

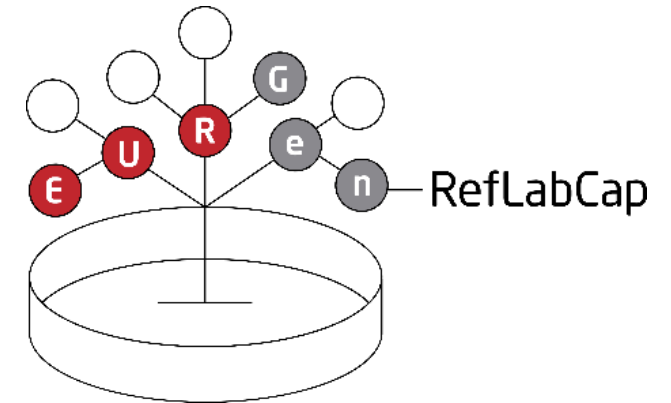
EURGen-RefLabCap

Technical training workshop # 1

First day (virtual)

Tuesday, 29 November 2022

10:00 - 12:30 CET





Please **turn off your cameras and microphones** unless you're speaking – this will help with bandwidth and maximise audibility.



Do frequently **use the chat function** to share your views, comments and challenges. Keep the chat constructive, respectful and on topic!



If you wish to make a comment for e.g. the discussion, please use the '**Raise hand**' function.

Agenda

First day (virtual) – Tuesday 29 November 2022, 10:00 - 12:30 CET

10:00 - 10:15: Introduction and agenda for the day (Ana Rita Rebelo, DTU)

10:15 - 11:00: From isolate to WGS - biochemical principles (Ana Rita Rebelo, DTU)

11:00 - 11:15: Coffee break

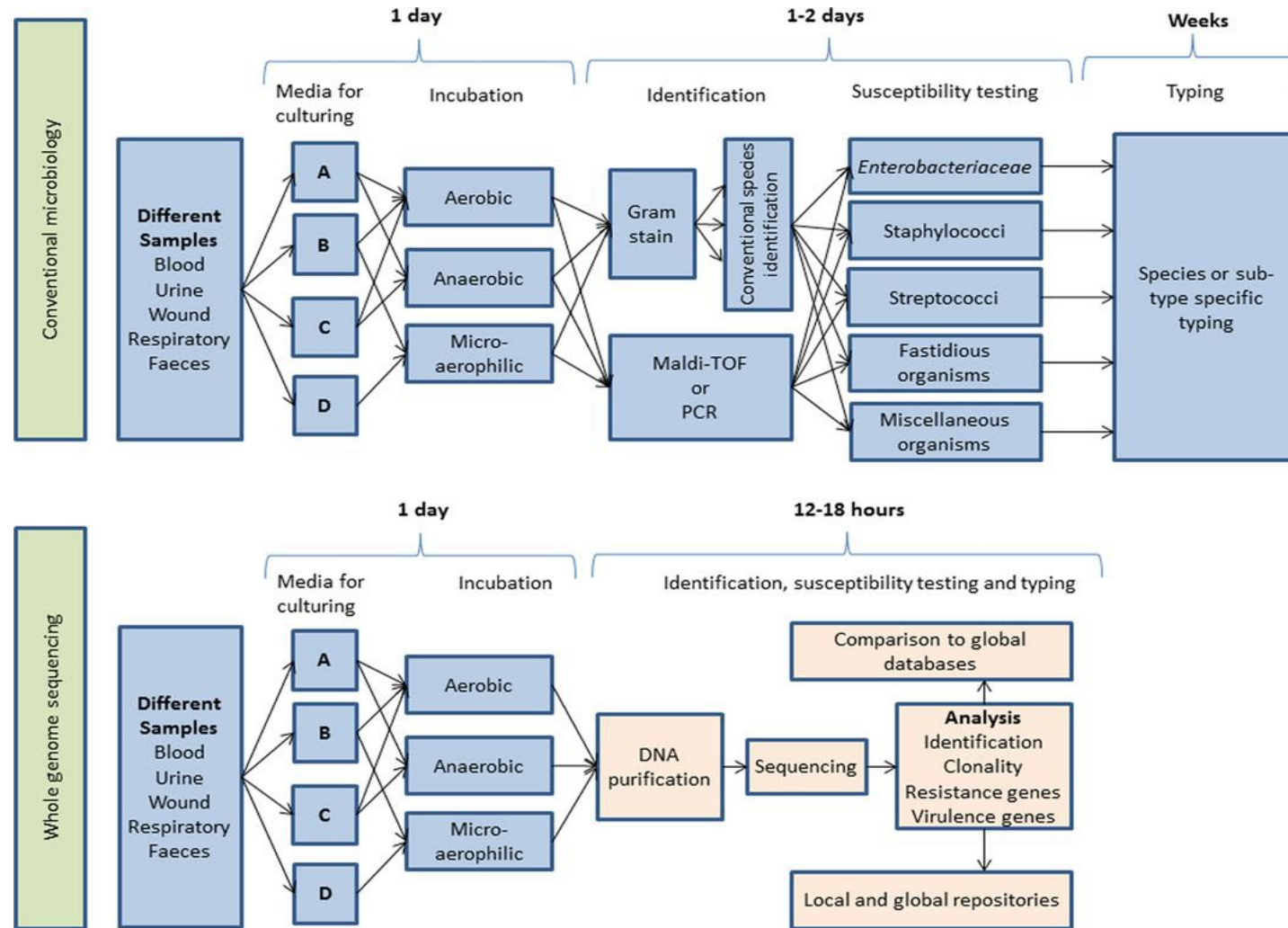
11:15 - 12:00: From isolate to WGS - WGS data and bioinformatics (Jette Sejer Kjeldgaard, DTU)

12:00 - 12:30: Quality control of WGS data (Ana Rita Rebelo, DTU)

Ana Rita Rebelo
anrire@food.dtu.dk

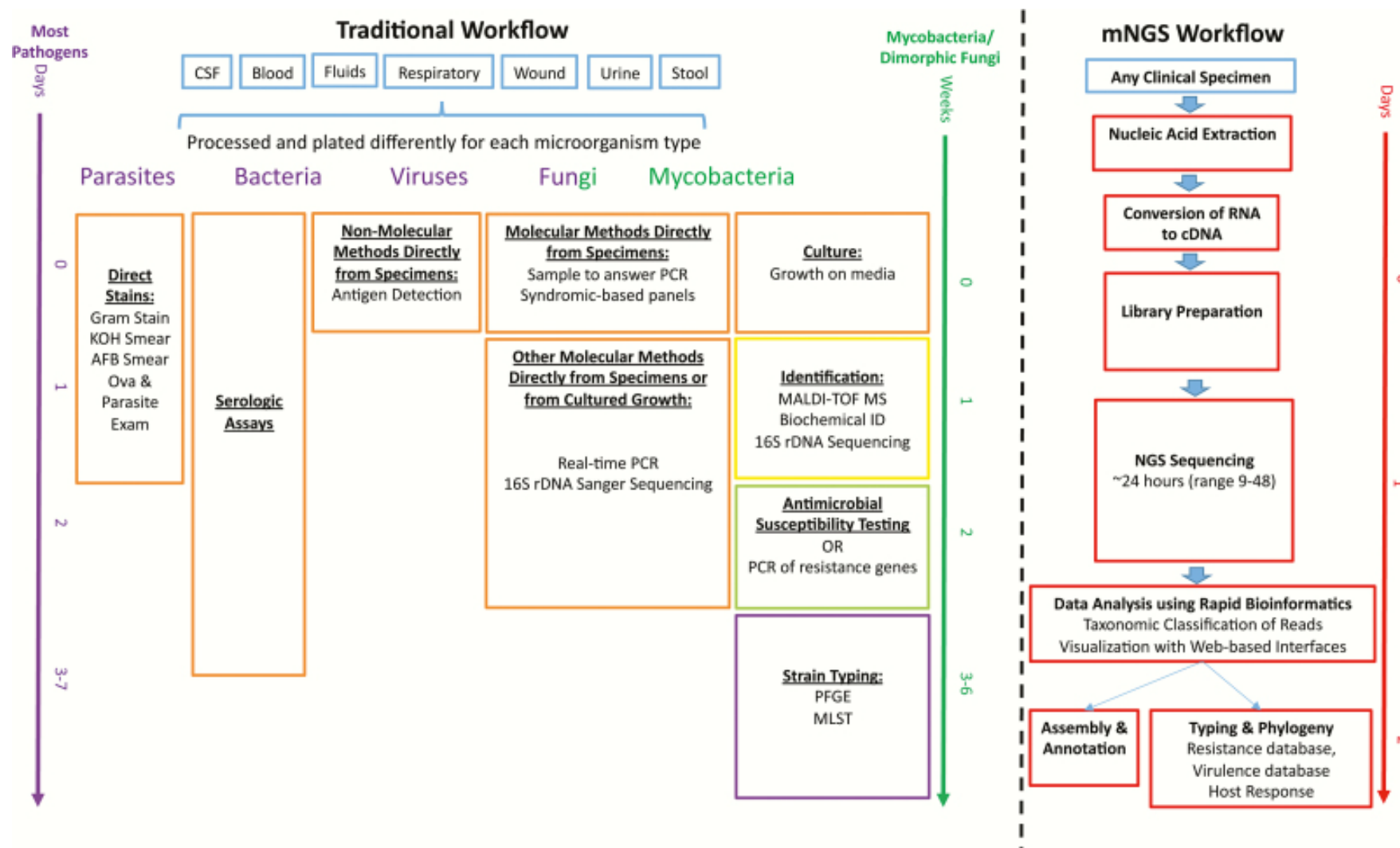
From isolate to WGS – biochemical principles

WGS vs. classical methods



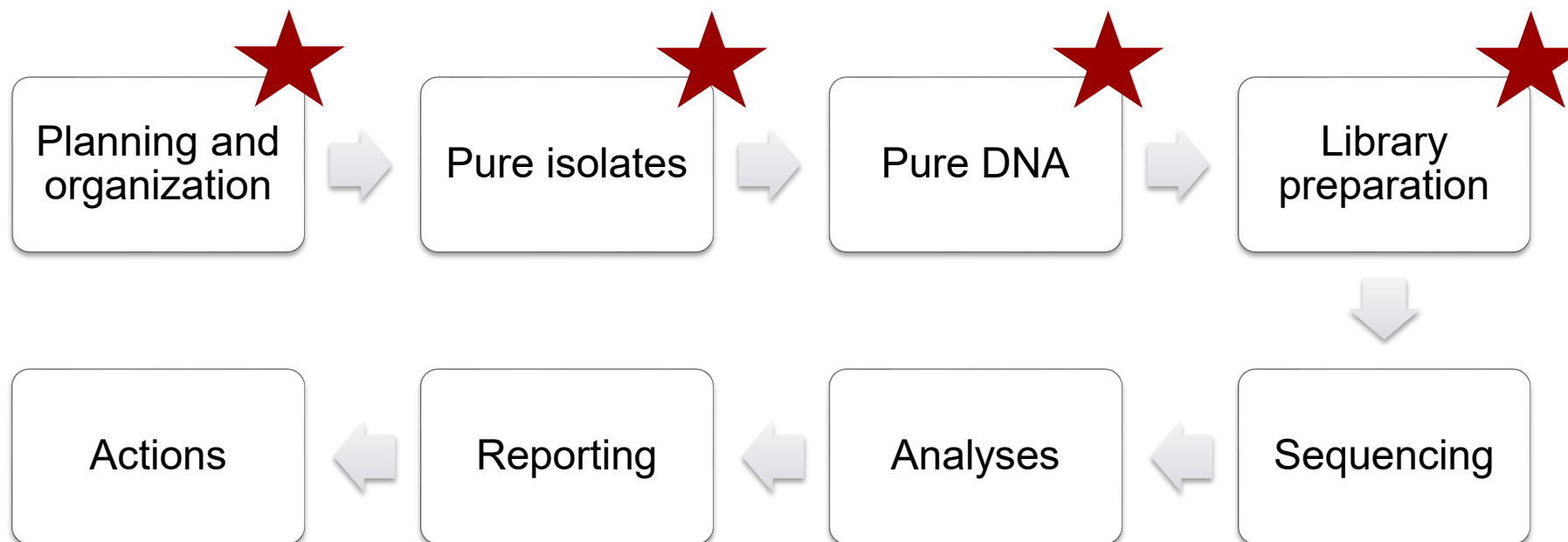
Hasman *et al.* 2013 (adapted)

WGS vs. classical methods

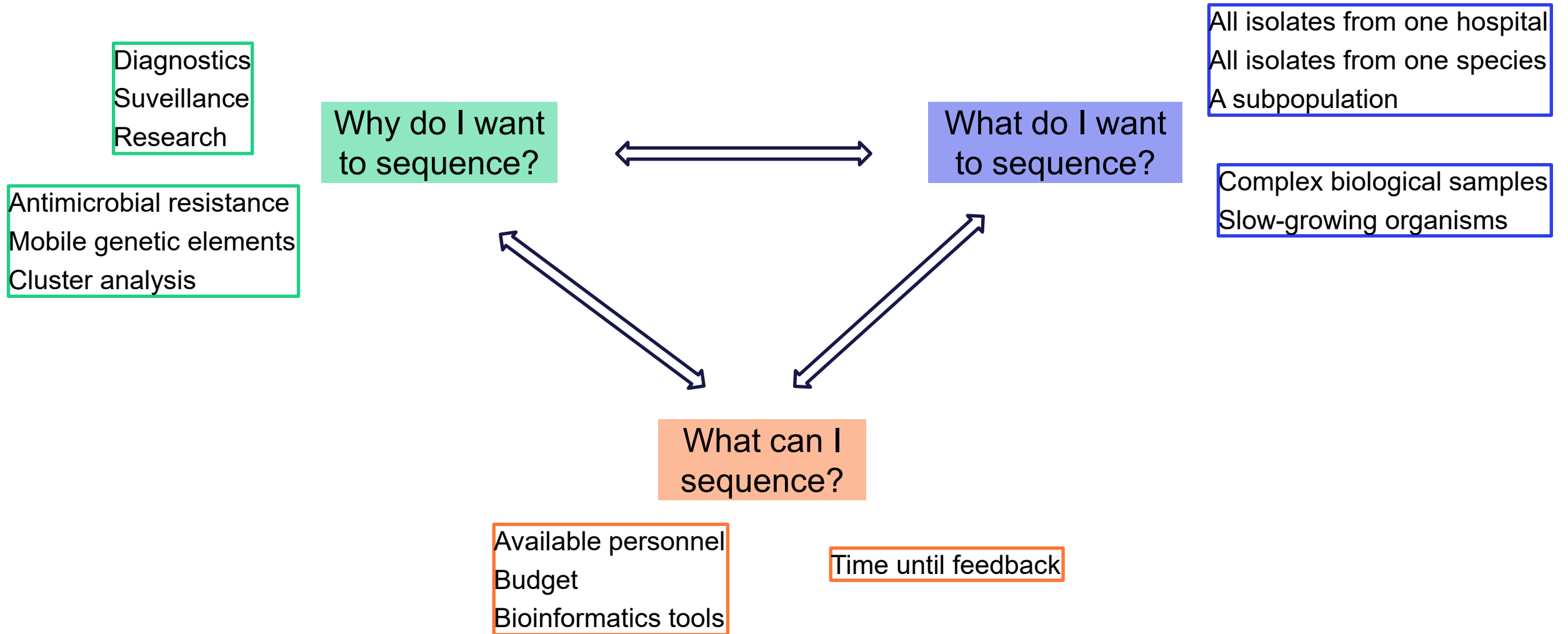


Simner *et al.* 2018

The complete workflow



Decisions...



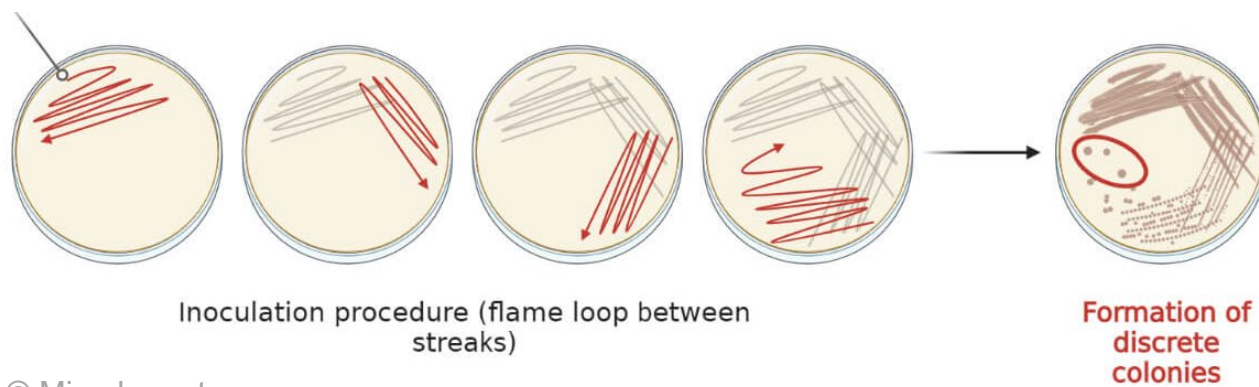
Biological samples and bacterial isolates

Cultures vs. complex samples

Selecting the adequate isolation methods

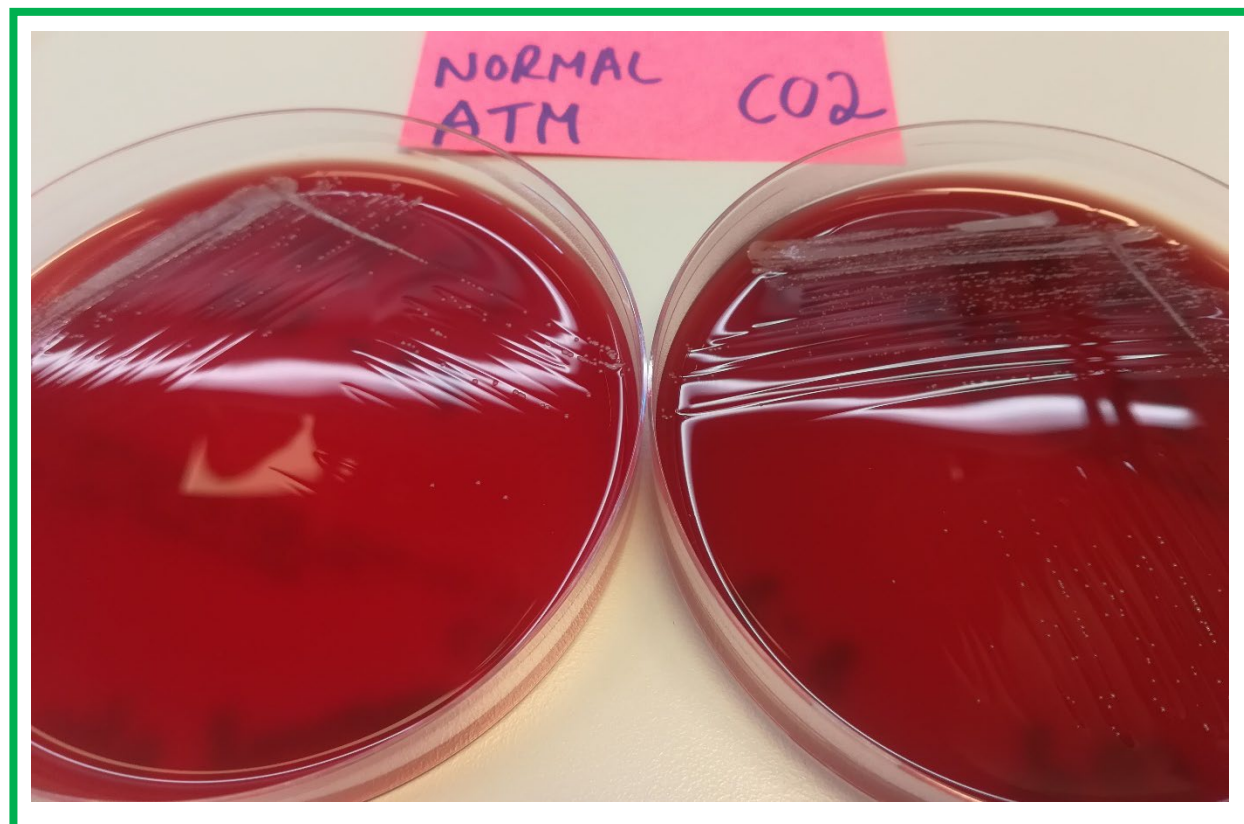
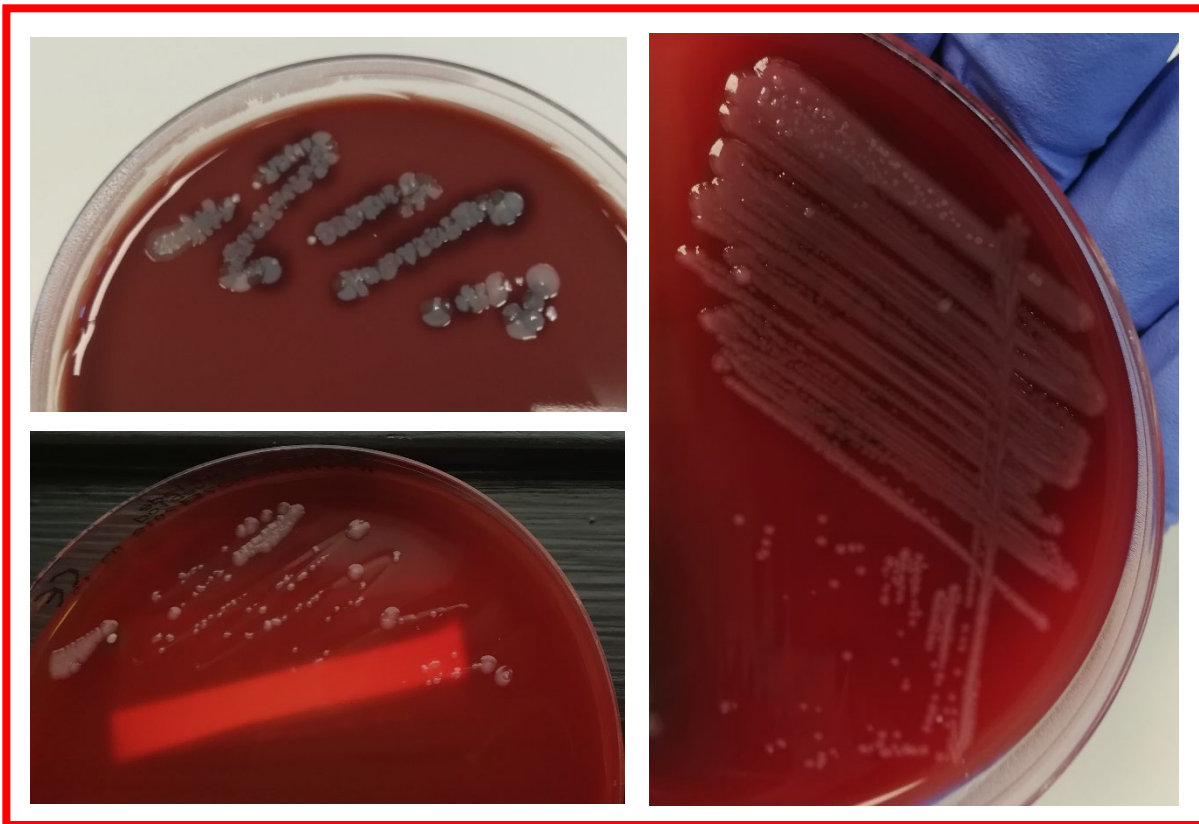
Selecting the correct isolates





© Microbe notes





In house protocols or commercial kits

Cell lysis

- Cell burst with release of intracellular components
- Enzymes, temperature, mechanical lysis, detergents, etc.

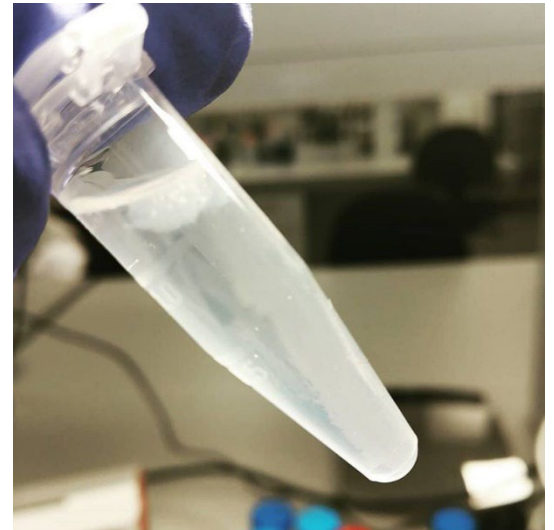
Precipitation

- Separation of the DNA and debris
- Organic solvents/alcohols and salts

Clean up

- Recovery of DNA and removal of remaining salts and reagents
- Organic solvents/alcohols

Resuspension



DNA dilution and quality control

UV 260/280 absorbance ratio values of the DNA samples should be in the interval 1.8 – 2.0

Bioanalyzer
(Agilent, Santa Clara, CA, USA)



© BioLabTech

Nanodrop spectrophotometers
(Thermo Scientific, Waltham, MA, USA)



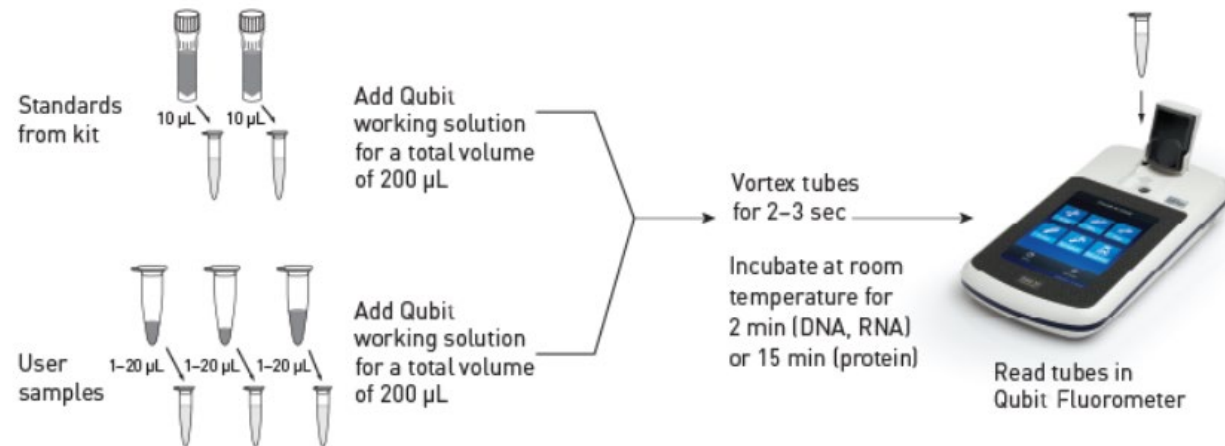
© Fischer Scientific

DNA dilution and quality control

Dilution of the extracted DNA

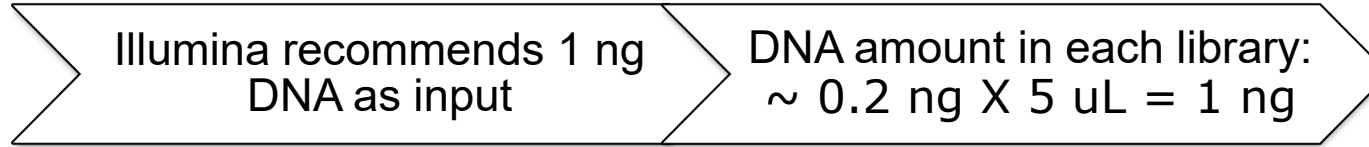
Qubit fluorometer
(Invitrogen, Carlsbad, CA, USA)

Accepted range: 0,18 – 0,28 ng/ μ L



Qubit 4 Fluorometer, Thermo Fisher Scientific, 2018

DNA dilution and quality control



Dependant on species, extraction method, operator, ...

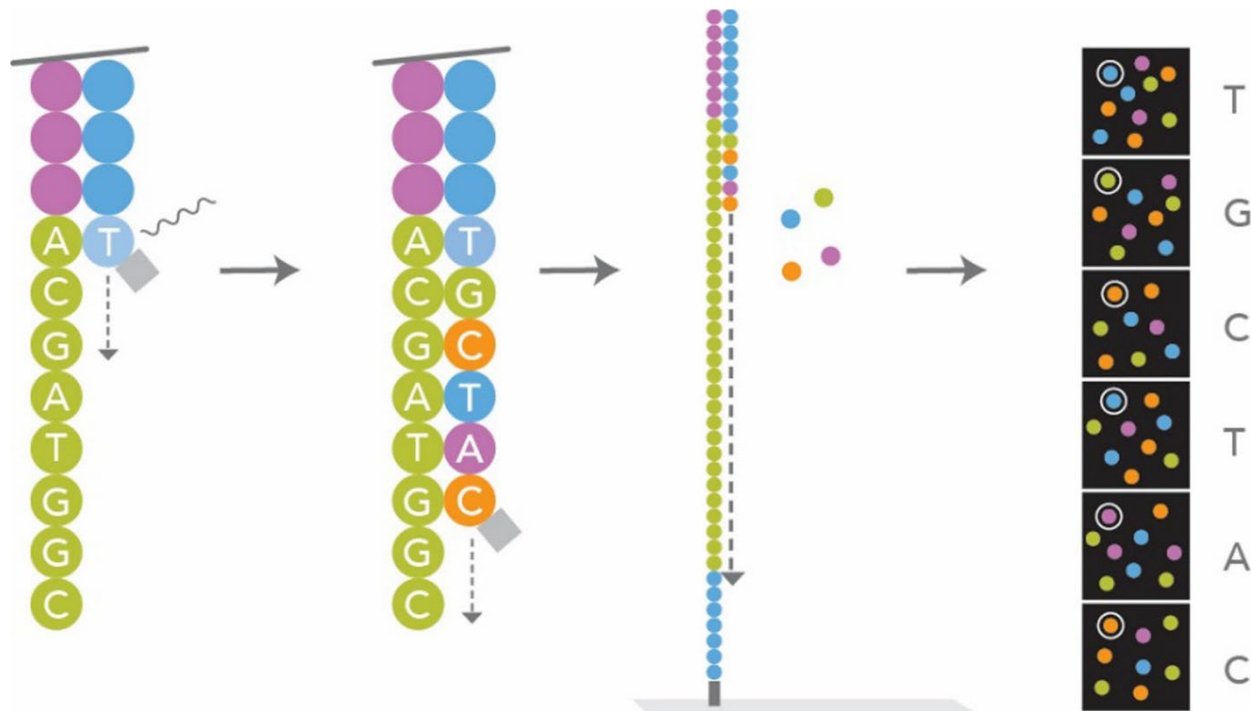
DILUTION



Sample (examples)	DNA concentration before dilution (ng/uL)	DNA concentration after dilution (ng/uL)
Isolate A	46,5	0,277
Isolate B	103	0,258

Sequencing methods

Sequencing by synthesis, sequencing by ligation, the chain termination method, pyrosequencing, ...



EFSA journal 2018;16(S1):e16086

Some concepts:

Short reads

The DNA fragments obtained by this process are shorter when compared to newer technologies

Sequencing by synthesis

The sequencing process works by using the DNA being analyzed – in the form of ssDNA - as a template to synthesize a complementary DNA strand with fluorescent nucleotides, which are then detected by the machine

Paired end

The complementary DNA is synthesized from both ends to ensure accuracy

Illumina platforms

(Illumina, Inc., San Diego, CA, USA)



iSeq 100



MiniSeq




















MiSeq Series +



NextSeq 550 Series +



NextSeq 1000 & 2000

Popular Applications & Methods	Key Application 	Key Application 	Key Application 	Key Application 	Key Application 
Large Whole-Genome Sequencing (human, plant, animal)					
Small Whole-Genome Sequencing (microbe, virus)					
Exome & Large Panel Sequencing (enrichment-based)					
Targeted Gene Sequencing (amplicon-based, gene panel)					

(...)

Run Time	9.5–19 hrs	4–24 hours	4–55 hours	12–30 hours	11–48 hours
Maximum Output	1.2 Gb	7.5 Gb	15 Gb	120 Gb	360 Gb *
Maximum Reads Per Run	4 million	25 million	25 million †	400 million	1.2 billion *
Maximum Read Length	2 × 150 bp	2 × 150 bp	2 × 300 bp	2 × 150 bp	2 × 150 bp

<https://www.illumina.com/systems/sequencing-platforms.html>



© Illumina

Library preparation

Pre-performed steps

Processes that allow us to obtain pure, single isolates and to extract and dilute their DNA for sequencing.

In here is also included the preparation of the sequencing run in BaseSpace or other platforms

Step 1

- Preparation of single isolates

Step 2

- DNA extraction

Library preparation

Laboratory procedures that allow for sequencing the extracted DNA.

They can be divided in several steps according to the biochemical processes taking place.

Step 3

- Library prep

Tagmentation
Amplification
Clean-up
Normalization
Pooling
Denaturing
Diluting

Step 4

- Loading and sequencing

Post-sequencing steps

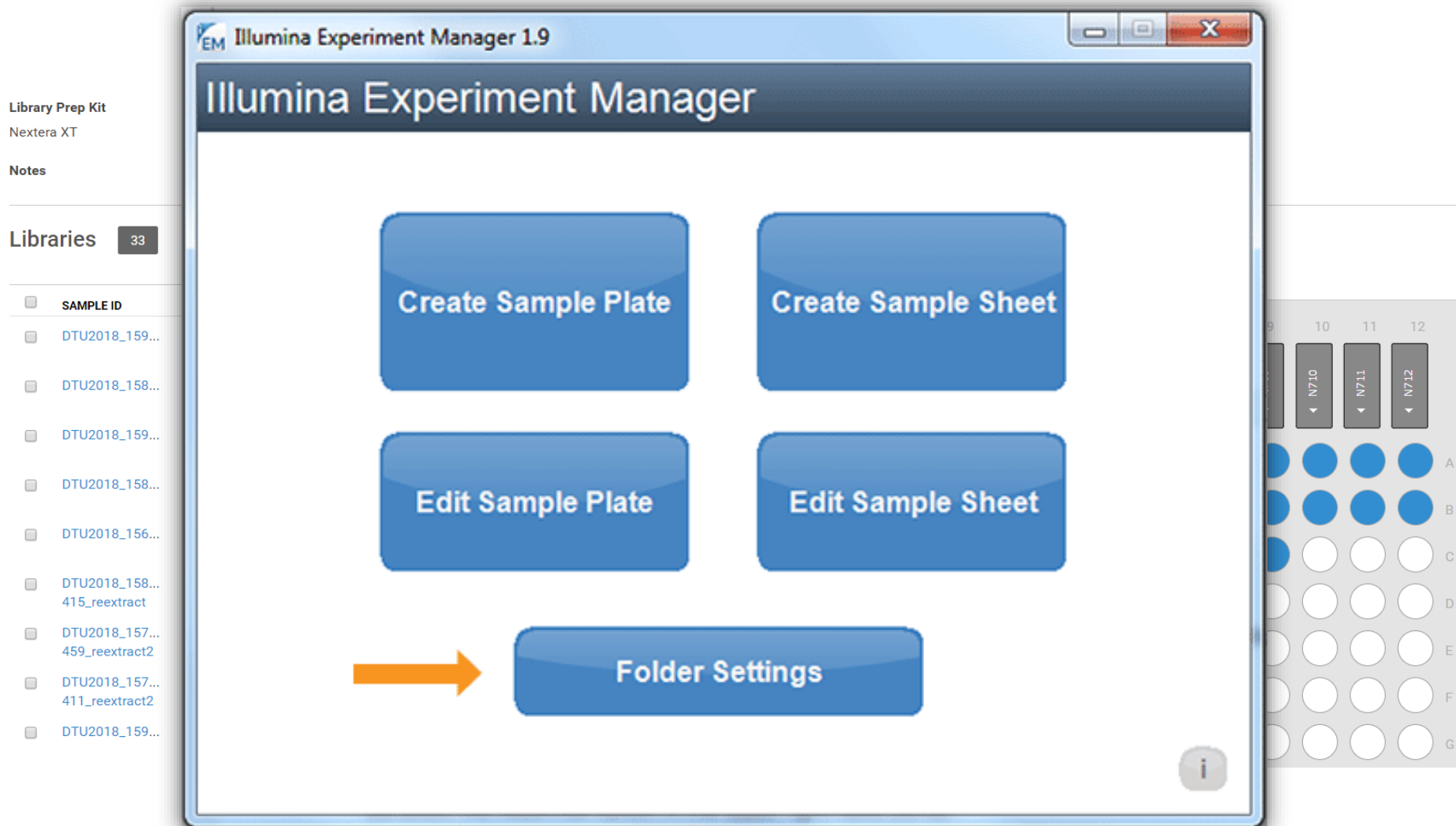
Processes that allow us control data quality and extract information from the DNA sequences

Steps 5 - ∞

- Downstream processing and analysis

Preparation of sample sheets

Organization of samples for the run



Illumina Experiment Manager 1.9

Illumina Experiment Manager

Library Prep Kit
Nextera XT

Notes

Libraries **33**

- ☐ SAMPLE ID
- ☐ DTU2018_159...
- ☐ DTU2018_158...
- ☐ DTU2018_159...
- ☐ DTU2018_158...
- ☐ DTU2018_156...
- ☐ DTU2018_158... 415_reextract
- ☐ DTU2018_157... 459_reextract2
- ☐ DTU2018_157... 411_reextract2
- ☐ DTU2018_159...

Create Sample Plate **Create Sample Sheet**

Edit Sample Plate **Edit Sample Sheet**

Folder Settings

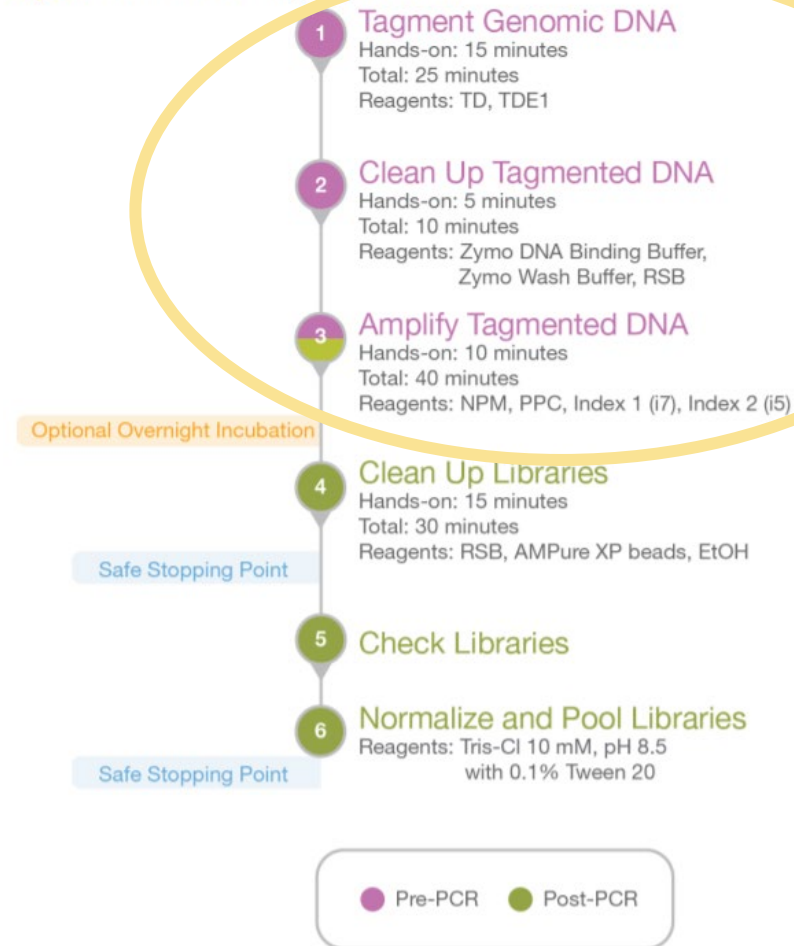
9 10 11 12

N710 N711 N712

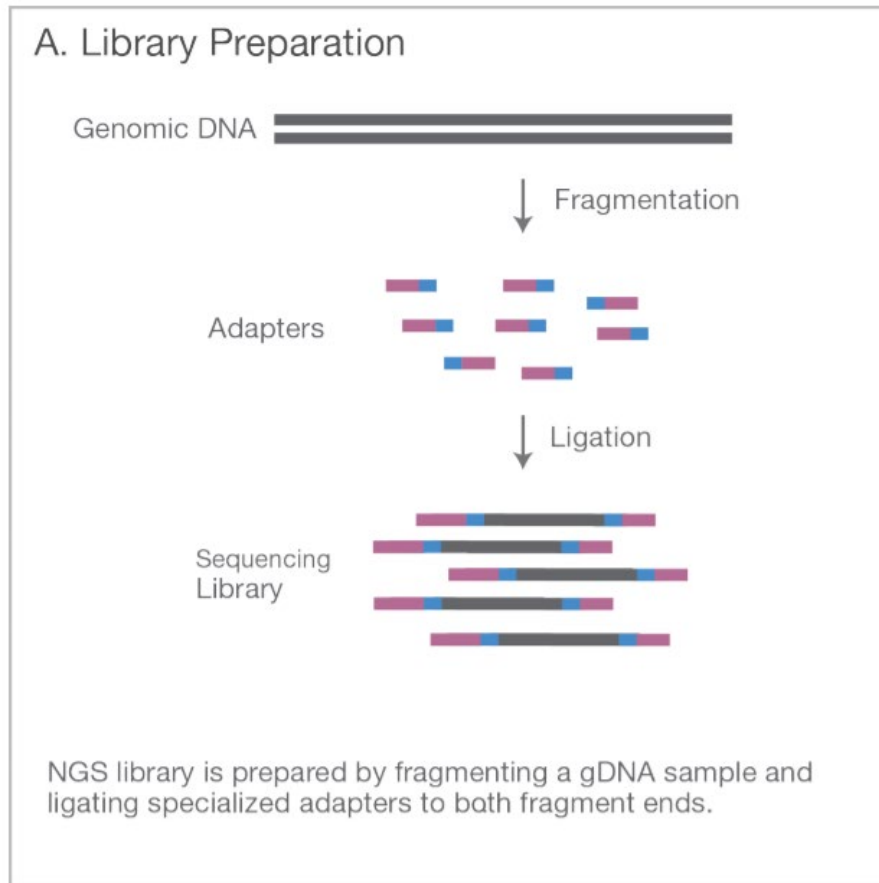
A B C D E F G

Library preparation – Tagmentation, indexing and amplification

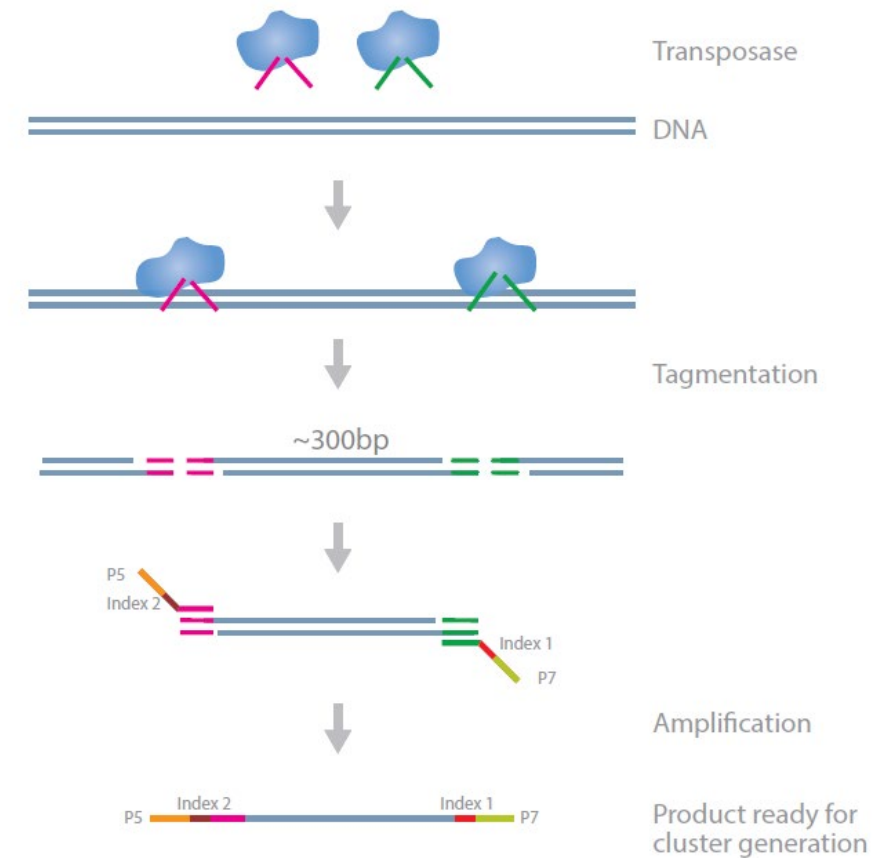
Figure 1 Nextera DNA Library Prep Workflow



Library preparation – Tagmentation, indexing and amplification



An introduction to Next-Generation Sequencing Technology, Illumina, Inc., 2017



NGS libraries construction, Biofidal, 2018

Library preparation – Tagmentation, indexing and amplification

Video: https://www.youtube.com/watch?v=womKfikWlxM&ab_channel=Illumina

Library preparation – Tagmentation, indexing and amplification

How important are these steps?

The most important!

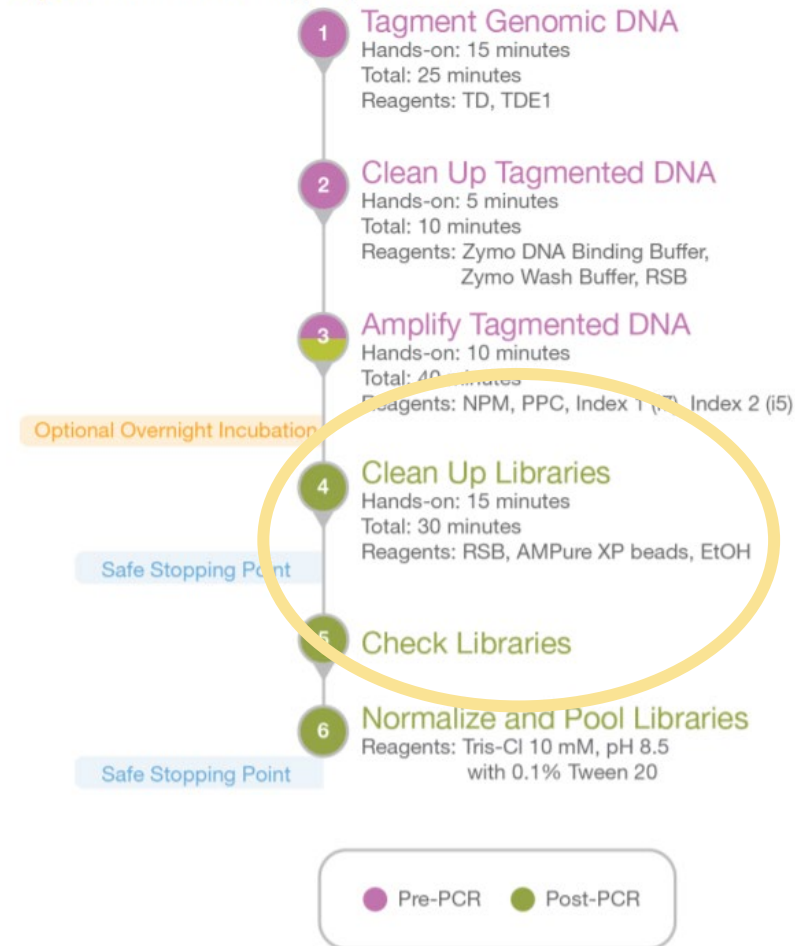
Fragmentation: If DNA is too fragmented it will be lost during clean-up

Tagmentation: Adapters provide binding sites for indexing. If tagmentation fails the DNA is useless

Indexing: Indexing marks each library independently of each other. If indexing fails the DNA is useless as it cannot be attributed to one specific library

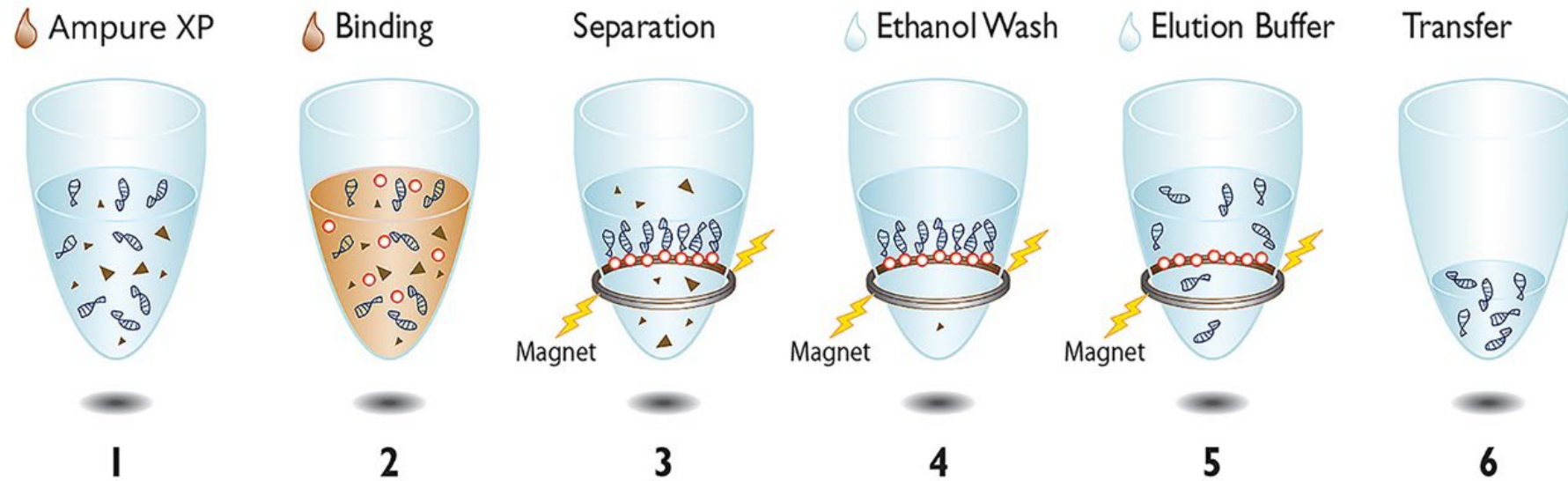
Library preparation – Clean-up

Figure 1 Nextera DNA Library Prep Workflow



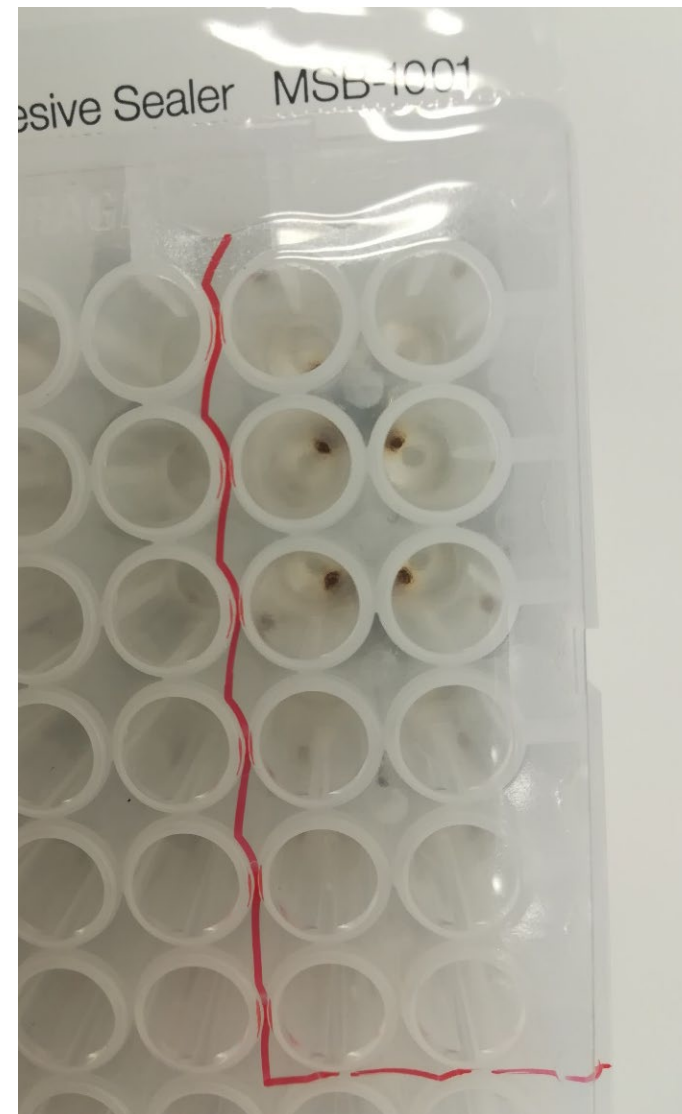
Library preparation – Clean-up

Size selection with AMPure XP beads



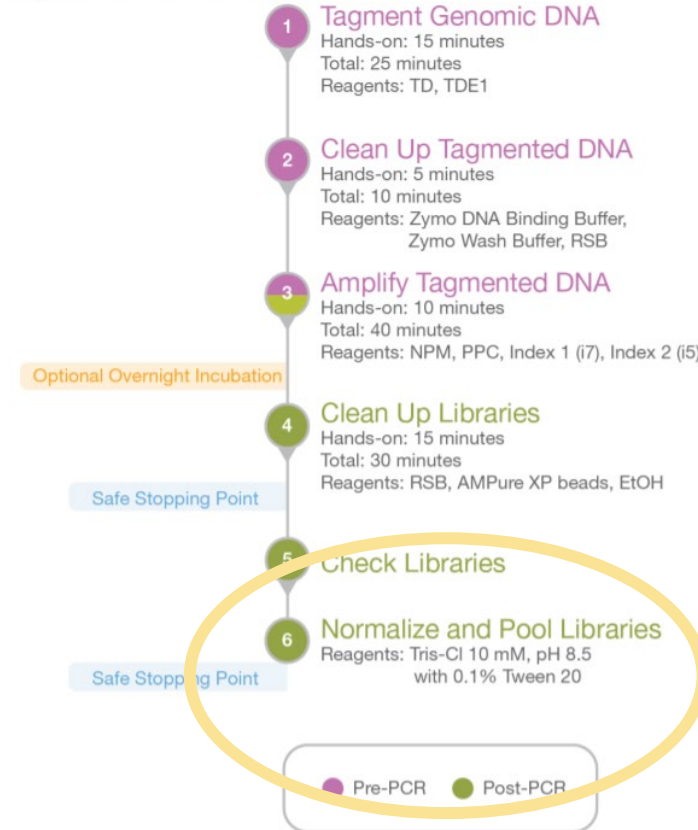
Ampure XP, Beckman Coulter Life Sciences, 2018

Library preparation – Clean-up



Library preparation - Normalization

Figure 1 Nextera DNA Library Prep Workflow



Library preparation - Normalization

Normalization: Adjusting library concentration to ensure a proper clustering and an even data distribution.

Recommended method in Nextera protocol: **bead normalization**

Somewhat similar to what happens in clean-up:

- DNA fragments are denatured and selected by using magnetic beads
- DNA fragments are bound to the surface of the beads in the same amount in all libraries
- Excess DNA is removed by washing
- The final concentration of DNA is the same in all libraries

Library preparation - Normalization

STANDARD NORMALIZATION

It's another way to adjust DNA input for the library - manually measure DNA and adjust DNA concentration using a spectrophotometric method

Illumina recommends loading pooled libraries at 6-20 pM (for MiSeq) or 1.8 pM (for NextSeq)

Individual library concentration before pooling should be known, normalized and measured

Dependant on fragment size, clean-up success, ...

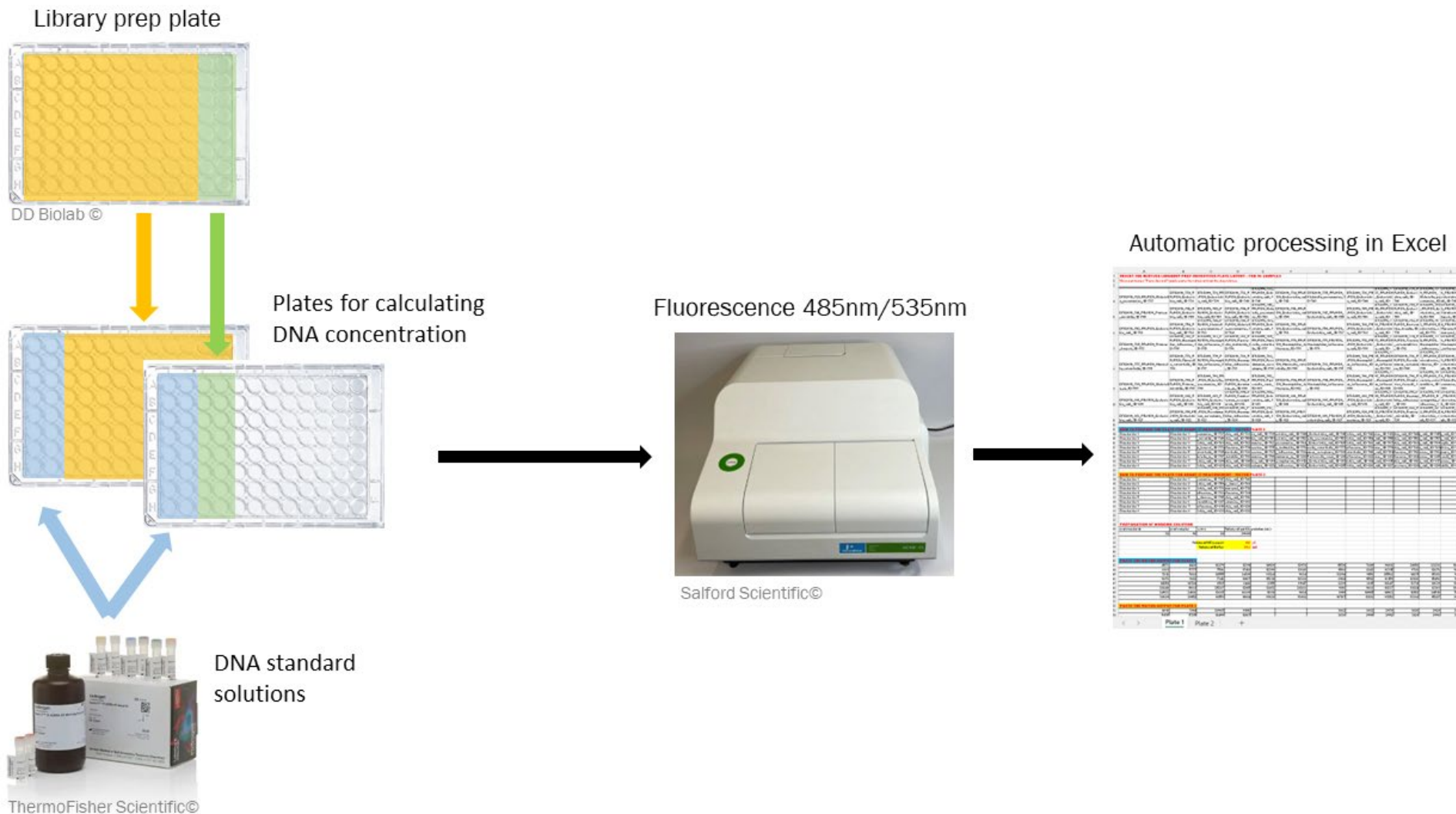
Fragment size has to be known to convert between units

Sample (examples)	Library concentration before normalization (ng/uL)	Library concentration before normalization (nM)
Isolate A	5,643	9,501
Isolate B	1,979	3,332

Final pooled library concentration: 0.71 nM

NORMALIZATION
AND POOLING

Library preparation - Normalization



Library preparation - pooling and diluting libraries

The normalization method will influence the last step of library prep: **pooling, denaturing and diluting libraries.**

Protocol B - Bead Normalization

It is the proper protocol to follow if the *library normalization* step was performed by bead normalization – but there are others

Dilution of the library followed by brief denaturation at 98 C

Protocol A - Standard Normalization ("manual")

Denaturation with NaOH followed by dilution of the library

Library preparation - pooling and diluting libraries

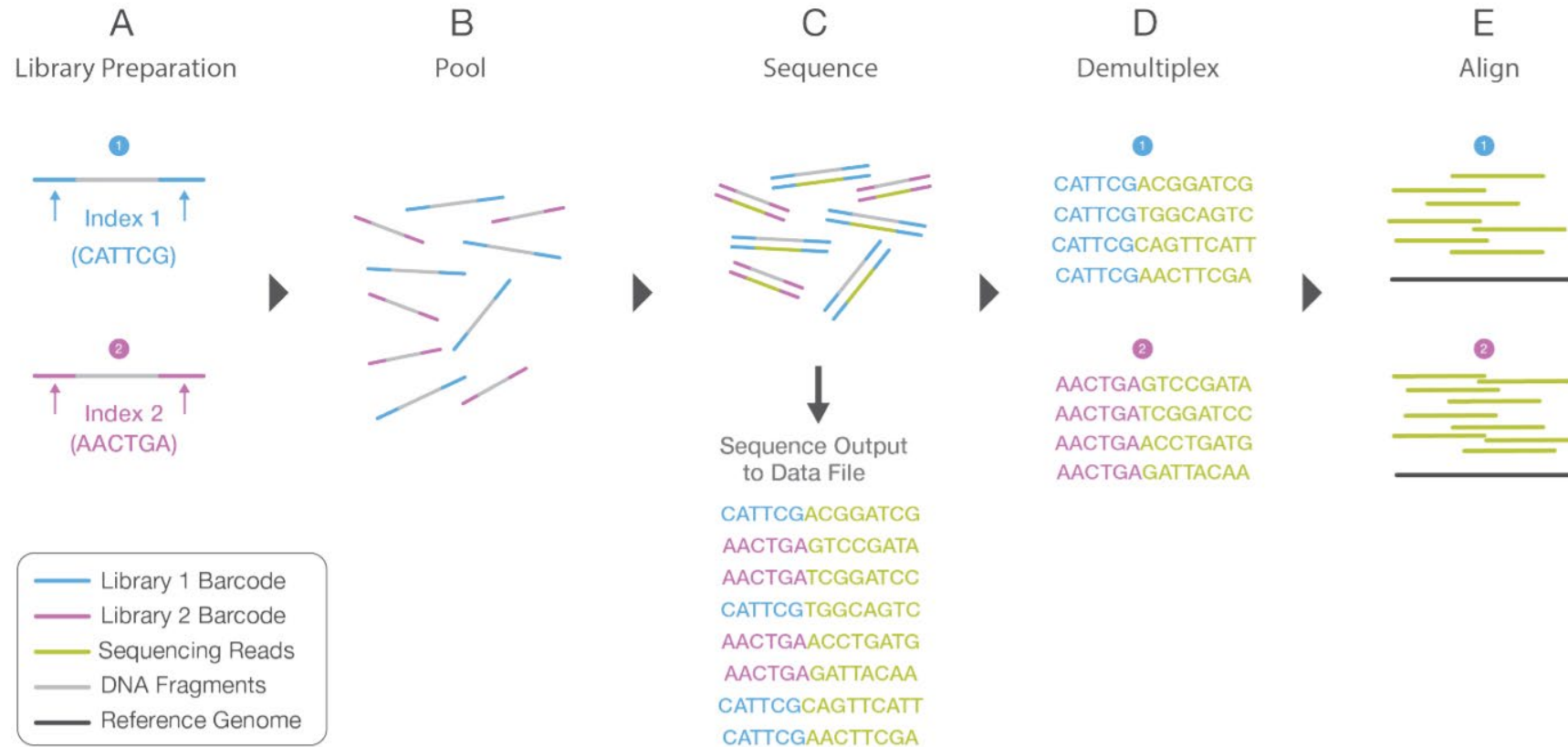
MiSeq VS. NextSeq

Main difference – final loading volume and library concentration

NextSeq: > 1 ml at 1.8 pM

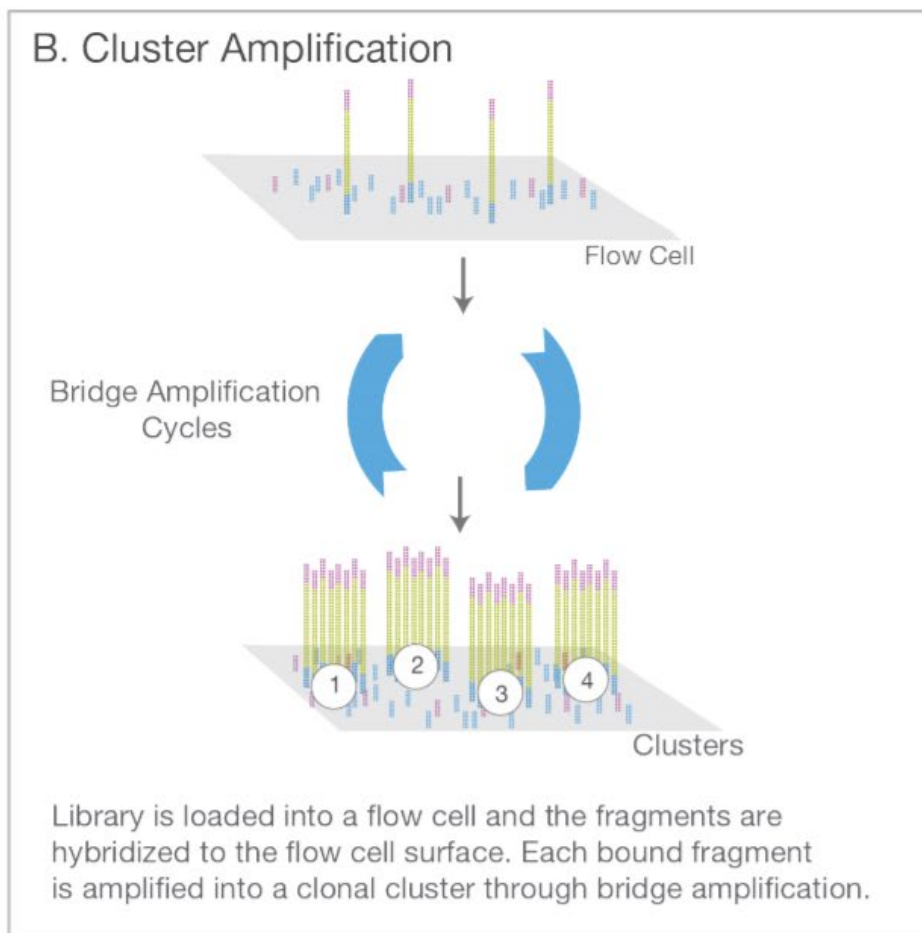
MiSeq: < 1 ml at 6 – 20 pM

Library preparation - pooling and diluting libraries

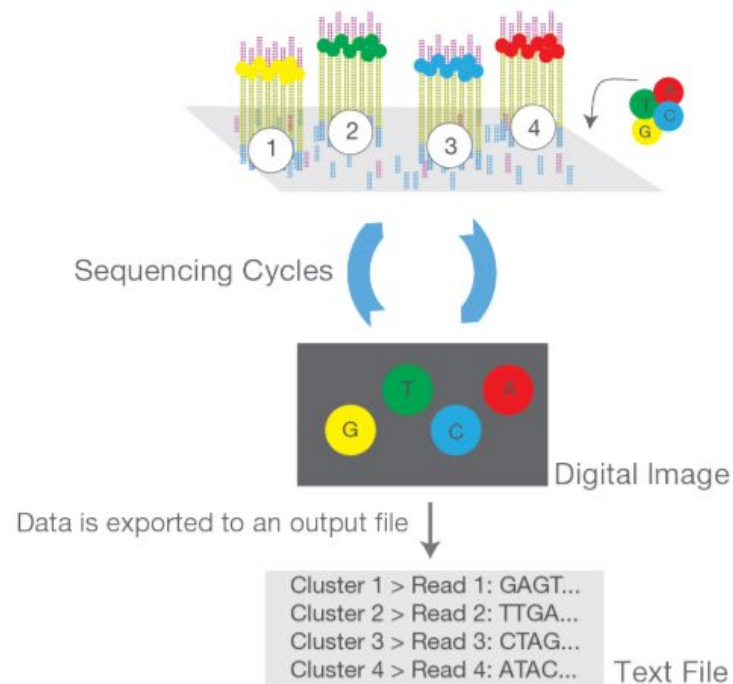


An introduction to Next-Generation Sequencing Technology, Illumina, Inc., 2017 (adapted)

Loading and sequencing



C. Sequencing



Sequencing reagents, including fluorescently labeled nucleotides, are added and the first base is incorporated. The flow cell is imaged and the emission from each cluster is recorded. The emission wavelength and intensity are used to identify the base. This cycle is repeated “n” times to create a read length of “n” bases.

An introduction to Next-Generation Sequencing Technology, Illumina, Inc., 2017

Loading and sequencing

- Video: https://www.youtube.com/watch?v=womKfikWlxM&ab_channel=Illumina

Troubleshooting

Low DNA quality	Presence of biological contaminants (improper isolate purification) Data can appear of good quality but results cannot be used for downstream analysis.
	Presence of chemical contaminants (left-overs from extraction) Can lead to undertagmentation and afterwards to underclustering.
Improper DNA dilution	Too much input DNA (>1ng) Can lead to undertagmentation and afterwards to underclustering.
	Too little input DNA (<1ng) Can lead to overtagmentation and afterwards to overclustering.
Improper clean-up	Can lead to overclustering.

Underclustering: Lower data output / Overclustering: Lower data quality

To keep in mind

Do not re-use materials – especially if you are not confident while doing it

- Can you use the same tips to distribute the indexes during amplification?
- Can you use the same tips to distribute ethanol during library clean up?

Pay attention to storage and thawing conditions

- Why?

Have a clear (even if basic) understanding of what is happening in each step

- This is the only way you will be able to modify the protocols according to your needs and resources.

Downstream processing

- Downloading data from platform
- Quality control
- Assembly
- Analysis
 - AMR genes
 - Virulence factors
 - MLST
 - Serotype
 - ...

Coffee break

Back at 11:20.



Jette Sejer Kjeldgaard

jetk@food.dtu.dk

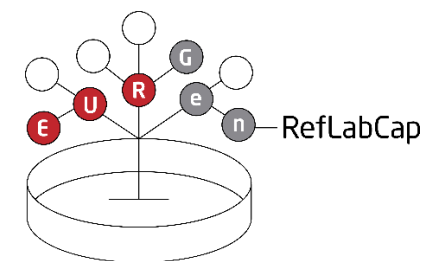
From isolate to WGS – WGS data and bionformatics

From isolate to WGS

- WGS data and bioinformatics

EURGen-RefLabCap
Technical training workshop #1
29 November 2022

Jette S. Kjeldgaard
(jetk@food.dtu.dk)

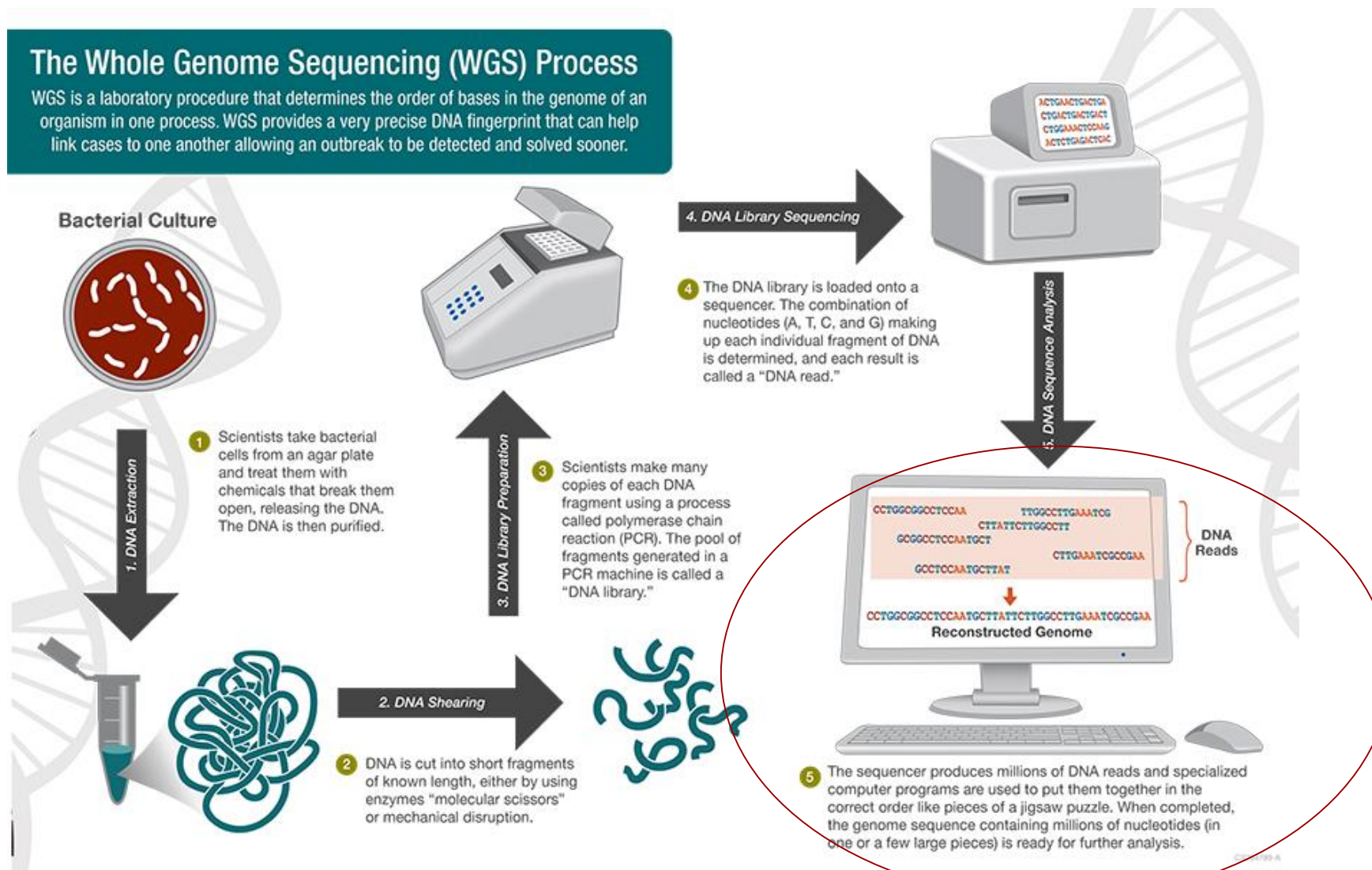


The **many** steps of sequencing

- Overview of workflow – bacterium to WGS result



WGS-based analysis of bacteria – Overview



What to do when you have a sequence?

- Illumina sequencing

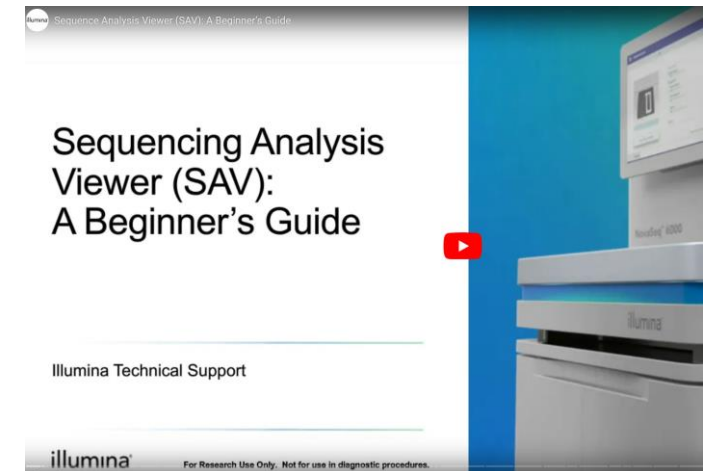
↓ →
Analysis tools

Illumina sequence viewer



Support.illumina.com

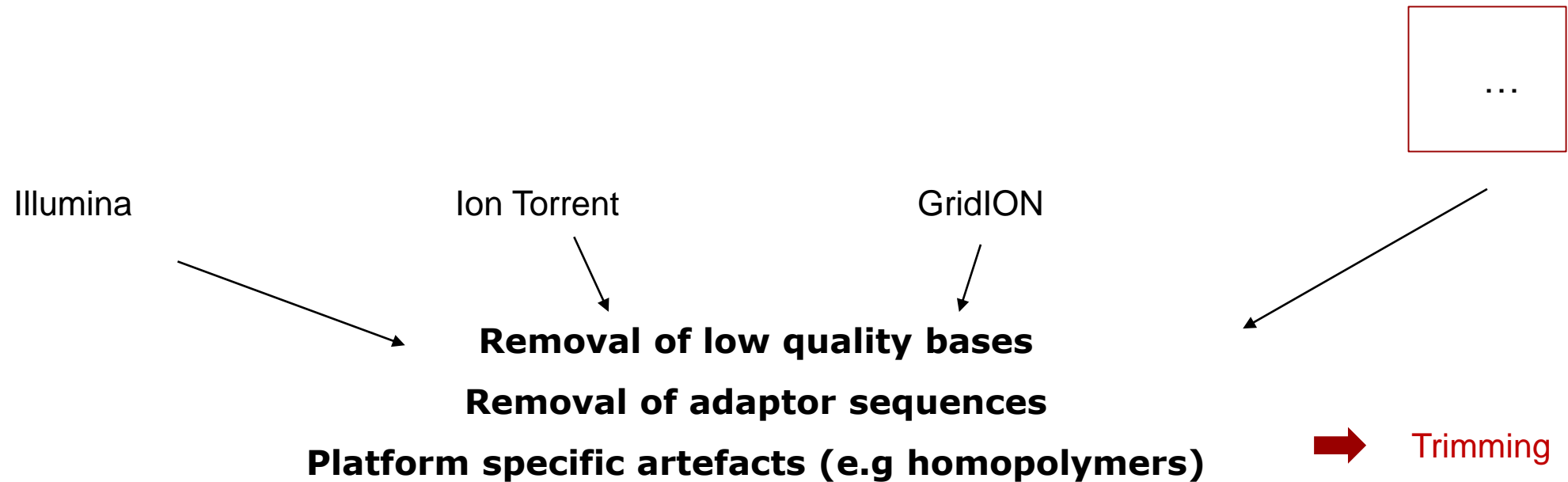
Illumina Featured Training



https://support.illumina.com/sequencing/sequencing_software/sequencing_analysis_viewer_sav.html

Sequencing in-house or outsourced;

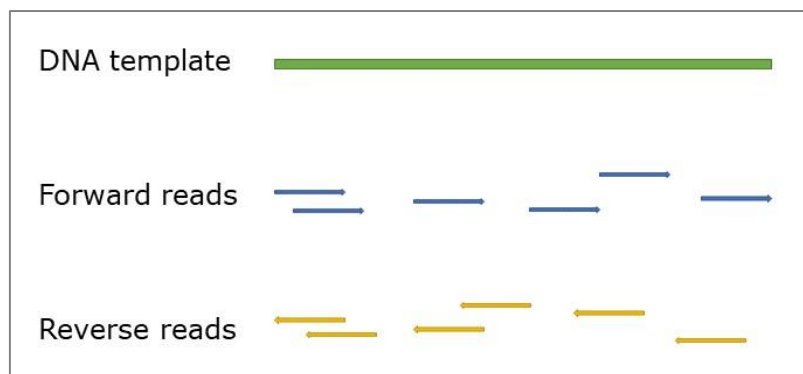
All platforms have errors and artefacts



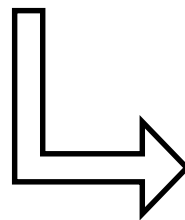
(Illumina) WGS-based analysis of bacteria

– How it works

- Illumina platforms sequence the DNA through a process called short-read paired-end massive parallel synthesis



For example, the sequencing cycles of NextSeq produce reads with a length of 150 bp, and MiSeq produces reads with lengths of 300 bp



Sequence identifier with information about the sequencing run and the cluster

```
@ML-P2-14:9:000H003HG:1:11102:17290:1073 1:N:0:TCCTGAGC+GCGATCTA  
TTTGGTAACAGCATGAATTATTCTAGCCACTAAACTCTATGAACATCTTGTGAAGGTTTCAGATAGAGCCTGAAGTACACAGAGAACAATTTCTAAAAA  
+  
AAAAAEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEEE<AEEEEEEEE
```

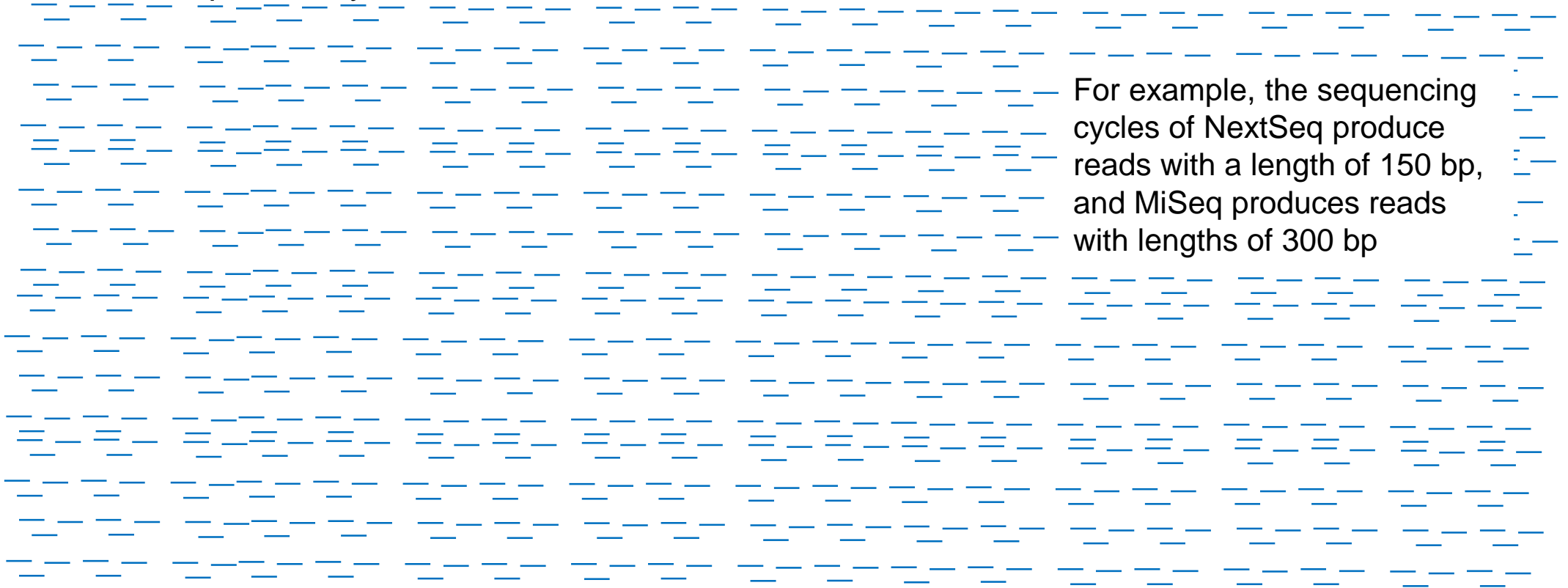
Uninterrupted
nucleotide sequence

Phred scores

(Illumina) WGS-based analysis of bacteria

– How it works

- Illumina platforms sequence the DNA through a process called short-read paired-end massive parallel synthesis



For example, the sequencing cycles of NextSeq produce reads with a length of 150 bp, and MiSeq produces reads with lengths of 300 bp

Paired-end short reads = fastq format

What is the data?

Fastq files

What is Fastq?

Fasta + quality scores

1 read, 4 lines

Fastq example:

```
@FCC0CD5ACXX:1:1101:1103:2048#ACCGT/1
ACAGTGTTTTTAGTTATTGTTTTGTTAAGTTGGGTTTTTTGTACCCAATAGCCAACAAGCCGCCTTTATGGCGGTTTTTTTGTGCCTGAAAAGTGGGCGCA
+
BP`ccceggcegihiighiifhihfd dgfhi^efgfhhhhhegiiaiiiiihiihihggeecddccccacWTT^acc[ab `][ b`^BBBBBBBB
@FCC0CD5ACXX:1:1101:1165:2058#ACGTT/1
ACGTTAGCAGAAATCGCTTTCTGTTCGTTTTCCACCTGCGACAGACGCACCGGACCACGGTTGGCGAGATCGTCGCGCAGAAATATCGGCGGCACGCTGCGAC
+
bb eeceefeqgehhda g fghhihfghighhffhifhfhcghfdhihafgdceba`a\aaccc^V]^baccaccXaaX^bbcccaac[ X]]a[aacXT
@FCC0CD5ACXX:1:1101:1135:2082#AGCGT/1
AGCGTGACAAACATTTTATTGCGCCCGGTTTTATCCAGCTTGAATGCCTGACGAAAGAAGATGATGGTGACGACGATGGAGAGAACAATCAGCACCAGATT
+
bbbeeeefggfqiighiigiiiiiiiffqifqeghiihhfefffhfhgh fhggdqegeaceeacbdbcc\^aa]``^bb]bccccbac a^bc
@FCC0CD5ACXX:1:1101:1239:2083#AGCGT/1
AGCGTCTGACTCACACAAAACGGTAACACAGTTATCCACAGAATCAGGGGATAAGGCCGGAAGAACATGTGAGCAAAAAGGCAAAGCCAGGACAAAAGG
+
bbbeeeegg ggiiaiiiiiiigifhhiighiiahhiiaiiiiiihiiaiiiiiihiigcdbbdcdcccccdcccccccccccccccccbcccaccccc
```

What is the data?

Fastq files

What is Fastq?

Fast + quality scores

Fastq example:

```
@FCC0CD5ACXX:1:1101:1103:2048#ACCGT/1
```

Header/ID

ACAGTGTTTTATGTTATTGTTTTGTTAAGTTGGGTTTTTGTACCCAATAGCCAACAAGCCGCCTTTATGGCGGTTTTTTTGTGCCTGAAAAGTGGGCGCA
+

BP`ccccggcgcgihiiighiifhifhfdgdgfhiefgfhhhhhhegiiaiiiiihihihihggeecddddcccccacWTT^acc[ab``][`b`^BBBBBBBBB

```
@FCC0CD5ACXX:1:1101:1165:2058#ACGTT/1
```

ACGTTAGCAGAATCGCTTTCTGTTCTGTTTTCCACCTGCGACAGACGCACCGGACCACGGTTGGCGAGATCGTCGCGCAGAATATCGGCGGCACGCTGCGAC
+

bb_eecceefegqgehdagfghhiihfghighhffhifhfhcgfhfdhiihafgdceba`a\aaccc^V]^baccaccXaaX^bbcccaac[X]]a[aacXT

```
@FCC0CD5ACXX:1:1101:1135:2082#AGCGT/1
```

AGCGTGACAAACATTTTATTGCGCCCGGTTTTATCCAGCTTGAATGCCTGACGAAAGAAGATGATGGTGACGACGATGGAGAGAACAATCAGCACCAGATT

+

```
bbbeeeeeffgafgajihgajigijiiiiiffaifgaghiijhhfefffhhhfgh fhggdgegeaceeacbdbcbcc\^aa]\`^bb]bcccccbac a^bc
```

```
@FCC0CD5ACXX:1:1101:1239:2083#AGCGT/1
```

AGCGTCTGACTCACACAAAACGGTAACACAGTTATCCACAGAATCAGGGGATAAGGCCGAAAGAACATGTGAGCAAAAAGGCAAAGCCAGGACAAAAGG
+

```
bbbeeeeeeggggggiiiiiiiiiiigifhhiighiiiihiiiiiiiihiiiiiiiiihiigcdbbdcdcccccdcccccccccccccccccbcccccccccc
```

What is the data?

Fastq files

What is Fastq?

Fasta + quality scores

Fastq example:

DNA sequence

@FCC0CD5ACXX:1:1101:1103:2048#ACCGT/1

ACAGTGTTTTTAGTTATTGTTTTGTAAAGTTGGGTTTTTTGTACCCAATAGCCAACAAGCCGCCTTTATGGCGGTTTTTTTGTGCCTGAAAAGTGGGCGCA

 $+$

BP`ccccggcgcegihihiighiifhihfddgfhi^efgfhhhhhhegiiaiiiiihihihihggeecddddccacWTT^acc[ab``][_b`^BBBBBBBB

@FCC0CD5ACXX:1:1101:1165:2058#ACGTT/1

ACGTTAGCAGAATCGCTTTCTGTTTCGTTTTCCACCTGCGACAGACGCACCGGACCACGGTTGGCGAGATCGTCGCGCAGAATATCGGGCGGCACGCTGCGAC

 $+$

```
bb eeceefeggehhdagfghhiihfghighhffhfhhhcgfhfdhiihafgdceba`a\aaccc^V]^baccaccXaaX^bbcccaac[ X]]a[aacXT
```

@FCC0CD5ACXX:1:1101:1135:2082#AGCGT/1

AGCGTGACAAACATTTTATTGCGCCCGGTTTTATCCAGCTTGAATGCCTGACGAAAGAAGATGATGGTGACGACGATGGAGAGAACAATCAGCACCAGATT

+

```
bbbeeeeeffggfgiighgiigiiiiiiiliffgfigfgeghiiihhfefffhfhfgh fhggdgegeaceeacbdbcbcc\^aa] `` ^bb] bcccccbac a^bc
```

@FCC0CD5ACXX:1:1101:1239:2083#AGCGT/1

AGCGTCTGACTCACACAAAAACGGTAACACAGTTATCCACAGAATCAGGGGATAAGGCCGGAAAGAACATGTGAGCAAAAAGGCCAAAGCCAGGACAAAAGG

+

```
bbbeeeeeeggggggiiiiiiiiiiigifhhiighiiiihiiiiiiiihiiiiiiiiihiigcdbbdcdcccccdcccccccccccccccccbcccccccccc
```

What is the data?

Fastq files

What is Fastq?

Fasta + quality scores

Fastq example:

```
@FCC0CD5ACXX:1:1101:1103:2048#ACCGT/1
ACAGTGTTTTTAGTTATTGTTTTGTAAAGTTGGGTTTTTGTACCCAATAGCCAACAAGCCGCCTTTATGGCGGTTTTTTTGTGCCTGAAAAGTGGGCGCA
+
_BP`ccceggcegihiighiifhfhfddgfhi^efgfhhhhhegiiaiiiiihihihiggeeccddccacWTT^acc[ab_`]'`[_b`^BBBBBBBB
@FCC0CD5ACXX:1:1101:1165:2058#ACGTT/1
ACGTTAGCAGAATCGCTTTCTGTTTCGTTTTCCACCTGCGACAGACGCACCGGACCACGGTTGGCGAGATCGTCGCGCAGAATATCGGCGGCACGCTGCGAC
+
bb_eceefeggeghdagfghhiihfghighhffhifhfhcgfhfdhiihafgdceba`a\aaccc^V]^baccaccXaaX^bbcccaac[_X])a[aacXT
@FCC0CD5ACXX:1:1101:1135:2082#AGCGT/1
AGCGTGACAAACATTTTATTGCGCCCGTTTTATCCAGCTTGAATGCCTGACGAAAGAAGATGATGGTGACGACGATGGAGAGAACAATCAGCACCAGATT
+
bbbeeeefggfgiighgiighiiaiiiiiiffgigfgeghiihhfhefffhfhfgh_fhggdgegeaceeacbdbcc\^aa]``_^bb]bccccbac_a^bc
@FCC0CD5ACXX:1:1101:1239:2083#AGCGT/1
AGCGTCTGACTCACACAAAACGGTAACACAGTTATCCACAGAATCAGGGGATAAGGCCGGAAGAACATGTGAGCAAAAAGGCAAAGCCAGGACAAAAGG
+
bbbeeeeeeeggggqiiiiiiiiiiigifhhiighiiaiiiiihiiaiiiiihiiaiiiiihiigcdbbddcdcccccdccccccccaccccccbccaccccc
```

What is the data?

Fastq files

What is Fastq?

Fasta + quality scores

Fastq example:

@FCC0CD5ACXX:1:1101:1103:2048#ACCGT/1

ACAGTGTTTTTAGTTATTGTTTTGTAAAGTTGGGTTTTTGTACCCAATAGCCAACAAGCCGCCTTTATGGCGGTTTTTTTGTGCCTGAAAAGTGGGCGCA
+

Quality scores

BP`ccceggcegihihiighiiifhfhfddgfhi^efgfhhhhhegiiaiiiiihiihihggeecdddcccacWTT^acc[ab`] `[_b`^BBBBBBBBB
@FCC0CD5ACXX:1:1101:1165:2058#ACGT/1

ACGTTAGCAGAATCGCTTTCTGTTTCGTTTTCCACCTGCGACAGACGCACCGGACCACGGTTGGCGAGATCGTCGCGCAGAATATCGGCGGCACGCTGCGAC
+

bb_eeceefeggehhdagfghhihfghhighhffhifhfhcgfhfdhiihafgdceba`a\aaccc^V]^baccaccXaaX^bbccaac[_X]]a[aacXT
@FCC0CD5ACXX:1:1101:1135:2082#AGCGT/1

AGCGTGACAAACATTTTATTGCGCCCGGTTTTATCCAGCTTGAATGCCTGACGAAAGAAGATGATGGTGACGACGATGGAGAGAACAATCAGCACCCAGATT
+

bbbeeeefggfgiighgiigiiiiiiiiffgifgeghiiahhfeffffhhhfgh_fhggdgegeaceeacbdbc\^aa]``_^bb]bccccbac_a^bc
@FCC0CD5ACXX:1:1101:1239:2083#AGCGT/1

AGCGTCTGACTCACACAAAAACGGTAACACAGTTATCCACAGAATCAGGGGATAAGGCCGGAAGAACATGTGAGCAAAAAGGCAAAGCCAGGACAAAAG
+

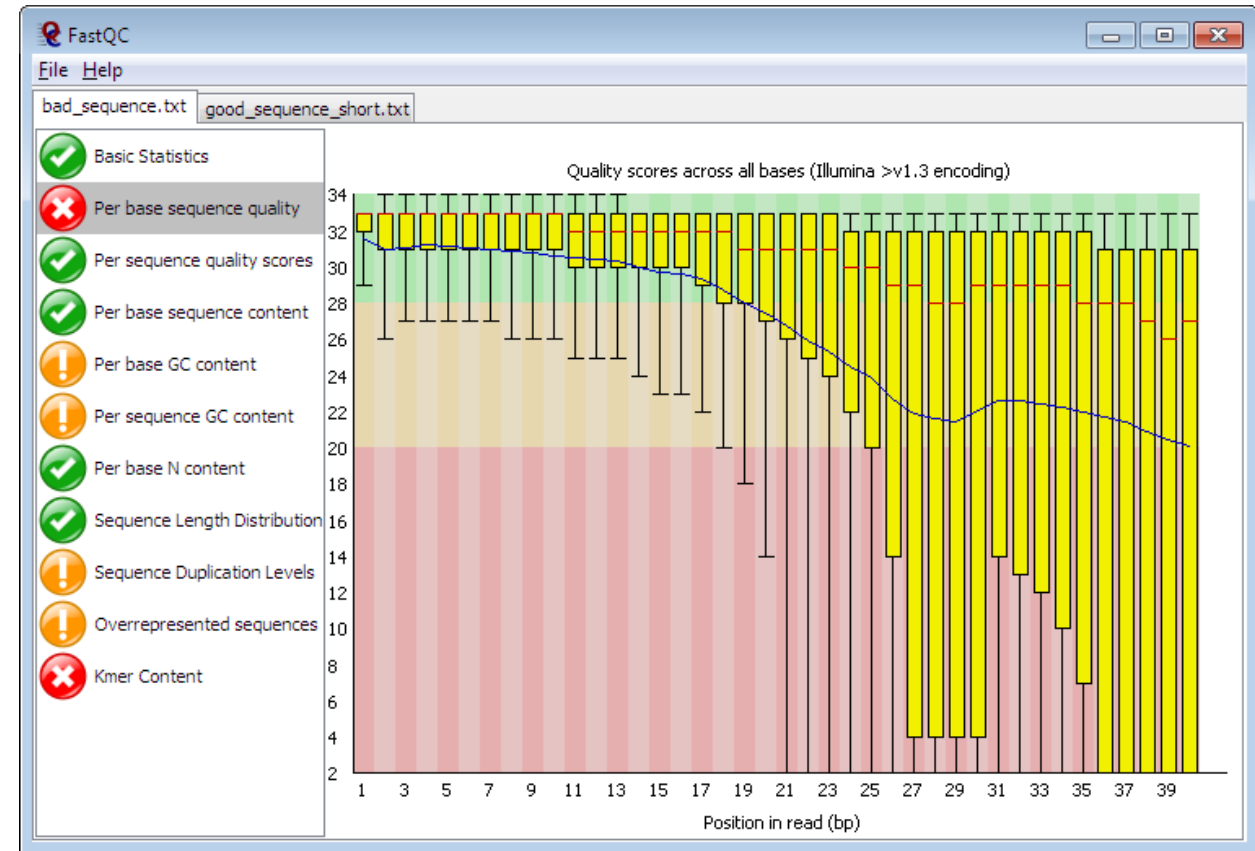
bbbeeeeeggggiiiiiiiiiiigifghhiighiiihhiiaaaaaahiiiiiiiiiihiigcdbbdedcccccdcccccccccccccccccbcccccccc

← Phred scores

Phred scores:
Show quality for
a single
nucleotides in
ASCII codes.
Should be ≥ 20

fastq format + quality scores

- Used for evaluation of quality of the sequences (QC tools)
- Used for trimming of poor quality reads
- Used for defining 'true' SNPs by SNP tools
- Trimming can be applied on raw reads
 - Also remove unpaired reads
- Trimmed reads = fastq format

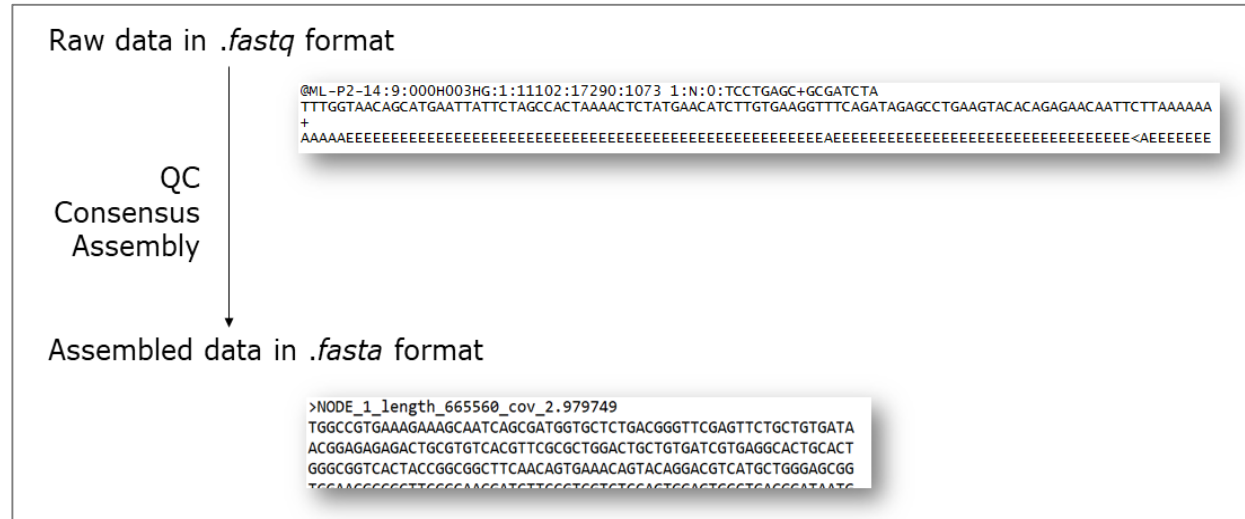


FastQC- Quality control tool (online)

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

(Illumina) WGS-based analysis of bacteria

Fastq or fasta?



File size?

Fastq:

E. coli – 100-500 MB

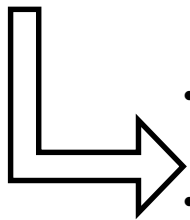
Two files: forward +
reverse file

Fasta:

E. coli – 5 MB

One file

=> different applications



- Compared with reference databases
- *What genes from the database are present in this genome?*

Fasta files

- Sequence data (only) is stored in fasta files

Header

```
>gi|218693476|ref|NC_011748.1| Escherichia coli 55989 chromosome, complete genome  
GTAAGTATTTTTCAGCTTTTCATTCTGACTGCAACGGGCAATATGTCTCTGTGTGGATTAAAAAAGAGT  
GTCTGATAGCAGCTTCTGAACTGGTTACCTGCCGTGAGTAAATTTAAATTTTATTGACTTAGGTCACTAA  
ATACTTTAACCAATATAGGCATAGCGCACAGACAGATAAAAATTACAGAGTACACAACATCCATGAAACG  
CATTAGCACCACCATTACCACCACCATCACCATTACCACAGGTAACGGTGCGGGCTGACGCGTACAGGAA  
ACACAGAAAAAAGCCCGCACCTGACAGTGCGGGGCTTTTTTTTCGACCAAAGGTAACGAGGTAACAACCAT  
GCGAGTGTTGAAGTTCGGCGGTACATCAGTGGCAAATGCAGAACGTTTTCTGCGTGTTGCCGATATTCTG  
GAAAGCAATGCCAGGCAGGGGCAGGTGGCCACCGTCCTCTCTGCCCCGCCAAAATCACCAACCACCTGG  
TGGCGATGATTGAAAAAACCATTAGCGGCCAGGATGCTTTACCCAATATCAGCGATGCCGAACGTATTTT  
TGCCGAACTTTTGACGGGACTCGCCGCCGCCAGCCGGGGTTCCCGCTGGCGCAATTGAAAACTTTCGTC  
GATCAGGAATTTGCCCAAATAAAACATGTCCTGCATGGCATTAGTTTGTGTTGGGGCAGTGCCCGGATAGCA
```

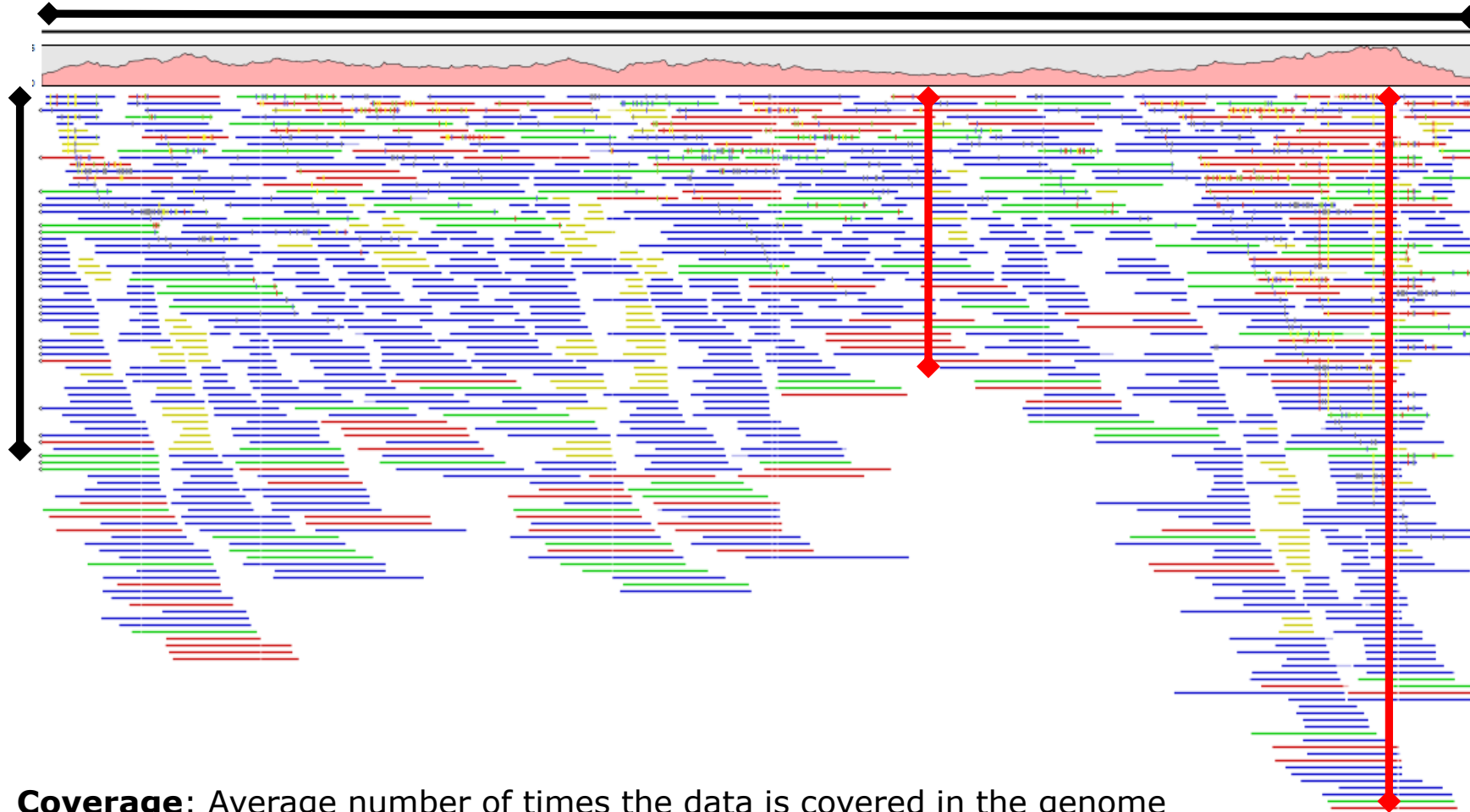
Sequence

Staph. aureus ~2.7 - 2.8 Mbp

E. coli ~ 4.5 - 5.5 Mbp

Human ~ 3.2 Gbp

Coverage vs. depth



The better coverage/higher depth

→ the better assembly

→ ...in theory and to a certain limit

Coverage: Average number of times the data is covered in the genome

Depth: Number reads that covers a particular nucleotide in each position in the genome.

Coverage

- Good coverage is important to ensure all of the genome is covered
 - High variation in local coverage over the genome
 - Low copy plasmids can be hard to find
 - (trimmed off)
- QC tools output average coverage
- Can also be calculated:

Coverage: The number of times the genome is covered by the data.

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

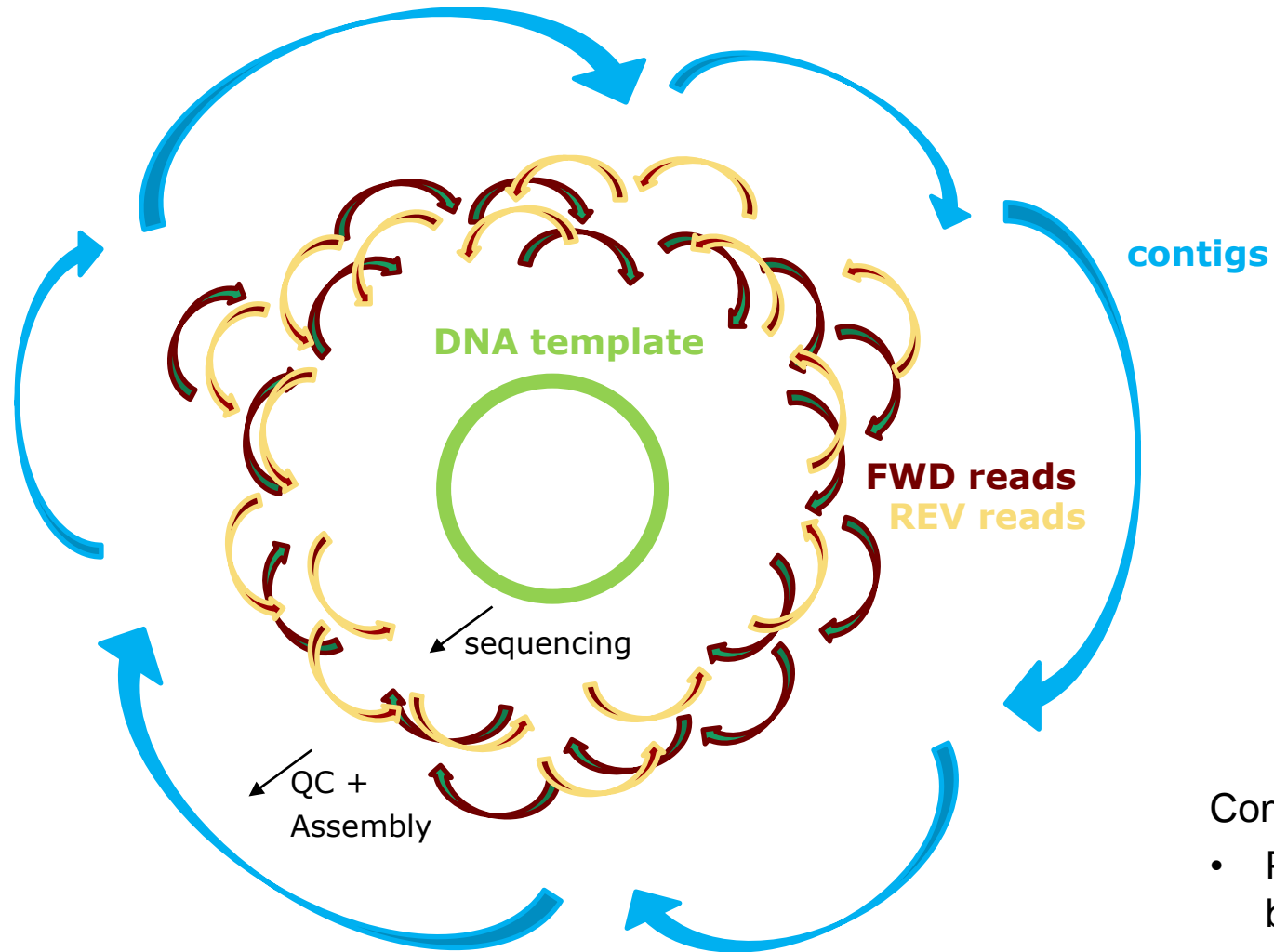
- N: Number of read
- L: Read length
- G: Genome size
(target **or** assembly)

Example:

N = 5 mill
L = 100 bp
G = 5 Mbp

$$C = 5 \cdot 100 / 5 = 100X$$

On average, 100 reads
covers each position
in the genome.



Contig size?

- Preferably 100,000's of bases
- Small contigs can (often) be removed

Assembly methods

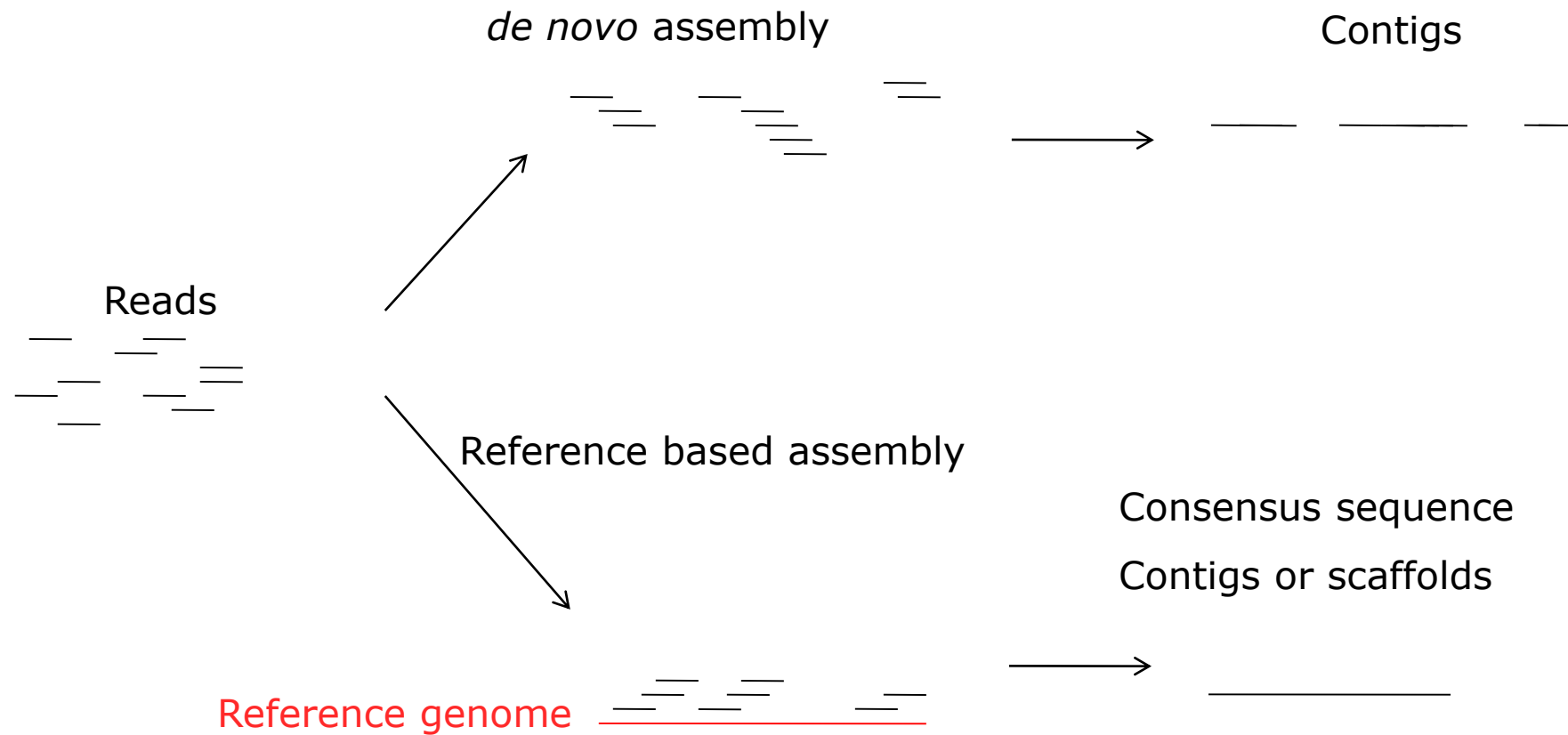
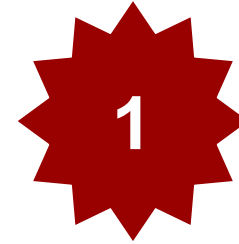
1. Mapping to reference

or

2. *de novo* assembly



Assembly methods



de novo assembly: short vs long reads

- You want as few and long contigs (or scaffolds) as possible
- Short reads are difficult to assemble to such long sequences
- Long reads have more errors
- Current state of the art is to use both short and long reads (hybrid assembly)
- Various software for different applications

Fasta vs fastq

- Fastq format
 - Much more data
 - Quality scores
 - Low depth reads included
- Better resolution for analyses
 - Quality parameters on SNPs
- Low quality reads can affect results
- Trimming is a benefit
- Fasta format
 - Smaller files to handle
 - No quality scores
 - Only consensus sequence
- Context of genes and up/downstream bases
- Fasta might be adequate for most processes

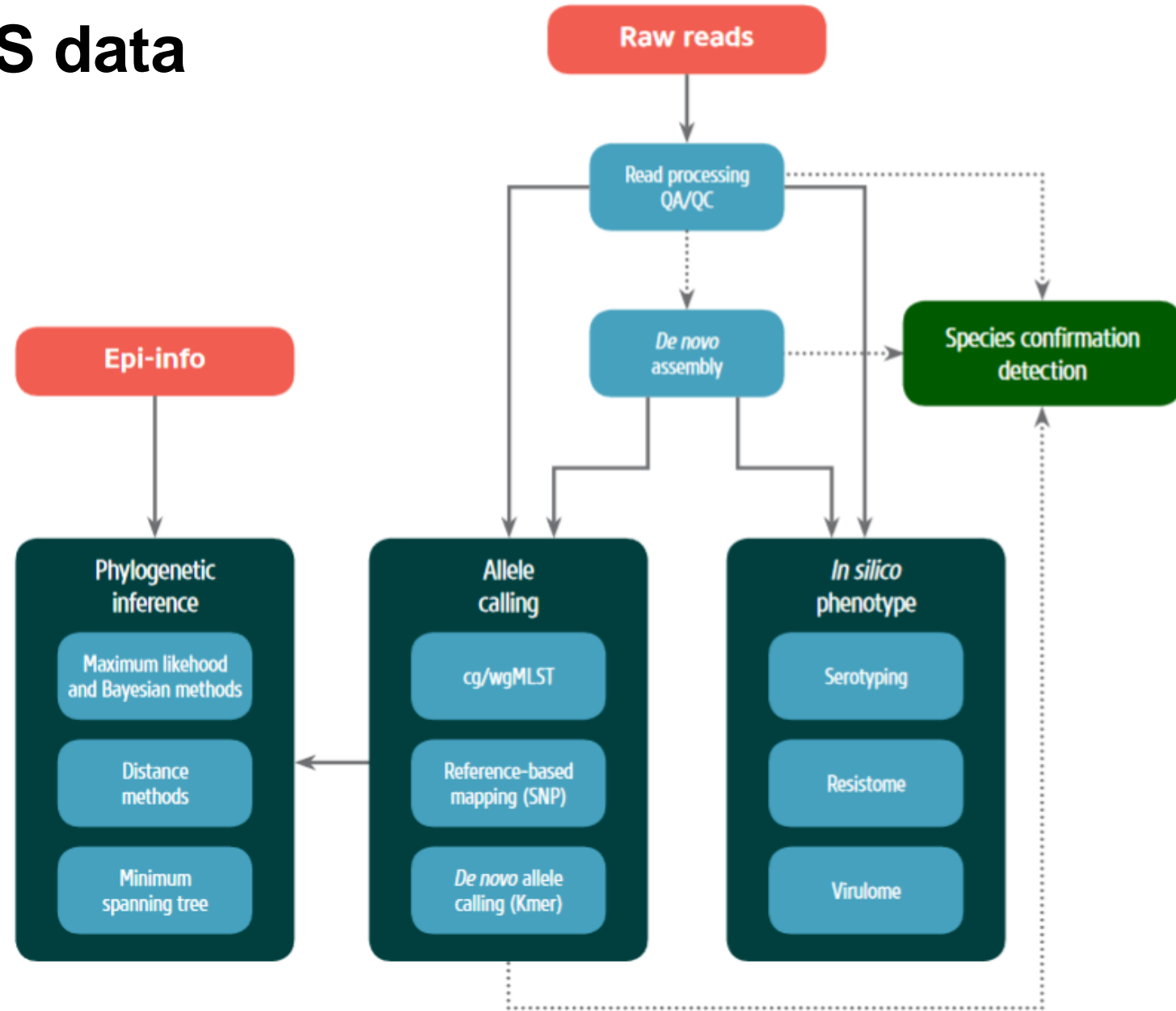
The **many** steps of sequencing

- Overview



Data analysis of WGS data

- Different approaches for different organisms (and subsets)
- Single-isolate analysis and/or phylogenetic analysis of all (relevant) isolates



Commercial pipelines

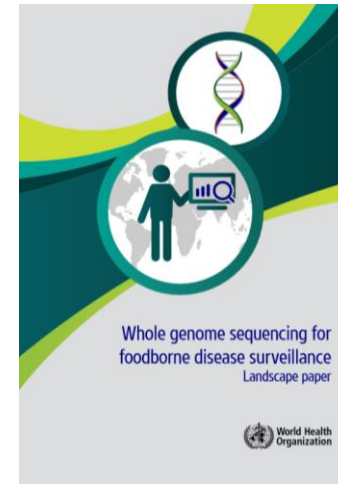
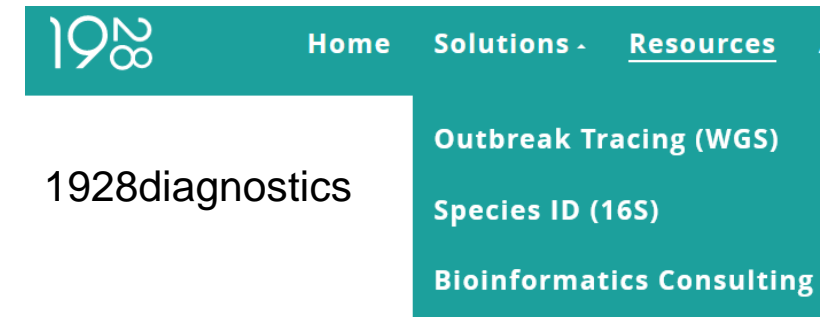
- Various solutions available – including:



Bionumerics



QIAGEN CLC Genomics
Workbench: NGS data analysis for
any species, any platform, any
workflow



**WHO Whole genome
sequencing for foodborne
disease surveillance:
landscape paper**

Great resource – also for
clinical labs!

<https://www.who.int/publications/i/item/789241513869>

Published pipelines and perspectives

– almost random examples

> [Microb Genom.](#) 2021 Jun;7(6):000583. doi: 10.1099/mgen.0.000583.

rMAP: the Rapid Microbial Analysis Pipeline for ESKAPE bacterial group whole-genome sequence data

Ivan Sserwadda^{1 2}, Gerald Mboowa^{3 2}

Article | [Open Access](#) | [Published: 03 July 2020](#)

PARGT: a software tool for predicting antimicrobial resistance in bacteria

[Abu Sayed Chowdhury](#) , [Douglas R. Call](#) & [Shira L. Broschat](#)

[Scientific Reports](#) **10**, Article number: 11033 (2020) | [Cite this article](#)

Volume 23, Number 9—September 2017

Perspective

Bioinformatic Analyses of Whole-Genome Sequence Data in a Public Health Laboratory

Kelly F. Oakeson , Jennifer Marie Wagner, Michelle Mendenhall, Andreas Rohrwasser, and Robyn Atkinson-Dunn

Author affiliations: Utah Department of Health, Utah Public Health Laboratory, Taylorsville, Utah, USA

> [Curr Protoc.](#) 2021 Sep;1(9):e242. doi: 10.1002/cpz1.242.

GALAXY Workflow for Bacterial Next-Generation Sequencing De Novo Assembly and Annotation

Soon Keong Wee¹, Eric Peng Huat Yap¹

Center for Genomic Epidemiology

Phenotyping

Phylogeny

Announcements

Nov 9 - 2022

Unstable services. Dear user of the CGE services. As you may have noticed, our services have been suffering from several periods of down time lately.

Coming soon! We have been working on an entirely **new platform for CGE**. This includes completely new servers and a completely new infrastructure, which will make our platform much **more stable**.

We will start moving services after New Year.

We are very sorry for the inconvenience these down times are causing, and we thank you for your patience. We are very excited about the new infrastructure, and we are working as hard as we can to get it online.

Prediction of a bacteria's pathogenicity towards human hosts.

phylogenetic trees with publicly available whole-genome sequencing data from foodborne, bacterial isolates that were deposited in the short sequencing read archives

WGS-based analysis of bacteria – Requirements

- Expertise on DNA extraction methods
- Expertise on library preparation methods

Not too technically demanding
Ideally a dedicated room

- Access to sequencing platform
- Access and expertise on bioinformatics tools
- Data management infrastructure

Main challenges: cost, implementation

Main challenges: cost, compatibility

Personnel...

Collaboration

- Microbiologist/Molecular biologist

- initial phenotypic/molecular identification and characterization of isolates, including culture purification and storage;
- genomic DNA extraction and purification, library preparation with appropriate quality controls;
- setting up of the sequencing run

- Bioinformatician

- computational analysis of sequencing data,
- Implementation and verification of tools/pipelines
- variant detection and isolate clustering through construction of phylogenetic trees;
- maintenance of accurate secure records of all procedures, including electronic databases of genome sequences and related quality control data;
- quality assessment of original and processed sequencing data

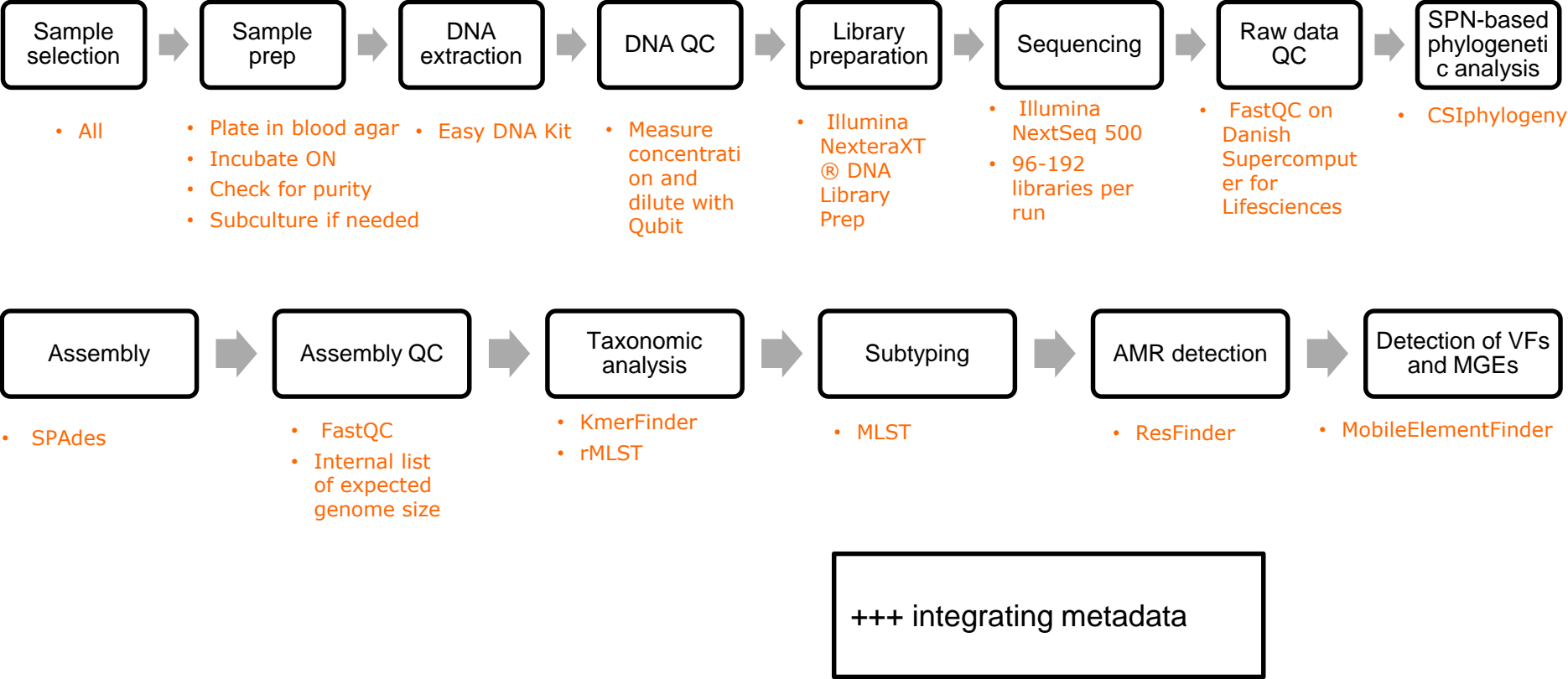
- Epidemiologist

- collecting epidemiological information and integrating it with WGS data
- setting of definitions for what constitutes a cluster to support epidemiological investigations
- determination of which cases need to be followed up to collect epidemiological information, including determination of what isolates are part of a cluster

- Questions – comments are welcome

–Looking forward to the hands-on training next week!

Example: a complete WGS workflow



Ana Rita Rebelo
anrire@food.dtu.dk

Quality control of WGS data

Many different:

- DNA extraction kits
- Sequencing platforms
- Bioinformatics approaches
- Bioinformatics tools



Well defined set of QC parameters

- For the raw data
E.g. nr. and length of raw reads, depth of coverage
- For the assembled genomes
E.g. N50, nr. of contigs, genome size
- For the performance of the tools
E.g. accurately detect PMs and ARGs in sets of benchmarking data

Raw data QC	Assembled data QC
<p>Number of reads Should be as high as possible. No assessed cut-off exist, but enough to obtain the desired coverage of the organism genome</p> <p>Average read length Should correspond to that expected from the sequencing platform and kit. Illumina MiSeq avg read length = 300 bps Illumina NextSeq avg read length = 150 bps</p> <p>Coverage Should as a minimum be 30x, and preferably even higher</p>	<p>Size of assembled genome <i>Enterobacterales</i>: 4.5 Mb - 5.5 Mb Deviation should not be higher than 10%</p> <p>Total number of contigs Should be less than 500</p> <p>N50 Should be over 15.000 bp</p>

$$\text{Coverage} = \text{Number of reads} \times \frac{\text{Read length}}{\text{Genome size}}$$

- **Number of contigs**

Is how many contigs (long sequences) were created during the assembly from good-quality raw reads. A low number of contigs means that the sequencing process was good enough to capture most of the genome and combine the raw data into long, uninterrupted sequences of nucleotides.

- **N50**



It's a parameter that describes the length of all contigs that compose a genome.

- **Depth of coverage of sequenced genome**

Is how many times each bp present in the assembly was sequenced.

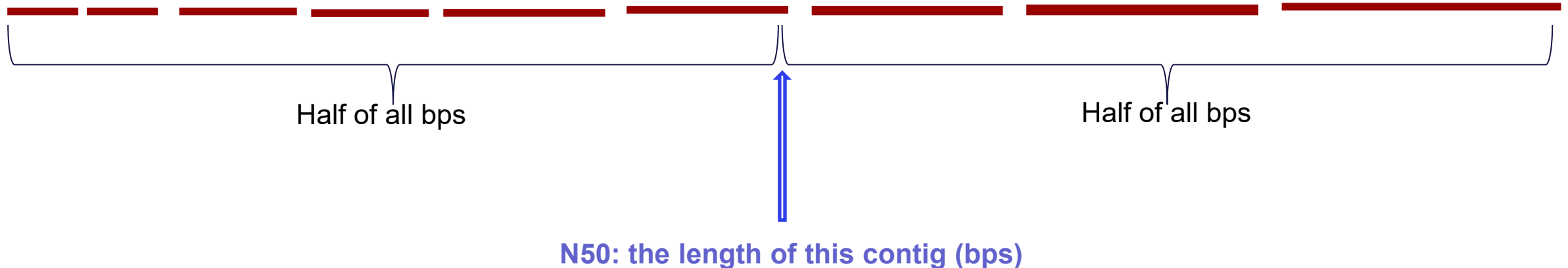
- **Genome size**

Is the number of individual bp that compose the assembled genome.

- **N50**

It's a parameter that describes the length of all contigs that compose a genome.

- 1) All contigs organized by size
- 2) Divide the total base-pairs in half
- 3) The contig that you "catch" has a certain length (bp) → that's the N50



Genome sizes

Gram-positive expected genome size: 1.6 - 3 Mbp

Gram-negative expected genome size: 1.7 – 7 Mbp

Genera or group	Expected genome size (million bps)
<i>Acinetobacter</i>	5
<i>Actinomyces</i>	2.2 – 3
<i>Aerococcus</i>	1.6 – 2
<i>Aeromonas</i>	4.5
<i>Anaerococcus</i>	2
<i>Bacteroides</i>	5
<i>Campylobacter</i>	1.7
<i>Clostridium</i>	4.2 – 5
<i>Corynebacterium</i>	2.5 – 3
<i>Enterobacterales</i> (excluding <i>Proteus</i>)	4.5 – 5.5
<i>Proteus</i>	4
<i>Enterococcus</i>	3
<i>Finnegoldia</i>	1.6 – 2
<i>Fusobacterium</i>	2
<i>Haemophilus</i>	1.8 – 2
<i>Micrococcus</i>	2.5
<i>Moraxella</i>	1.8 – 2
<i>Neisseria</i>	2
<i>Pasteurella</i>	2 – 2.2
<i>Peptoniphilus</i>	1.6 – 1.9
<i>Prevotella</i>	3 – 4
<i>Propionibacterium</i>	2 – 2.5
<i>Pseudomonas</i>	6.5 – 7
<i>Rothia</i>	2
<i>Staphylococcus</i>	2.5 – 2.8
<i>Stenotrophomonas</i>	4.5 – 5
<i>Streptococcus</i>	1.7 – 2.2

Troubleshooting



Usually poor **raw data** QC indicates:
Inadequate DNA extraction
Inadequate library preparation

Usually poor **assembly** QC indicates:
Inadequate DNA extraction
Contaminations

Re-sequence or re-extract?

Evaluation of QC becomes easier with experience + understanding the biochemical principles of the protocols.

QC of bioinformatics analysis

Thresholds for analysis and interpretation depend on the bioinformatics tools

In general:

- Be familiar with the recommended thresholds of each tool
- Use relevant control strategies
- Be critical when evaluating the results

Examples of recommended thresholds

Species identification with rMLST:

- at least 96% of support and absence of hits belonging to different species

Prediction of antimicrobial resistance with AMRFinder:

- minimum 90% identity and minimum 60% length

Thresholds for other tools and purposes:

- EURGen-RefLabCap WGS protocol
- Publication by the authors of the tool
- Publications by other professionals using the tool

Examples of relevant control strategies

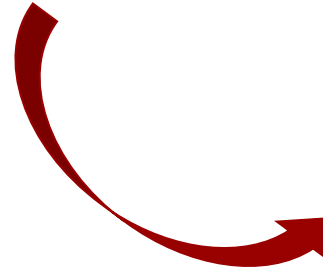
When analysing **isolates independently** (for example detecting AMR determinants in isolates):

- Use **control strains** that harbour the same or similar genetic determinants you want to find

When analysing **isolates together** (for example performing cluster analysis):

- Use **groups of isolates** with well-established genetic relatedness

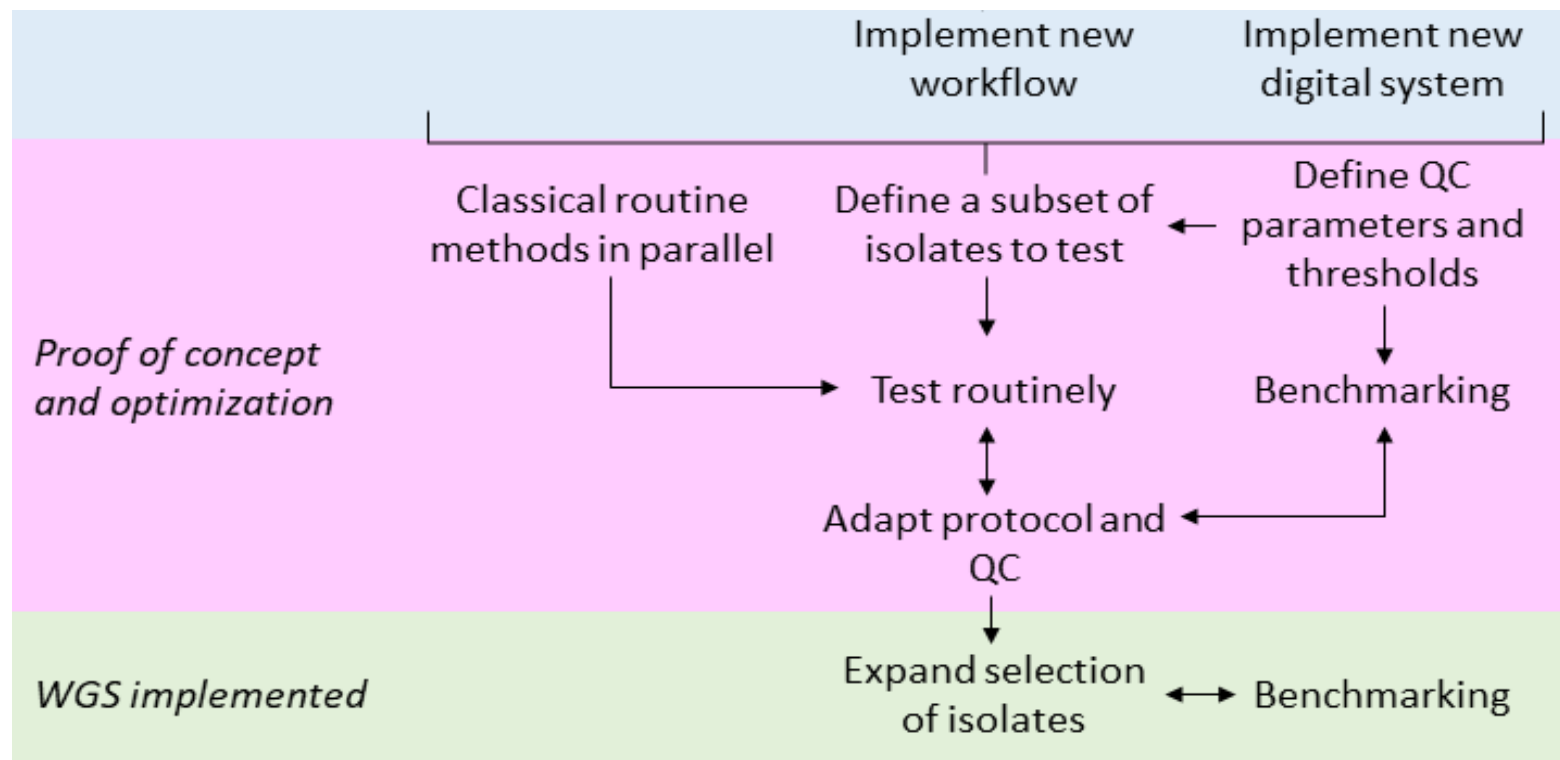
Examples of nonsense results



Why are these specific cases nonsense?

- Species identification tool detects a high proportion of hits belonging to different species
- Different *bla* genes in the same position of the genome with <100% coverage
- Lack of known AMR determinants when there is phenotypic resistance
- Zero SNPs between isolates very separated according to metadata (time/space)

Benchmarking in your settings



Ana Rita Rebelo
anrire@food.dtu.dk

Questions and wrapping up the day

Updated agenda for next days

Second day (physical) – Wednesday 7 December 2022

9:00 - 9:30: Introduction and agenda for the day (Rene S. Hendriksen, DTU)

9:30 - 15:00: Laboratory work - Illumina MiSeq library preparation and sequencing (including *ad hoc* coffee, snacks and lunch)

15:00 - 15:30: Coffee break

15:30 - 16:15: Exercise about quality control of WGS data (Ana Rita Rebelo, DTU)

16:15 - 17:00: Exercise about bioinformatics tools for species identification and serotyping (Jette Sejer Kjeldgaard, DTU)

EURGen-RefLabCap@food.dtu.dk

**Thank you on behalf of the
EURGen-RefLabCap team**