





Introduction to viral genome reconstruction using massive sequencing: Sars-cov-2 use case

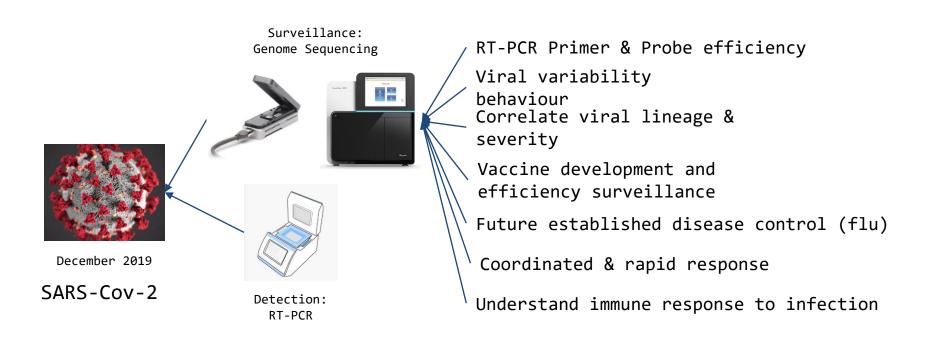
BU-ISCIII
Bioinformatics Unit, Institute of Health Carlos III
Madrid, Spain







1. Background



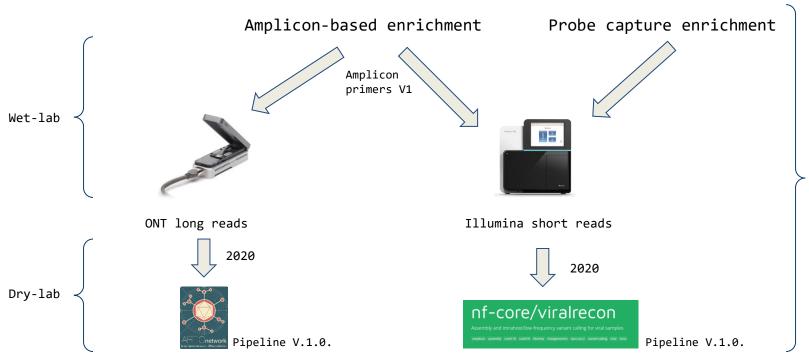








2. Sequencing Approaches





Data

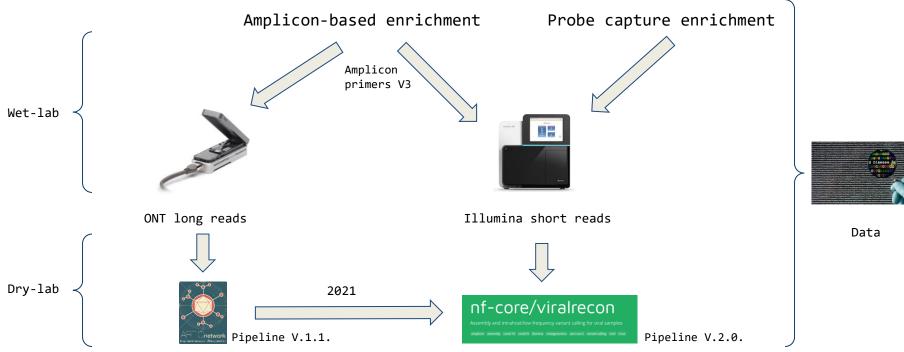








2. Sequencing Approaches



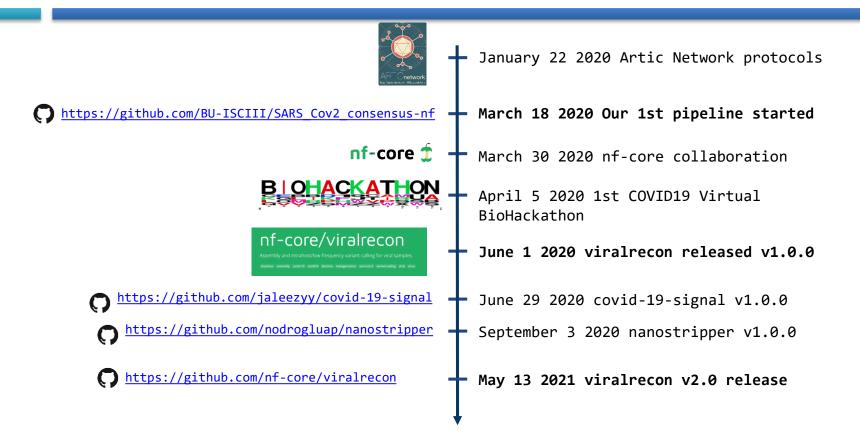








3. History











4. Viralrecon

nf-core/ **t** viralrecon

https://github.com/nf-core/viralrecon 🖸

nextflow

Multiple compute infrastructures

Portable 8 8 1

Easy to install

Reproducible

Stable code

















The need of standardisation I

Sequencing techniques are starting to be used in clinical diagnosis, and therefore workflows have to assure:

Reproducibility

Results always have to be reproducible

Portability

The analysis workflow must be executable in different platforms

Scalability

The analysis workflow must be able to work with different numbers of samples









Nextflow I

- **Nextflow** is a DSL for parallel and scalable computational pipelines.
- It enables scalable and reproducible scientific workflows using software containers.
- It allows the adaptation of pipelines written in the most common scripting languages.
- Its fluent **DSL** simplifies the implementation and the deployment of complex parallel and reactive workflows on clouds and clusters.









Nextflow II

Fast prototyping

Nextflow allows you to write a computational pipeline by making it simpler to put together many different tasks.

You may reuse your existing scripts and tools and you don't need to learn a new language or API to start using it.

Portable

Nextflow provides an abstraction layer between your pipeline's logic and the execution layer, so that it can be executed on multiple platforms without it changing.

It provides out of the box executors for SGE, LSF, SLURM, PBS and HTCondor batch schedulers and for Kubernetes and Amazon AWS cloud platforms.

Continuous checkpoints

All the intermediate results produced during the pipeline execution are automatically tracked.

This allows you to resume its execution, from the last successfully executed step, no matter what the reason was for it stopping.

Reproducibility

Nextflow supports Docker and Singularity containers technology.

This, along with the integration of the GitHub code sharing platform, allows you to write self-contained pipelines, manage versions and to rapidly reproduce any former configuration.

Unified parallelism

Nextflow is based on the dataflow programming model which greatly simplifies writing complex distributed pipelines.

Parallelisation is implicitly defined by the processes input and output declarations. The resulting applications are inherently parallel and can scale-up or scale-out, transparently, without having to adapt to a specific platform architecture.

Stream oriented

Nextflow extends the Unix pipes model with a fluent DSL, allowing you to handle complex stream interactions easily.

It promotes a programming approach, based on functional composition, that results in resilient and easily reproducible pipelines.





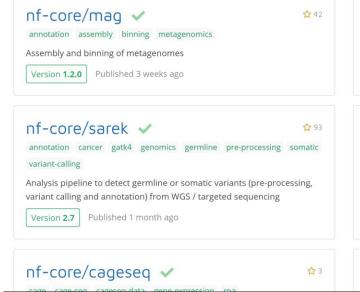


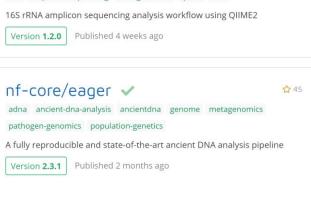


Nextflow III: nf-core

The nf-core project is a diverse project spread across many groups. It is A community effort to collect a curated set of analysis pipelines built using Nextflow.







\$ 54

☆ 298

nf-core/ampliseq <

nf-core/rnaseq 🗸

16s amplicon-sequencing metagenomics qiime rrna









Containers I

Linux containers is a generic term for an implementation of operating system-level virtualization for the Linux operating system.

Containers allow us to **port** pipelines and **replicate** their exact execution environments across different hardware.

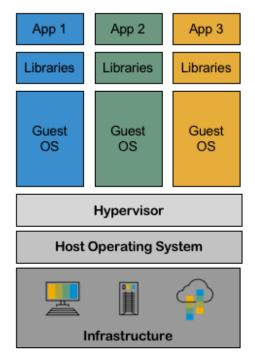




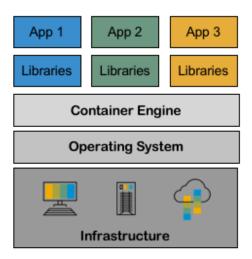




Singularity I



Hypervisor virtualization



Container virtualization









Singularity II

Singularity is a free, cross-platform and open-source computer program that performs operating-system-level virtualization.

One of the main uses of Singularity is to bring containers and reproducibility to scientific computing and the HPC world.

While Docker is broadly used, Singularity is fully compatible with Docker, plus Singularity does not require root permissions to be executed.







And now is when you think...

How all this hard to understand stuff can make my life easier???

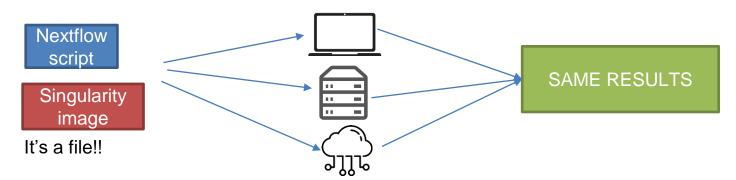






How all this hard to understand stuff can make my life easier?

- Reproducibility is imperative in clinical bioinformatics.
- Once this is all set you can just:
 - Run your pipeline ANYWHERE getting same results.
 - With ANYWHERE I mean your laptop, your home pc, your computation cluster, just one Amazon server you rent for some time...
 - You don't need to change ANYTHING in your code, just some config.
 - You don't need to install ANYTHING anywhere EVERYTHING is installed and in the same versión in your shiny container.

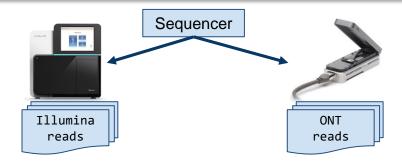








4. Viralrecon

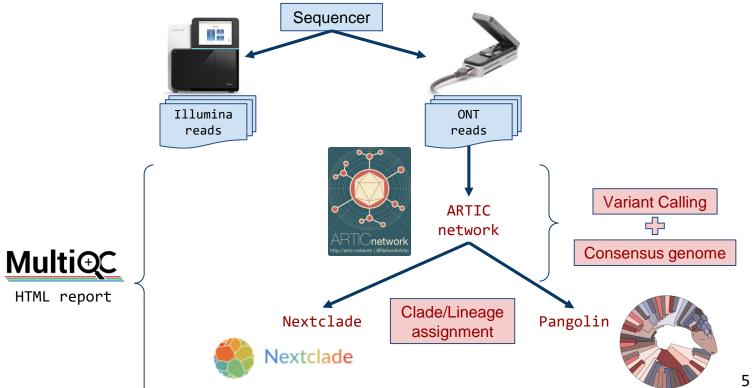








4. Viralrecon for ONT reads



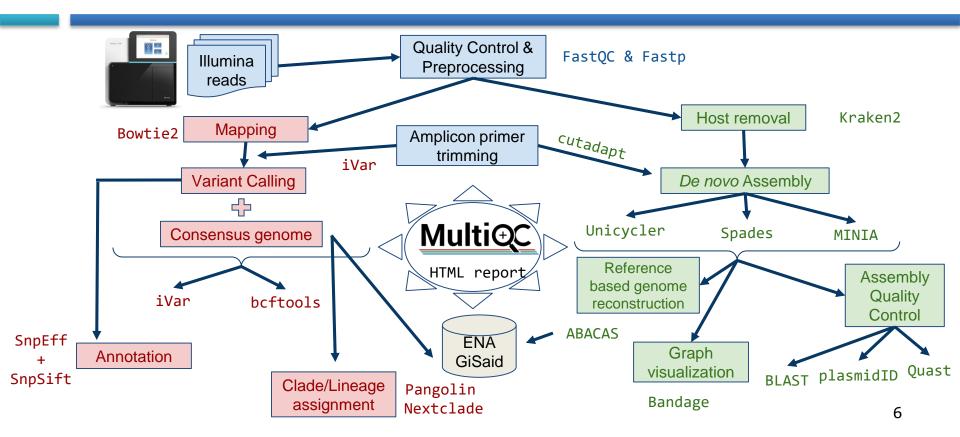








4. Viralrecon for Illumina reads



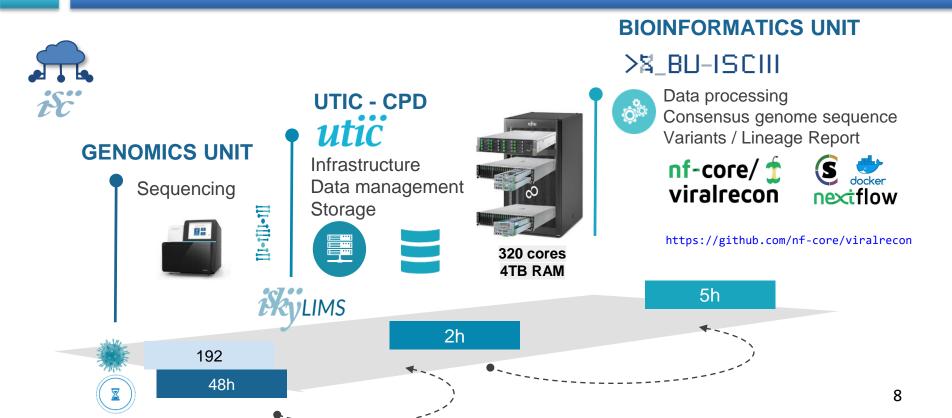








4. Viralrecon ISCIII









Input data

Which input format do we have from massive sequencing?







FASTQ format

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



Quality: must be 1 bit







Sequencing quality assessment

- To assess 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







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

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

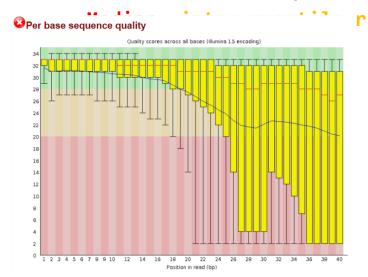


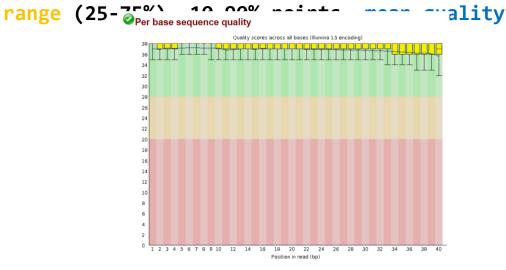




FastQC: Per base sequence quality

 Overview of the range of quality values across all bases at each position in the FastQ file









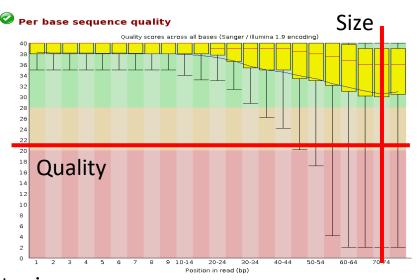


Sequence filtering

- 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



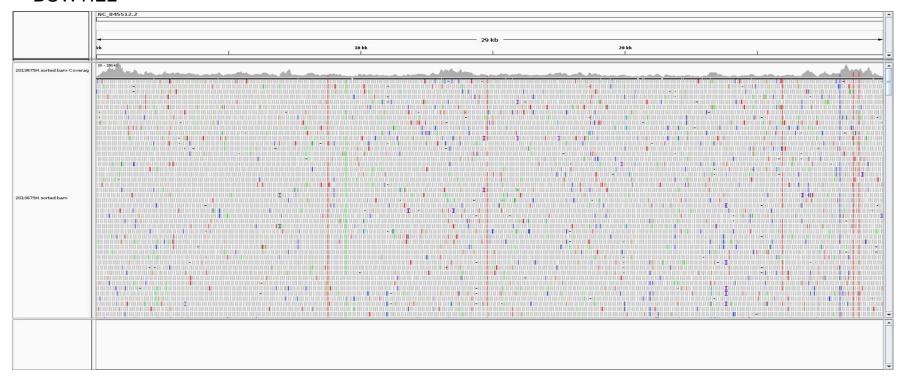






Mapping

BOWTIE2









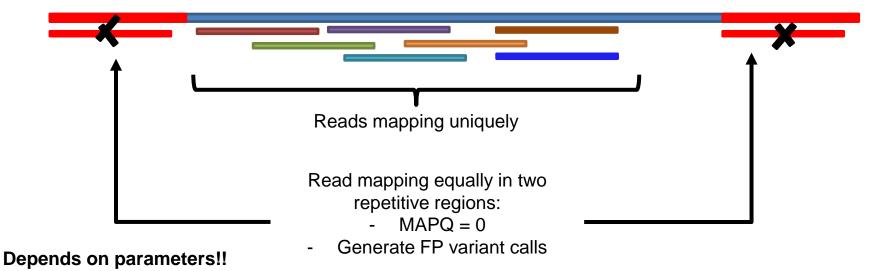
Mapping

BOWTIE2

Mapping software looks for the best match for each read in the genome.

Paired-end reads help the mapper to find the perfect spot!

Reference genome









Input data

Which output format do we have from mapping step?







SAM format

Col	Field	Type	Regexp/Range	Brief description
1	QNAME	String	[!-?A-~]{1,255}	Query template NAME
2	FLAG	Int	[0,2 ¹⁶ -1]	bitwise FLAG
3	RNAME	String	* [!-()+-<>-~][!-~]*	Reference sequence NAME
4	POS	$_{ m Int}$	[0,2 ³¹ -1]	1-based leftmost mapping PoSition
5	MAPQ	Int	[0,2 ⁸ -1]	MAPping Quality
6	CIGAR	String	* ([0-9]+[MIDNSHPX=])+	CIGAR string
7	RNEXT	String	* = [!-()+-<>-~][!-~]*	Ref. name of the mate/next read
8	PNEXT	Int	[0,2 ³¹ -1]	Position of the mate/next read
9	TLEN	Int	[-2 ³¹ +1,2 ³¹ -1]	observed Template LENgth
10	SEQ	String	* [A-Za-z=.]+	segment SEQuence
11	QUAL	String	[!-~]+	ASCII of Phred-scaled base QUALity+33

```
QHD VN:1.5 SO:coordinate
@SQ SN:ref LN:45
r001
      99 ref 7 30 8M2I4M1D3M = 37 39 TTAGATAAAGGATACTG *
r002 0 ref
              9 30 3S6M1P1I4M * 0 O AAAAGATAAGGATA
                              * O O GCCTAAGCTAA
r003 0 ref
              9 30 5S6M
                                                        * SA:Z:ref,29,-,6H5M,17,0;
r004
       0 ref 16 30 6M14N5M
                                    O ATAGCTTCAGC
r003 2064 ref 29 17 6H5M
                                    O TAGGC
                                                        * SA:Z:ref,9,+,5S6M,30,1;
r001
     147 ref 37 30 9M
                              = 7 -39 CAGCGGCAT
                                                        * NM:i:1
```







Mapping quality control

% mapping: number of reads mapping againts reference genome.

Picard Samtools



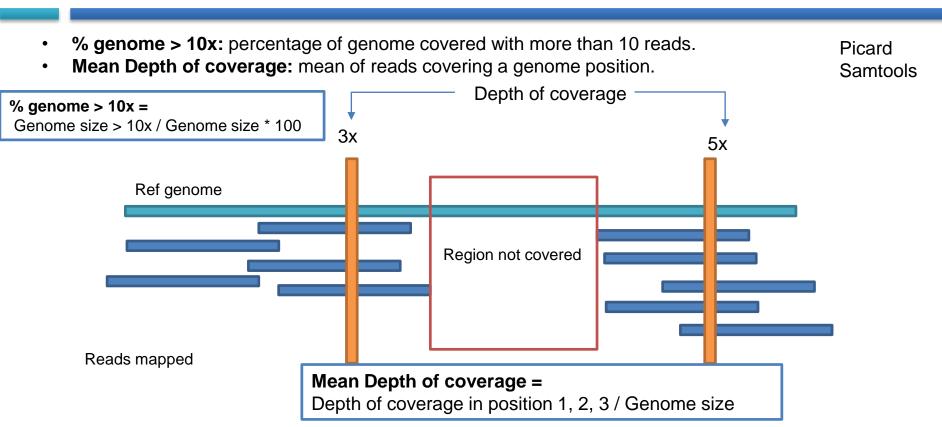
Mandatory parameter for microbial genomics!! It indicates us how many reads we have from our organism of interest. In human genomics this is almost always 99.99% unless something terrible happens. Not here!!!







Mapping quality control

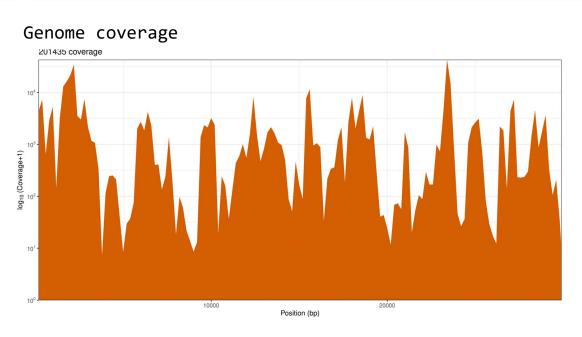


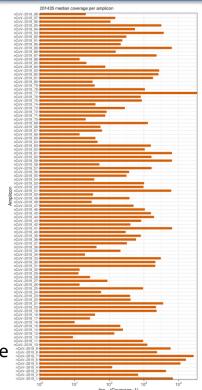






Amplicon QC results





Amplicon coverage

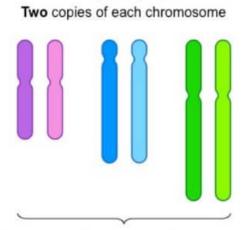






Diploid vs Haploid

Differences between Diploid and Haploid



Diploid (2n)

Three pairs of homologous chromosomes (of maternal and paternal origin)



Haploid (n)

One copy of each chromosome

Three non-homologous chromosomes







Diploid vs Haploid

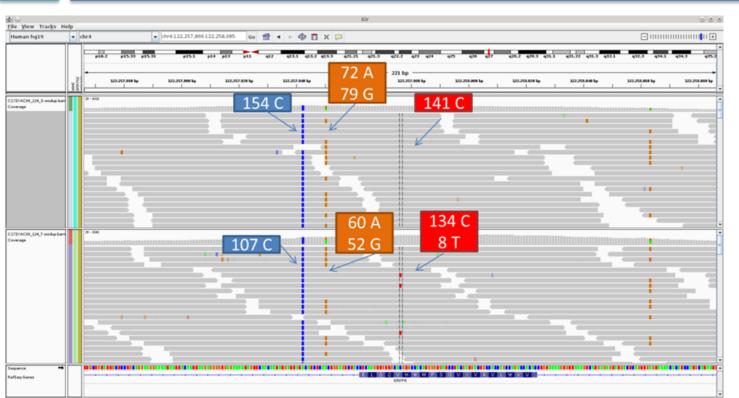
How do we see this differences on massive sequencing?







Diploid vs Haploid



We must say variant callers which type of genotypes we are looking for.

Diploid: 1/1,
0/1 or 1/1

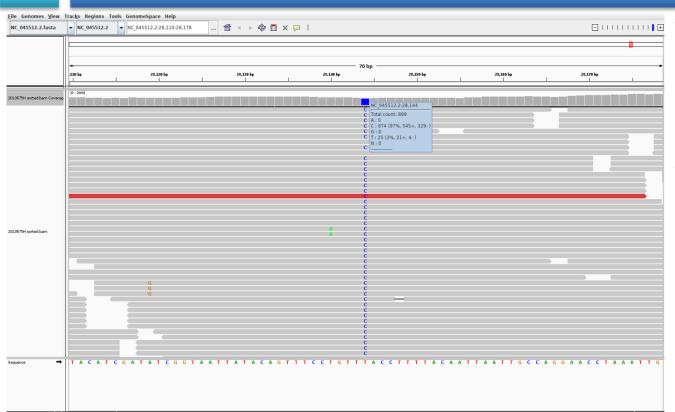
Haploid: 1 or 0.







Variant calling



Variant callers walk
through each alignment
column independently and
evaluate if there is a
variant.

A variant is called in haploid genomes when it's supported by at least 90% of the reads.

VARSCAN2 IVAR







Input data

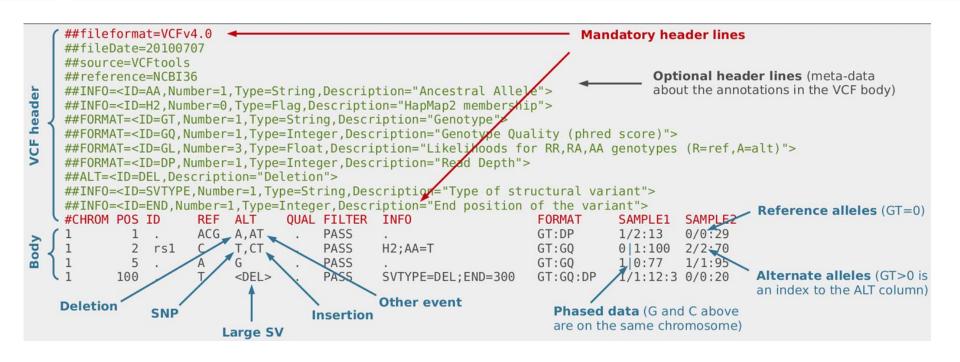
Which output format do we have from variant calling step?







VCF format



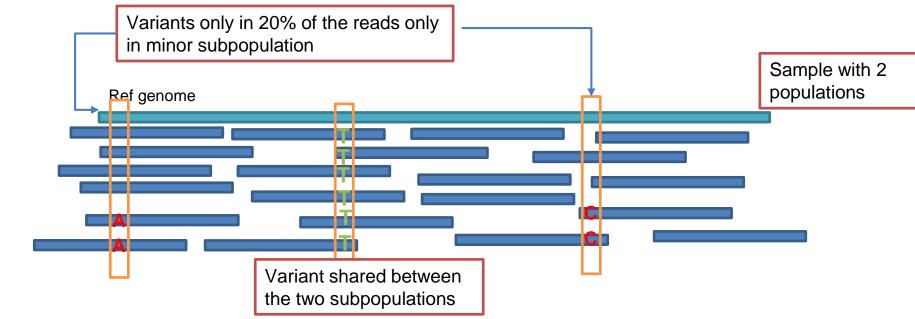






Viral subpopulation - Quasispecies

- Just as in clonal subpopulations in tumor samples, we can have viral subpopulations called quasispecies in viral samples.
- We detect them using the alternative allele frequency.



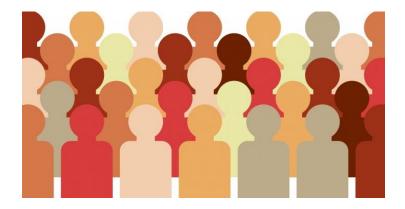






Population Allele frequency vs Sample Allele frequency

 <u>Population allele frequency:</u> probability of finding an allele in the population. Number of individuals carrying an allele vs total of individuals in the population.



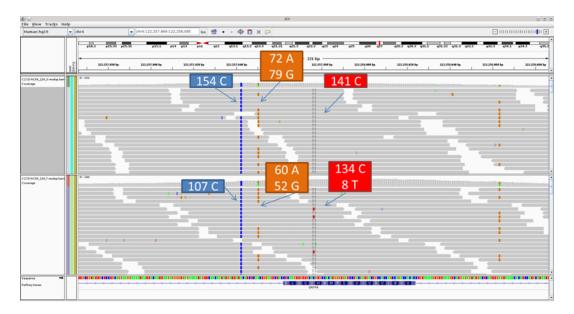






Population Allele frequency vs Sample Allele frequency

 Alternate/Base allele frequency: number of reads supporting the alternate allele vs total of reads.









Consensus genome

- Select variants: > 90% allele frequency
- Include variants in reference genome.
- Mask low frequency positions: <10x.

Ref genome Variants > 90% 265 1050 10233 A Consensus genome Mask low depth of coverage regions







Assembly

Reconstruct a representation of the original DNA from shorter DNA sequences or small fragments known as reads

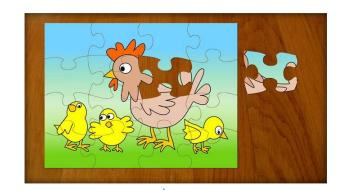
- De novo: with no previous knowledge of the genome to be assembled. It overlap the end of the end of each read in order to créate a longer sequence.
- Assembly with reference: A similar but not identical genome guides the assembly process. Map reads over supplied genome.

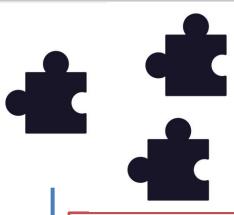






Assembly





The bigger the pieces the easier to reconstruct the puzzle!





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Assembly





Obviously this a little bit more complicated....

We have LOADS of really little pieces and a big puzzle in proportion





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Obviously this a little bit more complicated....

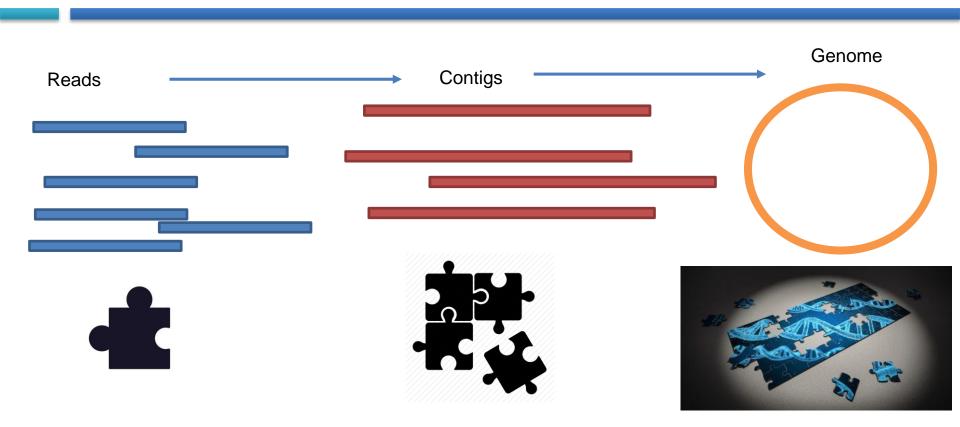
Actually we don't even have the box image most of the time





>¤_BU-ISCIII

Assembly









Assembly

Reconstruct a representation of the original DNA from shorter DNA sequences or small fragments known as reads

- De novo: with no previous knowledge of the genome to be assembled. It overlap the end of the end of each read in order to créate a longer sequence.
- Assembly with reference: A similar but not identical genome guides the assembly process. Map reads over supplied genome.







Assembly: contig y scaffold

- Contig: continuous sequence made up of overlaping shorter sequences
- Scaffold: two or more contigs located and rearranged according to spatial information(pair-end, mate pair, reference)



https://www.biostars.org/p/253222/







Assembly polishing and quality control

- Abacas sort contigs according to reference genome.
- Scaffolding
- Visualization



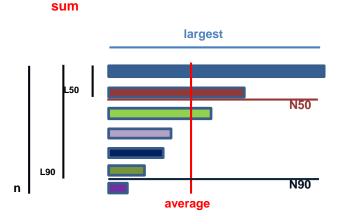




Assembly polishing and quality control

50%

- sum = total bases number
- n = contigs number
- average = average contig length
- largest = largest contig
- N50 = length of the shortest contig where 50% of sum is held
- L50 = number of contigs which have 50% of the genome
- N90 = length of the shortest contig where 90% of sum is held.
- L90 = number of contigs which have 90% of the genome



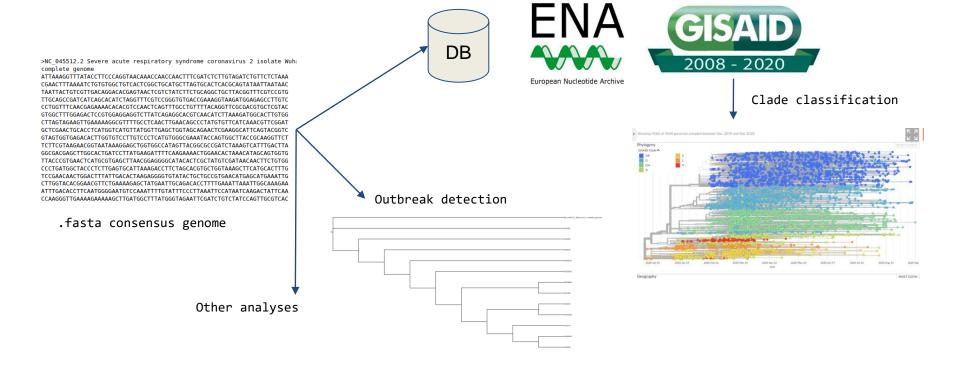
50%







And...after all this?









Linage assignment: Pangolin



pangolin assigns lineages to query sequences as described in Rambaut et al 2020.

https://cov-lineages.org/







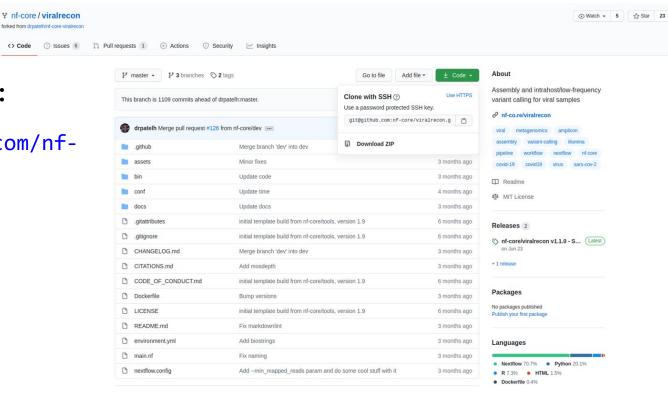


Conclusions

You can find viralrecon in:

https://github.com/nfcore/viralrecon













Thanks for your attention!

Thanks to all my colleges in BU-ISCII



Thanks to Genomics Unit and the Reference Laboratory Labs in CNM



Sarai Varona Luis Chapado Guillermo Gorines Alberto Lema **Isabel Cuesta**



(Find us in https://github.com/BU-ISCIII)

And special thanks to the nf-core community (https://nf-co.re/)

Harshil Patel

(https://github.com/drpatelh)



Bioinformatics & Biostatistics Group at The Francis Crick Institute, London, now in Segera Labs