

Central European Institute of Technology BRNO | CZECH REPUBLIC

Modern methods for genome analysis (PřF:Bi7420)

Lecture 7 : Single Cell RNA-seq analysis



## RNA-seq types

Bulk RNA-seq

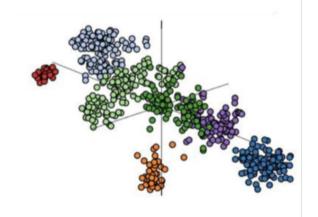
SC-RNA-seq

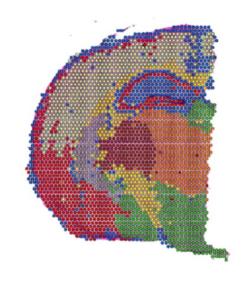
Spatially resolved RNA-seq













#### Bulk RNA-seq vs. SC-RNA-seq

- Bulk RNA-seq
  - Several samples
  - Difference in RNA levels between predetermined set of samples (conditions)

- SC-RNA-seq
  - Hundreds to thousand cells
  - Distinguish (cluster) cells based on the difference in RNA levels



### SC-RNA-seq primary analysis

- In principle similar to the bulk RNA-seq
  - Map to genomic reference, demultiplex and count reads per gene
- For 10x Genomics tool from the company Cell Ranger
- Cell Ranger report example



### SC-RNA-seq primary analysis results

per gene (feature) per cell read count

	AAACATACAACCAC-1	AAACATTGAGCTAC-1	AAACATTGATCAGC-1	AAACCGTGCTTCCG-1	AAACCGTGTATGCG-1	AAACGCACTGGTAC-1	AAACGCTGACCAGT-1	AAACGCTGGTTCTT-1
AL627309.1	0	0	0	0	0	0	0	0
AP006222.2	0	0	0	0	0	0	0	0
RP11-206L10.2	0	0	0	0	0	0	0	0
RP11-206L10.9	0	0	0	0	0	0	0	0
LINC00115	0	0	0	0	0	0	0	0
NOC2L	0	0	0	0	0	0	0	0
KLHL17	0	0	0	0	0	0	0	0
PLEKHN1	0	0	0	0	0	0	0	0
RP11-5407.17	0	0	0	0	0	0	0	0
HES4	0	0	0	0	0	0	0	0
RP11-5407.11	0	0	0	0	0	0	0	0
ISG15	0	0	1	9	0	1	0	0
AGRN	0	0	0	0	0	0	0	0
C1orf159	0	0	0	0	0	0	0	0
TNFRSF18	0	2	0	0	0	0	0	0
TNFRSF4	0	0	0	0	0	0	0	0
SDF4	0	0	1	0	0	0	0	0
B3GALT6	0	0	0	0	0	0	1	0
FAM132A	0	0	0	0	0	0	0	0
UBE2J2	0	0	0	0	0	0	0	0



### Pre-processing workflow

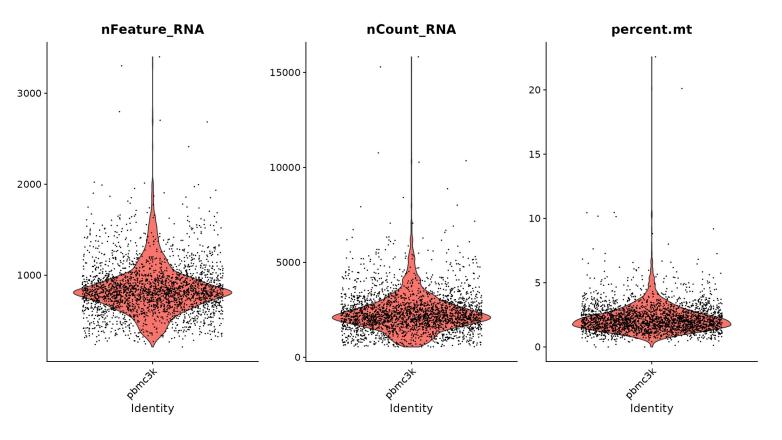
- Filtering of cells based on QC metrics
- Detection and filtering of highly variable features
- Data normalization and scaling

	AAACATACAACCAC-1	AAACATTGAGCTAC-1	AAACATTGATCAGC-1	AAACCGTGCTTCCG-1	AAACCGTGTATGCG-1	AAACGCACTGGTAC-1	AAACGCTGACCAGT-1	AAACGCTGGTTCTT-1
AL627309.1	0	0	0	0	0	0	0	0
AP006222.2	0	0	0	0	0	0	0	0
RP11-206L10.2	0	0	0	0	0	0	0	0
RP11-206L10.9	0	0	0	0	0	0	0	0
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RP11-5407.17	0	0	0	0	0	0	0	0
HES4	0	0	0	0	0	0	0	0
RP11-5407.11	0	0	0	0	0	0	0	0
ISG15	0	0	1	9	0	1	0	0
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SDF4	0	0	1	0	0	0	0	0
B3GALT6	0	0	0	0	0	0	1	0
FAM132A	0	0	0	0	0	0	0	0
UBE2J2	0	0	0	0	0	0	0	0



## Filtering of cells QC

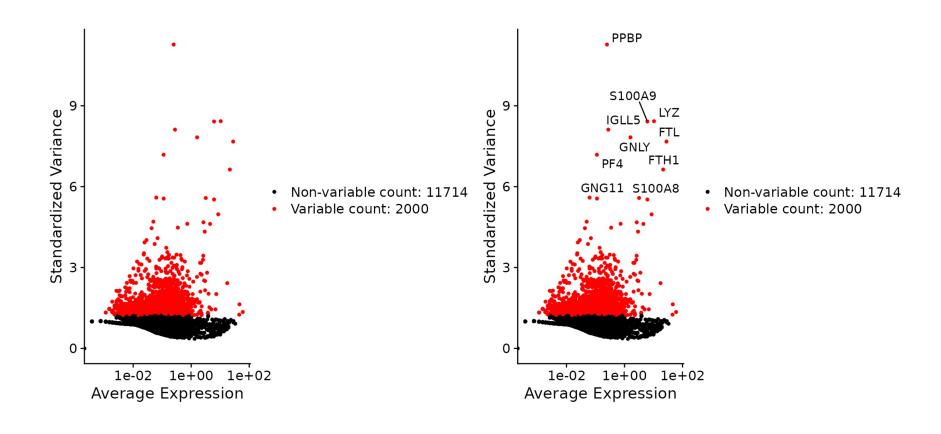
- Number of genes
- Number of reads
- % of mitochondrial RNA
  - Sign of dying cells



- Example filtering:
  - filter cells that have unique feature counts over 2,500 or less than 200
  - filter cells that have >5% mitochondrial counts



## Identification of highly variable features



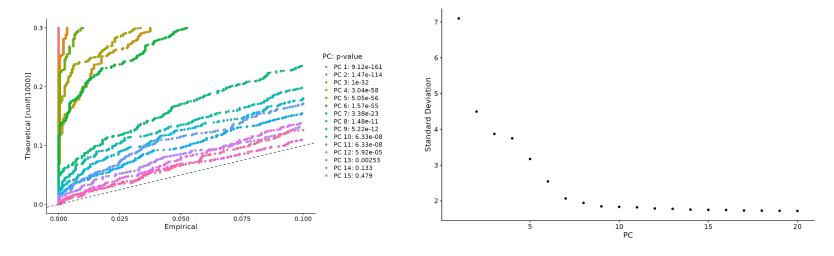


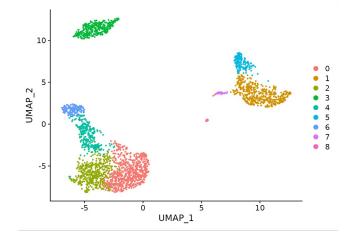
## Determine the 'dimensionality' of the dataset

How many clusters (cell types) is in the experiment (we want

to explore)

Analysis of PCA results



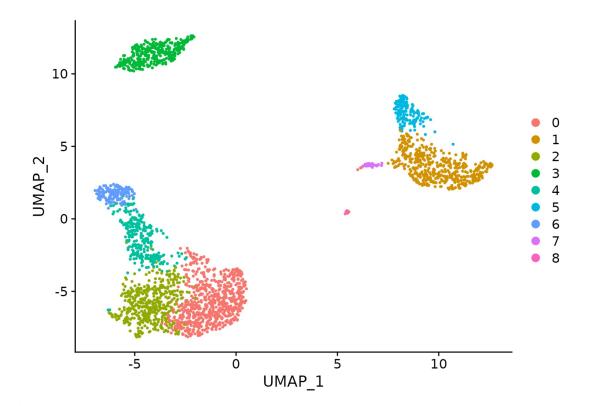


Better to choose higher value then smaller



#### Cluster the cells

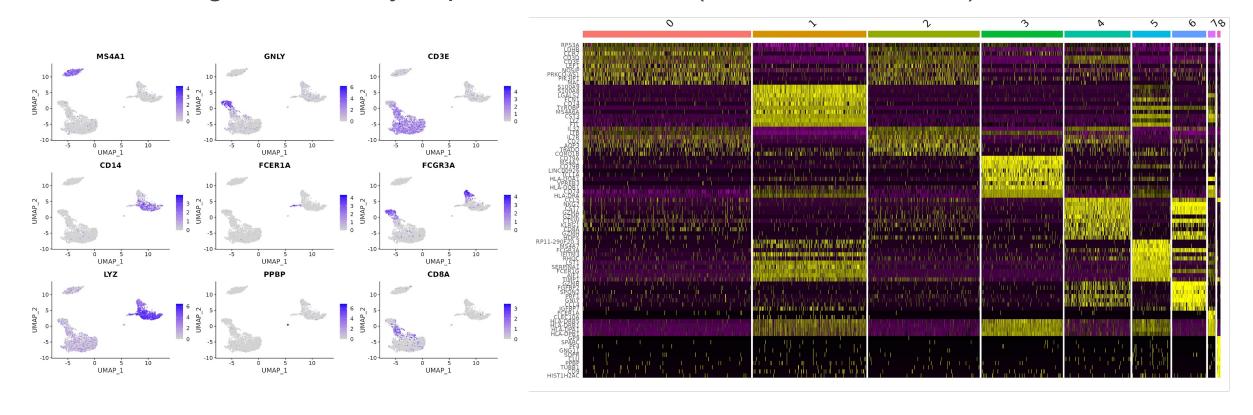
- graph-based algorithms
- UMAP/tSNE visualization





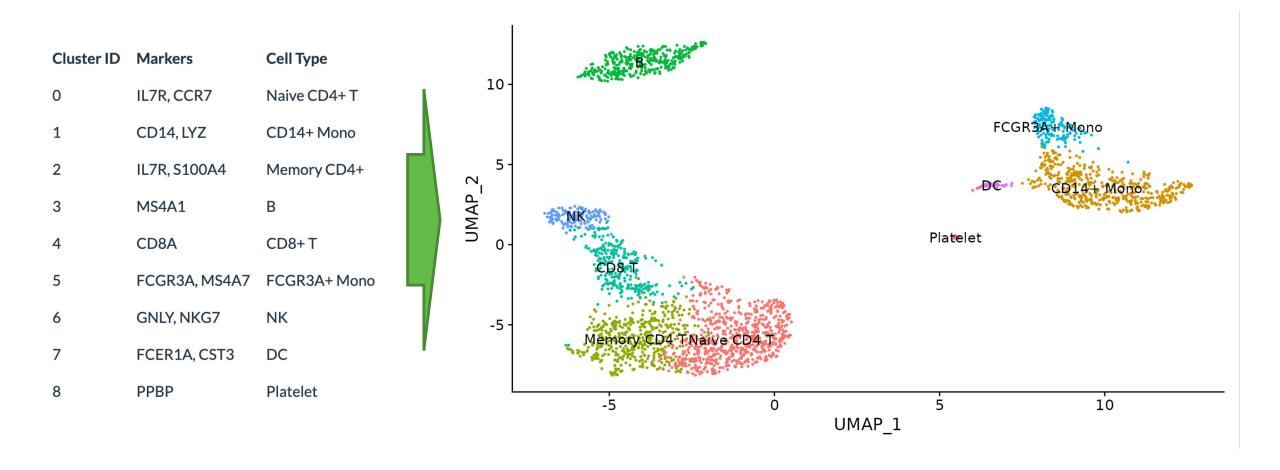
# **Analysing clusters**

Finding differentially expressed features (cluster biomarkers)





# Assigning cell type identity to clusters



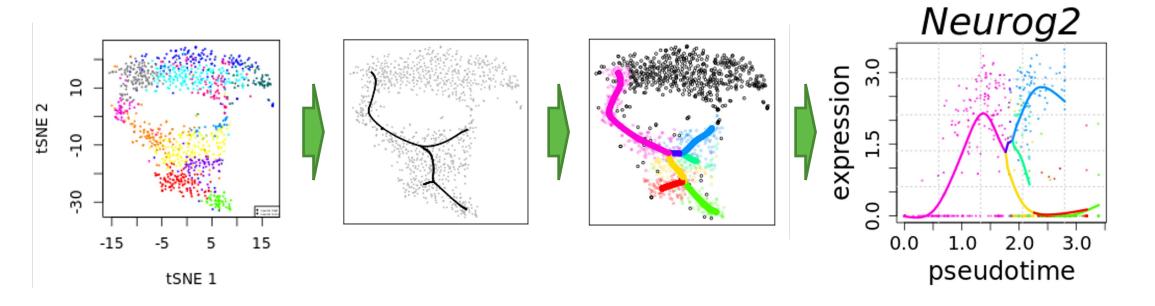


## Advanced analysis

- Cell state/development directions
  - Pseudo-time
  - Velocity

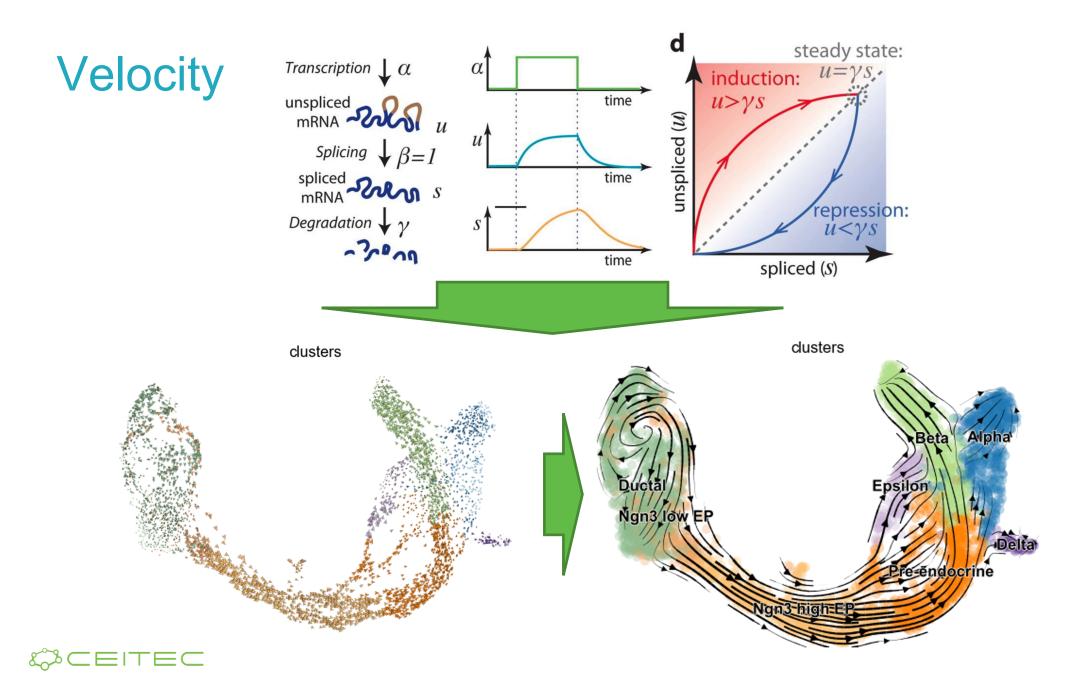


#### Pseudotime



does not provide information about directionality of dynamics





## RNA-seq types

Bulk RNA-seq

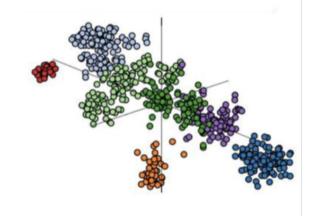
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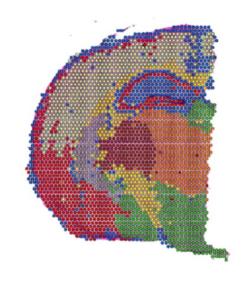
Spatially resolved RNA-seq





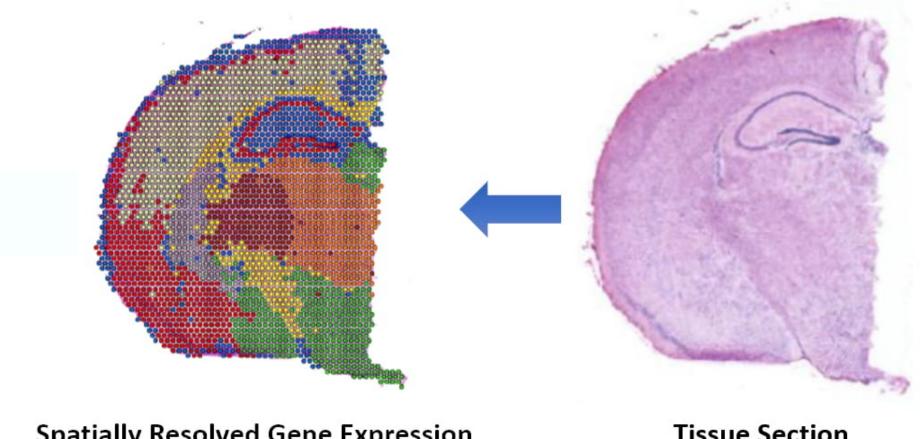








# **Spatial Transcriptomics**

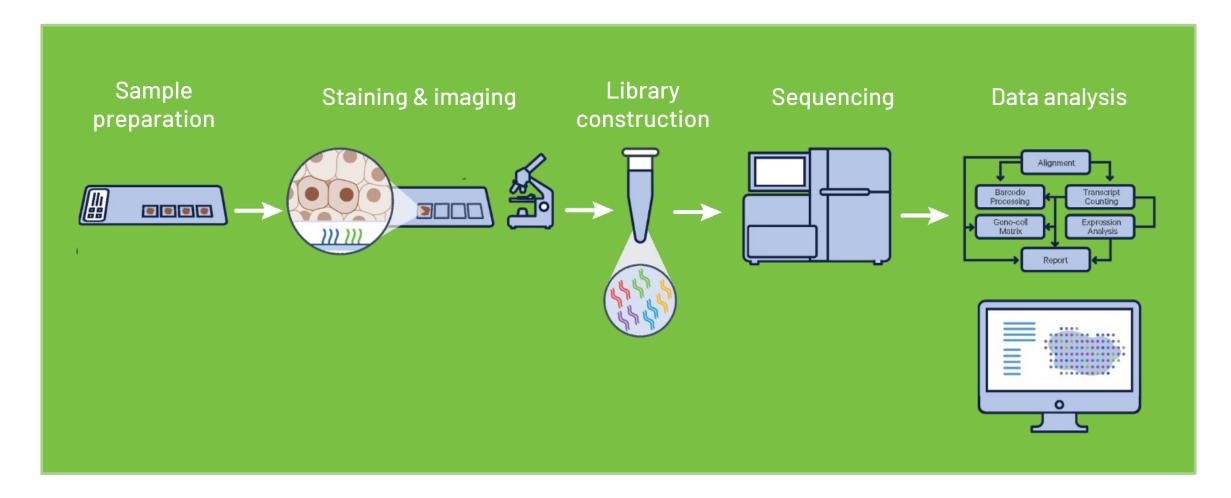




**Tissue Section** 

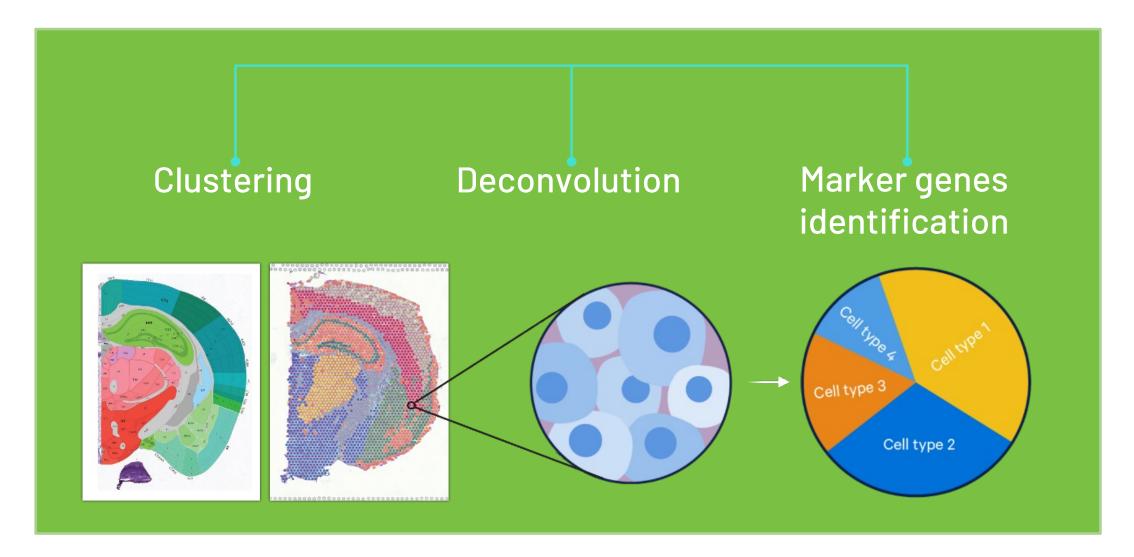


## Visium sequencing library





## Visium spatial data analyses





## Spatial Transcriptomics – Spatial Path



Vojtech Macala http://cfb.ceitec.muni.cz/spatialpath/



