



### Quality control and artefact removal

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Many thanks to Joanna Krupka, the original creator of these slides

### Why do we need quality control?

NGS generates highly accurate data, however there are still a few types of errors:

- Contamination with adapters
- Technical duplication in the library
- Failure at specific parts of the flow cell
- Amplification bias PCR duplicates
- •

# Quality scores in fastq files

Fastq files store quality scores in ASCII characters

$$e$$
 = probability of sequence base being  $\longrightarrow Q = -10 \cdot log_{10}(e) \longrightarrow$  Quality character = ASCII( $Q$  + 33) wrong

ASCII-quality score mapping:

# Quality scores in fastq files

Fastq files store quality scores in ASCII characters

$$\begin{array}{c} e = \text{probability} \\ \text{of sequence} \\ \text{base being} \end{array} \longrightarrow \begin{array}{c} \text{Phred score} \\ Q = -10 \cdot log_{10}(e) \end{array} \longrightarrow \begin{array}{c} \text{Quality character} \\ \text{= ASCII}(Q + 33) \end{array}$$
 wrong

ASCII-quality score mapping:

#### FastQC

#### So we use FastQC

- A tool to generate reports based on sequencing quality information from FASTQ or SAM/BAM files
- Command line and interactive mode
- Outputs an html report and a .zip file with the raw quality data
- Enables a quick look at the potential problems with your experiment



# FastQC report - summary

#### Good quality sequence

#### Summary

- Basic Statistics
- Per base sequence quality
- Per tile sequence quality
- Per sequence quality scores
- Per base sequence content
- Per sequence GC content
- Per base N content
- Sequence Length Distribution
- Sequence Duplication Levels
- Overrepresented sequences
- Adapter Content
- **Wer Content**

#### Bad quality sequence

#### Summary

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# FastQC report – basic statistics

#### Basic Statistics

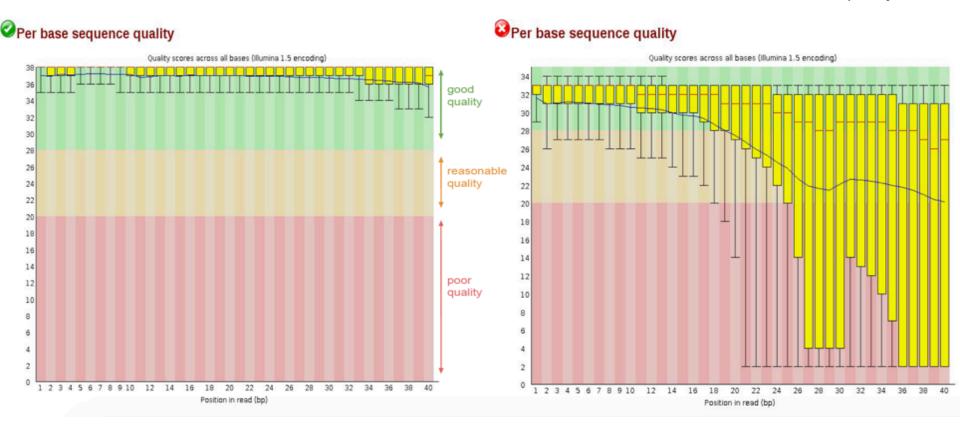
Measure	Value	
Filename	good_sequence_short.txt	
File type	Conventional base calls	
Encoding	Illumina 1.5	
Total Sequences	250000	
Sequences flagged as poor quality	0	
Sequence length	40	
%GC	45	

#### Basic Statistics

Measure	Value	
Filename	bad_sequence.txt	
File type	Conventional base calls	
Encoding	Illumina 1.5	
Total Sequences	395288	
Sequences flagged as poor quality	0	
Sequence length	40	
%GC	47	

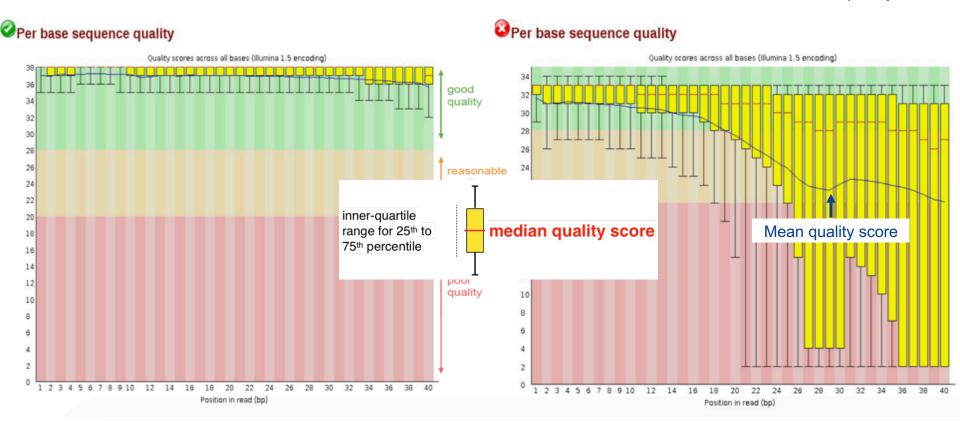
### FastQC report – per base sequence quality

Examines the Phred quality scores



### FastQC report – per base sequence quality

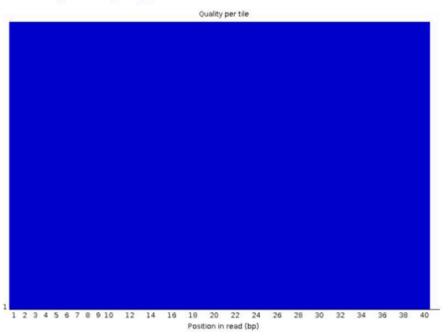
Examines the Phred quality scores



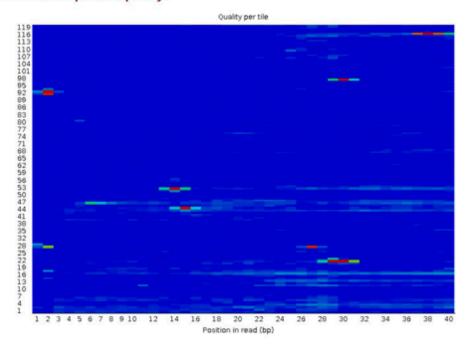
# FastQC report – per tile sequence quality

Examines the Phred quality scores





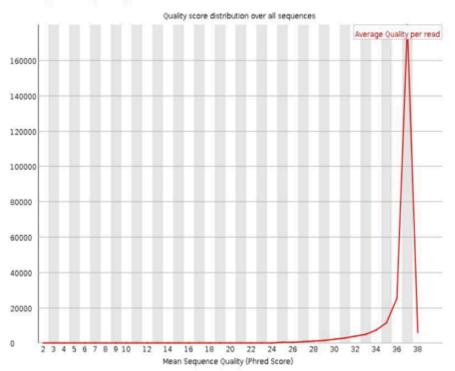
#### **3**Per tile sequence quality



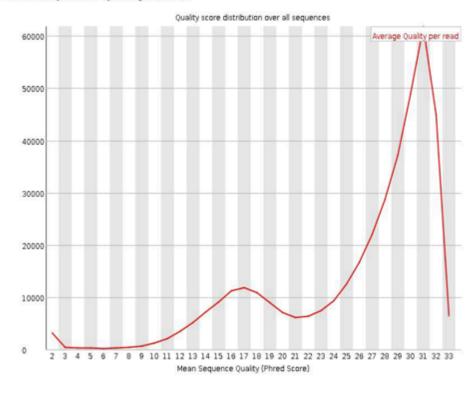
### FastQC report – per sequence quality scores

Examines the Phred quality scores





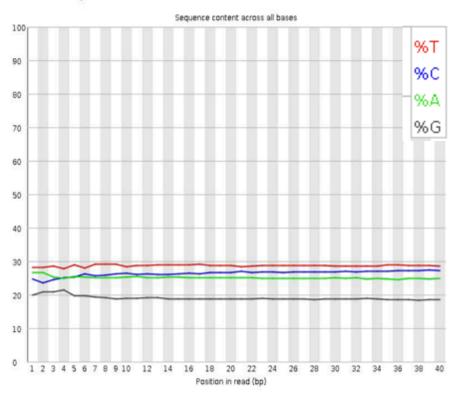
#### Per sequence quality scores



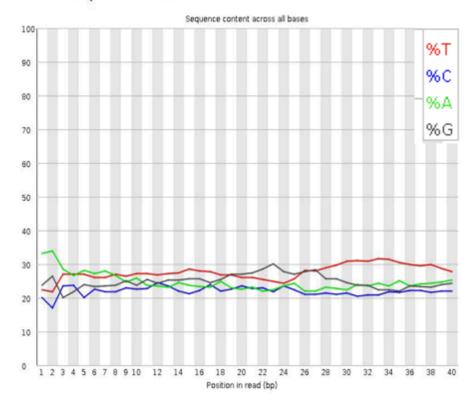
### FastQC report – per base sequence content

Examines the sequence base content



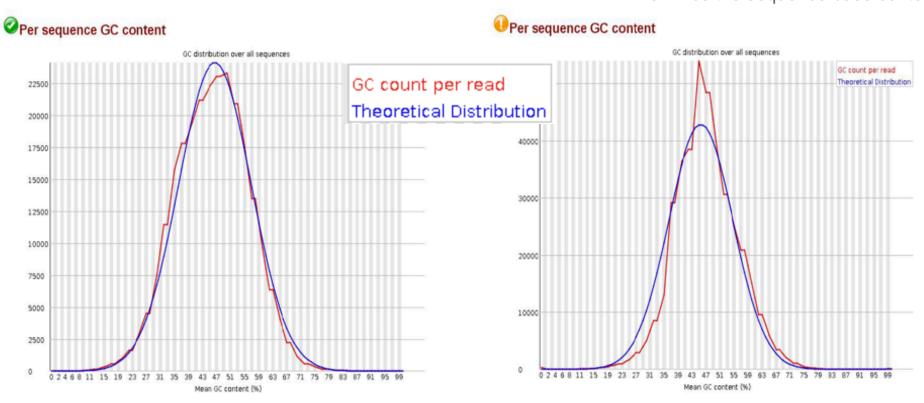


#### Per base sequence content

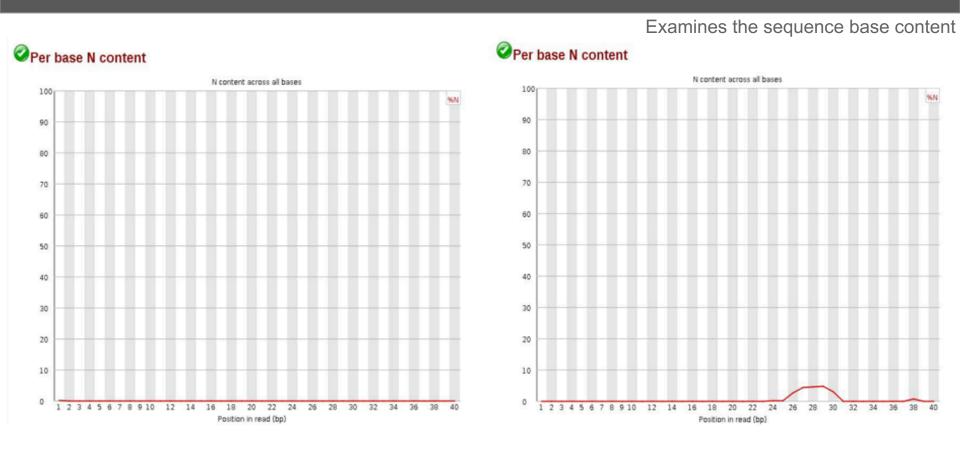


### FastQC report – per sequence GC content

Examines the sequence base content



# FastQC report – per base N content

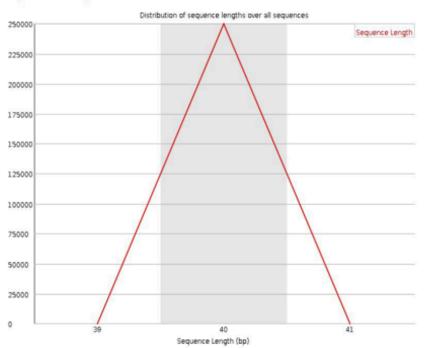


N is a base that could not be called

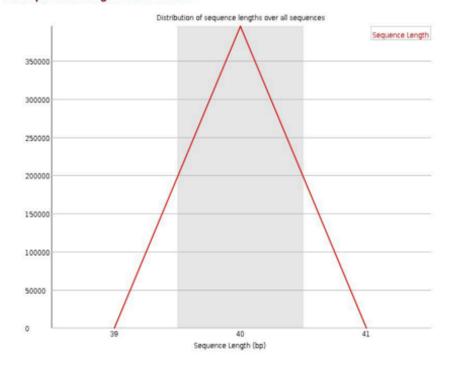
### FastQC report – sequence length distribution

Examines the sequence length



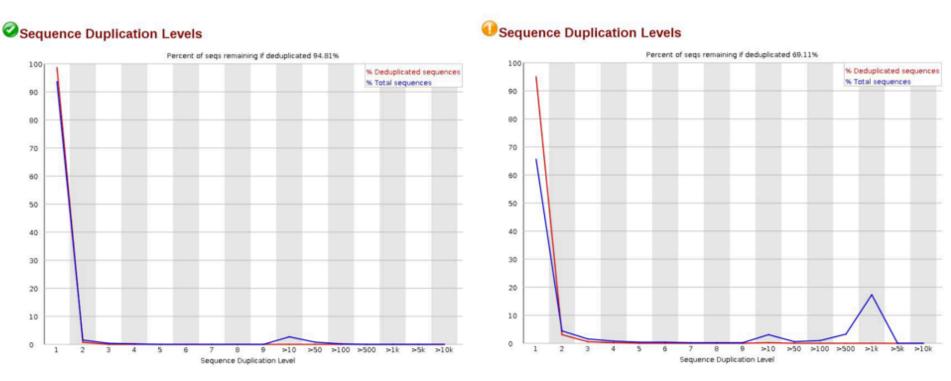


#### Sequence Length Distribution



### FastQC report – sequence duplication level

Examines potential unwanted sequences



Percentage of reads of a given sequence which are present a given number of times in the file.

### FastQC report – overrepresented sequences



A sequence is considered overrepresented if it accounts for  $\geq 0.1\%$  of the total reads.

Examines potential unwanted sequences

Overrepresented sequences

Sequence	Count	Percentage	Possible Source
GAGTTTTATCGCTTCCATGACGCAGAAGTTAACACTTTC	2065	0.5224039181558763	No Hit
ATT GGCGT AT CCAACCT GCAGAGTTTT AT CGCTT CCAT G	2047	0.5178502762542754	No Hit
TTGGCGTATCCAACCTGCAGAGTTTTATCGCTTCCATGA	2014	0.5095019327680071	No Hit
GATAAAAATGATTGGCGTATCCAACCTGCAGAGTTTTAT	1913	0.4839509420979134	No Hit
TATOCAACCTGCAGAGTTTTATCGCTTCCATGACGCAGA	1879	0.47534961850600066	No Hit
AAAAT GATT GGCGT AT CCAACCT GCAGAGTTTT AT CGCT	1846	0.4670012750197325	No Hit
GATT GGCGT AT CCAACCT GCAGAGTTTT AT CGCTT CCAT	1841	0.46573637449150995	No Hit
ACCTGCAGAGTTTTATCGCTTCCATGACGCAGAAGTTAA	1836	0.46447147396328753	No Hit
AT AAAAAT GATT GGCGT AT CCAACCT GCAGAGTTTT AT C	1831	0.4632065734350651	No Hit
AATGATTGGCGTATCCAACCTGCAGAGTTTTATCGCTTC	1779	0.45005160794155147	No Hit
TGATTGGCGTATCCAACCTGCAGAGTTTTATCGCTTCCA	1779	0.45005160794155147	No Hit
ATGATTGGCGTATCCAACCTGCAGAGTTTTATCGCTTCC	1760	0.4452449859343061	No Hit
AAATGATTGGCGTATCCAACCTGCAGAGTTTTATCGCTT		0.4374026026593269	No Hit
GT AT CCAACCT GCAGAGTTTT AT CGCTT CCAT GACGCAG		0.43335492096901496	
TCCAACCTGCAGAGTTTTATCGCTTCCATGACGCAGAAG		0.43209002044079253	
AGAGTTTTATCGCTTCCATGACGCAGAAGTTAACACTTT		0.42601849790532476	
			No Hit
GCAGAGTTTT AT CGCTT CCAT GACGCAGAAGTT AACACT			
AACCT GCAGAGTTTT AT CGCTT CCAT GACGCAGAAGTT A			
AT CCAACCT GCAGAGTTTT AT CGCTT CCAT GACGCAGAA	1630	0.4123575722005221	No Hit
GGTT CAGCAGGAAT GCCGAGAT CGGAAGAGCGGTT CAGC	599	0.15153508328105078	Illumina Paired End PCR Primer 2 (96% over 25b)
CTGCAGGTTGGATACGCCAATCATTTTTATCGAAGCGCG	585	0.1479933618020279	No Hit
OCTTANAGCT ACCAGTTATAT GOCT GGGGGGTTTTTTTT		0.13964501831575965	
TCTGCAGGTTGGATACGCCAATCATTTTTATCGAAGCGC		0.1345854162028698	
TGCGTCATGGAAGCGATAAAACTCTGCAGGTTGGATACG		0.13028475440691342	
TGCAGGTTGGATACGCCAATCATTTTTATCGAAGCGCGC	505	0.12775495335046852	No Hit

### FastQC report – overrepresented sequences

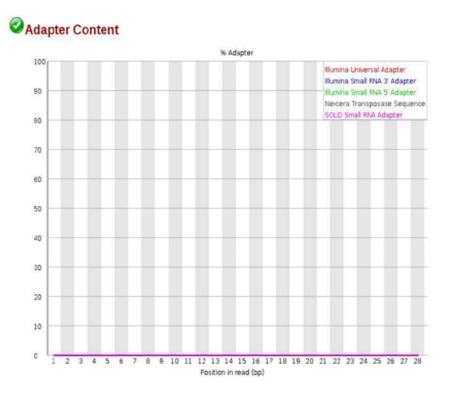
Examines potential unwanted sequences

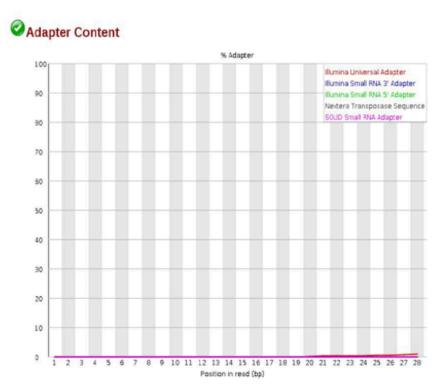
#### Overrepresented sequences

Sequence	Count	Percentage	Possible Source
${\tt GATCGGAAGAGCACACGTCTGAACTCCAGTCACGCCAATATCTCGTATGC}$	4156	0.20779999999999998	TruSeq Adapter, Index 6 (100% over 50bp)
${\tt TTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTT$	3490	0.1745	No Hit

# FastQC report – adapter content

Examines potential unwanted sequences

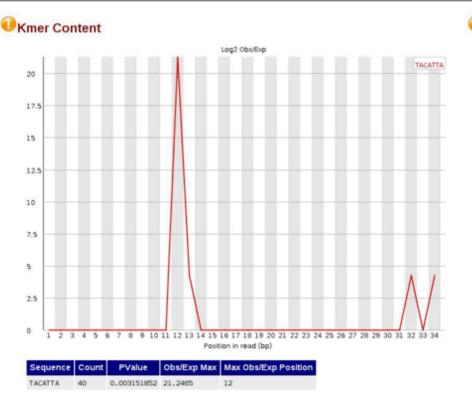




Cumulative plot of the percentage of reads where an adapter sequence has been identified at the indicated base position.

# FastQC report – Kmer content

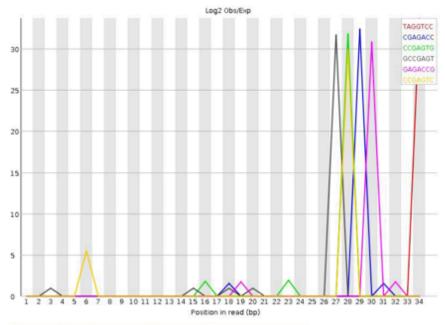
**Kmer Content** 



By default not displayed in the report

Measures the count of each k-mer (by default k
= 7) at each position along the read

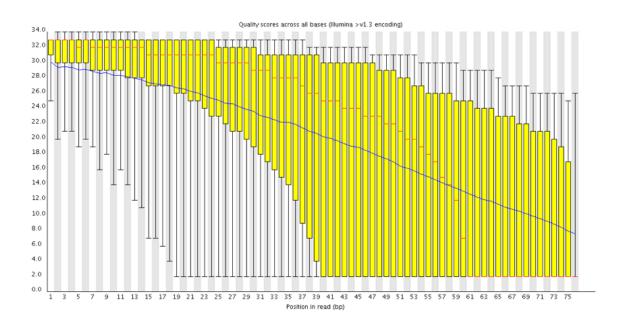
Examines potential unwanted sequences



Sequence	Count	PValue	Obs/Exp Max	Max Obs/Exp Position
TAGGTCC	30	1.5992917E-5	33.6211	34
CGAGACC	105	0.0	32.37975	29
CCGAGTG	90	0.0	31.803032	28
GCCGAGT	170	0.0	31.625078	27
GAGACCG	95	0.0	30.826315	30
CCGAGTC	30	4.3762376E-4	29.815344	28

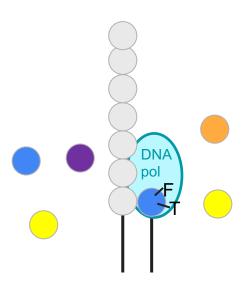
A drop in sequence quality towards the 3'end of the read is common

Normally, you can trim the reads so that you just keep the parts that have a Phred score > 20



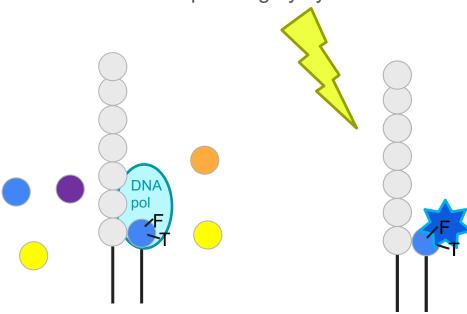
This drop in sequence quality is often caused by phasing

Normal Illumina sequencing by synthesis:



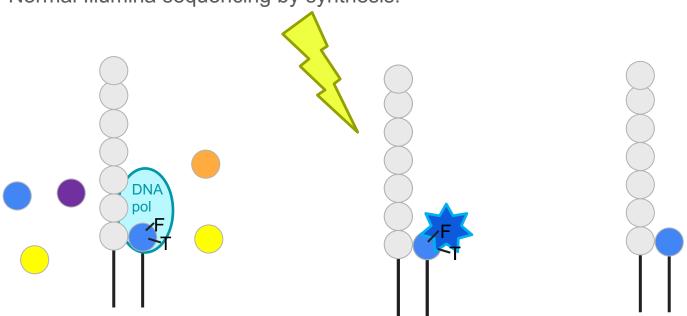
This drop in sequence quality is often caused by **phasing** 

Normal Illumina sequencing by synthesis:



This drop in sequence quality is often caused by phasing

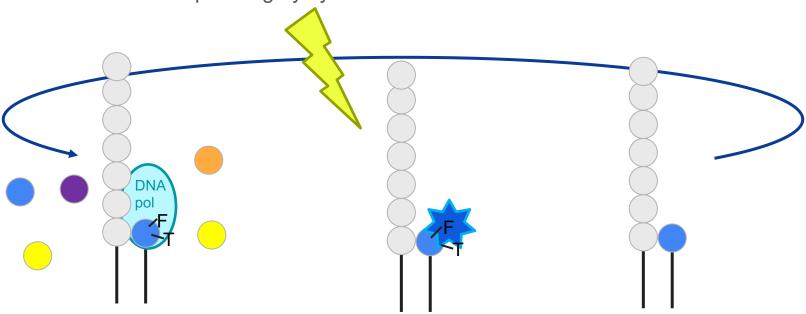
Normal Illumina sequencing by synthesis:



https://www.ebi.ac.uk/training/online/courses/functional-genomics-ii-common-technologies-and-data-analysis-methods/next-generation-sequencing/illumina-sequencing/

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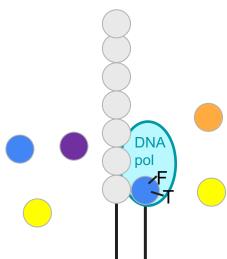
Normal Illumina sequencing by synthesis:



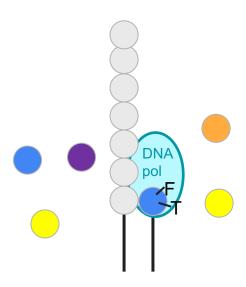
https://www.ebi.ac.uk/training/online/courses/functional-genomics-ii-common-technologies-and-data-analysis-methods/next-generation-sequencing/illumina-sequencing/

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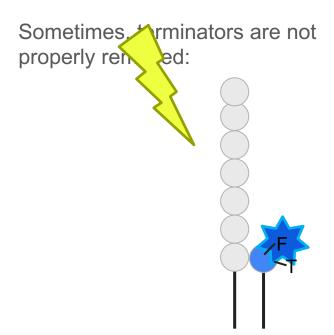
Sometimes, terminators are not properly removed:

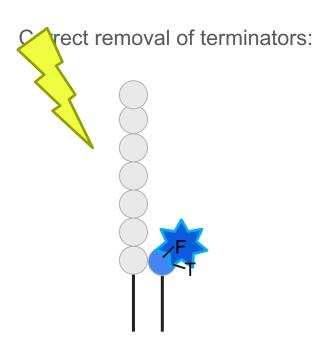


Correct removal of terminators:



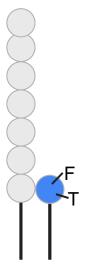
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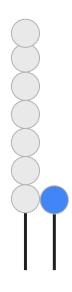


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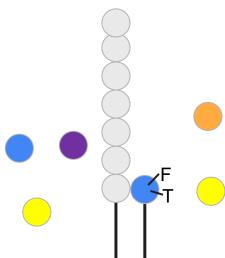


Correct removal of terminators:

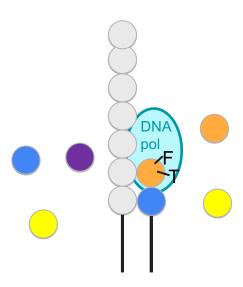


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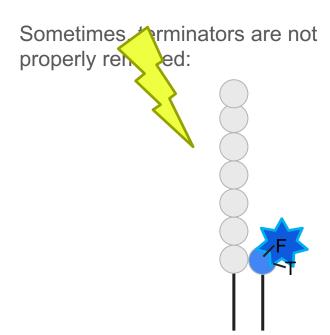
Sometimes, terminators are not properly removed:

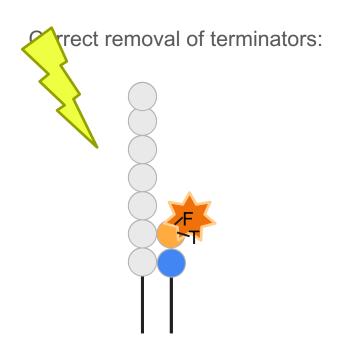


Correct removal of terminators:



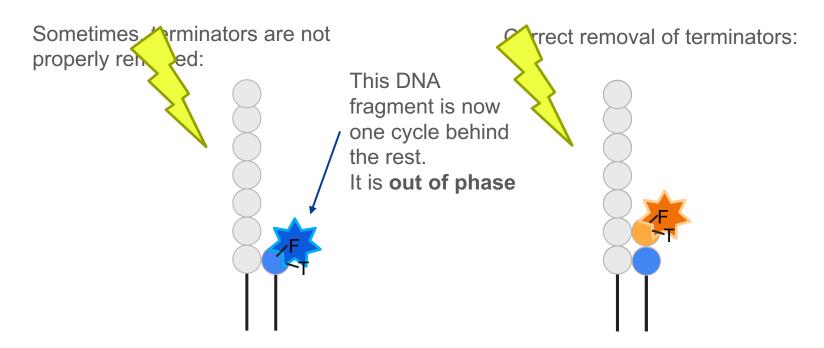
This drop in sequence quality is often caused by **phasing** 





https://www.ebi.ac.uk/training/online/courses/functional-genomics-ii-common-technologies-and-data-analysis-methods/next-generation-sequencing/illumina-sequencing/

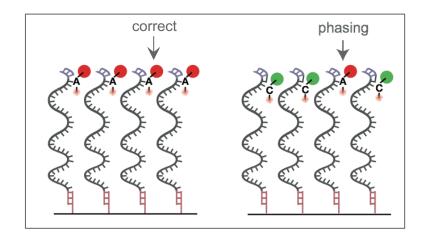
This drop in sequence quality is often caused by **phasing** 



https://www.ebi.ac.uk/training/online/courses/functional-genomics-ii-common-technologies-and-data-analysis-methods/next-generation-sequencing/illumina-sequencing/

#### This drop in sequence quality is often caused by phasing

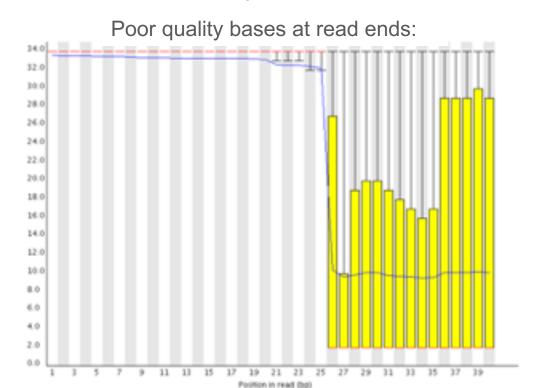
- This fragment will pollute the light signal that the sequencer's camera has to read
- Over time, with increasing read length, they add up and pollute the light signal more and more, leading to lower and lower quality scores



 Defect terminator caps can also cause a similar effect, where two nucleotides can bind in one cycle (called prephasing)

So what do we do if the quality isn't good enough?

Often we need to remove the bad parts



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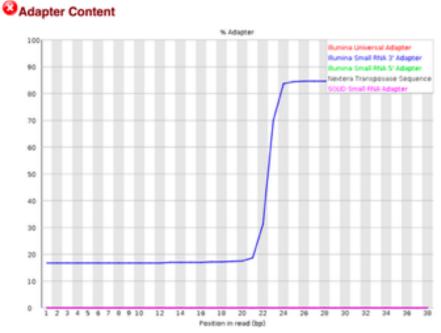
Often we need to remove the bad parts



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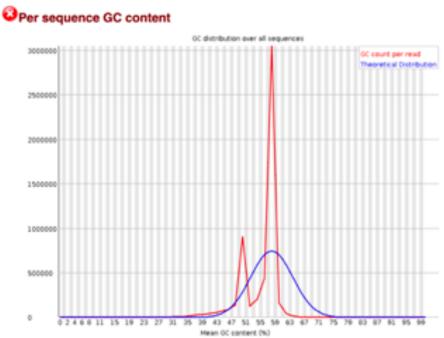
Leftover adapter sequences:



So what do we do if the quality isn't good enough?

Often we need to remove the bad parts

Known contaminants:



So what do we do if the quality isn't good enough?

Often we need to remove the bad parts

In the practical, we will use **Cutadapt** to perform quality trimming of our sample dataset

#### Practical time

# Let's practice!