



Quality assessment and read preprocessing

Sarai Varona

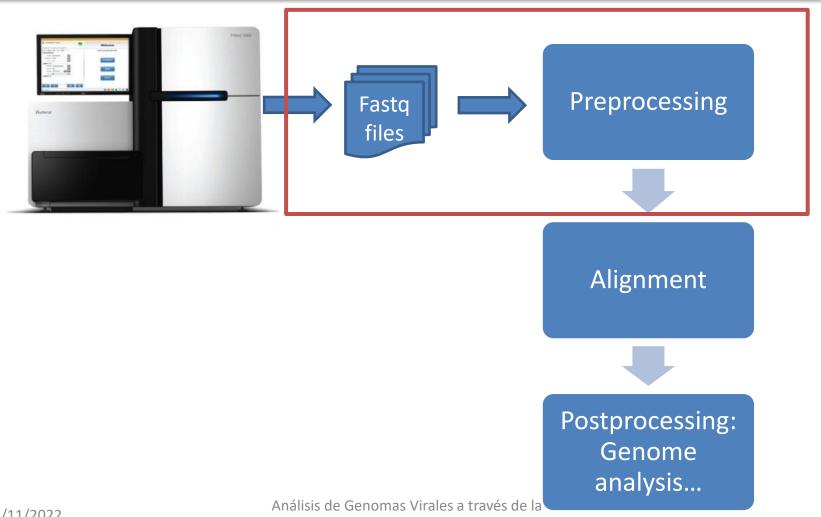
<u>BU-ISCIII</u> <u>Unidades Comunes Científico Técnicas - SGSAFI-ISCIII</u>

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Step in the process







Raw output files format





.bcl .fastq



454 .sff





Nanopore .fast5 .fastq



Bax.h5 fasta





- Is a FASTA file with quality information
- Within HTS, FASTA contain genomes y FASTQ reads

Quality: must be 1 bit





- Each base has an assigned quality score
 - Sequencing quality scores measure the probability that a base is called incorrectly
- How is it calculated?

Phred transforming

| '''*((((***+))%%++)(%%%).1***-+*''))**55CCF>>>>>CCCCCCC65

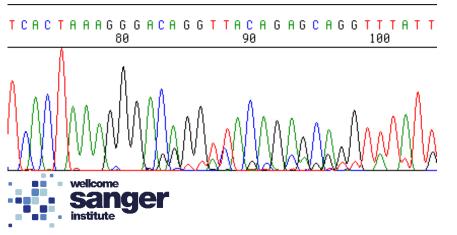


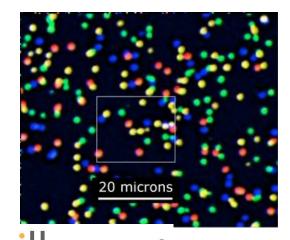


- Light intensity is used to calculate the error probabilities
- Convert error probability into Phred score quality -Ewing B, Green P. (1998)

 Phred originated as an algorithmic approach that considered Sanger sequencing metrics, such as peak

resolution and shape









- Convert error probability into Phred score quality in real time on Illumina platforms
- Q scores are defined as a property that is logarithmically related to the base calling error probabilities (P)
- Phred quality range between 0-40 for Sanger and Illumina
 1.8+

$$Q = -10 \log_{10} P$$

Phred Quality Score	Probability of Incorrect Base Call	Base Call Accuracy
10	1 in 10	90%
20	1 in 100	99%
30	1 in 1,000	99.9%
40	1 in 10,000	99.99%
50	1 in 100,000	99.999%

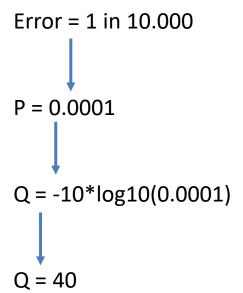




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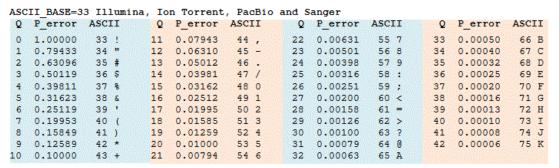
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50	1 in 100,000	99.999%







 Convert Phred quality score into ASCII, a compact form, which uses only 1 byte per quality value



 Phred+33 (Sanger and current Illumina). O Phred quality correspond to decimal 33, which is the symbol!

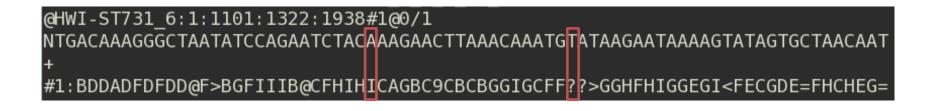
ASC	II BASE=6	4 Old Illu	umina								
Q	Perror	ASCII	Q	P_error	ASCII	Q	P_error	ASCII	Q	P_error	ASCII
0	1.00000	64 @	11	0.07943	75 K	22	0.00631	86 V	33	0.00050	97 a
1	0.79433	65 A	12	0.06310	76 L	23	0.00501	87 W	34	0.00040	98 b
2	0.63096	66 B	13	0.05012	77 M	24	0.00398	88 X	35	0.00032	99 c
3	0.50119	67 C	14	0.03981	78 N	25	0.00316	89 Y	36	0.00025	100 d
4	0.39811	68 D	15	0.03162	79 0	26	0.00251	90 Z	37	0.00020	101 e
5	0.31623	69 E	16	0.02512	80 P	27	0.00200	91 [38	0.00016	102 f
6	0.25119	70 F	17	0.01995	81 Q	28	0.00158	92 \	39	0.00013	103 g
7	0.19953	71 G	18	0.01585	82 R	29	0.00126	93]	40	0.00010	104 h
8	0.15849	72 H	19	0.01259	83 S	30	0.00100	94 ^	41	0.00008	105 i
9	0.12589	73 I	20	0.01000	84 T	31	0.00079	95	42	0.00006	106 j
10	0.10000	74 J	21	0.00794	85 U	32	0.00063	96 -			

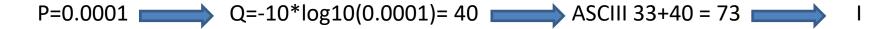
Phred+64 (Solexa and Illumina 1.3-1.5)

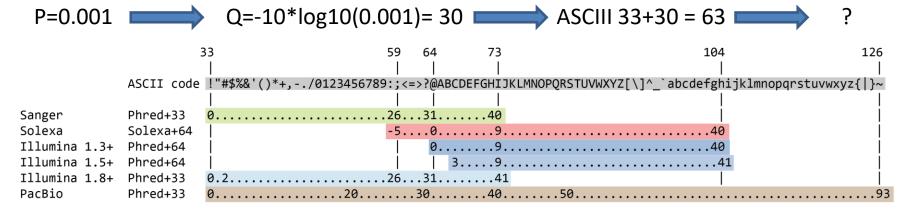




Phred 33 example

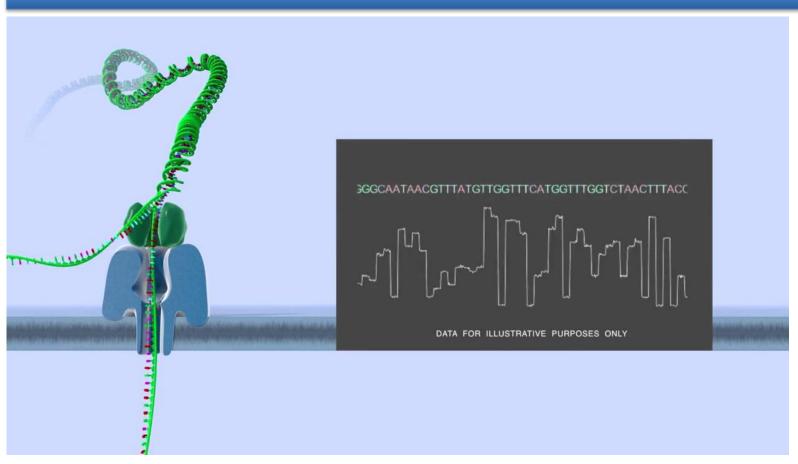








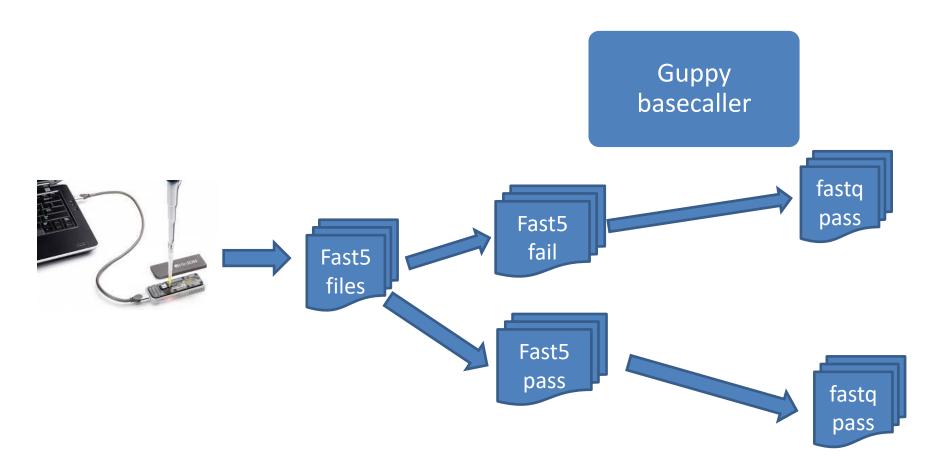




Clara Delahaye, Jacques Nicolas. Nanopore MinION long read sequencer: an overview of its error landscape. 2020. ffhal-03123133f

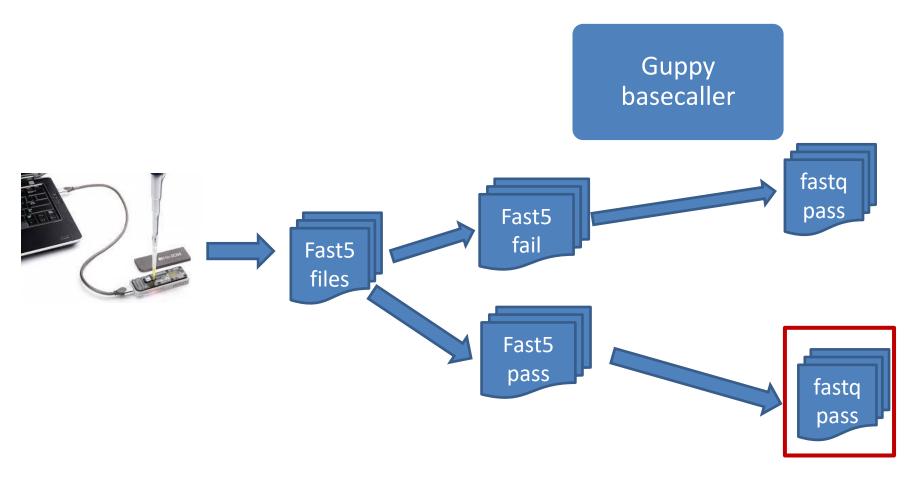








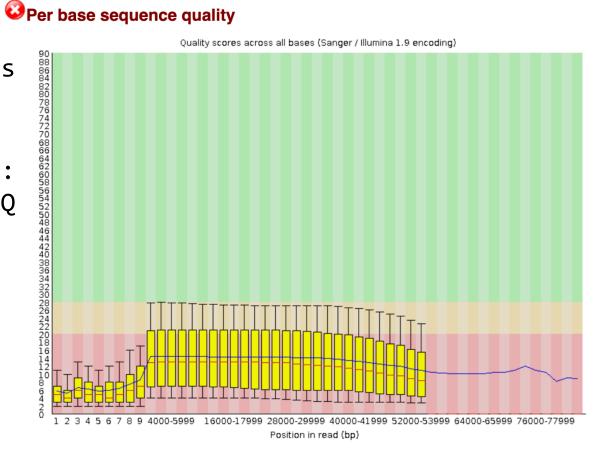








- Nanopore quality score (Q) does not follow Phred scores
- To estimate error rate (E) (locally and at read level):
 E = 0.015Q2 1.15Q + 24



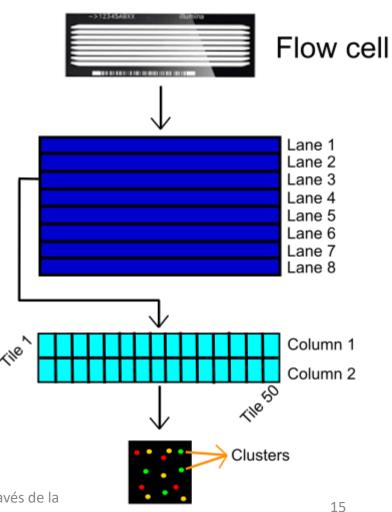




Illumina read header

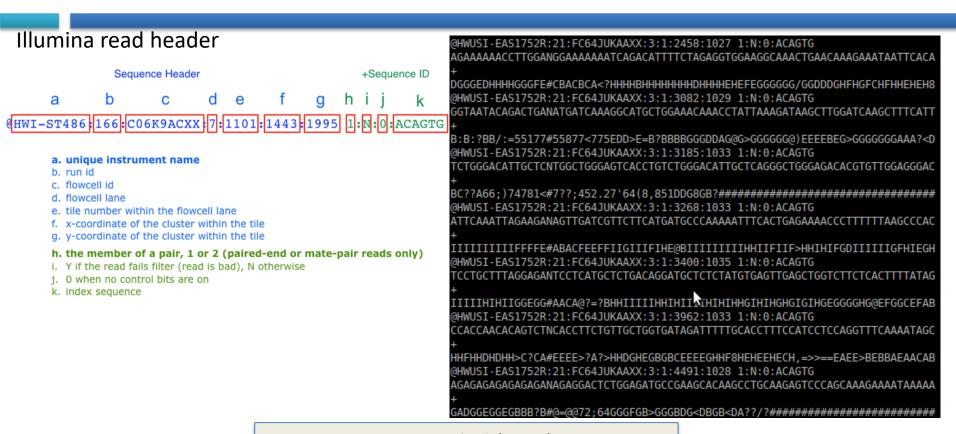
Sequence Header +Sequence ID a @HWI-ST486 166 C06K9ACXX 7:1101:1443 1995 1:N:0:ACAGTG

- a. unique instrument name
- b. run id
- c. flowcell id
- d. flowcell lane
- e. tile number within the flowcell lane
- f. x-coordinate of the cluster within the tile
- g. y-coordinate of the cluster within the tile
- h. the member of a pair, 1 or 2 (paired-end or mate-pair reads only)
- i. Y if the read fails filter (read is bad), N otherwise
- i. 0 when no control bits are on
- k. index sequence









ASCII-coded (0-40):

- "!"#\$%" lowest quality
- "FGHI" highest quality

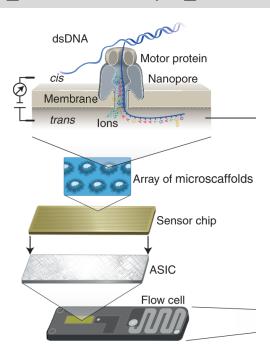




Nanopore read header

@d76be4fb-11a9-47e7-90be-c4f15591e0d9 runid=ba02134f00f2059e7b2dc248113c02f76577b101 read=11 ch=142 start_time=2019-06-27T11:09:03Z flow_cell_id=FAH59799 protocol_group_id=k6963 sample_id=k6963

- 1. @read identifier
- 2. run-id
- 3. read-id
- 4. channel
- 5. start_time
- 6. flow cell id
- protocol_group_id
- 8. sample_id







Nanopore read header

@d76be4fb-11a9-47e7-90be-c4f15591e0d9 runid=ba02134f00f2059e7b2dc248113c02f76577b101 read=11 ch=142 start_time=2019-06-27T11:09:03Z flow_cell_id=FAH59799 protocol_group_id=k6963 sample_id=k6963

- @read identifier
- run-id
- read-id
- channel
- start_time
- flow_cell_id
- protocol_group_id
- sample_id

```
b101 read=11 ch=142 start_time=2019-06-27T11:09:03Z flow_cell_id=FAH59799 protoc
ol_group_id=k6963 sample_id=k6963
CATTGTACTGATTCAGTTACAATATTGCTGCTTTTCATCAAGGAGAAAGTAATGACAGCGCATCGCAGTGAAAAGAGACT
TCGACCGAAAAAATGCAGAGGCAATGCCACGCCAGCATGACCAGGCAGCAGCGAAAGGGTGCTCGAGATCTCGGACTGTG
ACAACGGAAAAAAGCCAGGGCGATCGTCGCCGCCATGACGAAGTGCTGGCCGAAAGCACGCGCAGGAGTTCCCAGCTATC
TGGCAGACCGGCTCCAGCAAAAGCAATATGAGAATATGAACGAGGTGCTGGCCAGCCGCCTCAATGAACTGCTCGGCGGA
GAACGGGGATAGCGCGAAAATTAATAATGACGATGTAGATAAACCAGAAGCTCAAATGATGTATTCCC
##$&&"#%$$&$17)$#$%#%#'''**)*%"'3679*%*((70>->>B>;>'&,400+&''89?344.&'0=N61%+$33
*)1>;7/++))&##%%(38?;=@?8-?A>4432&(,35*+;6%%$###+'321%+%$)*+$#%')$((158;;%2/10..
8+>66A:9?>79-+*-$$%+,,.-/-*$**,1680('(+2('**,%%.))))600/(+.*)$#&'#2222,<==B:9,6+
-%%$.//*1B9<;)=@&20.--53729</99246##+5))/-;>:;7(*41$#+6&33*'%(*13-$8'9;8/'++*)46
8/)'+,+56%;2207#$(0.7;6:A2--('+-,".%%"%&%'',=<A74973/.'%&'()$$+)$*,;5'%#")5$()*+)
%*610&3>2++%((0366*&#&)$)8:=@2-20%&"$$)$-,1)=8/+&&9/D3C>446%%'&(*+1,
@6d14c02c-1950-46f3-804c-3391a8020324 runid=ba02134f00f2059e7b2dc248113c02f76577
b101 read=6 ch=451 start_time=2019-06-27T11:09:04Z flow_cell_id=FAH59799 protoco
l_group_id=k6963 sample_id=k6963
GGTATTACTTCGTTCAGTTACGTGTGCTCGCTTCGGTTTGATCGCCTGTTAAACGACGCGCGCCACCCGAGGTGATATCT
CCCTGCGCAGCGCGATTGCCAGAACCACCGGCGCAGCAGTAGTTCTGCATGAATAACCGAGCAGGCCAGTAGAAATCGGG
```

@d76be4fb-11a9-47e7-90be-c4f15591e0d9 runid=ba02134f00f2059e7b2dc248113c02f76577





Sequencing quality assessment

- To asses quality, software uses Phred per-base quality score is used
- Is the **first quality control step** after sequencing. There should be one after every step of the analysis
- After quality assessment user can know how reliable are their datasets
- QC will determine the next filtering step
- Filtering decisions will impact directly in further analysis
- Many other steps also use this quality as variable in their algorithms





Sequencing quality assessment: Artifacts

HTS methods are bounded by their technical and theoretical limitations and sequencing errors cannot be completely eliminated (Hadigol M, Khiabanian H. 2018)

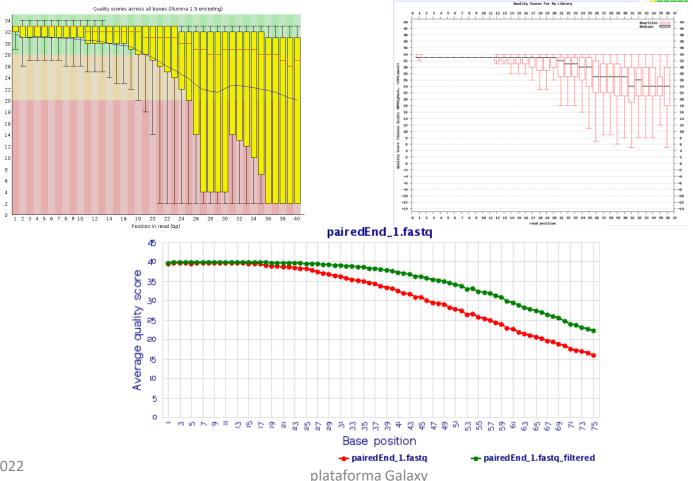
- Artifacts in library preparation
 - Remaining adapters
 - High rate of duplicates
 - GC regions bias
 - Polymerase error rate
 - DNA damage during breakdown
- Artifacts during secuencing
 - Low quality in sequence ends(Phasing: cluster loose sync)
 - Complication in certain regions:
 - Repetitions
 - Homopolymers
 - High CG content





Sequencing quality assessment

Short reads: FastQC, fastx-toolkit, NGSQCToolkit, etc...



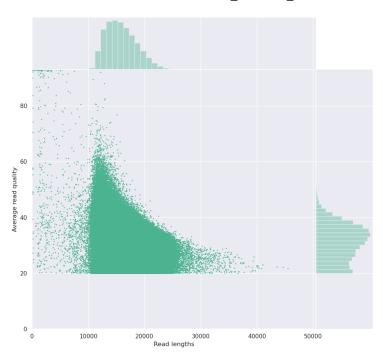


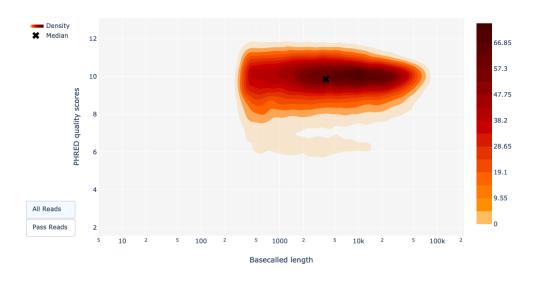


Sequencing quality assessment

Long reads: Nanoplot, PycoQC, etc.

PacBio Hifi reads for m64244_210612_174252





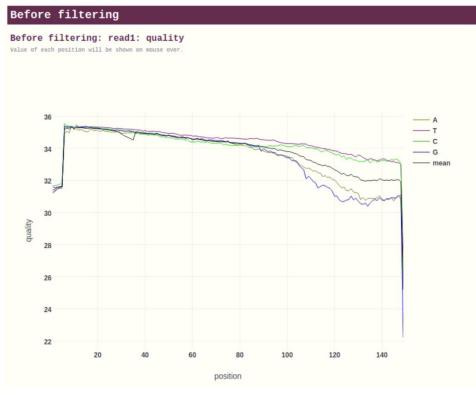




Sequencing quality assessment: fastp

Fastp









Sequencing quality assessment: FastQC



https://www.bioinformatics.babraham.ac.uk/projects/fastqc/





FastQC: Basic Statistics

- Self defined overall stats
 - Encoding: Phred33 or Phred64

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		

Basic Statistics

Measure	Value		
Filename	<pre>good_sequence_short.txt</pre>		
File type	Conventional base calls		
Encoding	Illumina 1.5		
Total Sequences	250000		
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Sequence length	40		
%GC	45		











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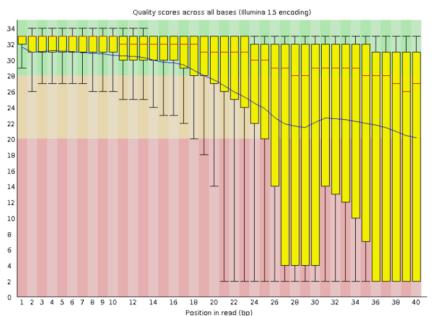




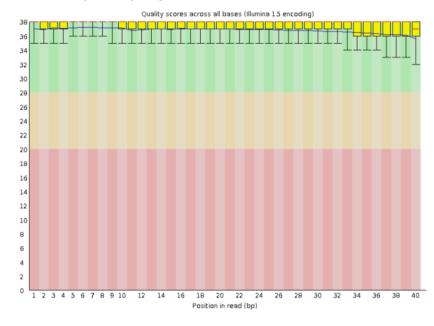
FastQC: Per base sequence quality

- Overview of the range of quality values across all bases at each position in the FastQ file
- Median, inter-quartile range (25-75%), 10-90% points, mean quality

Per base sequence quality



Per base sequence quality







FastQC: Per sequence quality score

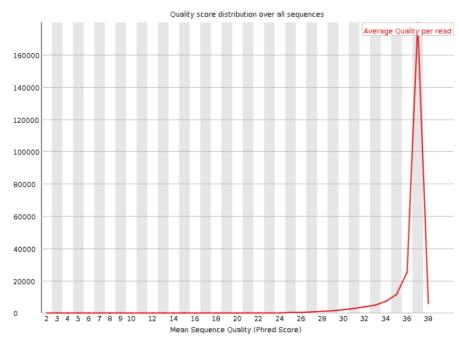
Number of sequences with the same mean quality

Per sequence quality scores Quality score distribution over all sequences Average Quality per read 40000 20000 10000

Mean Sequence Quality (Phred Score)

12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33



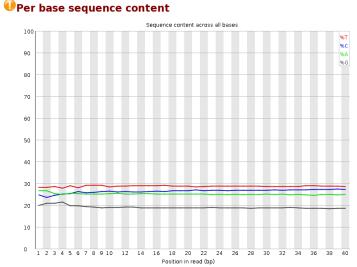


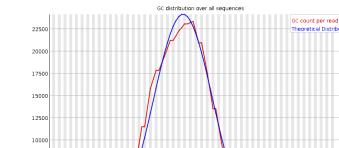




FastQC: Nucleotide related errors

- How expected nucleotide distribution deviates from expected
 - Per base sequence content
 - Per base GC content
 - Per sequence GC content
 - Per base N content





0 2 4 6 8 11 15 19 23 27 31 35 39 43 47 51 55 59 63 67 71 75 79 83 87 91 95 99

Per sequence GC content

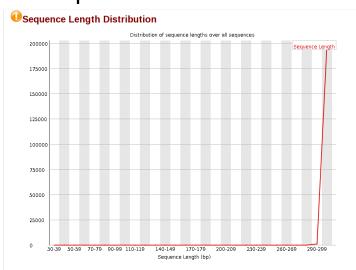
7500

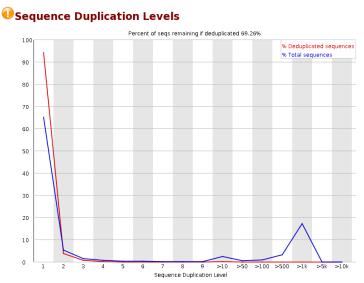




FastQC: Sequence related errors

- How expected nucleotide distribution deviates from expected
 - Sequence Length Distribution Fragments
 - Sequence Duplication Levels
 - Overrepresented sequences
 - Adapter Content





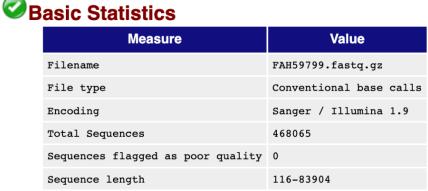




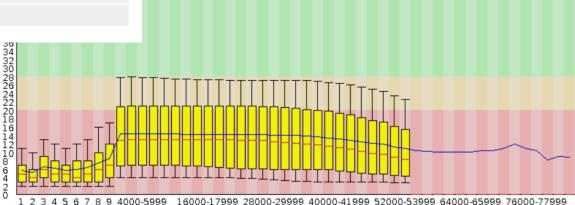
FastQC: Per base sequence quality

Nanopore

Per base sequence quality



54



Quality scores across all bases (Sanger / Illumina 1.9 encoding)

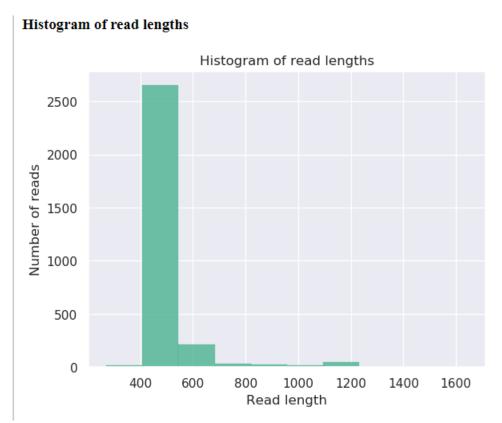
%GC





Sequencing quality assessment: NanoPlot

NanoPlot



NanoPlot report

Summary statistics

feature	
General summary	
Mean read length	537.5
Mean read quality	13.9
Median read length	516.0
Median read quality	14.0
Number of reads	3,000.0
Read length N50	517.0
Total bases	1,612,409.0
Number, percentage and megabases of reads above quality cutoffs	
>Q5	3000 (100.0%) 1.6Mt
>Q7	3000 (100.0%) 1.6Mb
>Q10	2865 (95.5%) 1.5Mb
>Q12	2461 (82.0%) 1.3Mb
>Q15	905 (30.2%) 0.5Mb
Top 5 highest mean basecall quality scores and their read lengths	
1	21.3 (504)
2	20.2 (517)
3	20.1 (509)
4	20.0 (526)
5	19.9 (530)
Top 5 longest reads and their mean basecall quality score	
1	1643 (13.5)
2	1641 (16.7)
3	1533 (12.5)
4	1427 (13.2)
5	1383 (15.0)



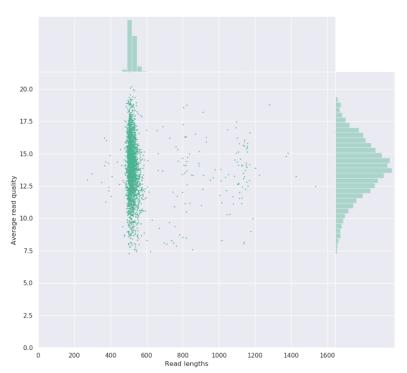


Sequencing quality assessment: NanoPlot

NanoPlot

Read lengths vs Average read quality plot using dots

Read lengths vs Average read quality plot



NanoPlot report

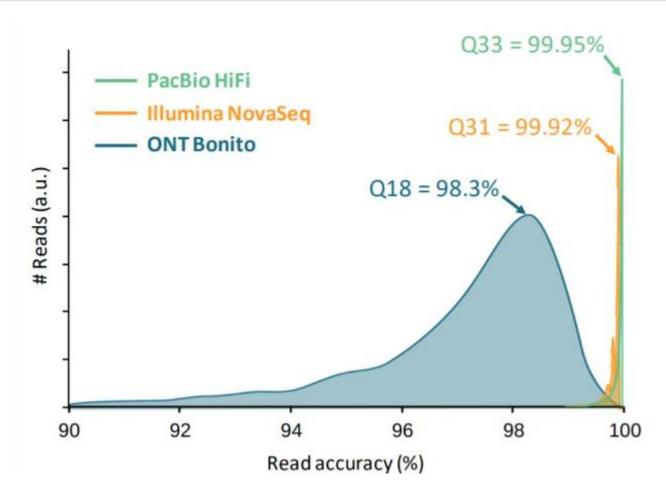
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Sequencing quality assessment

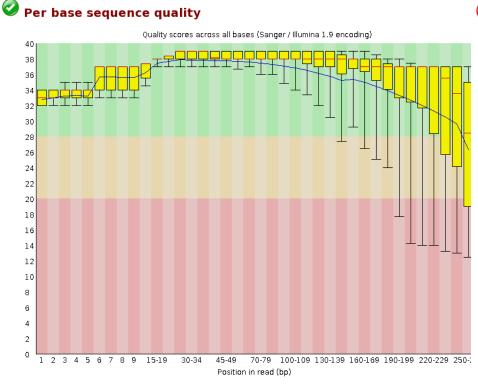




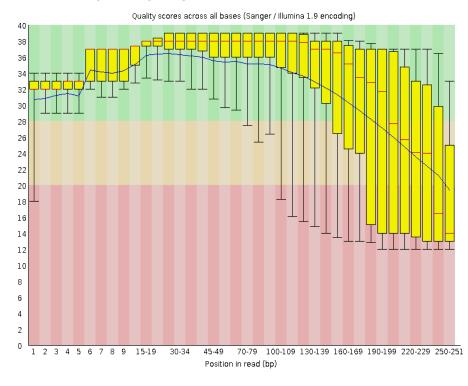


Per base sequence quality: Example

Miseq assymetry











Sequencing quality assessment: Bias

- <u>Small/micro RNA</u>: In small RNA libraries, we typically have a relatively small set of unique, short sequences. Small RNA libraries are not randomly sheared before adding sequencing adapters to their ends: all the reads for specific classes of microRNAs will be identical. It will result in:
 - Extremely biased per base sequence content
 - Extremely narrow distribution of GC content
 - Very high sequence duplication levels
 - Abundance of overrepresented sequences
 - Read-through into adapters
- <u>Amplicon:</u> Amplicon libraries are prepared by PCR amplification of a specific target. For example, the V4
 hypervariable region of the bacterial 16S rRNA gene. All reads from this type of library are expected to be
 nearly identical. It will result in:
 - Extremely biased per base sequence content
 - Extremely narrow distribution of GC content
 - Very high sequence duplication levels
 - Abundance of overrepresented sequences
- <u>Bisulfite or Methylation sequencing:</u> With Bisulfite or methylation sequencing, the majority of the cytosine (C) bases are converted to thymine (T). It will result in:
 - Biased per base sequence content
 - Biased per sequence GC content
- <u>Adapter dimer contamination</u>: Any library type may contain a very small percentage of adapter dimer (i.e. no insert) fragments. They are more likely to be found in amplicon libraries constructed entirely by PCR (by formation of PCR primer-dimers) than in DNA-Seq or RNA-Seq libraries constructed by adapter ligation. If a sufficient fraction of the library is adapter dimer it will become noticeable in the FastQC report:
 - Drop in per base sequence quality after base 60
 - Possible bi-modal distribution of per sequence quality scores
 - Distinct pattern observed in per bases sequence content up to base 60
 - Spike in per sequence GC content
 - Overrepresented sequence matching adapter
 - Adapter content > 0% starting at base 1



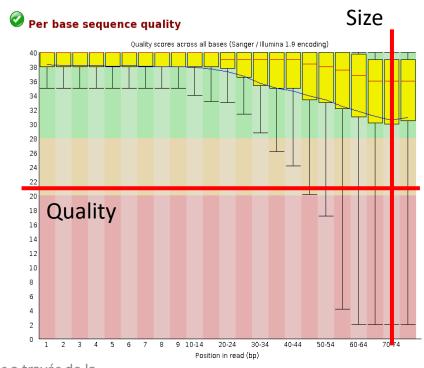


Sequence filtering

- Remove residual adapters
 - Depending on used library

Filtering parameters

- Quality filtering
 - Overall mean quality
 - Local mean quality
 - Sequence end
 - Sliding window
- Size filtering
 - Overall sequence size
 - Remaining sequence size after filtering







Sequencing quality filtering

- Illumina:
 - Fastp
 - Trimmomatic
 - Trim Galore!
- Nanopore:
 - Nanofilt
 - Porechop

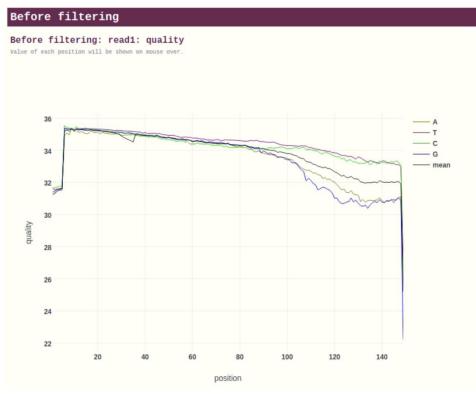




Sequencing quality filtering: fastp

Fastp











Sequence filtering: Trimmomatic with MultiQC

Trimmomatic

Trimmomatic is a flexible read trimming tool for Illumina NGS data.







Sequence filtering: NanoFilt





Sequencing quality filtering: trimmomatic

Example of quality filtering





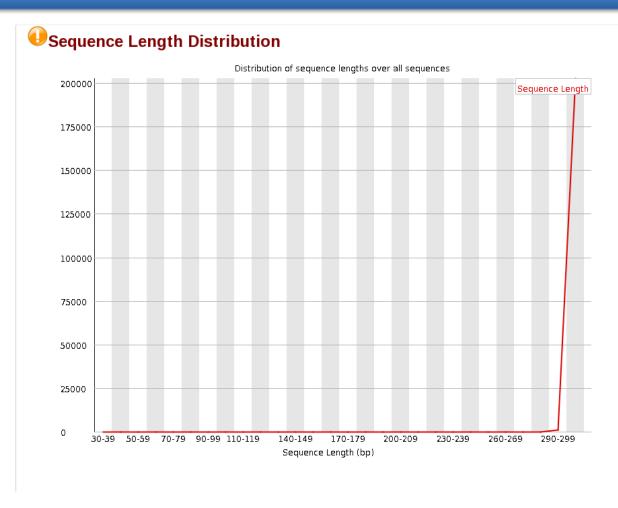


Sequence filtering: stats with MultiQC



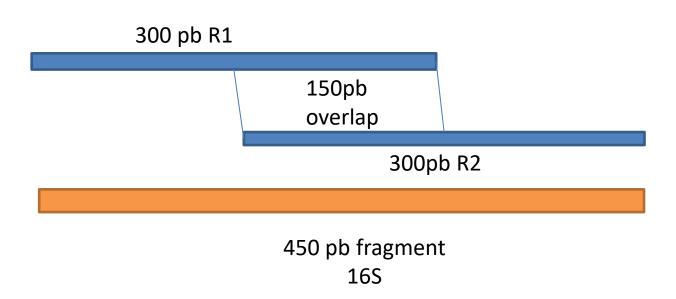






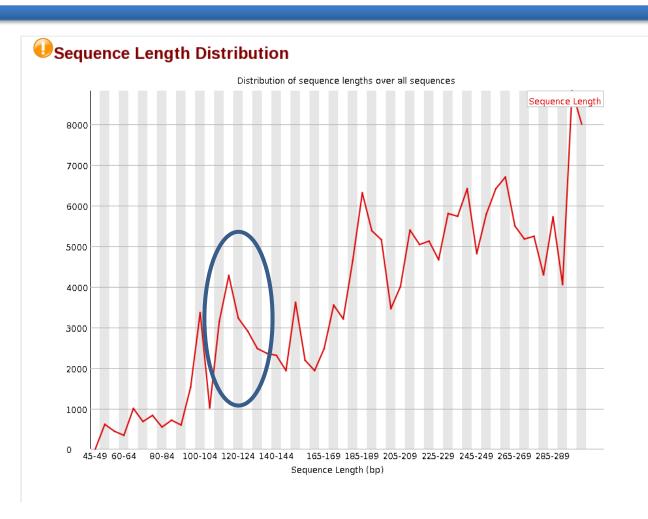






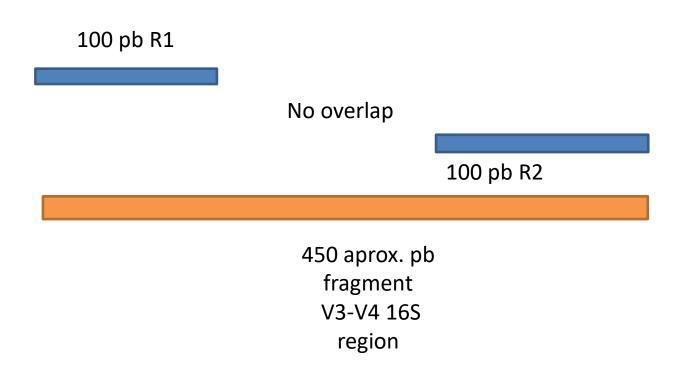
















Questions?