

Atelier scRNA-seq

Technology for scRNA-seq and data processing

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Morgane Thomas-Chollier - GenomiqueENS, Paris + IFB-core

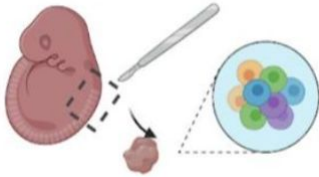


Organisation of this session

- From cells to nucleotide sequences (reads)
 - focus on the 10X genomics technology
 - how are the reads organised
- Preprocessing : from reads to raw count matrix
 - quality check (FASTQC)
 - mapping (STAR)
 - how is annotation used
 - barcode and UMI treatment

Global overview of a scRNA-seq experiment

Tissue dissection + cell dissociation



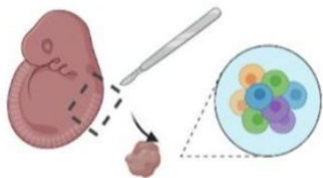
Cell partitioning + mRNA capture

Library preparation + sequencing



Global overview of a scRNA-seq experiment

Tissue dissection + cell dissociation



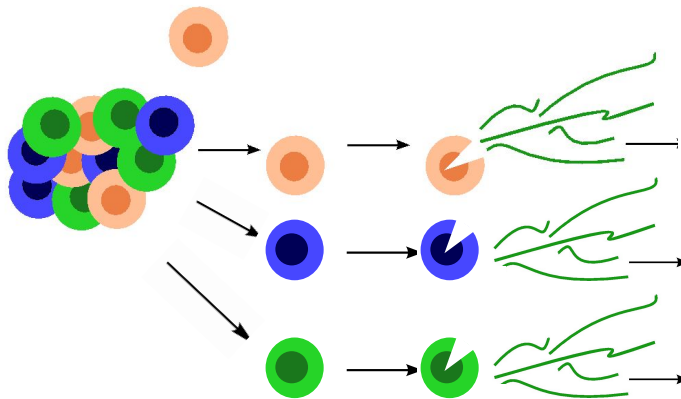
Cell partitioning + mRNA capture

this step enables to treat each cell separately, and capture its RNA while retaining from which cell it originates

Library preparation + sequencing

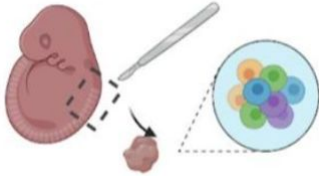


scRNA-seq



Global overview of a scRNA-seq experiment

Tissue dissection + cell dissociation



Cell partitioning + mRNA capture

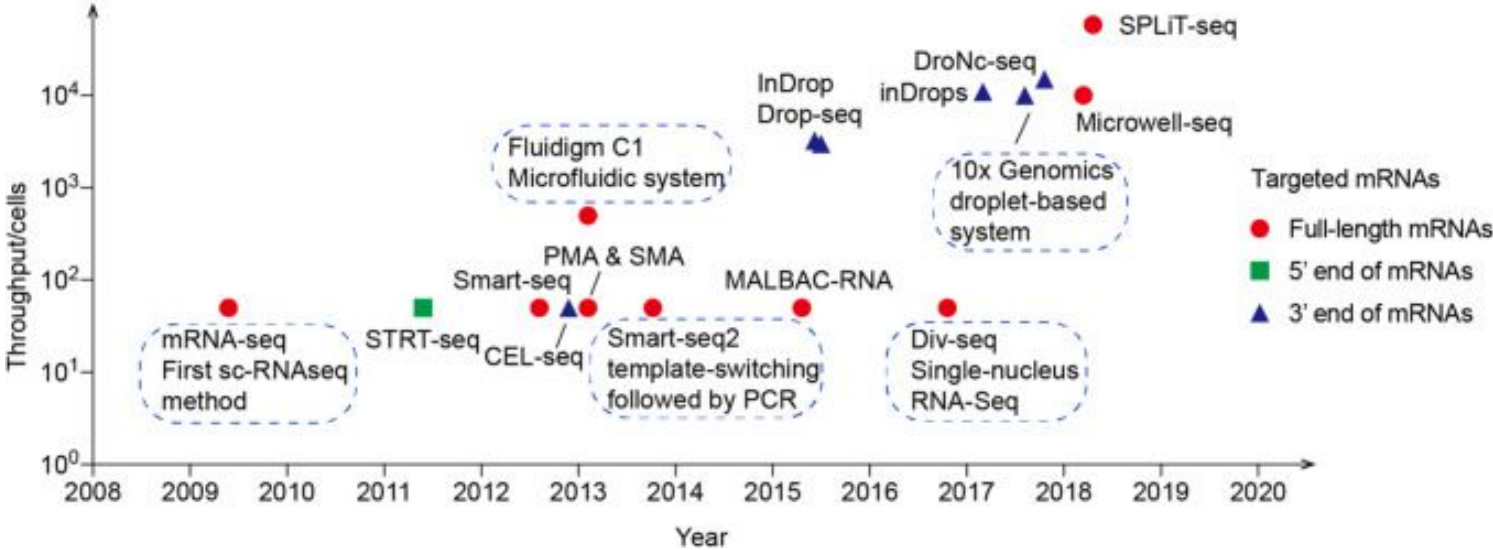
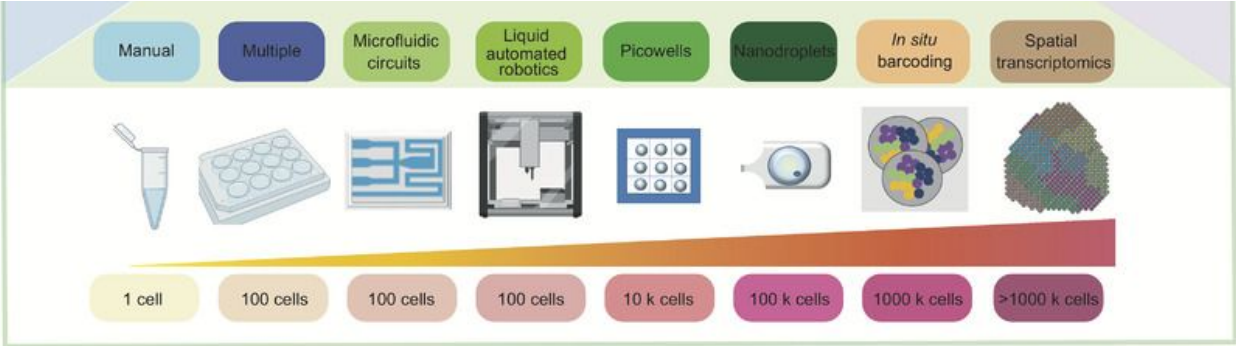
various technologies
developed over time for
this specific step



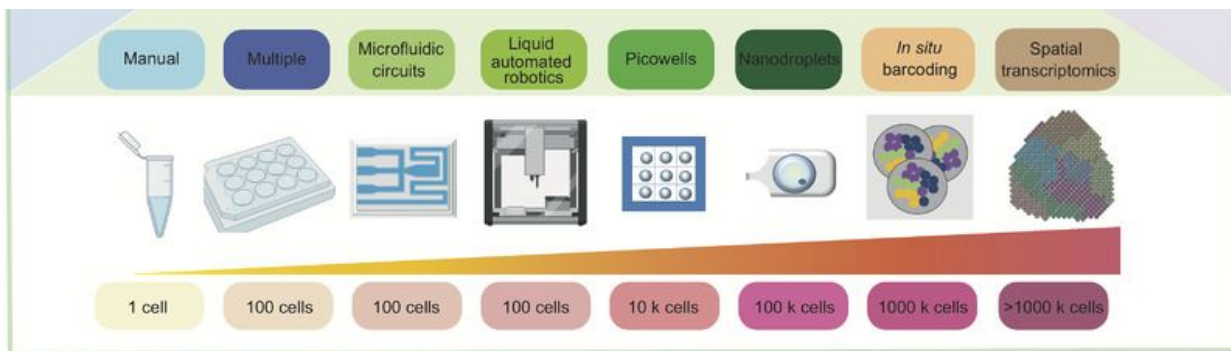
Library preparation + sequencing



Technologies over the last decade



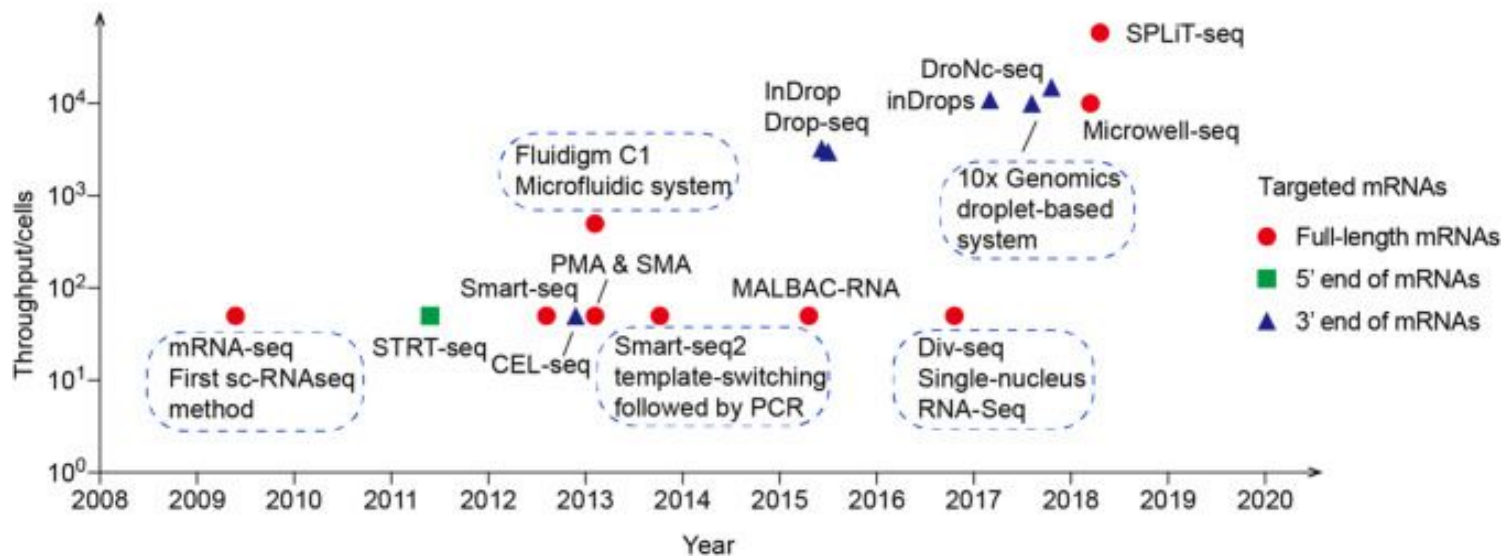
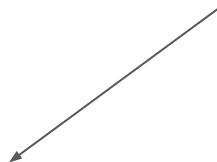
Technologies over the last decade



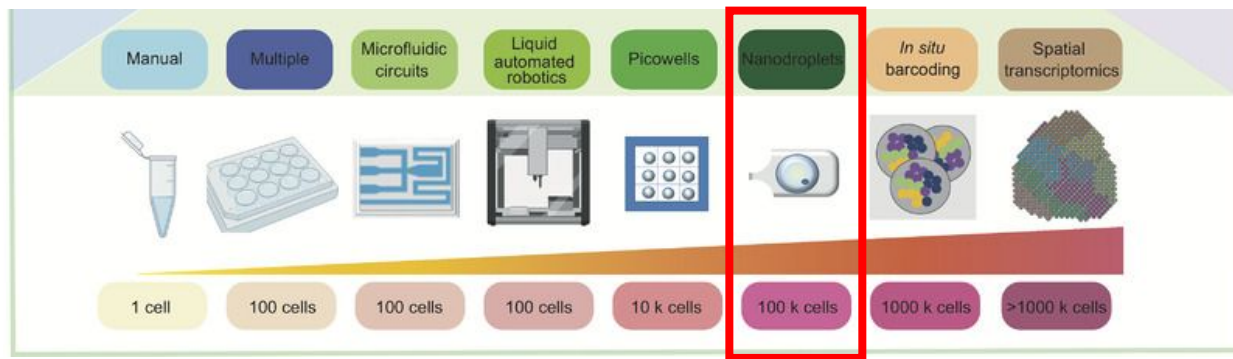
Differences in cell isolation/partitioning : the device can be a simple plate to complex microfluidic machines

Technologies over the last decade

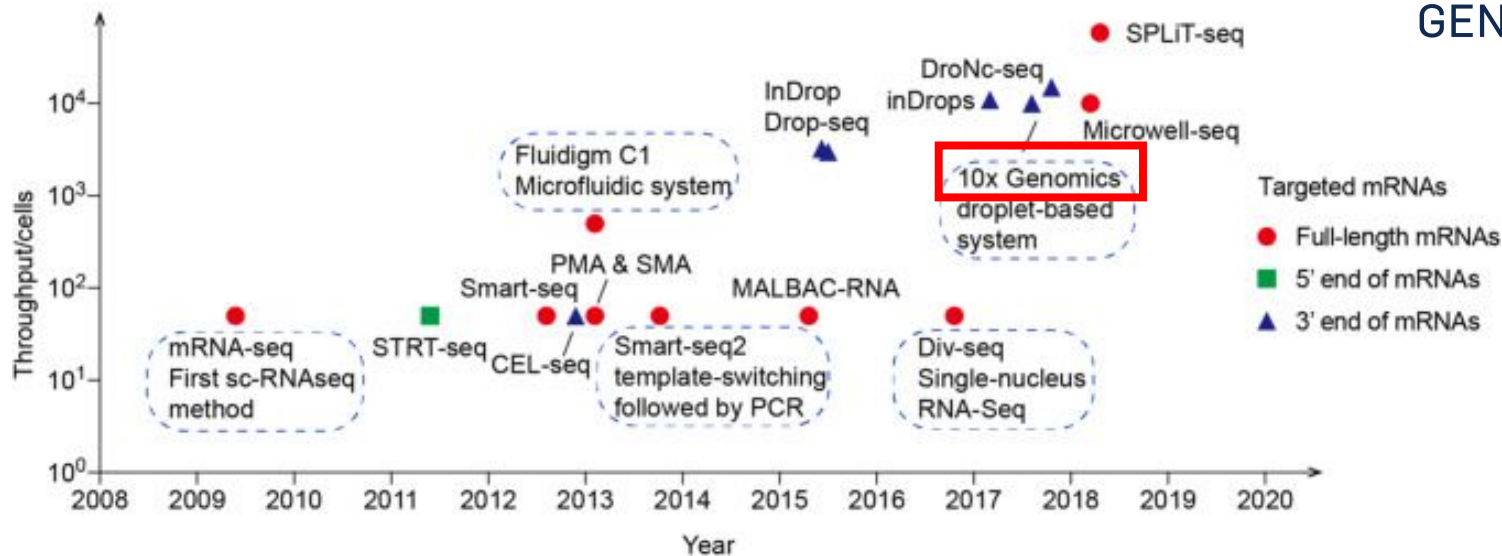
The number of cells that can be studied has grown from a handful to >10,000 in 10 years



Technologies over the last decade

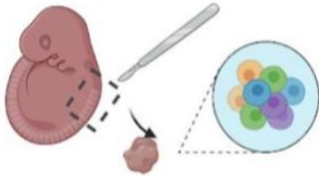


The technology that has enabled widespread usage of scRNA-seq approach is the droplet-based approach proposed by the company 10X Genomics.



Single-cell transcriptomics with 10X genomics technology

Tissue dissection + cell dissociation



Cell partitioning + mRNA capture

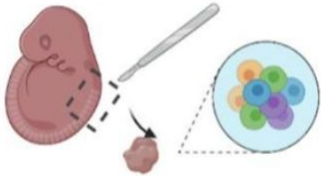


Library preparation + sequencing

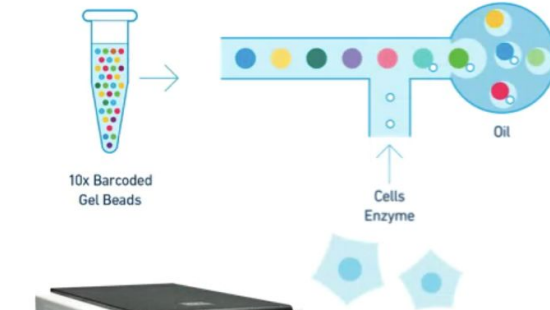


Single-cell transcriptomics with 10X genomics technology

Tissue dissection + cell dissociation



Cell partitioning + mRNA capture



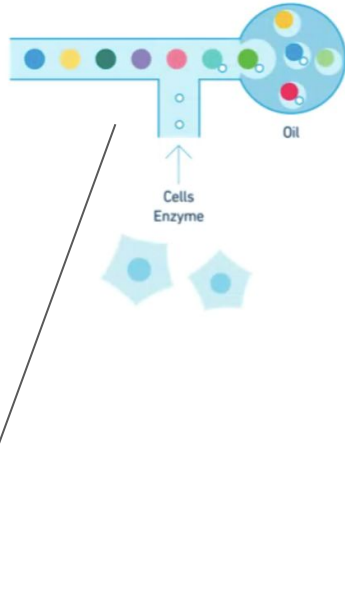
Library preparation + sequencing



How is the 10X Genomics droplet-based system working ?

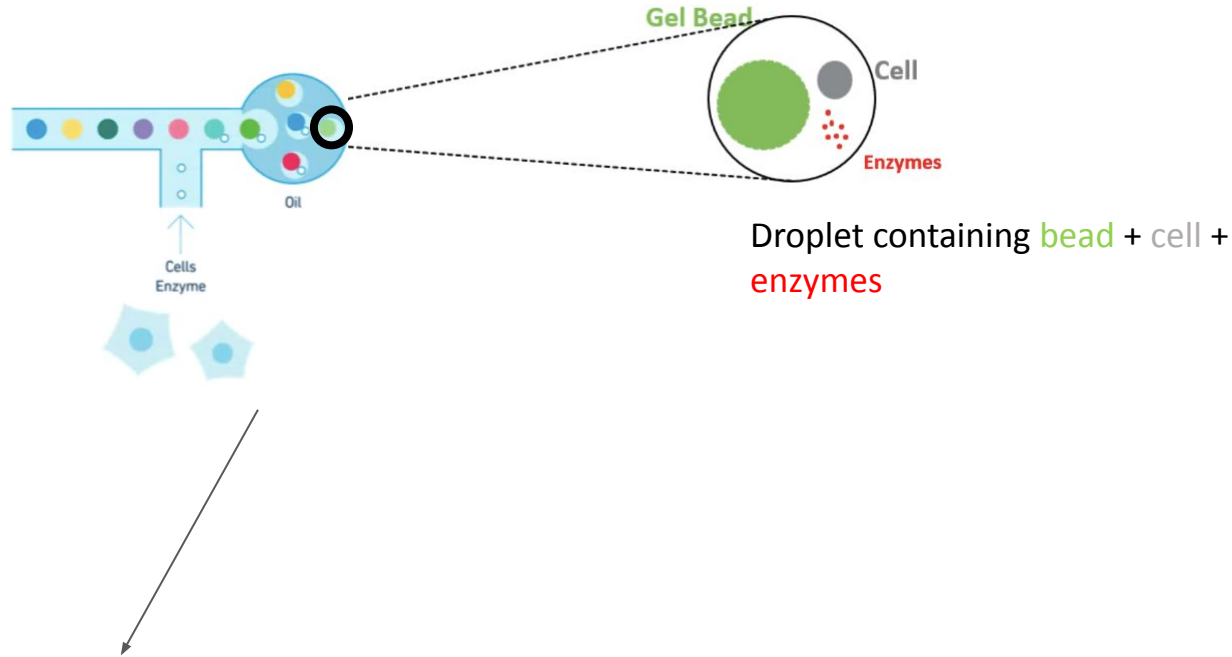


Single-cell transcriptomics with 10X genomics technology



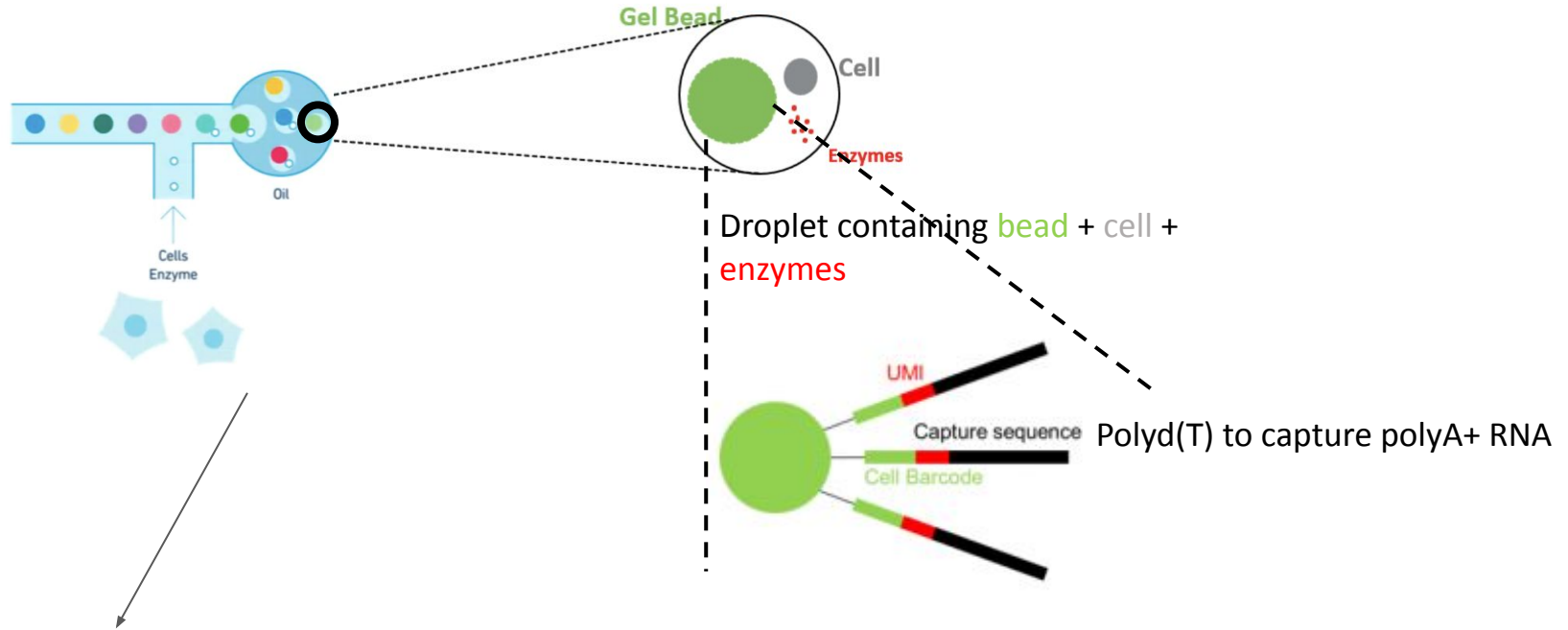
Cells and gel beads arrive
in the device from 2
separate channels

Single-cell transcriptomics with 10X genomics technology



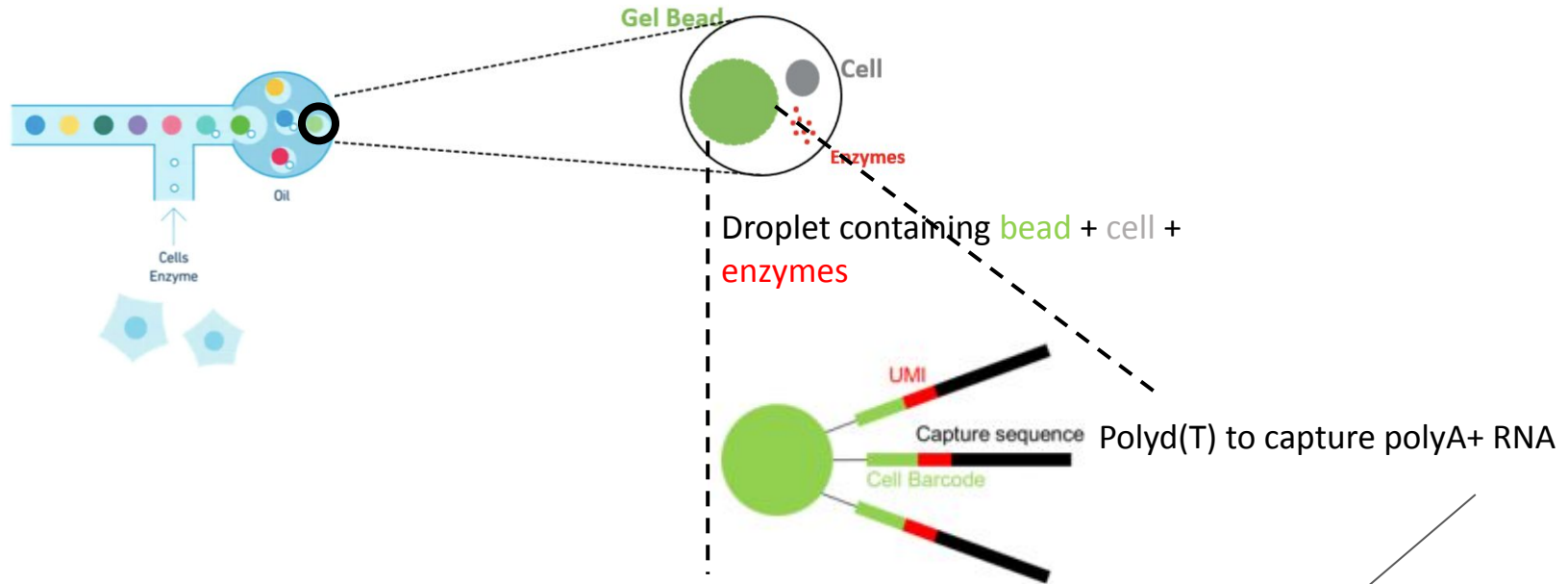
A single cell and a single gel bead (+ enzymes) are then encapsulated in a droplet

Single-cell transcriptomics with 10X genomics technology



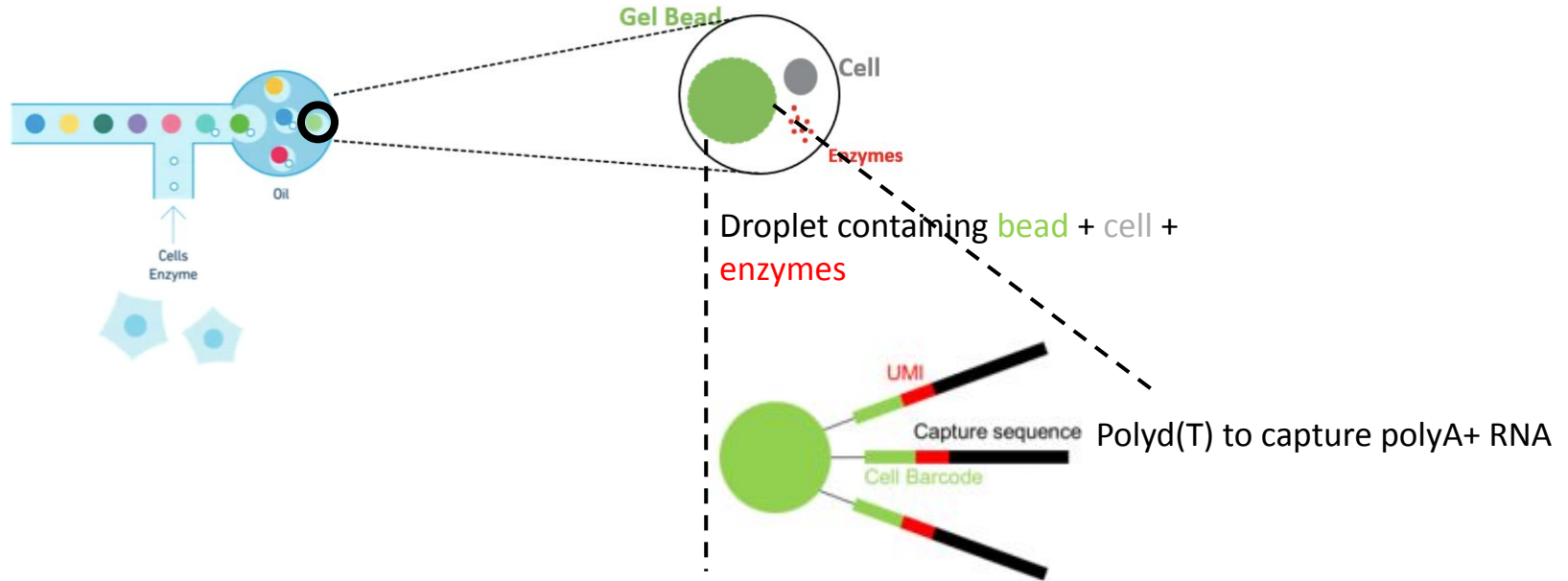
The gel bead is special : it is covered with molecules made of 3 parts

Single-cell transcriptomics with 10X genomics technology



The black part is the **capture sequence** (to “catch” the RNA). 10X has various capture sequences. Here the sequence is polyd(T) to capture RNA that are polyadenylated polyA+.

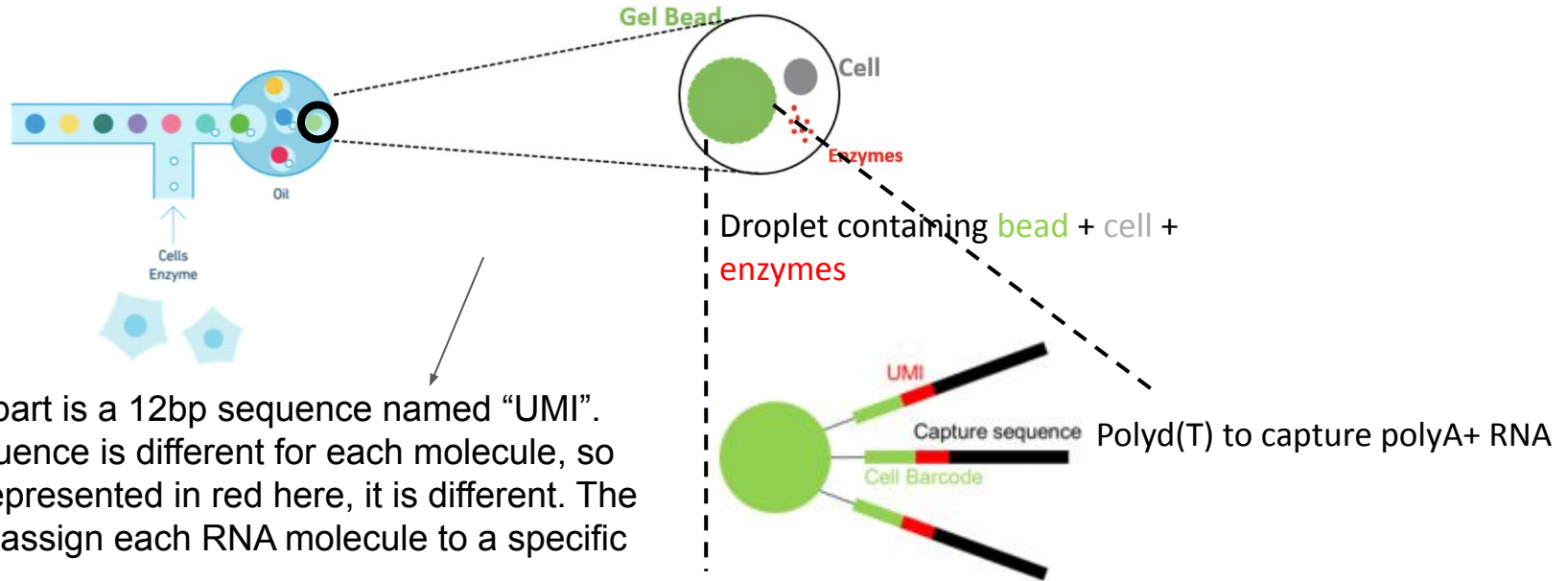
Single-cell transcriptomics with 10X genomics technology



Cell Barcode (16bp) = sequence specific to each bead (so each cell)

The green part is a 16bp sequence named "barcode". This same sequence is all over the bead. These barcodes are created by 10X and the list is available.

Single-cell transcriptomics with 10X genomics technology

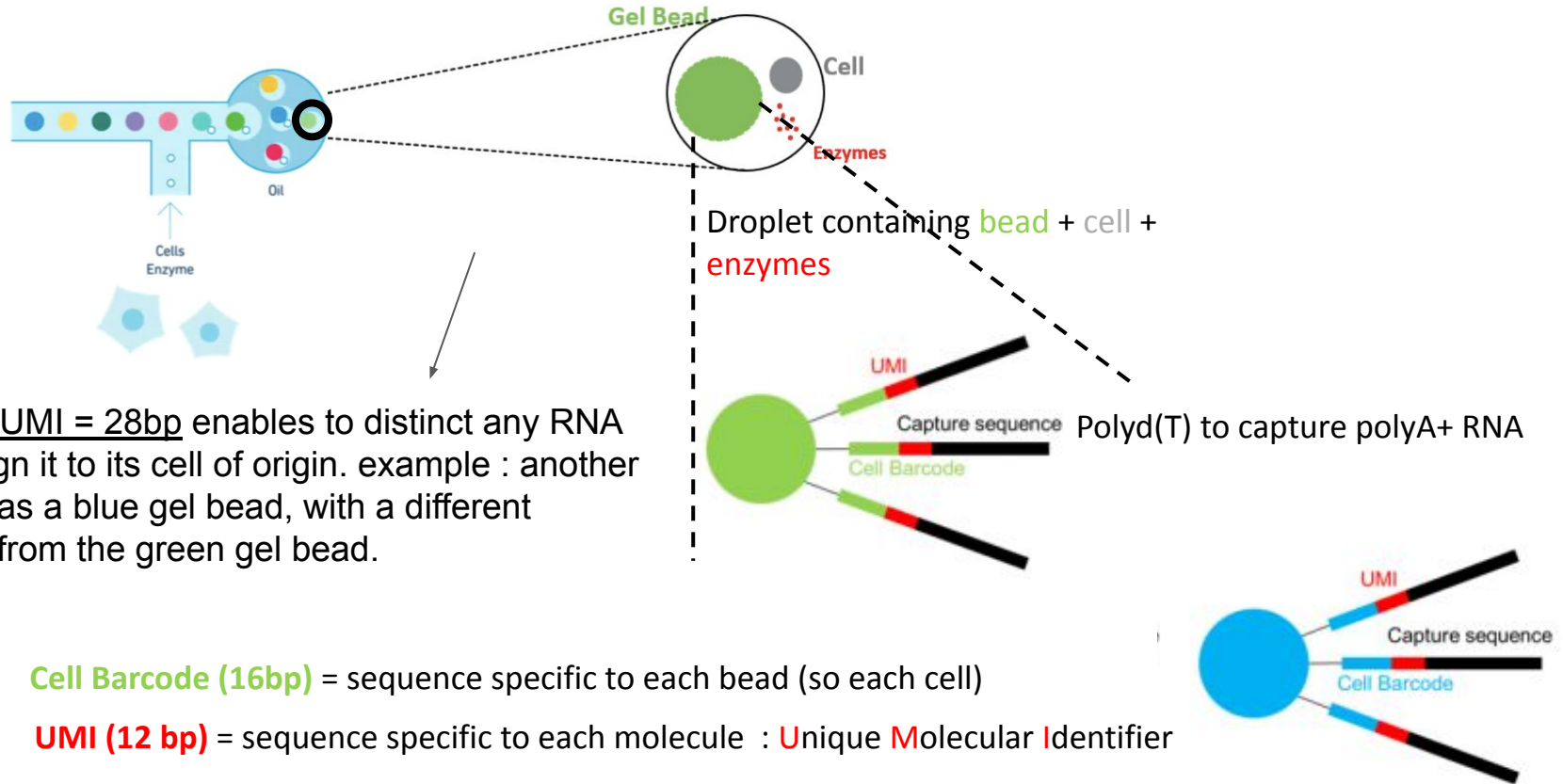


The red part is a 12bp sequence named “UMI”. This sequence is different for each molecule, so even if represented in red here, it is different. The aim is to assign each RNA molecule to a specific UMI.

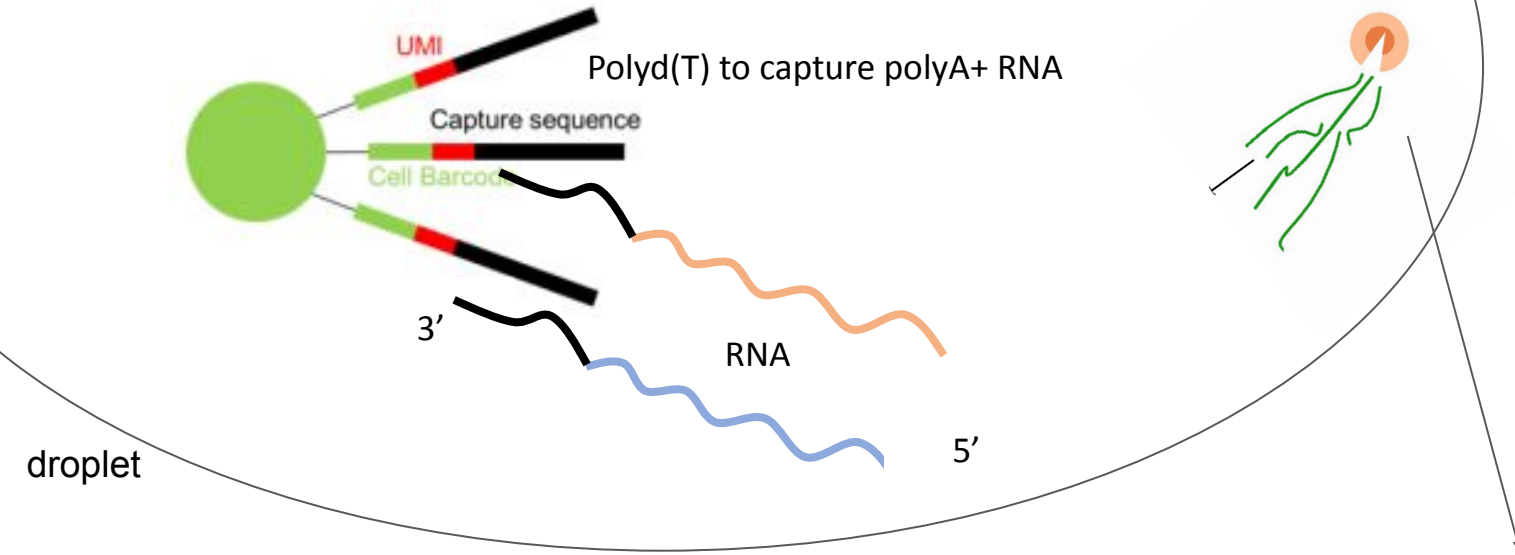
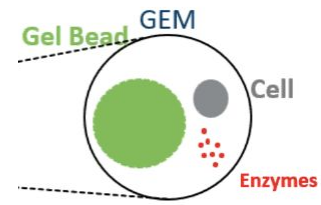
Cell Barcode (16bp) = sequence specific to each bead (so each cell)

UMI (12 bp) = sequence specific to each molecule : **U**nique **M**olecular **I**dentifier

Single-cell transcriptomics with 10X genomics technology

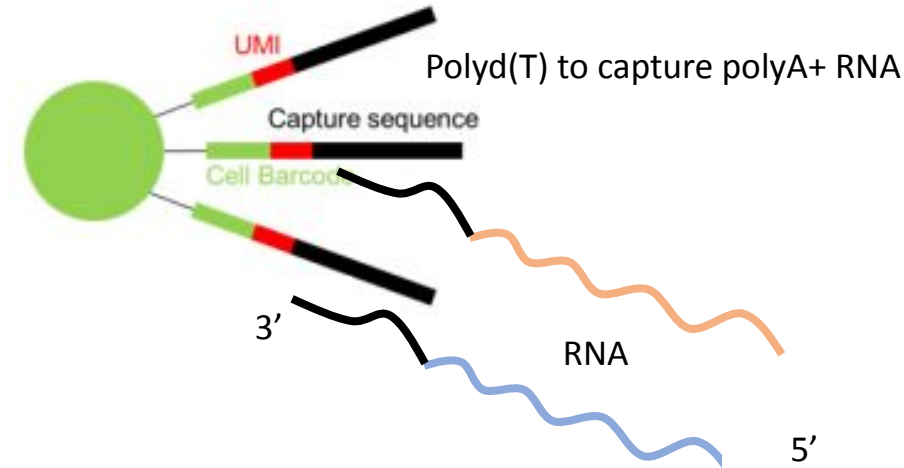


scRNA-seq with 10X targets 3' end of the transcripts

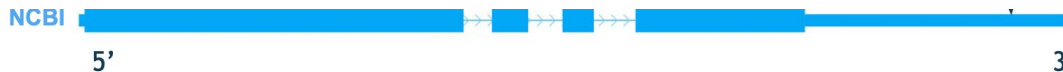
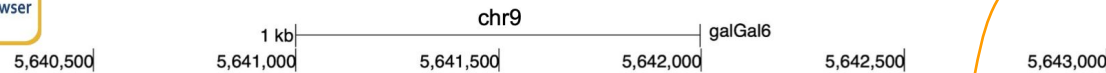


The cell is then lysed within the droplet. RNA is released (but contained in the droplet). polyA+ RNA are captured from the 3' end on the polyd(T) sequence

scRNA-seq with 10X targets 3' end of the transcripts



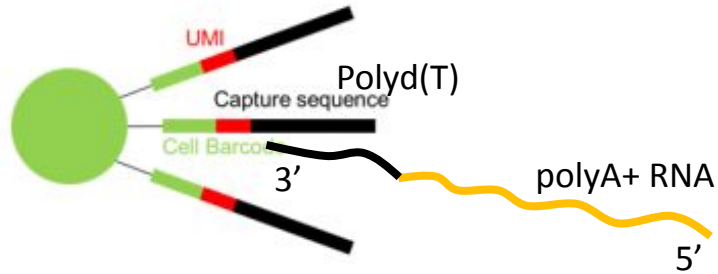
The signal is biased towards the 3' end of the transcript (more about that tomorrow)



10X scRNA-seq sequenced in short reads



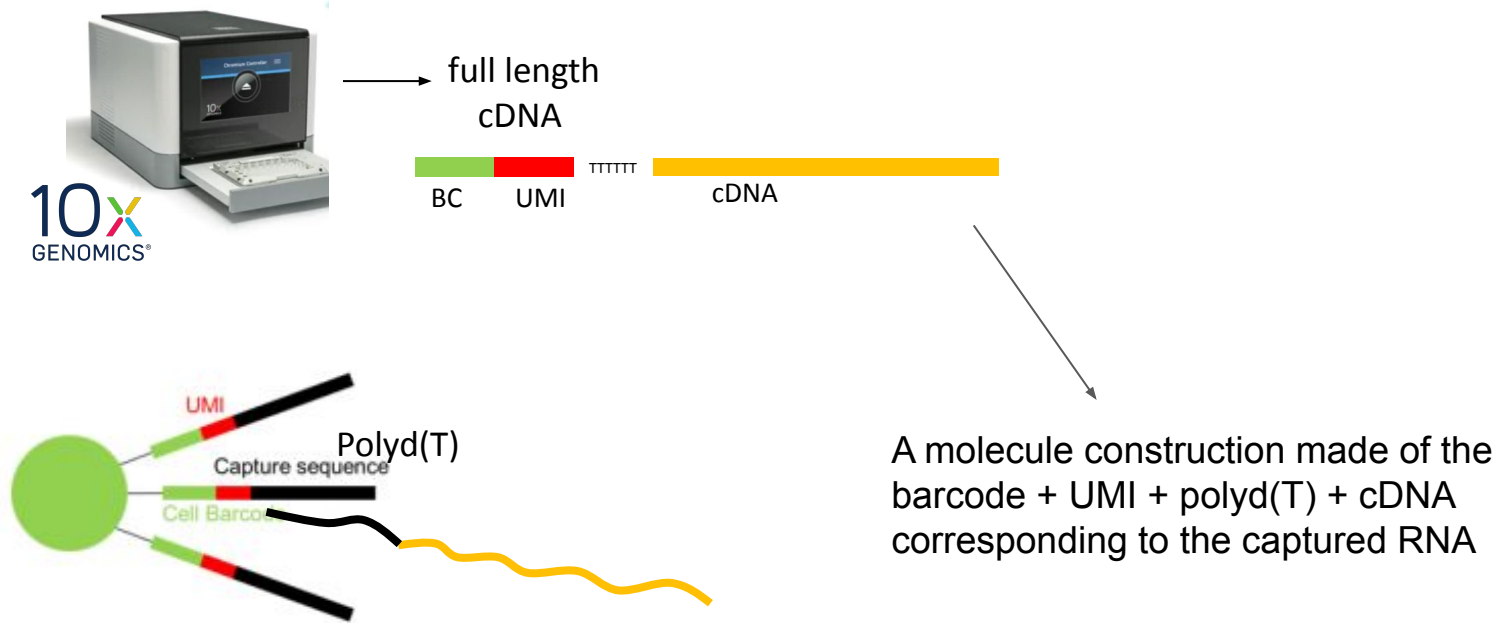
At the end of this step, what actually comes out of the 10X Genomics device ?



Cell Barcode = sequence specific to each cell

UMI = sequence specific to each molecule : **U**nique **M**olecular **I**dentifier

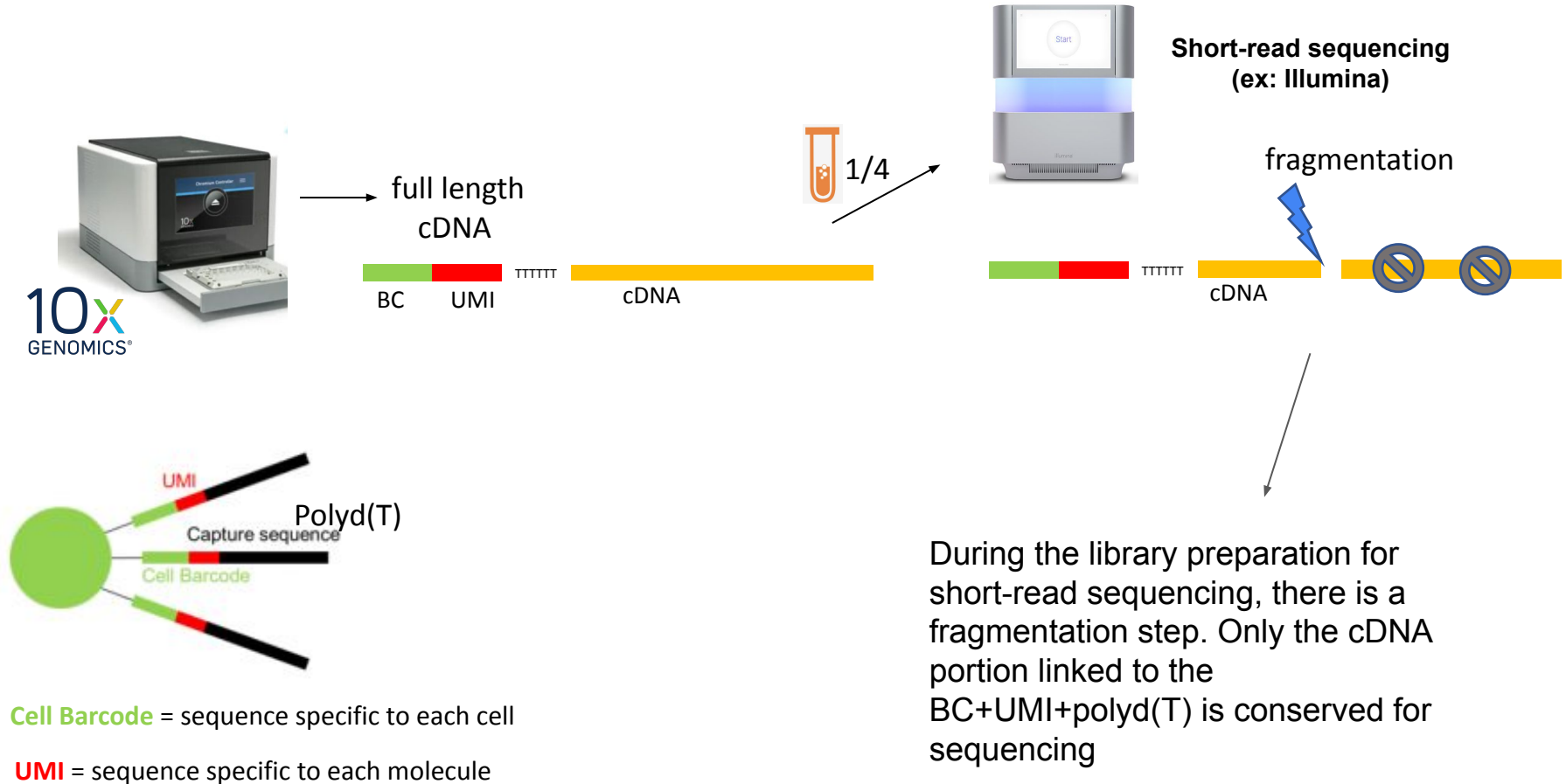
10X scRNA-seq sequenced in short reads



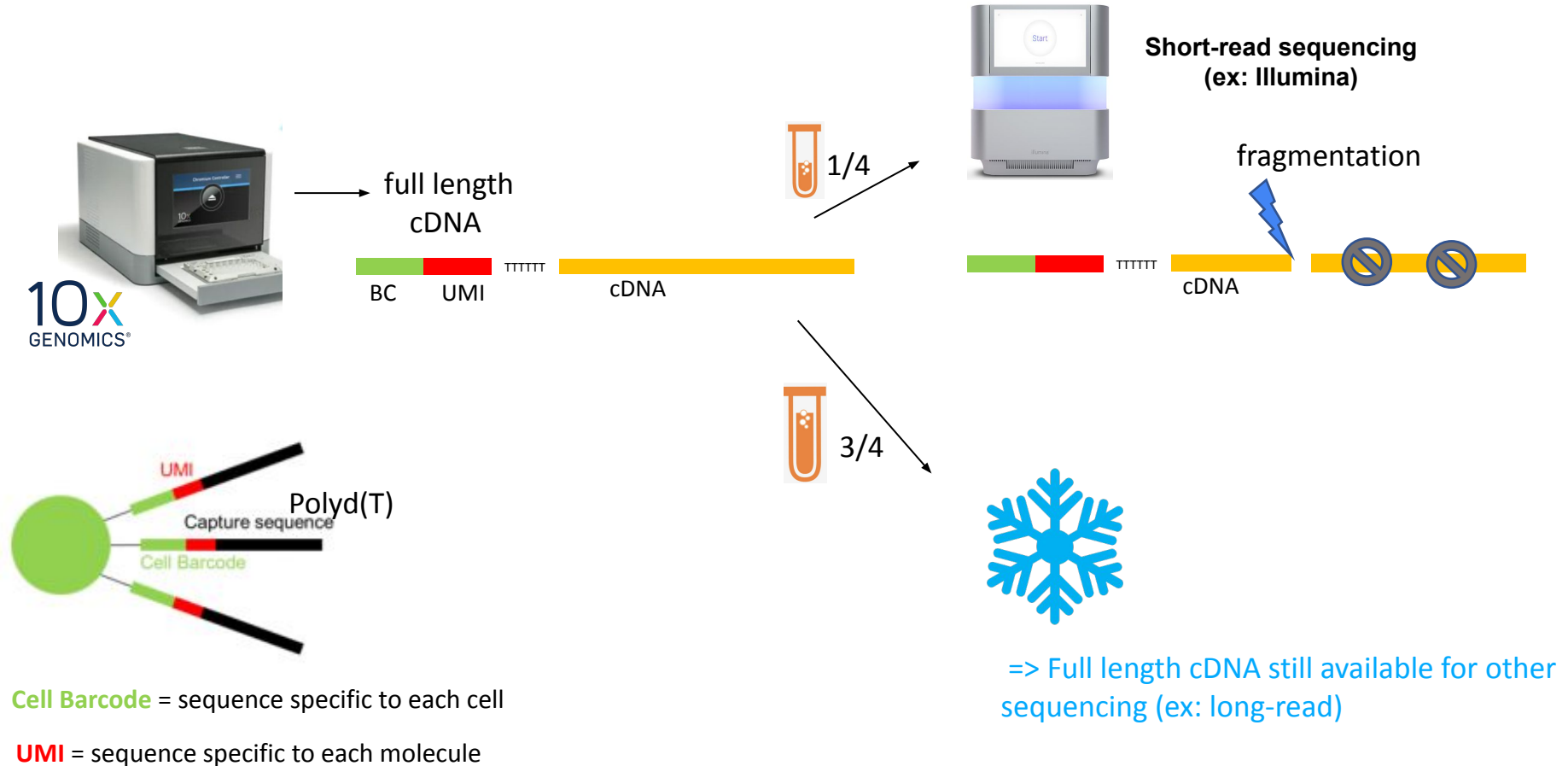
Cell Barcode = sequence specific to each cell

UMI = sequence specific to each molecule

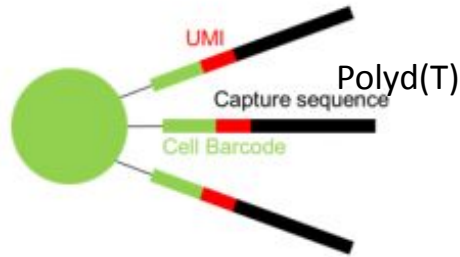
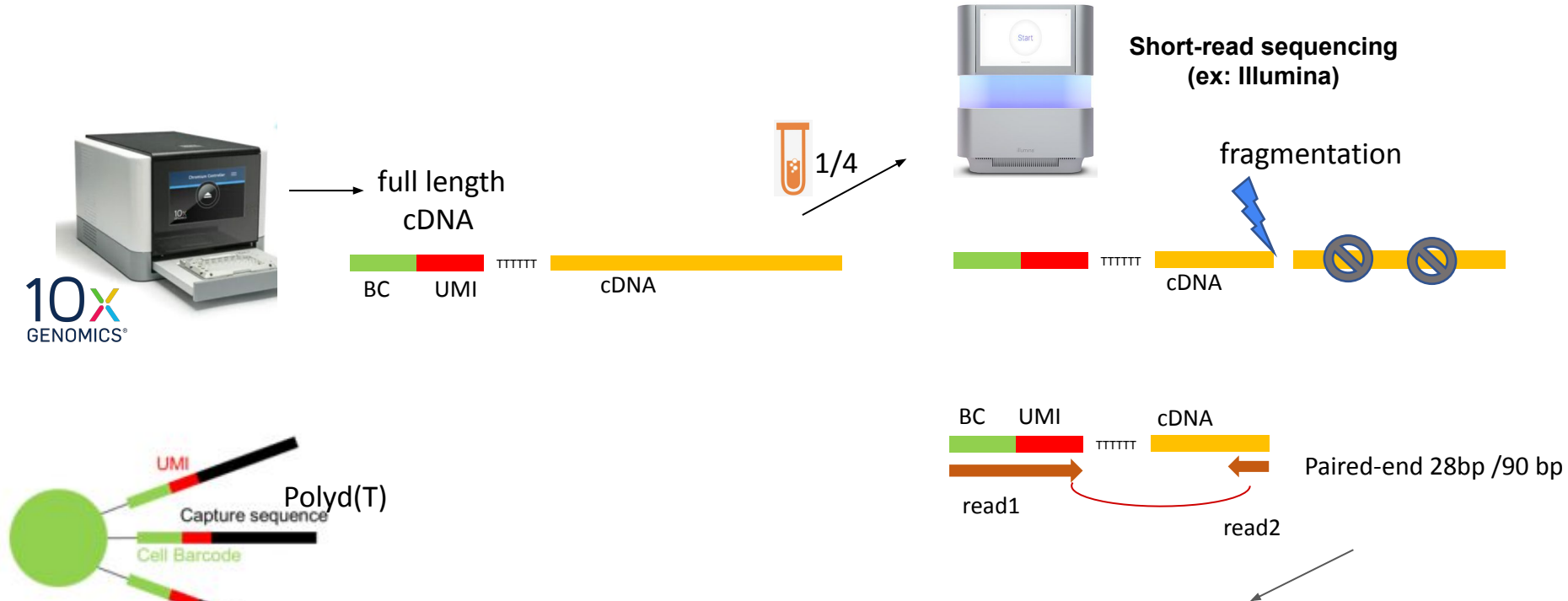
10X scRNA-seq sequenced in short reads



10X scRNA-seq sequenced in short reads



10X scRNA-seq sequenced in short reads



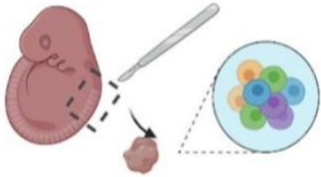
Cell Barcode = sequence specific to each cell

UMI = sequence specific to each molecule

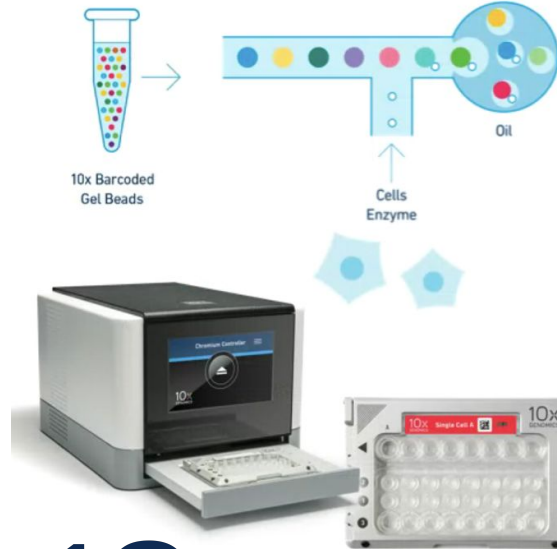
The library is sequenced in paired-end. The read 1 contains the BC+UMI (28bp). The read2 contains a 90bp portion of cDNA. Only read2 corresponds to genomic/biological DNA. Read1 stems from synthetic molecules, not the transcriptome.

10X scRNA-seq in a nutshell

Tissue dissection + cell dissociation



Cell partitioning + mRNA capture

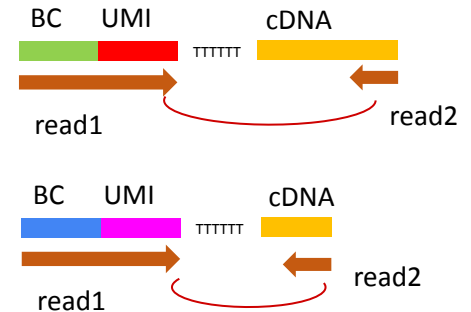


10x
GENOMICS®

Library preparation + sequencing



Paired-end 28bp / 90 bp



Biases/limitations of 10X Genomics technology

- Only the 3' end is sequenced (with short-read protocol + 3' kit)
- Cell size < 30um otherwise clog microfluidic channels
- 30% polyA+ transcripts captured per cell
- A droplet may contain 2 cells (= doublet)
- Some cell sub-population may be completely depleted/unfound

Considerations on experiments

- **Fresh cells** : time between dissociation and 10X experiment should be <30min, otherwise cells start to die and result in RNAs wrongly assigned to cells (RNA “soup”) and many expressed genes linked to cell death
- **Frozen cells** : does not work on all cells
- **FFPE** : only in human + mouse, restricted to certain tissues
- **Dissociation + Fixation** with ACME protocol (acetic acid + methanol + glycerol): requires optimisation but successful on exotic species (GenomiqueENS)
- Charge a bit more cells (25,000)
- Many tests have been done on **PBMCs** (immune cells) that are natively dissociated. Results do not necessarily reproduce on cells dissociated from tissues
- **Q&A** section of 10X website is very informative :
<https://kb.10xgenomics.com/hc/en-us/categories/360000149952-Single-Cell-Gene-Expression>

Bioinformatics analysis of 10X Genomics scRNA-seq dataset



Which result file(s) did you obtain from the sequencing core facility ?

Bioinformatics analysis of 10X Genomics scRNA-seq dataset



Cell Ranger · pbmc_1k_v3 · Peripheral blood mononuclear cells (PBMCs) from a healthy donor

SUMMARY ANALYSIS

Estimated Number of Cells

1,222

Mean Reads per Cell

54,502

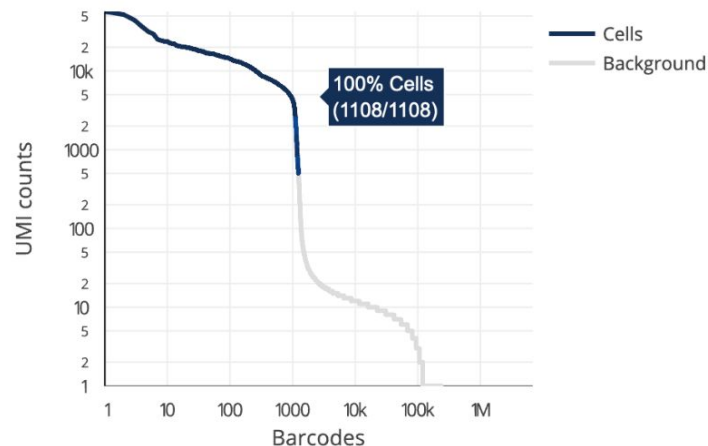
Median Genes per Cell

1,919

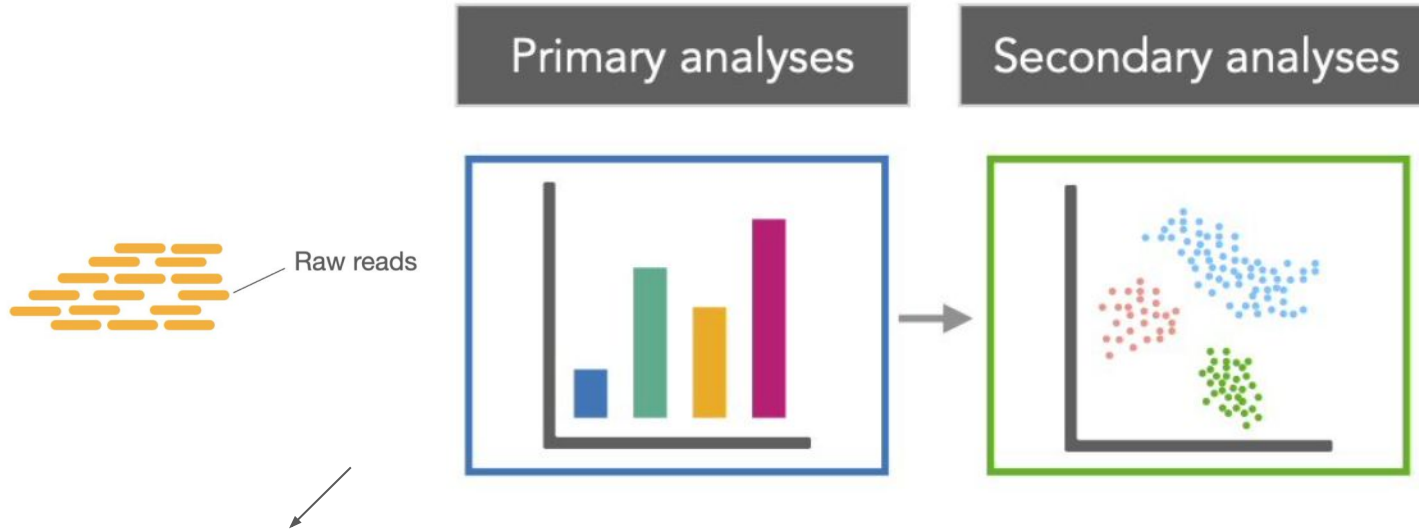
Sequencing

| | |
|-----------------------|------------|
| Number of Reads | 66,601,887 |
| Valid Barcodes | 97.4% |
| Sequencing Saturation | 70.8% |

Cells

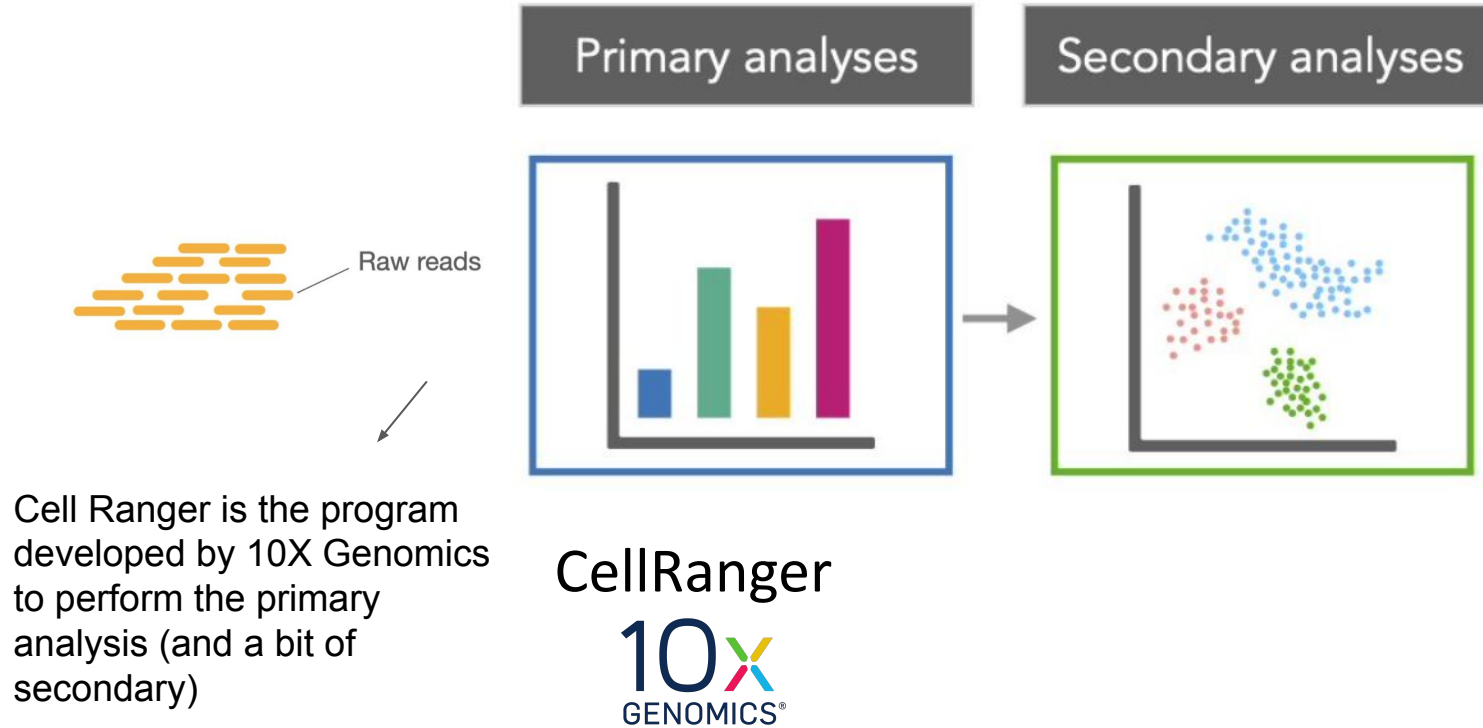


Bioinformatics analysis of 10X Genomics scRNA-seq dataset



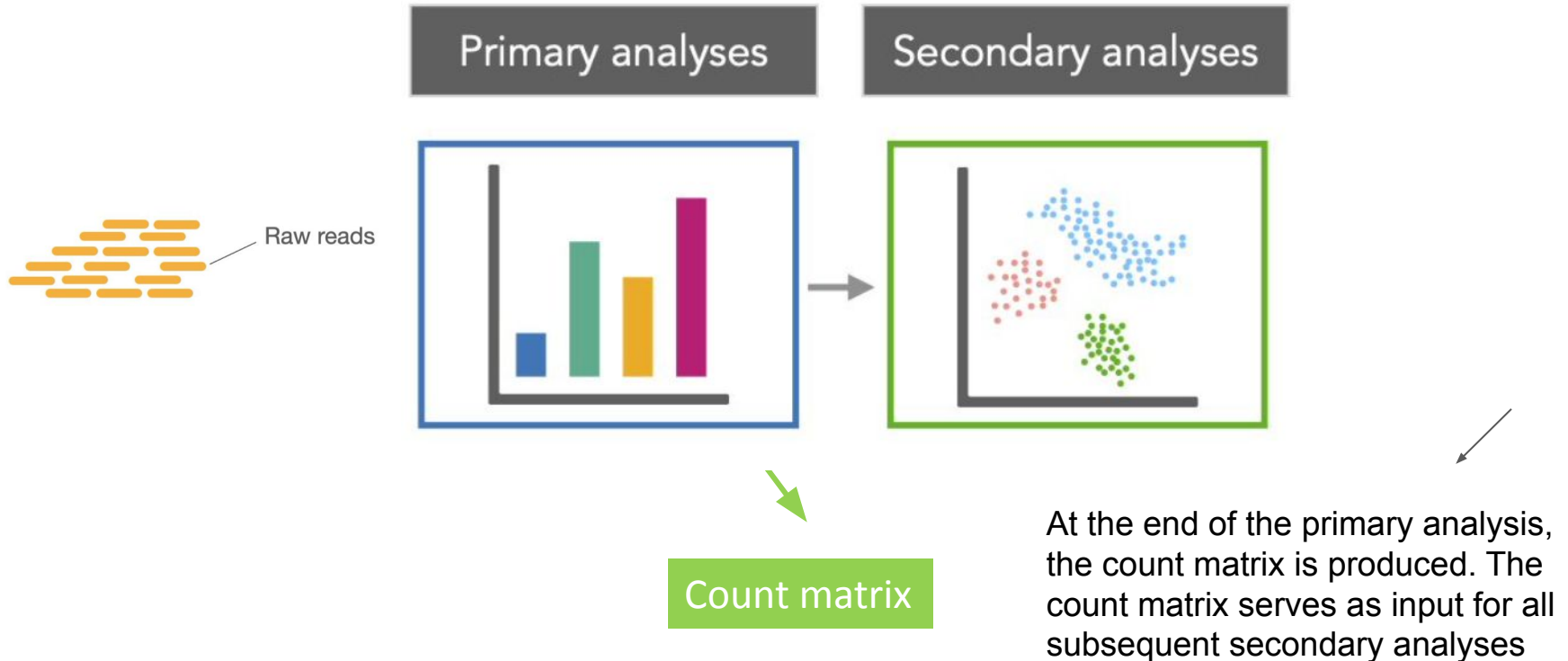
Raw data are the sequence reads. Then the bioinformatics analysis are in 2 phases :
Primary (= preprocessing) and secondary

Bioinformatics analysis of 10X Genomics scRNA-seq dataset

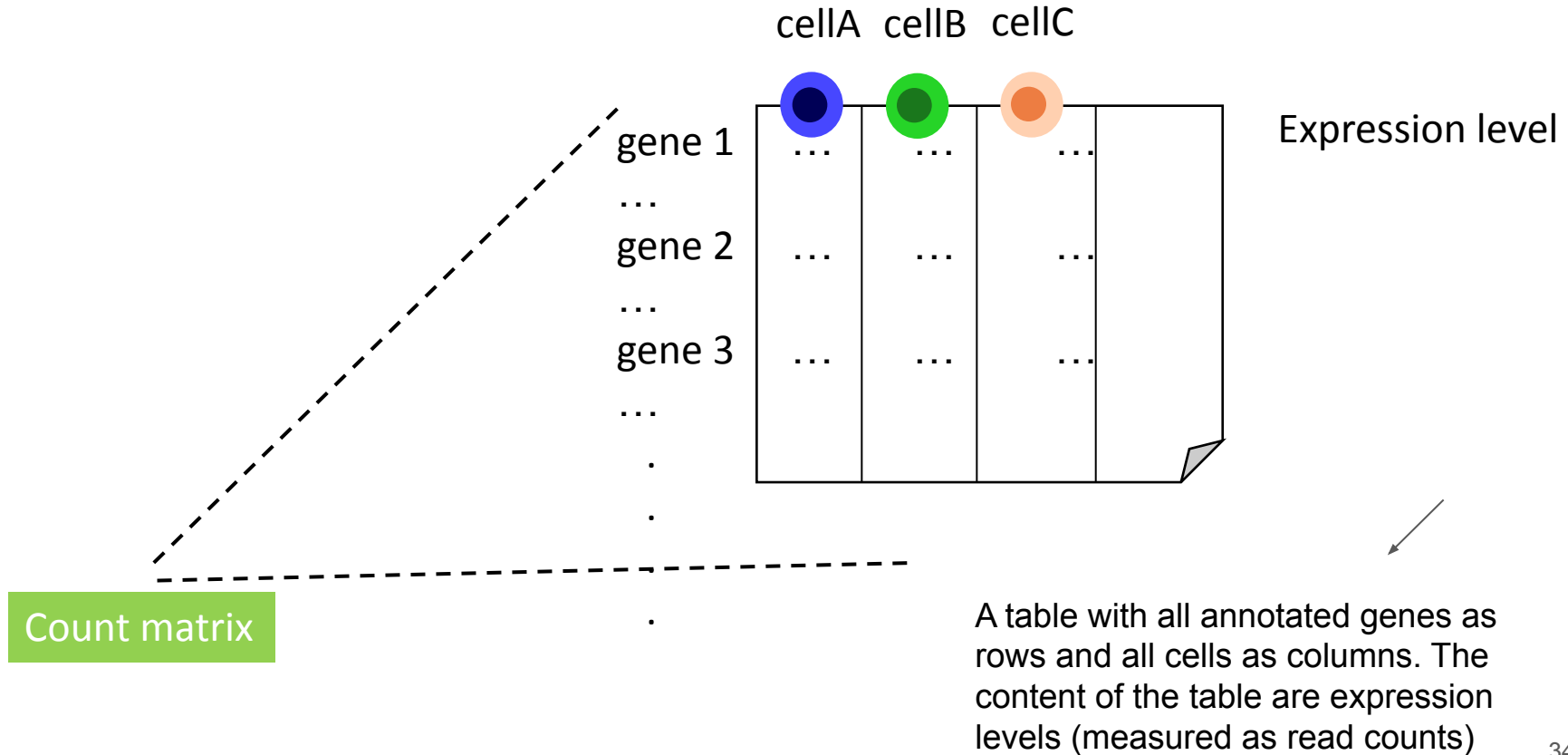


Cell Ranger is the program developed by 10X Genomics to perform the primary analysis (and a bit of secondary)

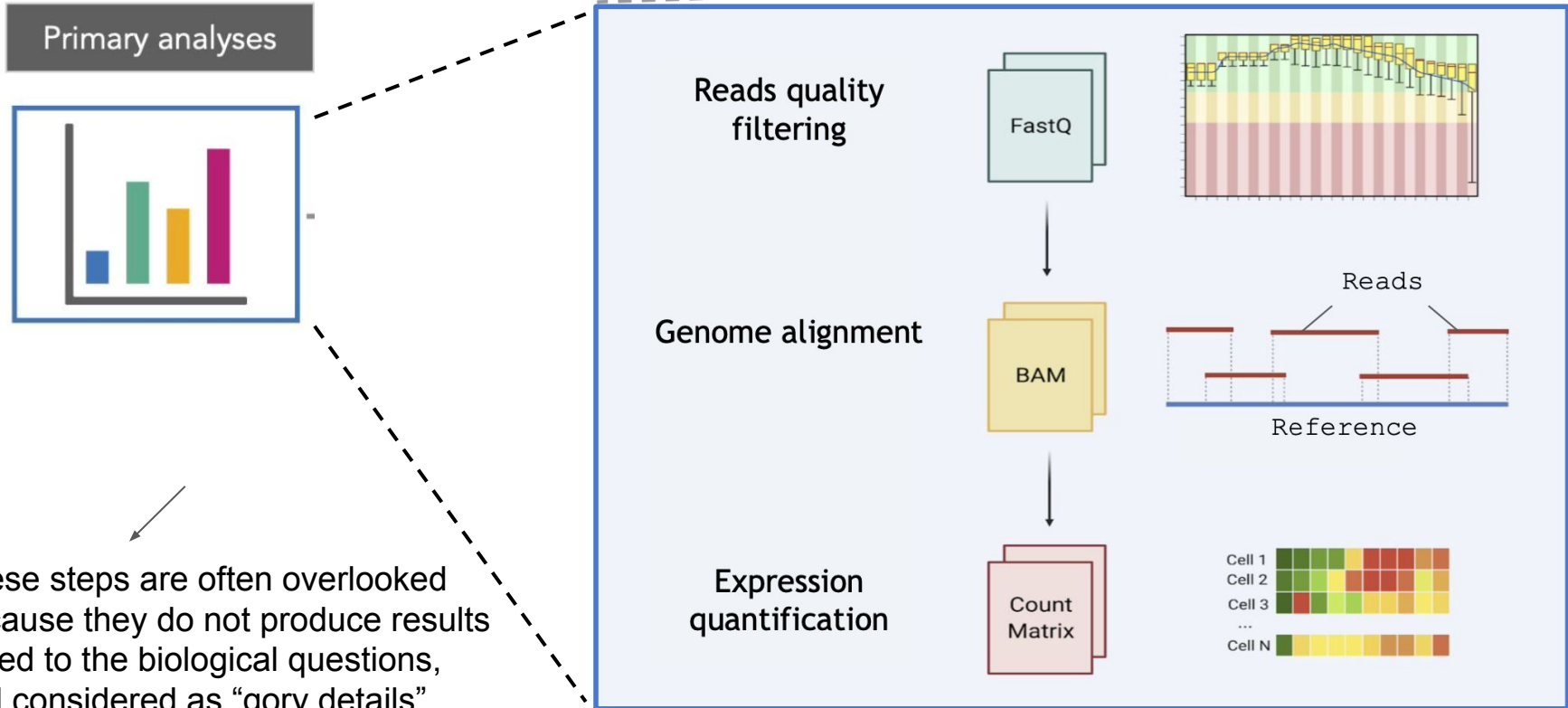
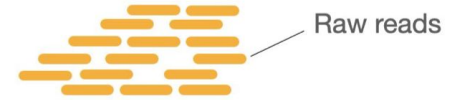
Bioinformatics analysis of 10X Genomics scRNA-seq dataset



What is a count matrix ?



The processing steps that are often overlooked



Primary analyses : Reads



Primary analyses



- Results starts by a **BCL** file (raw base calling from the sequencer). This file needs to be treated to produce the **FASTQ** files containing the reads
- This steps is done by the program bcl2fastq from Illumina (step “mkfastq” in CellRanger)

this is a detail, we just indicate it here in case you read further about Cell Ranger and step upon the notion of BCL

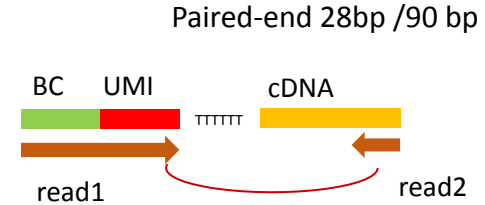
Primary analyses : Reads quality checking



Primary analyses



- 2 FASTQ files :
 - one contains all the read1
 - one contains all the read2





Example dataset from 10X Genomics



- **Dataset name** : pbmc_1k_v3 => 1000 human peripheral blood mononuclear cells (PBMCs) in human, freely available from 10X genomics website
- **2 files** :
 - pbmc_1k_v3_S1_L001_R1_001.fastq.gz
 - pbmc_1k_v3_S1_L001_R2_001.fastq.gz



10X provides other public datasets, for each application, chemistry...

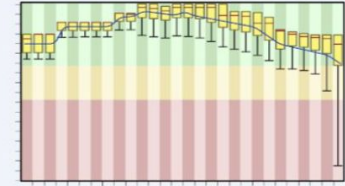
Primary analyses : Reads quality checking

Primary analyses

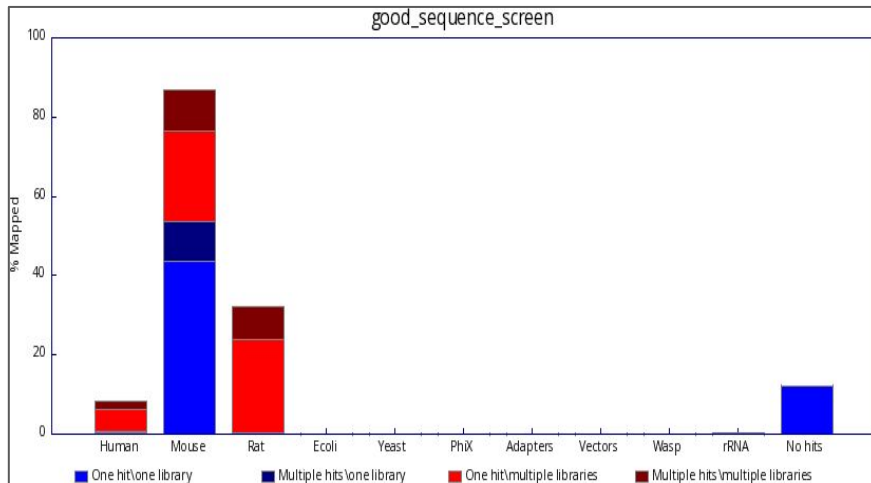


Reads quality
filtering

FastQ



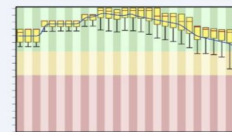
- As for any other NGS experiment, check the quality of the reads with FASTQC.
- FastqScreen enables to check for contaminations with other organisms
- These steps are usually done by the sequencing core facility, ask for these results if not provided





Example dataset from 10X Genomics

Reads quality
filtering



FastQC Report

Thu 6 Oct 2022
2022-006sc_S1_L001_R1_001.fastq.gz

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](#)

Basic Statistics

| Measure | Value |
|-----------------------------------|------------------------------------|
| Filename | 2022-006sc_S1_L001_R1_001.fastq.gz |
| File type | Conventional base calls |
| Encoding | Sanger / Illumina 1.9 |
| Total Sequences | 494792037 |
| Sequences flagged as poor quality | 0 |
| Sequence length | 28 |
| %GC | 49 |



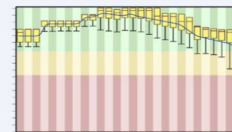
Read1 : 28bp
494M reads



Example dataset from 10X Genomics

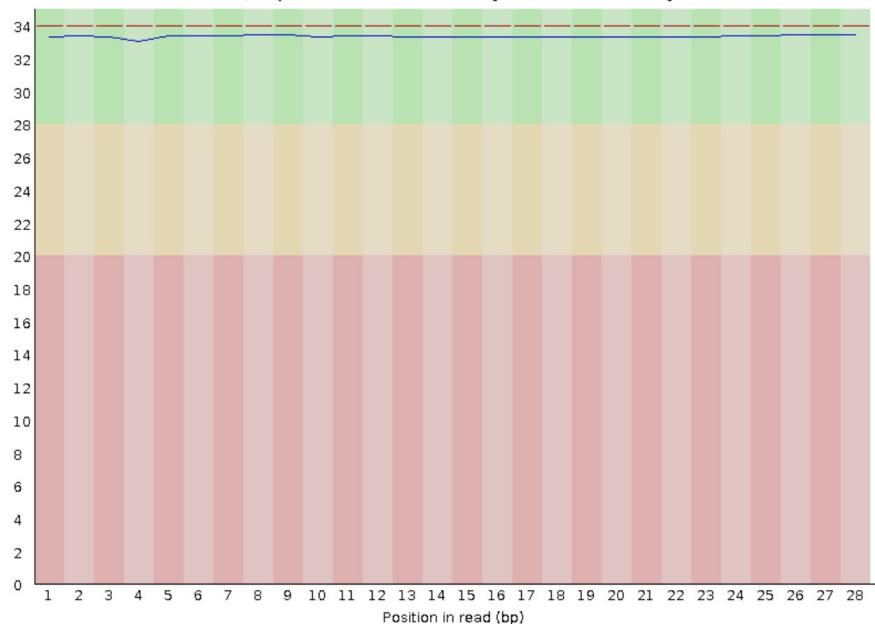
Reads quality
filtering

FastQ



✓ Per base sequence quality

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



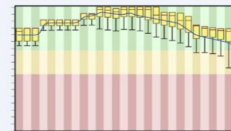
quality is excellent



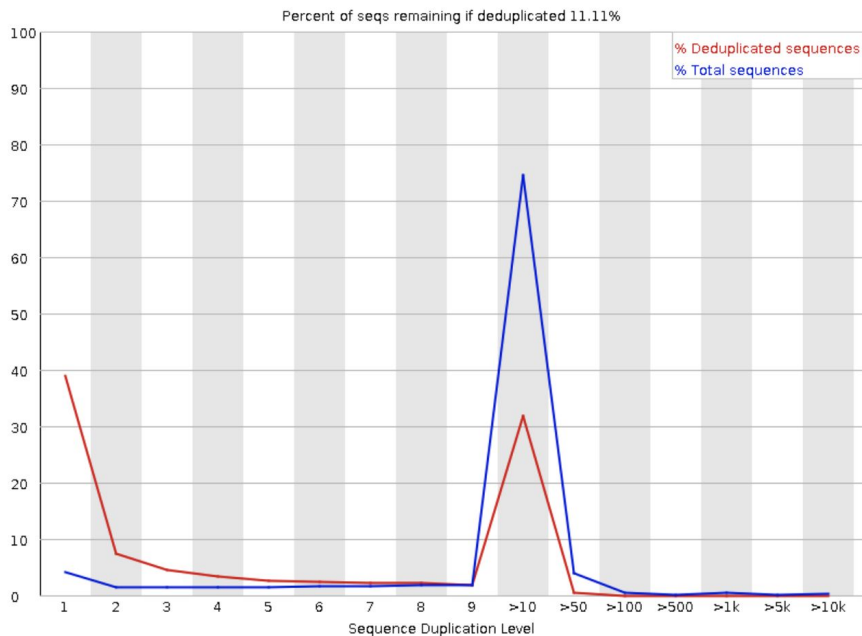
Example dataset from 10X Genomics

Reads quality
filtering

FastQ



Sequence Duplication Levels



normal to have
duplication level
because some
BC+UMI have
amplification biases



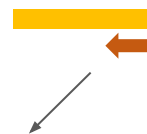
Example dataset from 10X Genomics



Basic Statistics

| Measure | Value |
|-----------------------------------|------------------------------------|
| Filename | 2022-006sc_S1_L001_R2_001.fastq.gz |
| File type | Conventional base calls |
| Encoding | Sanger / Illumina 1.9 |
| Total Sequences | 494792037 |
| Sequences flagged as poor quality | 0 |
| Sequence length | 90 |
| %GC | 46 |

cDNA



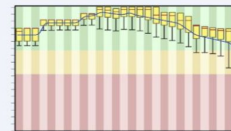
Read2 : 90bp
494M reads



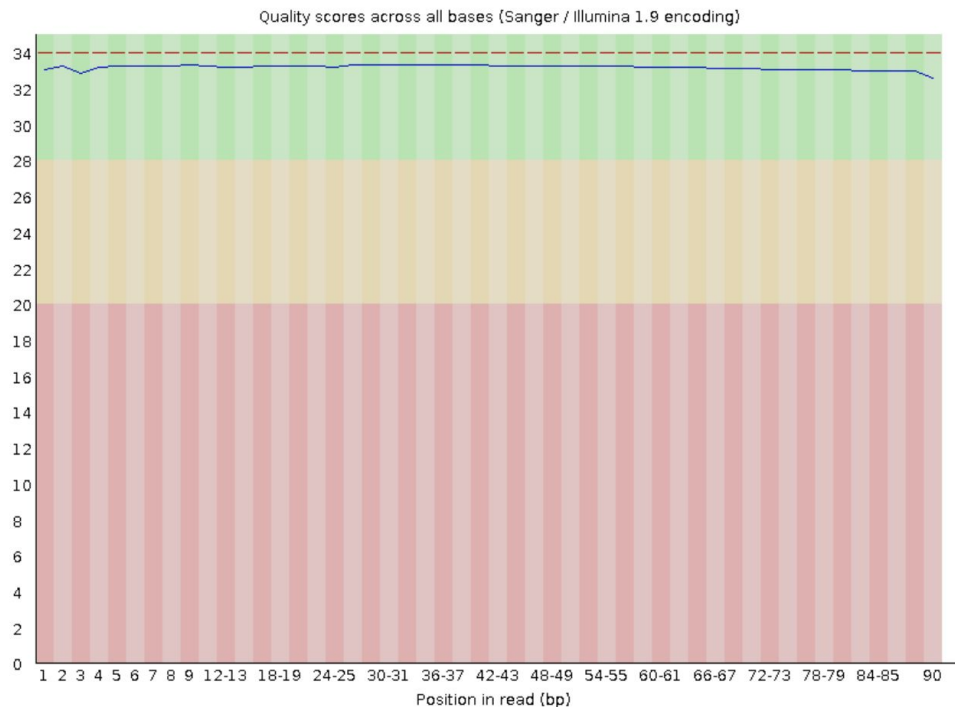
Example dataset from 10X Genomics

Reads quality
filtering

FastQ



Per base sequence quality



quality is excellent



Example dataset from 10X Genomics

In **CellRanger** report

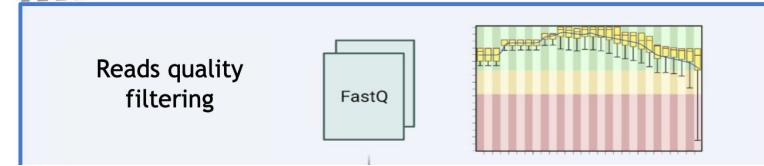


Sequencing

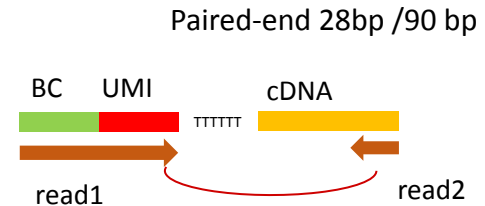
| | |
|---------------------------|------------|
| Number of Reads | 66,601,887 |
| Valid Barcodes | 97.4% |
| Sequencing Saturation | 70.8% |
| Q30 Bases in Barcode | 94.1% |
| Q30 Bases in RNA Read | 90.2% |
| Q30 Bases in Sample Index | 91.1% |
| Q30 Bases in UMI | 92.7% |

check the
“sequencing” section
of the report. The
Q30 means “very
high quality of bases”

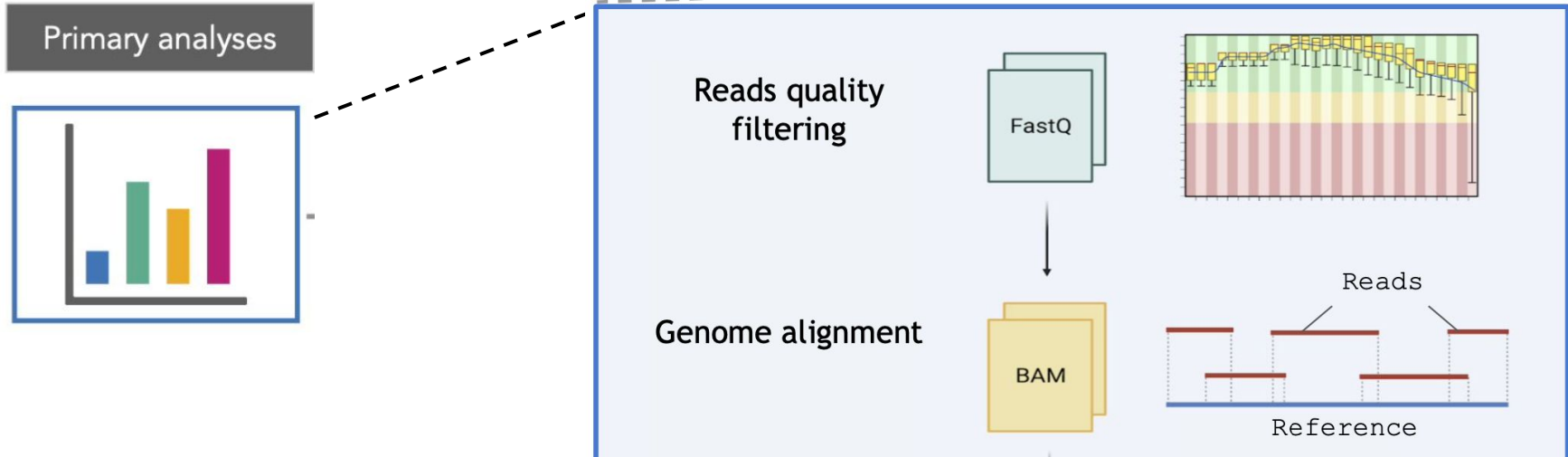
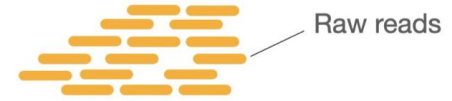
Primary analyses : Reads quality checking



- Make sure read1 is of high quality because it contains the BC + UMI, later used to trace back the cell from which originates the RNA
- Ns and highly repeated sequences would impair read assignment.
- Any wrong base => lost read and barcode



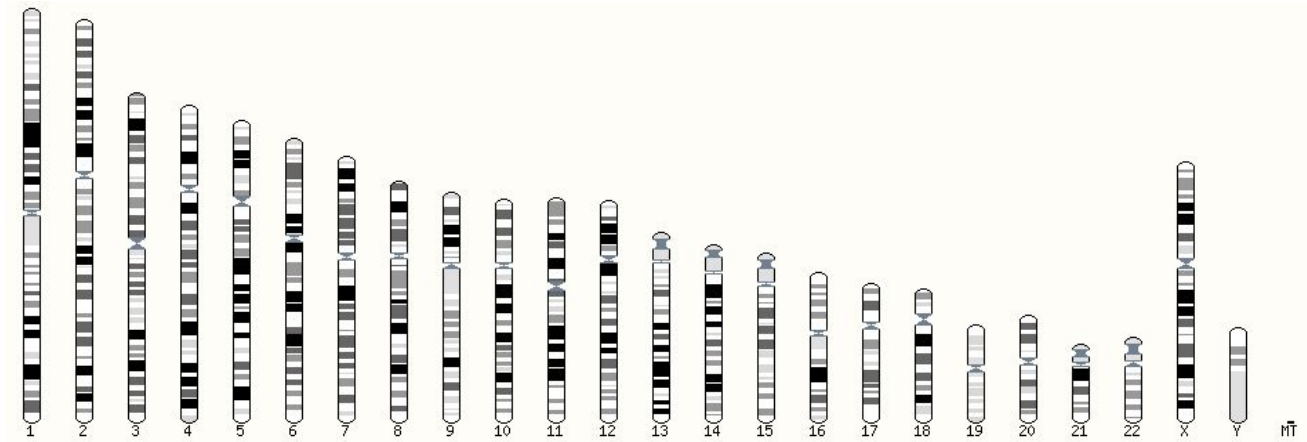
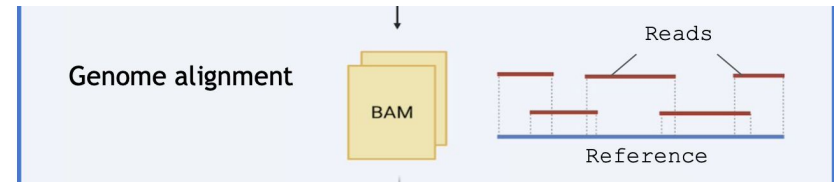
Primary analyses : Mapping



- Read1 and Read2 are then treated separately
- **Read2** corresponds to genomic sequence => **mapping step** (=infer the position on the genome from which the read originates)



Primary analyses : Mapping

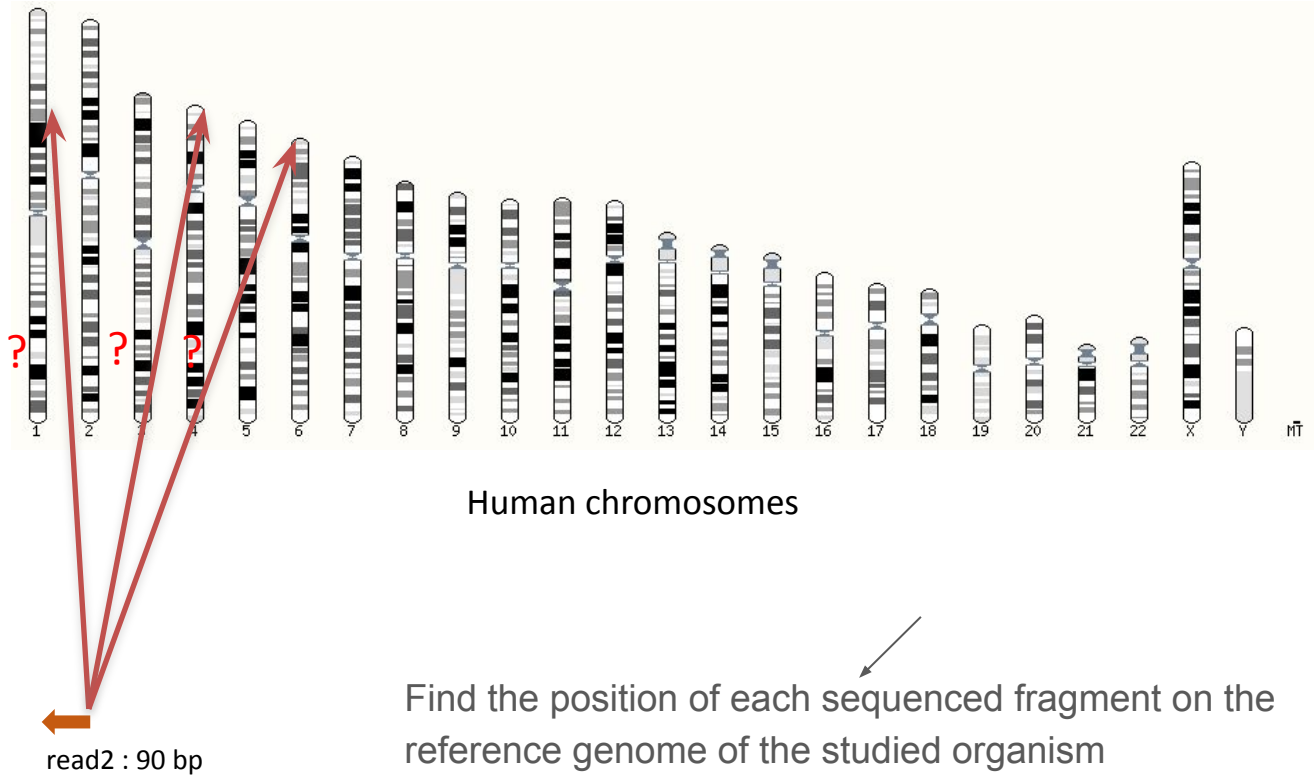
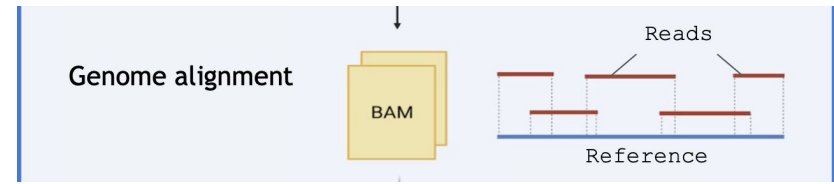


Human chromosomes

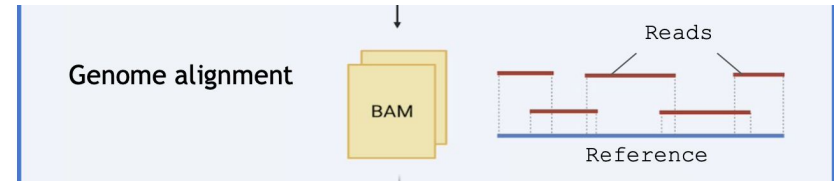
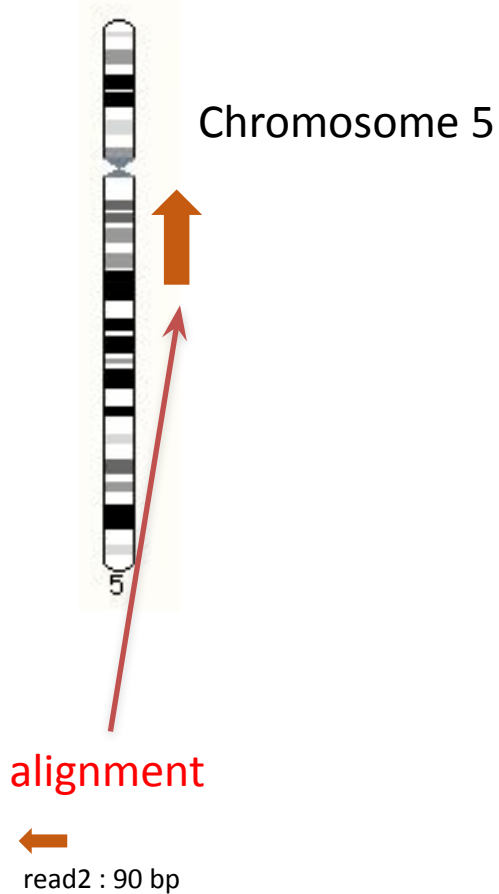


read2 : 90 bp

Primary analyses : Mapping

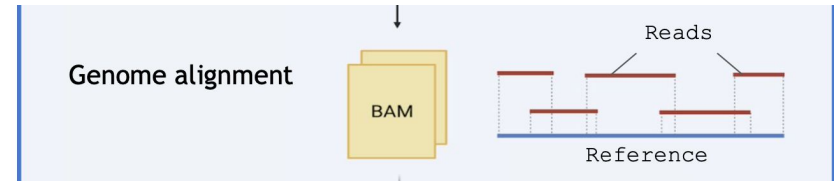
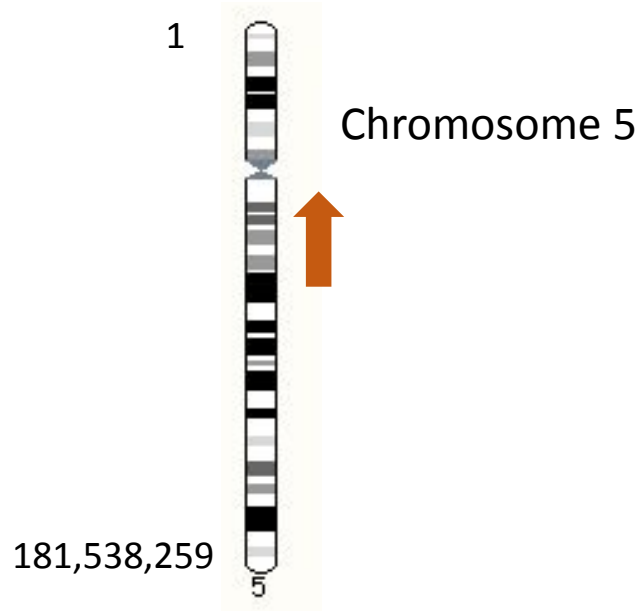


Primary analyses : Mapping



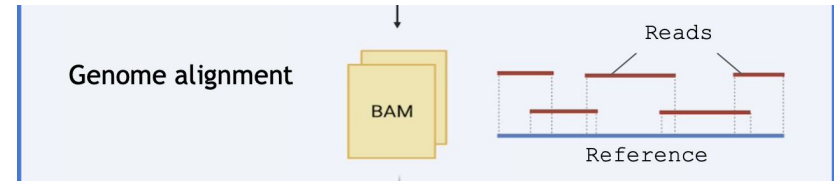
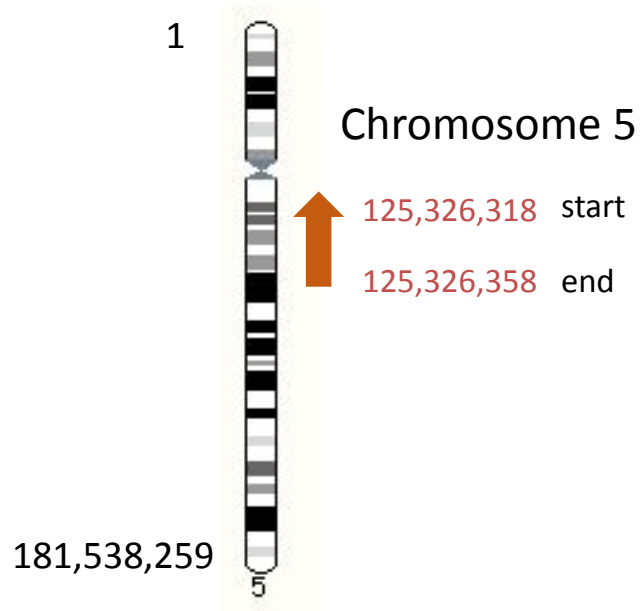
The best alignment is found for this read over the whole genome. Here it is on chromosome 5

Primary analyses : Mapping



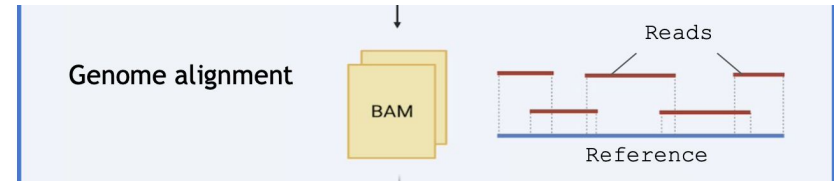
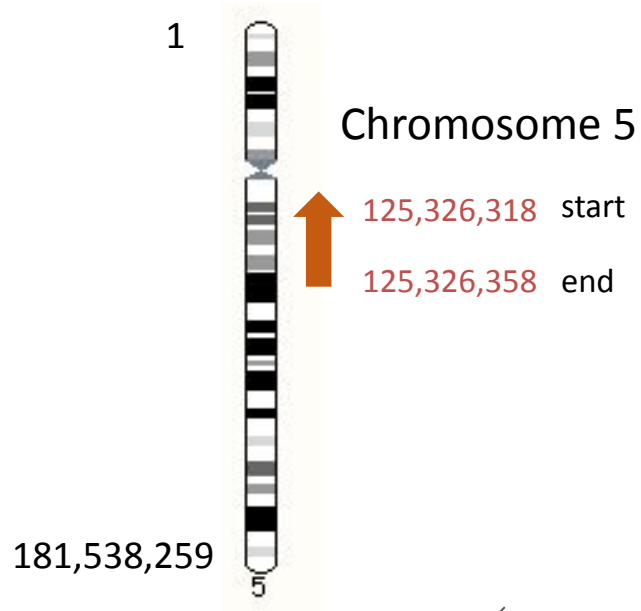
To precise the localisation of the best alignment, a coordinate system is used. First, each position of the chromosome has a particular value, corresponding to its distance from the beginning of the chromosome

Primary analyses : Mapping



The region of alignment has a **start** and **end** position + strand orientation

Primary analyses : Mapping



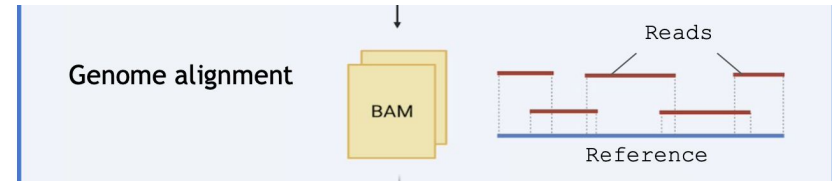
Genomic coordinates:

chr5 125326318 125326358 -



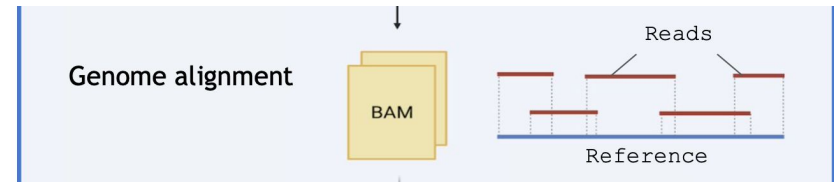
The genomic coordinates is like “GPS coordinates” to locate regions on a genome. The format is :
chromosome start end strand

Primary analyses : Mapping

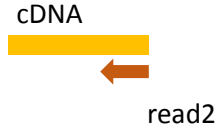
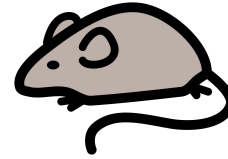


- The mapping step enables to obtain the genomic coordinates of all reads2 for which an alignment has been found.
- The output file is in **BAM** format
- Not all reads can be aligned (contaminations, differences between the sample and reference genome, ...)
- Programs that perform this mapping step are often called “mappers”

Primary analyses : Mapping



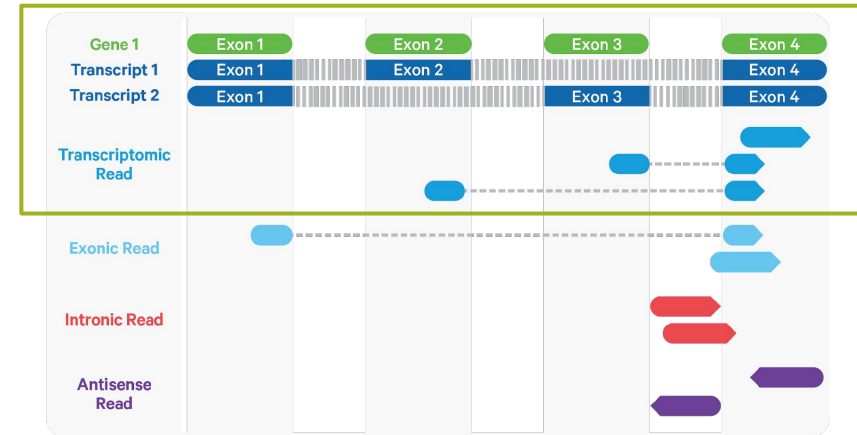
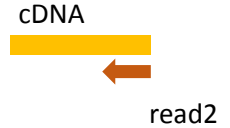
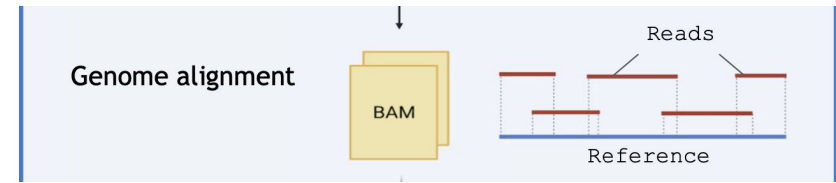
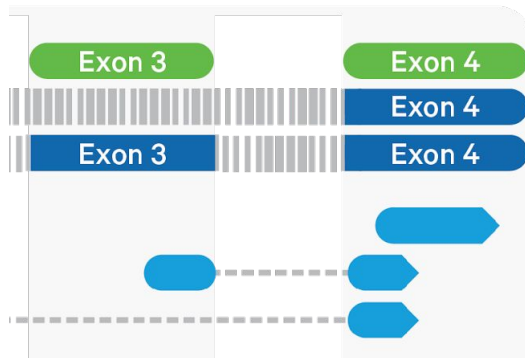
- CellRanger internally uses **STAR** as the program to align the reads on the reference genome
- The reference genome must be provided in the form of an **index**
- Ready-to-use genomes index:
 - human (hg19, GRCh38)
 - mouse (mm10)
 - both (xenografts)
- For other organisms :
 - Use the genome in FASTA format
 - convert it with *cellranger mkgtf* and *cellranger mkref*.
- If you use some specific sequences (transgenes), don't forget to provide the sequence and rebuild the index ! (otherwise, no reads will be mapped to this region)



working with user-specific sequences or genomes requires more work because the genome index must be built (computer-intensive)

Primary analyses : Mapping

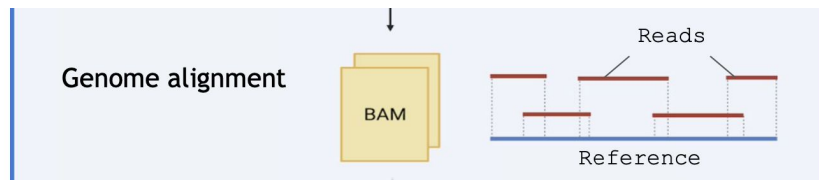
- STAR deals with RNA splicing, a read can be artificially “cut” to map to distant regions from which it originates (=different exons)





Example dataset from 10X Genomics

In CellRanger report



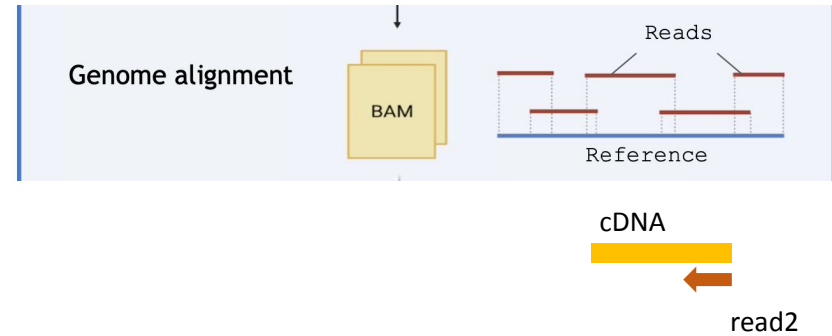
Mapping

| | |
|--|-------|
| Reads Mapped to Genome | 95.4% |
| Reads Mapped Confidently to Genome | 92.4% |
| Reads Mapped Confidently to Intergenic Regions | 4.8% |
| Reads Mapped Confidently to Intronic Regions | 31.1% |
| Reads Mapped Confidently to Exonic Regions | 56.5% |
| Reads Mapped Confidently to Transcriptome | 53.7% |
| Reads Mapped Antisense to Gene | 1.0% |

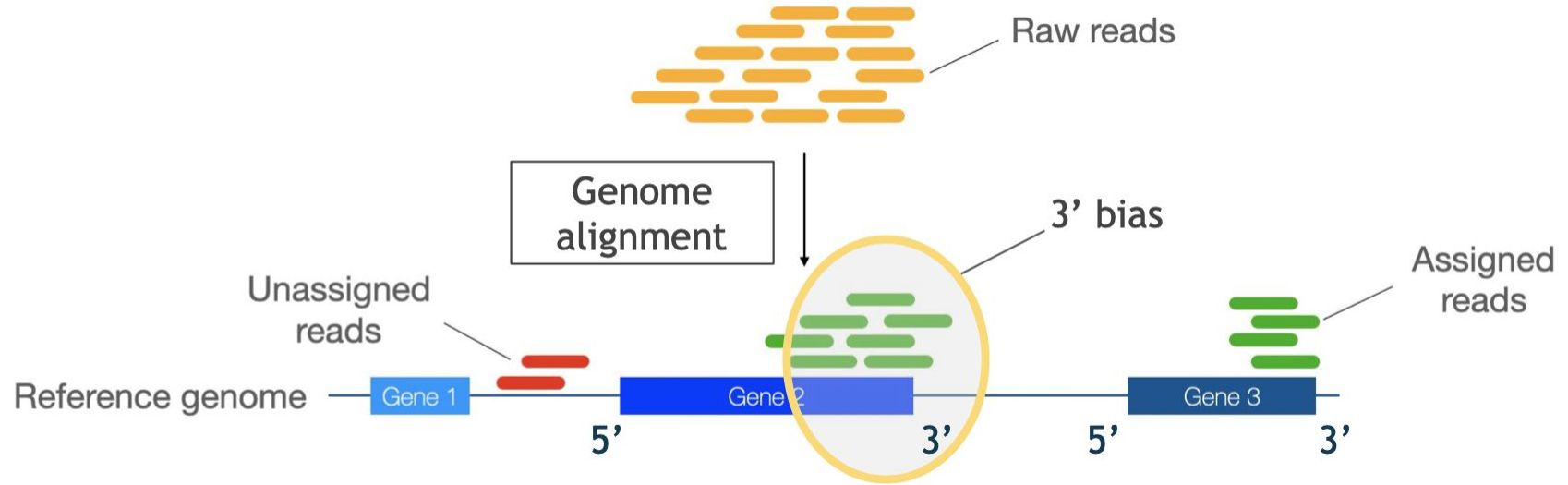
It is normal to have <100% reads aligned to the genome, because the reference genome is not exactly the genome of the studied sample. % will decrease with huge rearrangements (cancer or cell lines) or many SNPs (wild animals)

Primary analyses : Mapping

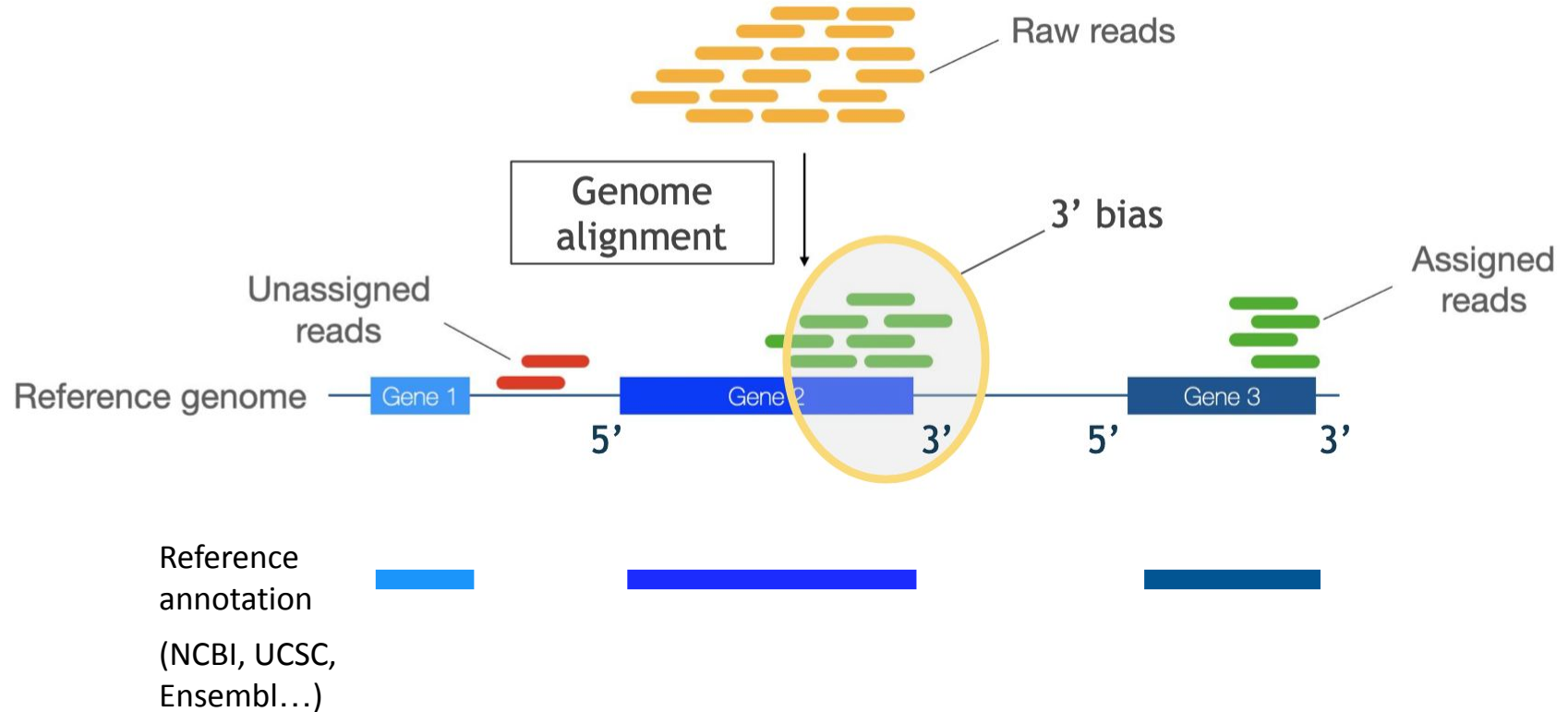
- Then, the **genome annotation** is used to assign the reads to genes
- Annotation is provided by genome portals (NCBI, Ensembl, UCSC) or consortiums of researchers working on a same organism
- genome annotation is generally provided as a file in the format **GFF** or **GTF**



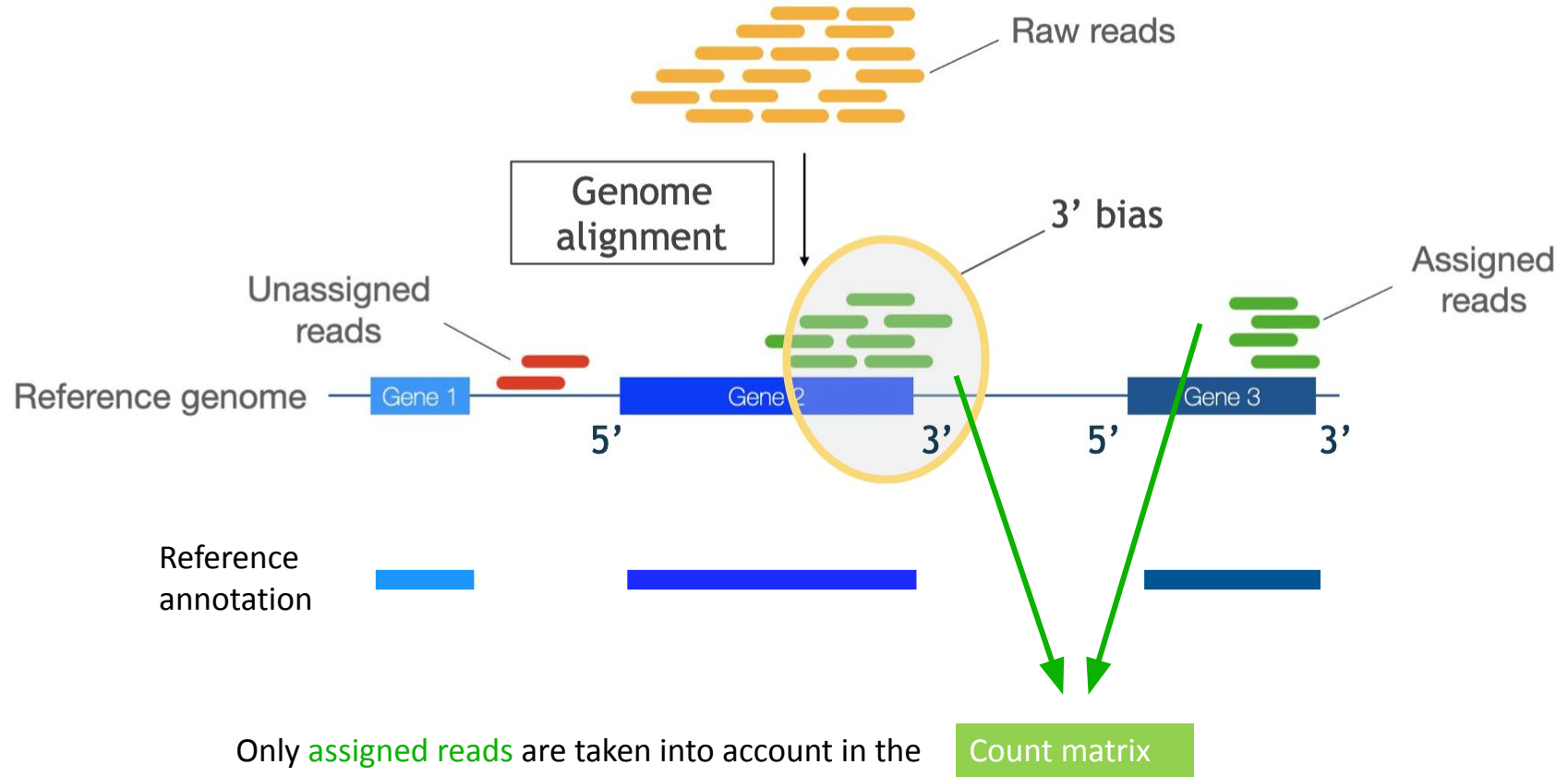
How is the genome reference annotation used ?



How is the genome reference annotation used ?



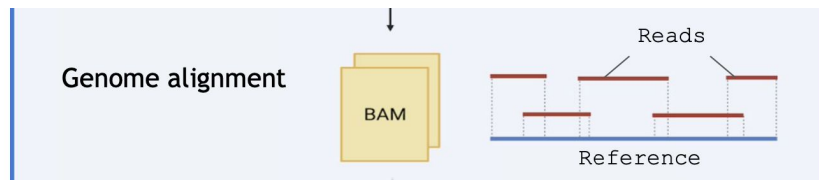
How is the genome reference annotation used ?





Example dataset from 10X Genomics

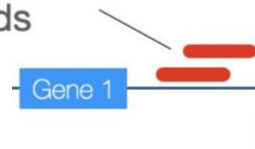
In CellRanger report



Mapping

| | |
|--|-------|
| Reads Mapped to Genome | 95.4% |
| Reads Mapped Confidently to Genome | 92.4% |
| Reads Mapped Confidently to Intergenic Regions | 4.8% |
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| Reads Mapped Confidently to Exonic Regions | 56.5% |
| Reads Mapped Confidently to Transcriptome | 53.7% |
| Reads Mapped Antisense to Gene | 1.0% |

Unassigned reads



Assigned reads






Important point on annotation

- Annotation is a crucial parameter (largely *underestimated*), as reads outside the annotated exons will not be taken into account !
- CellRanger will warn you on the report with the Alert below. In such cases, you need to visualise your signal in a genome browser (more on this tomorrow) and suspect the annotation may be problematic

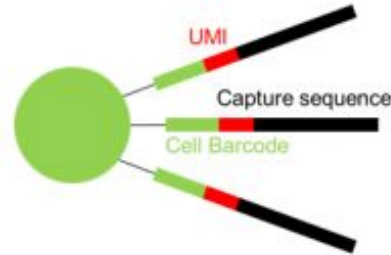
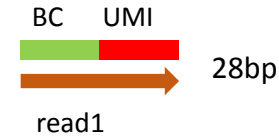
Alerts

The analysis detected  1 warning.

| Alert | Value | Detail |
|---|-------|---|
|  Low Fraction Reads Confidently Mapped To Transcriptome | 51.5% | Ideal > 30%. This can indicate use of the wrong reference transcriptome, a reference transcriptome with overlapping genes, poor library quality, poor sequencing quality, or reads shorter than the recommended minimum. Application performance may be affected. |

Primary analyses : barcode and UMI

- Read1 is made of BC + UMI
- Barcode and UMI are treated separately



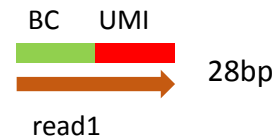
Reminder : barcode enables to trace back the read to the cell of origin ; UMI enables to distinguish each individual molecule

Cell Barcode (16bp) = sequence specific to each bead (so each cell)

UMI (12 bp) = sequence specific to each molecule : **U**nique **M**olecular **I**dentifier

Primary analyses : barcode

- Barcode is extracted (16bp)
- 10X provides a **whitelist** containing all possible barcodes used on the gel beads (~3 million barcodes for the v3 chemistry)
- All barcodes are compared to this whitelist
- Correction:** barcodes with 1 difference (1 mismatch) from the whitelist are corrected.
- Filtering:** keep only BC in the whitelist.



Expected barcodes in whitelist

AACGTCGTGAGTGCAT

CCGCTGACTGAGTTCA

...

Observed barcodes in R1

AACGTCGTGAGTGCAT

AACGTCGT**C**AGTGCAT

AAGG**T**GGTGAGTGGCAT

CCGCTGACTGAGTTCA

GGGCTGAGGTGGGTCA

...



Example dataset from 10X Genomics

In CellRanger report

Sequencing

| | |
|-----------------|------------|
| Number of Reads | 66,601,887 |
|-----------------|------------|

| | |
|----------------|-------|
| Valid Barcodes | 97.4% |
|----------------|-------|

| | |
|-----------------------|-------|
| Sequencing Saturation | 70.8% |
|-----------------------|-------|

| | |
|----------------------|-------|
| Q30 Bases in Barcode | 94.1% |
|----------------------|-------|

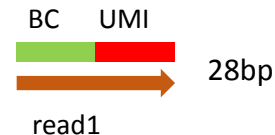
| | |
|-----------------------|-------|
| Q30 Bases in RNA Read | 90.2% |
|-----------------------|-------|

| | |
|---------------------------|-------|
| Q30 Bases in Sample Index | 91.1% |
|---------------------------|-------|

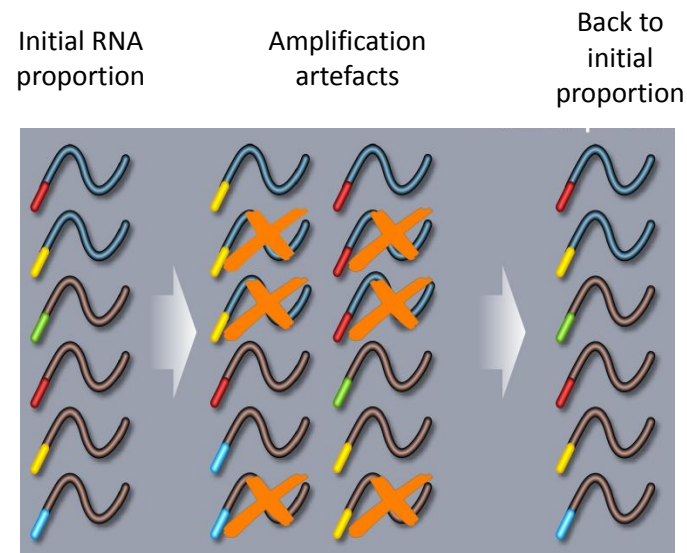
| | |
|------------------|-------|
| Q30 Bases in UMI | 92.7% |
|------------------|-------|

↙
% of valid barcodes
is indicated in the
report

Primary analyses : UMI

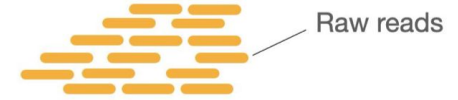


- UMI is extracted (12bp)
- UMIs are randomised sequences, there is no whitelist
- **Correction:** UMIs with 1 difference (1 mismatch) from a higher-count UMI are corrected to the higher count UMI if they share a cell barcode.
- **Filtering:** remove incorrect UMIs:
 - homopolymers (e.g. AAAAAAAAAA)
 - Contains 1 or several N
 - contains any base with BASEQ < 10



UMI aims at correcting amplification artefacts (more details tomorrow)

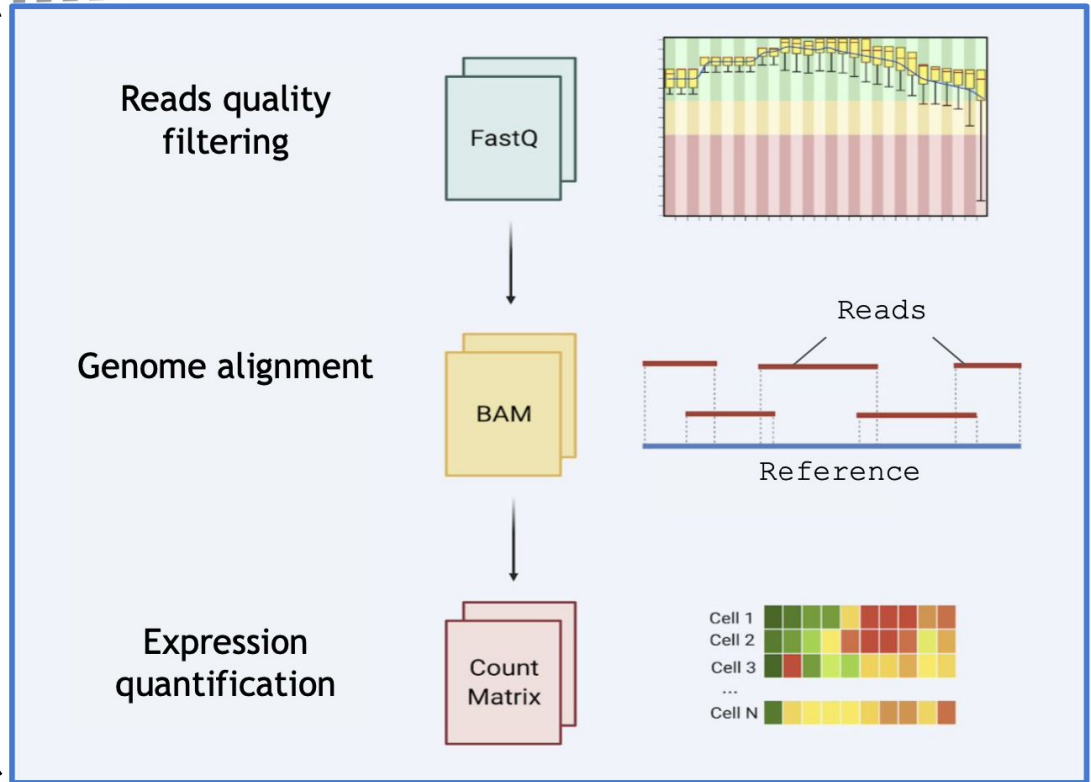
Last step : generation of the count matrix



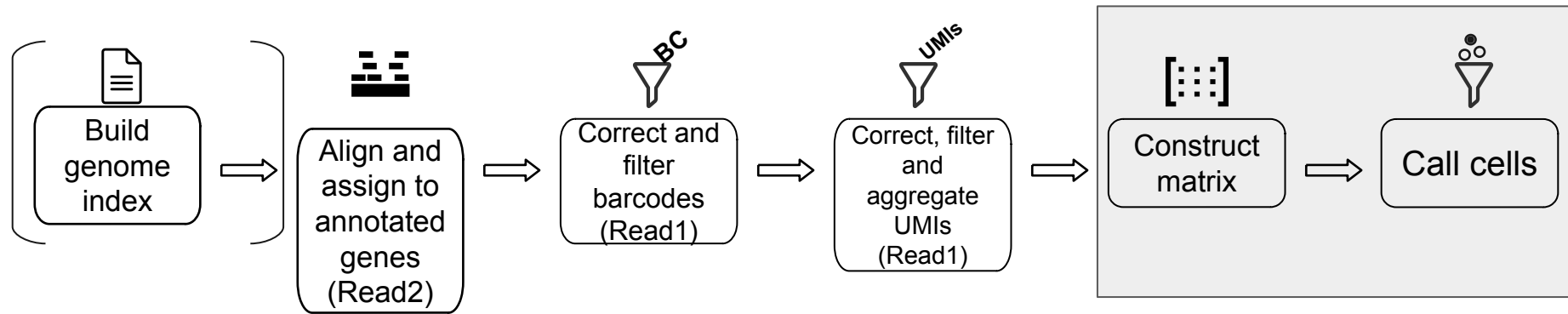
Primary analyses




will be presented in slides part 2



Overview of the workflow for primary analysis



Take-home messages

- **Primary analysis is important !** If this step has issues, the resulting count matrix will have issues that will be propagated to all downstream analyses
- These steps are often overlooked
- **Cell Ranger** : program provided by 10X Genomics that perform primary analysis (and a bit more). Cell Ranger is reliable but it is necessary to understand what it does and its limits
- You will hear that “the raw data is the count matrix” => this is wrong, remember **the raw data are the reads** The diagram shows several horizontal orange lines of varying lengths, representing sequencing reads. Some lines are solid, while others have gaps, illustrating the structure of raw sequencing data.
- Only **read2** is mapped to the genome ; read1 is synthetic Barcode+UMI
- Alternative ways to perform primary analysis exist

Acknowledgements

- Some illustrations were created by Nathalie Lehmann and Rémi Montagne