

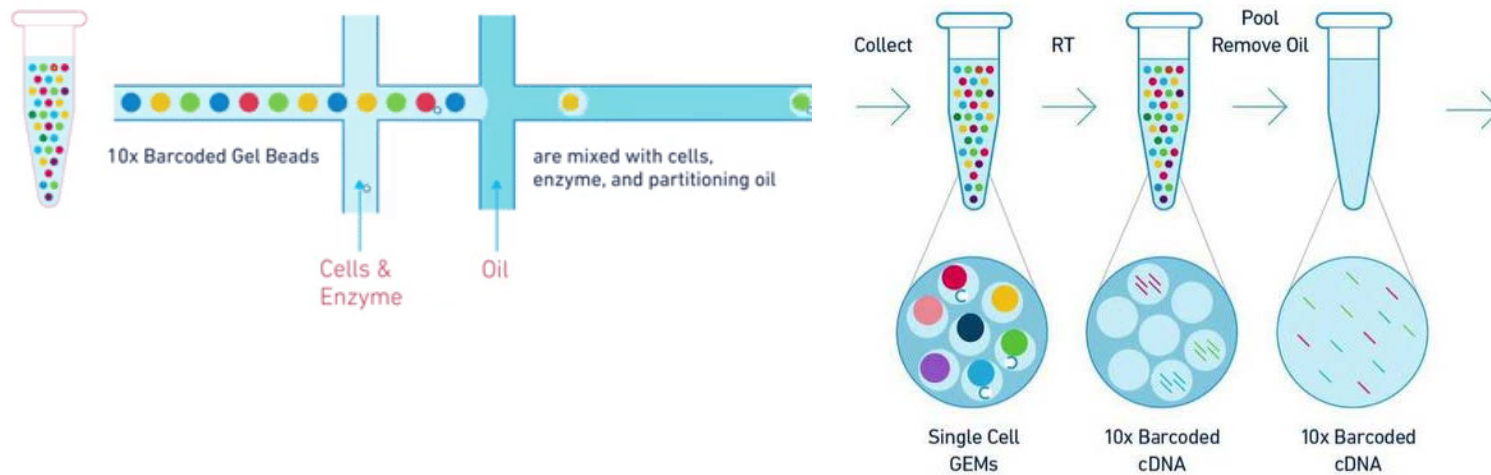
ELIXIR-IIB Training Platform
Single-Cell RNA Sequencing and Data Analysis

Theory Refresher and Software Overview: Cell Ranger

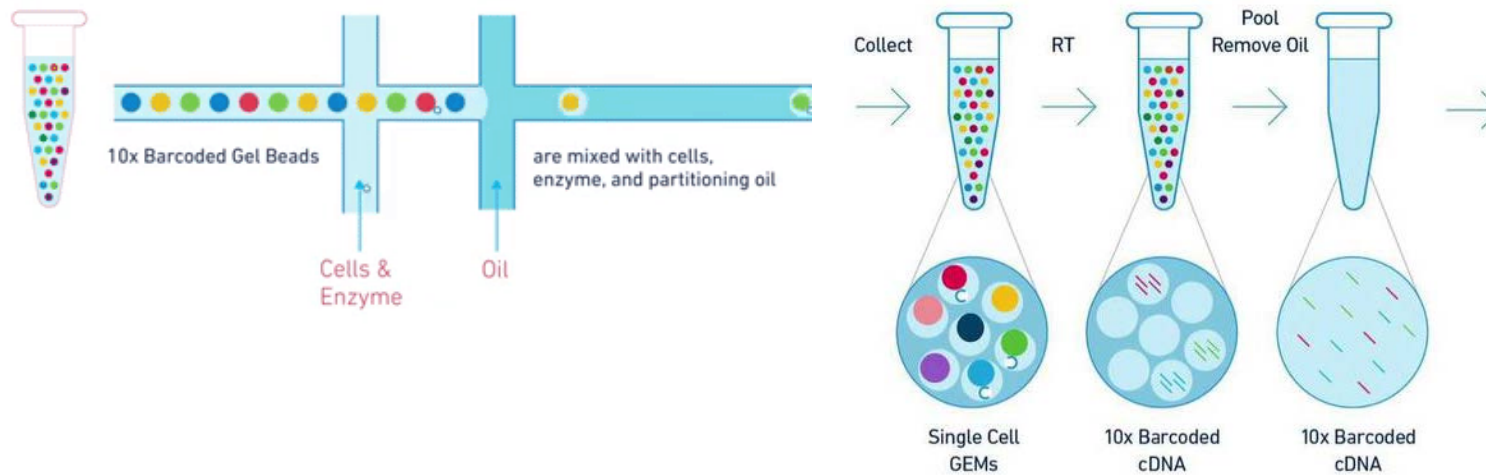
Francesco Panariello, Bioinformatician



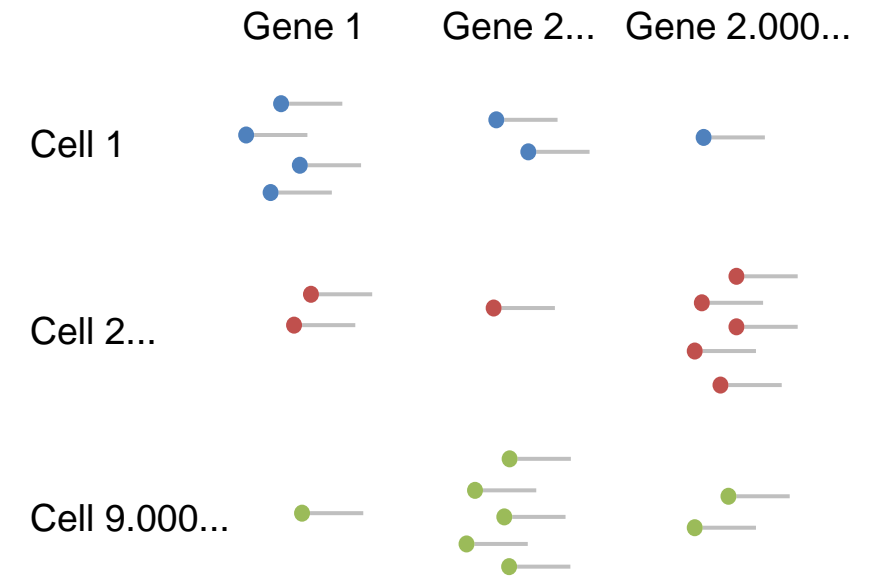
10x GemCode™ Technology for Single Cell Partitioning

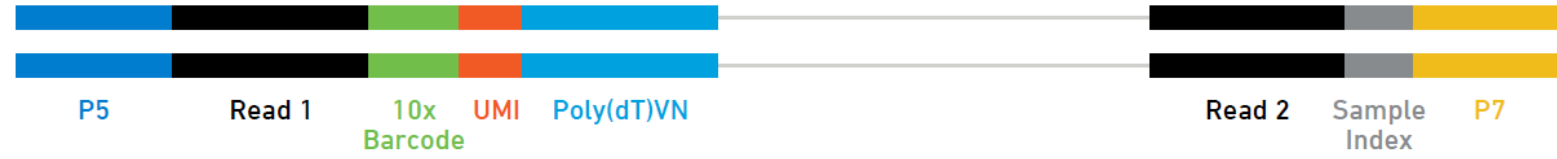


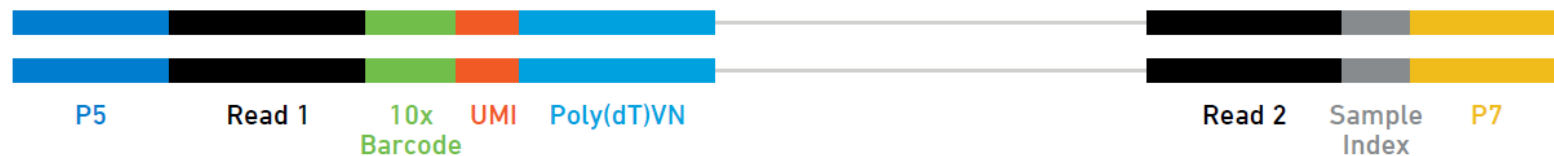
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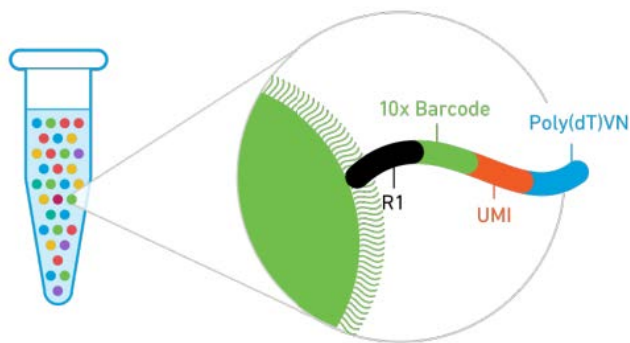
Transcriptional profiling of individual cells



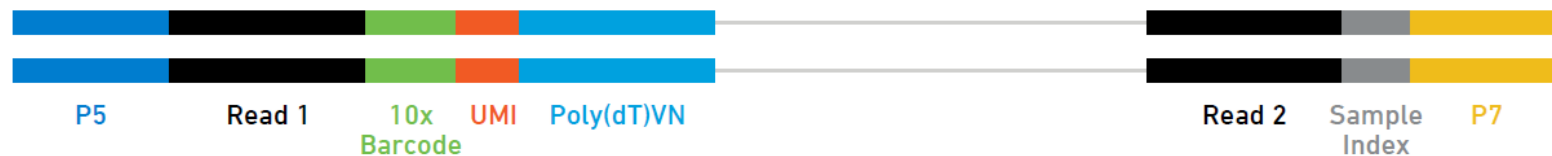




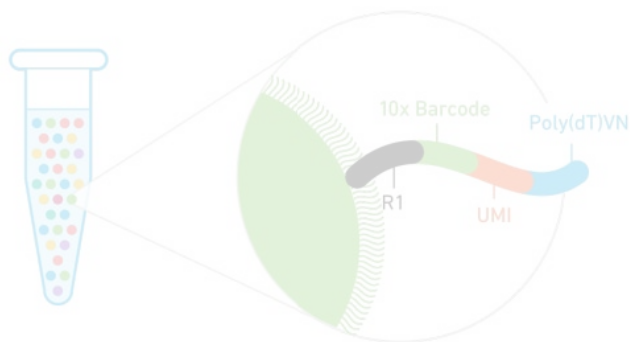
10x™ Barcode



The 16bp 10x barcode is unique to each Gel Bead and tells you which cell the transcript is from.

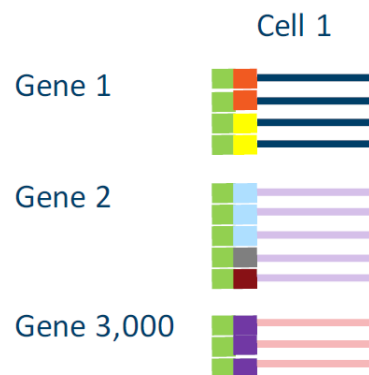


10x™ Barcode

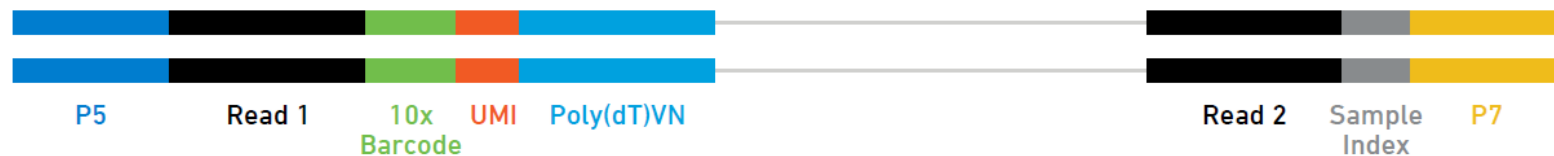


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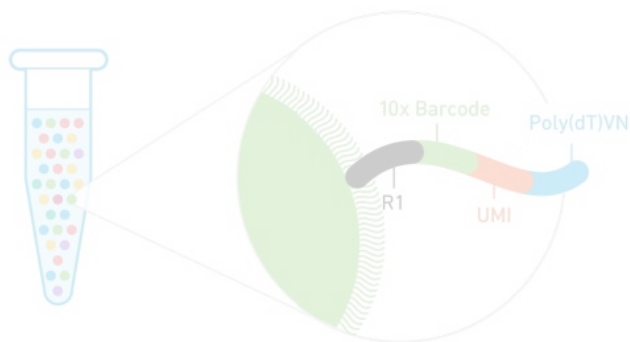
Unique Molecular Identifier (UMI)



The 10bp UMI enables accurate quantitation of cell expression levels.

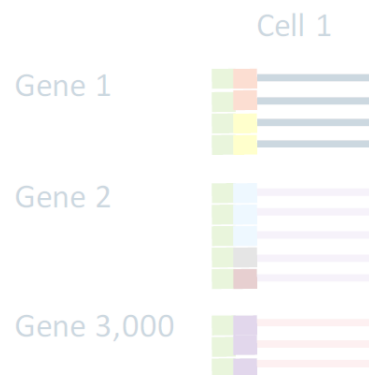


10x™ Barcode



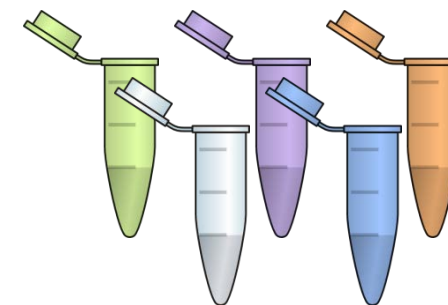
The 16bp 10x barcode is unique to each Gel Bead and tells you which cell the transcript is from.

Unique Molecular Identifier (UMI)



The 10bp UMI enables accurate quantitation of cell expression levels.

Sample index



The 8bp sample index allows to assign each barcoded read to its sample of origin.

ELIXIR-IIB Training Platform

Single-Cell RNA Sequencing and Data Analysis

**Theory Refresher and Software Overview:
Cell Ranger**

Cell Ranger™ Pipeline

RNA-seq bulk

Demultiplexing



FastQC
MultiQC

Quality control / Trimming

TrimGalore
Trimmomatic
BBDuK



Reads Alignment

STAR
BWA
Kalist
TopHat



Reads Quantification

HT-seq
featureCounts
RSEM



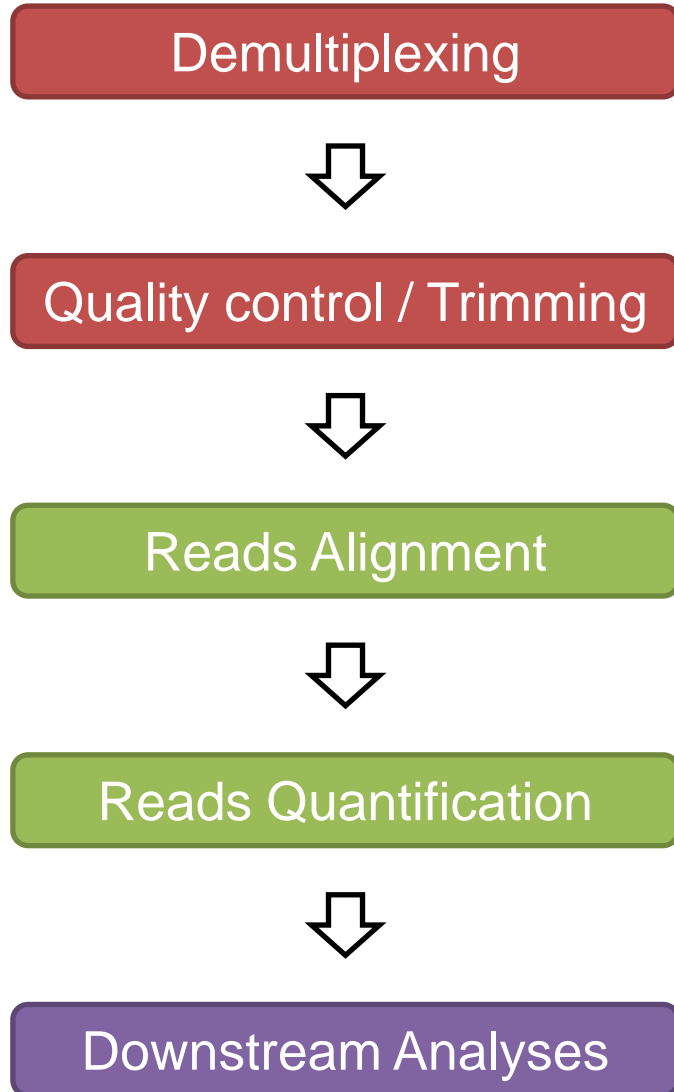
Downstream Analyses

Primary Analysis

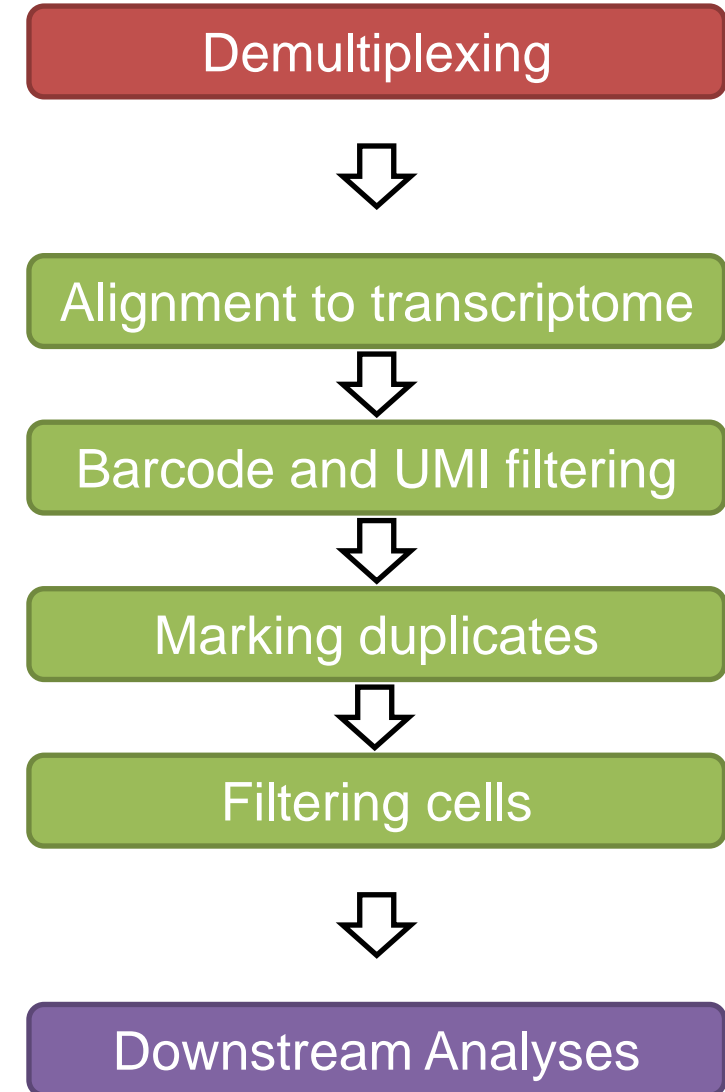
Secondary Analysis

Tertiary Analysis

RNA-seq bulk



scRNA-seq – Cell Ranger



Cell Ranger is a set of analysis pipelines that process Chromium single-cell RNA-seq output to align reads, generate feature-barcode matrices and perform clustering and gene expression analysis.

Cell Ranger includes four pipelines relevant to single-cell gene expression experiments

cellranger mkfastq

cellranger count

cellranger aggr

cellranger reanalyze

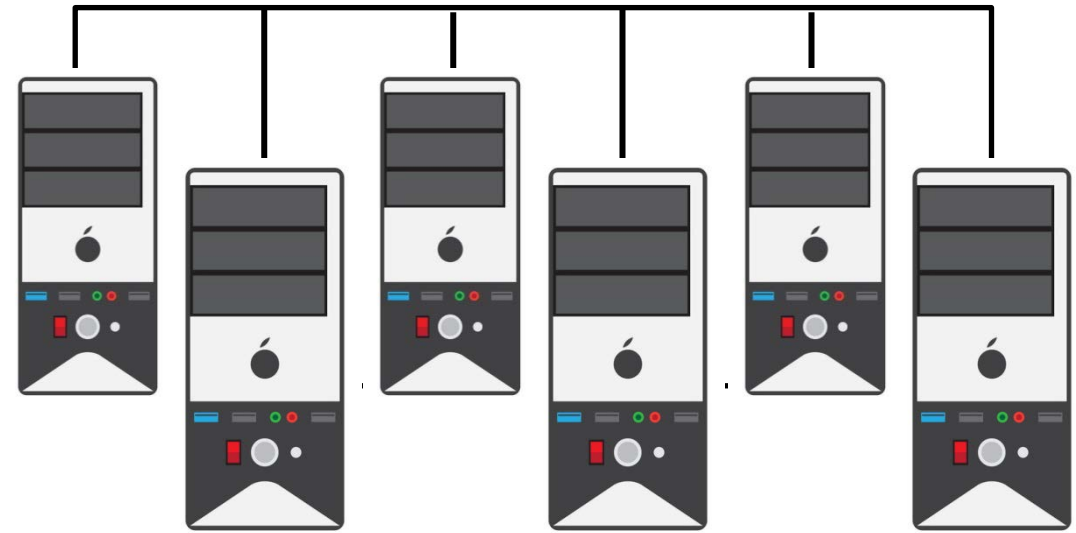


Cluster Mode

- Run on SGE and LSF
- Each node must have 8+ cores and 8GB+ RAM/core
- Shared filesystem between nodes (e.g. NFS)

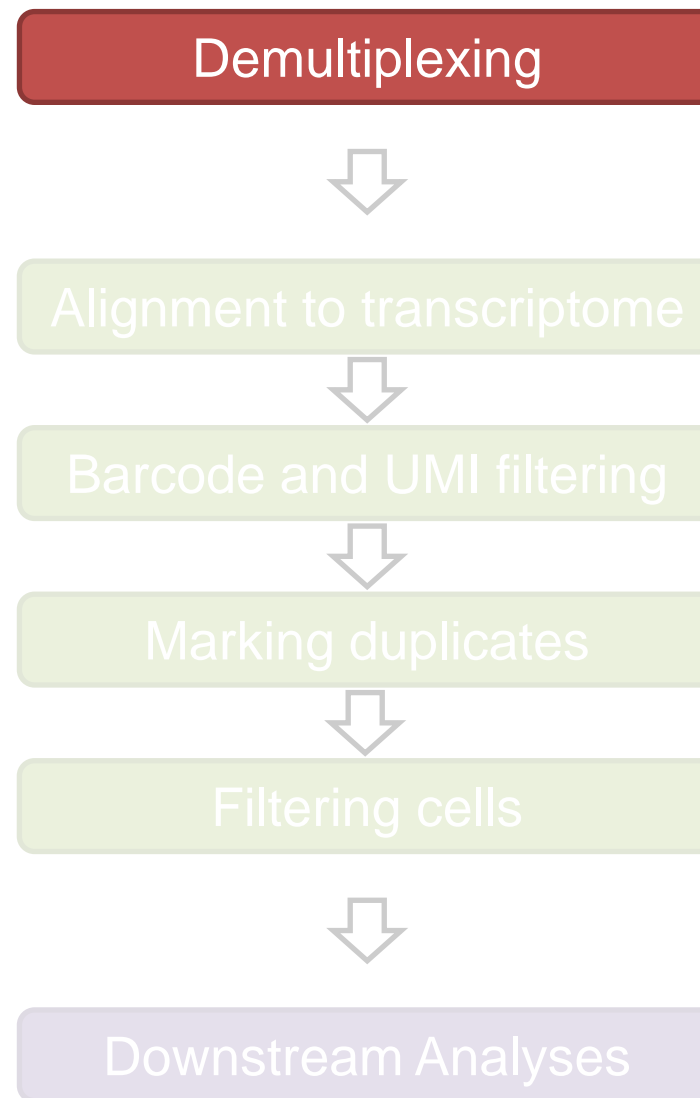
Local Mode

- Run on single, standalone Linux system
- CentOS/RedHat 5.2+ or Ubuntu 8.04+
- 8+ cores, 64GB RAM



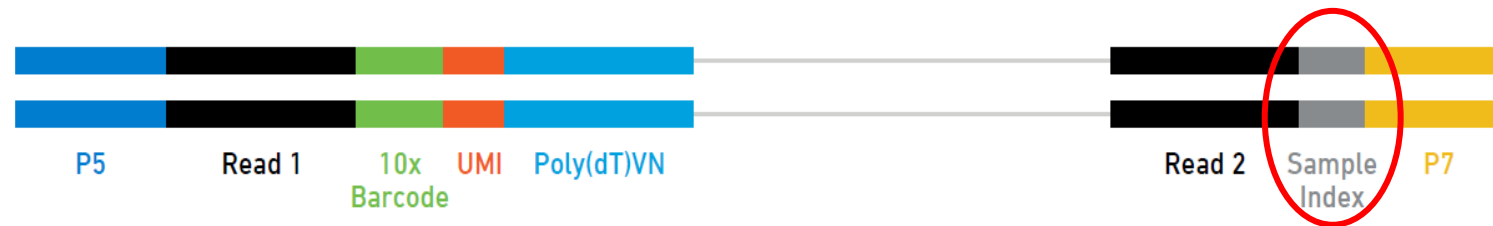
Demultiplexing is the step through which sequencing reads are divided into separate **fastq files** for each sample index.

```
@SN1083:379:H8VA1ADXX:2:1101:1248:2144 1:N:0:12  
CCTTAAAATGTACCCATTAGGCCTAAGTAGCTAGCTGGGCC  
+  
BBBBFFFFFIIIIFIII<FFFFFIIBBBUUUUFIIIIIDDDDIIFIFIFII
```



The *cellranger mkfastq* pipeline allows to demultiplex an Illumina sequencing run folder into FASTQ files.

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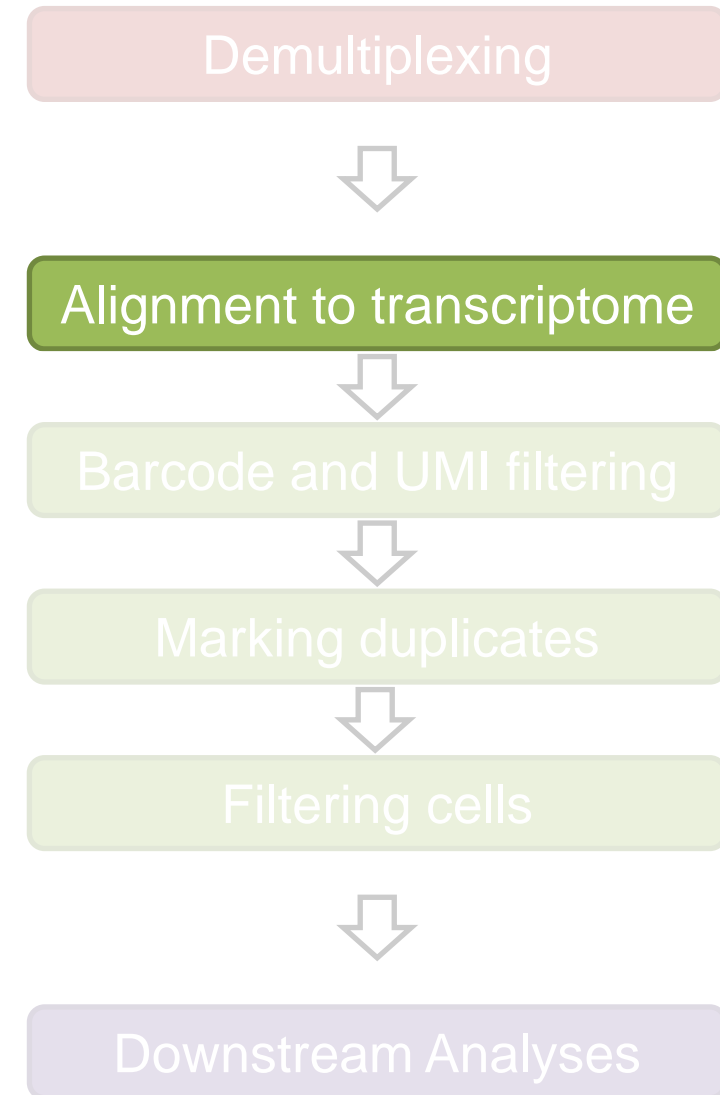
- 10x samples indices are used to assign reads to their sample of origin
- They are supplied on a 96 well plate
- Each 10x sample index is composed of 4 oligos
 - 10x index: SI-GA-A1
 - 4 Oligos: "GGTTTACT", "CTAAACGG", "TCGGCGTC", "AACCGTAA"

The *cellranger mkfastq* pipeline allows to demultiplex an Illumina sequencing run folder into FASTQ files.

The association between indexes and samples is provided through a samplesheet in .csv (comma-separated values) format.

```
Lane,Sample,Index  
*,Sample_1,SI-GA-A1  
*,Sample_2,SI-GA-A2  
*,Sample_3,SI-GA-A3
```


- **Read alignment** consists in the assignment of sequencing reads to the most likely locus of origin.
- Reads mapping can be performed on the genome, as well as on the transcriptome, in a **fasta** format.
- To speed up the process, fasta files are usually indexed.
- A **GTF** file is also used to provide information about gene location, biotype, etc.

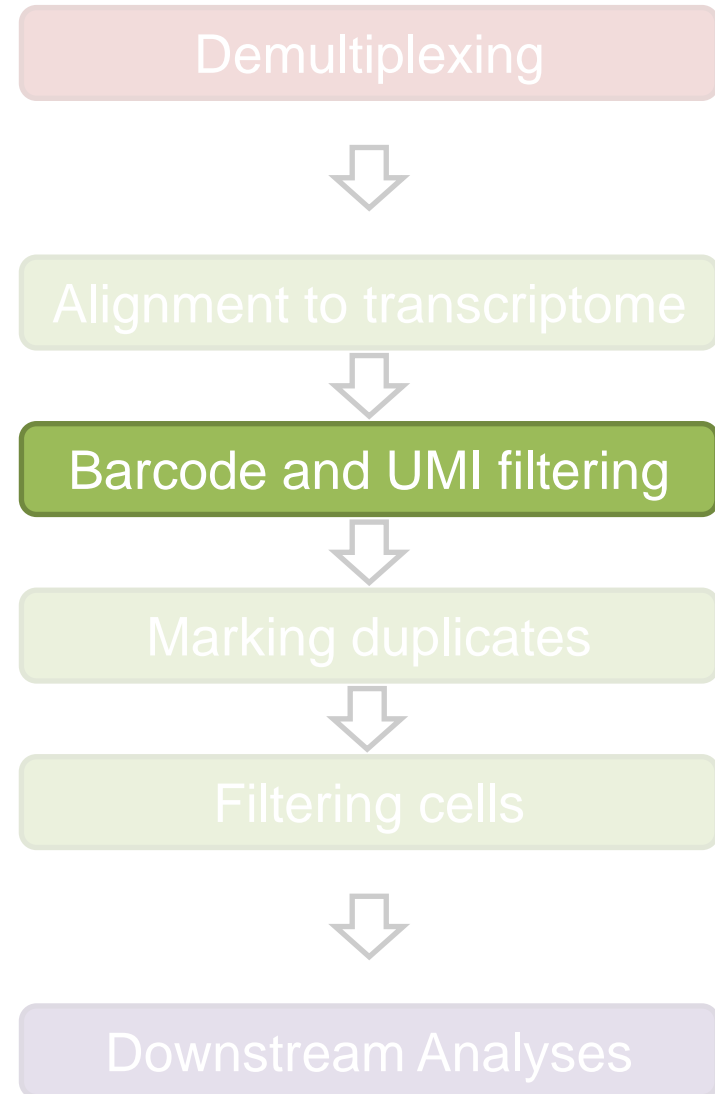


References

- 10x annotation uses ENSEMBL genomes and gene annotations.
- 10x pre-built references: Human (hg19 and GRCh38), Mouse, Human and Mouse.
- Bundled *mkgtf* utility filters a GTF file by key value pairs in the attributes column for transcript biotype (e.g. protein-coding, non-coding, linc RNA).
- Bundled *mkref* utility generates a 10x reference package from any FASTA and GTF gene file (STAR compatible).

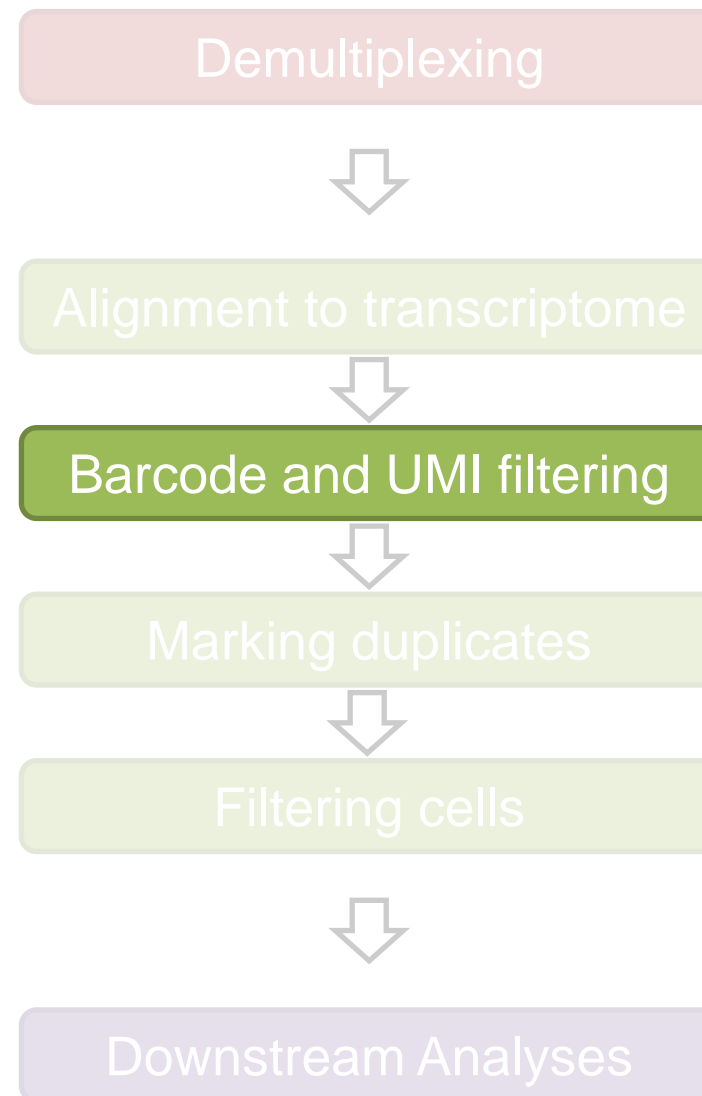
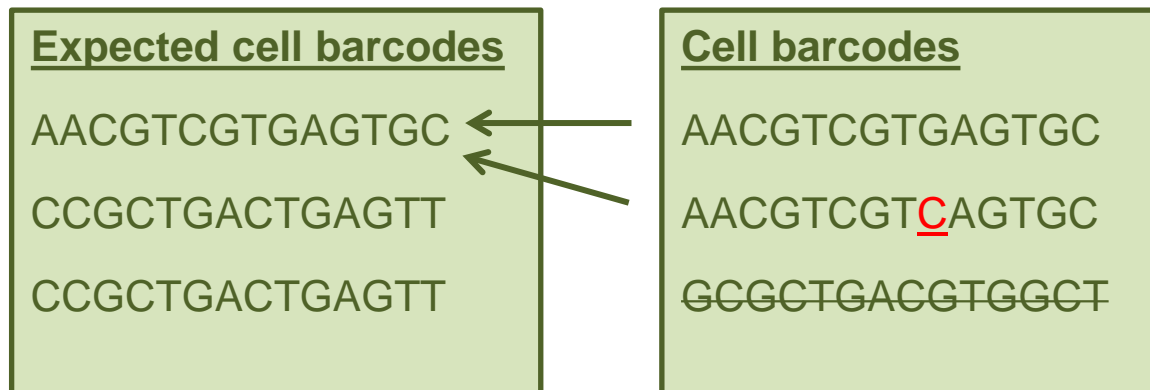
Alignment

- Alignment done via **STAR** (Spliced Transcripts Alignment to a Reference):
 - Robust, open-source, junction-aware RNA-seq aligner.
 - Aligns reads to the genome and transcriptome simultaneously.
- STAR memory usage:
 - *mkref* script builds STAR reference such that it uses max 16 GB of memory.
- Only use confidently mapped reads aligning to transcriptome.



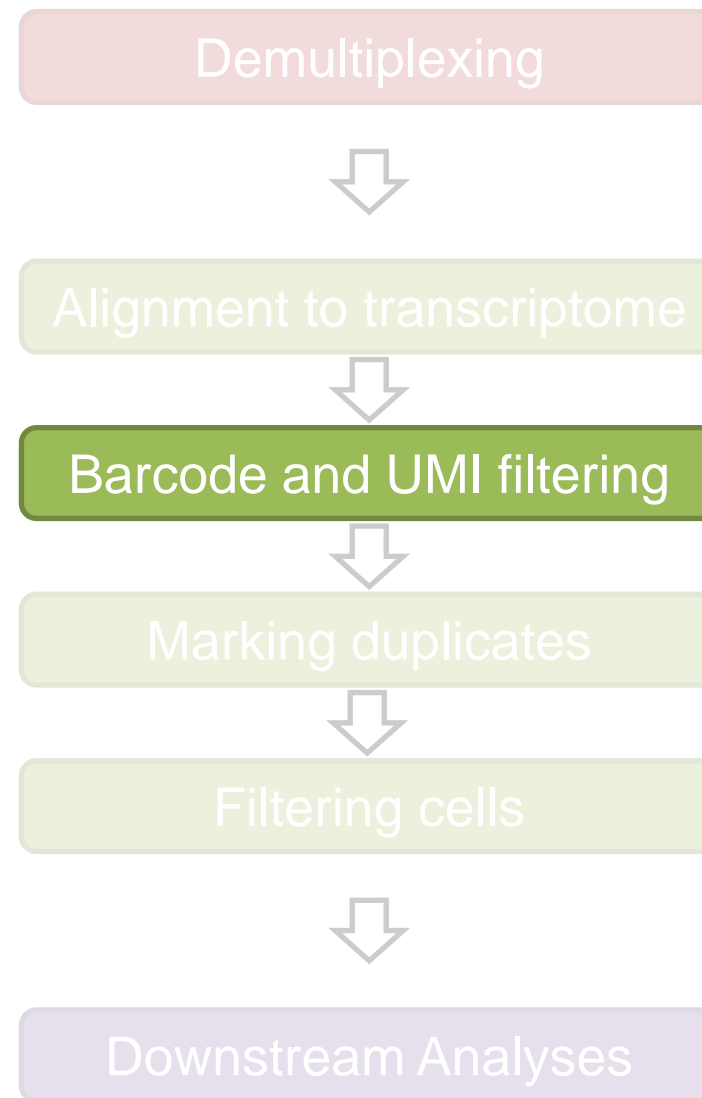
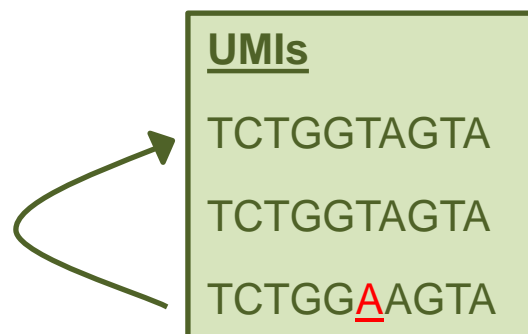
Cell barcodes:

- Must be on a static list of known cell barcode sequences
- May be **one** mismatch away from the list **ONLY IF** the mismatch occurs at a low-quality position (the barcode is then corrected)



UMIs:

- Must not be a homopolymer, e.g. AAAAAAAAAA
- Must not contain N
- Must not contain bases with base quality < 10
- UMIs that are 1 nucleotide mismatch away from a higher-count UMI are corrected to that UMI if they share a cell barcode and gene.



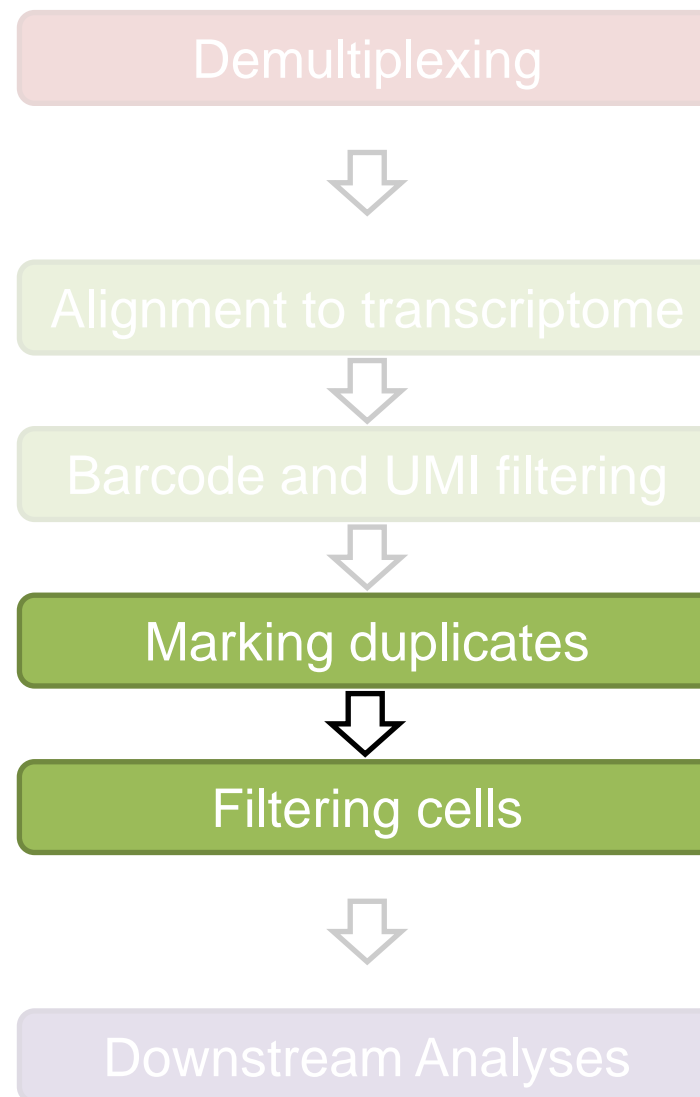
Marking duplicates:

- Record which reads are duplicates of the same RNA molecule.
- Count only the unique UMIs as unique RNA molecules.



Filtering cells:

- Sum UMI counts for each barcode.
- Select barcodes with total UMI count \geq 10% of the 99th percentile of the expected recovered cells.



The *cellranger count* pipeline allows to run all the previous steps (secondary analysis), once per sample.

```
cellranger count --sample=Sample_1 --transcriptome=refdata-cellranger/GRCh38 --fastqs=HAD58ADXX/Sample_1
```

BAM – Genome-Aligned Reads

- Indexed BAM containing position-sorted, aligned reads
- Barcodes and UMIs attached as standard tags

Cell Ranger™ Pipeline: Output files

BAM – Genome-Aligned Reads

- Indexed BAM containing position-sorted, aligned reads
- Barcodes and UMIs attached as standard tags

MEX – Gene/Barcode Matrix

- “Market Exchange” format, a sparse matrix representation
- Suitable for downstream analysis in Python and R

Gene	ATCAGGGACAGA	AGGGAAAATTGA	TTGCCTTACGCG	TGGCGAAGAGAT	TACAATTAAGGC
LOXL4	0	0	0	0	0
PYROXD2	1	0	1	1	0
HPS1	23	12	9	8	3
CNNM1	0	2	1	0	0
GOT1	22	6	7	9	3

Cell Ranger™ Pipeline: Output files

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.cloupe File - Analysis

- 2D projections
- Cell clustering
- Differential expression
- Interactive exploration

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Cell Ranger™ Pipeline: Output files

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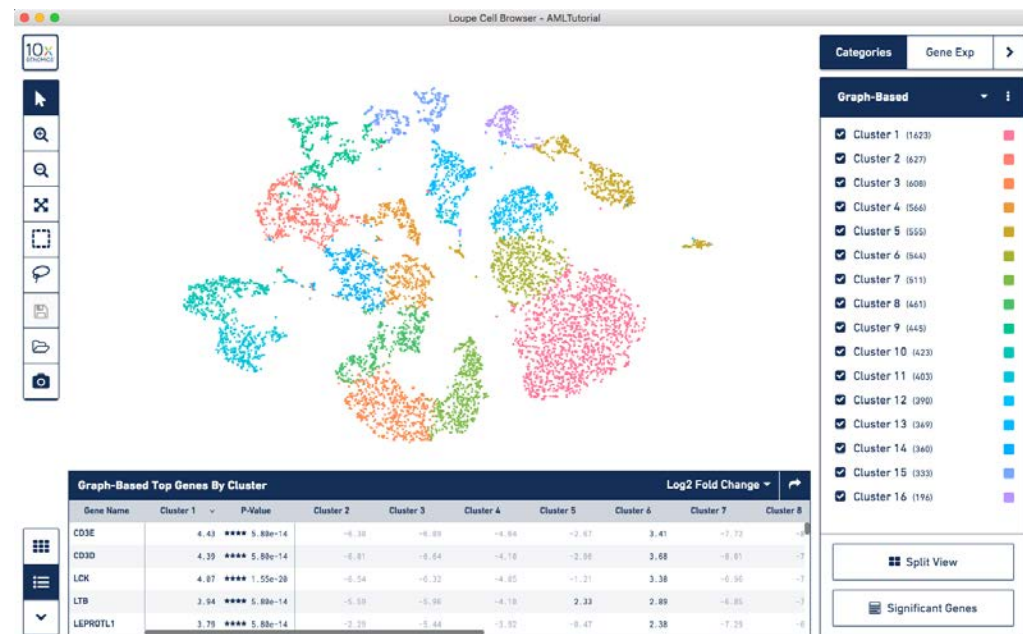
.cloupe File - Analysis

- 2D projections
- Cell clustering
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- Interactive exploration

HTML, CSV – Run Summary

- Run metrics and basic static visualizations

Gene	ATCAGGGACAGA	AGGGAAAATTGA	TTGCCTTACGCG	TGGCGAAGAGAT	TACAATTAAGGC
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Estimated Number of Cells

8,391

Mean Reads per Cell

93,441

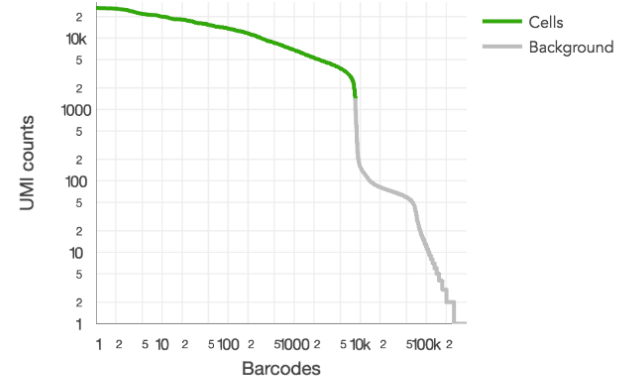
Median Genes per Cell

1,296

Sequencing

Number of Reads	784,064,148
Valid Barcodes	98.5%
Reads Mapped Confidently to Transcriptome	61.4%
Reads Mapped Confidently to Exonic Regions	65.3%
Reads Mapped Confidently to Intronic Regions	24.0%
Reads Mapped Confidently to Intergenic Regions	3.4%
Sequencing Saturation	90.5%
Q30 Bases in Barcode	98.2%
Q30 Bases in RNA Read	78.9%
Q30 Bases in Sample Index	96.4%
Q30 Bases in UMI	98.2%

Cells

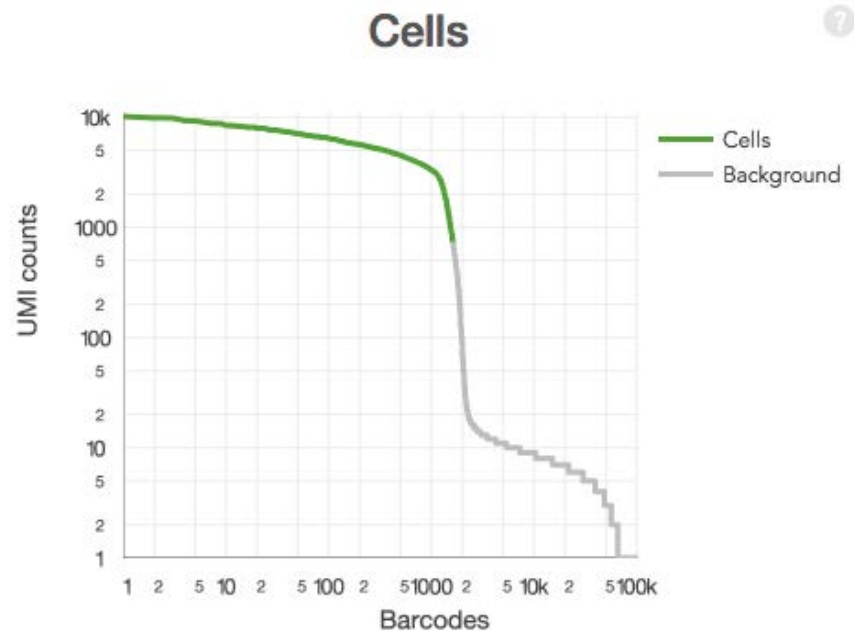


Estimated Number of Cells	8,391
Fraction Reads in Cells	93.2%
Mean Reads per Cell	93,441
Median Genes per Cell	1,296
Total Genes Detected	21,424
Median UMI Counts per Cell	4,082

Sample

Name	pbmc8k
Description	Peripheral blood mononuclear cells (PBMCs) from a healthy donor
Transcriptome	GRCh38
Chemistry	Single Cell 3' v2
Cell Ranger Version	1.3.0

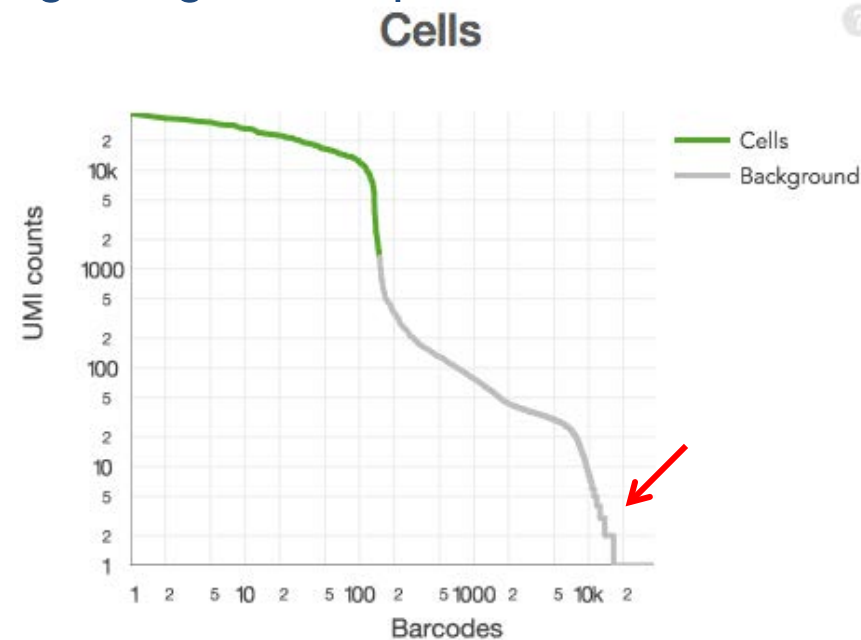
Typical Sample Profile



Defined cliff and knee

Metric	Value
Barcodes	> 90,000
Cell Barcodes	> 1,000
UMIs	> 10,000

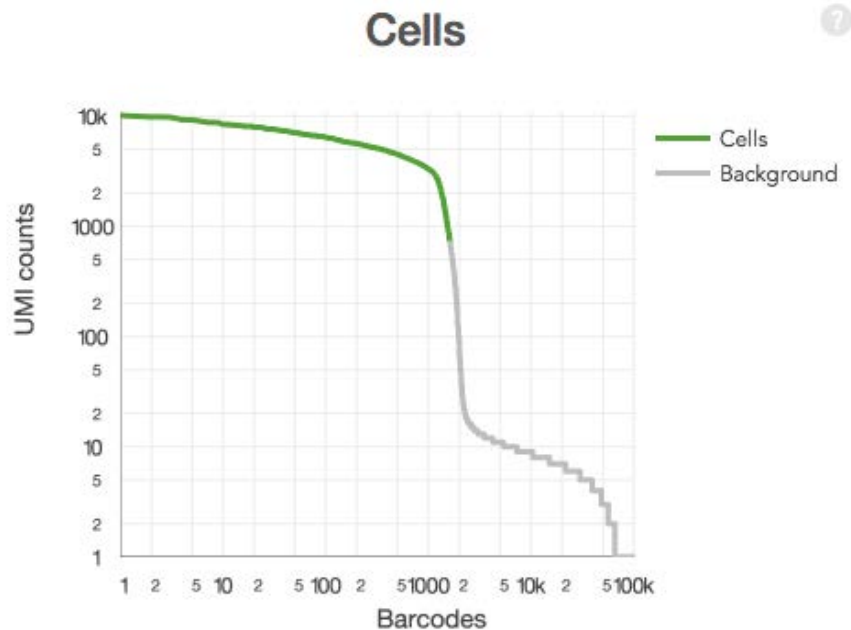
Low Barcode Counts (e.g. Clog, low-depth, low ambient RNA)



Low number of barcodes detected

Metric	Value
Barcodes	~ 15,000
Cell Barcodes	> 100
UMIs	> 10,000

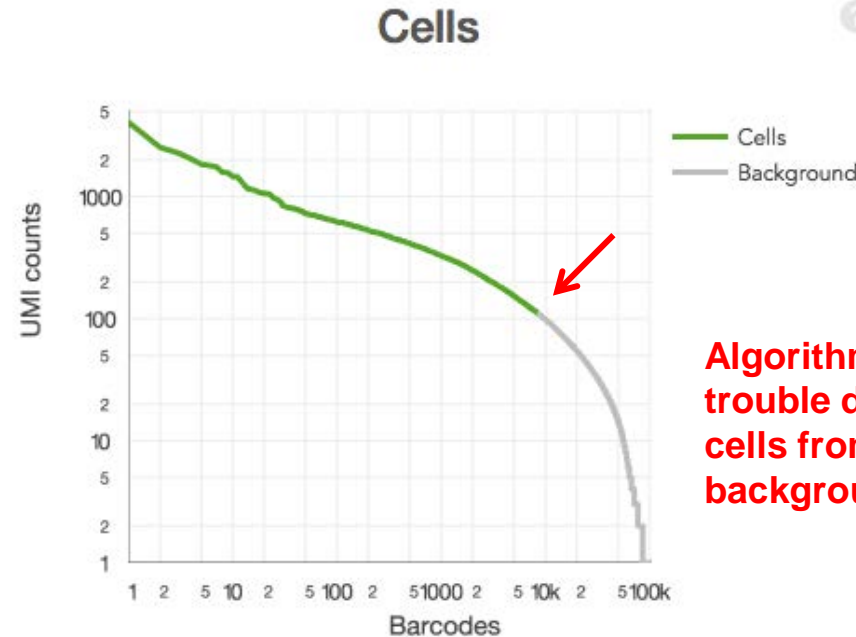
Typical Sample Profile



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Loss of Single Cell Behaviour (e.g. Lysis or Wetting failure)

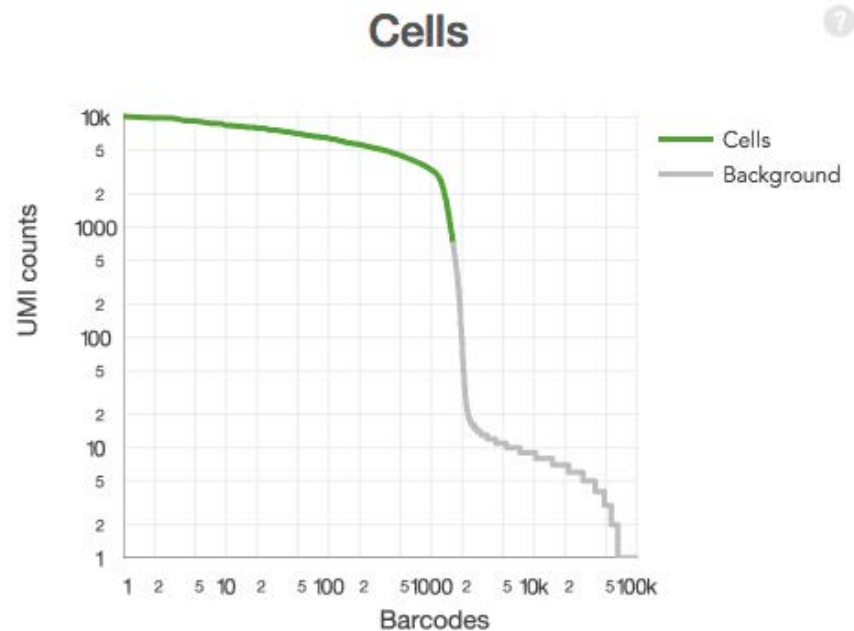


Algorithm has trouble discerning cells from the background

Lack of defined cliff and knee

Metric	Value
Barcodes	> 90,000
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UMIs	> 10,000

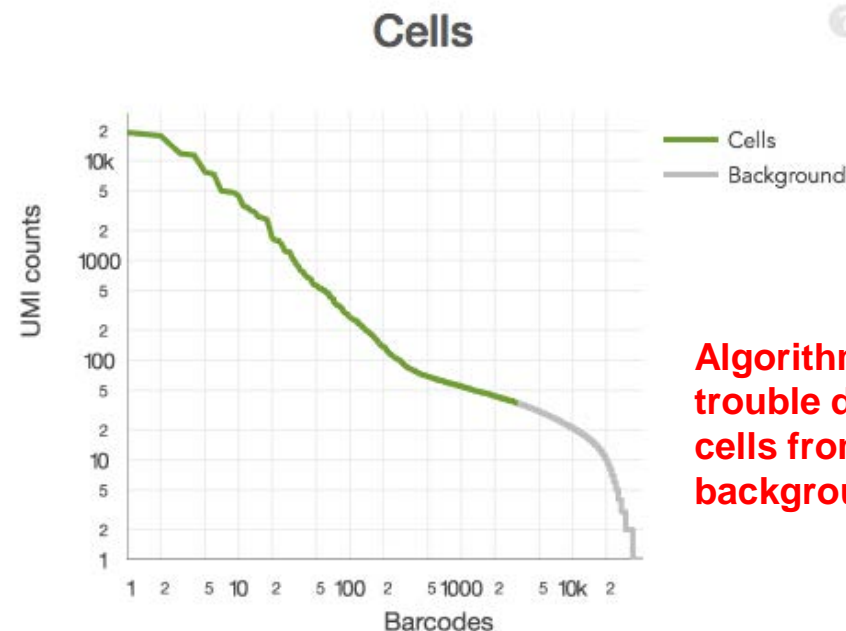
Typical Sample Profile



Defined cliff and knee

Metric	Value
Barcodes	> 90,000
Cell Barcodes	> 1,000
UMIs	> 10,000

Loss of Single Cell Behaviour (e.g. High fraction of ambient RNA)



Algorithm has trouble discerning cells from the background

Lack of defined cliff and knee

Metric	Value
Barcodes with > 1,000 UMIs	Few

Estimated Number of Cells

8,391

Mean Reads per Cell

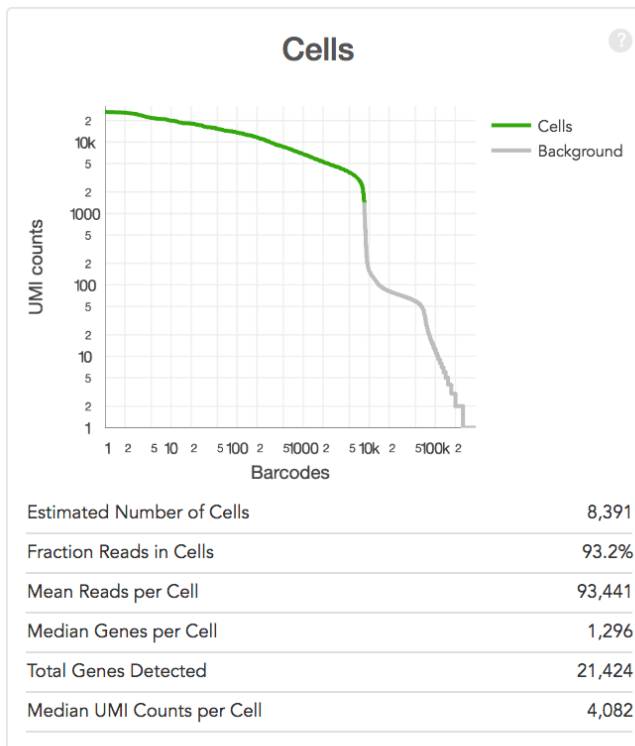
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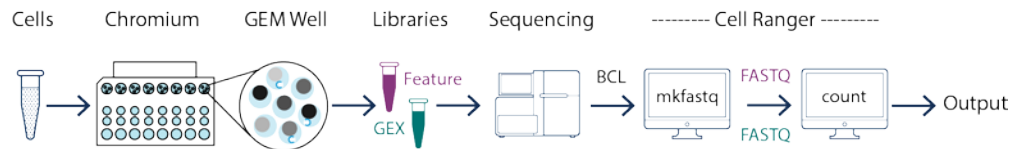
Reads confidently mapped to transcriptome (<30%)

- Reads mapped to wrong genome or different strain
- Read length is too short
- Custom reference contains overlapping genes

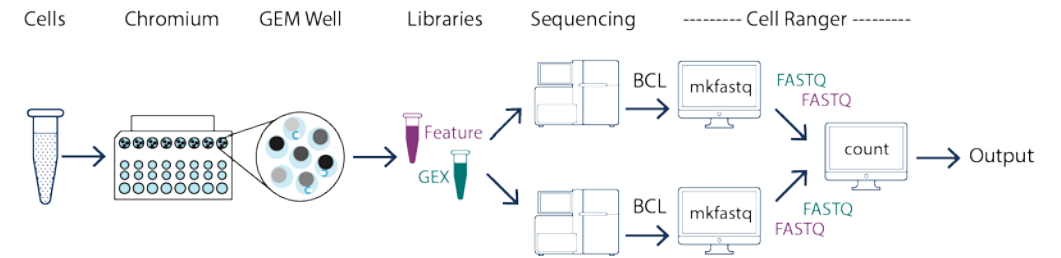
Read2 Q30 metrics are low (<70%)

- Sequencing problems
- Suboptimal loading concentration on sequencer

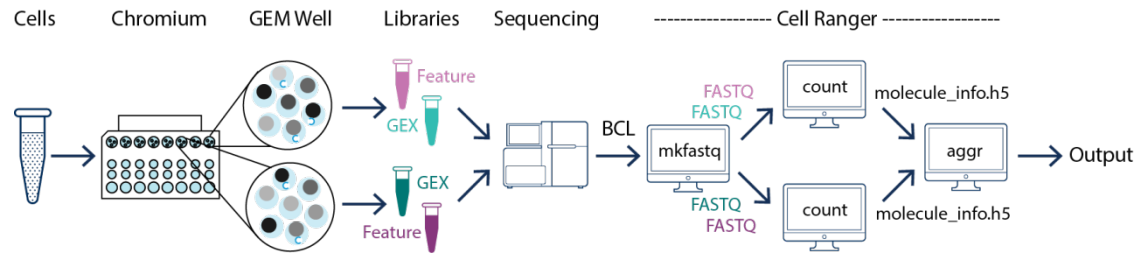
One Sample, One GEM Well, One Flowcell



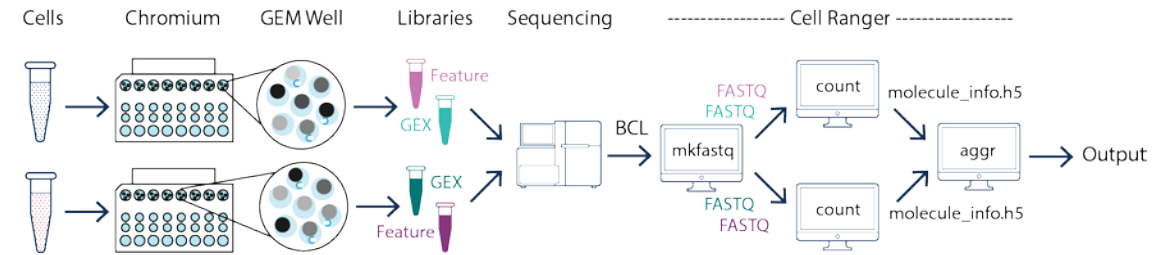
One Sample, One GEM Well, Multiple Flowcells



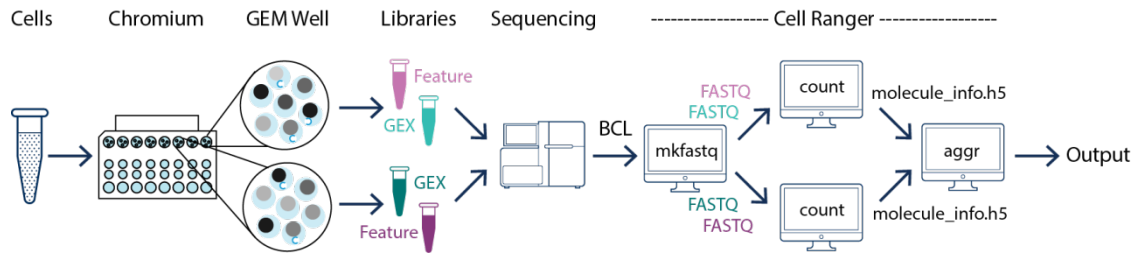
One Sample, Multiple GEM Wells, One Flowcell



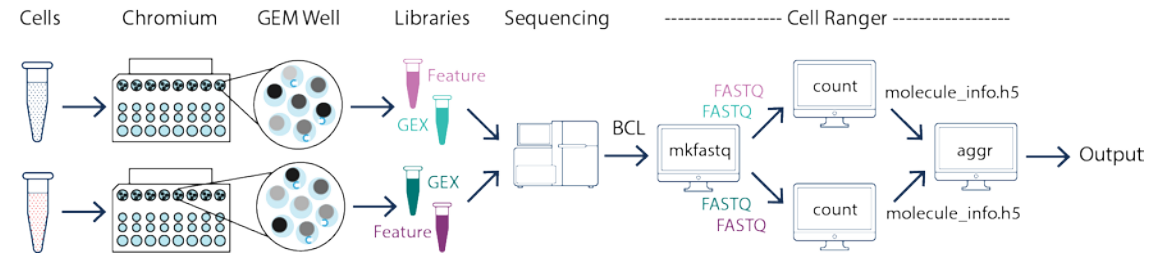
Multiple Samples, Multiple GEM Wells, One Flowcell



One Sample, Multiple GEM Wells, One Flowcell

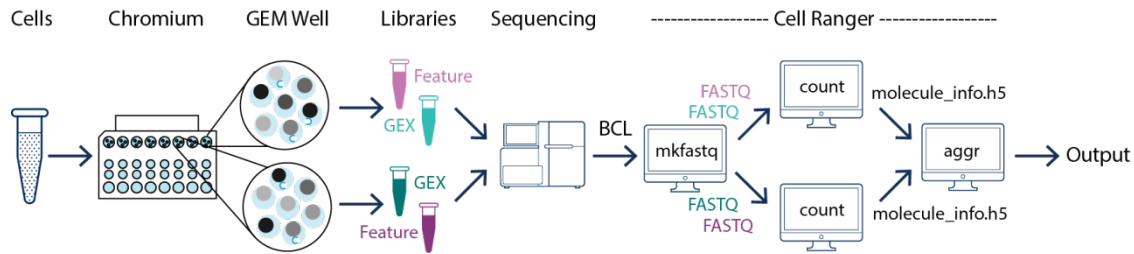


Multiple Samples, Multiple GEM Wells, One Flowcell

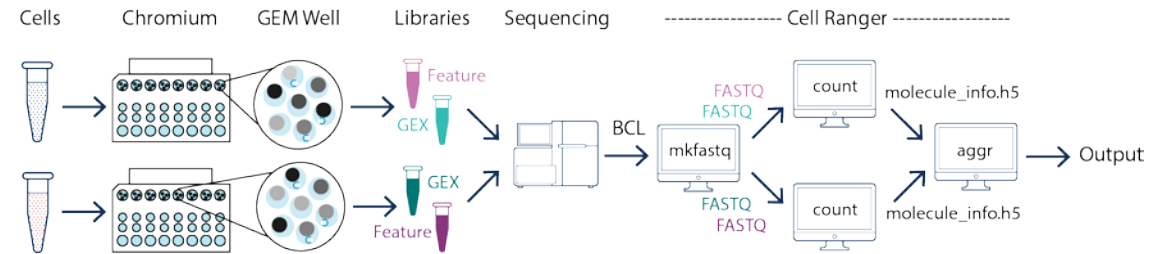


The *cellranger aggr* pipeline pools the results from single runs of cellranger counts, using the *molecule_info.h5* files

One Sample, Multiple GEM Wells, One Flowcell



Multiple Samples, Multiple GEM Wells, One Flowcell



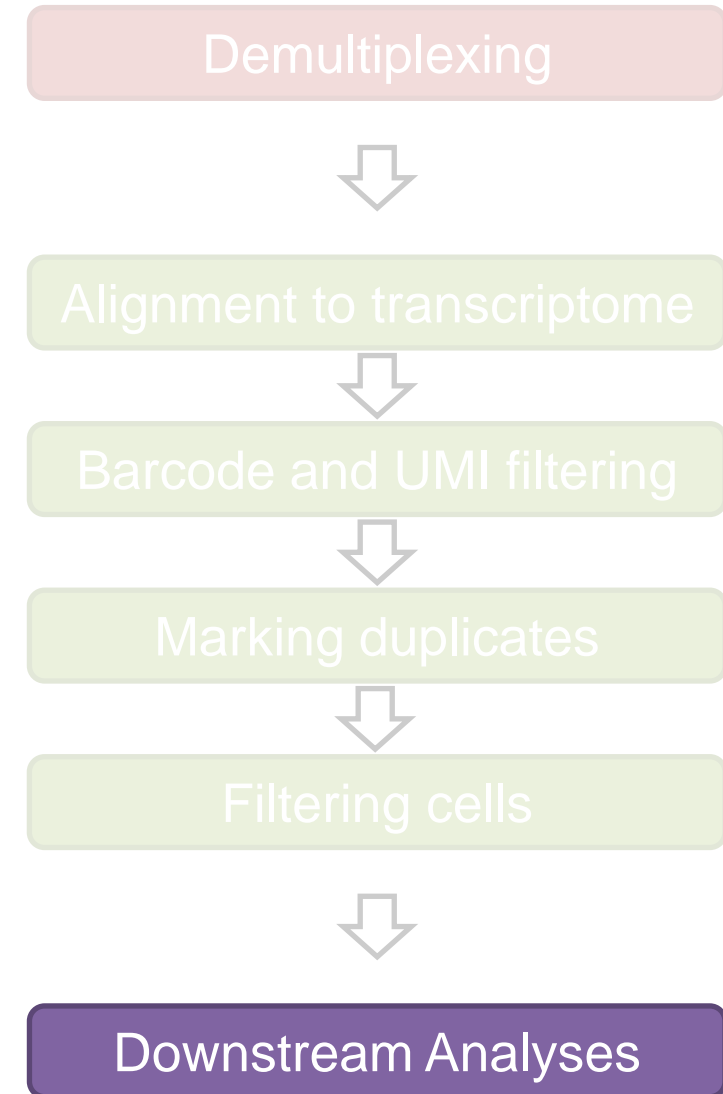
The *cellranger aggr* pipeline pools the results from single runs of cellranger counts, using the *molecule_info.h5* files

WARNING!!

By default, the reads from each GEM well are subsampled such that all GEM wells have the same effective sequencing depth, measured in terms of confidently mapped reads per cell.

Cellranger counts produces a .cloupe file (accessible through the Desktop app **LOUPE**) containing standard downstream analyses, run with default parameters:

- 2D projections
- Cell clustering
- Differential expression
- Interactive exploration

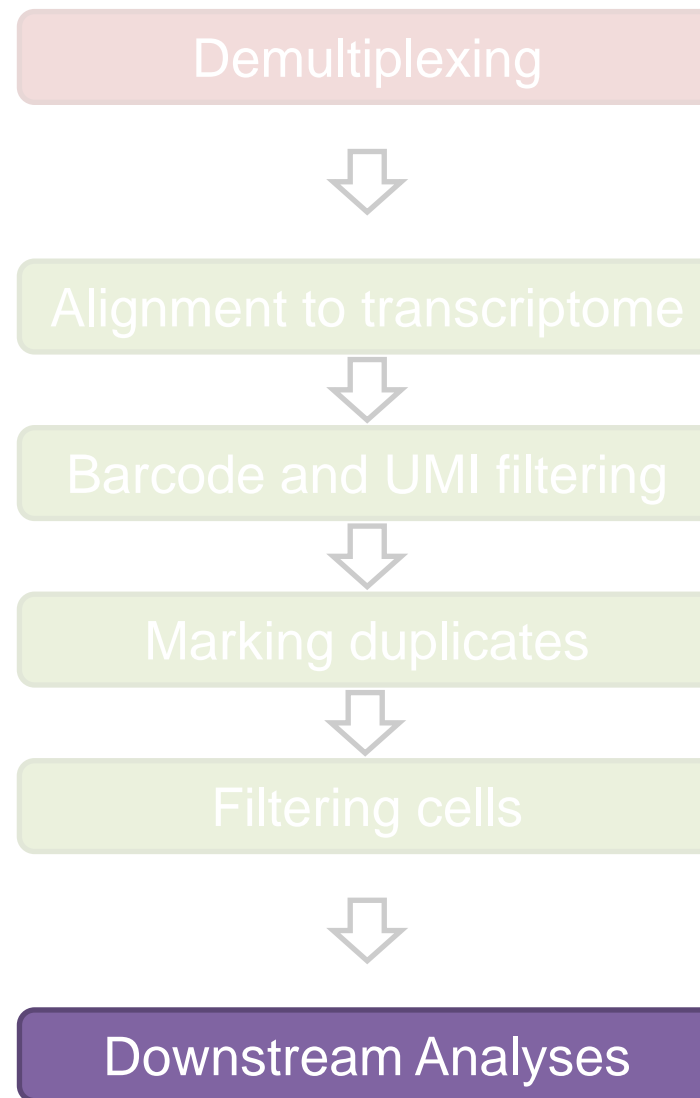


Cell Ranger™ Pipeline: Downstream analysis – cellranger reanalyze

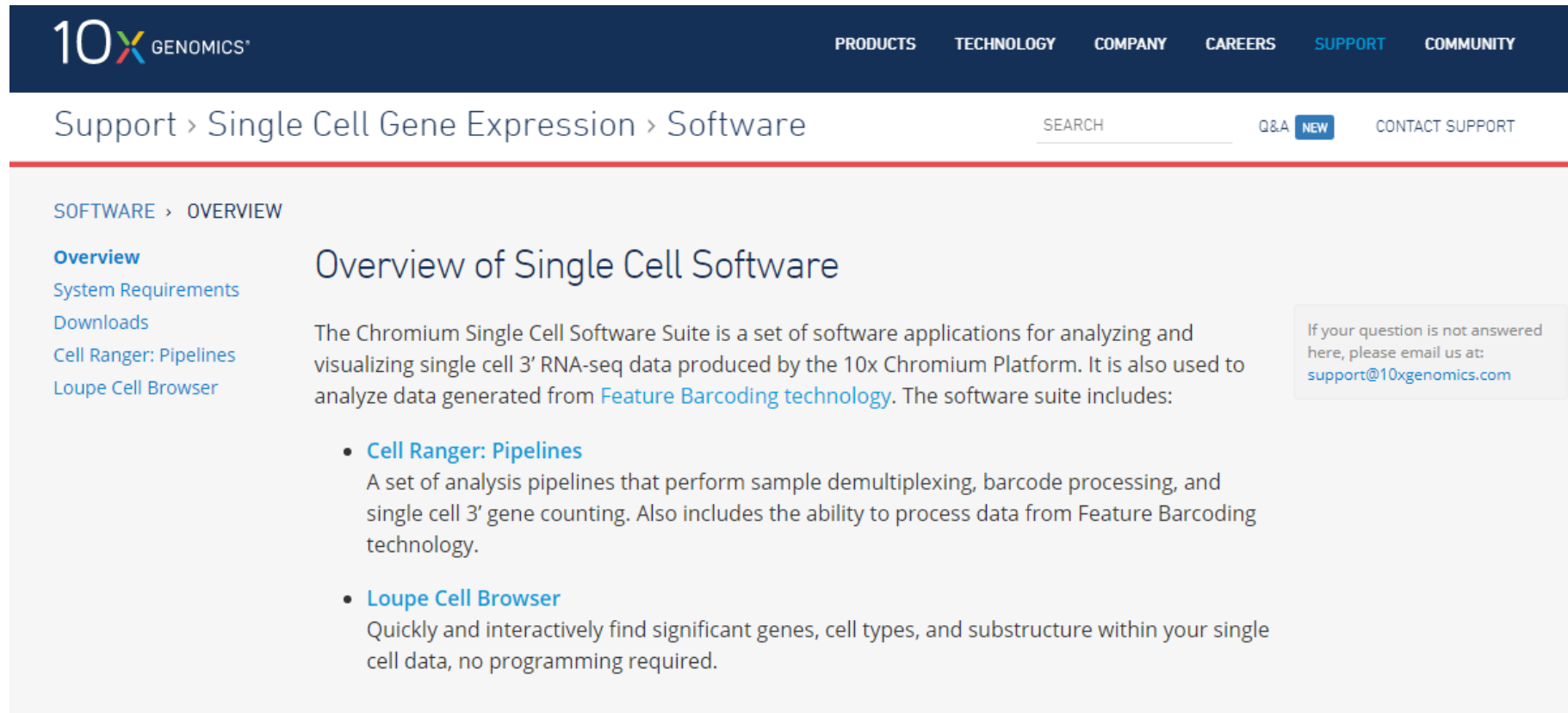
Cellranger counts produces a .cloupe file (accessible through the Desktop app **LOUPE**) containing standard downstream analyses, run with default parameters:

- 2D projections
- Cell clustering
- Differential expression
- Interactive exploration

The *cellranger reanalyze* pipeline re-runs tertiary analyses performed on the feature-barcode matrix using custom parameter settings.



<https://support.10xgenomics.com/single-cell-gene-expression/software/overview/welcome>



The screenshot shows the 10x Genomics support page for Single Cell Software Overview. The page has a dark blue header with the 10x Genomics logo on the left and navigation links for PRODUCTS, TECHNOLOGY, COMPANY, CAREERS, SUPPORT, and COMMUNITY on the right. Below the header, the breadcrumb trail reads "Support > Single Cell Gene Expression > Software". To the right of the breadcrumb is a search bar and a "Q&A NEW" button. The main content area has a sub-breadcrumb "SOFTWARE > OVERVIEW" and a list of links: Overview (highlighted), System Requirements, Downloads, Cell Ranger: Pipelines, and Loupe Cell Browser. The main heading is "Overview of Single Cell Software". The text describes the Chromium Single Cell Software Suite as a set of applications for analyzing and visualizing single cell 3' RNA-seq data. It lists two software components: Cell Ranger: Pipelines and Loupe Cell Browser, each with a brief description of their functions. A callout box on the right side of the page provides contact information for support.

10x GENOMICS™

PRODUCTS TECHNOLOGY COMPANY CAREERS SUPPORT COMMUNITY

Support > Single Cell Gene Expression > Software

SEARCH Q&A NEW CONTACT SUPPORT

SOFTWARE > OVERVIEW

Overview
System Requirements
Downloads
Cell Ranger: Pipelines
Loupe Cell Browser

Overview of Single Cell Software

The Chromium Single Cell Software Suite is a set of software applications for analyzing and visualizing single cell 3' RNA-seq data produced by the 10x Chromium Platform. It is also used to analyze data generated from [Feature Barcoding technology](#). The software suite includes:

- **Cell Ranger: Pipelines**
A set of analysis pipelines that perform sample demultiplexing, barcode processing, and single cell 3' gene counting. Also includes the ability to process data from Feature Barcoding technology.
- **Loupe Cell Browser**
Quickly and interactively find significant genes, cell types, and substructure within your single cell data, no programming required.

If your question is not answered here, please email us at: support@10xgenomics.com

ELIXIR-IIB Training Platform

Single-Cell RNA Sequencing and Data Analysis

**Theory Refresher and Software Overview:
Cell Ranger**

Data pre-filtering

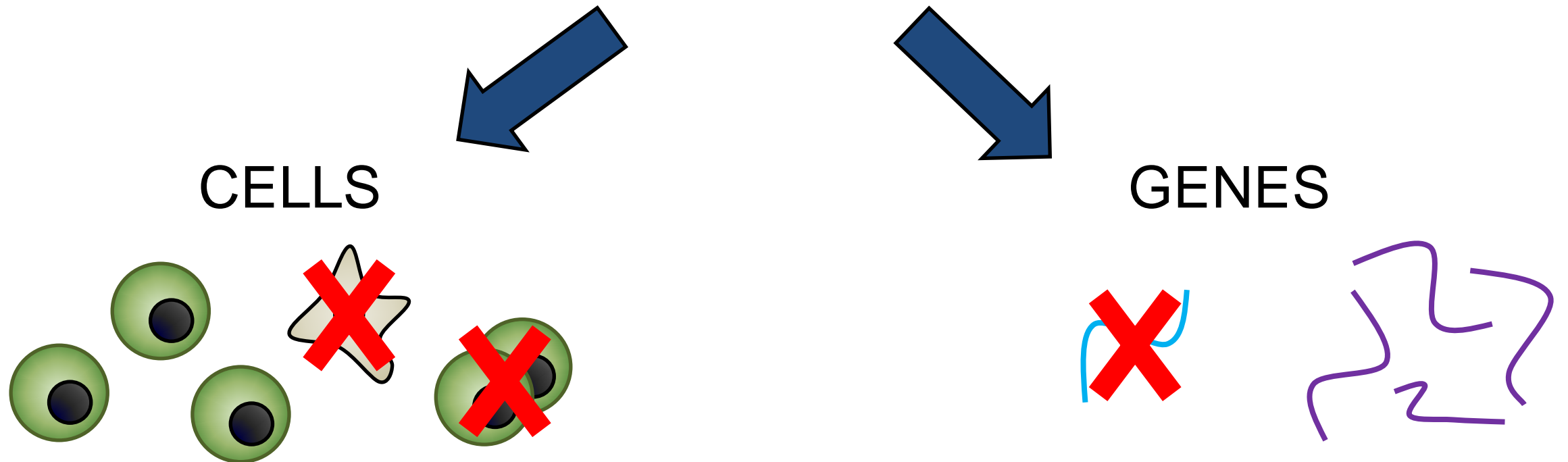
Several factors (variables) influence scRNA-seq data:

- Multiplets
- Apoptotic cells
- Drop-out effect

Data pre-filtering overview

Several factors (variables) influence scRNA-seq data:

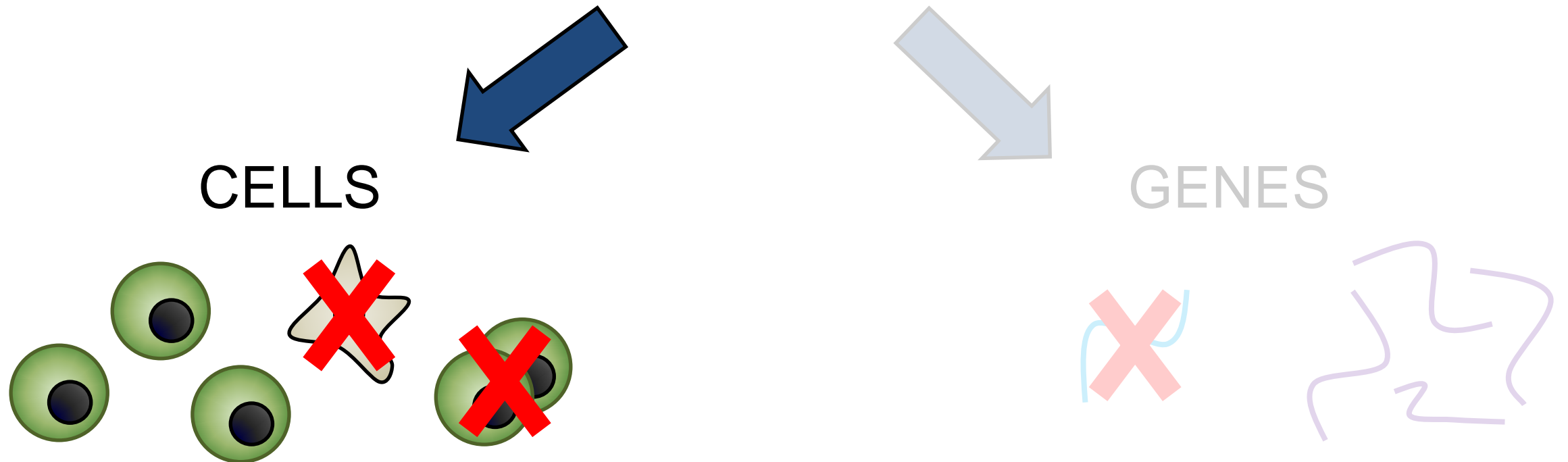
- Multiplets
- Apoptotic cells
- Drop-out effect



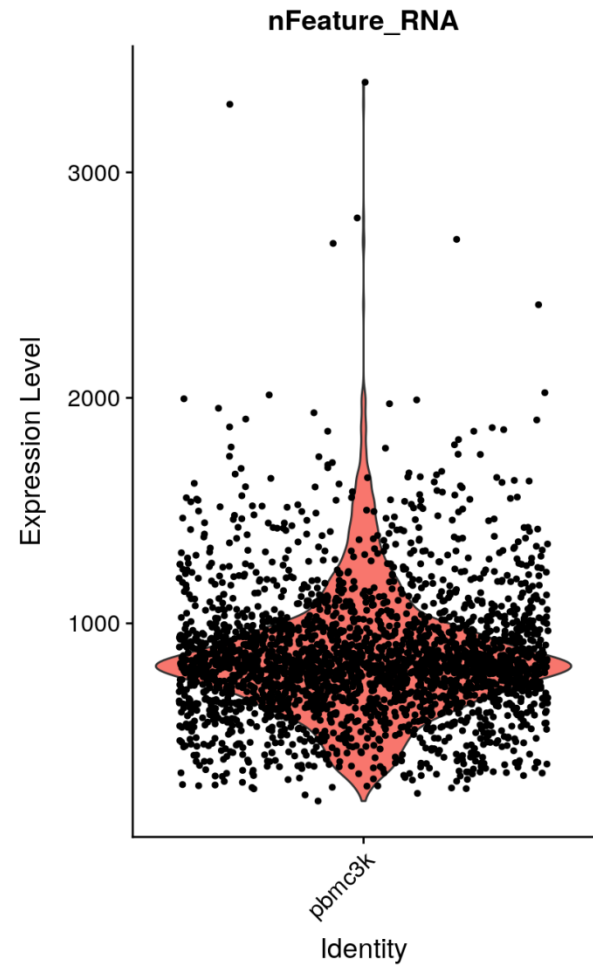
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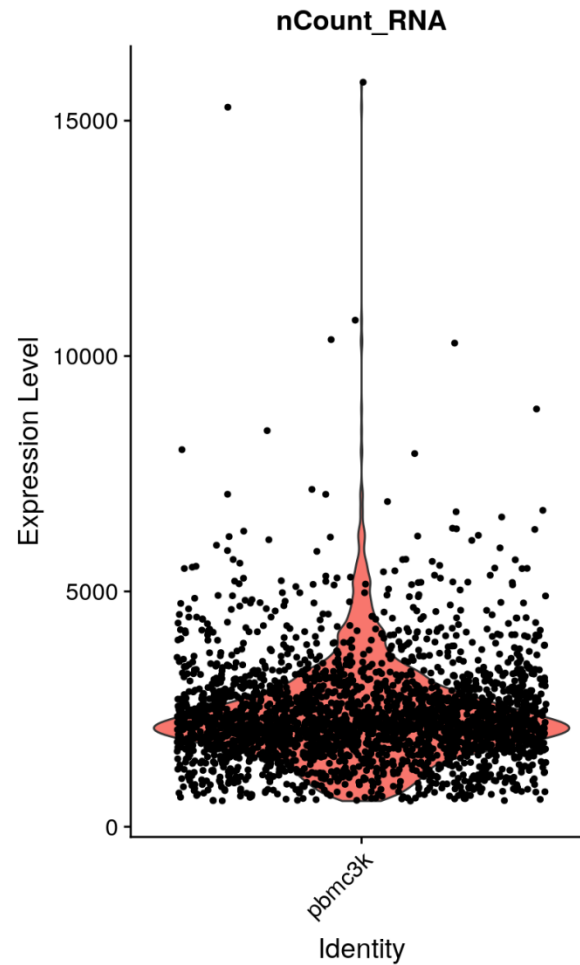
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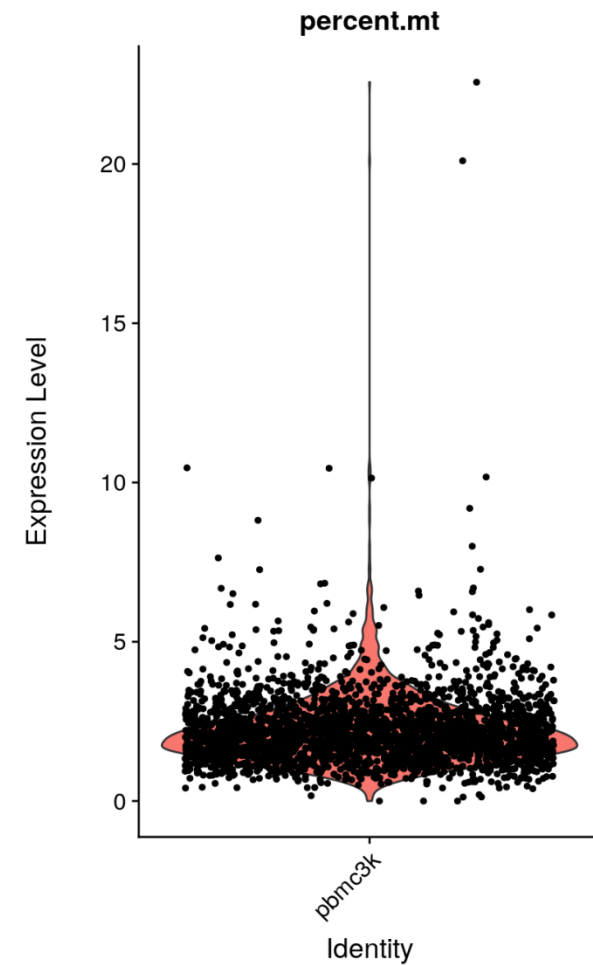
Data pre-filtering: CELLS



Number of
detected genes

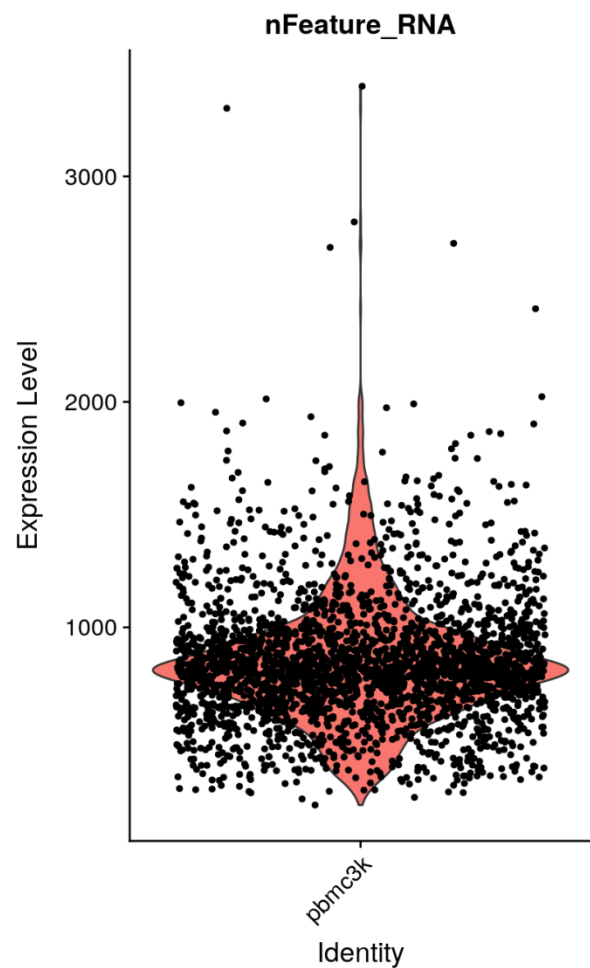


Number of
total reads per cell

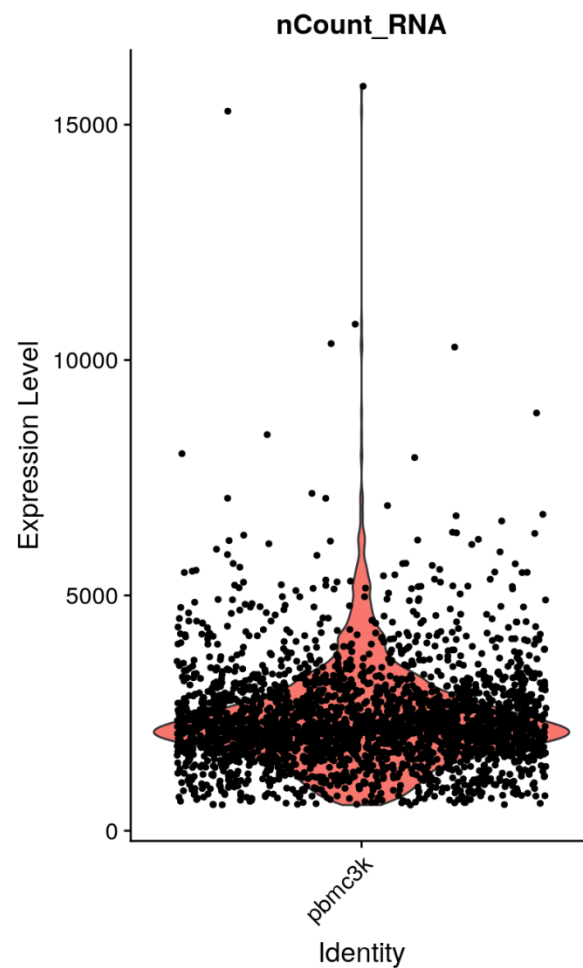


% of reads aligned to
mitochondrial genes

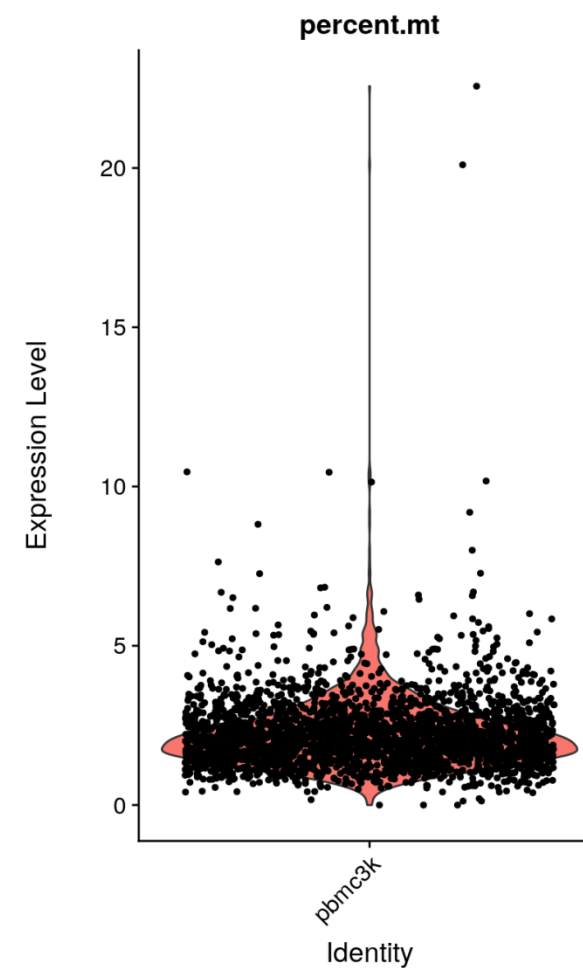
Data pre-filtering: CELLS



Number of detected genes

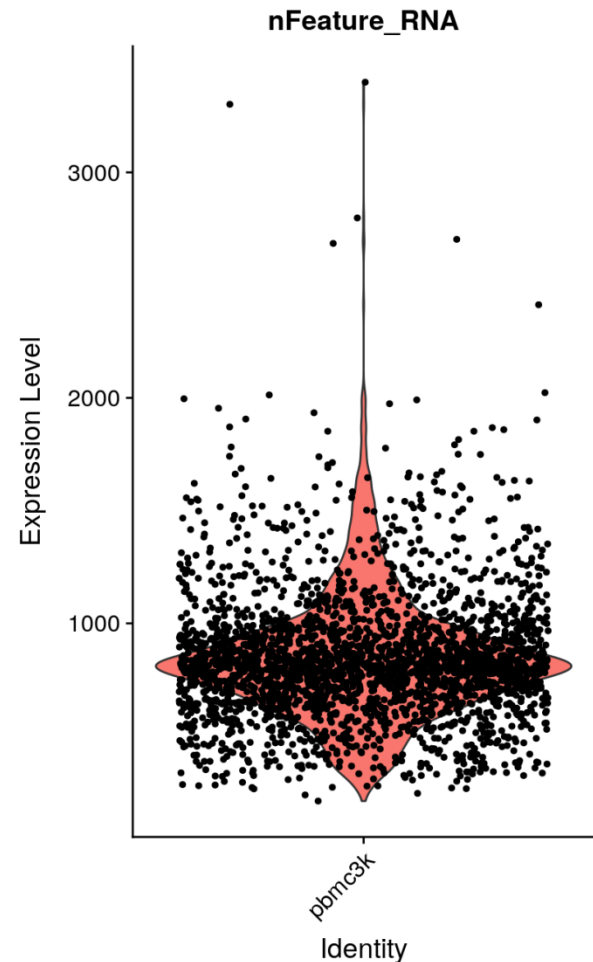


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% of reads aligned to mitochondrial genes

Data pre-filtering: CELLS



Number of
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Suggestion from several tools (Seurat, Scanpy):

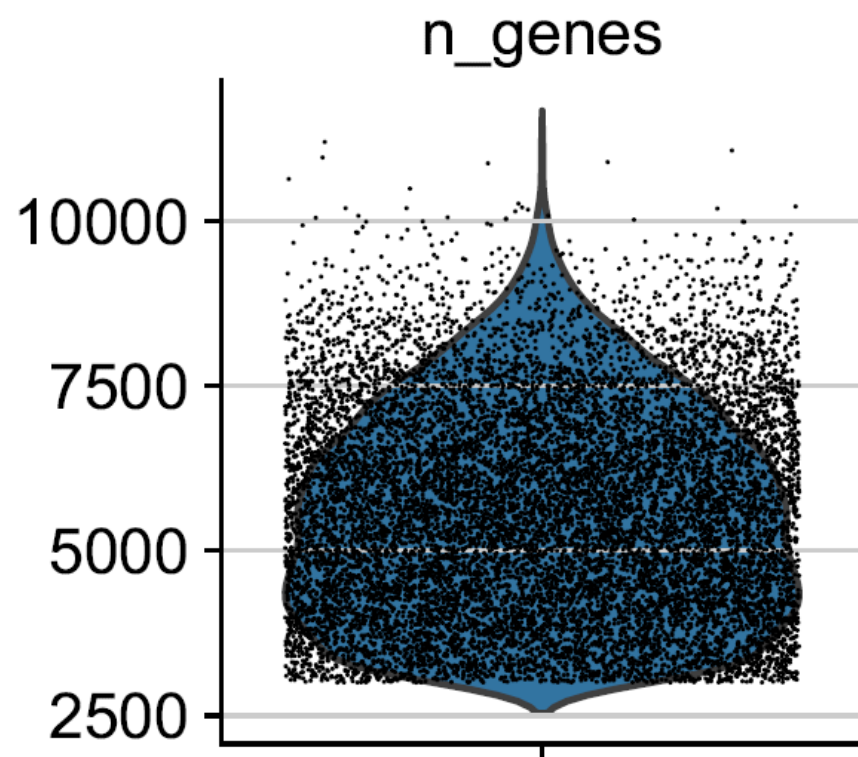
- Lower limit → More than 200

Remove cells poorly informative

- Upper limit → Less than 2,500-3,000

Remove outlier cells/multiplets (?)

- Doublet Detection
<https://github.com/JonathanShor/DoubletDetection/blob/master/docs/DoubletDetection.pdf>
- Mixed gene expression
- Classification of low quality cells from single-cell RNA-seq data (Illicic et al. 2016)



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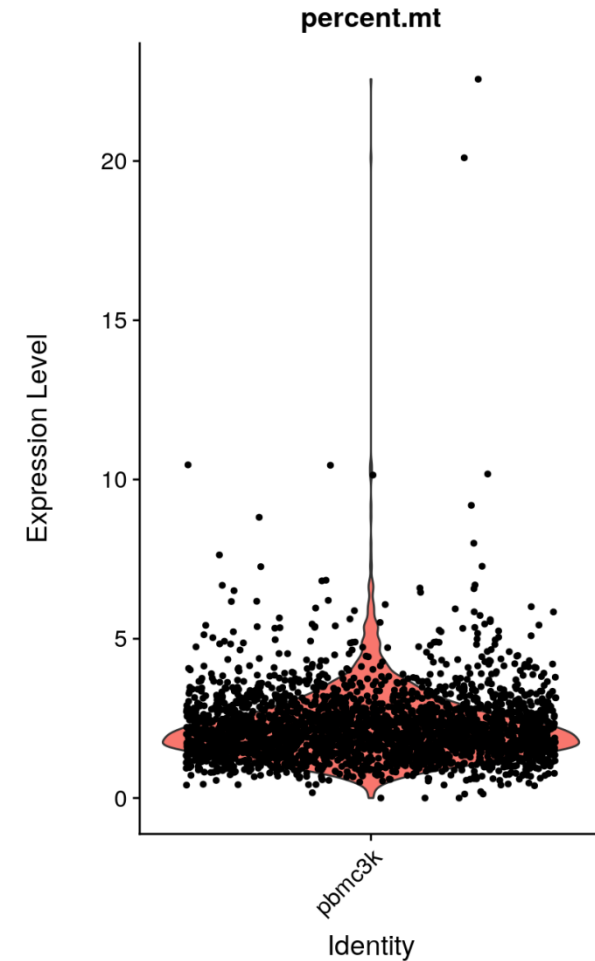
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Data pre-filtering: CELLS

High percentage of mitochondrial gene expression may be due to:

- Apoptotic cells
- Lysed cells



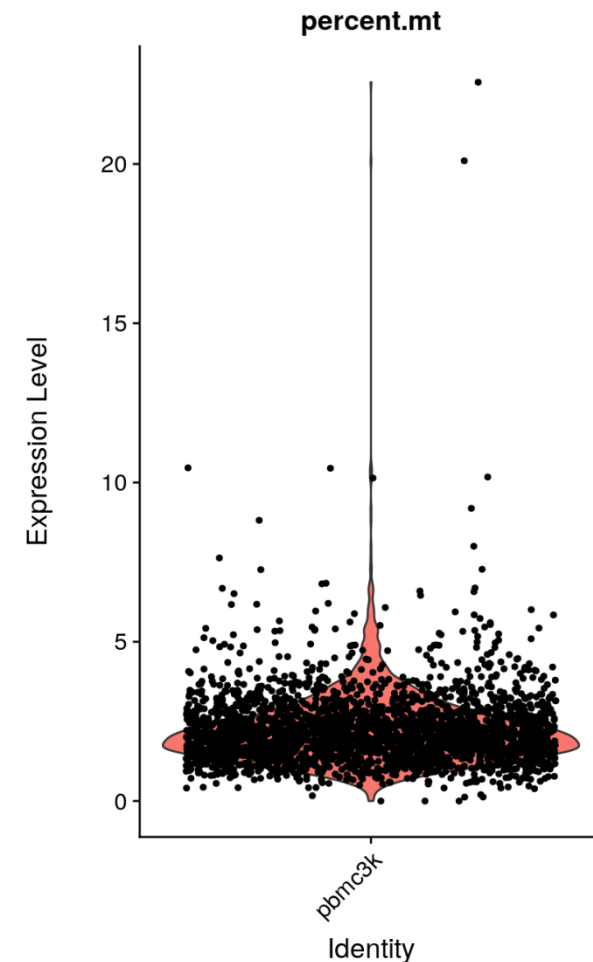
% of reads aligned to mitochondrial genes

Data pre-filtering: CELLS

High percentage of mitochondrial gene expression may be due to:

- Apoptotic cells
- Lysed cells

Cells with more than 5-7% of mtRNAs should be removed



% of reads aligned to mitochondrial genes

Acknowledgements

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