

The background of the slide is a complex, abstract pattern composed of numerous small, overlapping circles and triangles in various colors including red, green, blue, yellow, and purple. These shapes are arranged in a way that creates a sense of depth and movement, resembling a digital or biological data visualization.

# **Lecture 2**

## **Intro to Data processing: From bcl to count matrix**

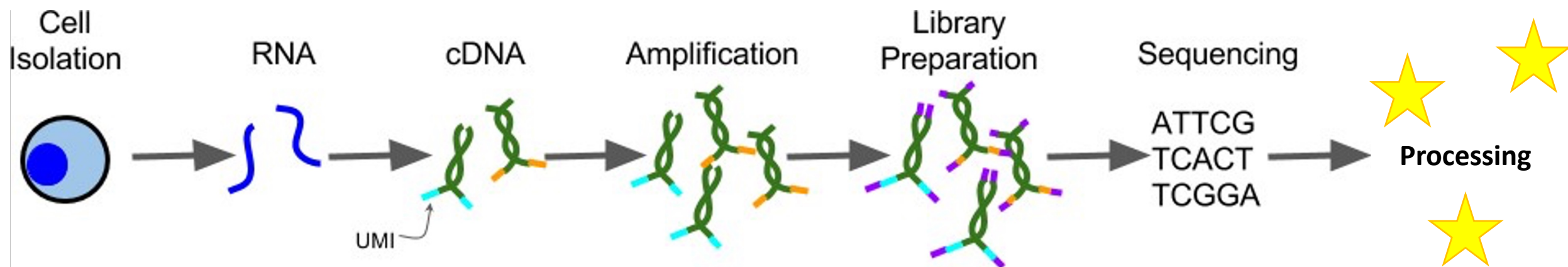
**Physalia course 2023**

—

**Single-cell RNA-seq with R/Bioconductor**

**Instructors:** Orr Ashenberg & Jacques Serizay

# Experimental pipeline

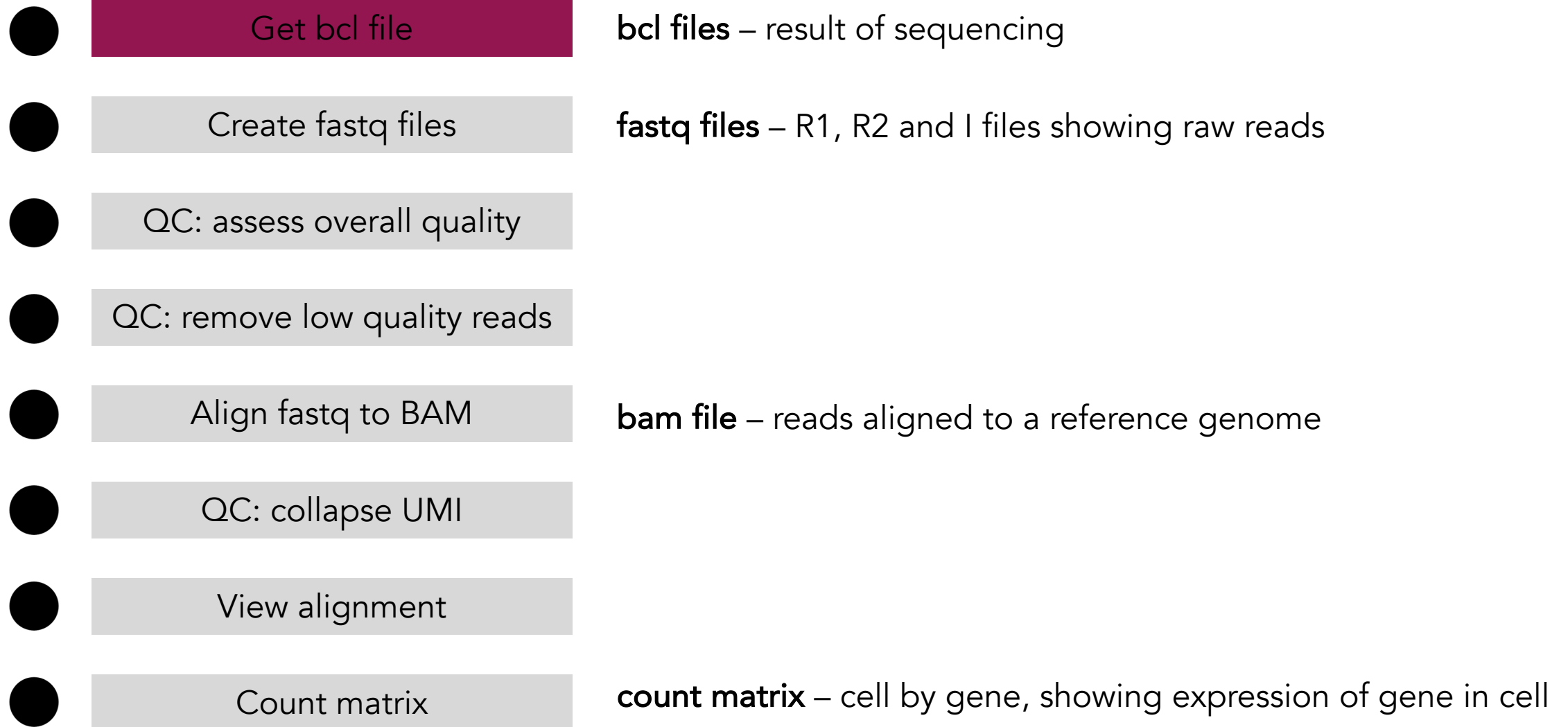


Regardless of who is going to do the experimental steps, **DISCUSS WITH THEM !!!**

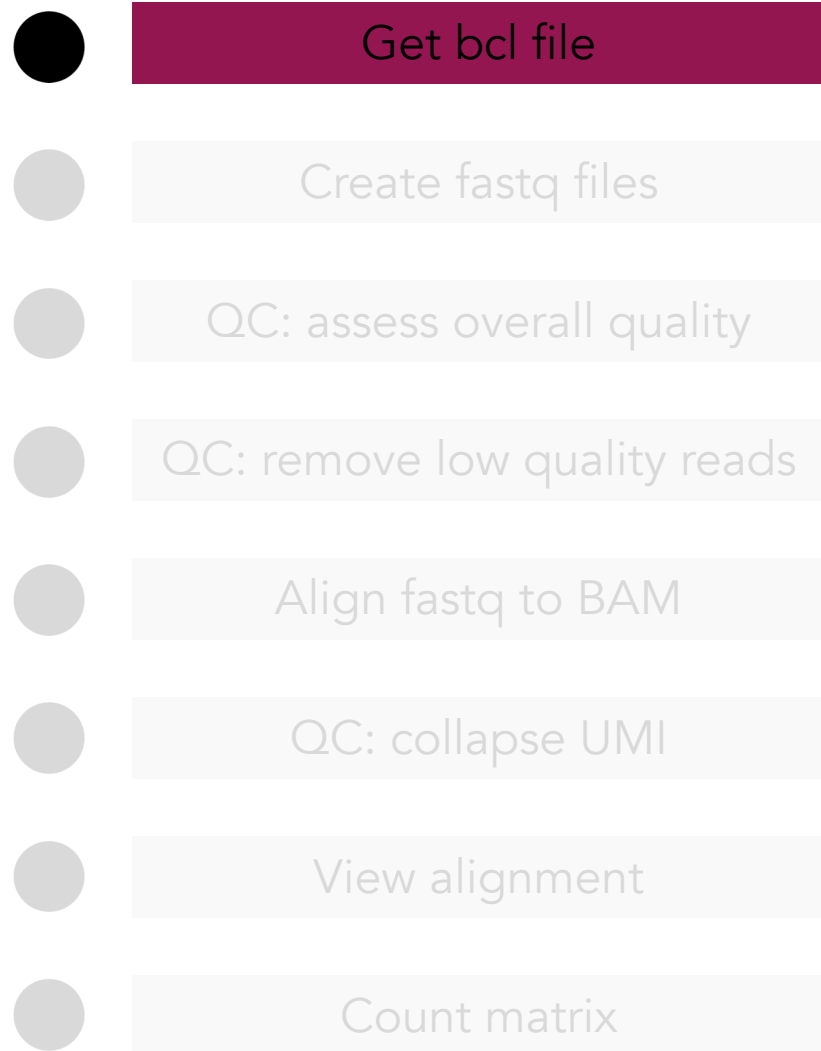
Experimental design is crucial for the success of a single-cell project !!!

**Systematic comparison of single-cell and single-nucleus RNA-sequencing methods, Ding et al., Nat. Biotech. 2020**

## Flowchart: from .bcl to count matrix



## Flowchart: from .bcl to count matrix





.bcl:

- **Raw data** output of a sequencing run
- Binary, non-human-readable file
- Contains the base calling and quality score per cluster per sequencing lane

## Flowchart: from .bcl to count matrix



Get bcl file

That is the role of the sequencing machine!



Create fastq files



QC: assess overall quality



QC: remove low quality reads



Align fastq to BAM



QC: collapse UMI



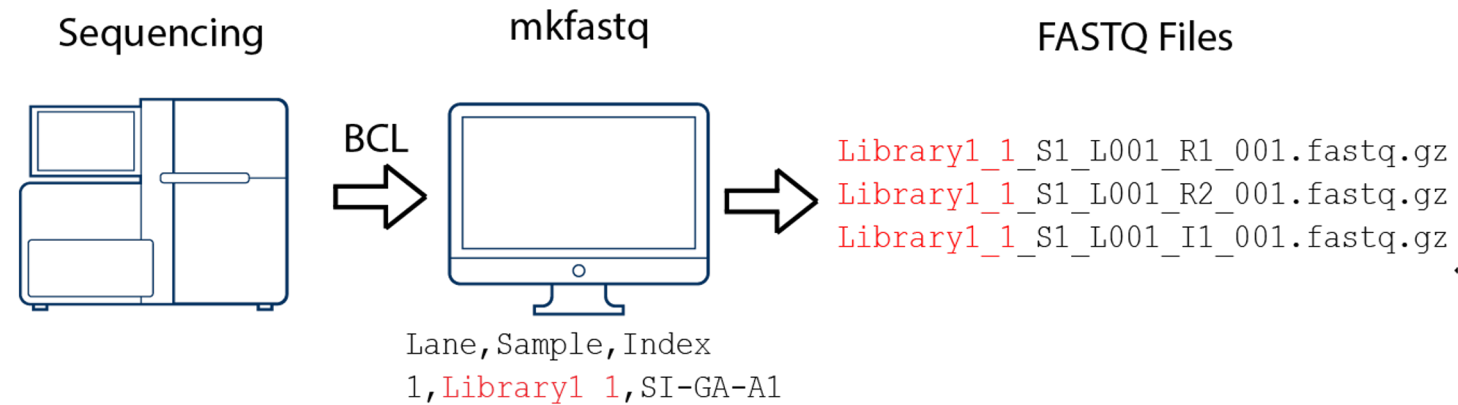
View alignment



Count matrix

Command:

```
bcl2fastq --run-folder-dir <bcl_files_folder> -p 12 --output-dir <fastq_files_folder>
```

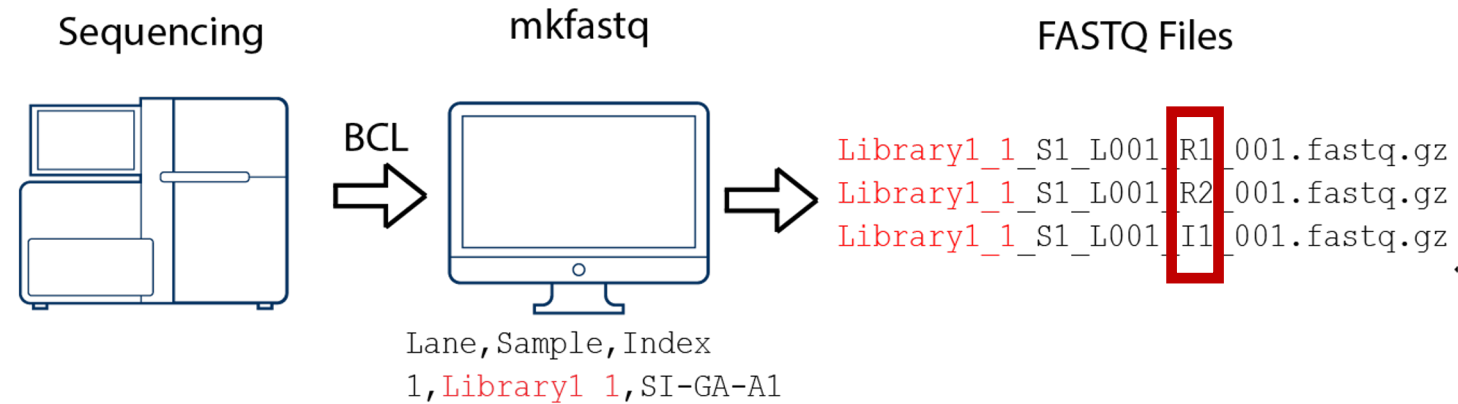


User guide:

[https://support.illumina.com/content/dam/illumina-support/documents/documentation/software\\_documentation/bcl2fastq/bcl2fastq\\_letterbooklet\\_15038058\\_brpmi.pdf](https://support.illumina.com/content/dam/illumina-support/documents/documentation/software_documentation/bcl2fastq/bcl2fastq_letterbooklet_15038058_brpmi.pdf)

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# How is a .fastq organized?

Each fastq file contains reads, each read is composed of 4 lines:

1. A sequence identifier with information about the sequencing run
2. The sequence (the base calls; A, C, T, G and N).
3. A separator, which is simply a plus (+) sign.
4. The base call quality scores, using ASCII characters to represent the numerical quality scores.

```

jacquesserizay@LOCAL[12:46:19]:~ $ cat SRR11575369_1.fastq.gz | zcat | head -n 8
@SRR11575369.1 1/1
ANCAACAGTGAATTCGTTTTGATGAAAAAATAAATTTGTTCTTCAAAGCAGAGTGAATGATGCAGTACGAGCTCTTGCTCTTGAAAACCCATCACAACCTTTATAATTTAATAATTAGTGAAAAATTAAAAAATAATATTTCTTATATT
+
F#F:FFFFF:FFFFFFFFF:F,FF,FF::F,::FFFFFFFFF:FF:FF:F,FFF,F:F,FFF,FF:FF:F:FFF:FF:FFFF,:F::FFFF:FF,FF:F:FF,,:F:F,::,F:F:FF,::,FF:,,F,::,::FFFF::F::,FF
@SRR11575369.2 2/1
TNGCCAGTCAATACCGCCCTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTGTGTTTTTTTTTTTTTATATAATAAAATTTTTTTTTTTTTTAATAAAATTTTTTTTTTATTTT
+
F#FFFFFFFF::FF:FF::FFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFF,FFFFFFFFF:FFFFFF:FFFFFFFFFFFF:F,FFFF,,::,::FFFFFFFFF,,::,F,,,,,FFFFFF,F,,FF:,F:,F,FFFFF,,::,F::F

```

# Why do we end up with so many fastq files?

We sequence the paired ends of each DNA fragment molecule, in 3 different sequencing “runs”.



# I1 is important for demultiplexing multiple samples simultaneously sequenced

Each fastq generated by **bcl2fastq** contains a different information:

- I1.fastq contains sample index
- R1.fastq contains cell barcode + UMI
- R2.fastq contains transcript information



## R1 contains information on the cell of origin (as well as a UMI)

Each fastq generated by **bcl2fastq** contains a different information:

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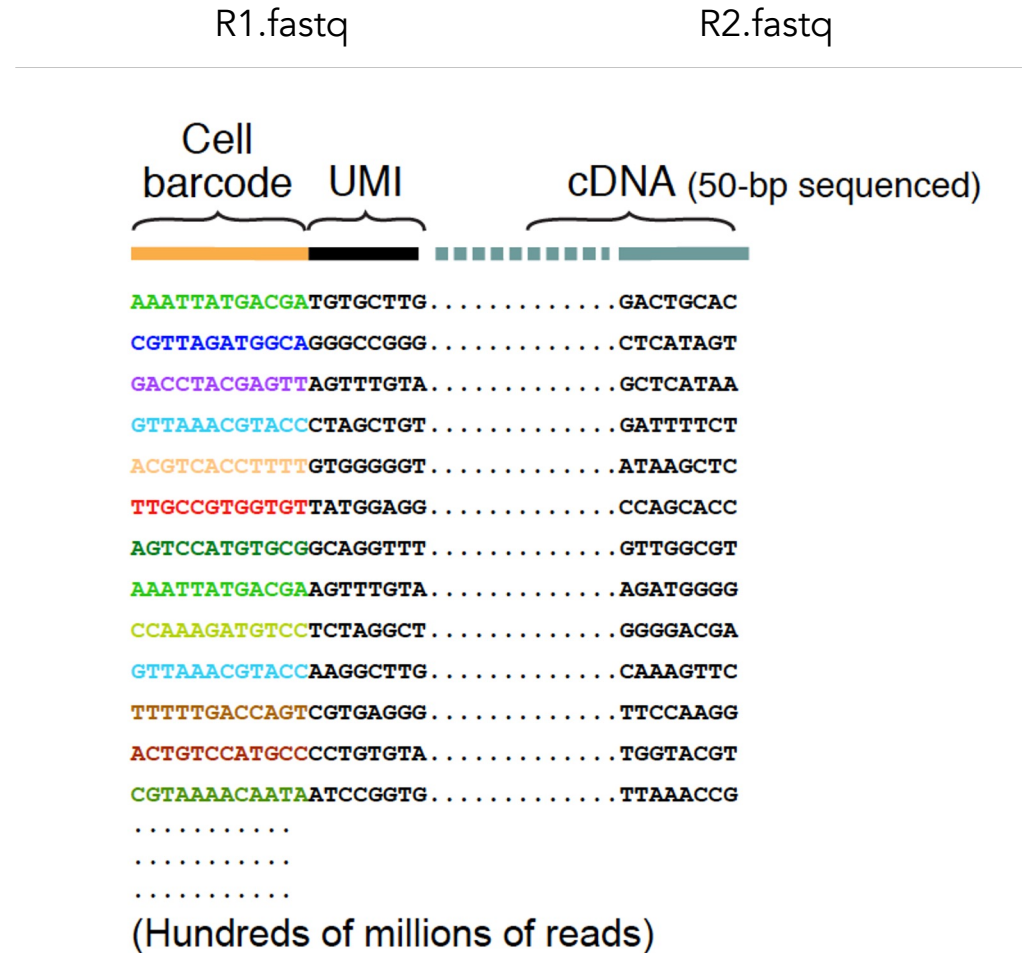
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# Reconstructing actual molecules loaded in the sequencer

After demultiplexing, here is what you have:



## Flowchart: from .bcl to count matrix



Get bcl file

That is the role of the sequencing machine!



Create fastq files

bcl2fastq



QC: assess overall quality



QC: remove low quality reads



Align fastq to BAM



QC: collapse UMI



View alignment

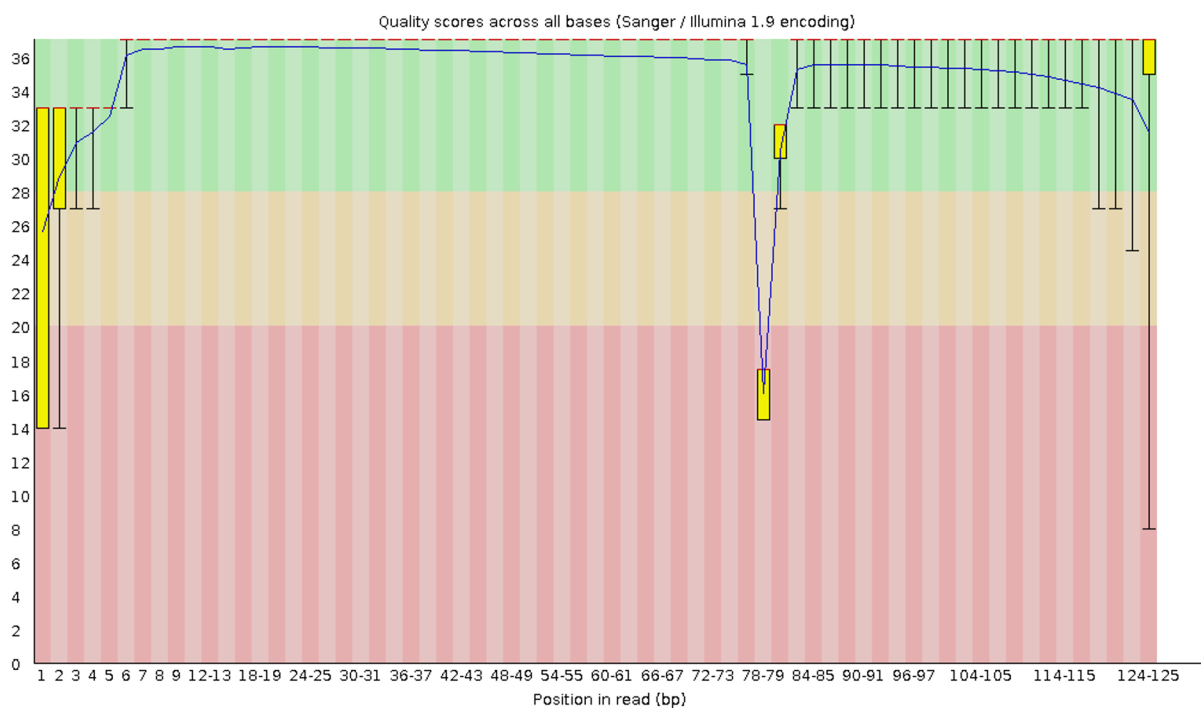
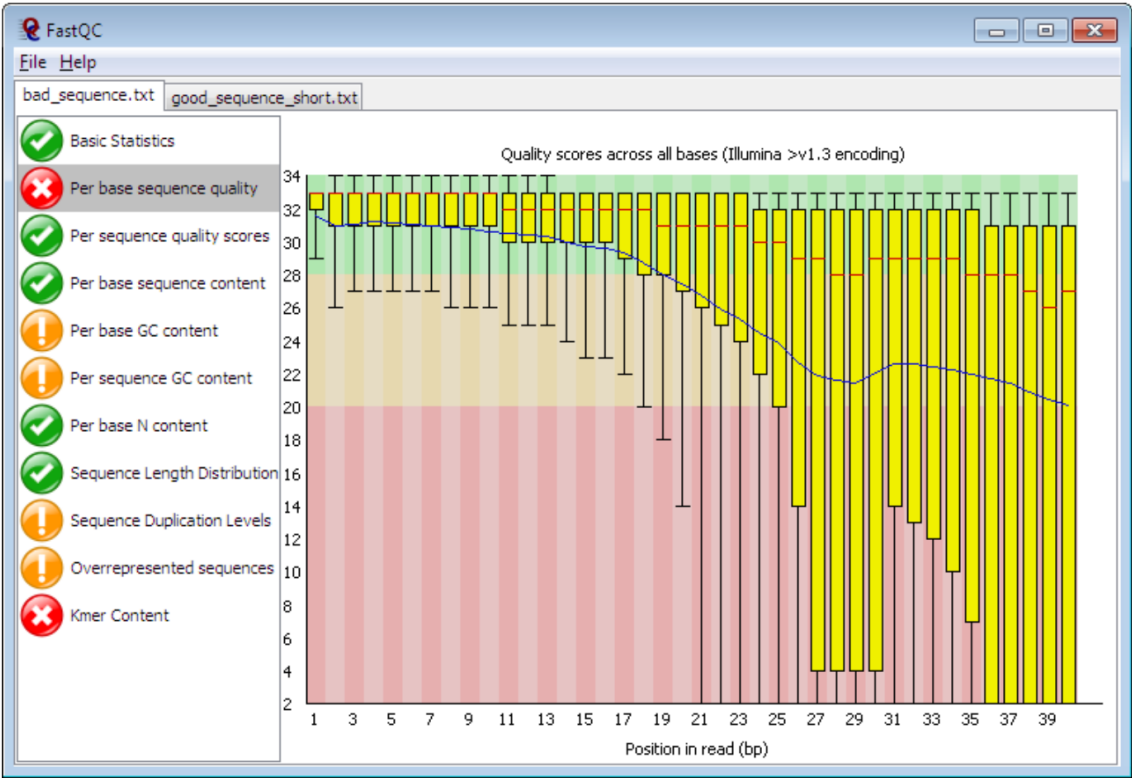


Count matrix

# Raw sequencing QC is typically done with Fastqc

## FastQC

FastQC is a program designed to spot potential problems in high throughput sequencing datasets. It runs a set of analyses on one or more raw sequence files in fastq or bam format and produces a report which summarises the results.



FastQC will highlight any areas where this library looks unusual and where you should take a closer look. The program is not tied to any specific type of sequencing technique and can be used to look at libraries coming from a large number of different experiment types (Genomic Sequencing, ChIP-Seq, RNA-Seq, BS-Seq etc etc).



## Flowchart: from .bcl to count matrix



Get bcl file

That is the role of the sequencing machine!



Create fastq files

bcl2fastq



QC: assess overall quality

fastqc



QC: remove low quality reads



Align fastq to BAM



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View alignment



Count matrix

Cellranger

## Filtering low quality reads

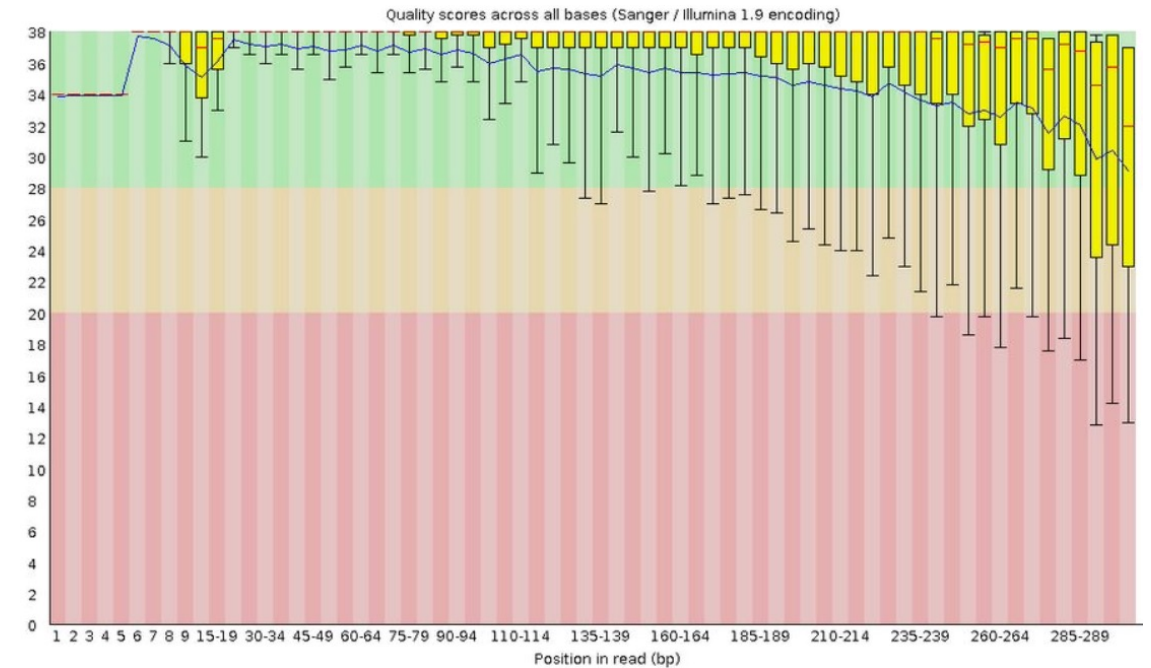
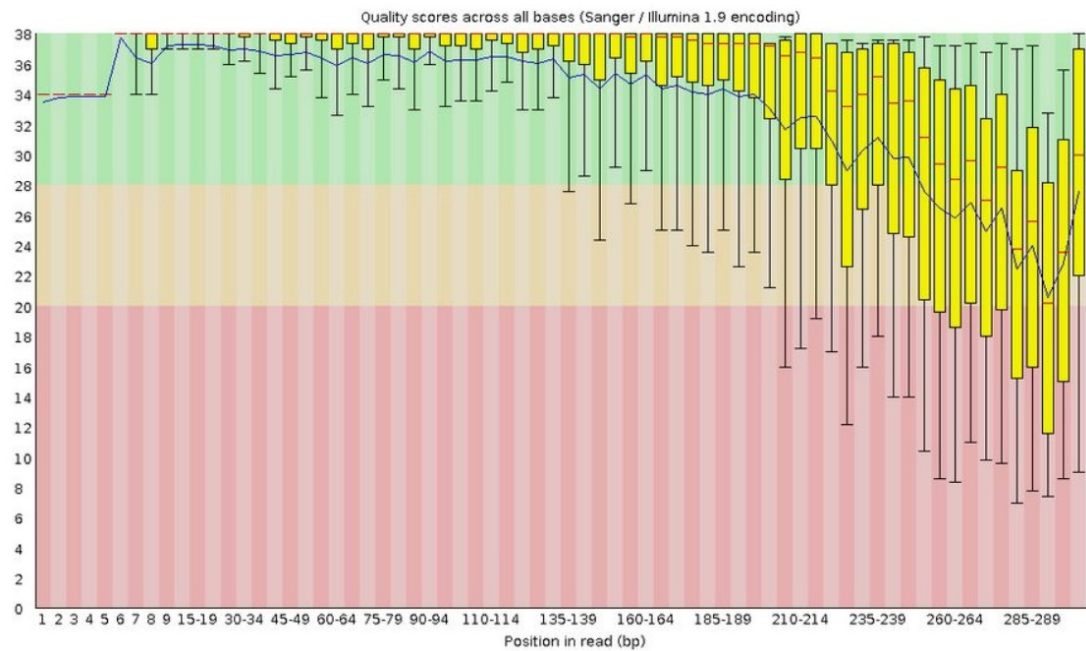
It is important to consider (and remove!) :

1. Reads with overall low quality
2. Unrecognized cell barcode

# Filtering low quality reads

It is important to consider (and remove!) :

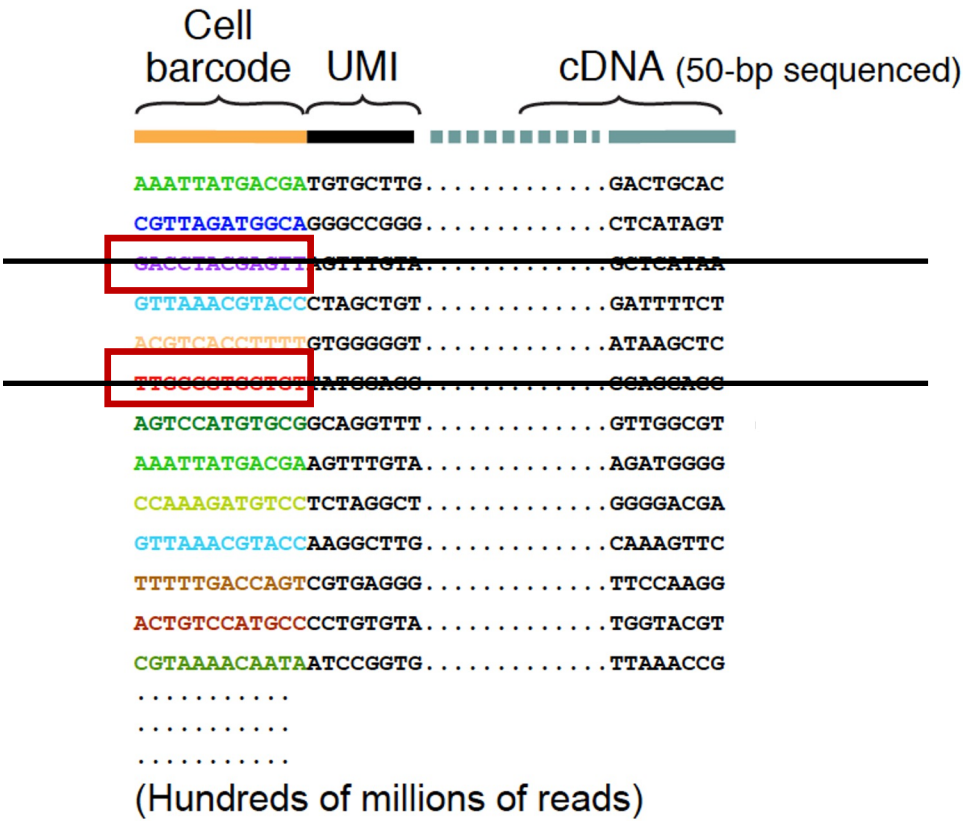
1. Reads with overall low quality
2. Unrecognized cell barcode



# Filtering low quality reads

It is important to consider (and remove!) :

1. Reads with overall low quality
2. Unrecognized cell barcode



## What is a barcode whitelist?

List of all known barcode sequences that have been included in the assay kit and are available during library preparation.

For example, there are roughly 737,000 cell barcodes in the whitelist for Cell Ranger's Single Cell 3' applications. Here are the first 10 lines of the corresponding barcode whitelist 737K-august-2016.txt:

```
AAACCTGAGAAACCAT
AAACCTGAGAAACCGC
AAACCTGAGAAACCTA
AAACCTGAGAAACGAG
AAACCTGAGAAACGCC
AAACCTGAGAAAGTGG
AAACCTGAGAACAAC
AAACCTGAGAACAATC
AAACCTGAGAACTCGG
AAACCTGAGAACTGTA
```

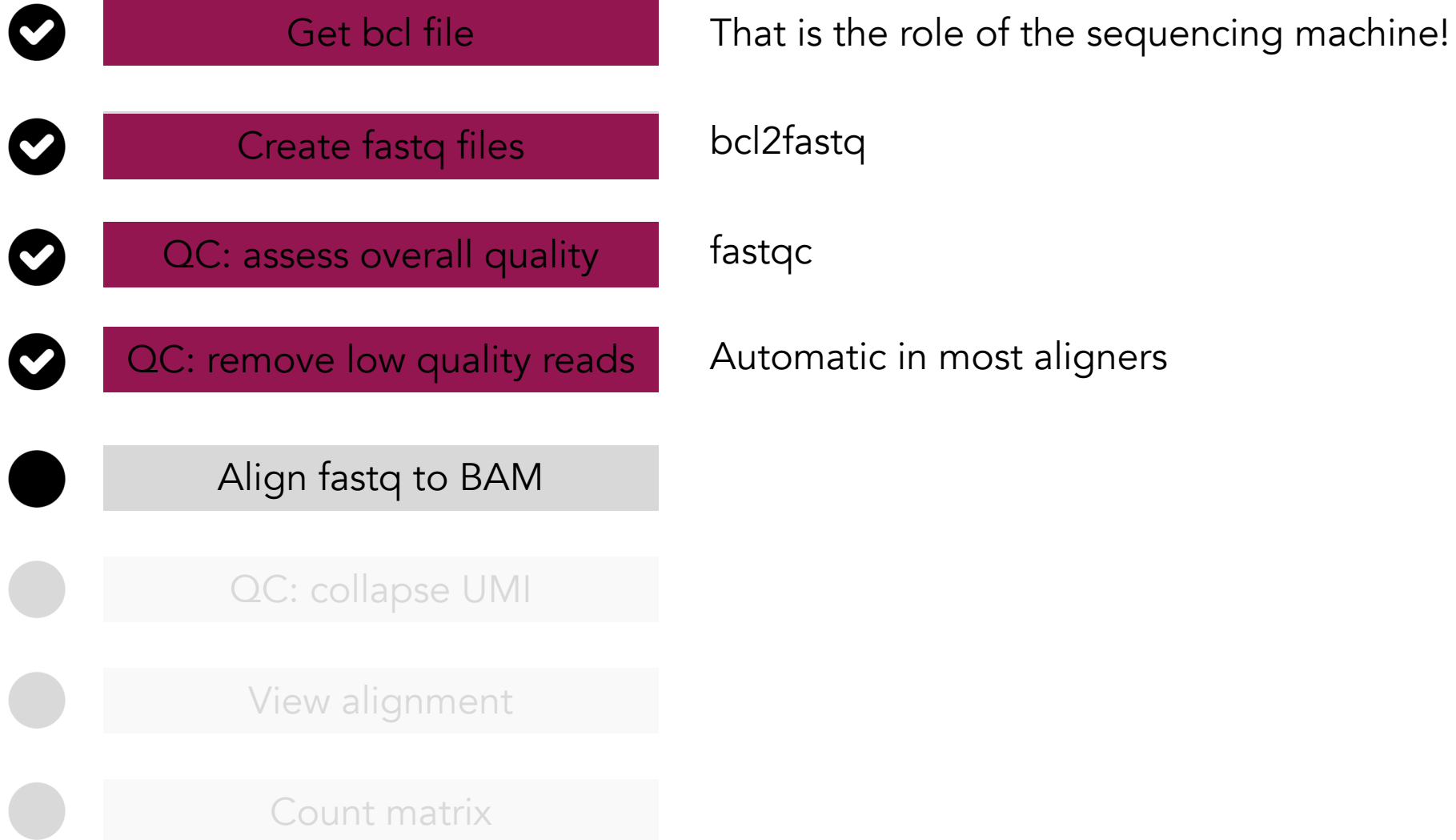


### How to correct barcode sequencing errors?

For every observed barcode in the dataset not on the whitelist, but 1-Hamming-distance away (i.e. 1 mismatch) from a whitelist barcode:

- Compute the posterior probability that the observed barcode originated from the whitelist barcode with a sequencing error at the differing base (based on the base Q score).
- Replace the observed barcode with the whitelist barcode with the highest posterior probability that exceeds 0.975.

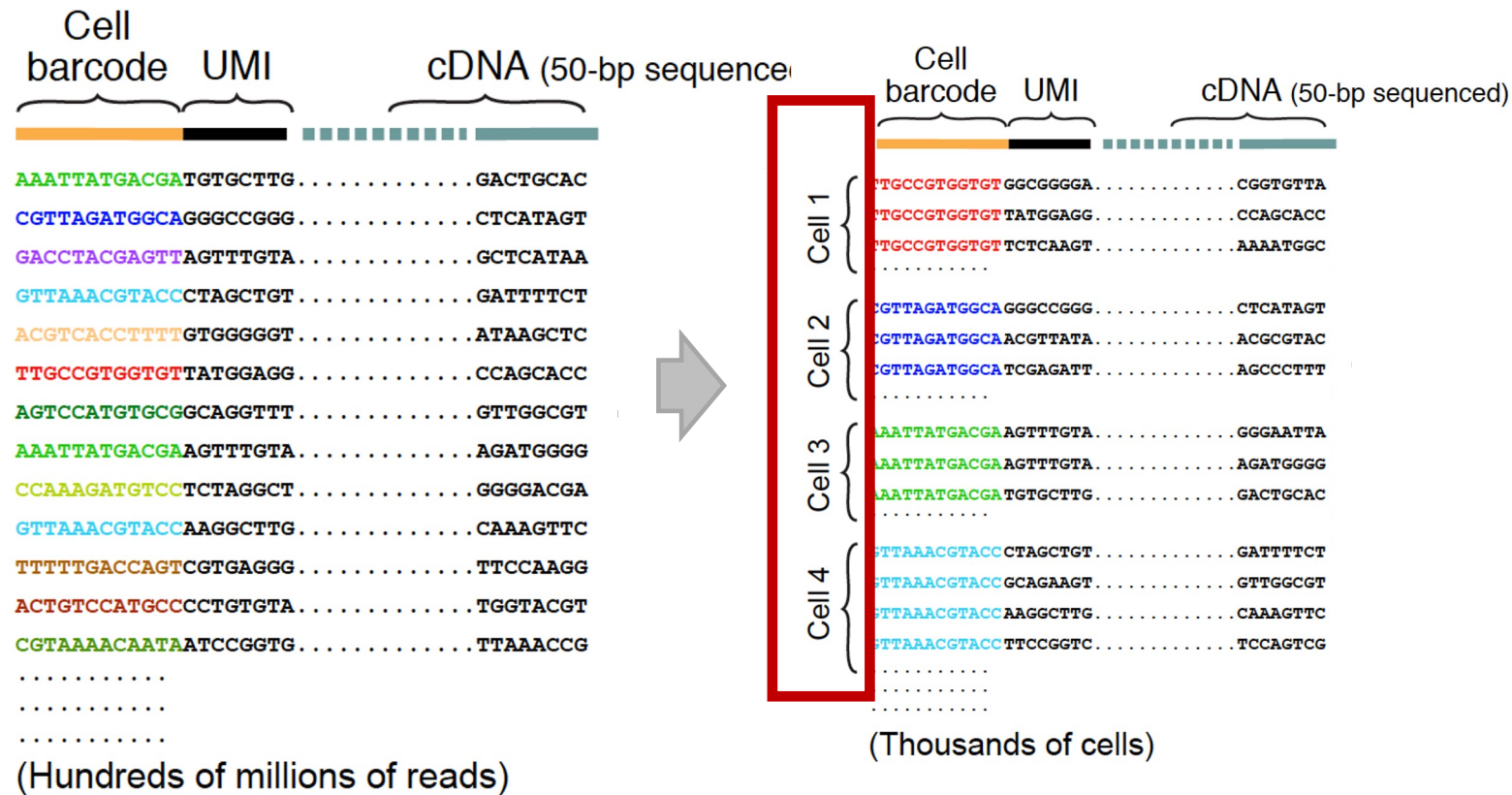
## Flowchart: from .bcl to count matrix



Cellranger

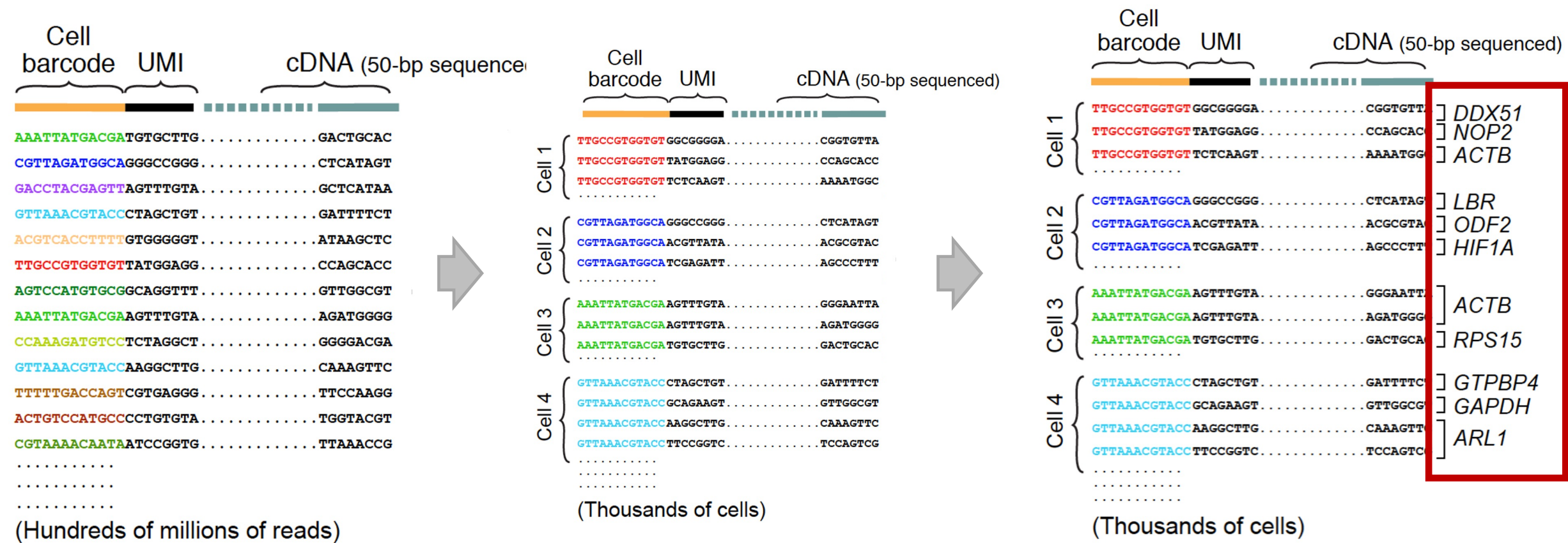
# Aligning reads to a transcriptome reference

## 1. Group reads by cell-of-origin (using the cell barcodes)



# Aligning reads to a transcriptome reference

1. Group reads by cell-of-origin (using the cell barcodes)
2. Recover which transcript the cDNA sequence aligns to





Most aligners (included STAR-based cellranger) will map scRNAseq reads on a transcriptome index.

A transcriptome index consists of:

# FASTA example

1. A genome sequence reference

[illegible]

# What do I need to align my reads?

Most aligners (included STAR-based cellranger) will map scRNAseq reads on a transcriptome index.

A transcriptome index consists of:

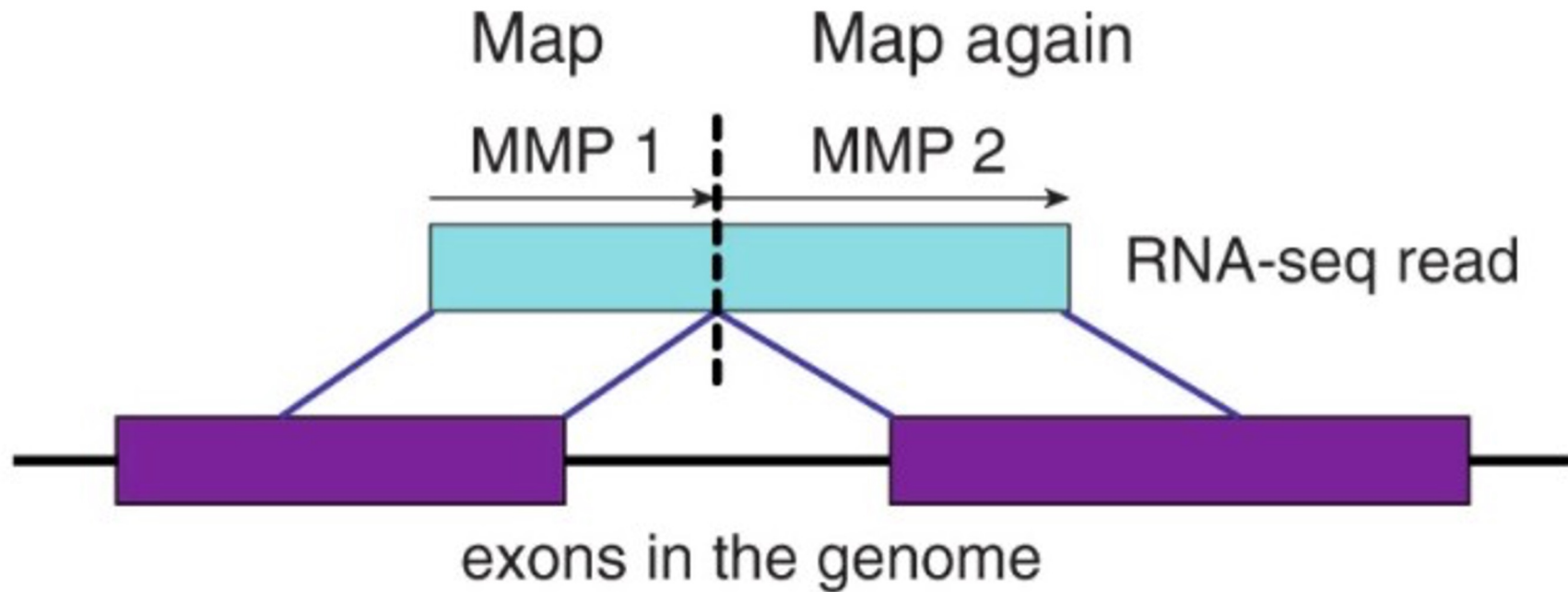
1. A genome sequence reference
2. A gene feature annotation reference

## GTF example

Seqname	Source	Feature	Start	End	Score	Strand	Frame	Attributes
YHet	protein_coding	exon	311	424	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 1
YHet	protein_coding	CDS	311	424	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 1
YHet	protein_coding	exon	540	799	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 2
YHet	protein_coding	CDS	540	799	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 2
YHet	protein_coding	exon	857	1196	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 3
YHet	protein_coding	CDS	857	1196	.	+	1	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 3
YHet	protein_coding	exon	1254	1519	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 4
YHet	protein_coding	CDS	1254	1519	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 4
YHet	protein_coding	exon	1576	1729	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 5
YHet	protein_coding	CDS	1576	1729	.	+	1	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 5
YHet	protein_coding	exon	1816	2154	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 6
YHet	protein_coding	CDS	1816	2154	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 6
YHet	protein_coding	exon	2212	2324	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 7
YHet	protein_coding	CDS	2212	2324	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 7
YHet	protein_coding	exon	2376	2667	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 8
YHet	protein_coding	CDS	2376	2667	.	+	1	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 8
YHet	protein_coding	exon	2726	2879	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 9
YHet	protein_coding	CDS	2726	2879	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 9
YHet	protein_coding	exon	15564	15931	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 10
YHet	protein_coding	CDS	15564	15931	.	+	2	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 10
YHet	protein_coding	exon	16461	16907	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 11
YHet	protein_coding	CDS	16461	16907	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 11
YHet	protein_coding	exon	16954	19761	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 12
YHet	protein_coding	CDS	16954	19761	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 12
YHet	protein_coding	exon	30303	30469	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 13
YHet	protein_coding	CDS	30303	30469	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 13
YHet	protein_coding	exon	30522	31622	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 14
YHet	protein_coding	CDS	30522	31622	.	+	1	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 14
YHet	protein_coding	exon	33215	33413	.	+	.	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 15
YHet	protein_coding	CDS	33215	33410	.	+	1	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 15
YHet	protein_coding	stop_codon	33411	33413	.	+	0	gene_id "FBgn0001315"; transcript_id "FBtr0113891"; exon_number 15

## STAR aligner

STAR is a traditional aligner that works by trying to find the longest possible sequence which matches one or more sequences in the reference genome.



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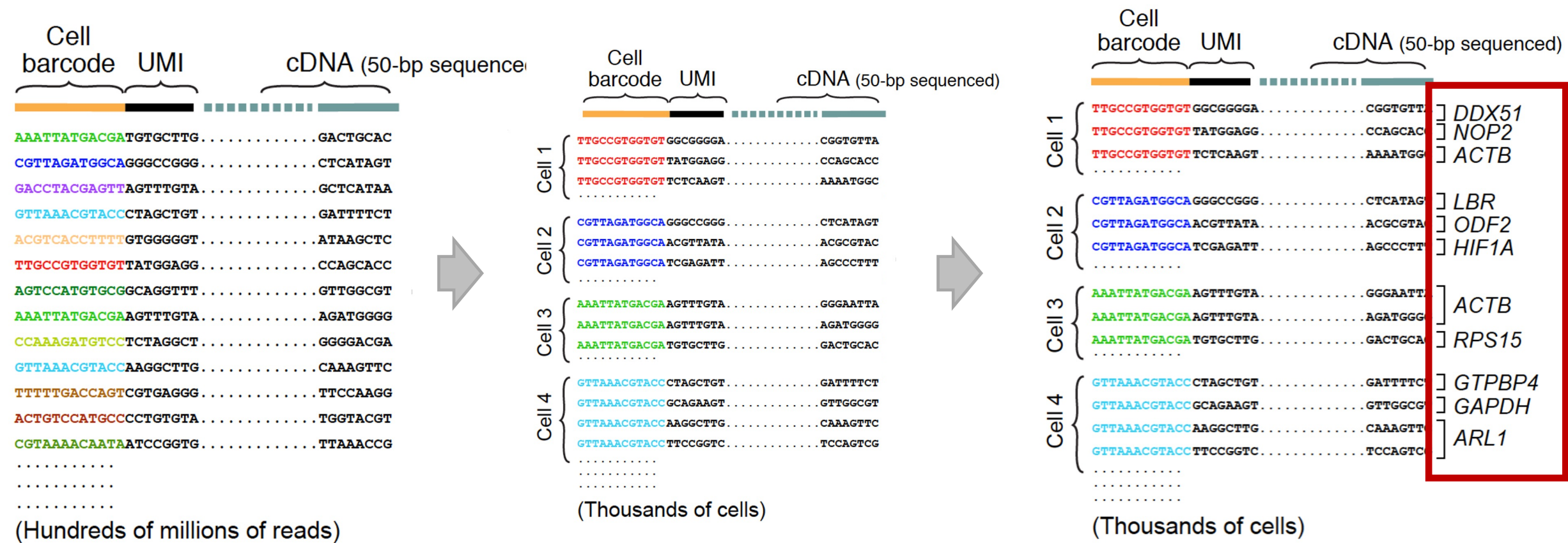
- Advantages: STAR is a splice aware aligner, making it suitable if you are interested in studying alternative splicing.
- Disadvantage: STAR requires a lot of RAM.

An alternative requiring less RAM is HISAT2.

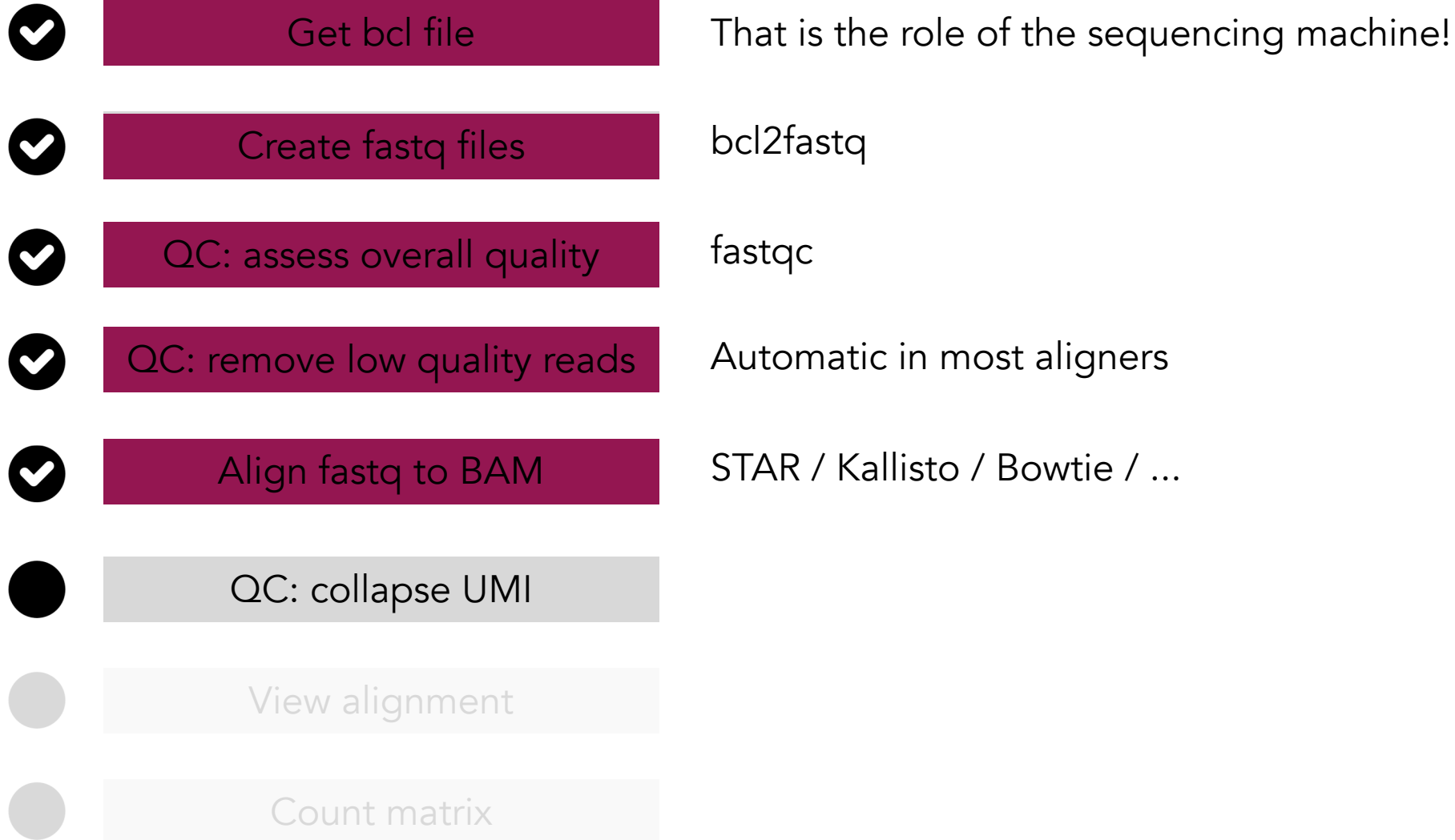


# Aligning reads to a transcriptome reference

1. Group reads by cell-of-origin (using the cell barcodes)
2. Recover which transcript the cDNA sequence aligns to



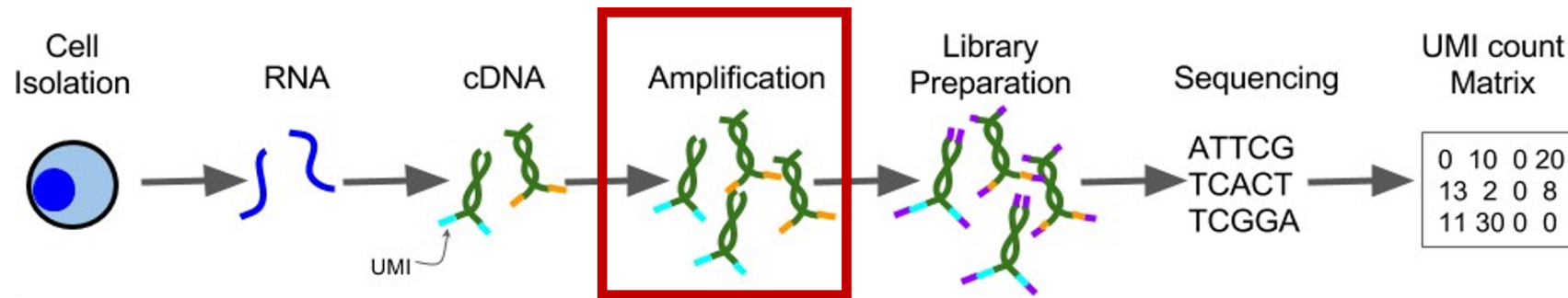
## Flowchart: from .bcl to count matrix



Cellranger

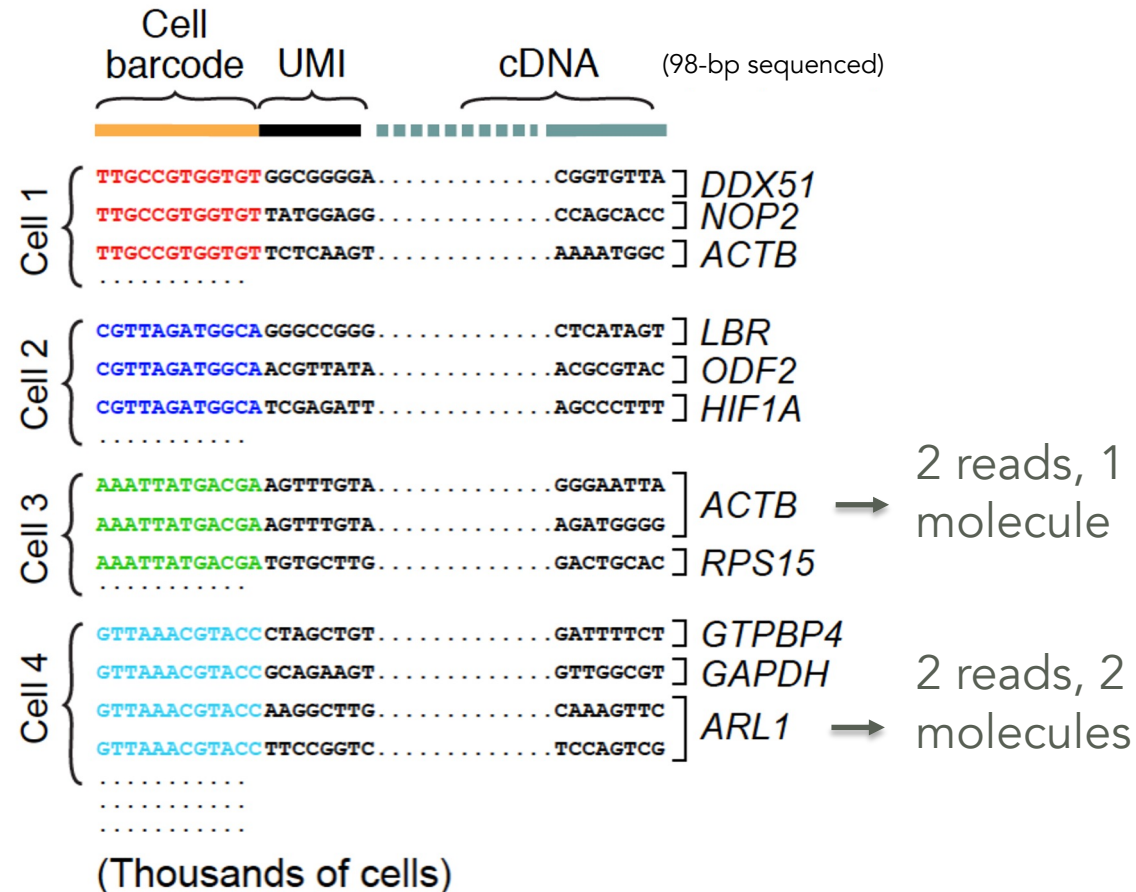
# PCR duplicates

- PCR amplifications is an issue for both bulk and scRNAseq.
- During the PCR amplification process, some transcripts become over represented in the final library compared to their true abundance.
- The problem worsened when a high number of PCR cycles are used to generate the sequencing library, for example in scRNAseq due to the low amount of starting material.



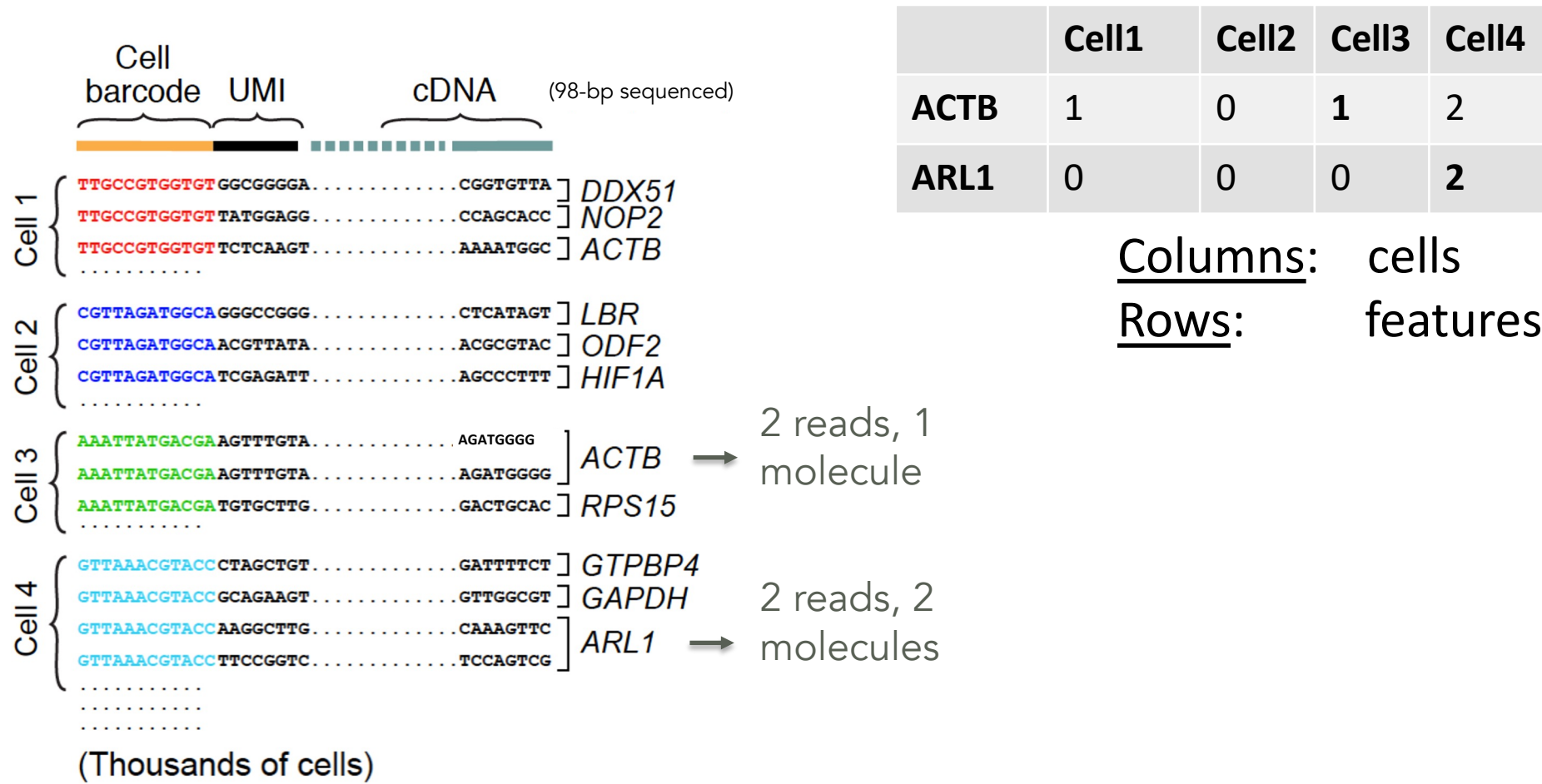
## Solution: using UMIs

UMIs enable sequencing reads to be assigned to individual transcript molecules and then the removal of amplification noise and biases from scRNAseq data.



Solution: using UMIs

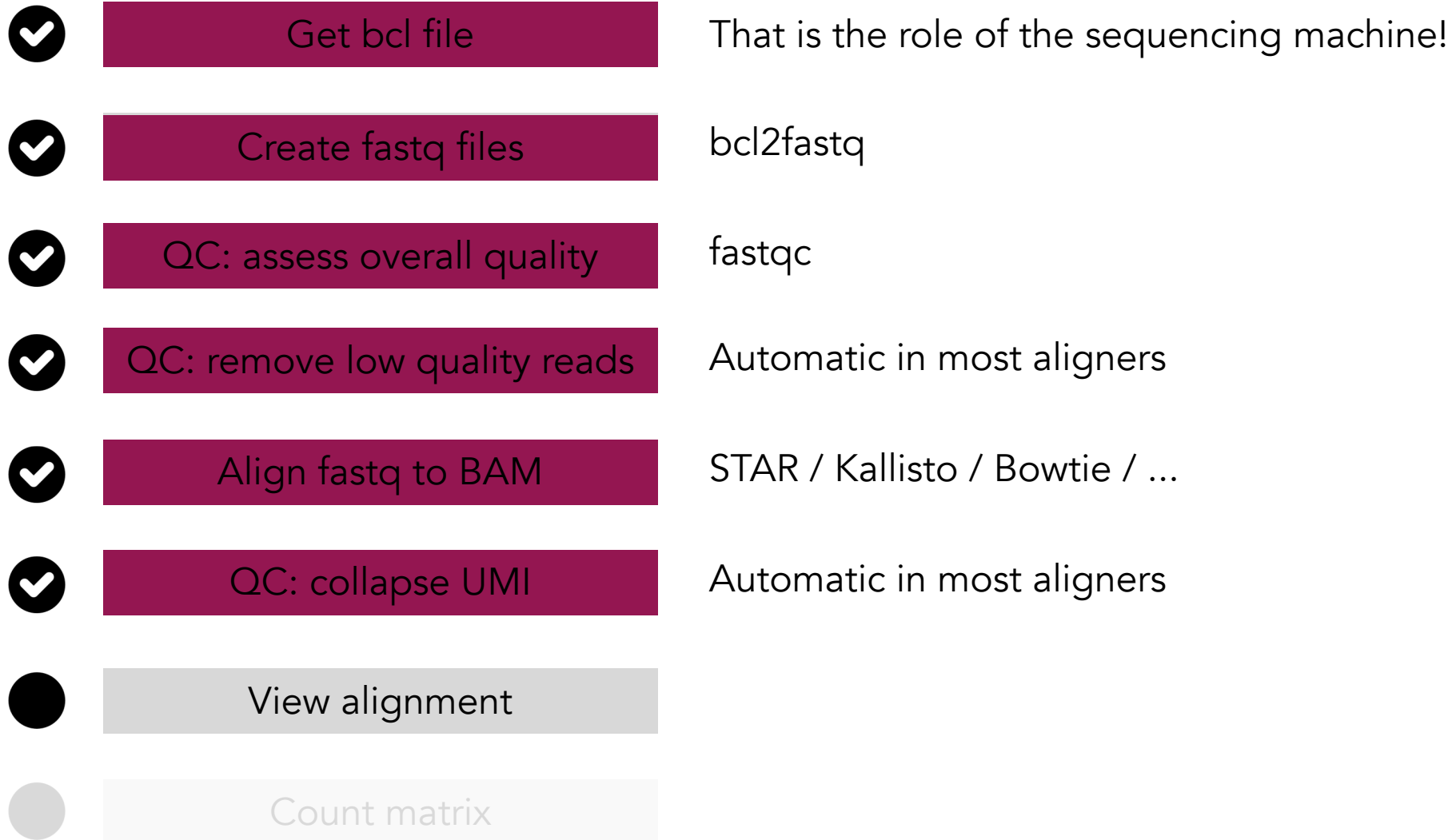
UMIs enable sequencing reads to be assigned to individual transcript molecules and then the removal of amplification noise and biases from scRNAseq data.



**“Error-correcting” almost-correct UMIs:**

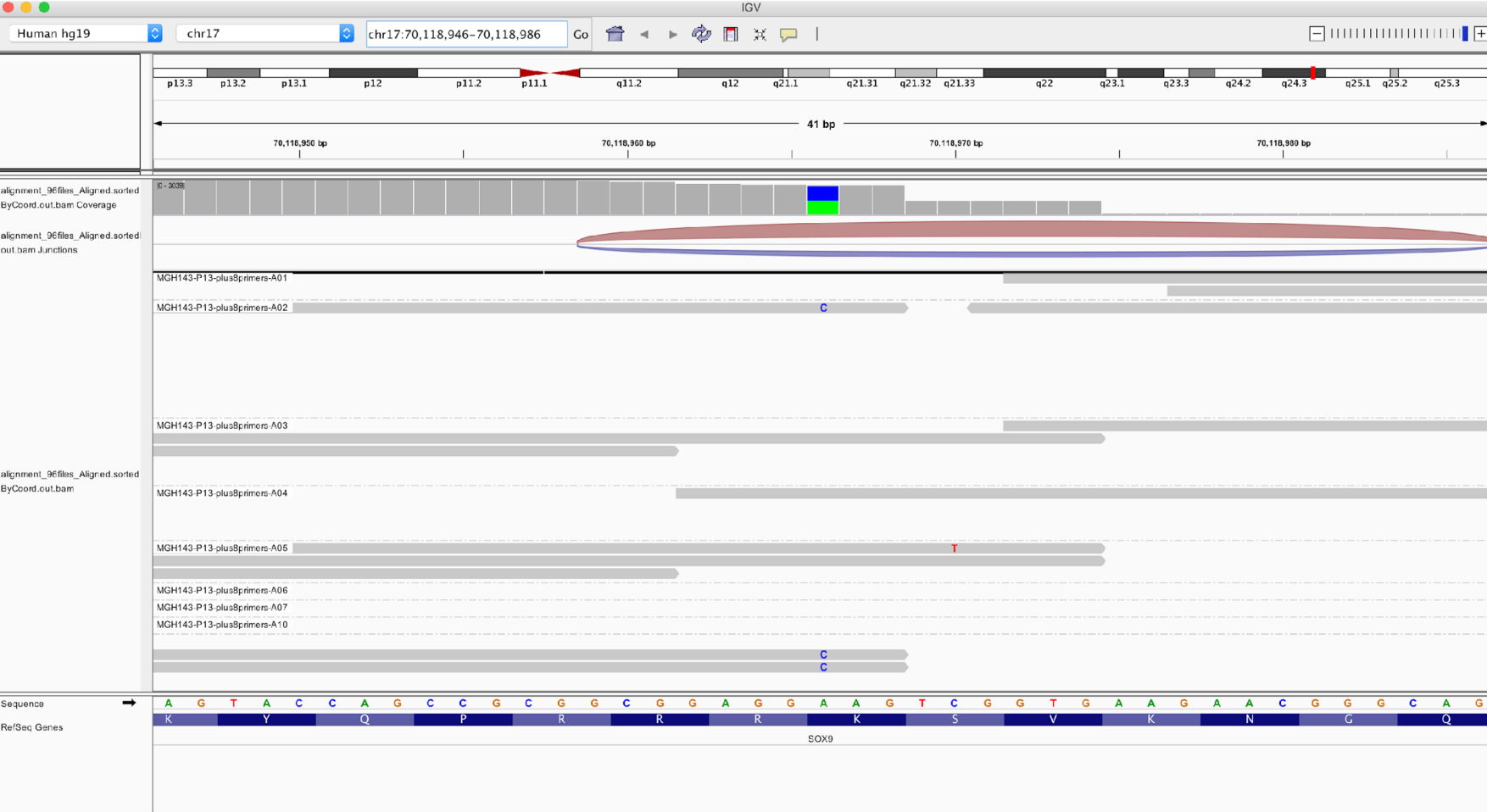
A UMI that is 1-Hamming-distance away from another UMI (with more reads) for the same cell barcode and gene is corrected to the UMI with more reads.

## Flowchart: from .bcl to count matrix



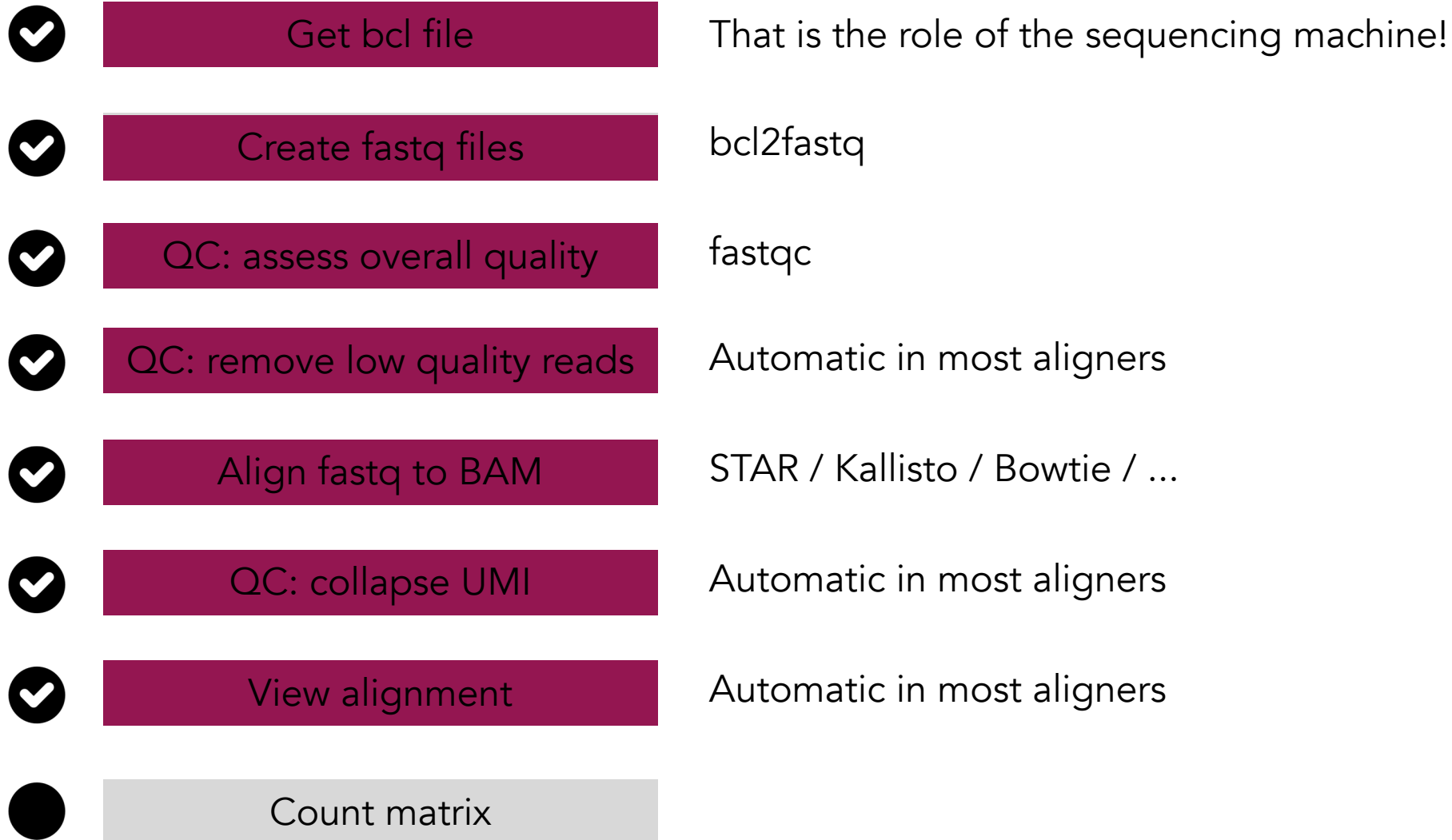
Cellranger

Point mutation, splice junction, ...



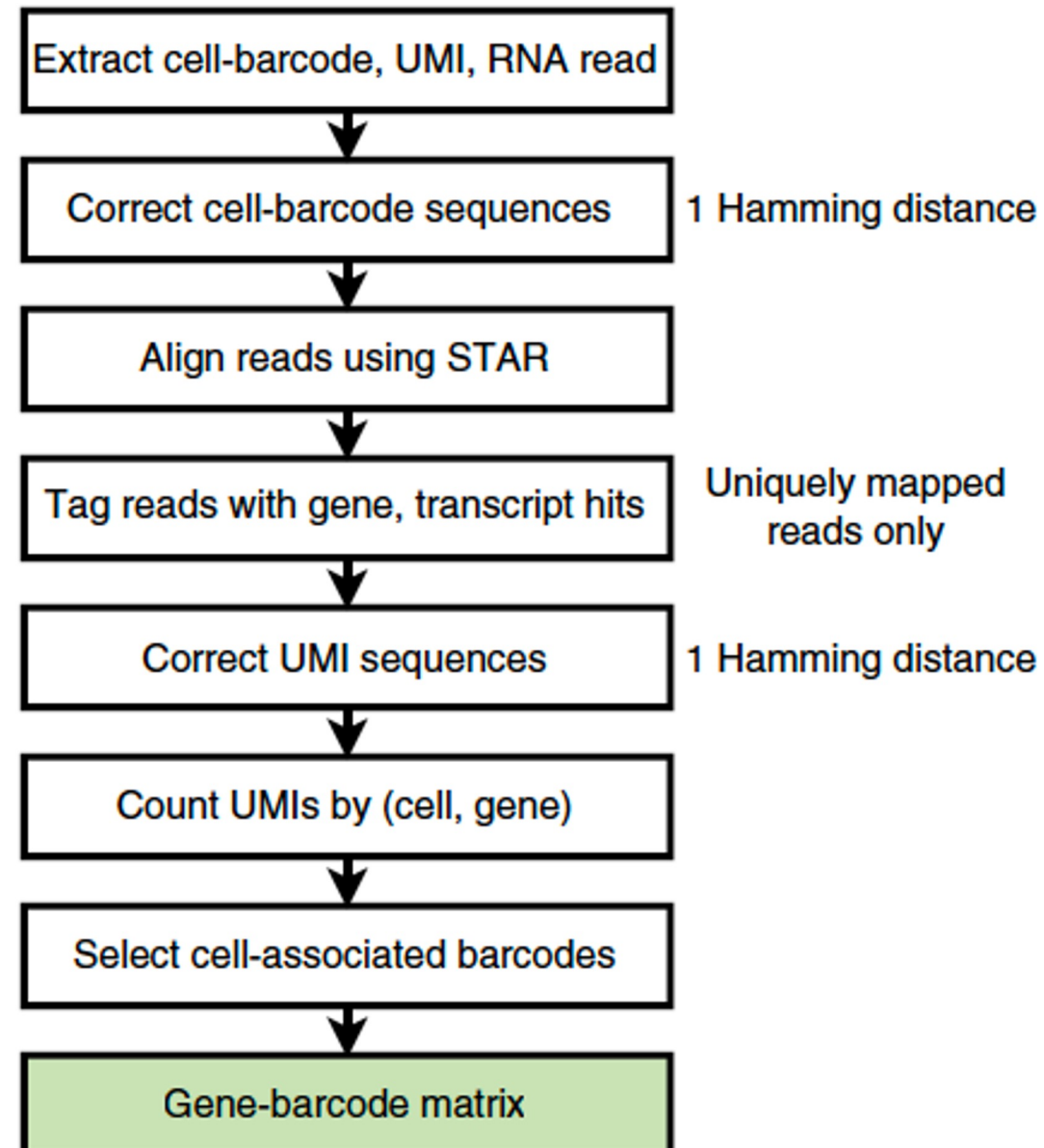


## Flowchart: from .bcl to count matrix



Cellranger

# Cellranger count pipeline



### Which reads are considered for UMI counting by Cell Ranger?

1. Only reads with a valid UMI and a valid 10x barcode.
2. No bases with base quality < 10.
3. Read maps to exactly one gene.
4. Overlaps an exon by at least 50% in a way consistent with annotated splice junctions and strand annotation.
5. Multiple reads that map to the same UMI will only count once.

# Cellranger output

Estimated Number of Cells

4,053

Mean Reads per Cell

23,971

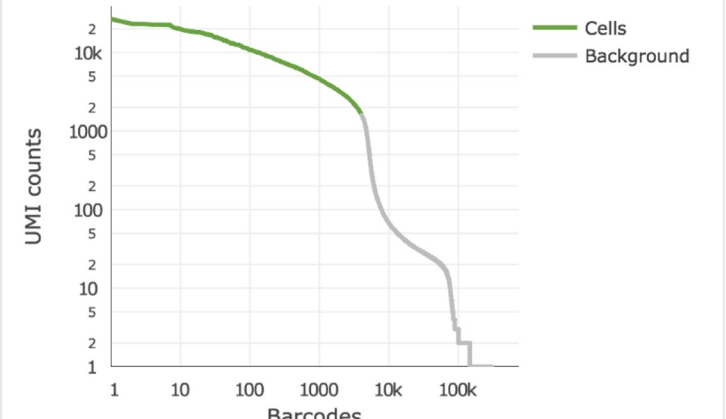
Median Genes per Cell

1,806

Sequencing	
Number of Reads	97,156,575
Valid Barcodes	94.5%
Sequencing Saturation	49.1%
Q30 Bases in Barcode	97.1%
Q30 Bases in RNA Read	93.2%
Q30 Bases in Sample Index	96.1%
Q30 Bases in UMI	97.0%

Mapping	
Reads Mapped to Genome	96.1%
Reads Mapped Confidently to Genome	93.0%
Reads Mapped Confidently to Intergenic Regions	9.3%
Reads Mapped Confidently to Intronic Regions	0.0%
Reads Mapped Confidently to Exonic Regions	83.7%
Reads Mapped Confidently to Transcriptome	42.1%
Reads Mapped Antisense to Gene	38.4%

Cells

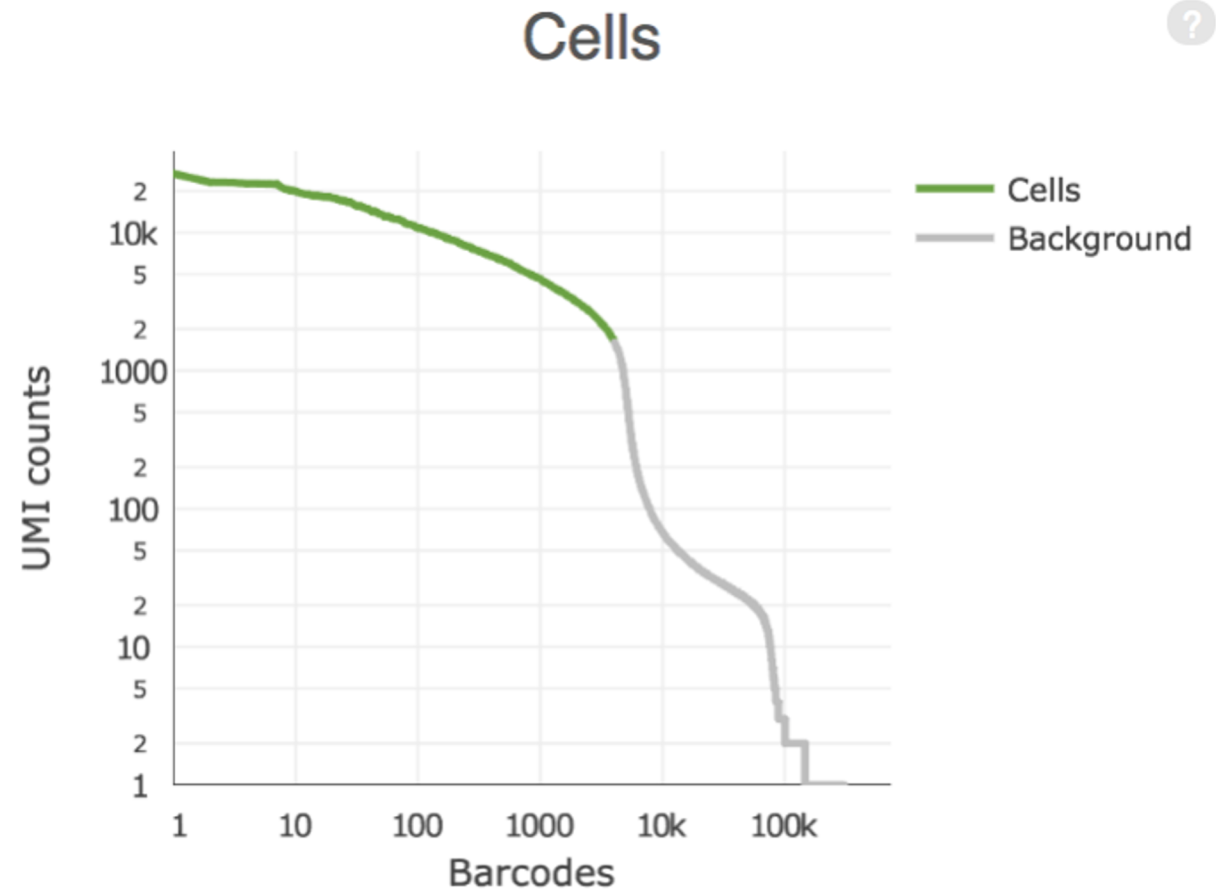


Estimated Number of Cells	4,053
Fraction Reads in Cells	81.0%
Mean Reads per Cell	23,971
Median Genes per Cell	1,806
Total Genes Detected	26,347
Median UMI Counts per Cell	3,117

Sample	
Name	results
Description	
Transcriptome	GRCh38_premrna
Chemistry	Single Cell 3' v2
Cell Ranger Version	2.2.0

## Cellranger output

Some cell barcodes have many UMIs, but most do not.

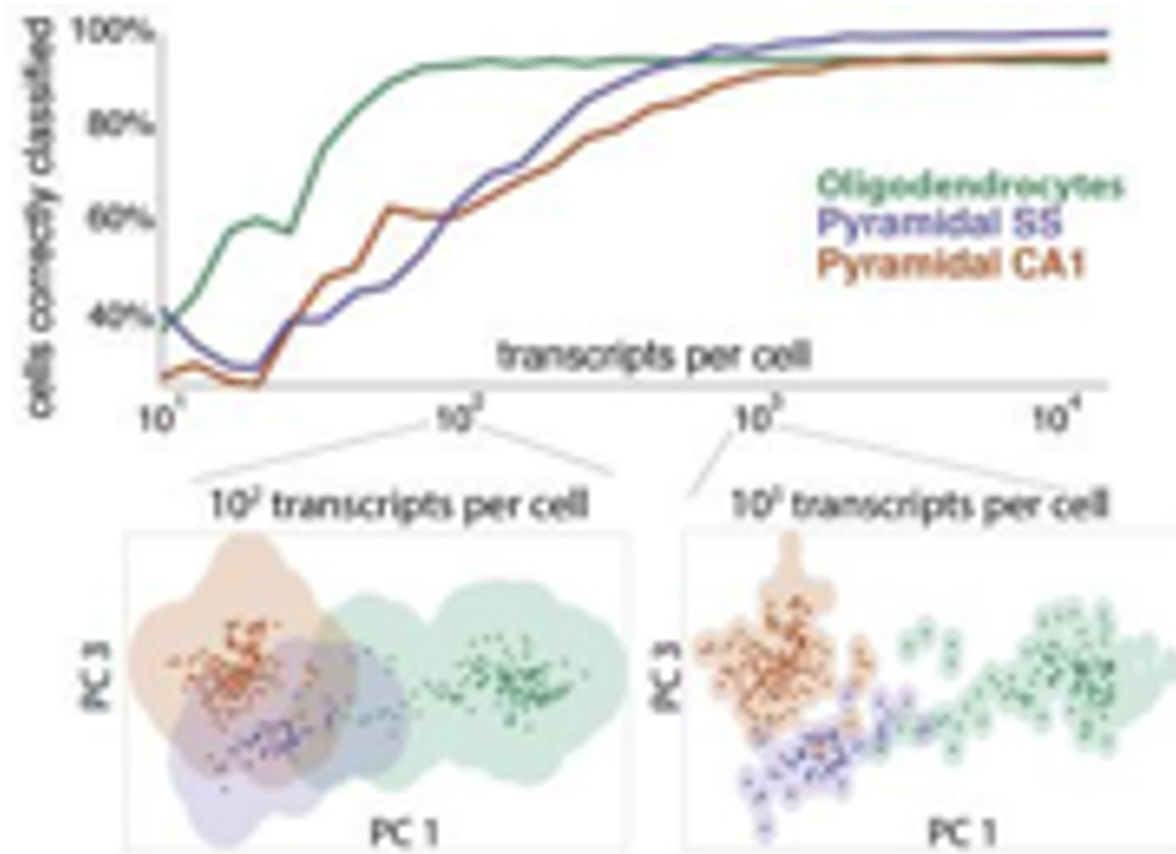


### How deeply do I need to sequence?

- At high sequencing depth, we can detect:
  - The expression of rare splice variants.
  - Quantitative modulations in transcript abundance.
- Lower sequencing depth can still be very powerful :
  - Catalog of cell types.
  - Catalog of transcriptional programs.

# Sequencing depth

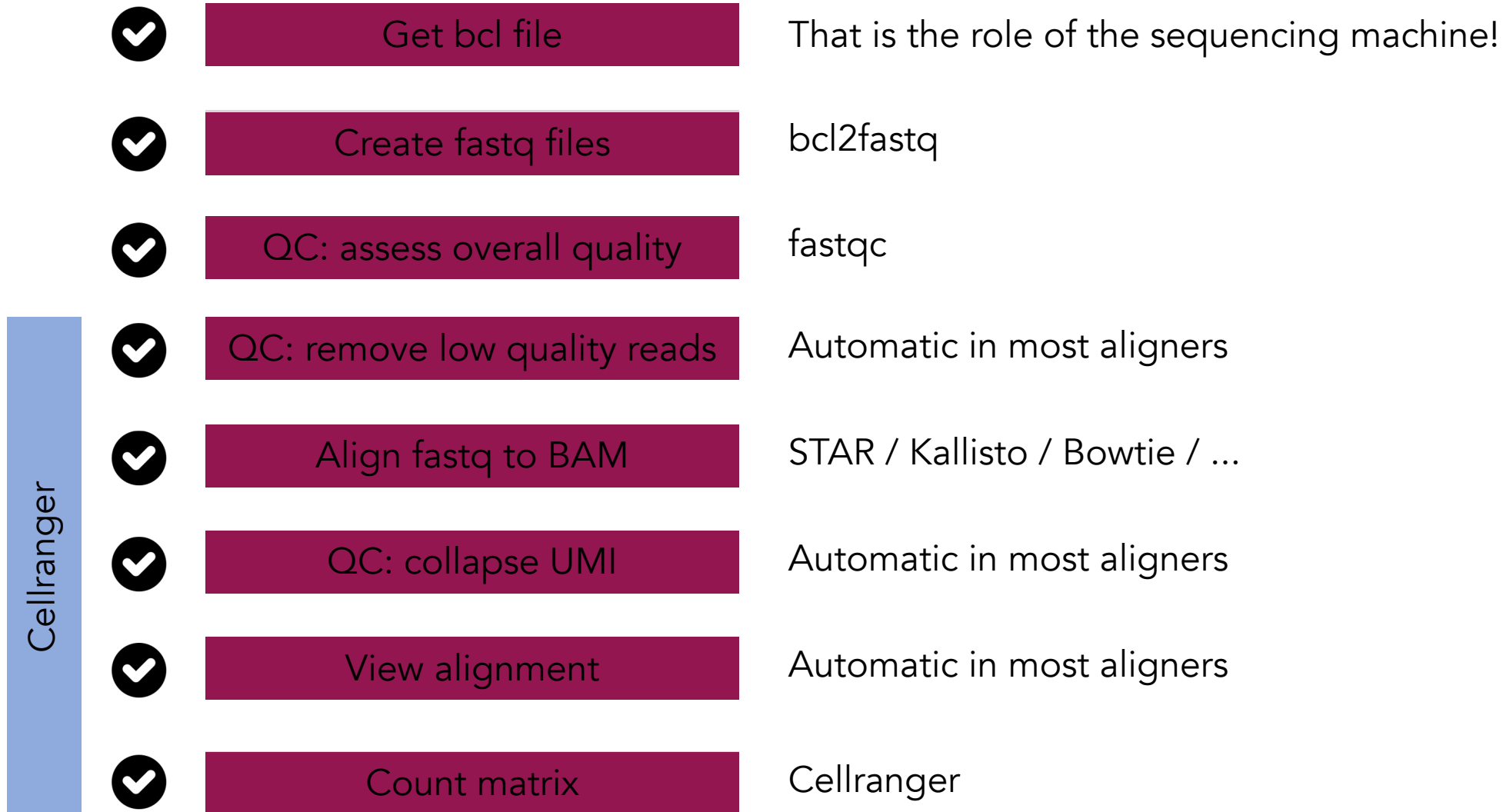
How deeply do I need to sequence?



Heimberg et al. Low Dimensionality in Gene Expression Data Enables the Accurate Extraction of Transcriptional Programs from Shallow Sequencing. Cell Systems 2016.

<https://satijalab.org/costpercell/>

## Flowchart: from .bcl to count matrix





**Any question?**

