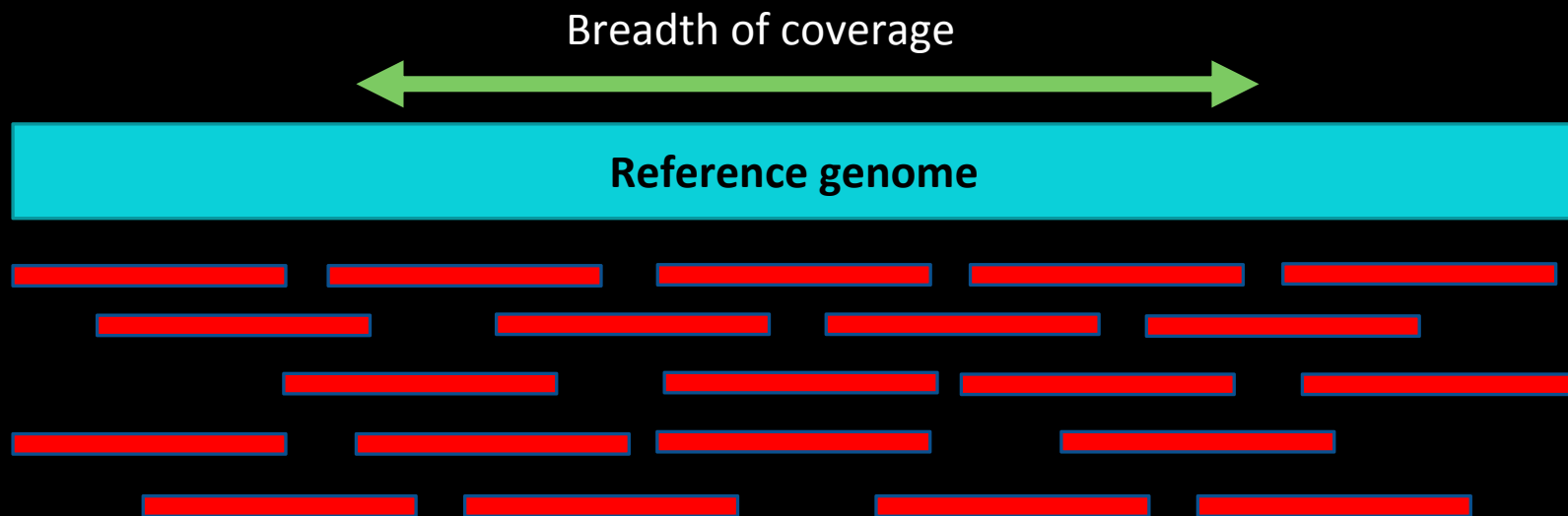
$$C = 5.000.000 \times \frac{100}{5.000.000}$$

$$C = 5 \times \frac{100}{5}$$

$$C = 100X$$

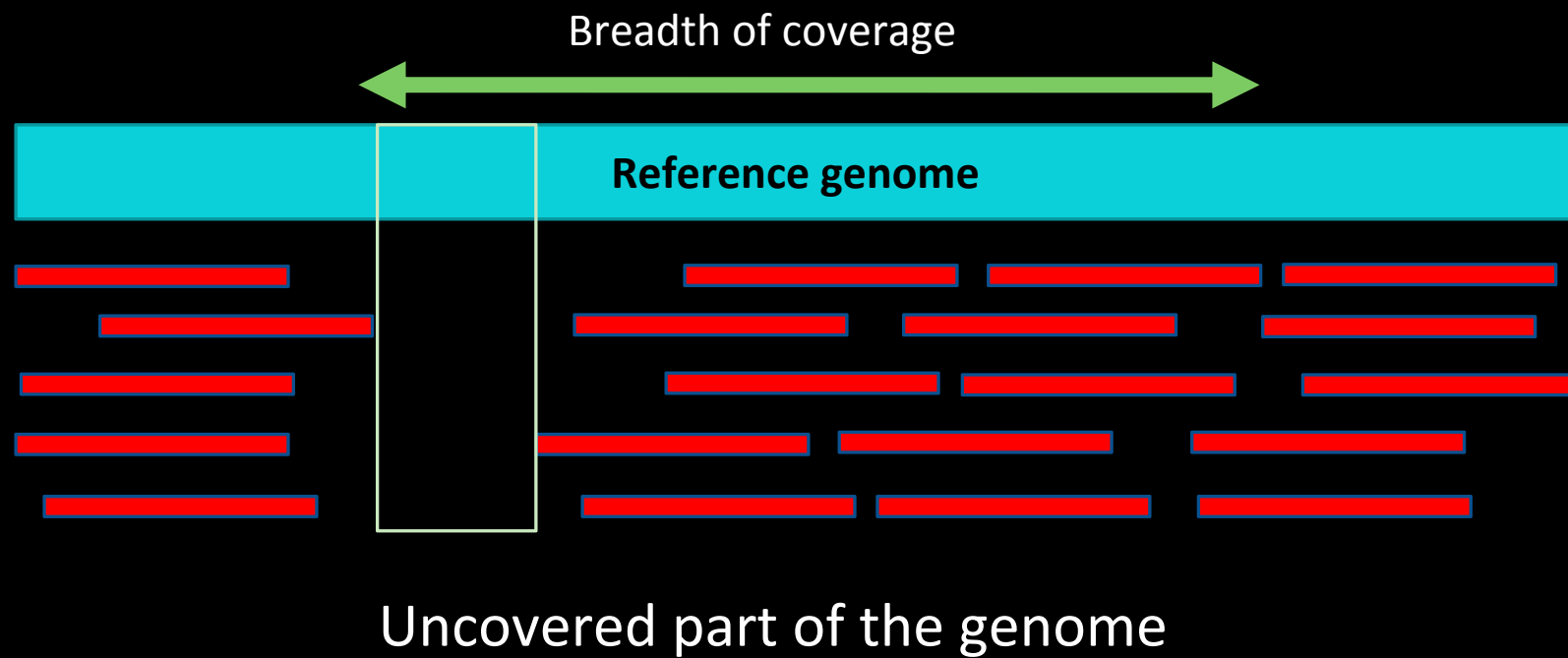
On average there are 100 reads covering each position in the genome

GENOME COVERAGE

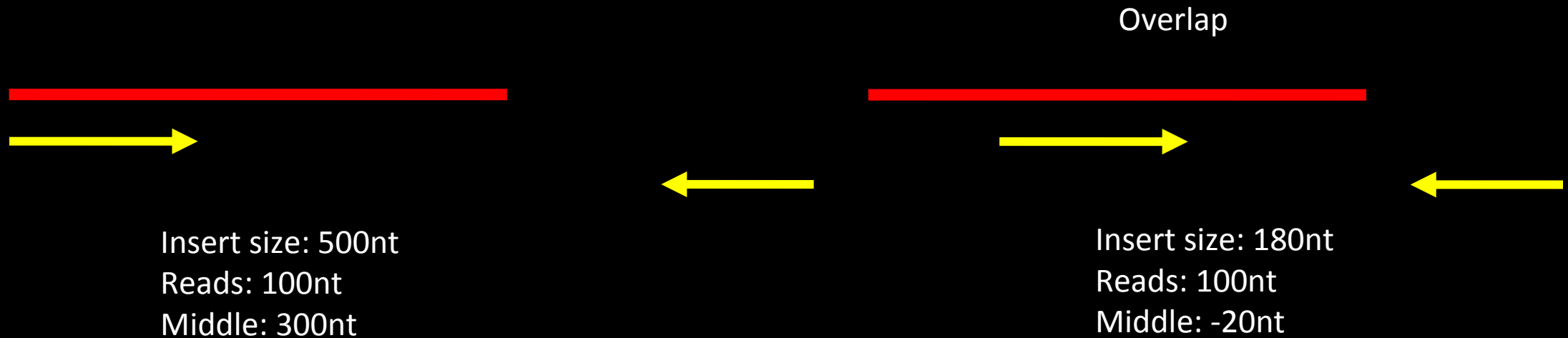


How much of the reference genome is covered by your data

GENOME COVERAGE



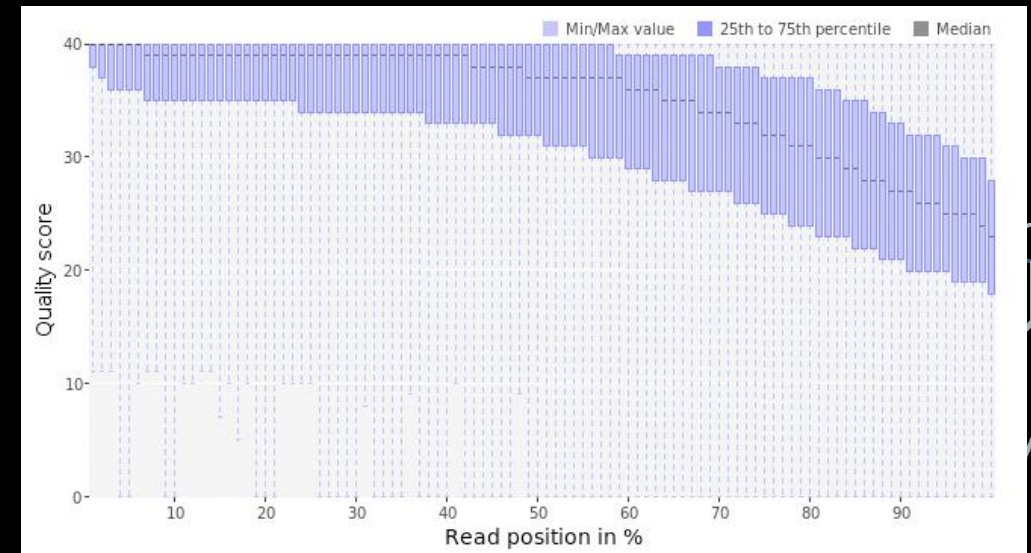
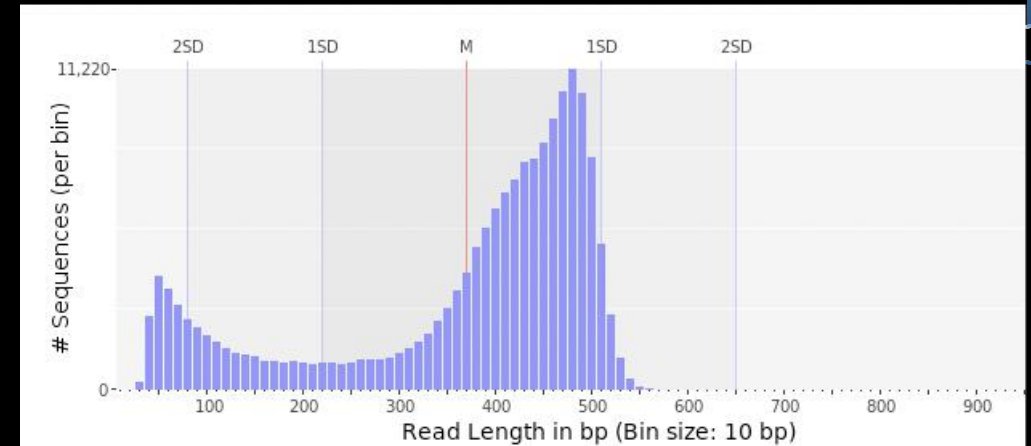
MERGE PAIRED END READS



- Merge overlapping pairs into single longer read
- Smart because Illumina reads have low quality in the 3'
- Very useful for *de novo* assembly

454 / ION TORRENT DATA

- Main problem is indels at homopolymer runs
- (Trim homopolymers), trim trailing poor quality bases
- Remove very short reads
- For *de novo* assembly, adapters should be removed (prinseq)
- For alignment we use Smith- Waterman (local) so less important



Quality control for other technologies

- We heard about other newer technologies yesterday
 - Pac bio, Nanopore etc.
 - How can we do quality control on reads from these technologies?
- Long reads quality control

LongQC: A Quality Control Tool for Third Generation Sequencing Long Read Data

 Yoshinori Fukasawa,  Luca Ermini, Hai Wang, Karen Carty and Min-Sin Cheung

G3: GENES, GENOMES, GENETICS April 1, 2020 vol. 10 no. 4 1193-1196;

<https://doi.org/10.1534/g3.119.400864>

NanoPack: visualizing and processing long-read sequencing data

Wouter De Coster , Sven D'Hert, Darrin T Schultz, Marc Cruts, Christine Van Broeckhoven

FINAL – BUT IMPORTANT NOTE

- Lots of data - storage is expensive!
- Keep data compressed whenever possible (gzip, bzip, bam)
- Remove intermediate files and files that can easily be re-created

