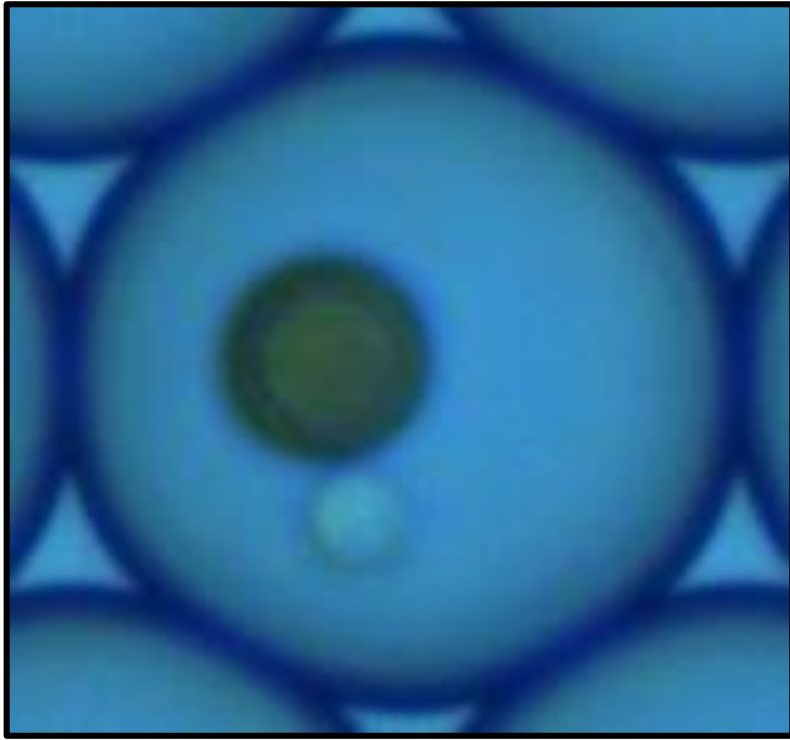
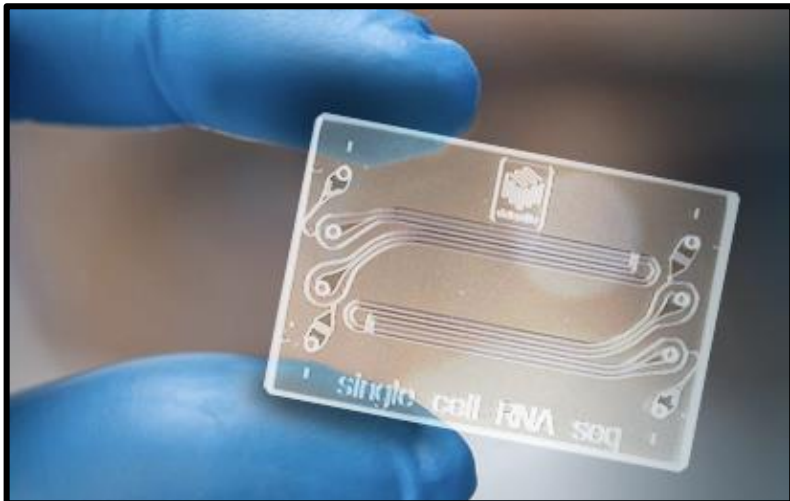




Microfluidics in biomedicine (single cell trapping)

3.3.2021

(Microfluidics and BioMEMS course)



Päivi Saavalainen, PhD, docent

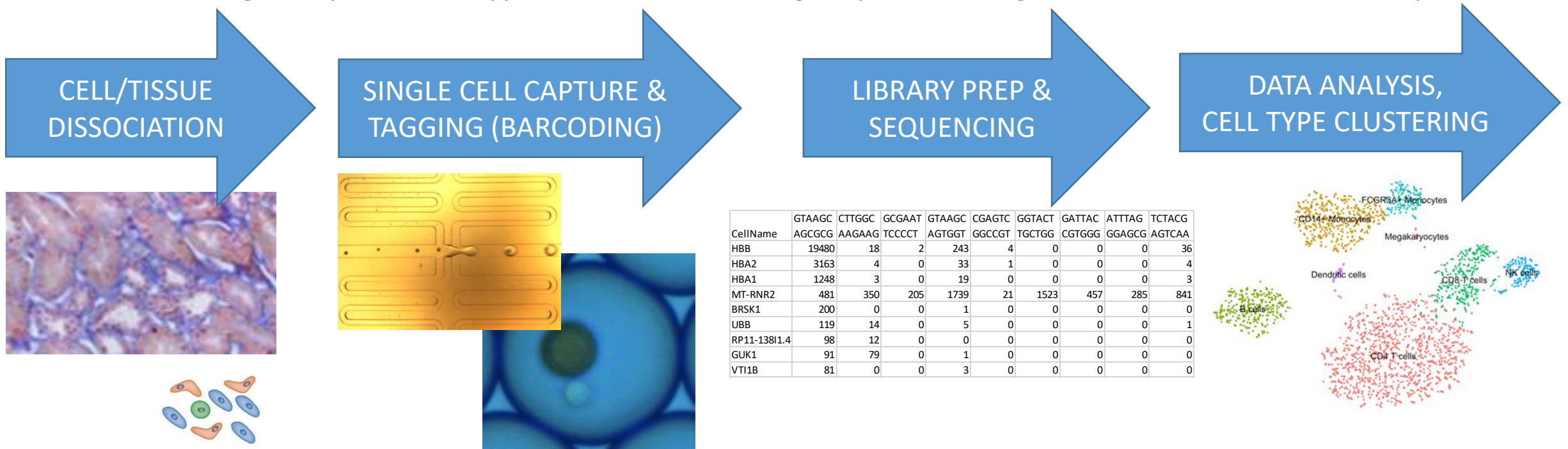
PI, Immunomics group, University of Helsinki

Single cell transcriptomics (scRNAseq)

Cells are not identical – Instead of **average** gene expression of millions of cells you now can measure it from **1000's of single cells individually**

With single cell resolution:

-> Cell heterogeneity, rare cell types, cancer cells, drug responses, diagnostics, individualized therapies etc



Traditional way to study and isolate single cells: Flow cytometry (FACS=fluorescence assisted cell sorting)

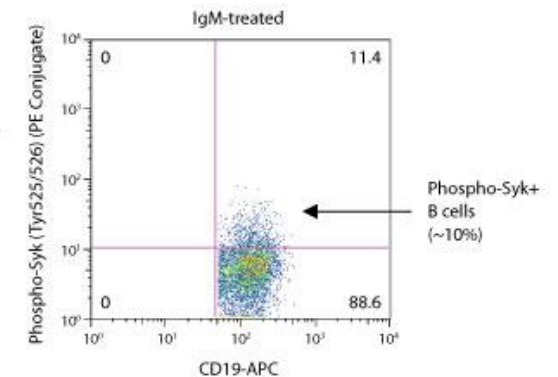
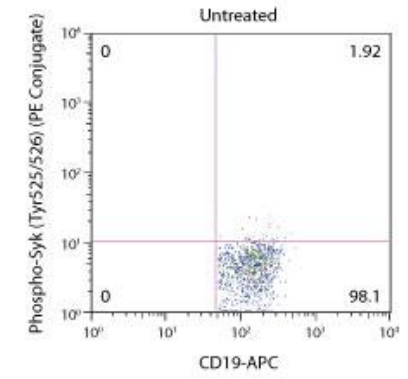
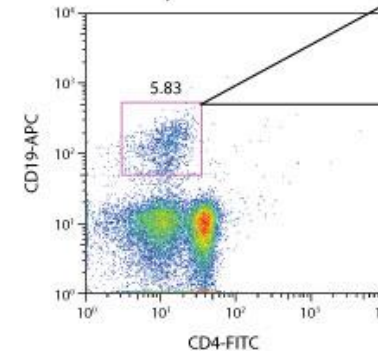
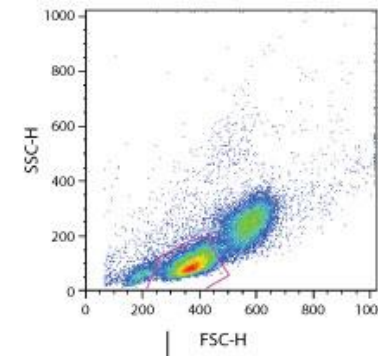
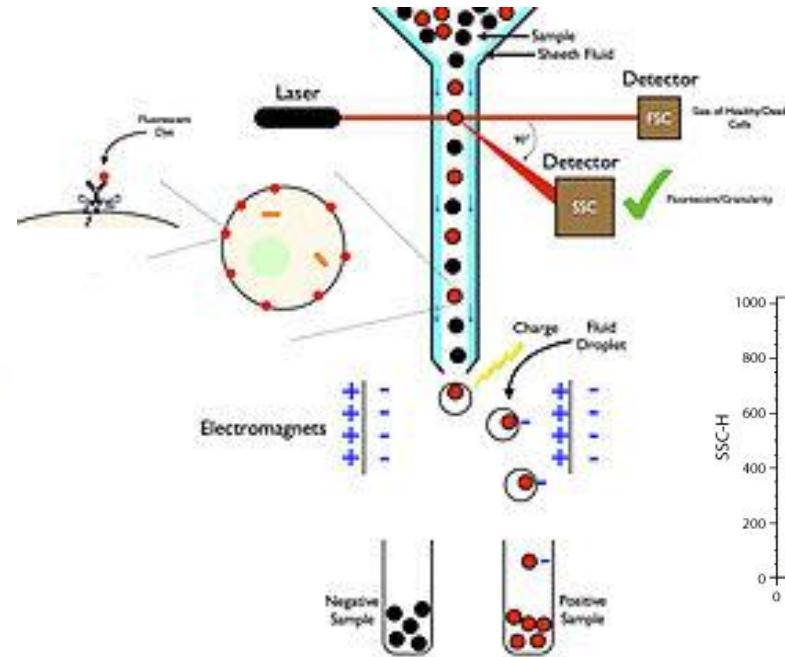


Table 1 Notable advancements in single-cell techniques

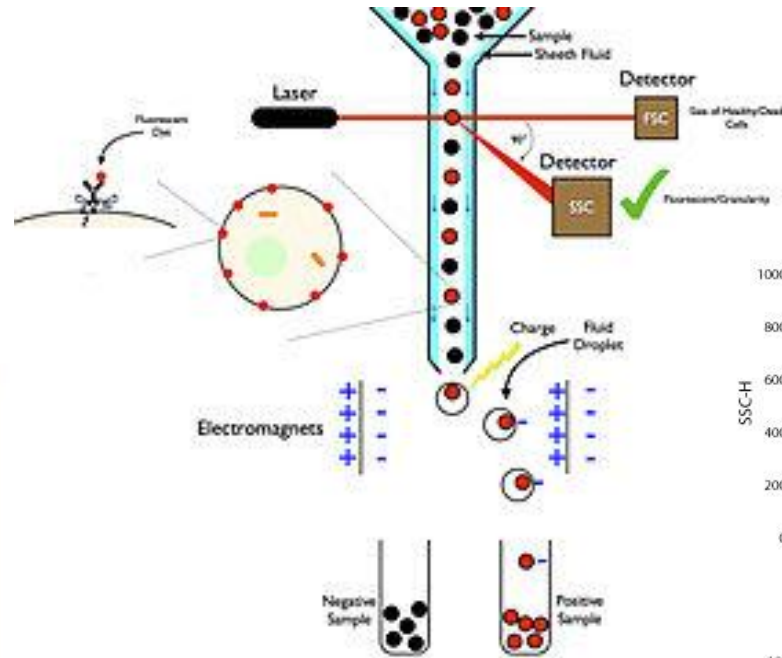
Year introduced	Notable technology advancements	Method cell range ^a
2009	Tang et al. [10]	1 ^b
2011	STRT-seq [23]	< 100
2012	SMART-seq [24]	< 100
2012	CEL-Seq [25]	< 100
2013	Fluidigm C1 (IFC) [26]	< 800
2013	Smart-seq 2 [27]	< 1000
2014	MARS-seq [28]	10,000 s
2015	Drop-seq [29]	10,000 s
2015	inDrop [30]	10,000 s
2016	Chromium (10× Genomics) [31]	10,000 s
2017	ddSeq (Bio-Rad) [32]	10,000 s
2017	SPLiT-seq [33]	10,000 s
2017	Seq-well [34]	10,000 s

This is a non-comprehensive list of peer-reviewed studies that advanced single-cell isolation and preparation techniques

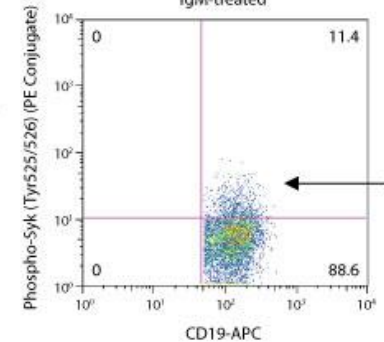
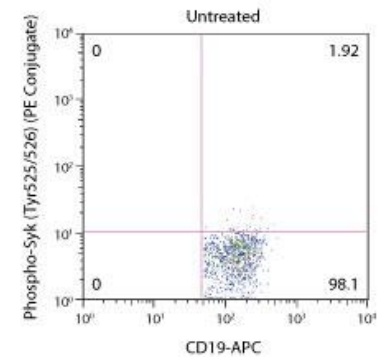
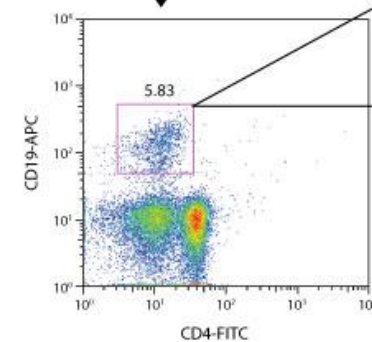
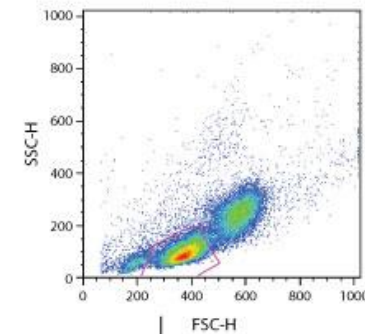
^a The "range" lists the largest relative population that can or has been studied using this technique



Traditional way to study and isolate single cells: Flow cytometry (FACS=fluorescence assisted cell sorting)

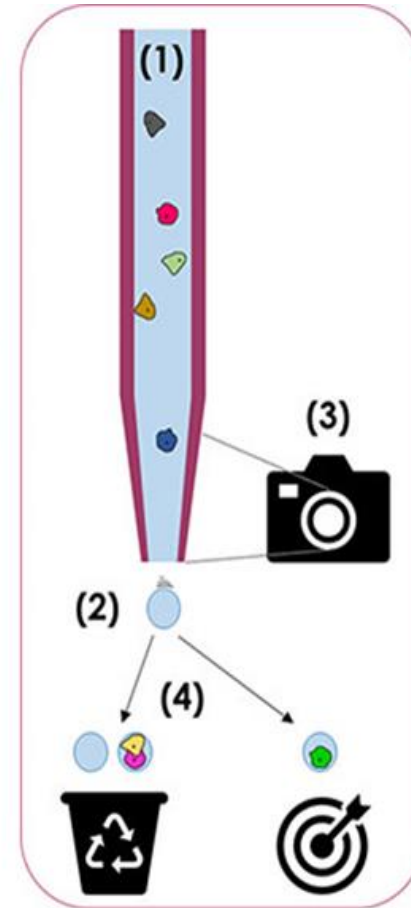
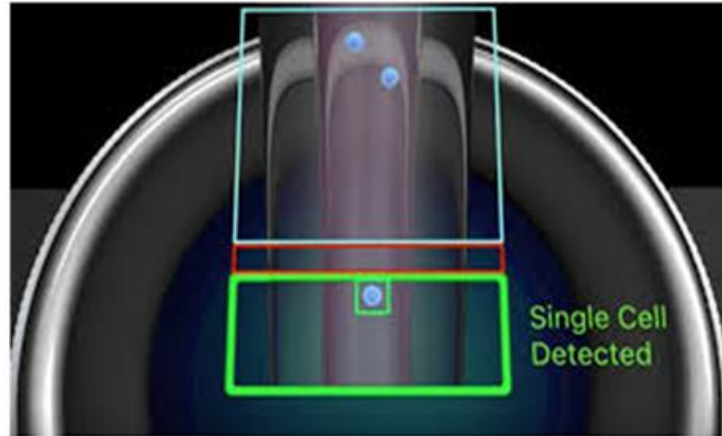
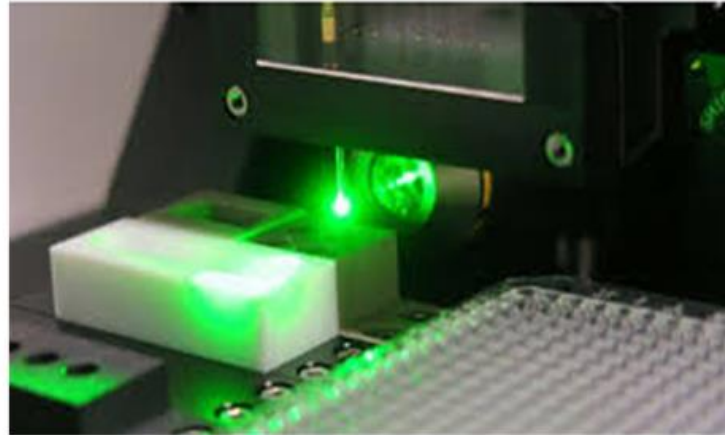


Sorting to tubes or 1 cell per
Plate well

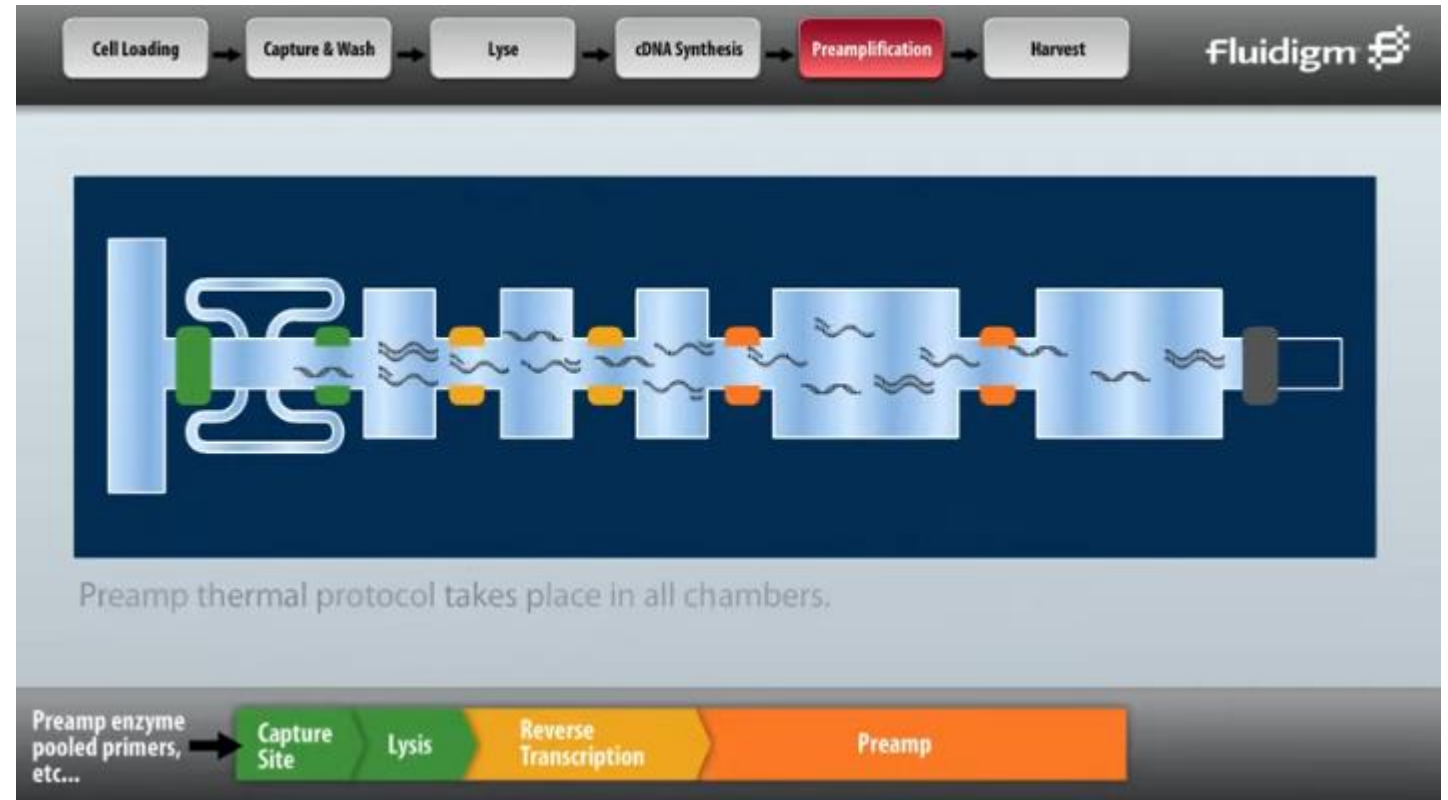


Phospho-Syk+
B cells
(~10%)

CellenONE single cell dispenser (at FIMM)



Non-Droplet systems: Fluidigm C1 (not very popular anymore)

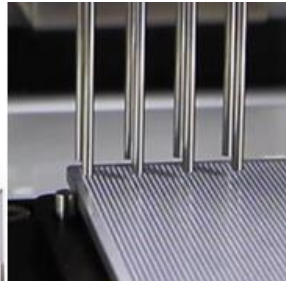


Non-Droplet systems: Wafergen ICELL8 nanodispenser



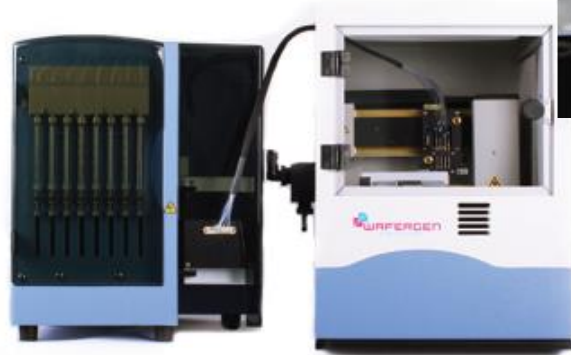
SmartChip:
72x72=5184 wells

ICELL8 Chips and Reagents

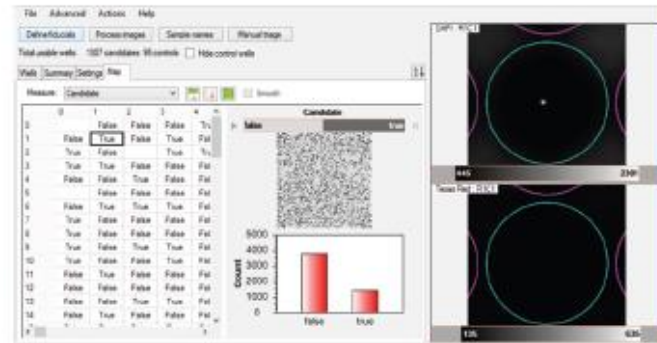


Imaging Station

Acquired by Takara



MultiSample NanoDispenser



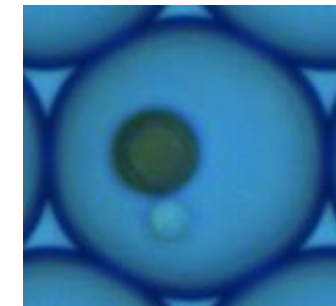
CellSelect Software

Table 1 Notable advancements in single-cell techniques

Year introduced	Notable technology advancements	Method cell range ^a
2009	Tang et al. [10]	1 ^b
2011	STRT-seq [23]	< 100
2012	SMART-seq [24]	< 100
2012	CEL-Seq [25]	< 100
2013	Fluidigm C1 (IFC) [26]	< 800
2013	Smart-seq 2 [27]	< 1000
2014	MARS-seq [28]	10,000 s
2015	Drop-seq [29]	10,000 s
2015	inDrop [30]	10,000 s
2016	Chromium (10× Genomics) [31]	10,000 s
2017	ddSeq (Bio-Rad) [32]	10,000 s
2017	SPLiT-seq [33]	10,000 s
2017	Seq-well [34]	10,000 s

This is a non-comprehensive list of peer-reviewed studies that advanced single-cell isolation and preparation techniques

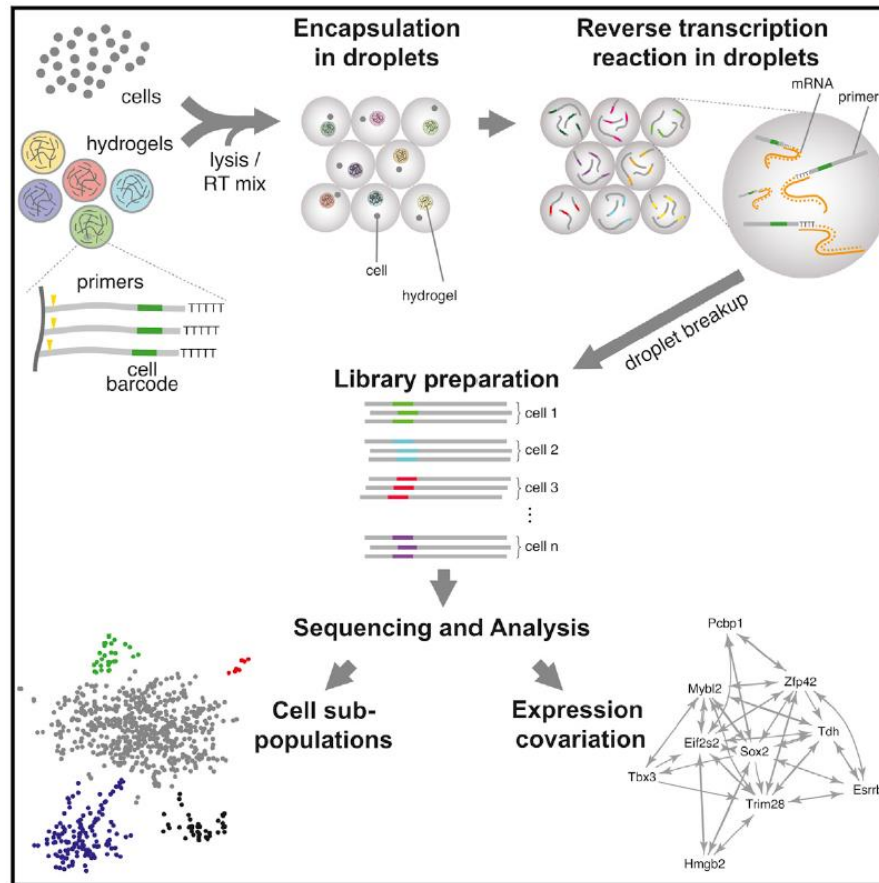
^a The "range" lists the largest relative population that can or has been studied using this technique



New level of throughput: 2 droplet papers 2015 in Cell

Droplet Barcoding for Single-Cell Transcriptomics Applied to Embryonic Stem Cells

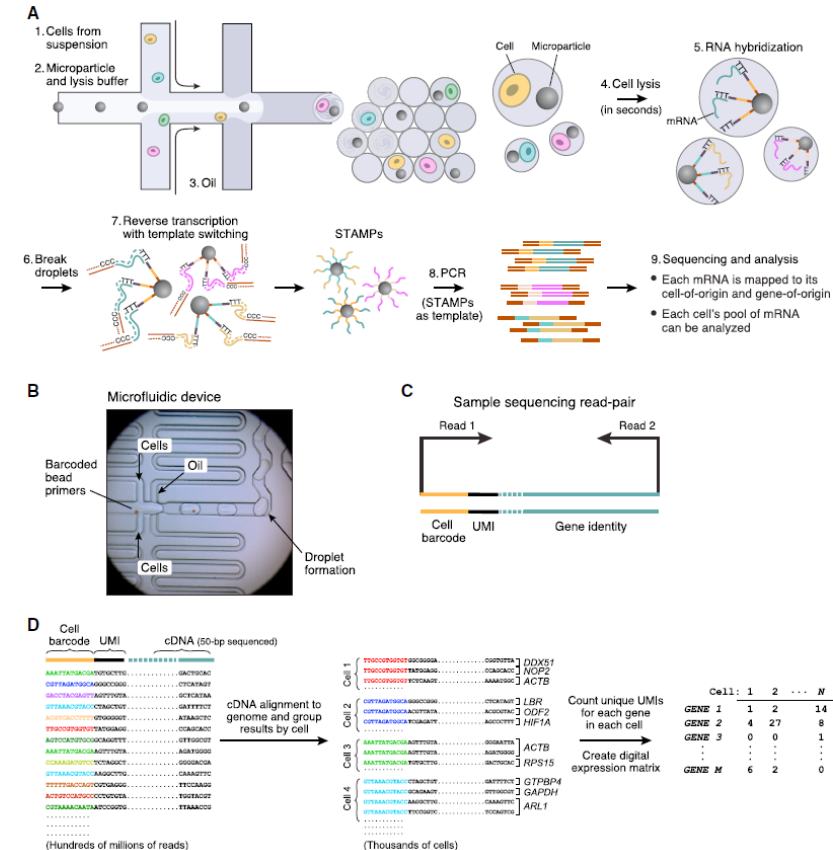
Allon M. Klein,^{1,6} Linas Mazutis,^{2,3,6} Ilke Akartuna,^{2,6} Naren Tallapragada,¹ Adrian Veres,^{1,4,5} Victor Li,¹ Leonid Peshkin,¹ David A. Weitz,^{2,*} and Marc W. Kirschner^{1,*}
¹Department of Systems Biology, Harvard Medical School, Boston, MA 02115, USA



inDrop (indexing droplets)

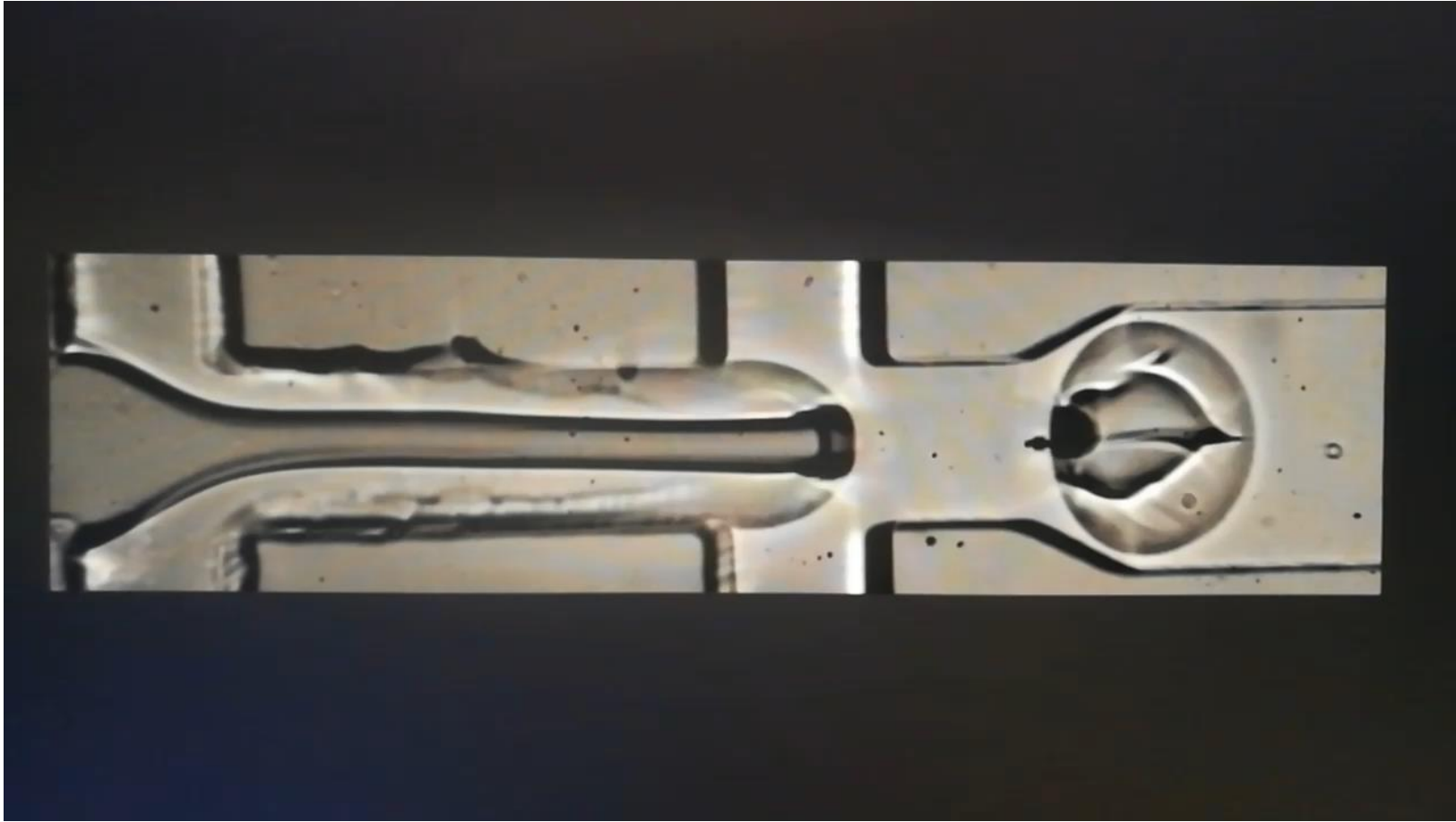
Highly Parallel Genome-wide Expression Profiling of Individual Cells Using Nanoliter Droplets

Evan Z. Macosko,^{1,2,3,*} Anindita Basu,^{4,5} Rahul Satija,^{4,6,7} James Nemesh,^{1,2,3} Karthik Shekhar,⁴ Melissa Goldman,^{1,2} Itay Tirosh,⁴ Allison R. Bialas,⁸ Nolan Kamitaki,^{1,2,3} Emily M. Martersteck,⁹ John J. Trombetta,⁴ David A. Weitz,^{5,10} Joshua R. Sanes,⁹ Alex K. Shalek,^{4,11,12} Aviv Regev,^{4,13,14} and Steven A. McCarroll^{1,2,3,*}
¹Department of Genetics, Harvard Medical School, Boston, MA 02115, USA



Drop-Seq: 10,000 cells, 12h, 6.5cents/cell
 44,808 cells -> 39 cell populations in the retina

Droplet cell / bead capture video (DropSeq)



- <http://mccarrolllab.org/wp-content/uploads/2015/05/Substack-4455-4504.gif>

Commercial droplet based devices

10x Genomics Chromium
120.000/50.000eur device



Bio-Rad Illumina ddSEQ
~50.000eur

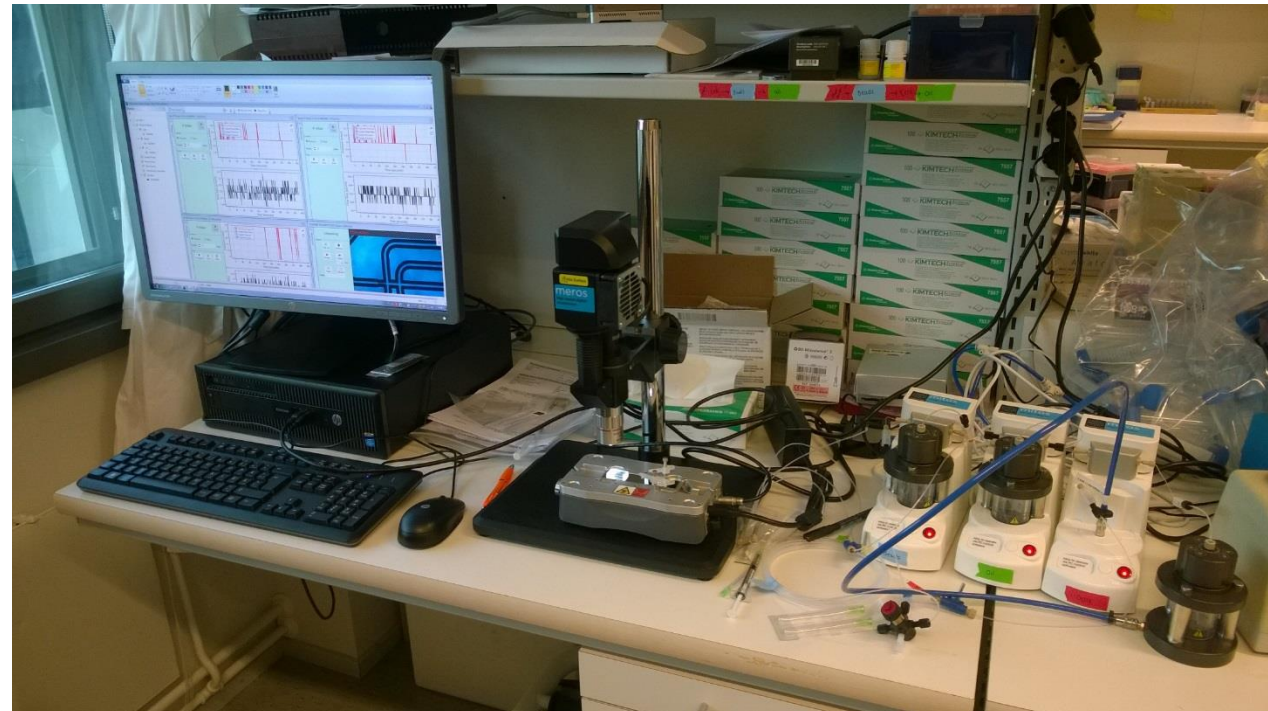
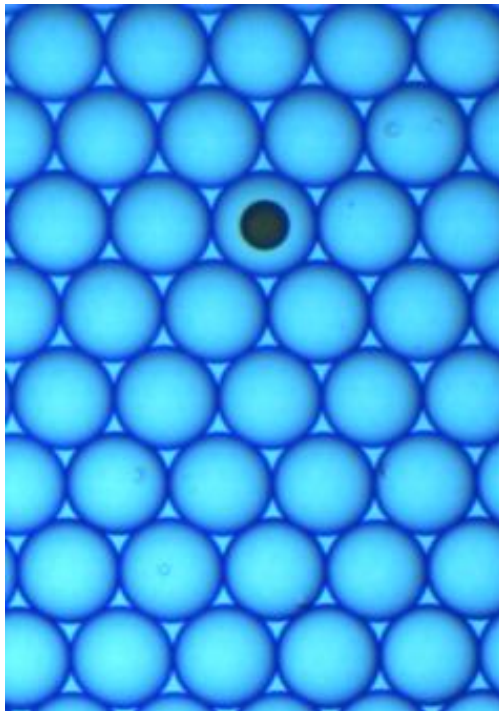
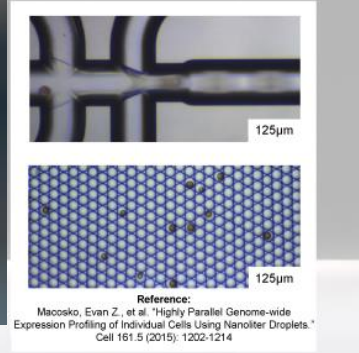
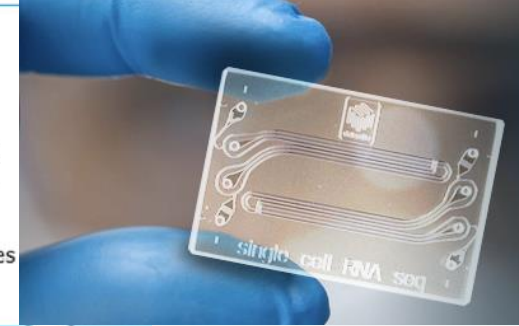
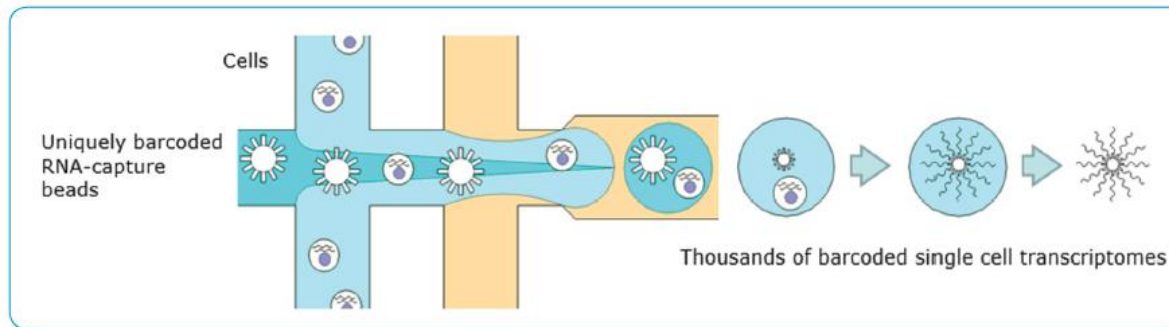


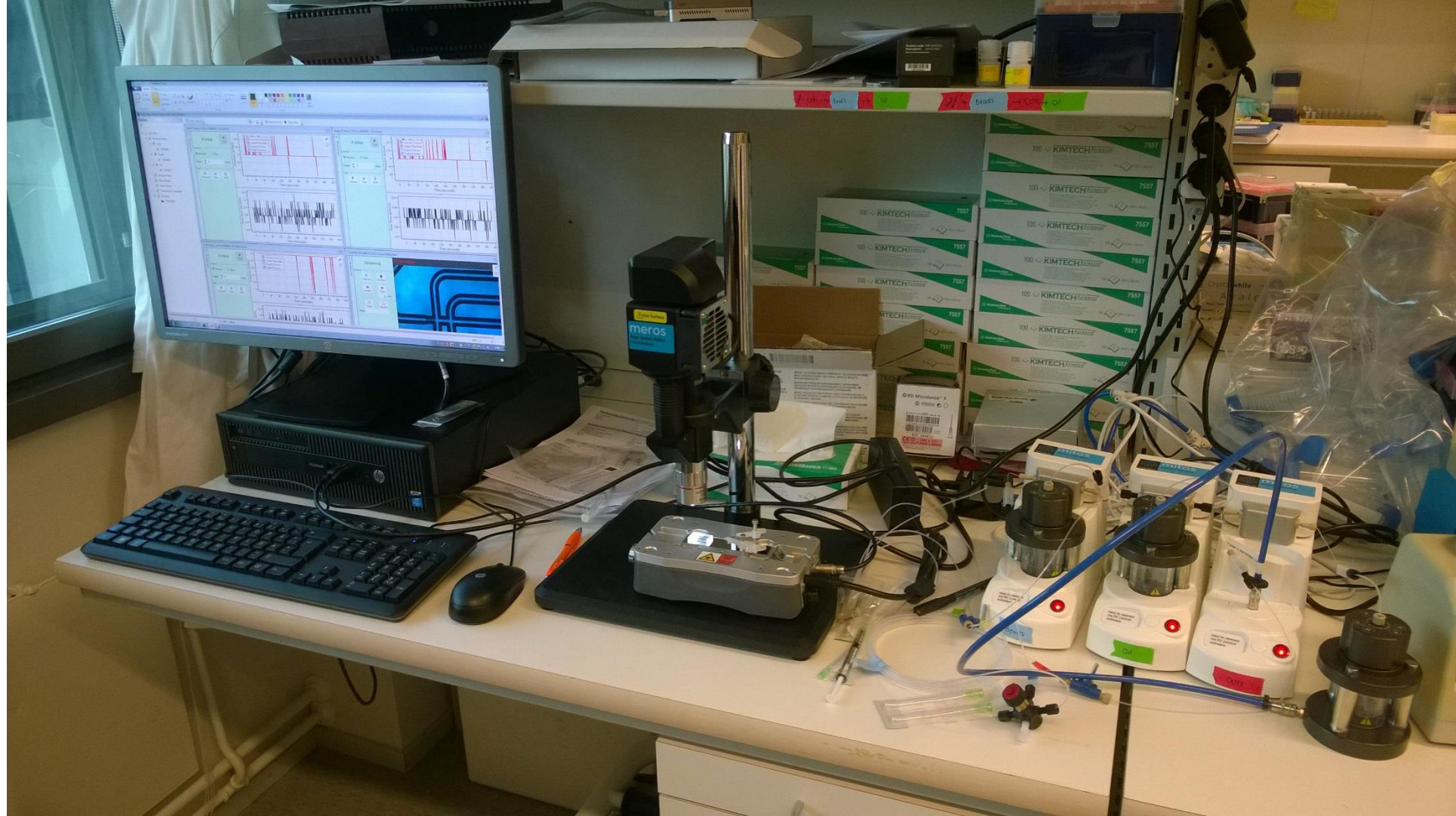
Dolomite Nadia (NEW)
~50.000eur



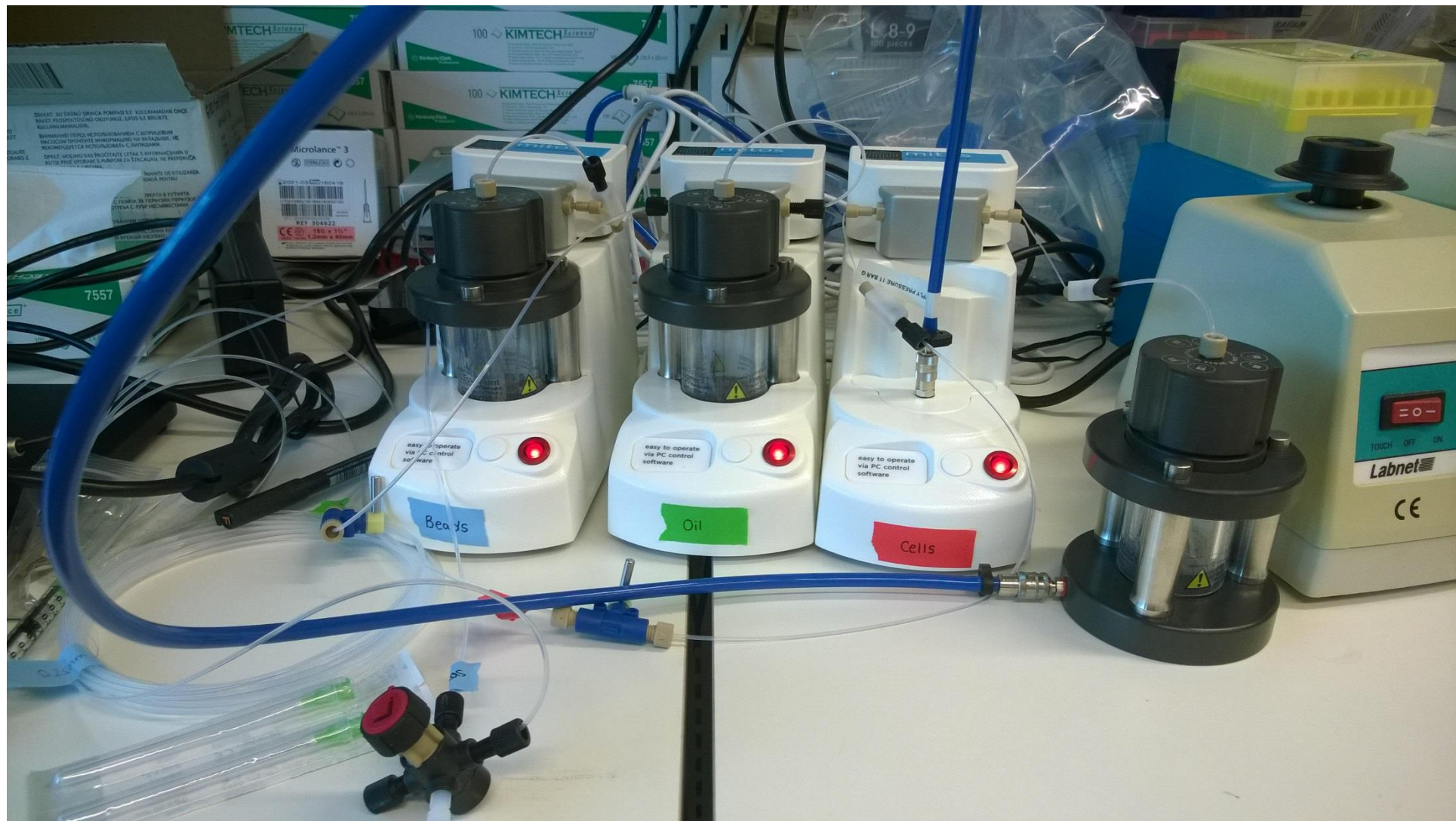


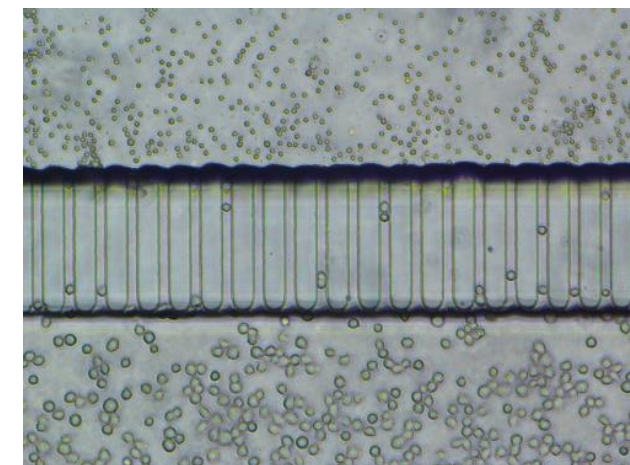
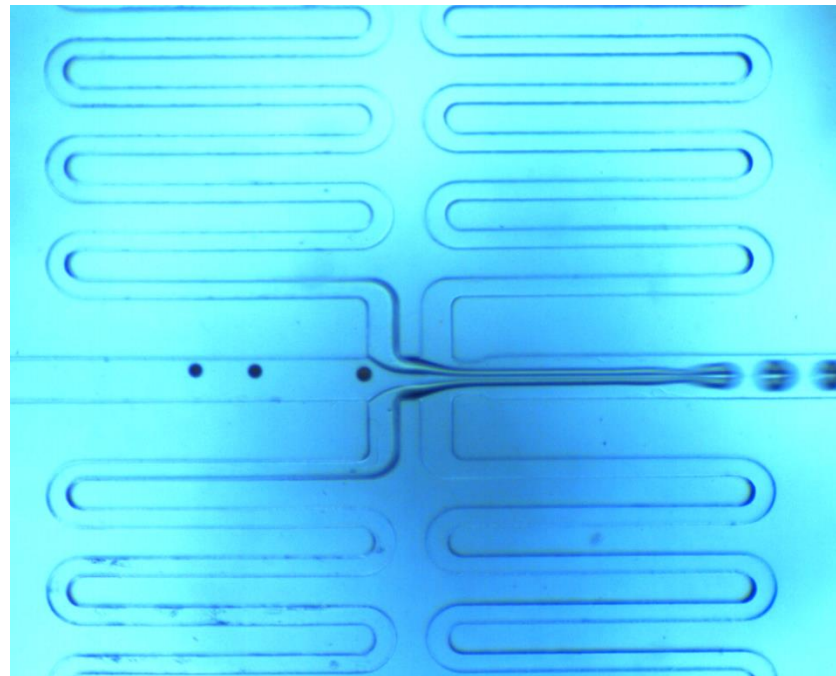
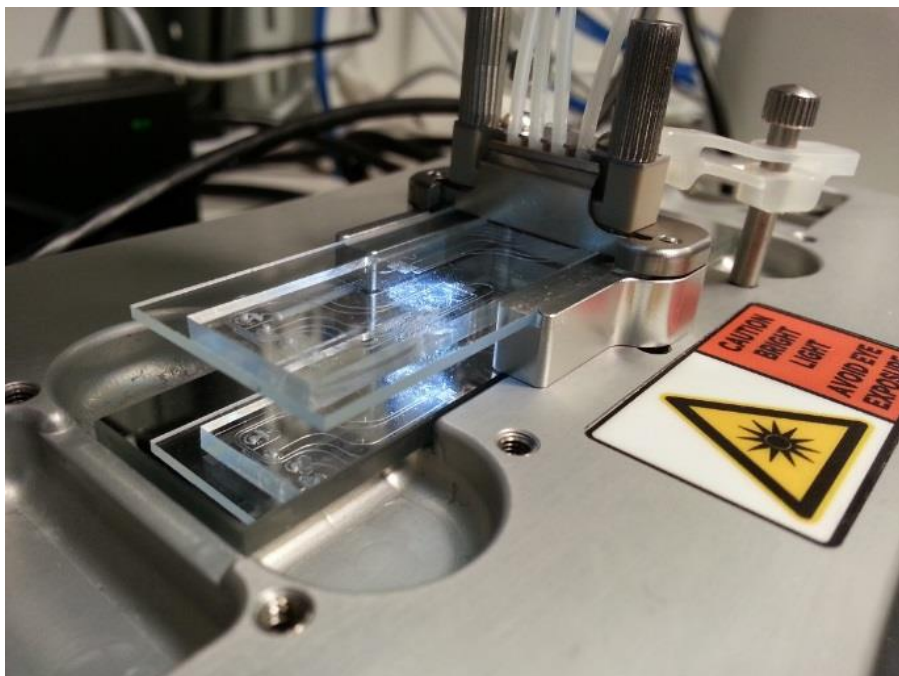
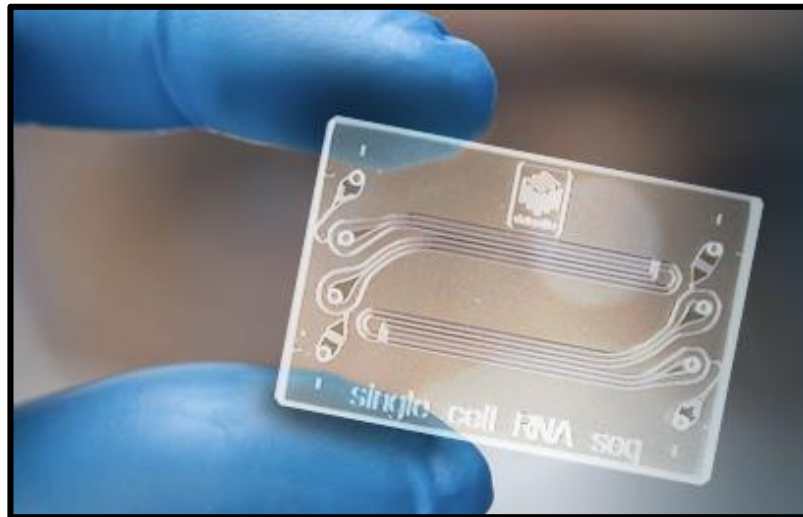
In our lab: the Harvard Drop-seq protocol & Dolomite Bio Single Cell RNAseq device





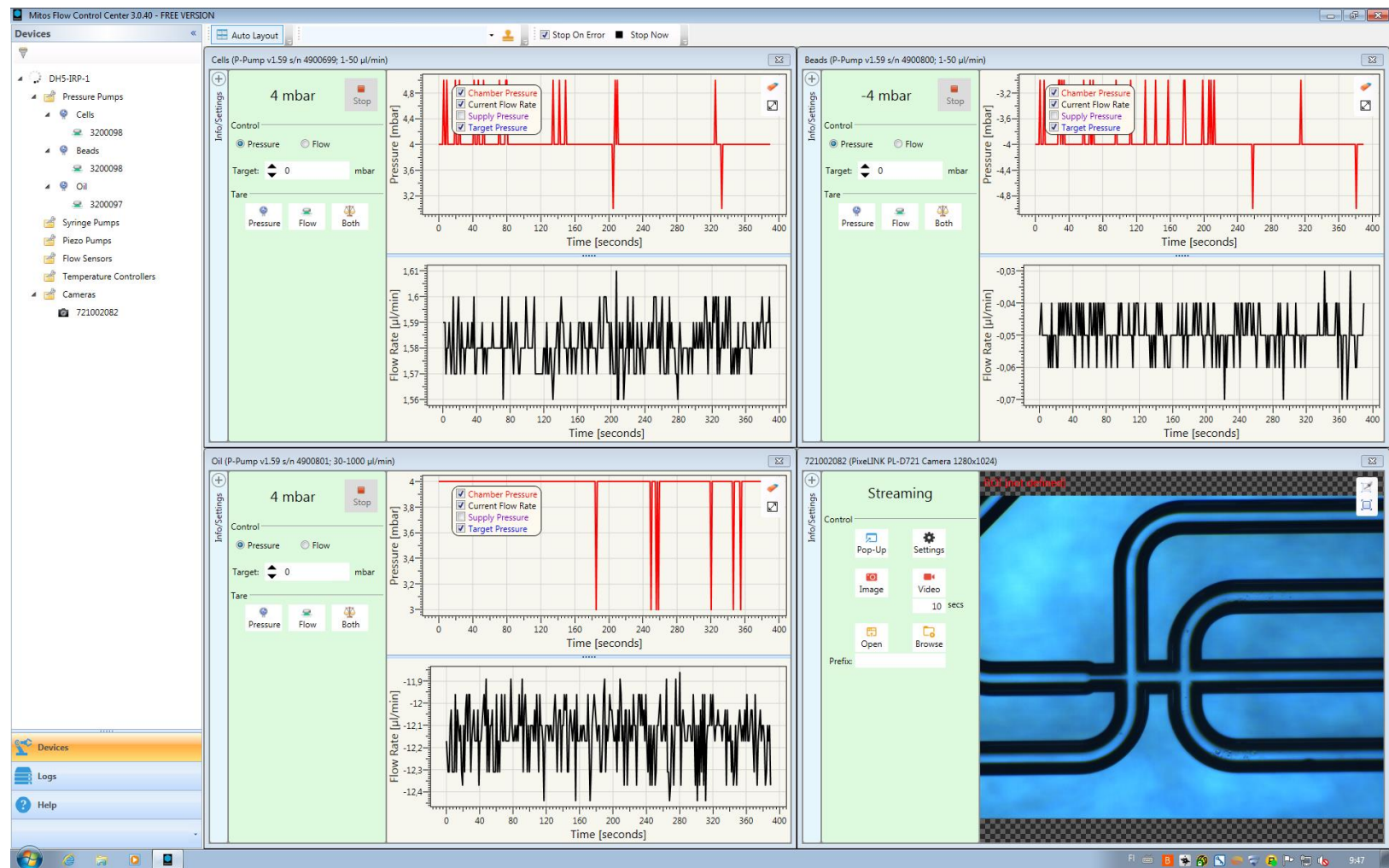


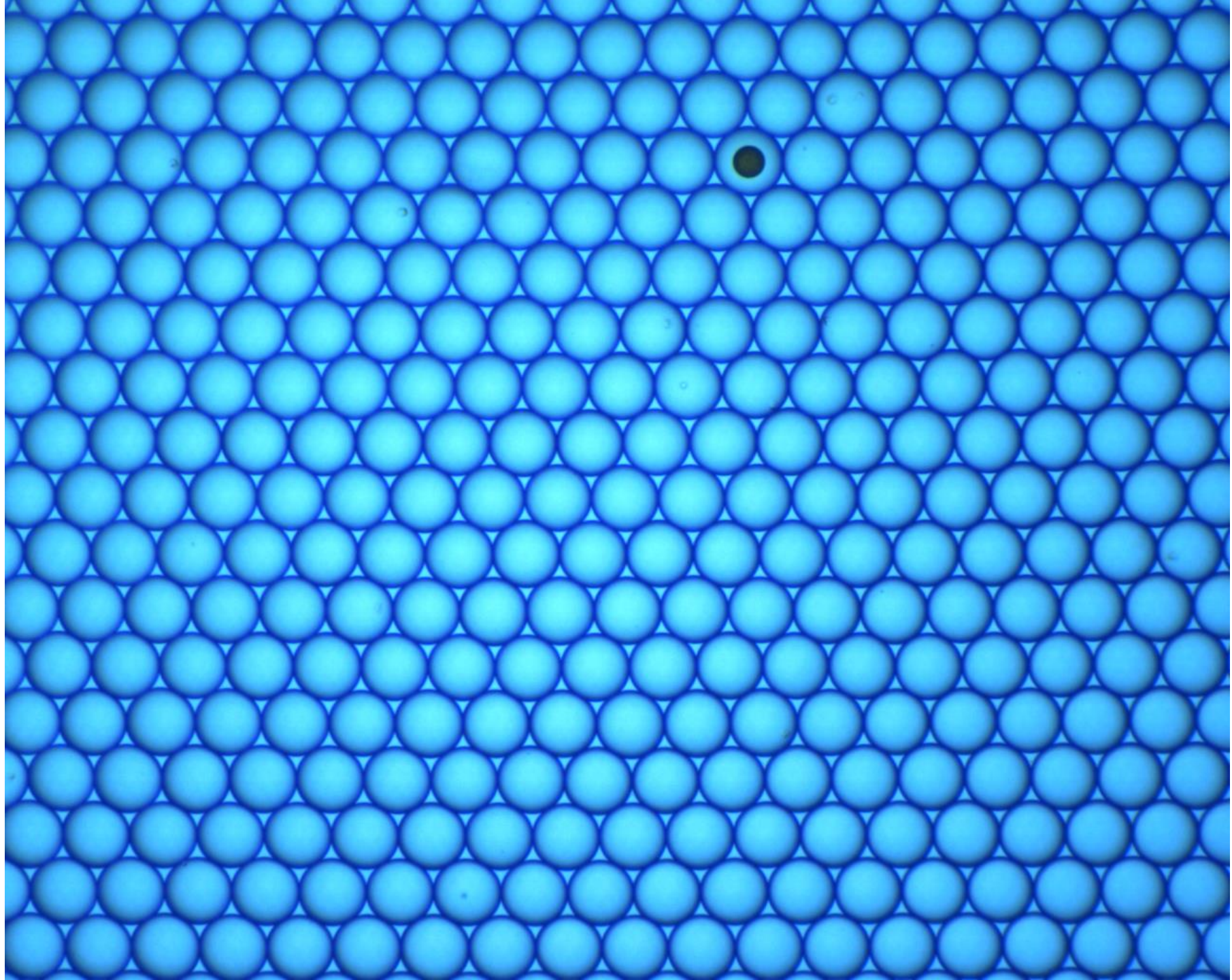




Cell migration/co-culture chip
modified from Businaro et al,
Lab Chip 2013

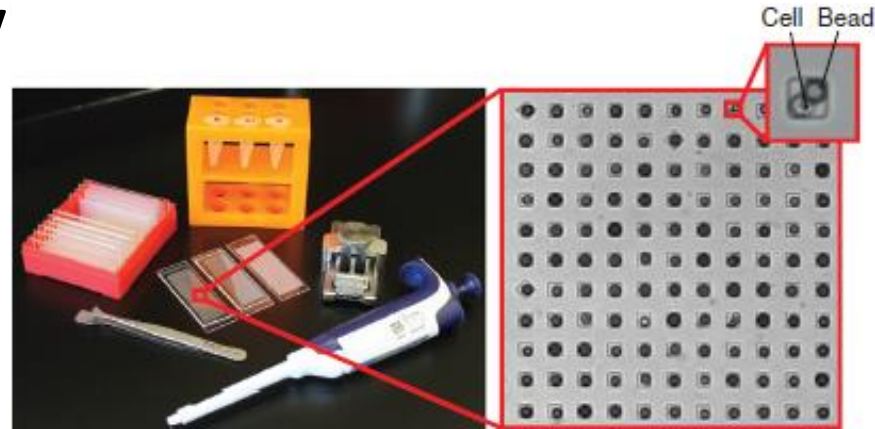
Droplet chips (Dolomite glass chip and in-house PDMS chip)



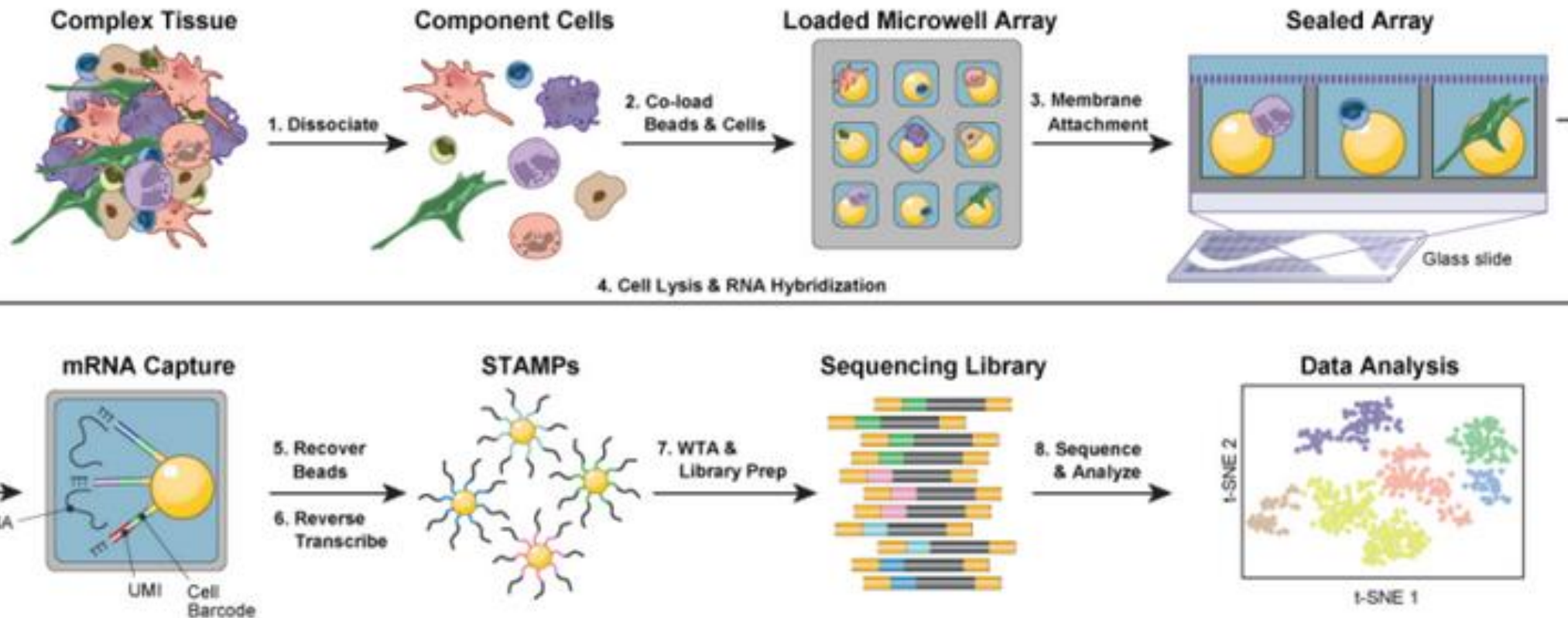
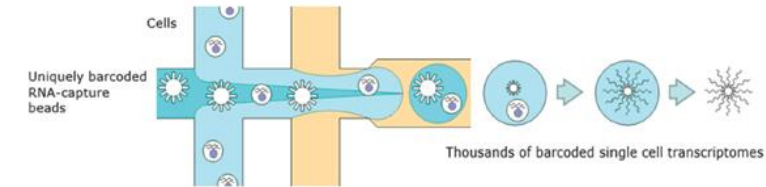


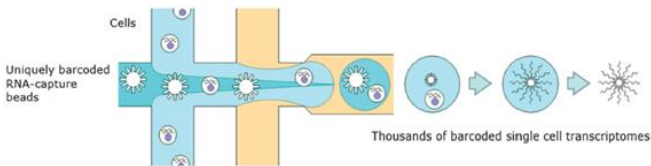
SeqWell - DropSeq chemistry but in microwells instead droplets:

Portable, academic system
(MIT, Gierahn et al 2017)



Comparison to DropSeq:





**Within droplets
Or Array wells**

12bp cell
Barcode
~17 mio

8bp mol
barcode
(UMI)

DropSeq: 180.000 cells + 180.000 beads

->5% captured + bead loss ~4000 cells

SeqWell: 10.000 cells + 110.000 beads -> 50% captured

5'TTTTTTTAAGCAGTGGTATCAACGCAGAGTACNJJJJJJJJJJNNNNNNNTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTT-3'

TTTTTTTAAGCAGTGGTATCAACGCAGAGTACNJJJJJJJJJJNNNNNNNTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTT-3'

5'TTTTTTTAAGCAGTGGTATCAACGCAGAGTACNJJJJJJJJJJNNNNNNNTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTT-3'

TTTTTTTAAGCAGTGGTATCAACGCAGAGTACNJJJJJJJJJJNNNNNNNTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTT-3'

3'AAAAAAAAAAAAAAAAAAAAAAAAAAAA---geneA-5'

3'AAAAAAAAAAAAAAAAAAAAAAAAAAAA---geneB-5'

**Collected droplets/emulsion
Library prep in bulk**

Reverse transcription (42°C), with template switch oligo -> washes

ExoSap treatment (37°C) -> washes

Full length cDNA amplification PCR

AmpureXP purification of PCR product

Tagmentation (shortening+adapters) + PCR

AmpureXP purification

Illumina NextSeq/NovaSeq/MiSeq sequencing

(20bp+55bp, from 3' end)

Data analyses (QC filtering/trimming, demultiplexing, UMI counting, alignments to reference genome/transcriptome, cell type clustering, differential expression etc)



polyA-RNA from
the lysing cell

Template switching oligo

...gcatgacCCC (cDNA)

...cgtactg (RNA)

rGrGrGtgcatgctca (TSO)

...gcatgacCCC

...cgtactgGGGtgcatgctca

...gcatgacCCCacgtacgagt

...cgtactgGGGtgcatgctca

Single cell libraries -> Illumina sequencing

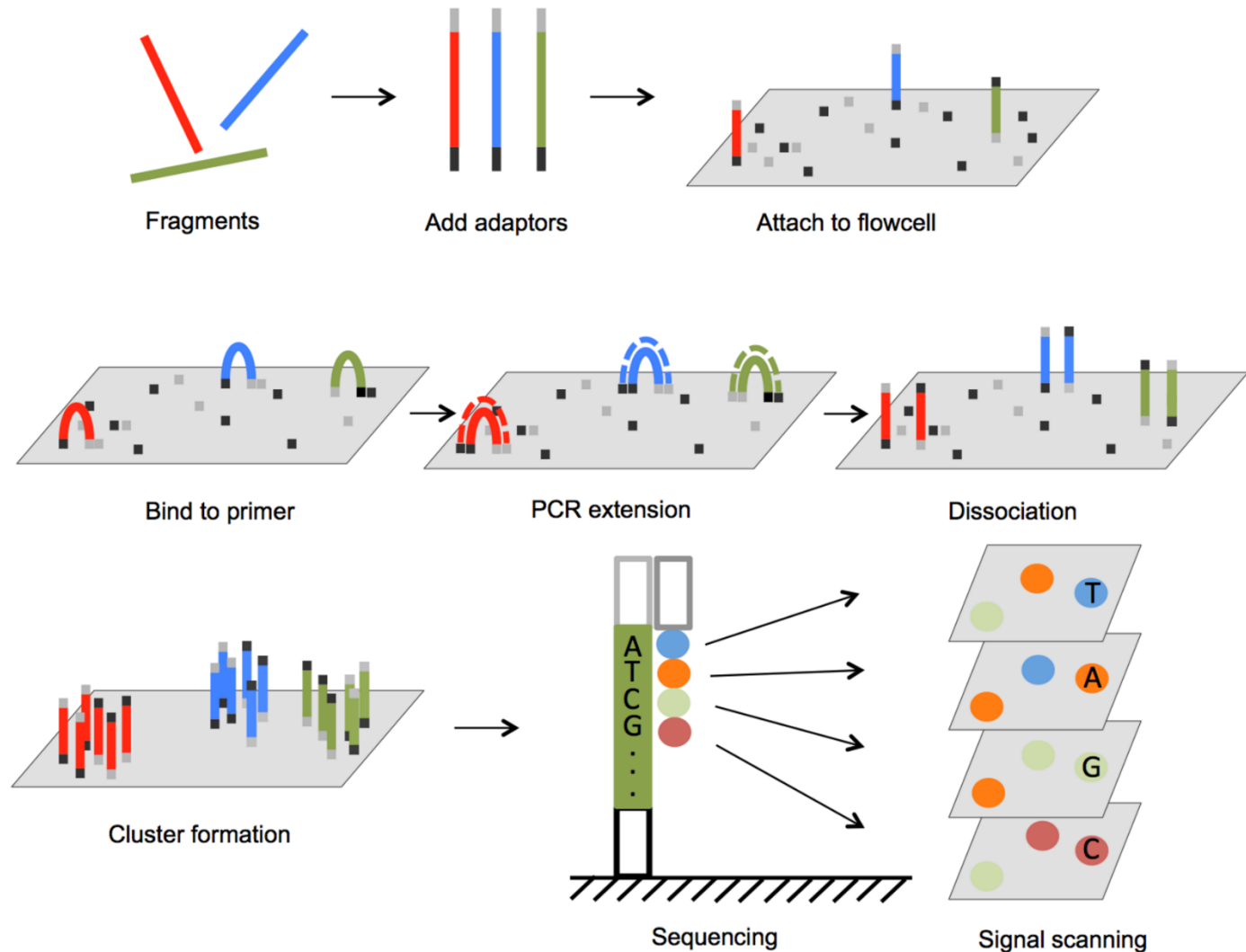
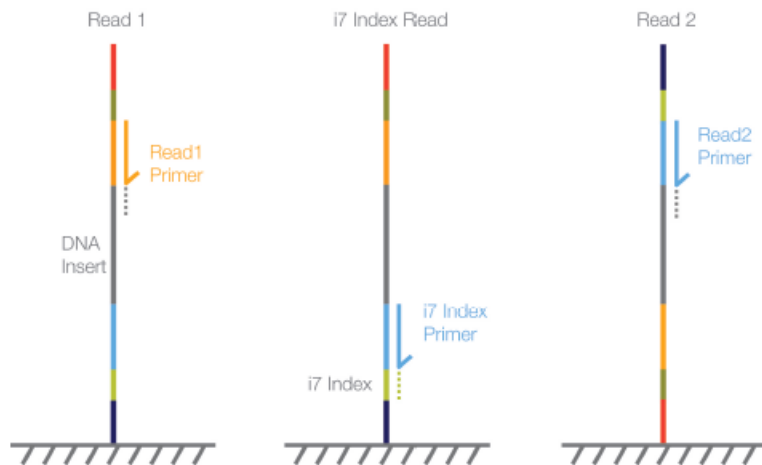
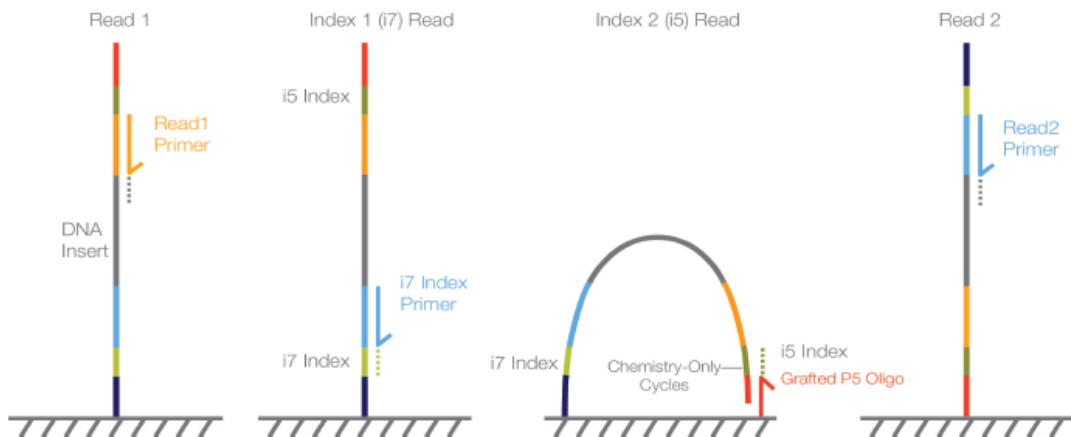


Figure 1 Single-Indexed Sequencing



- 1 **Read 1**—Read 1 follows the standard Read 1 sequencing protocol using SBS reagents. The Read 1 sequencing primer is annealed to the template strand during the cluster generation step.
- 2 **Index Read preparation**—The Read 1 product is removed and the Index 1 (i7) sequencing primer is annealed to the same template strand, producing the Index 1 (i7) Read.
- 3 **Index 1 (i7) Read**—Following Index Read preparation, the Index 1 (i7) Read is performed. The read length depends on the system and run parameters.
- 4 **Read 2 resynthesis**—The Index Read product is removed and the original template strand is used to regenerate the complementary strand. Then, the original template strand is removed to allow hybridization of the Read 2 sequencing primer.
- 5 **Read 2**—Read 2 follows the standard paired-end sequencing protocol using SBS reagents.

Figure 2 Dual-Indexed Sequencing on a Paired-End Flow Cell (Workflow A)



MiSeq

- **Workflow A**—The Index 2 Read is performed before Read 2 resynthesis, so the Index 2 (i5) adapter is sequenced on the forward strand.
- **Workflow B**—The Index 2 Read is performed after Read 2 resynthesis, which creates the reverse complement of the Index 2 (i5) index adapter sequence.

Table 1 Dual-Index Paired-End Sequencing Workflows

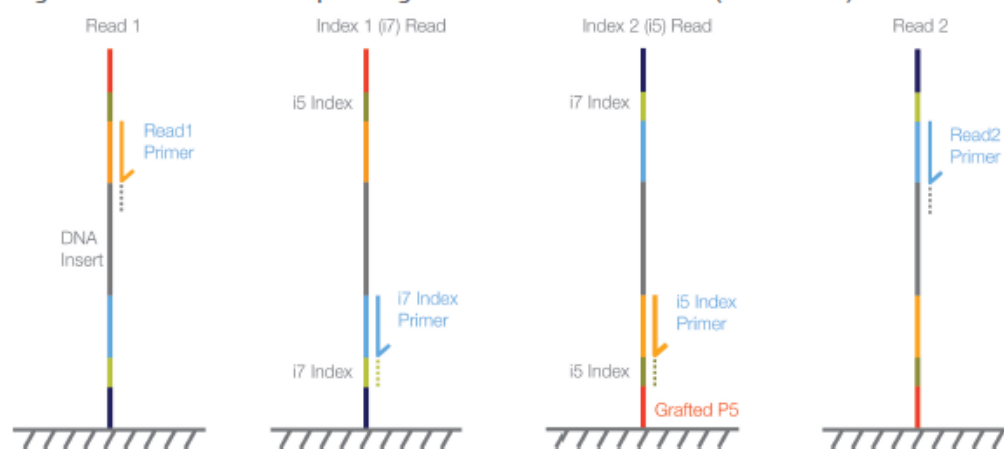
Workflow A	Workflow B
Read 1	Read 1
Index Read preparation	Index Read preparation
Index 1 Read	Index 1 Read
Index 2 Read	Read 2 resynthesis
Read 2 resynthesis	Index 2 Read
Read 2	Read 2 preparation
	Read 2

Dual-Indexed Workflow on a Paired-End Flow Cell

Dual-index sequencing on a paired-end flow cell follows 1 of 2 workflows, depending on the system:

- Workflow A is performed on the NovaSeq™, MiSeq®, HiSeq 2500, and HiSeq 2000.
- Workflow B is performed on the MiniSeq™, NextSeq®, HiSeq 4000, and HiSeq 3000.

Figure 3 Dual-Indexed Sequencing on a Paired-End Flow Cell (Workflow B)



NextSeq

Drop-seq protocols, analysis programs, example data sets, discussion forum etc:

<http://mccarrolllab.com/dropseq/>

McCarroll Lab



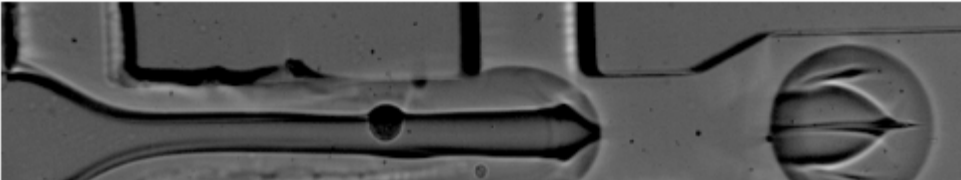
Welcome to Drop-seq!

Drop-seq is a technology that allows biologists to analyze genome-wide gene expression in thousands of individual cells in a single experiment. This work is described in [Macosko et al., Cell, 2015](#). This site provides interested users with resources to implement Drop-seq in their own labs. We hope you do amazing things with Drop-seq. Tell us about it!

If you would like to be informed of new protocol optimizations, discussion forums, etc., please send an email to dropseq@gmail.com and we'll put you on the list.

Do you currently have Drop-seq working?

We are trying to gauge the level of adoption of the technology, and would very much appreciate hearing about your experiences. If you have Drop-seq working, please email dropseq@gmail.com with a picture of your first "Barneyard" plot and, if possible, the broad scientific direction/question you are hoping to address with the technology. We would really love to hear from all of you!



Drop-Seq Laboratory Protocol

version 3.1 (12/28/15)

[Evan Macosko](#) and [Melissa Goldman](#)

[Steve McCarroll's lab](#), Harvard Medical School

Bioinformatics of Drop-seq data

<http://www.tongji.edu.cn/~zhanglab/Drseq/>

Dr. SEQ

quality control and analysis for Drop-seq

Getting started

Overview

Manual

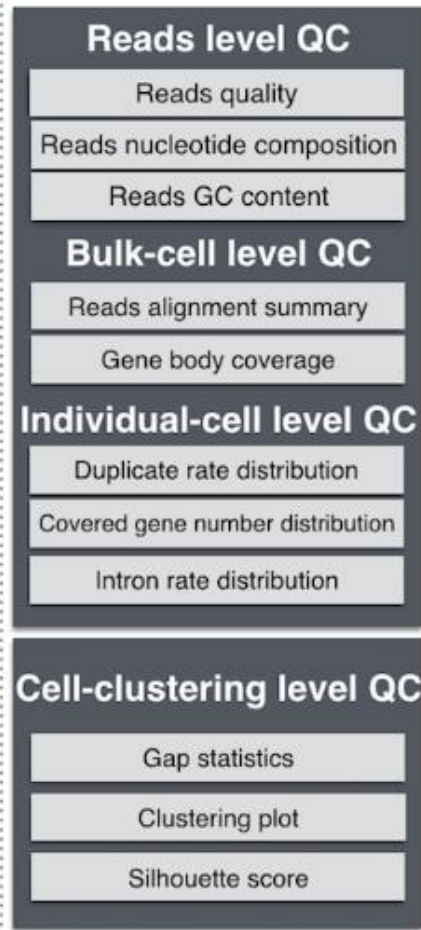
Output

Download

FAQ

Quality control

Analysis



Drop-seq Core Computational Protocol

version 1.0.1 (6/11/15)

James Nemesh

Steve McCarroll's lab, Harvard Medical School



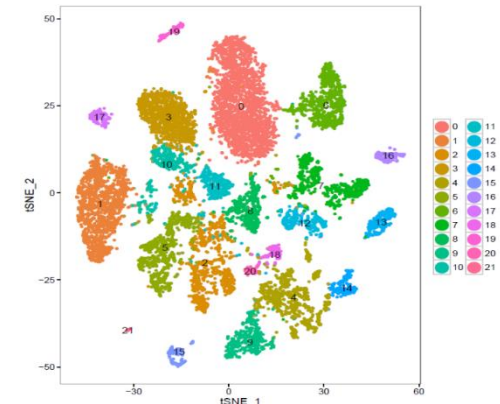
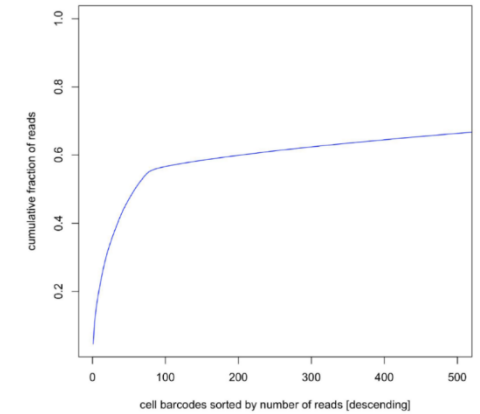
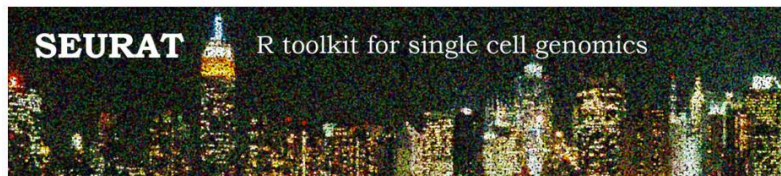
CellName	GTAAGC AGCGCG	CTTGCC AAGAAG	GCGAAT TCCCCT	GTAAGC AGTGGT	CGAGTC GGCCGT	GGTACT TGCTGG	GATTAC CGTGGG	ATTTAG GGAGCG	TCTACG AGTCAA
HBB	19480	18	2	243	4	0	0	0	36
HBA2	3163	4	0	33	1	0	0	0	4
HBA1	1248	3	0	19	0	0	0	0	3
MT-RNR2	481	350	205	1739	21	1523	457	285	841
BRSK1	200	0	0	1	0	0	0	0	0
UBB	119	14	0	5	0	0	0	0	1
RP11-1381.4	98	12	0	0	0	0	0	0	0
GUK1	91	79	0	1	0	0	0	0	0
VTI1B	81	0	0	3	0	0	0	0	0



<http://satijalab.org/seurat/>

SATIJA LAB

HOME NEWS PEOPLE RESEARCH PUBLICATIONS SEURAT JOIN/CONTACT



CSC/Chipster: Drop-seq pipeline + Seurat

The screenshot displays the Chipster 3.12.2 software interface, which is used for managing and executing bioinformatics workflows. The interface is divided into several panels:

- Datasets:** A list of input files including `hisat.log`, `drseq_read_1_merged.bam`, `drseq_read_1_merged.bam.bai`, `merged_tagged.bam`, `merged_tagged.bam.bai`, `inflectionPoint.pdf`, `cell_readcounts.txt.gz`, `synthesis_stats_summary.txt`, `digital_expression.tsv`, `cleaned.bam`, `synthesis_stats.txt`, `digital_expression_summary.txt`, `inflectionPoint.pdf`, and `cell_readcounts.txt.gz`.
- Analysis tools:** A list of tools categorized by type (Microarrays, NGS, Misc). The **NGS** category is selected, showing tools like **Create digital gene expression matrix**, **BETA Seurat - Setup and QC**, **BETA Seurat - Filtering, regression and detection of variable genes**, **BETA Seurat - PCA**, **BETA Seurat - Clustering**, and **BETA Seurat - Visualize biomarkers**. The **Create digital gene expression matrix** tool is highlighted, and its description is shown on the right: "Corrects bead synthesis errors and extracts gene expression values from a BAM file where reads have been tagged with gene names. The resulting Digital Gene Expression matrix, DGE, can be used for further analysis with the Seurat tools."
- Workflow:** A visual representation of the Drop-seq pipeline. It shows a sequence of steps: `html` (yellow) → `gz` (green) → `html` (yellow) → `gz` (green) → `bam` (blue) → `txt` (blue) → `pdf` (blue) → `bam` (blue) → `bai` (blue) → `log` (red) → `html` (yellow) → `bam` (blue) → `bai` (blue) → `bam` (blue) → `bai` (blue) → `pdf` (blue) → `gz` (green) → `txt` (blue) → `tsv` (blue) → `bam` (blue) → `txt` (blue) → `txt` (blue) → `pdf` (blue) → `gz` (green).
- Visualisation:** A spreadsheet view showing the results of the pipeline. It displays 1997 rows of data (genes) and 501 columns (genes). The first column is labeled **GENE**, and the subsequent columns are labeled with gene names: `TCATTTAGT...`, `GATCGGTACATG`, `TAGTCCCAACGA`, `TAACAACCTCTA`, `GTTGTCGTCTGA`, `ACTGCAAGGAAT`, `TCTAGGTCATAG`, `CCGTCAAATTGG`, and `CCGGTATC`. The data is presented as a table of counts for each gene across the different samples.

At the bottom of the interface, there is a status bar showing the connection to `chipster.csc.fi` as `pamaho`, a **View jobs** button, a **0 jobs running** status, and a **Used memory 156M / 870M** indicator. The Windows taskbar at the very bottom shows various application icons and the system clock at 12:07.

10x Genomics Chromium (droplet based platform in FIMM) tools

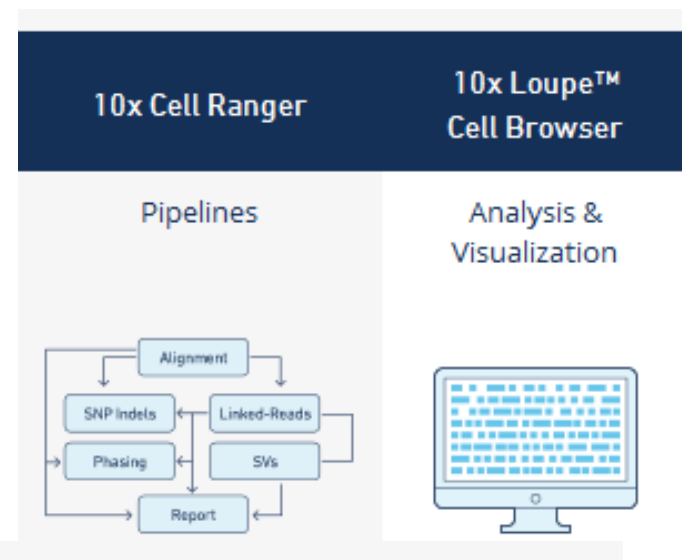
Cell Ranger:

- **cellranger mkfastq** wraps Illumina's bcl2fastq to correctly demultiplex Chromium-prepared sequencing samples and to convert barcode and read data to FASTQ files.
- **cellranger count** takes FASTQ files from `cellranger mkfastq` and performs alignment, filtering, and UMI counting. It uses the Chromium cellular barcodes to generate gene-barcode matrices and perform clustering and gene expression analysis. `count` can take input from **multiple sequencing runs** on the same library.
- **cellranger aggr** aggregates outputs from multiple runs of `cellranger count`, normalizing those runs to the same sequencing depth and then recomputing the gene-barcode matrices and analysis on the combined data. `aggr` can be used to combine data from multiple samples into an experiment-wide gene-barcode matrix and analysis.
- **cellranger reanalyze** takes gene-barcode matrices produced by `cellranger count` or `cellranger aggr` and reruns the dimensionality reduction, clustering, and gene expression algorithms using tunable parameter settings.

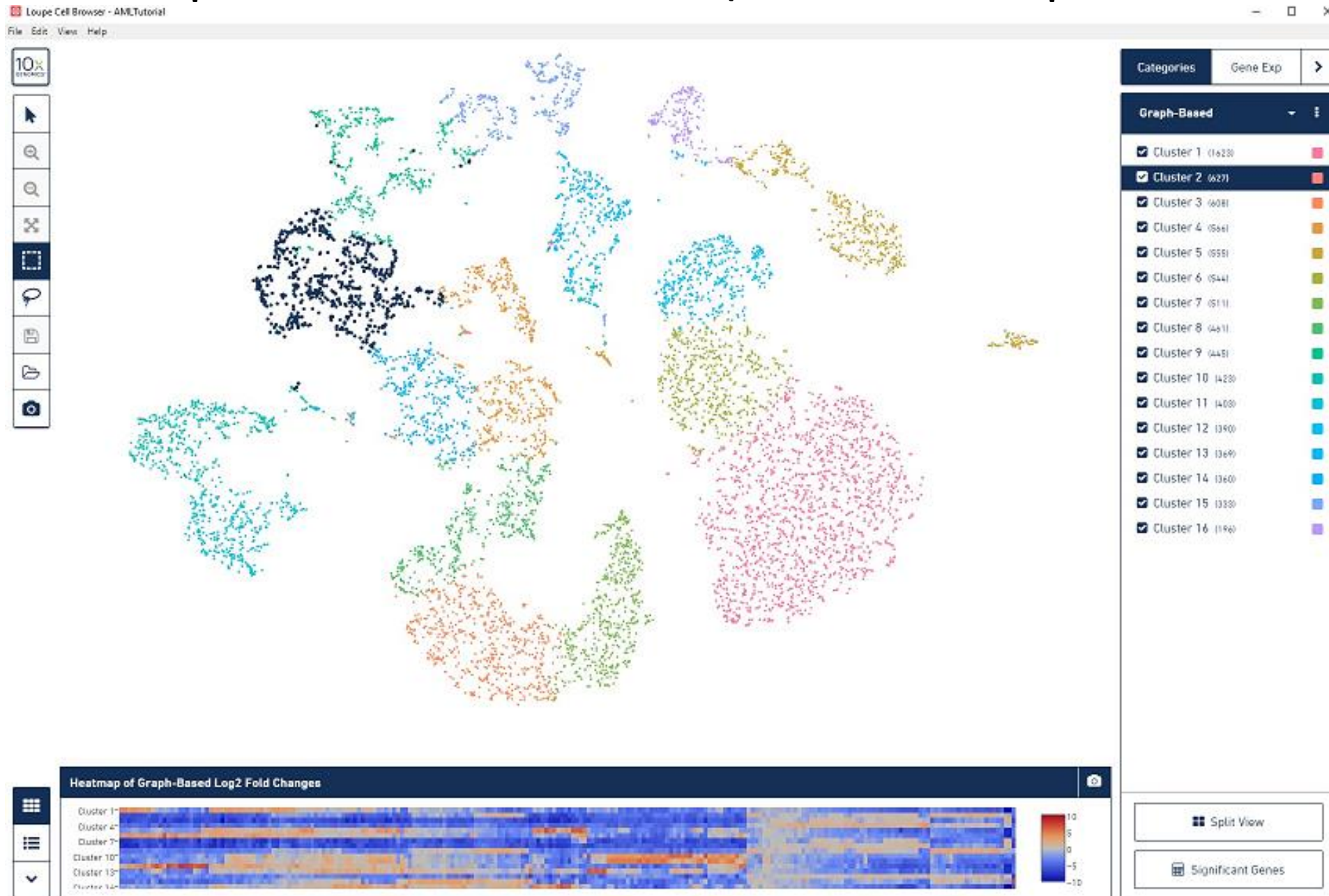
Loupe Cell Browser:

Loupe Cell Browser opens `.cloupe` files generated by the Cell Ranger 1.3 pipeline, or by `cellranger mkloupe` for older pipelines. Once your data is in Loupe Cell Browser, you can rapidly explore and gain insights from the data without writing a line of code:

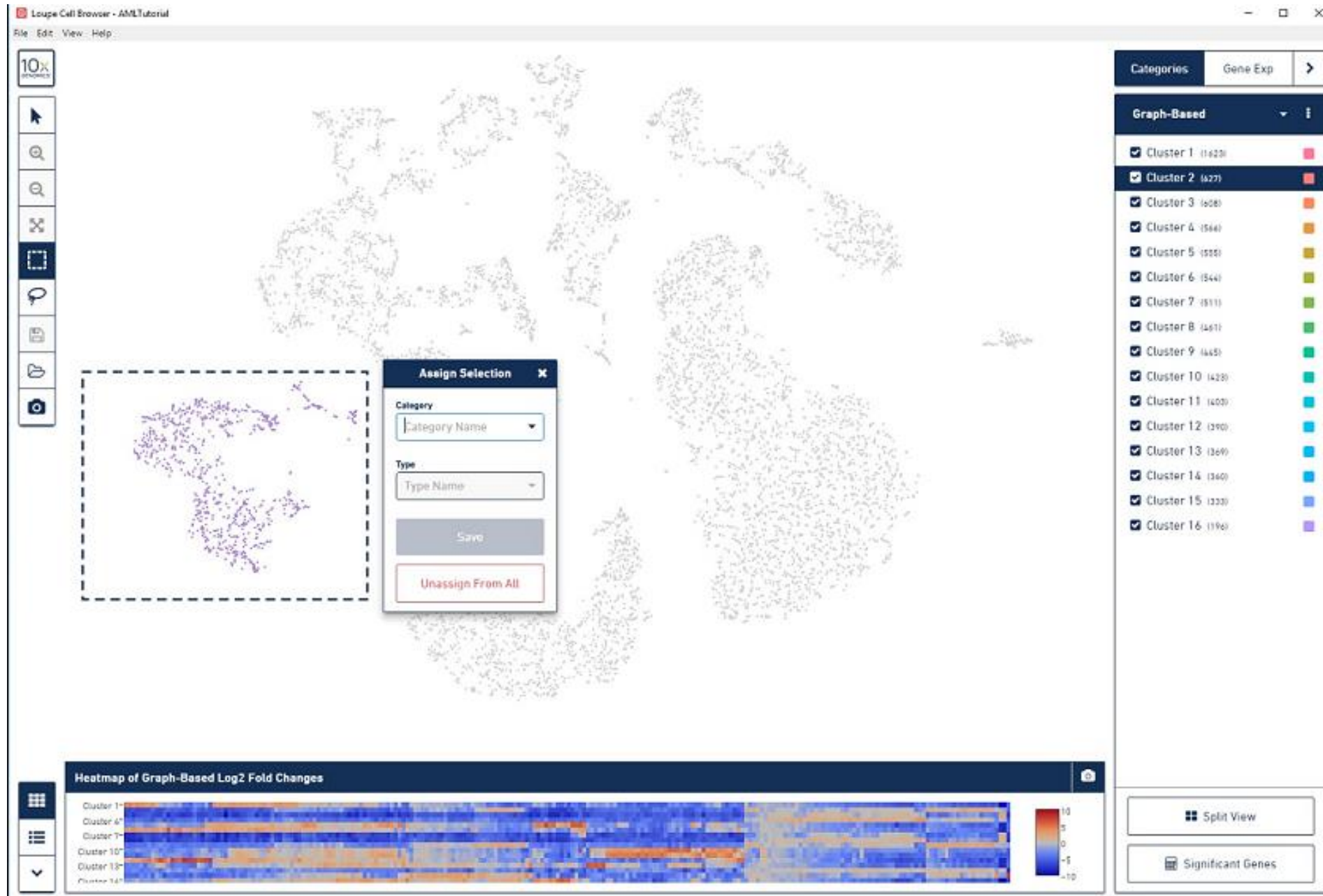
- **Finding Significant Genes**
Determine genes that uniquely characterize clusters at the press of a button.
- **Identifying Cell Types**
Use gene lists and the gene expression view to locate different cell types and functional groups.
- **Exploring Substructure**
Create custom clusters and use differential expression tools to identify even smaller subgroups in your data.
- **Sharing Results**
Save genes of interest, export data tables, and capture screenshots of your single-cell data.



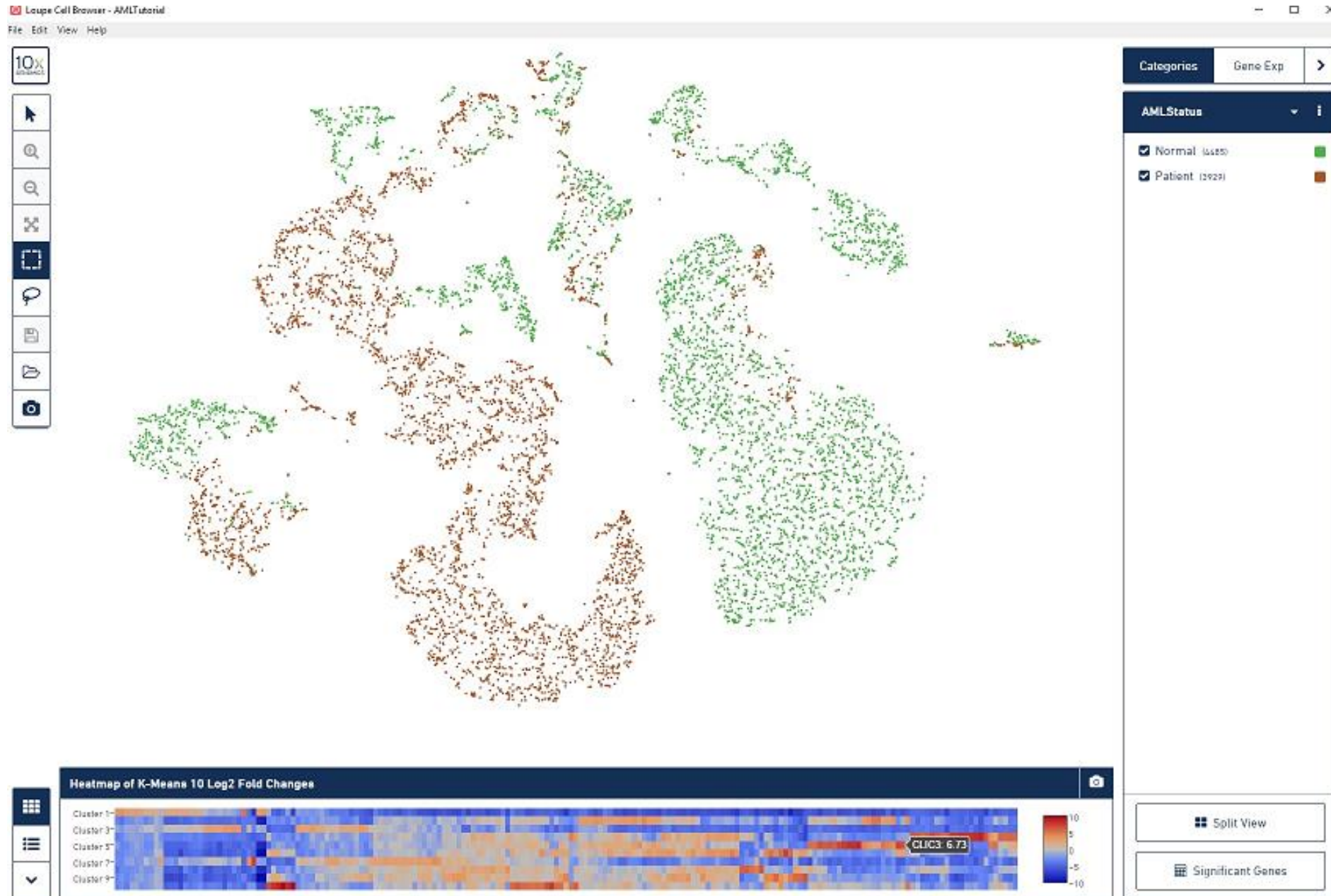
10x Loupe Cell Browser (also nice public data sets!)



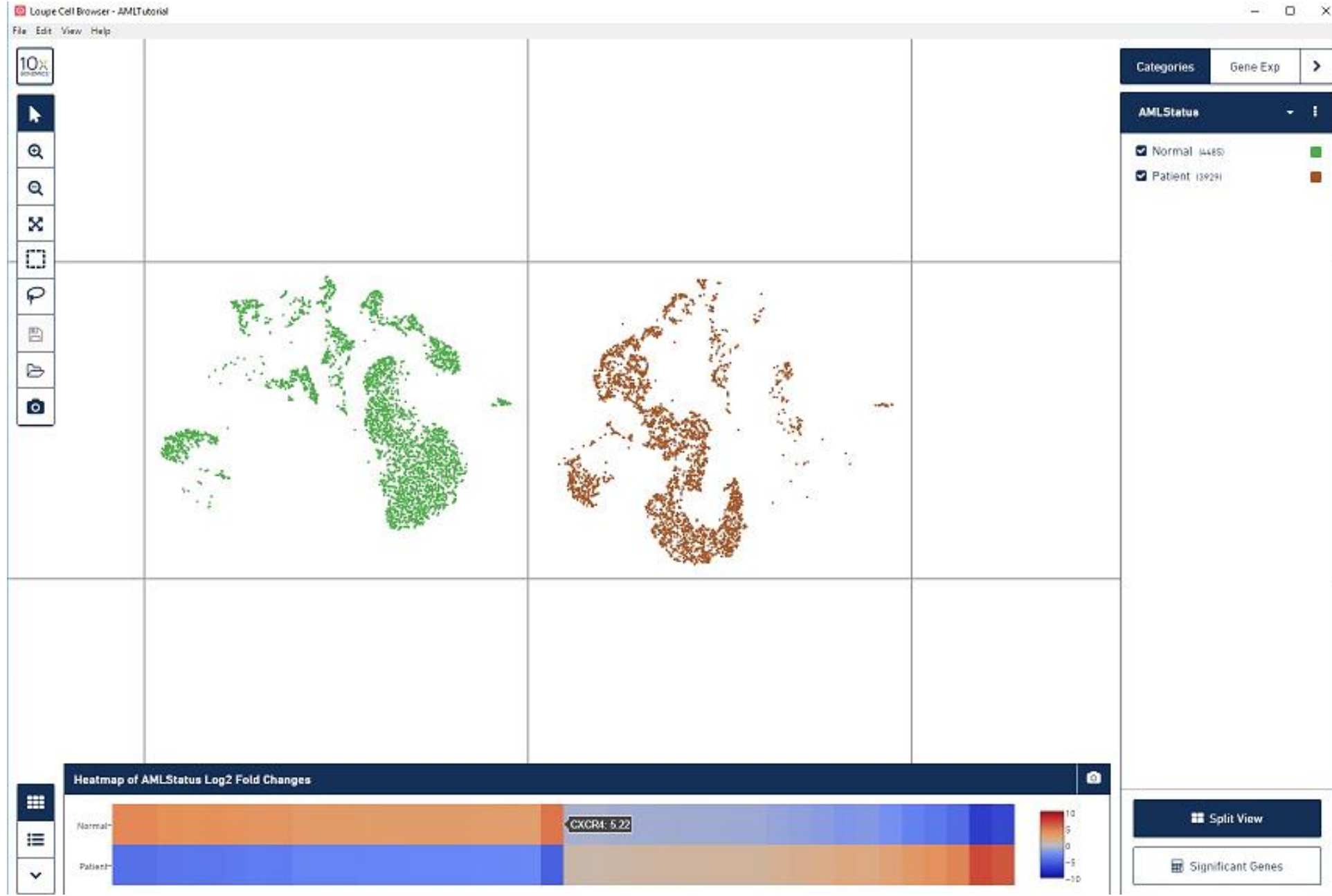
10x Loupe Cell Browser



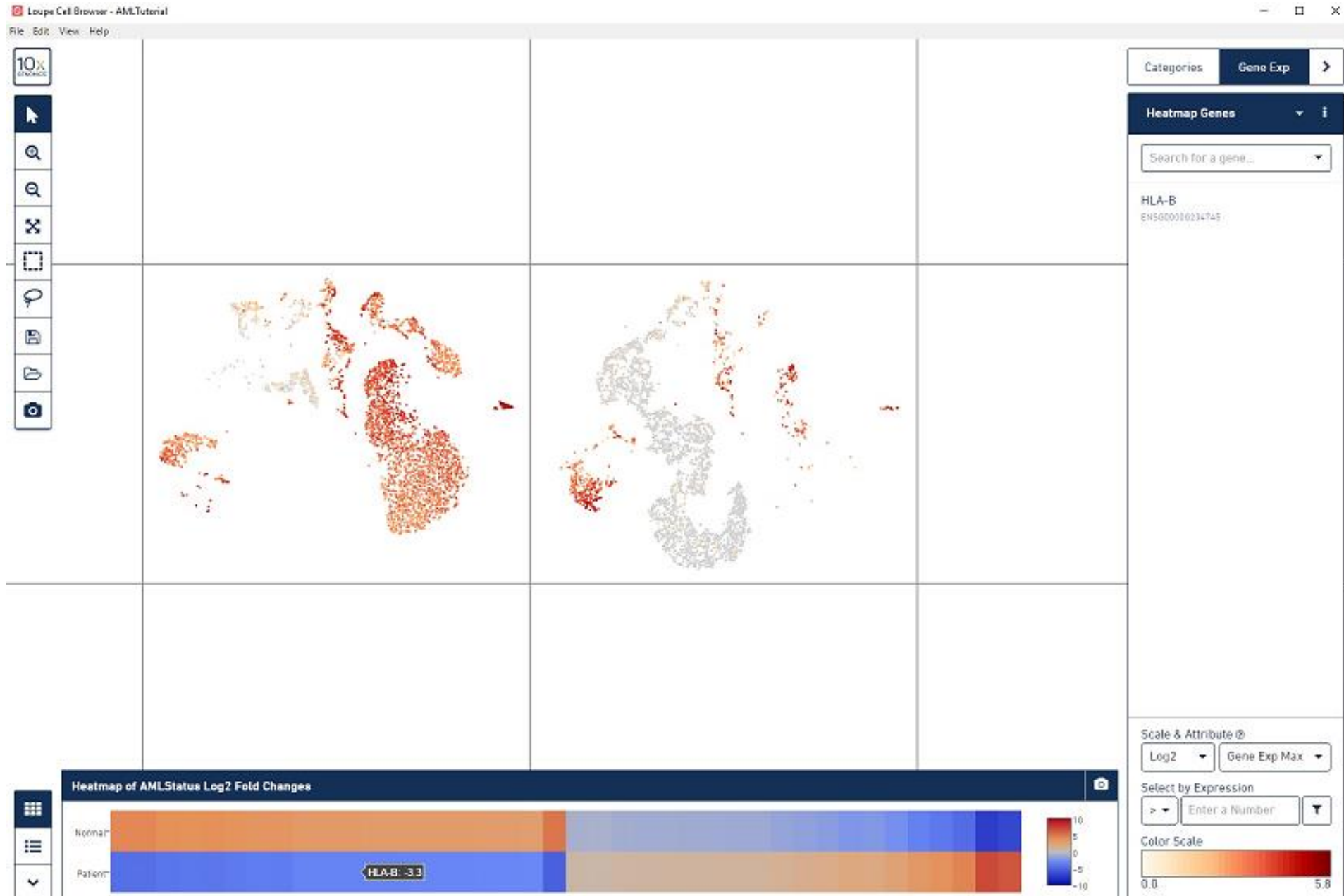
10x Loupe Cell Browser



10x Loupe Cell Browser



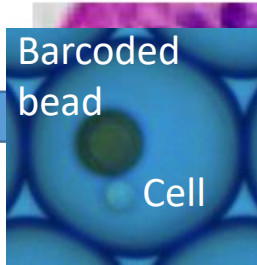
10x Loupe Cell Browser



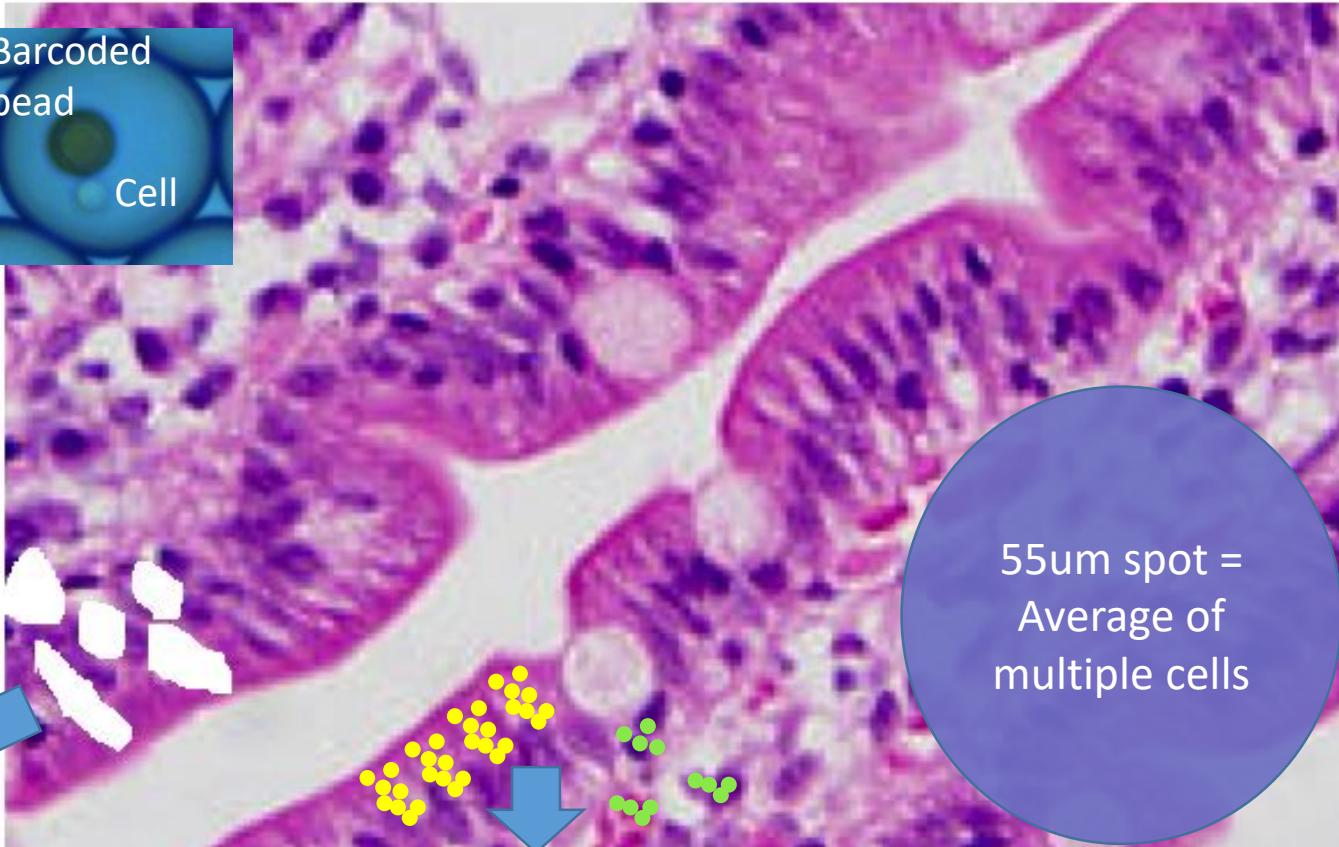
Spatial single cell transcriptomics, directly from tissues

-> original location of cells, cell-to-cell interactions...

Enzymatic/mechanic dissociation-> suspension-> barcode in droplets (e.g. 10x Genomics Chromium)



Laser microdissection + sequencing



Gene probing/in situ hybridization (e.g. Merfish, NanoString)

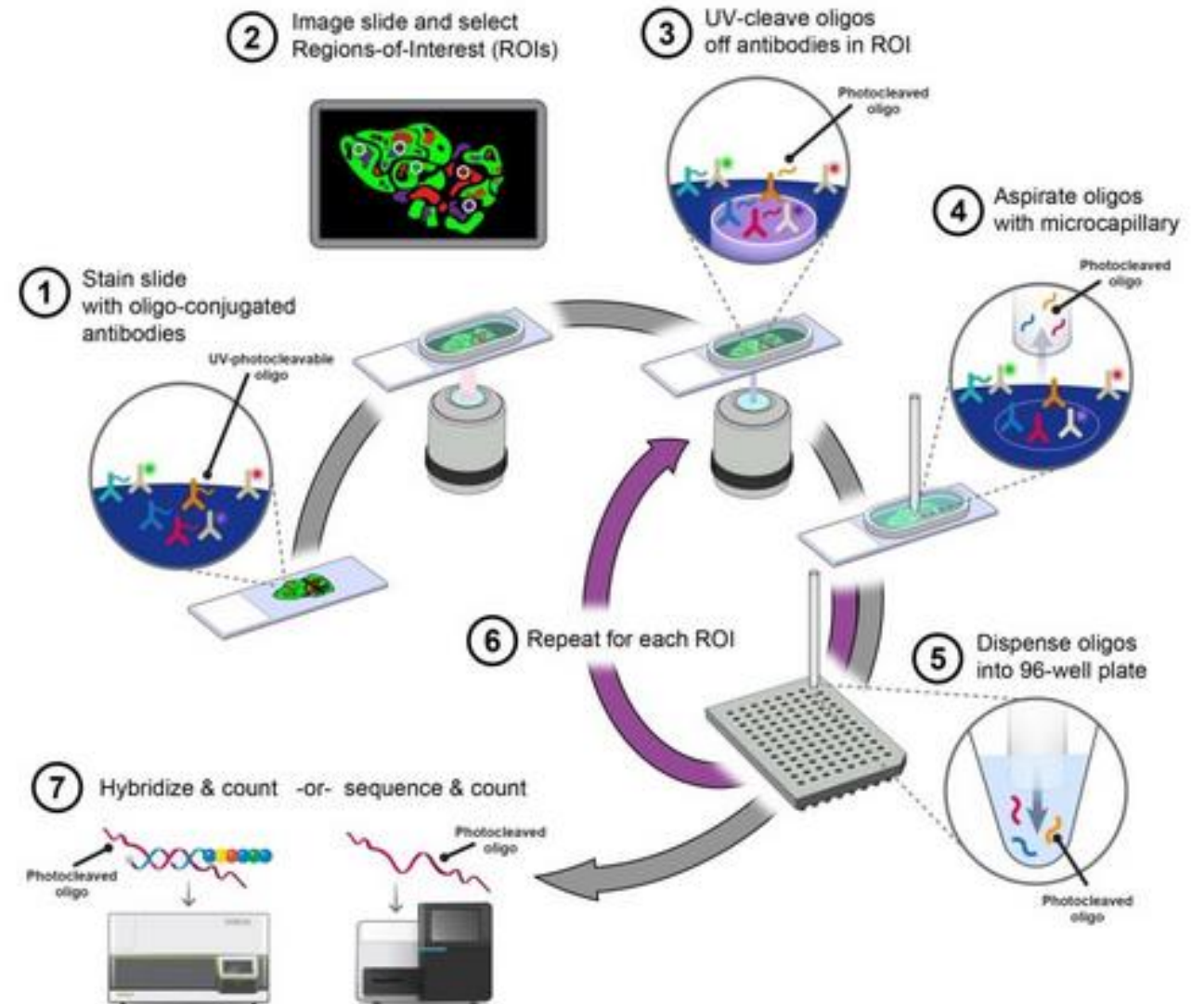
Spatial microarrayed barcode slides (e.g. Spatial Transcriptomics -> acquired by 10x Genomics -> Visium system)

NanoString nCounter and GeoMx

GeoMx® Digital Spatial Profiler Your GPS for Biology

Morphological context and high-plex protein or RNA expression data – now possible from just one FFPE slide.

- Quantify up to 96 proteins and over 1000 RNA targets with spatial context
- Process 10-20 slides /day with readout on the **nCounter® System** or with NGS
- Preserve precious samples with non-destructive processing

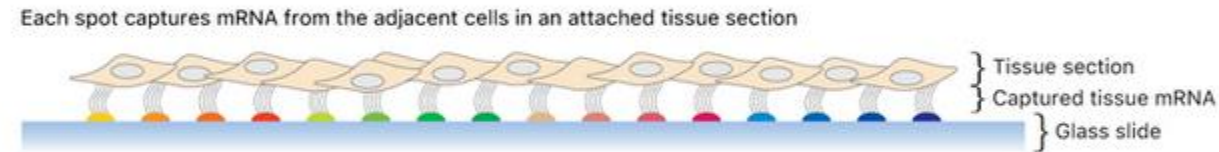
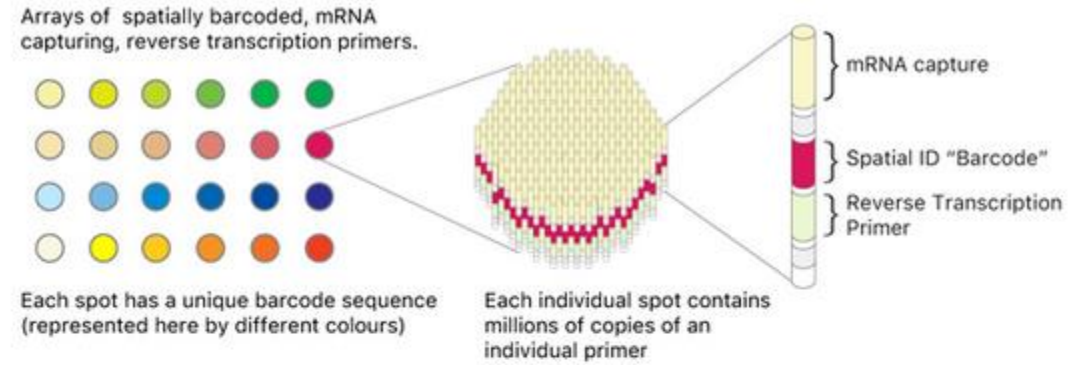
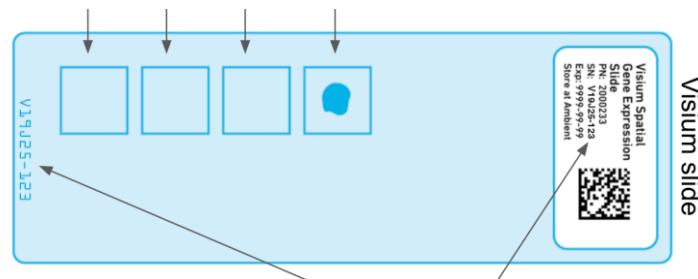




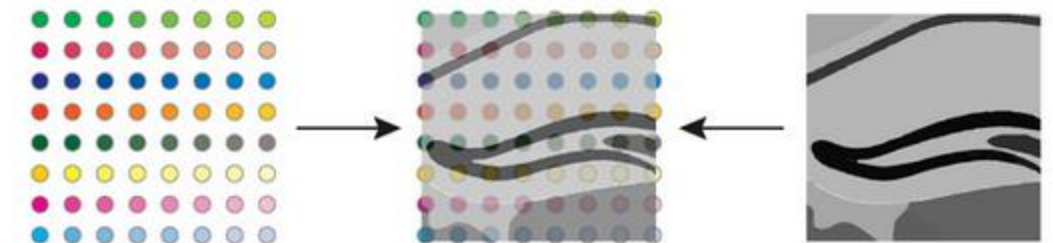
Spatial Transcriptomics

(Lundeberg, Frisen, Sweden)

Acquired by 10x Genomics
 -> Visium system
 (at FIMM)
 4 x 5000 spots
 Diameter 55um each

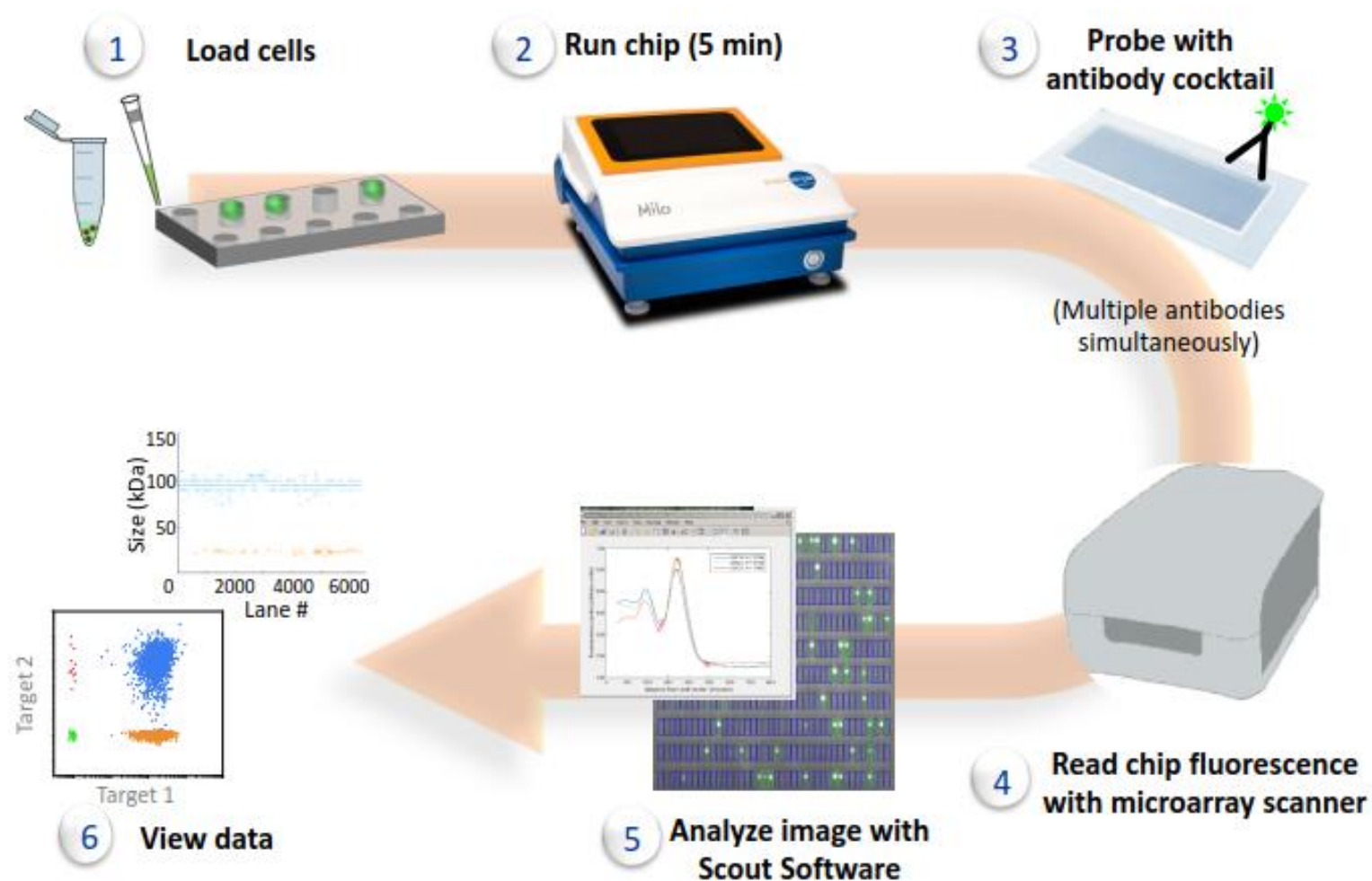


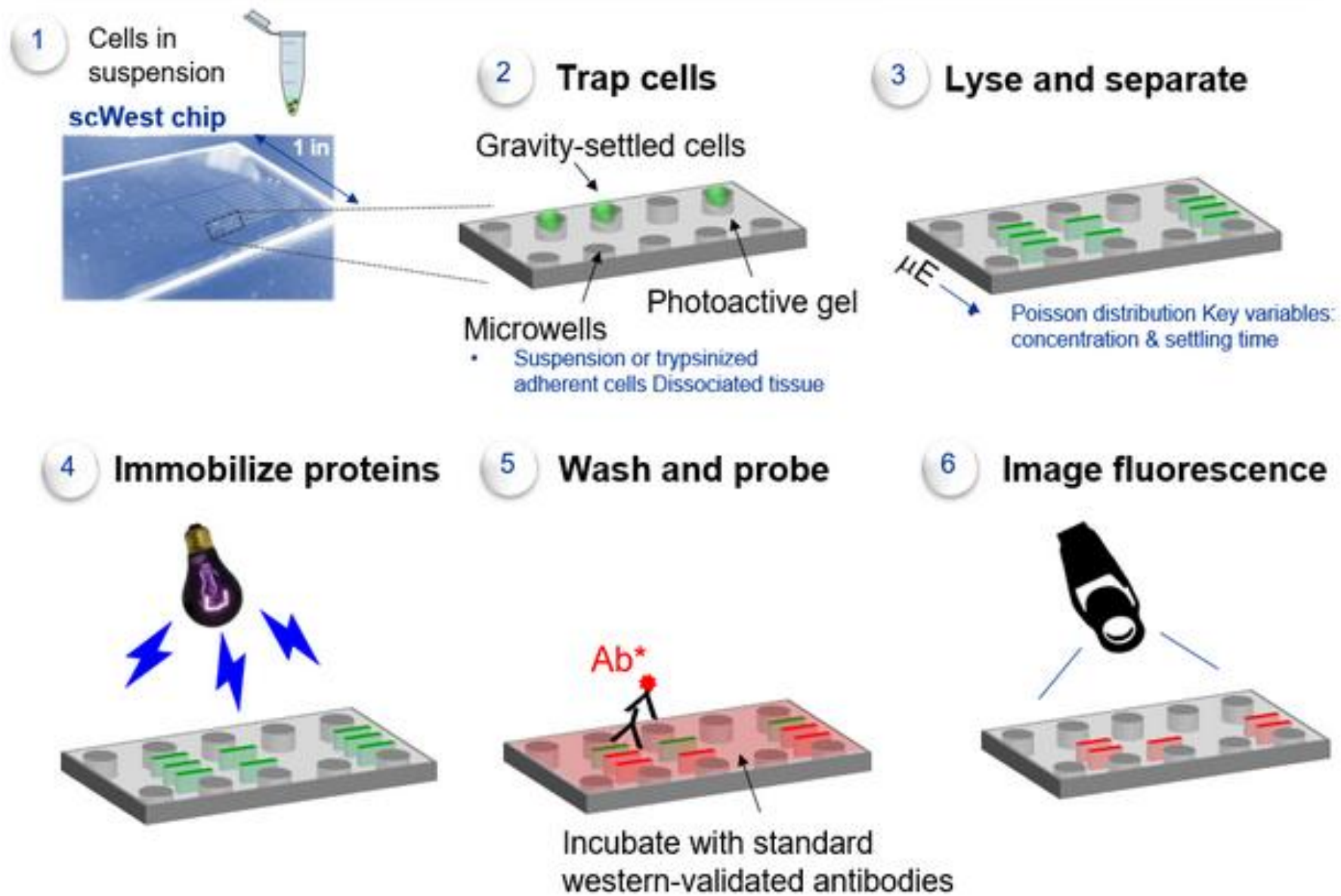
Sequencing data for each individual mRNA can be mapped back to its spot of origin on the array. Alignment of the array with a previously captured image of the tissue section then allows the spatial analysis of all tissue mRNAs.



MILO single cell Western blotting

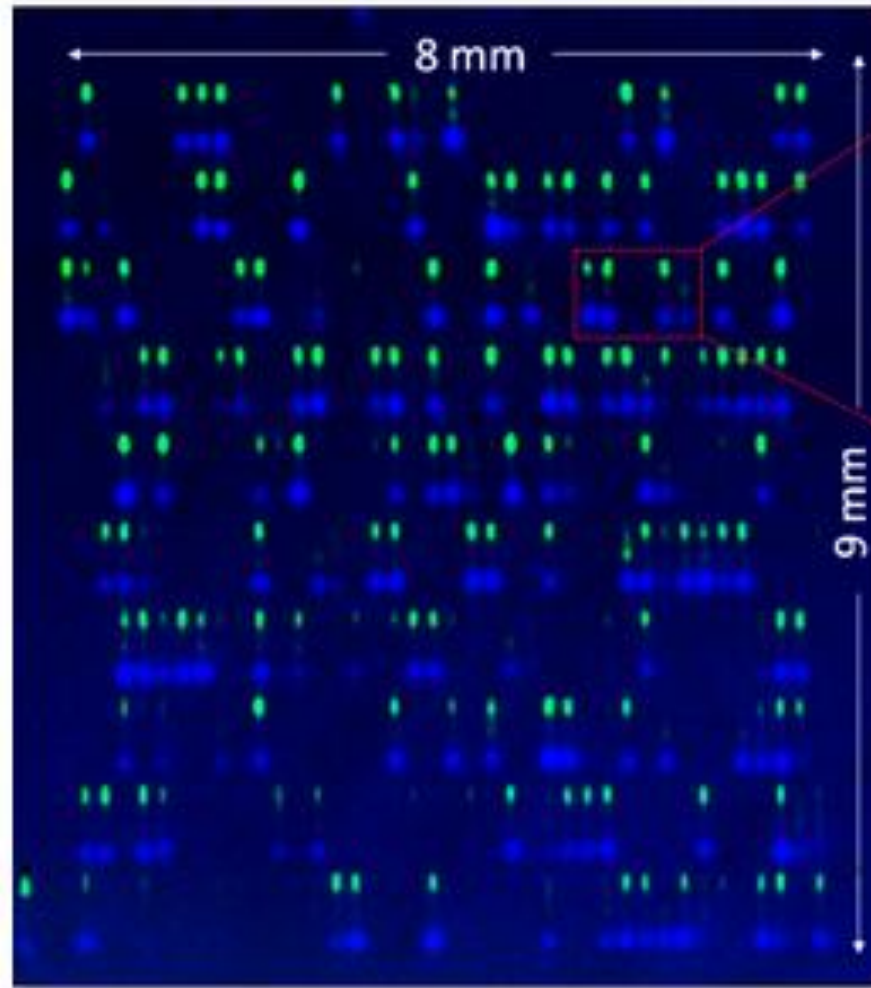
Single-Cell Western Workflow



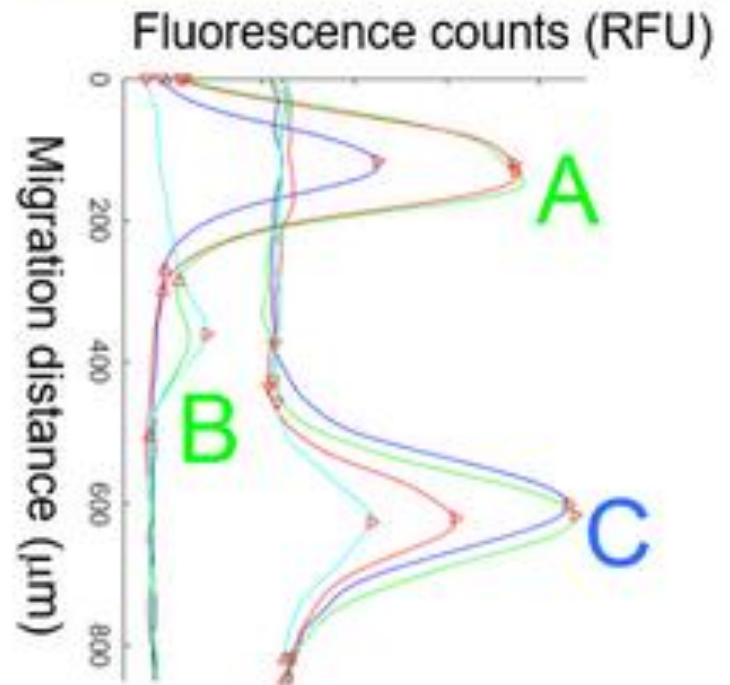
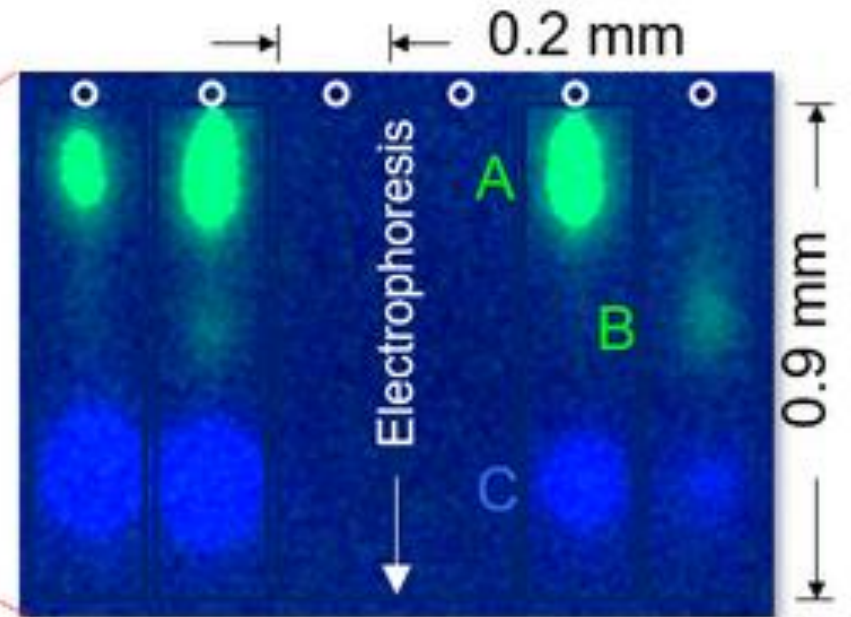


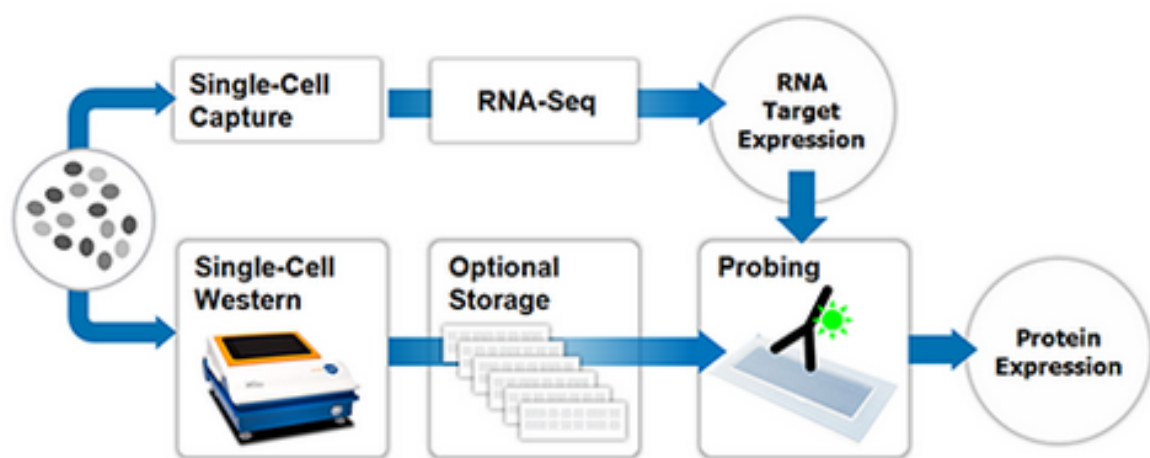
Slide scanning
(2 colours)
at FuGU Agilent
microarray
scanner

Scout software
for slide image
analyses
(free at
ProteinSimple
and also in UH
software
center)



1/16th of scWest chip





I need protein validation of my single-cell RNA-seq data

Use conventional western antibodies to validate your single-cell RNA-seq data with single-cell protein data. Plus, scWest chips can be archived for up to 9 months after you run them so you have plenty of time to get your sequencing results back before you have to probe for your targets of interest.

[Learn More >](#)

Available single cell platforms in Helsinki area

Single cell omics core facilities in Helsinki

<https://www.helsinki.fi/en/infrastructures/single-cell-omics>

FIMM: 10x Chromium, index sorting. CellenOne dispenser

Saavalainen/Functional Genomics Unit (FuGU):

Drop-seq, Seq-well, index sorting

Viikki: inDrops, Polaris, Helios CyTOF (mass cytometer)

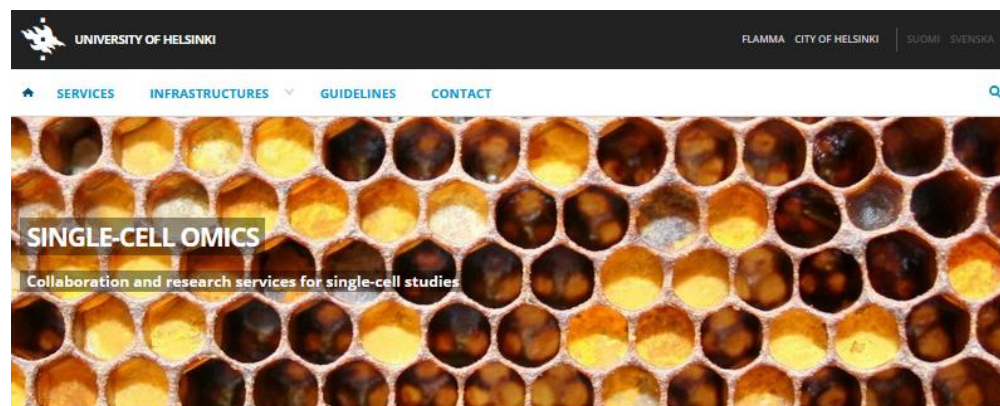
FACS cores for single cell plate sorting

Other devices elsewhere:

Fluidigm C1: Anna Vähärautio/Biomedicum, Kristiina Iljin/VTT,

Seppo Ylä-Herttuala/UEF

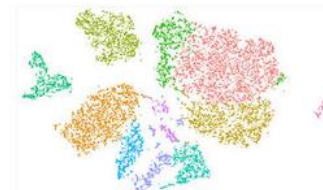
10x Chromiums in Helsinki: FIMM, Viikki, Aaltonen and Alitalo labs



Single-Cell omics platform aims at bringing the latest single cell omics tools available for the research community. Services based on novel instruments are now available for single cell genomics and single cell proteomics to enable cutting-edge research in many areas of molecular biology and molecular medicine.

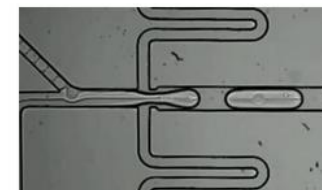


SC OMICS LIFE SCIENCE RESEARCH INFRASTRUCTURES



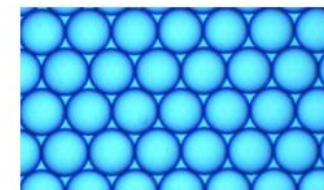
Single-cell Analytics, FIMM

FIMM Single Cell Analytics (SCA) unit develops and provides single cell transcriptomics services nationally and internationally. SCA is...



Single-cell Proteomics and Genomics, BI

Single-Cell Proteomics and Genomics (ProtGen) unit develops and provides single cell services to both University of Helsinki internal and...



Single Cell Microfluidics Service Lab

Cost-effective RNA sequencing from thousands of single cells



SC OMICS bioinformatics

Bioinformatics services for multidimensional scRNA-seq readout

Single-Cell Analytics

Jenni Lahtela

+358 50 319 8725

jenni.lahtela@fimm.fi

Single-Cell Proteomics and Genomics

Mantas Survila

+358 45 357 1991

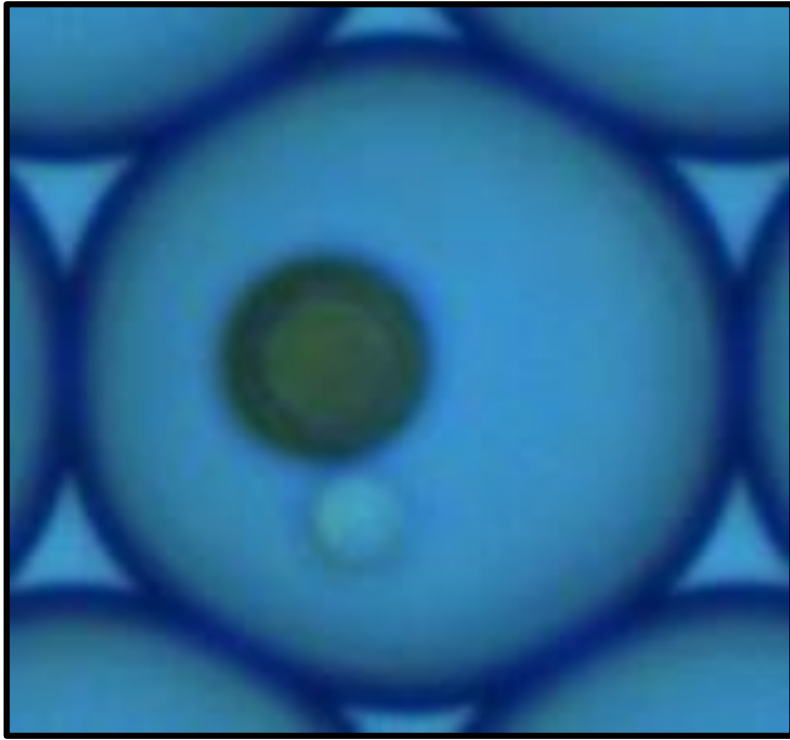
mantas.survila@helsinki.fi

Single-Cell Microfluidics Service Lab

Päivi Saavalainen

+358 50 4487487

paivi.saavalainen@helsinki.fi



Thank you for your attention!

