

Scheine (best 4 out of 5), first week of March

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From fluorescence microscopy to sequencing and back

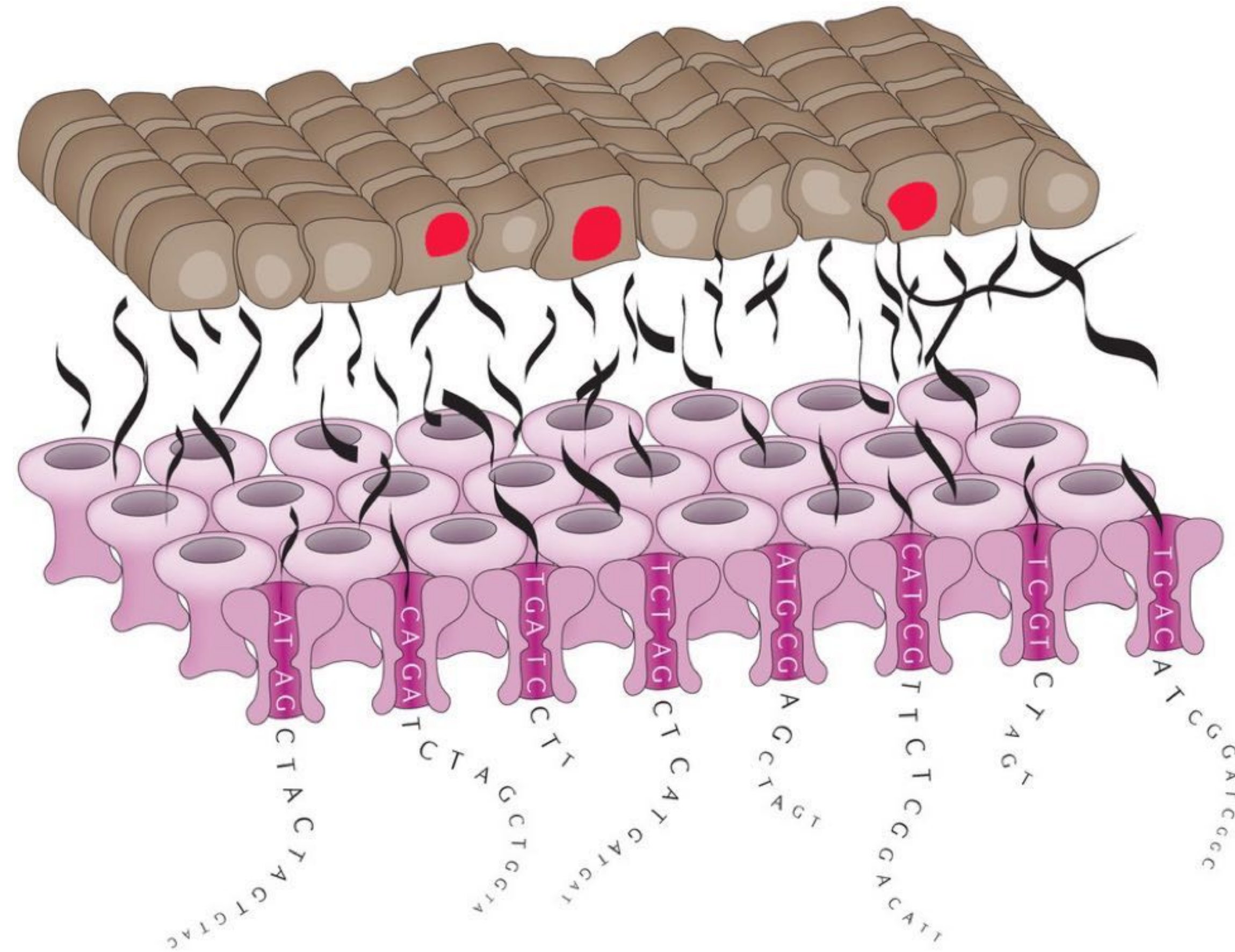
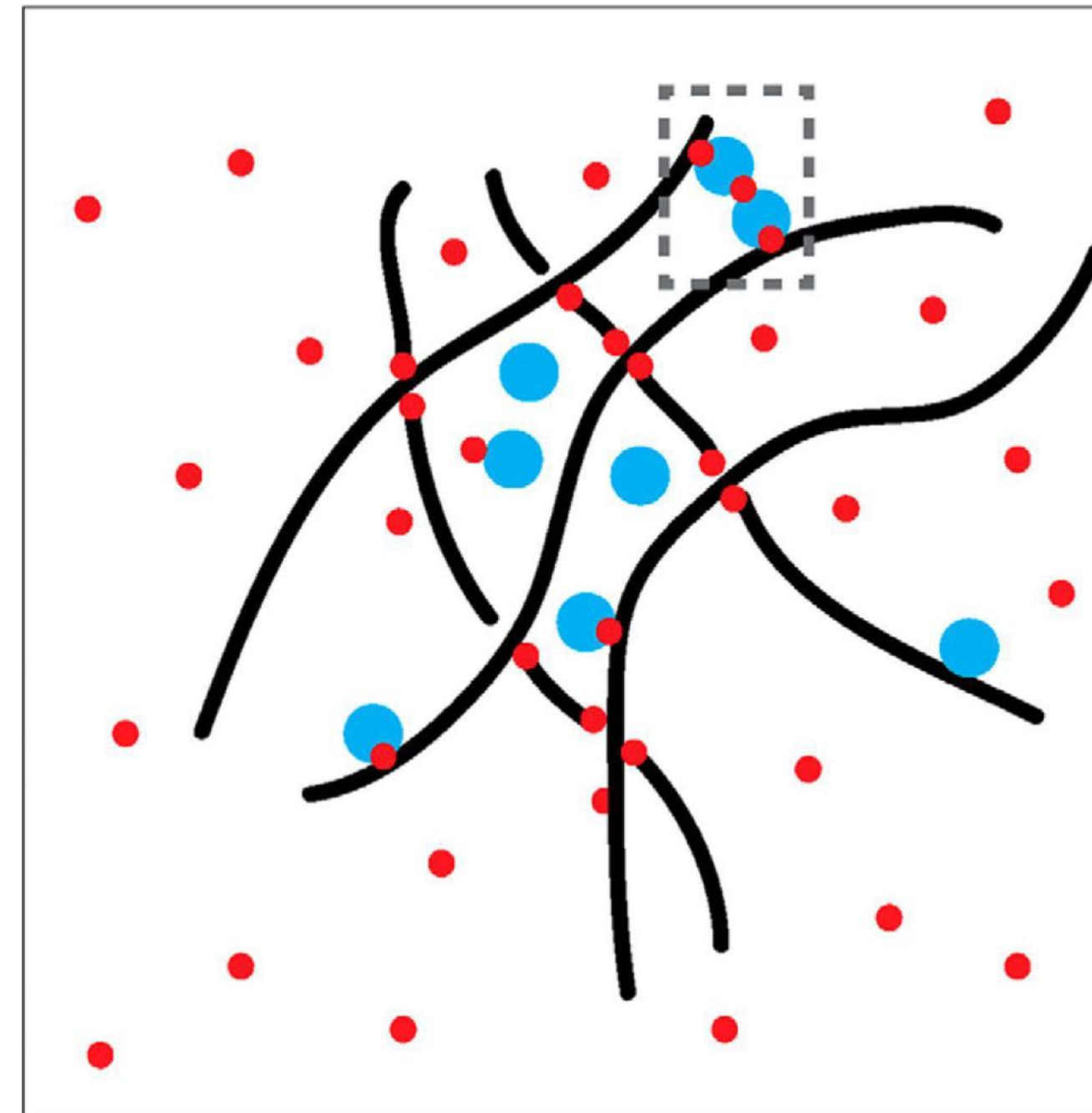
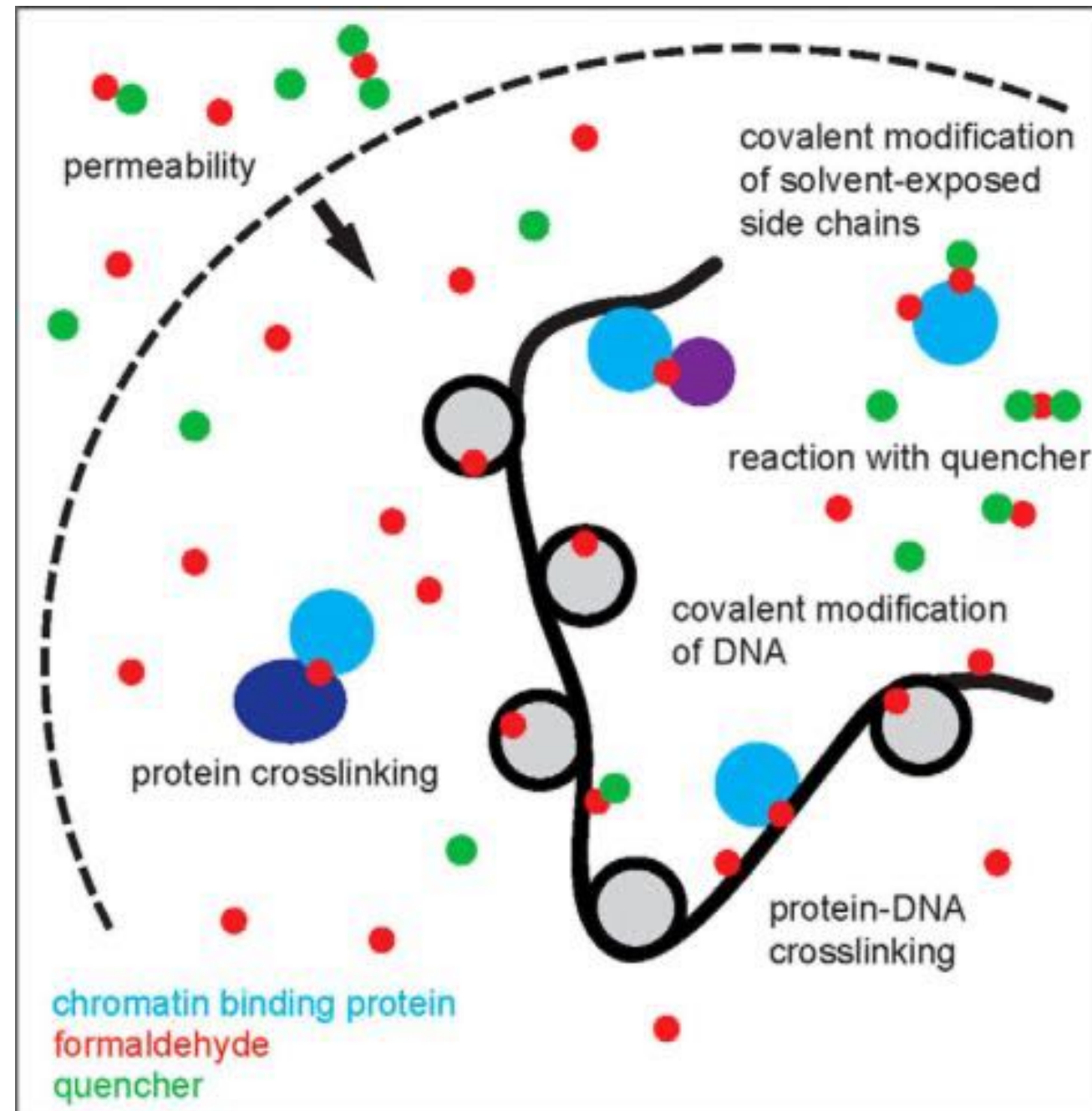
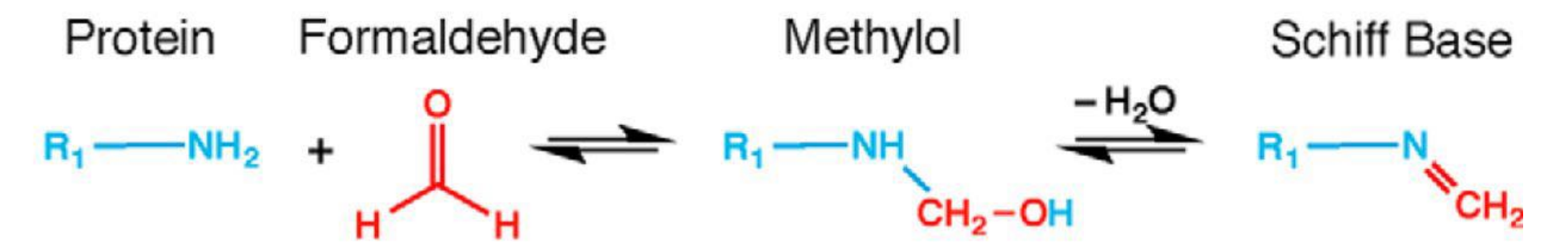


Image from Crosetto,
Nat Rev Genet, 2015

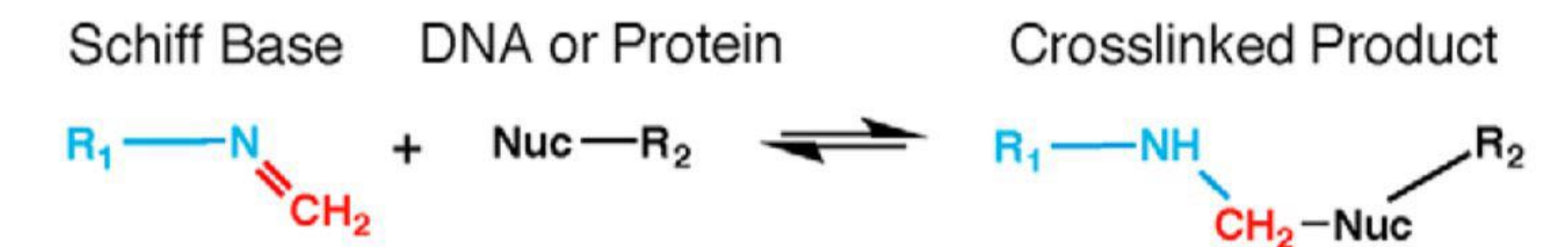
Formaldehyde cross-linking of proteins and DNA



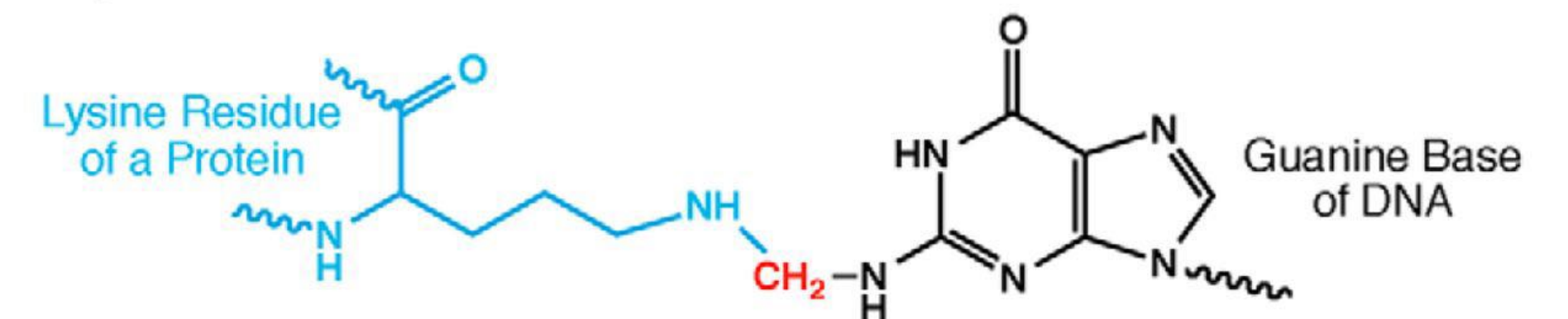
Step 1



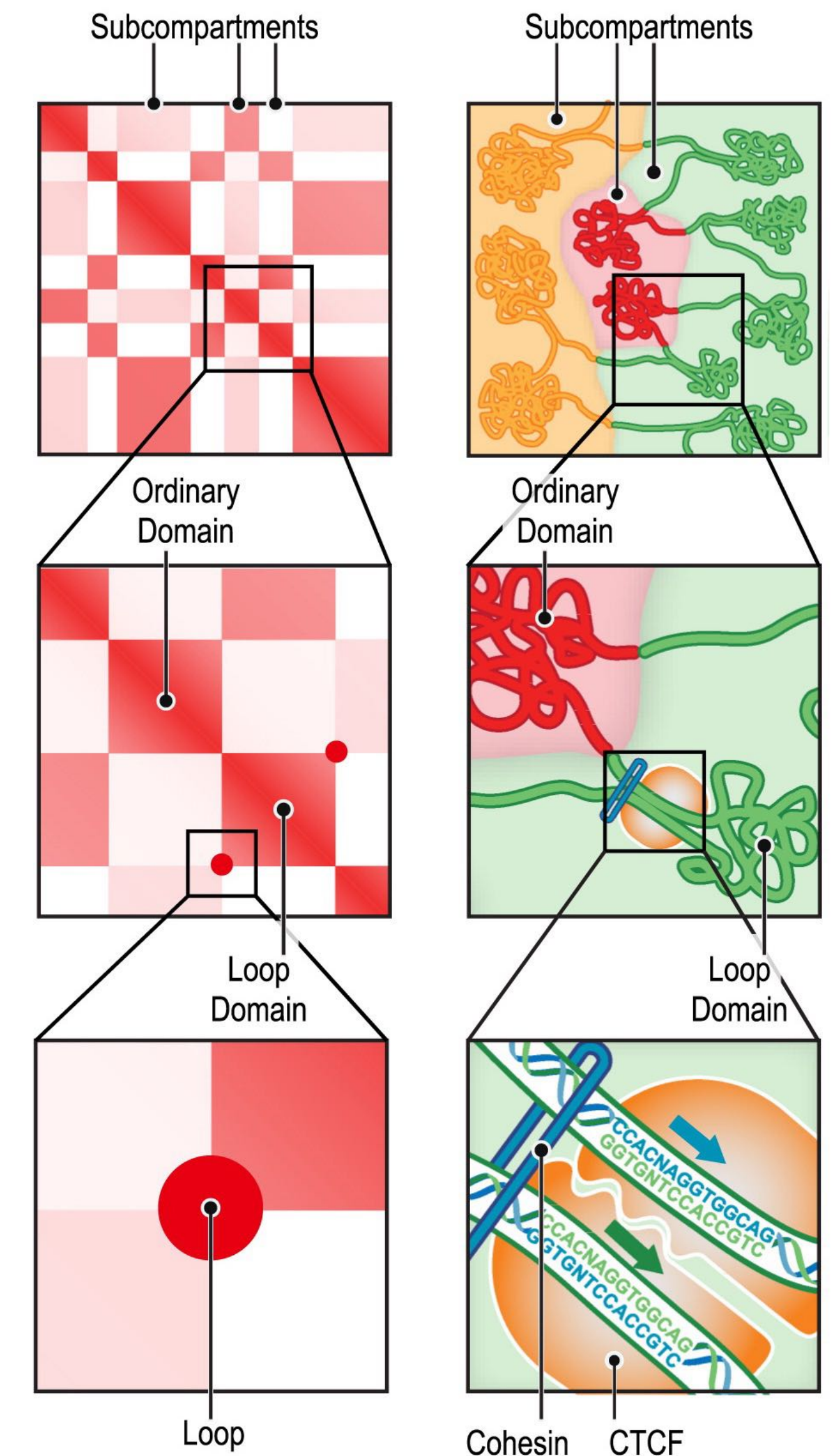
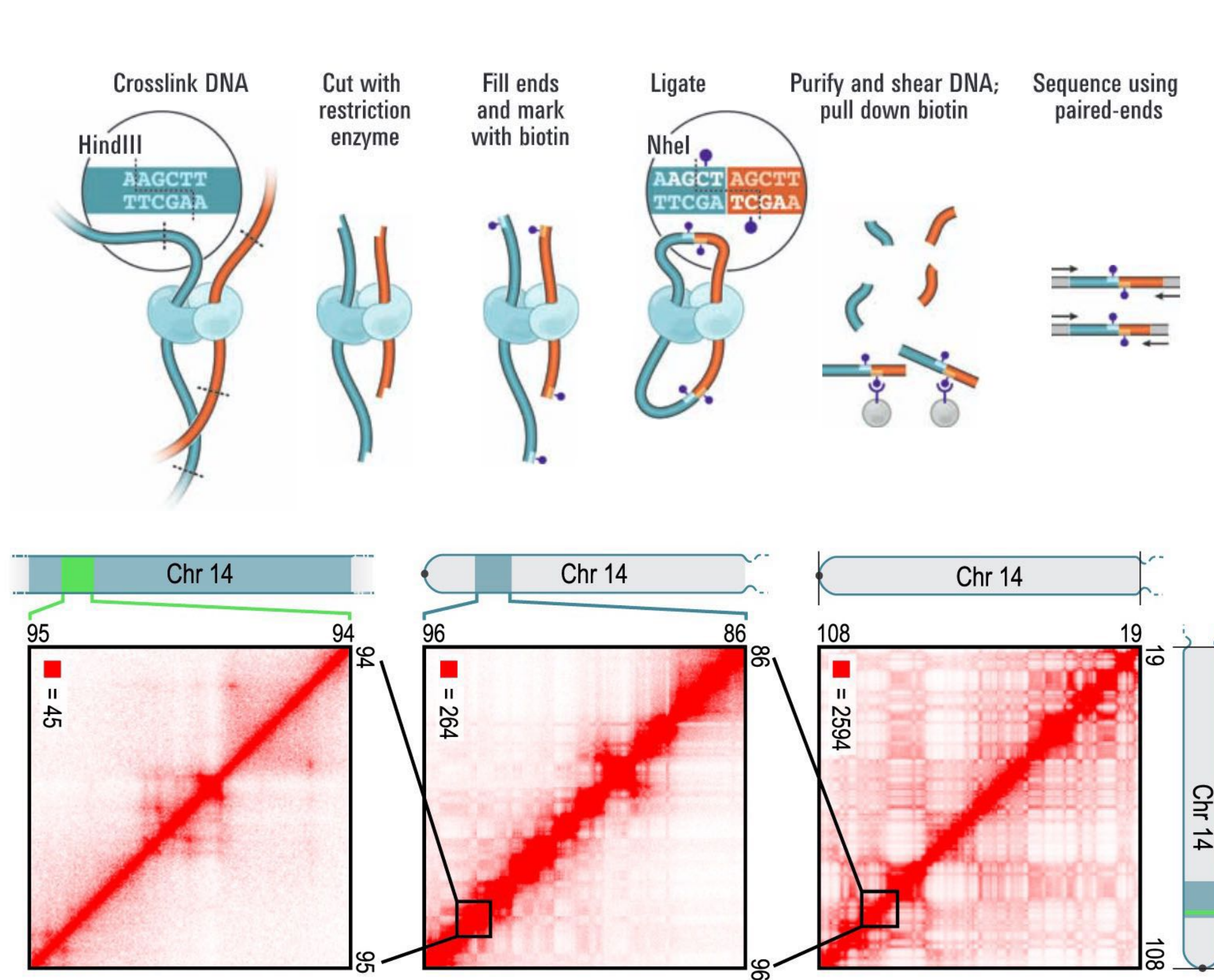
Step 2



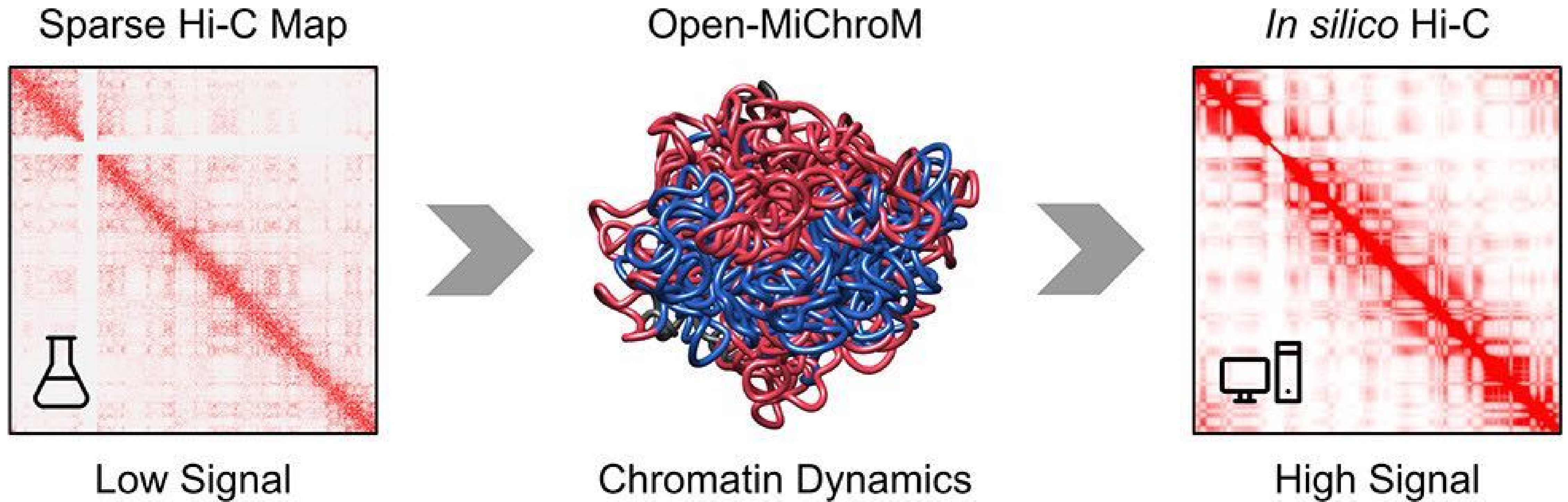
Example Protein-DNA Crosslink



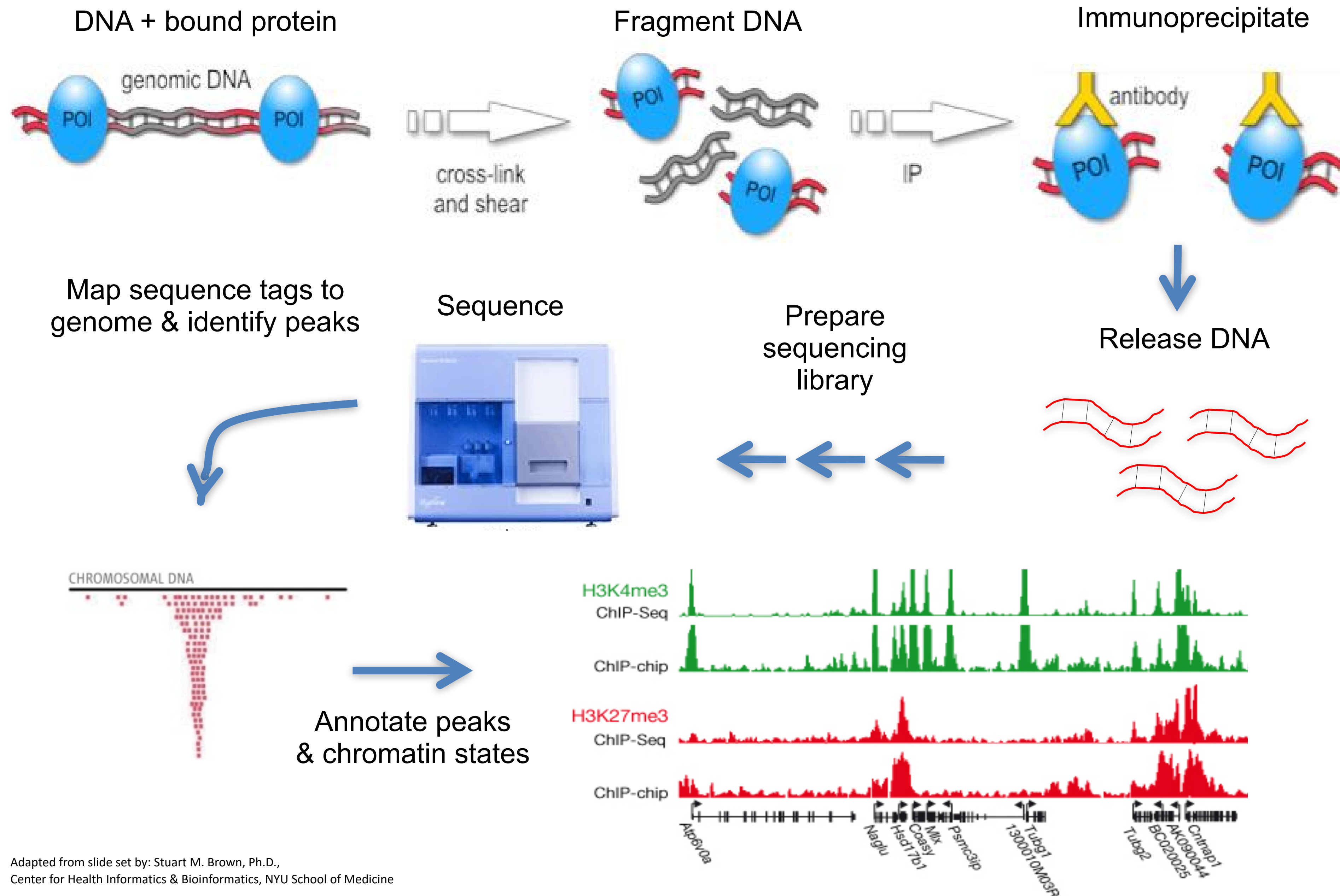
Chromatin Conformation Capture (3C) analysis of chromatin organization *in situ* by cross-linking and sequencing



Single cell genome folding maps from Hi-C



Quantitative information on chromatin binding from ChIP-seq

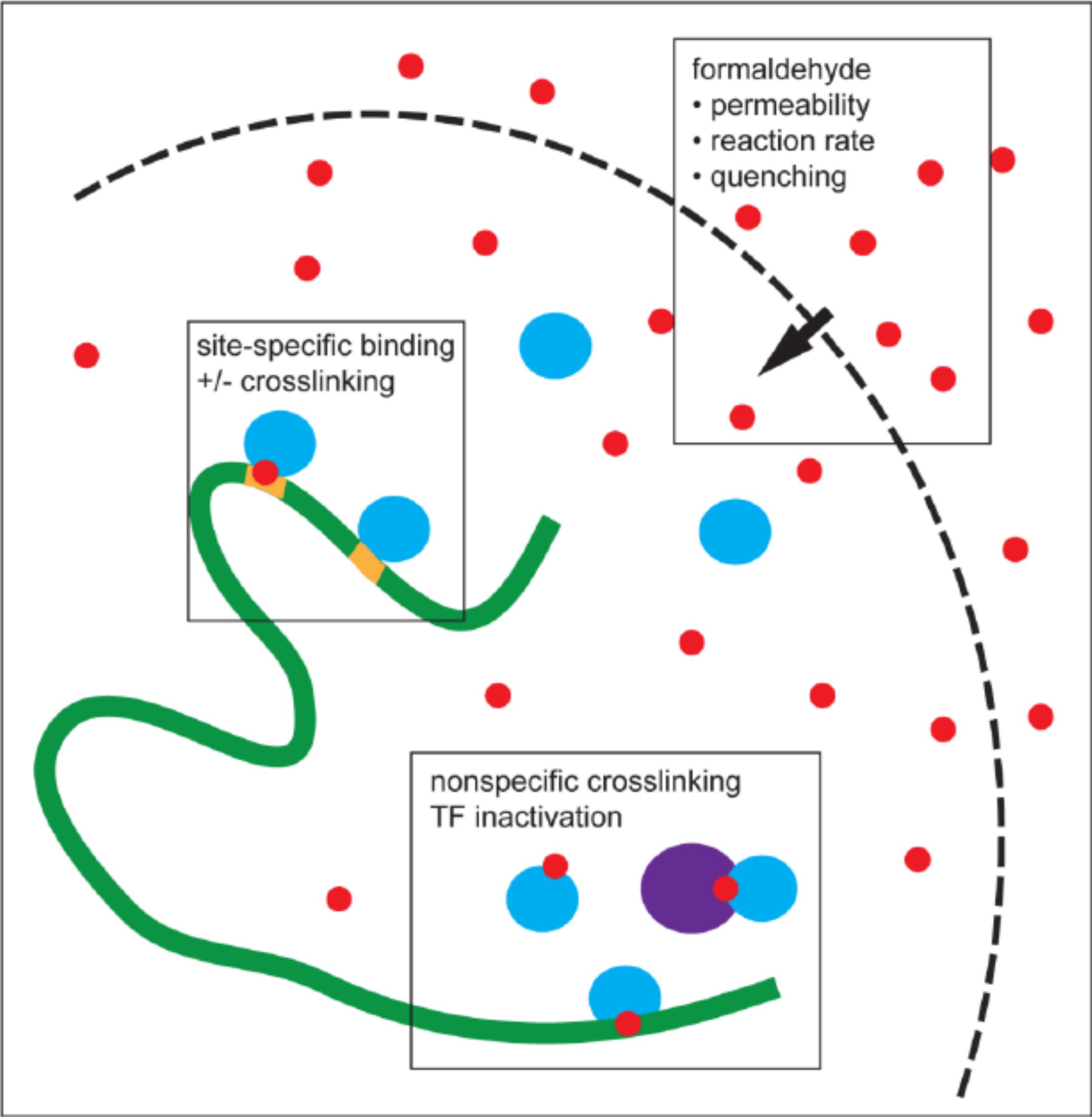


Measuring Chromatin Interaction Dynamics on the Second Time Scale at Single-Copy Genes

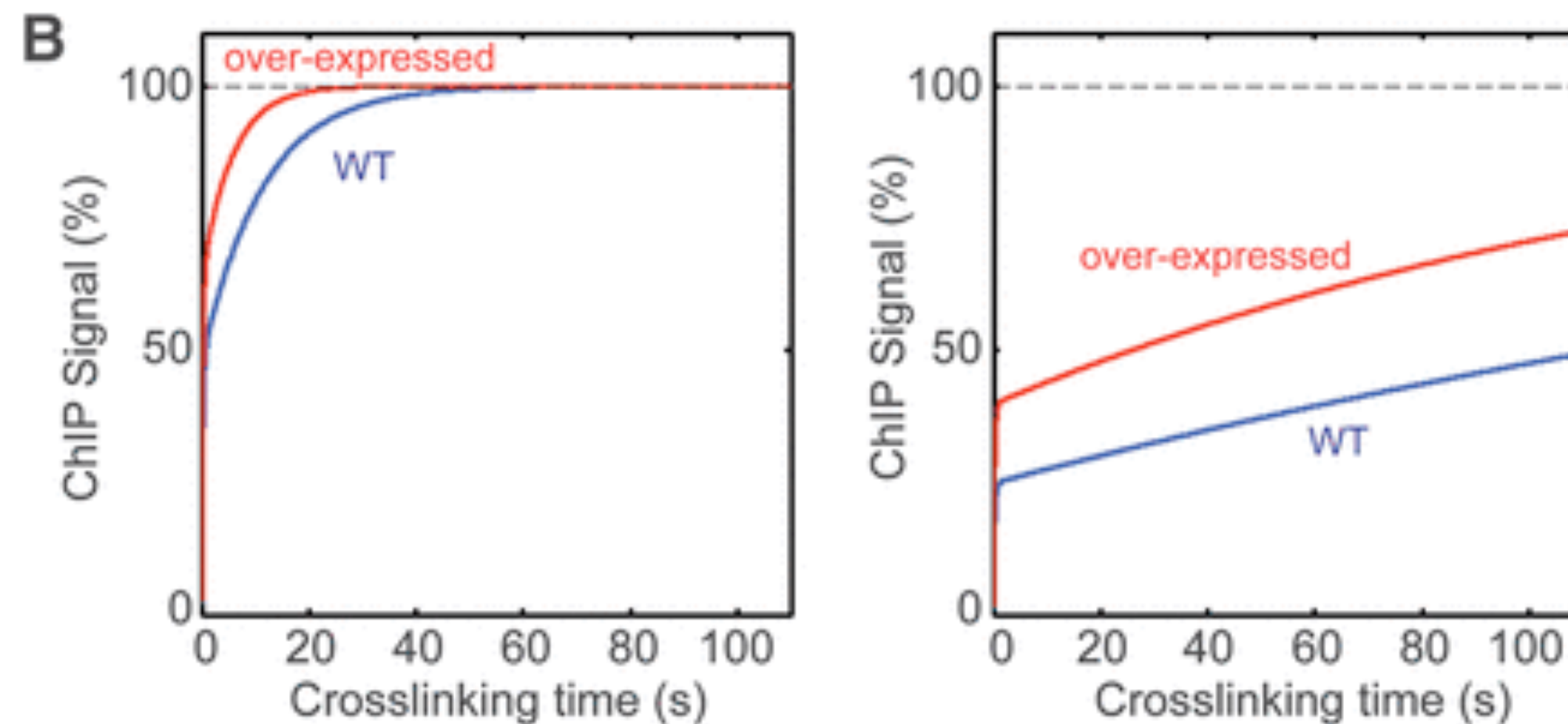
Kunal Poorey,^{1*} Ramya Viswanathan,^{1*} Melissa N. Carver,¹ Tatiana S. Karpova,² Shana M. Cirimotich,¹ James G. McNally,² Stefan Bekiranov,^{1†} David T. Auble^{1†}

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Measure binding kinetics of transcription factors at single binding sites from ChIP-seq analysis by the CLK (cross-linking kinetic) method



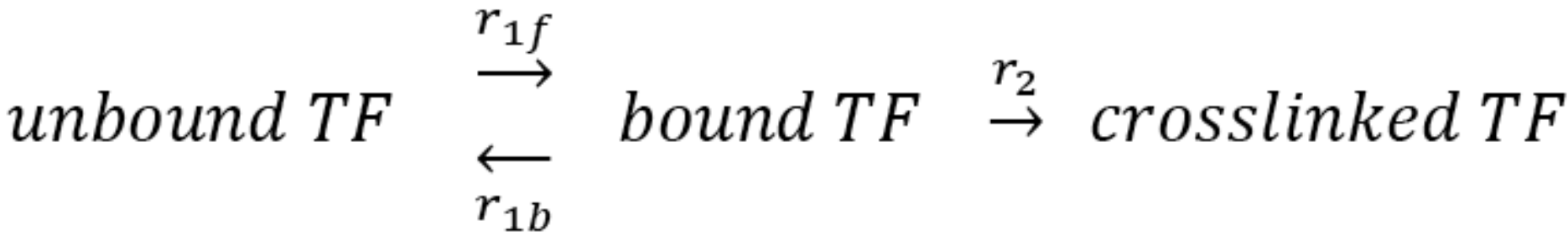
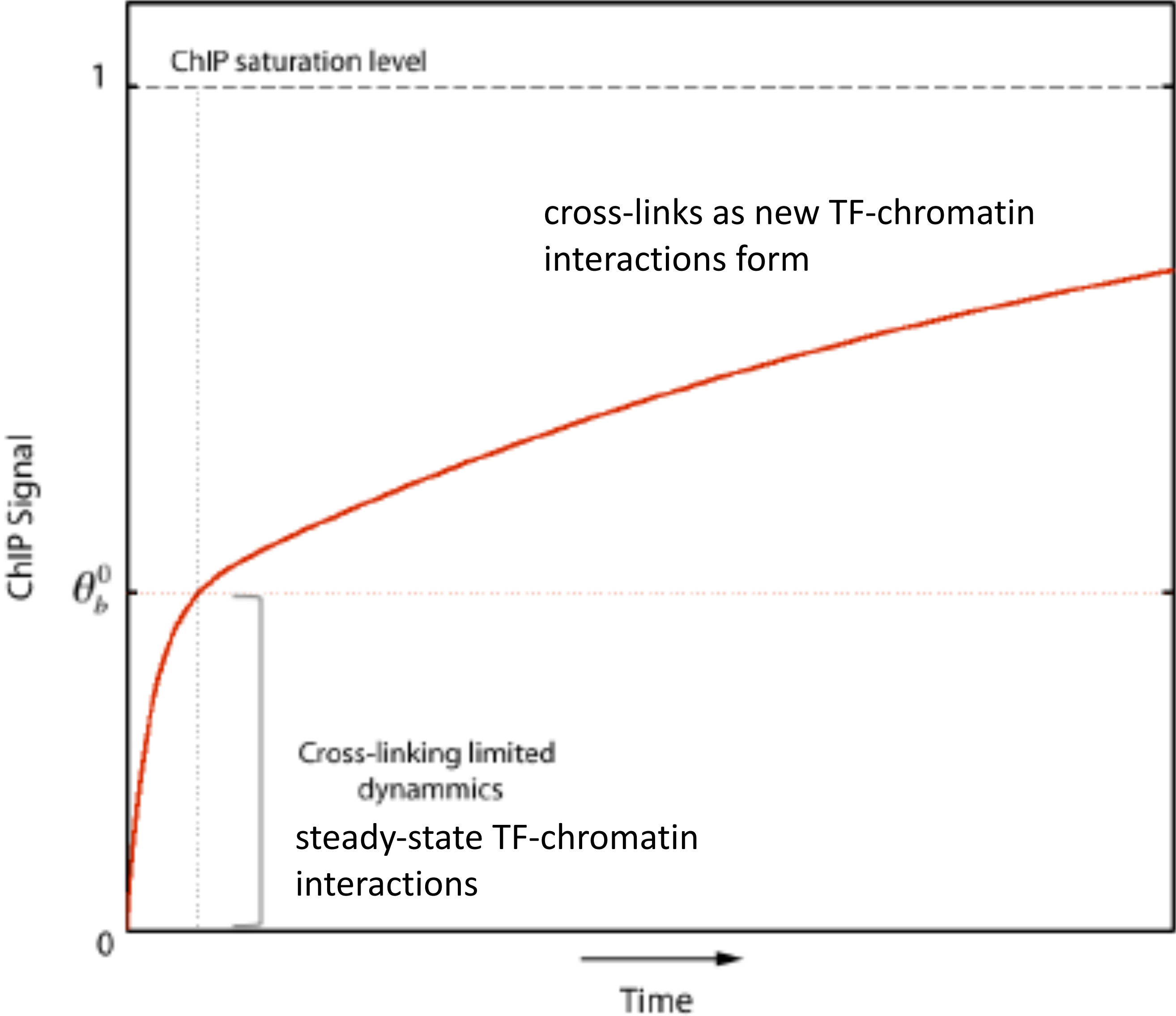
Measuring Chromatin Interaction Dynamics on the Second Time Scale at Single-Copy Genes by ChIP-seq



Assumptions

- TF in excess over binding sites
- TFs are cross-linked to specific and un-specific sites
- unbound TFs are not inactivated by cross-linking
- formaldehyde is not limiting
- ChIP signal is proportional to the number of TF-chromatin interactions

CLK curve



On rate limiting chromatin dynamics

short cross-linking time

$$\theta_{xl}(t) \approx \theta_b^0 (1 - e^{-k_{xl} C_{FH} t})$$

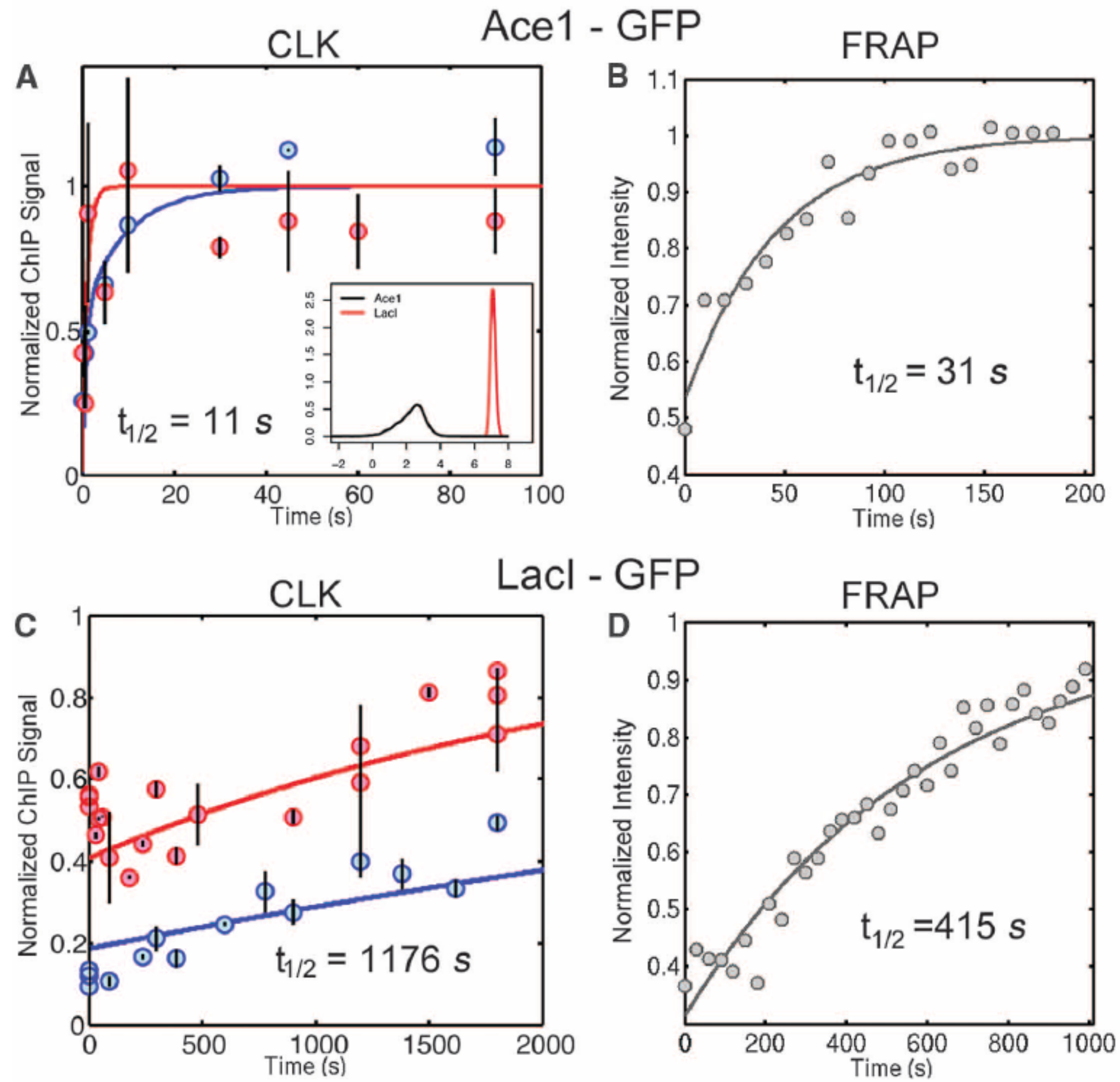
Fraction cross-linked

Fraction bound at $t = 0$

$$\theta_{xl}(t) \approx 1 - (1 - \theta_b^0) e^{-k_a C_{TF} t}$$

TF concentration

Comparison of FRAP and CLK



Ace1 at the
CUP1 gene array

LacI at the
lacO array

From fluorescence microscopy to sequencing and back

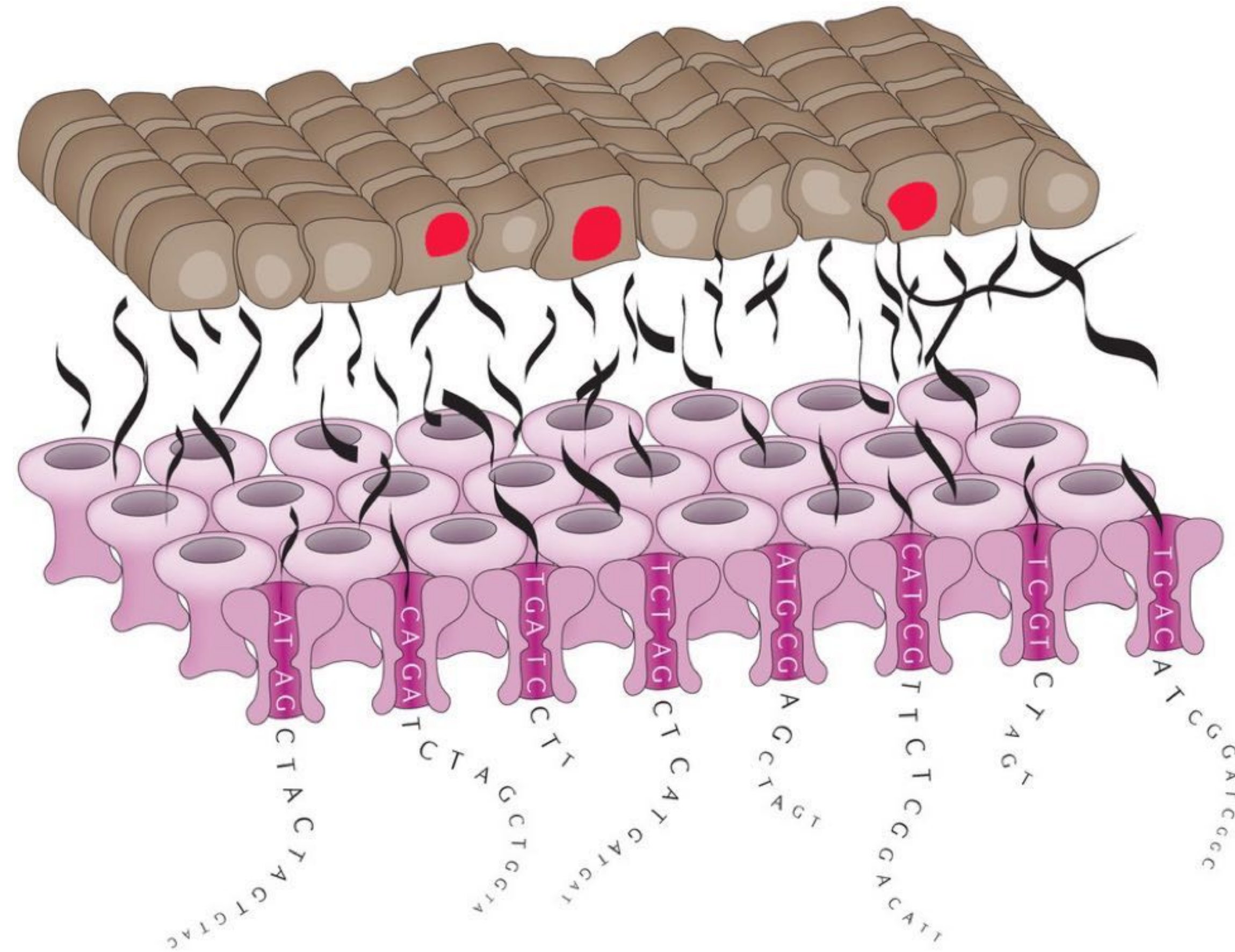
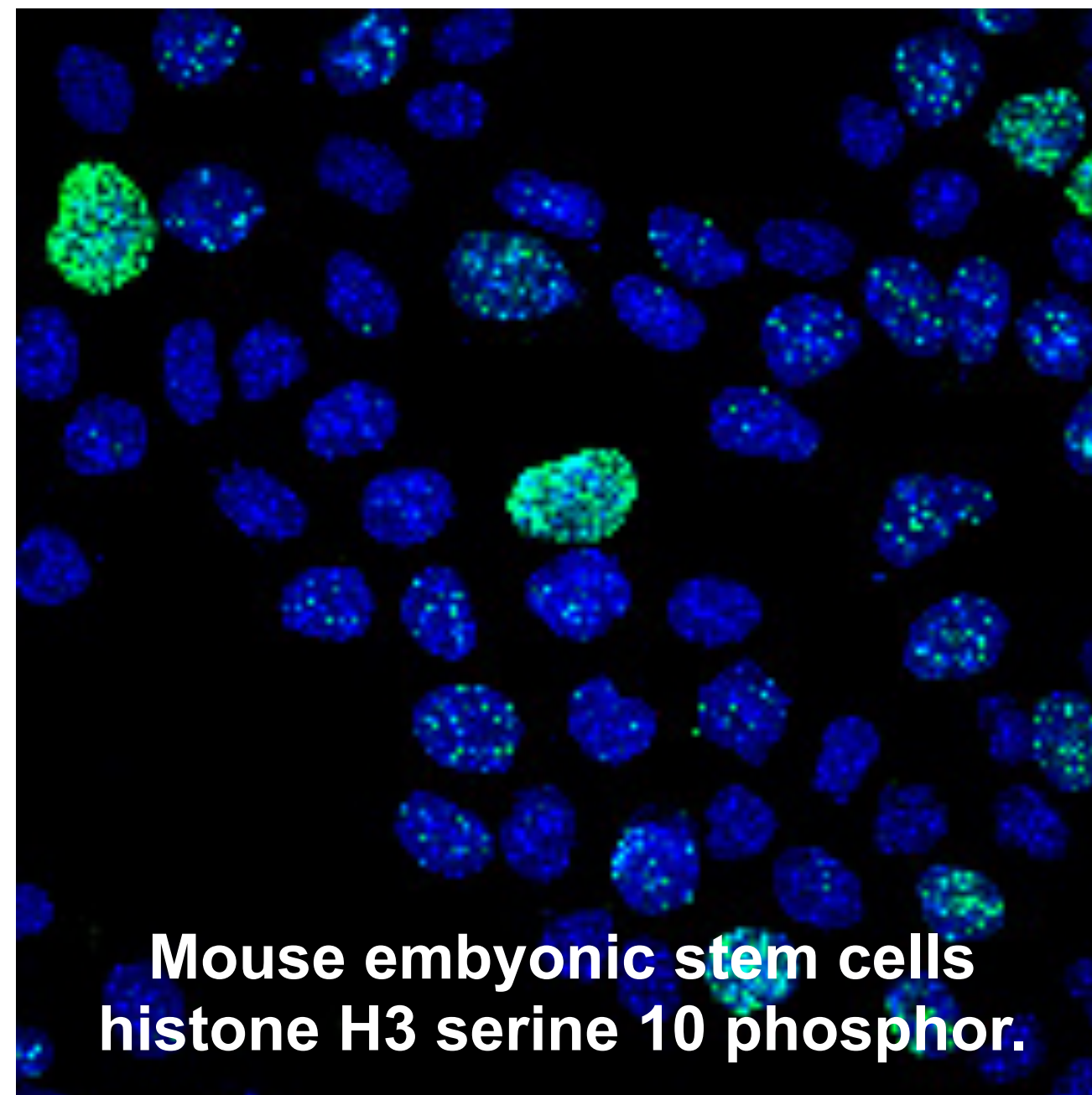


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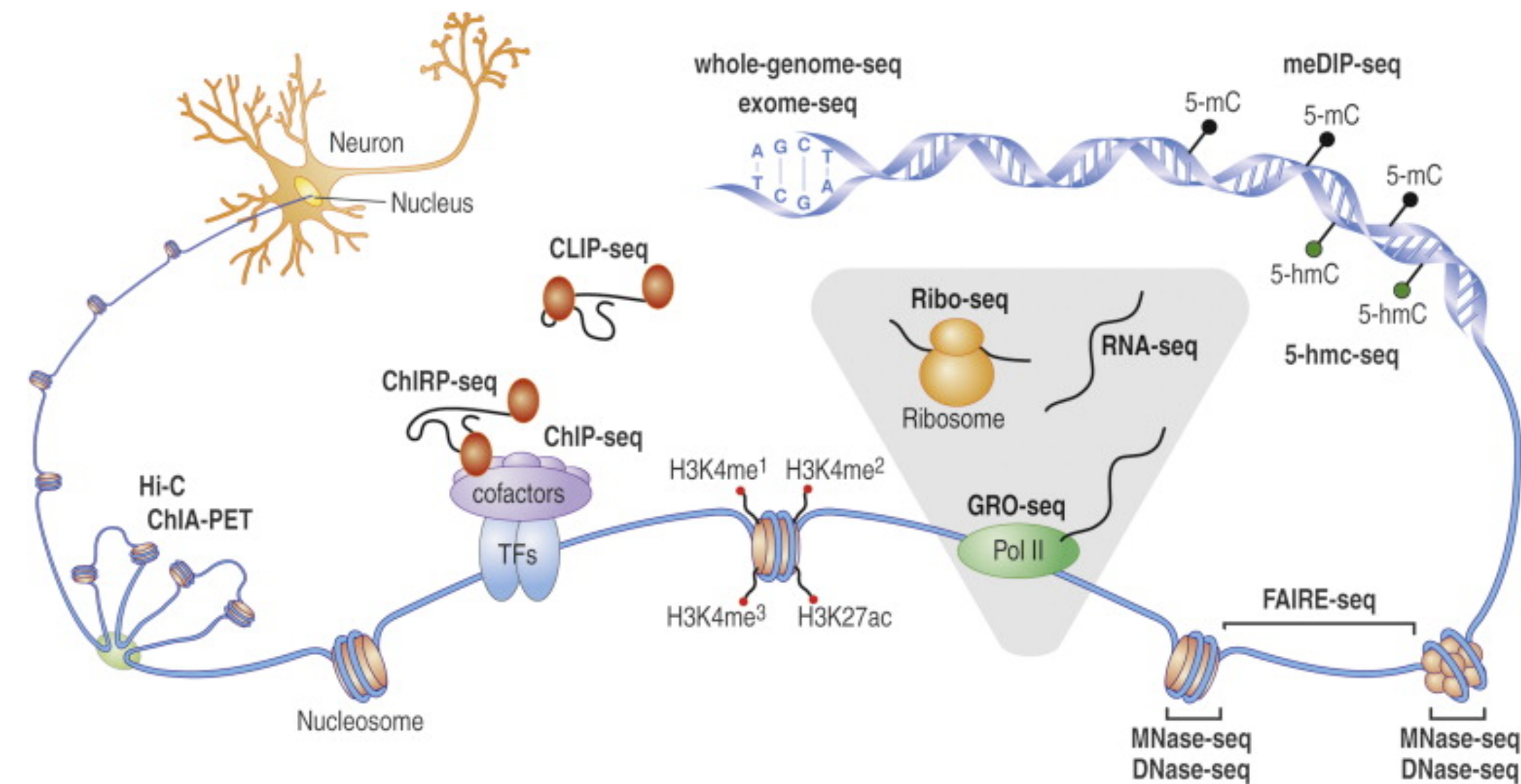
How to integrate microscopy with sequencing?

Fluorescence microscopy



- Single (living) cells
- No molecular resolution
- Lacks genome reference

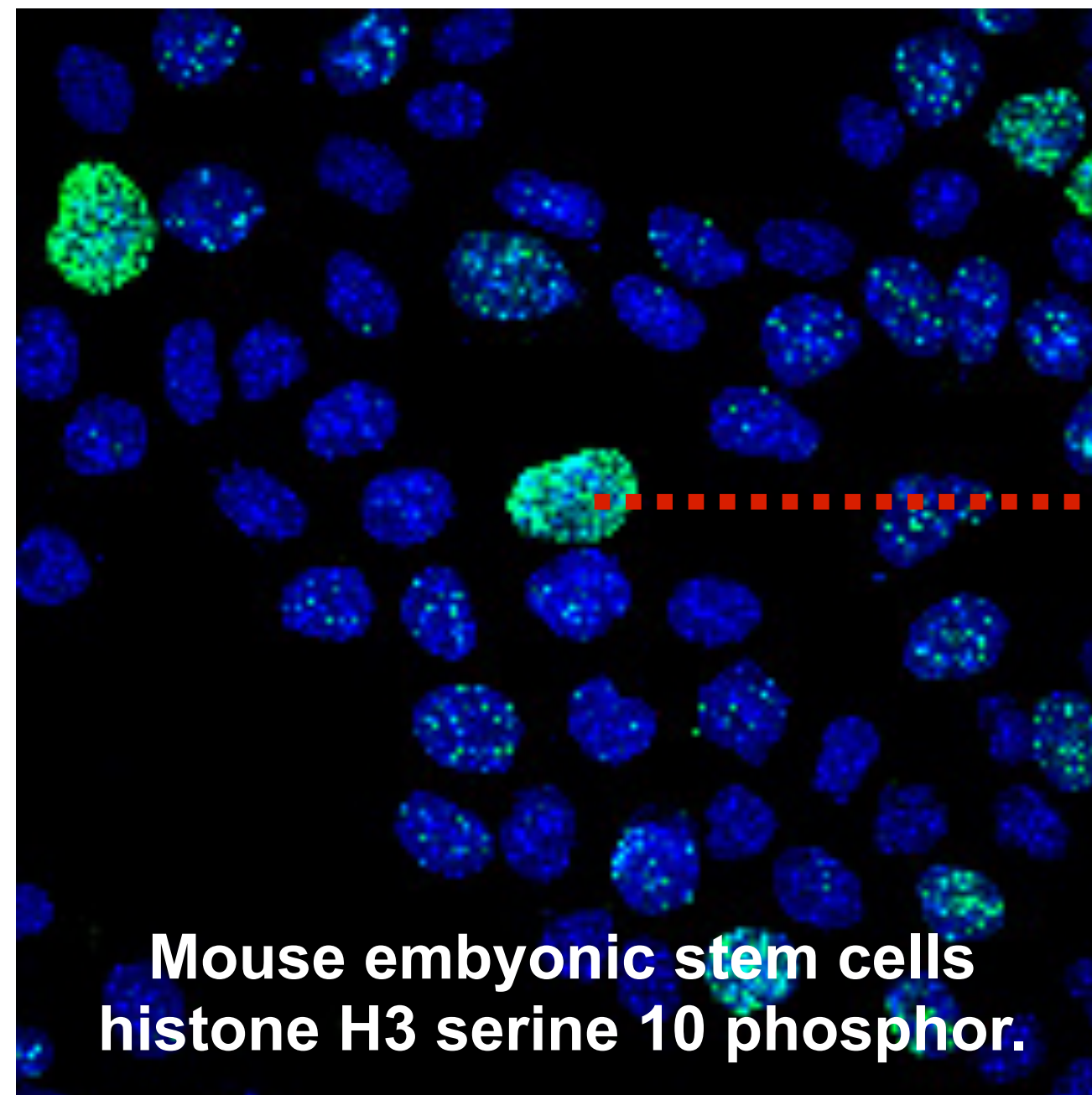
Sequencing



- 10^4 - 10^7 dead cells
- Base pair resolution (0.34 nm)
- Genome sequence reference

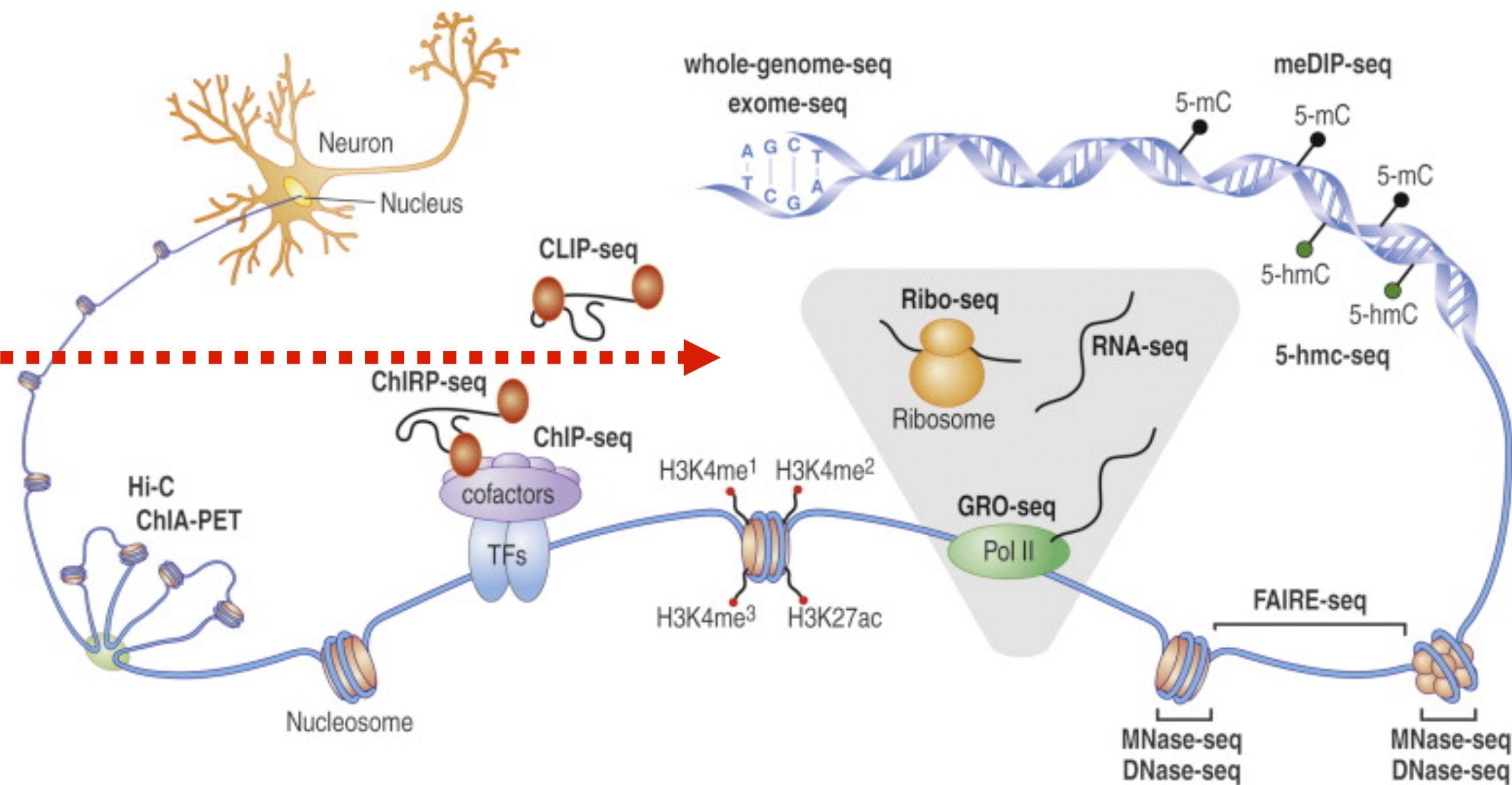
Microscopy meets sequencing

Fluorescence microscopy



- Single (living) cells
- No molecular resolution
- Lacks genome reference

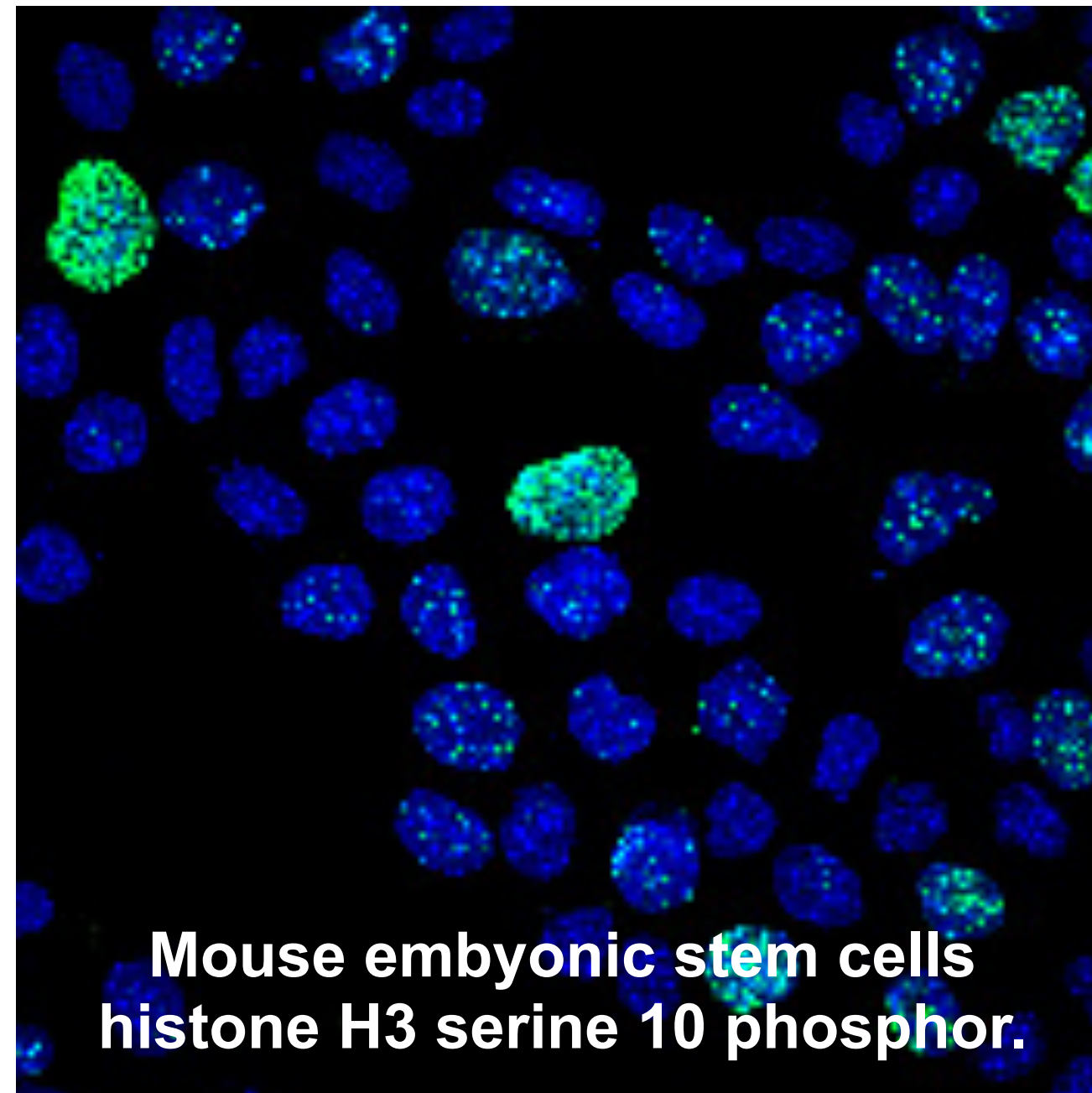
Sequencing



- **Single cells**
- Base pair resolution (0.34 nm)
- Genome sequence reference

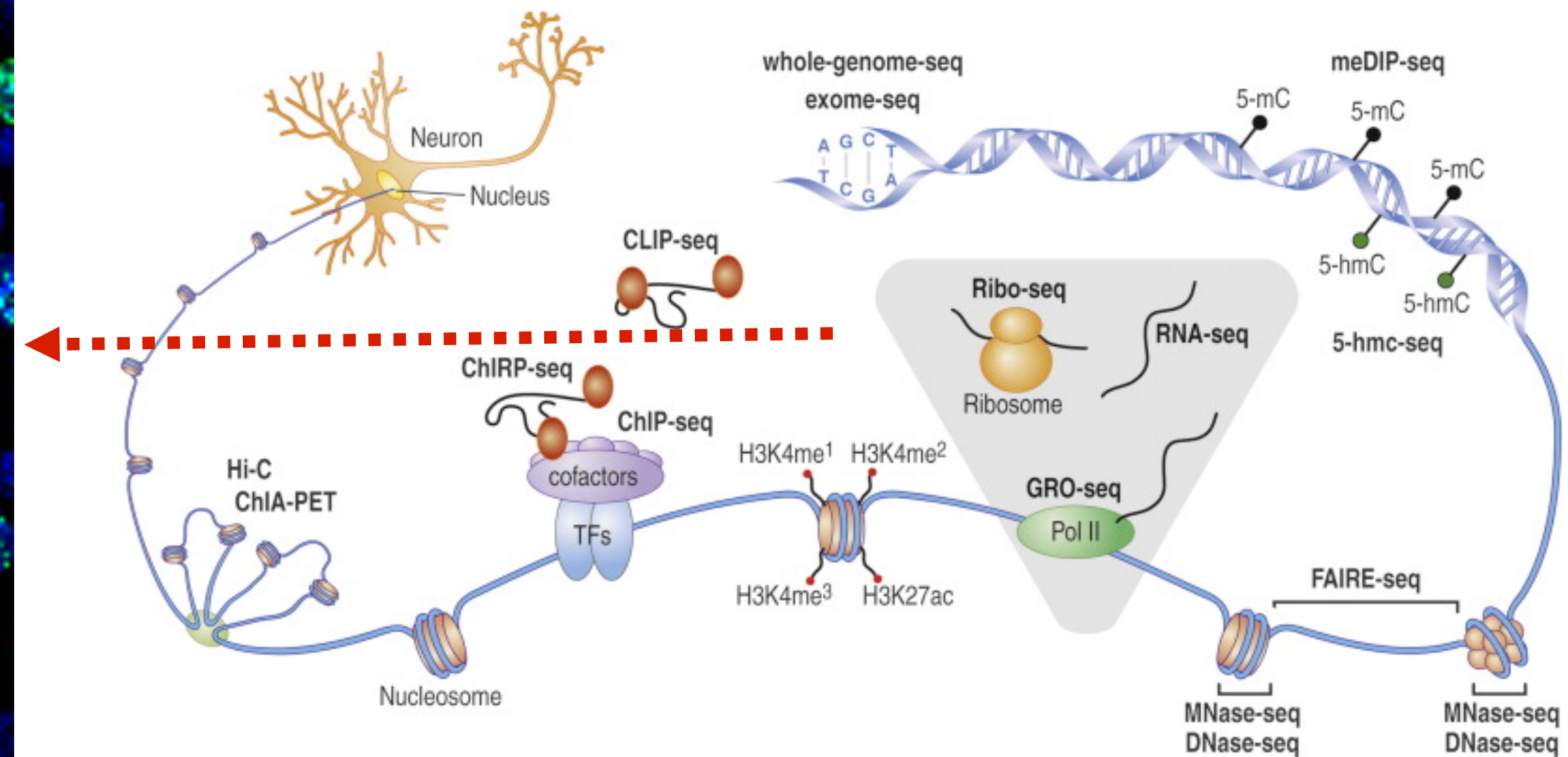
Microscopy meets sequencing

Fluorescence microscopy



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Sequencing

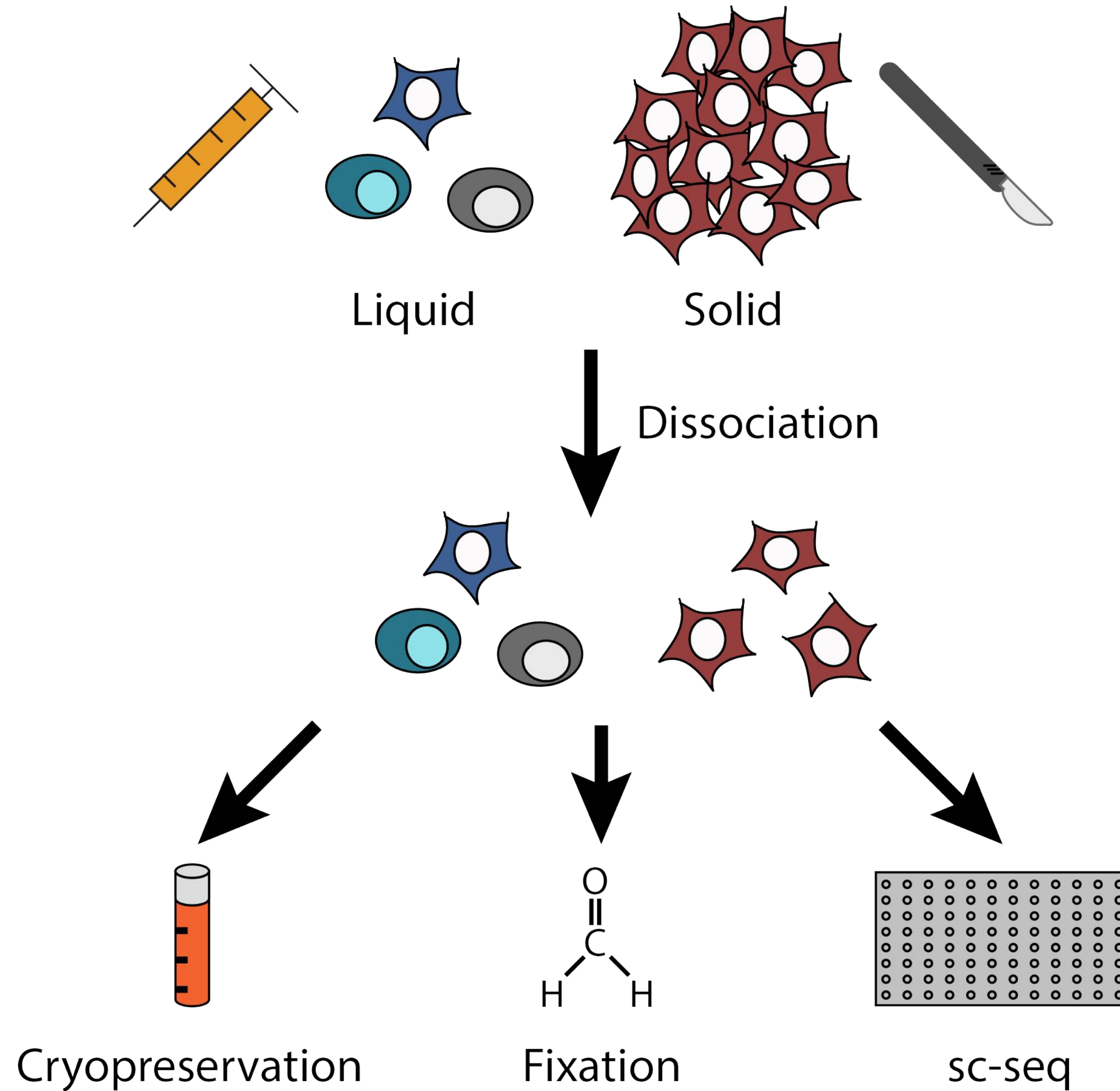


- **Single cells**
- Base pair resolution (0.34 nm)
- Genome sequence reference

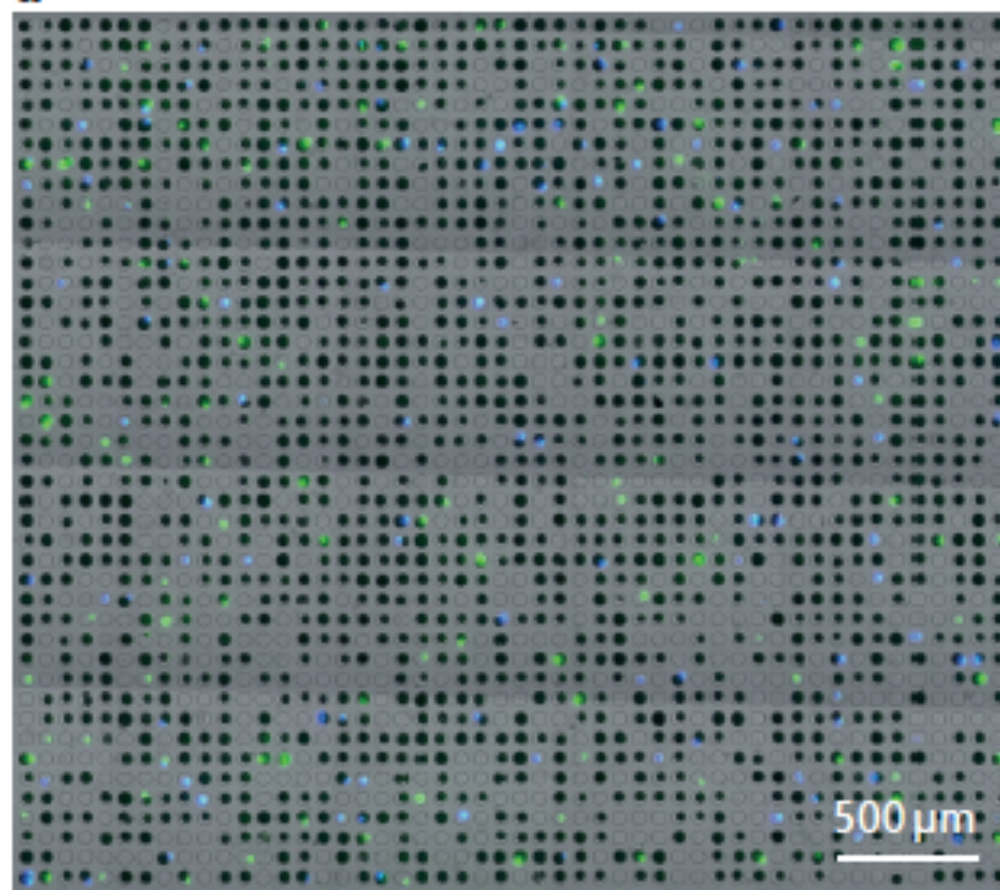
Single-cell sequencing based readouts

	Isolation: stability versus sensitivity of molecules during cell isolation and processing	Recovery: yield and efficiency of sequencing libraries	Scalability: automation-readiness for many single cells	Scalability: sequencing efficiency	Refs
<i>Single-cell epigenomics approaches</i>					
DNA methylation	Very stable	4–20% of each haplotype; low error rate	Good	Low (~20 million reads typically required per cell, although fewer reads can be used for lower-depth screens) = 10 cells / lane	41,42
Histone modifications	Reasonable	~5% for H3K4me3 50% at validated sites	Good	~100,000 reads per cell	Rotem 2015 Nat Biotechnol
Chromatin accessibility and nucleosome positioning	Sensitive	500–70,000 tags per cell	Good (for example, using combinatorial labelling)	High (500,000–1,500,000 reads typically required per cell) = 100 cells / lane	64,65
Chromosomal contacts	Sensitive (fixation required)	10,000–100,000 ligations; <5% are spurious ligations	Medium (material is fixed)	High (1 million reads typically required per cell)	86
<i>Other single-cell approaches</i>					
RNA sequencing (RNA-seq)	Ultra-sensitive (RNA degradation during cell acquisition)	1–20% of mRNAs (but information is quantitative)	Very good (using fluorescence-activated cell sorting or microfluidic droplets)	High (50,000–1,000,000 reads typically required per cell) = 100-5000 cells / lane	96,113
Genomics	Ultra stable	40–84%	Good	Low for whole-genome analysis at single-nucleotide resolution (200 million reads typically required per cell); higher for copy number-variation analysis, as fewer reads are required	94
Schwartzman & Tanay 2015, Nat Rev Genet 16, 716.					

Sample processing for single cell sequencing

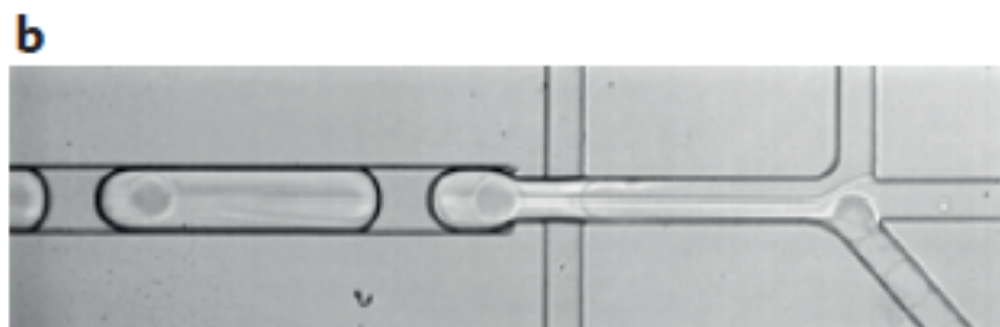


Capturing single cells



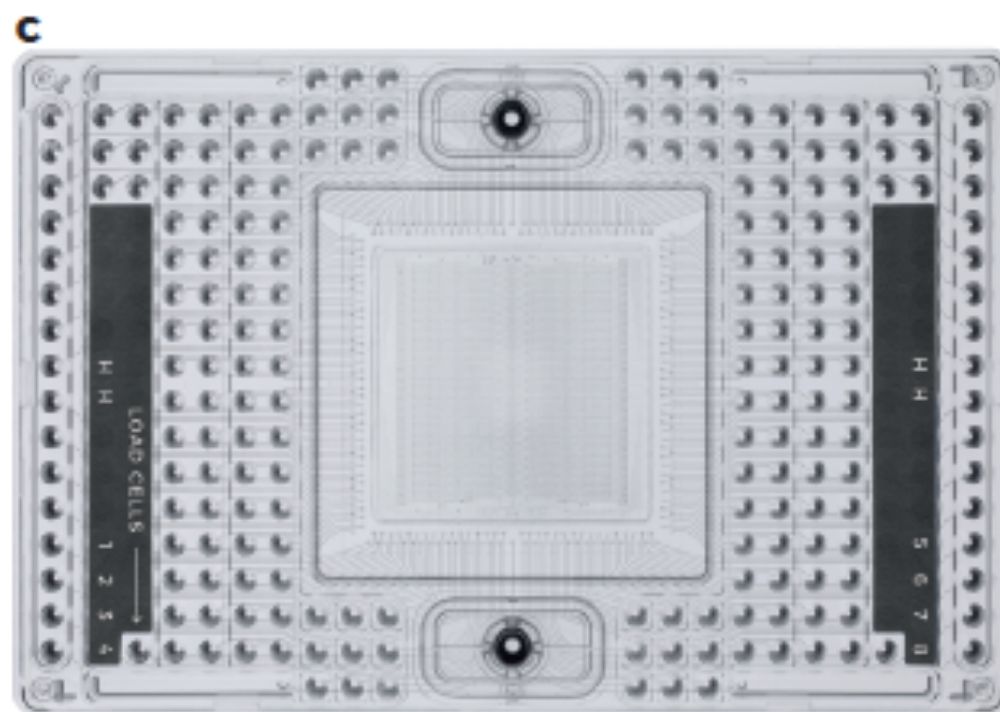
Microwells

FACS based
commercial systems (dispenser)



Droplets

emulsion pumps
commercial systems (drop-seq)



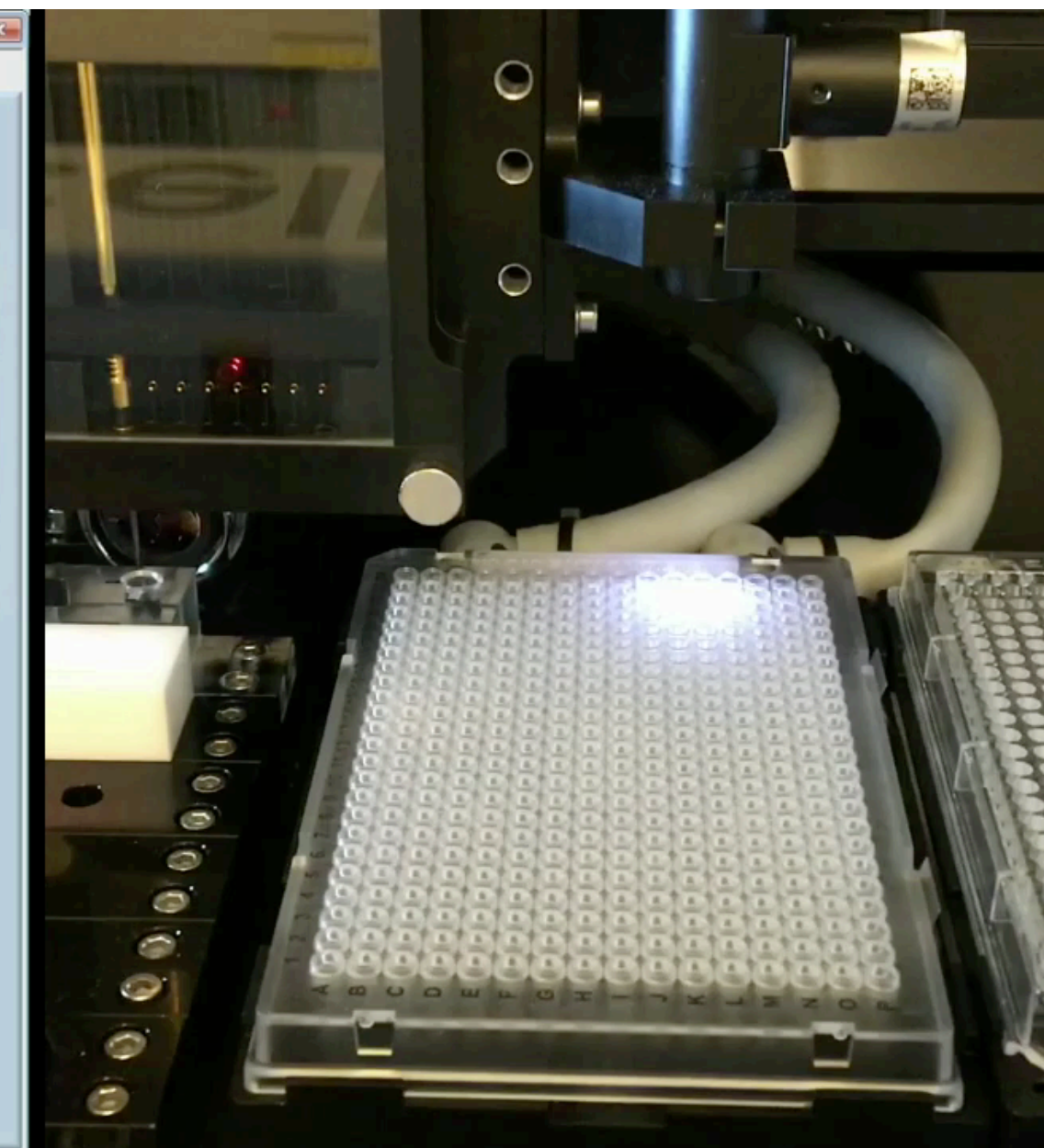
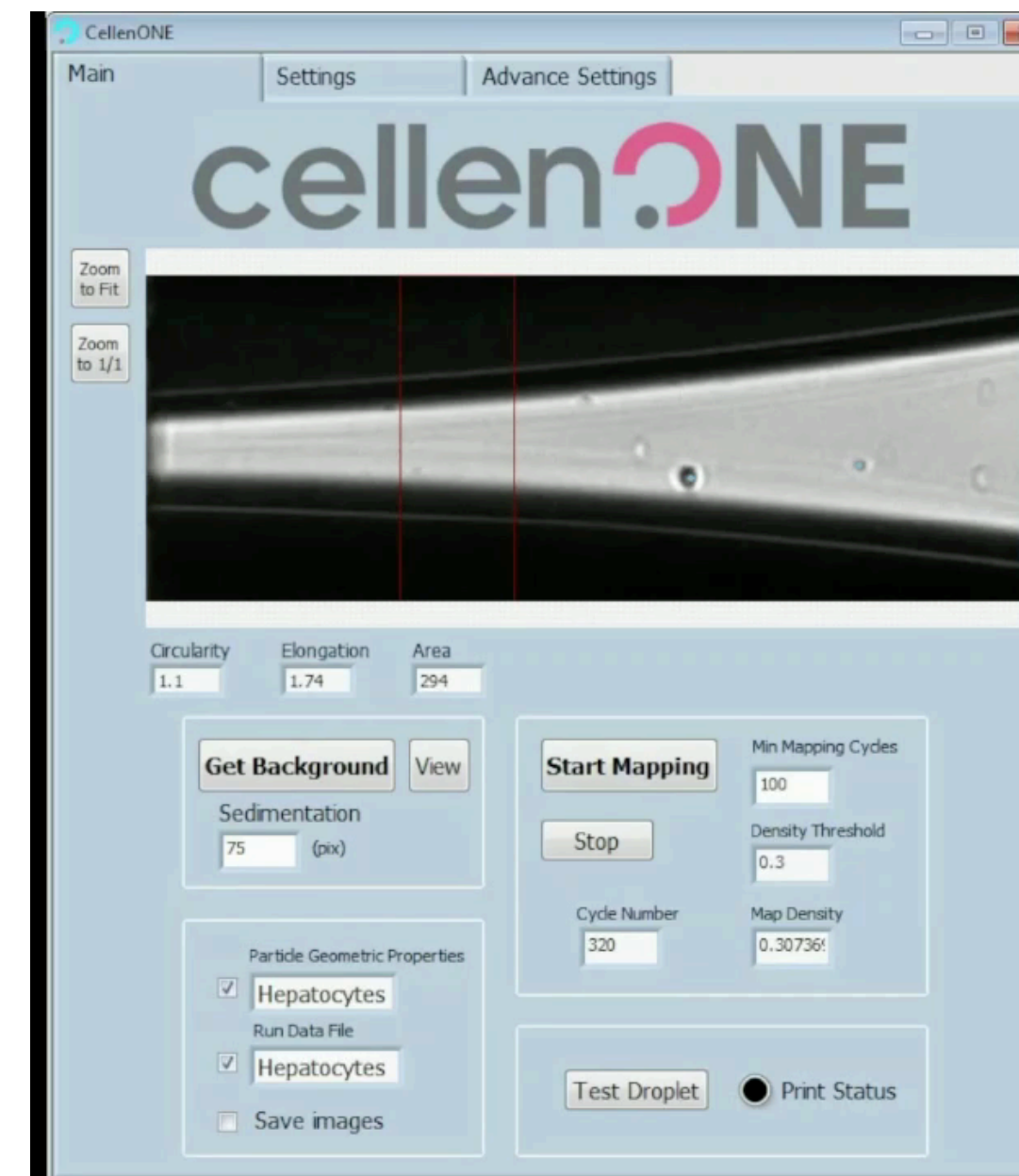
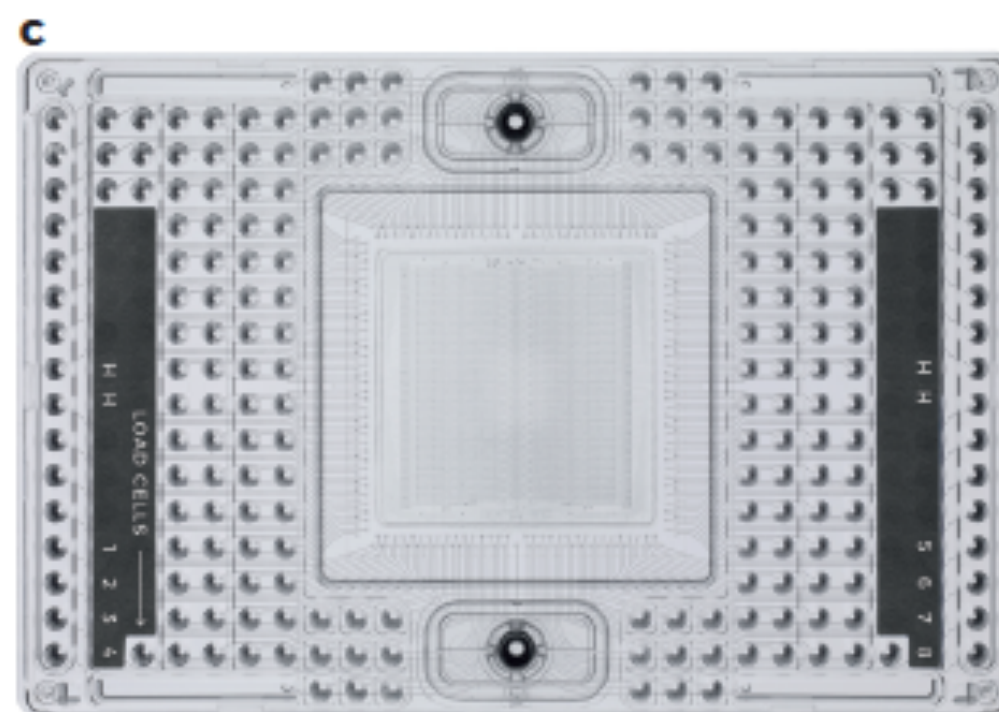
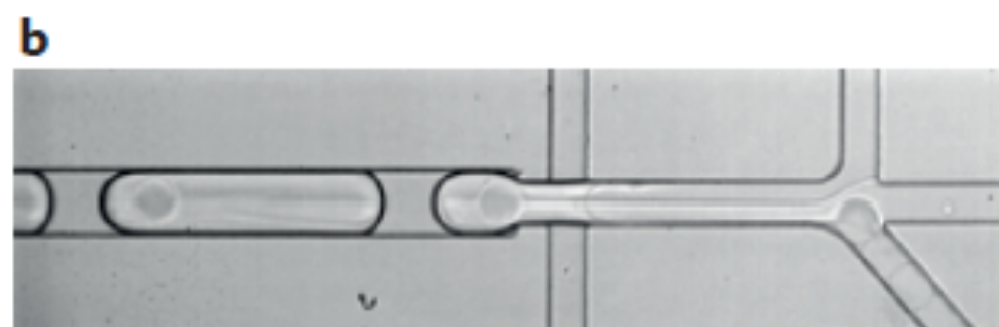
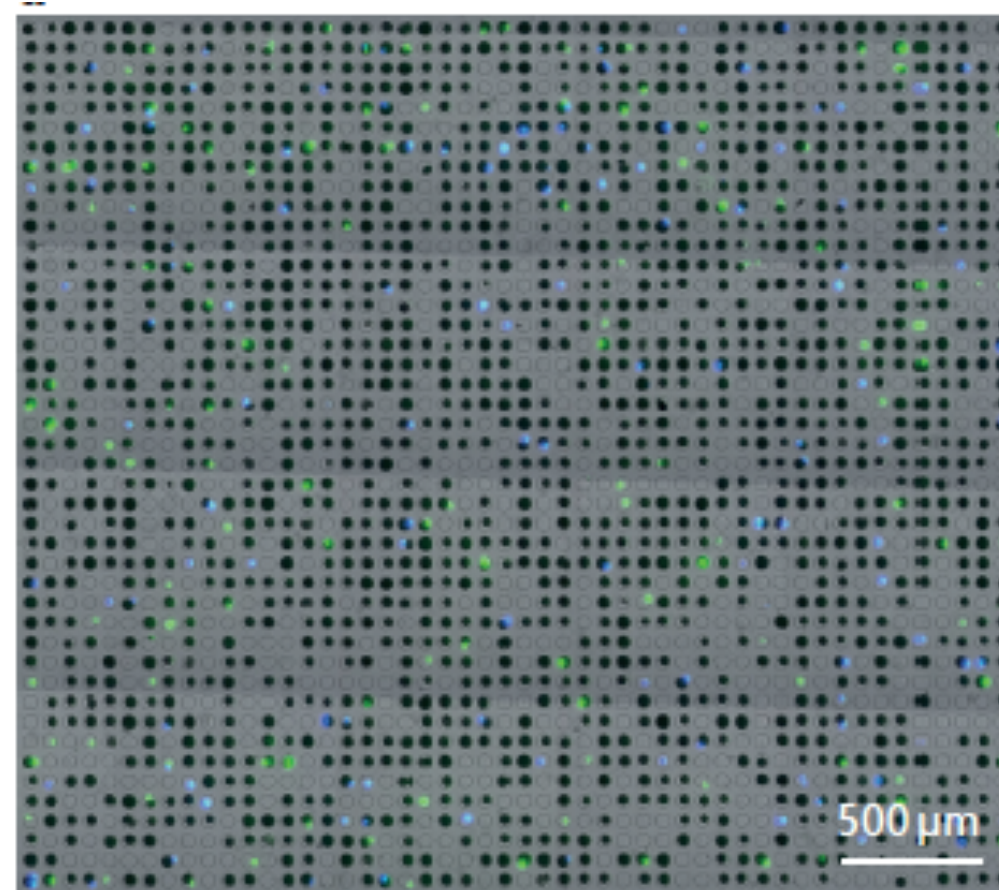
Microfluidics

cell capture
commercial systems (nano traps)

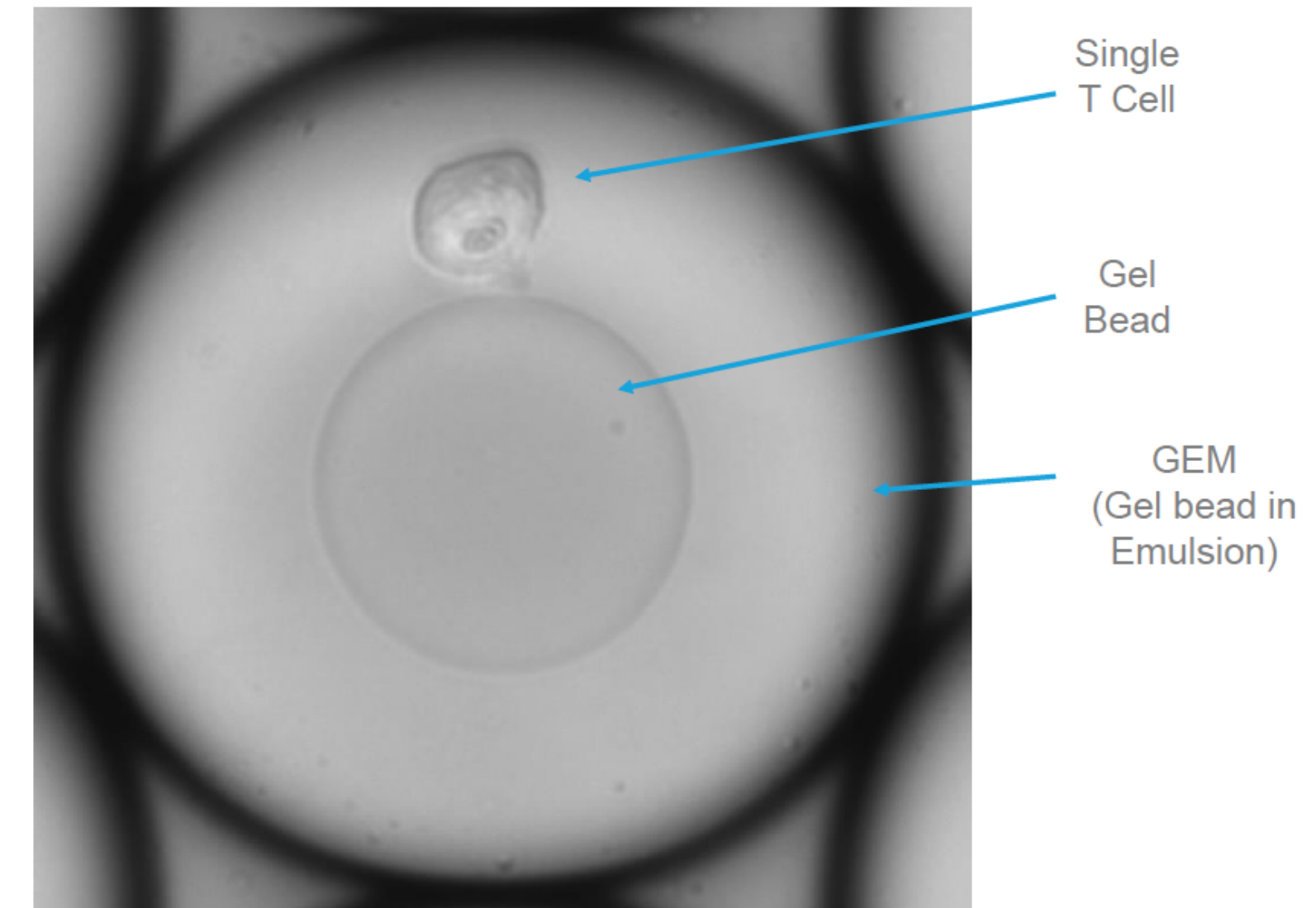
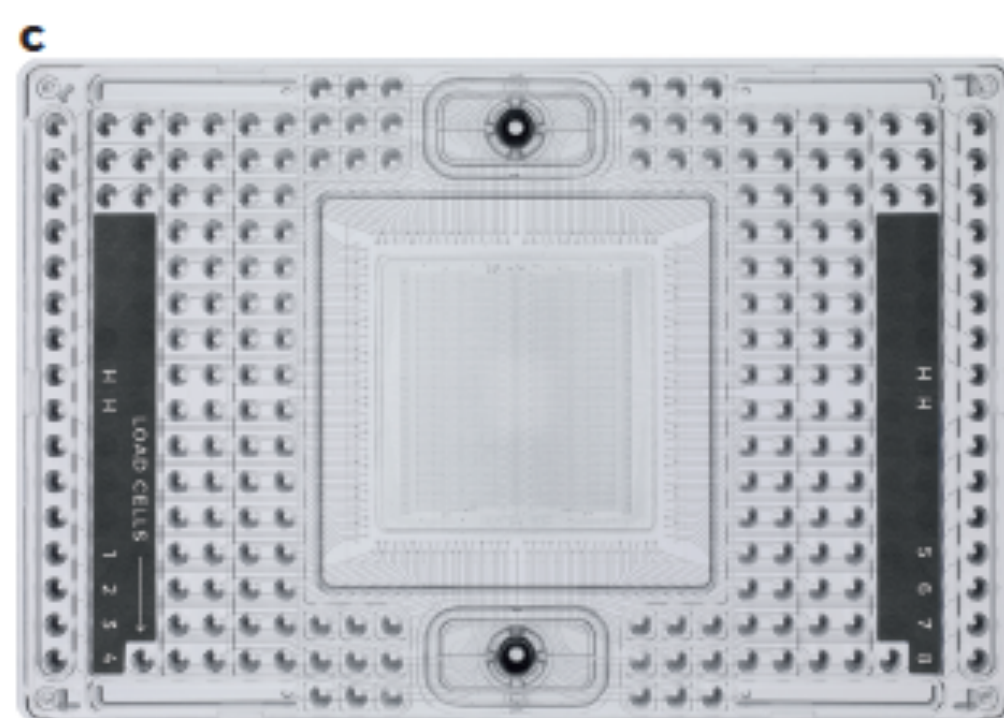
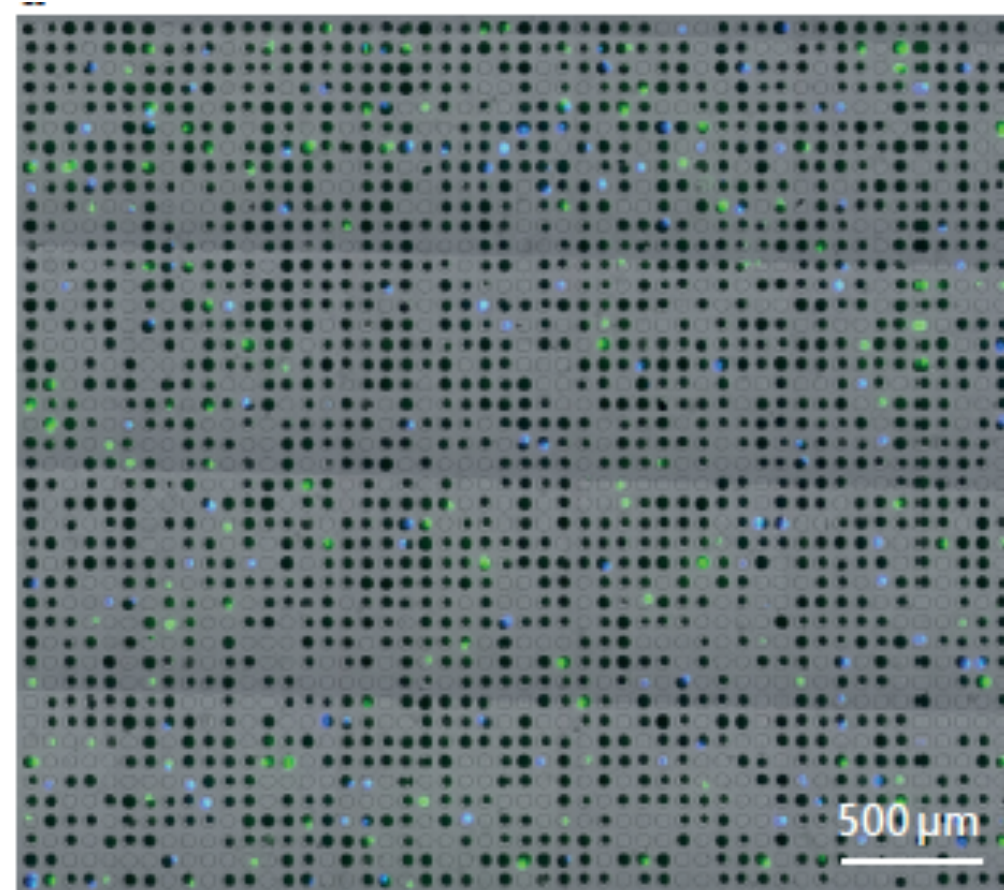
SINGLE CELL SEQUENCING - CAPTURING

Microwells

FACS based
commercial systems (dispenser)



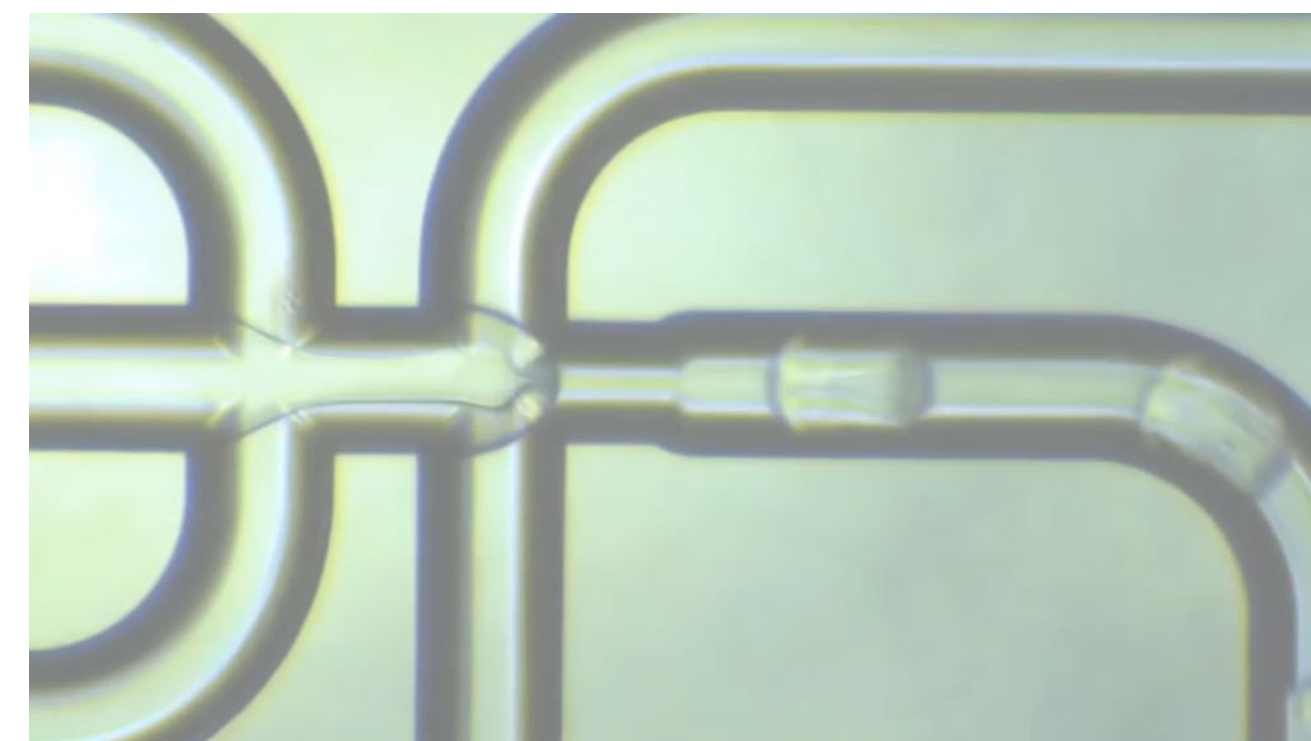
SINGLE CELL SEQUENCING - CAPTURING



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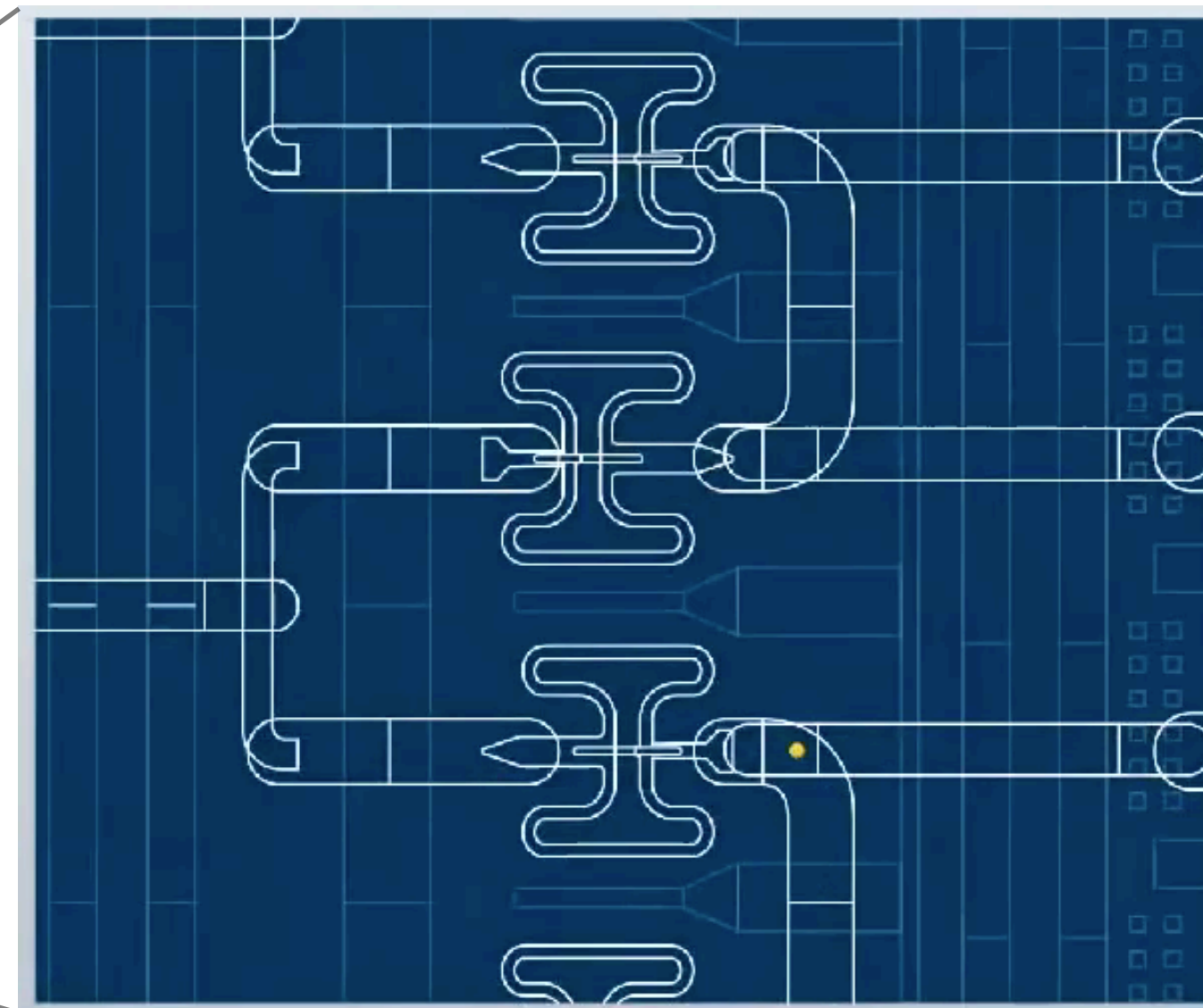
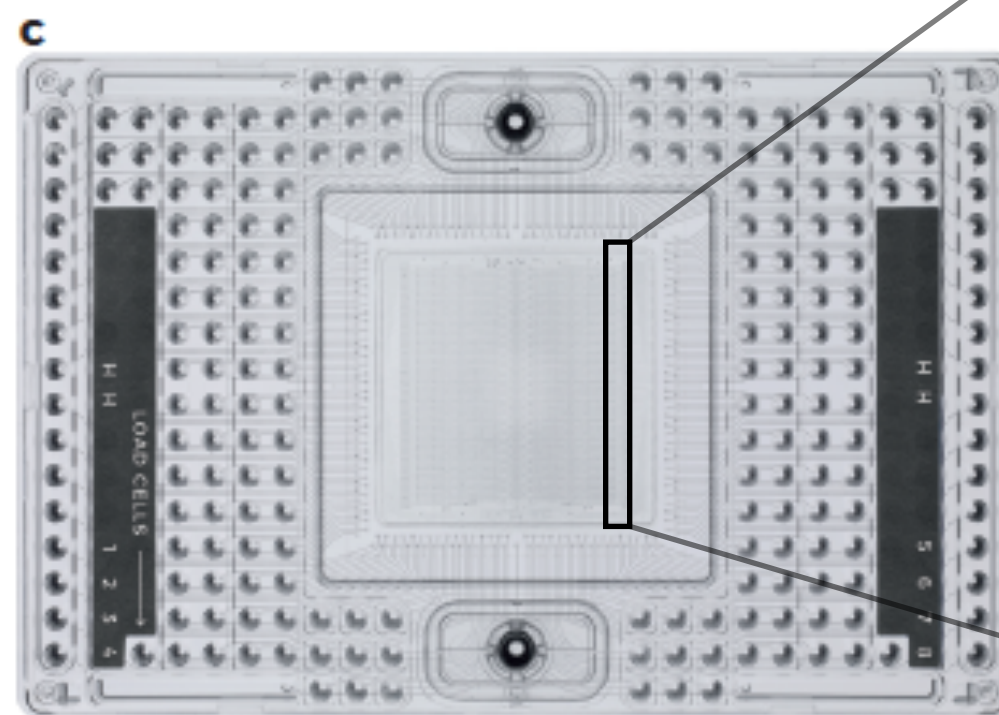
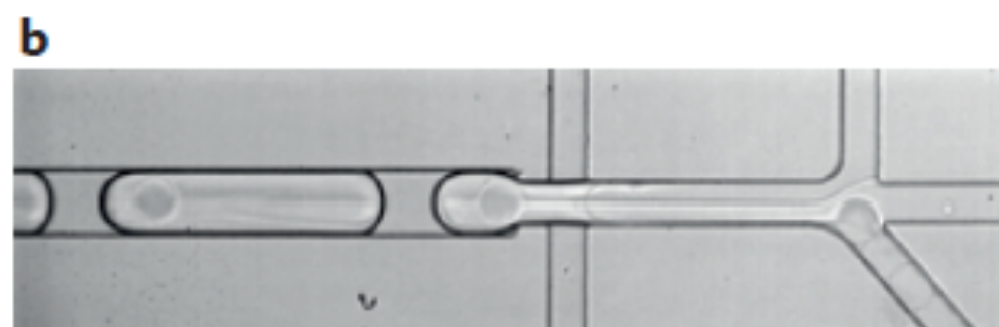
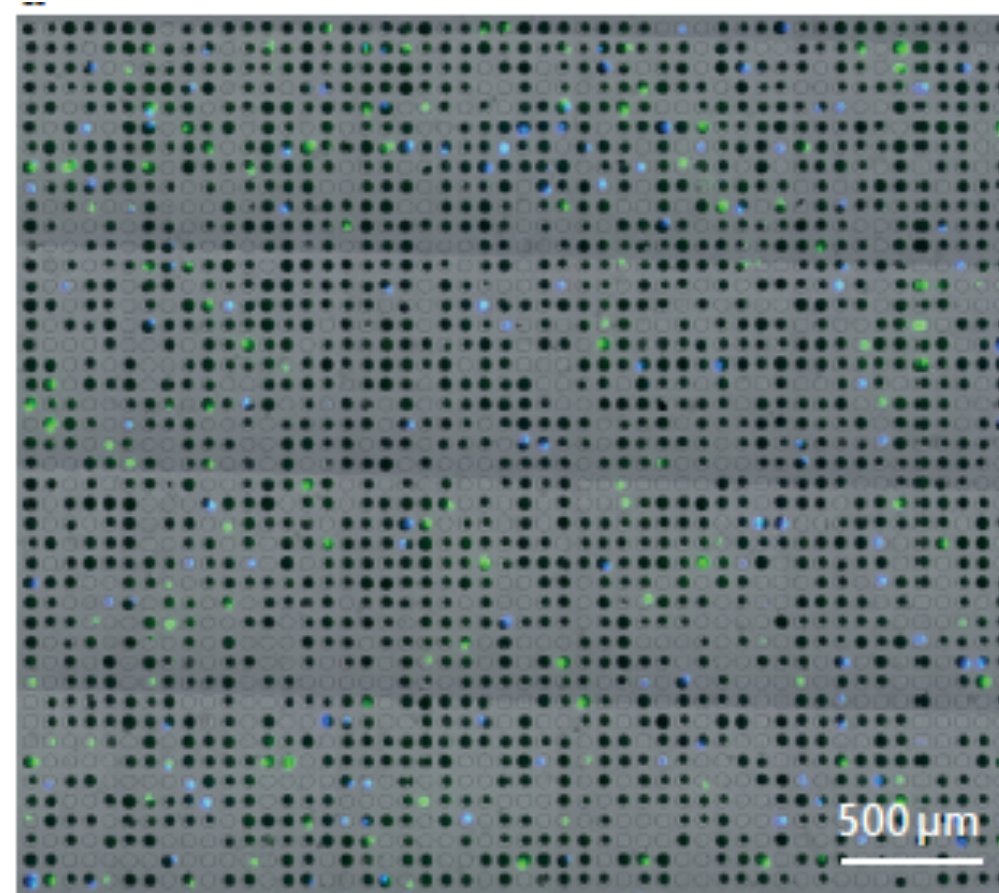


SINGLE CELL SEQUENCING - CAPTURING

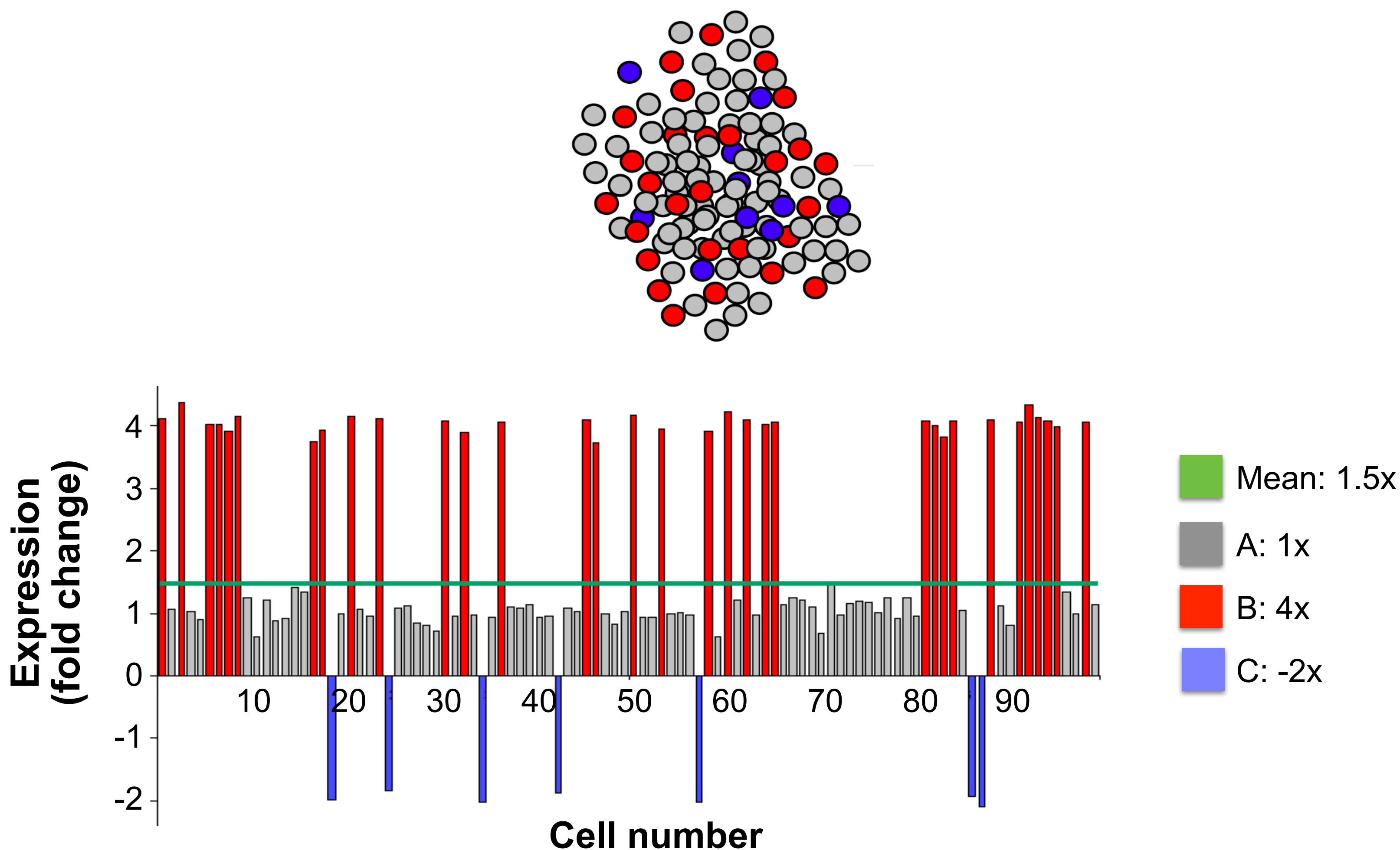
Microfluidics

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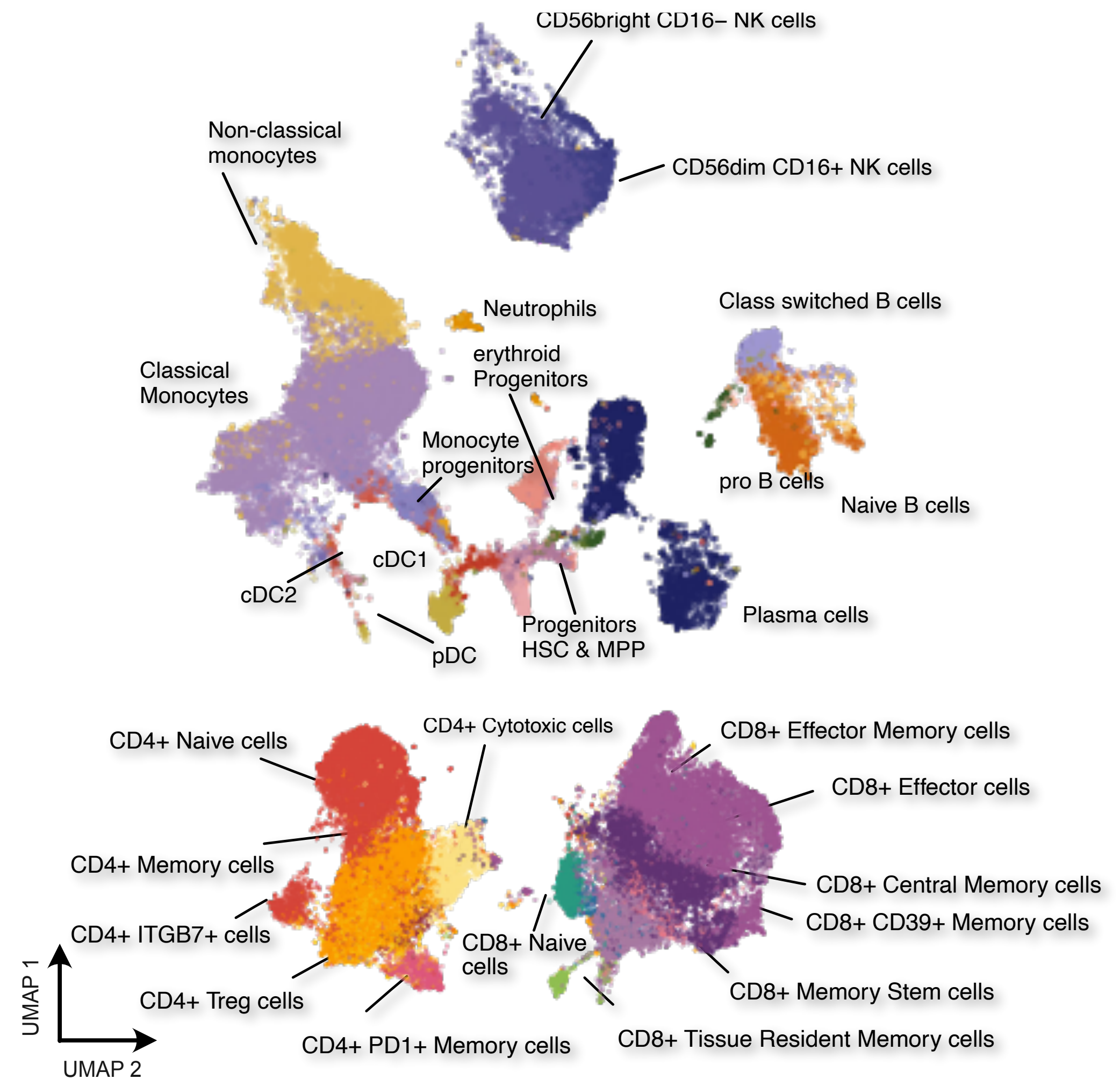
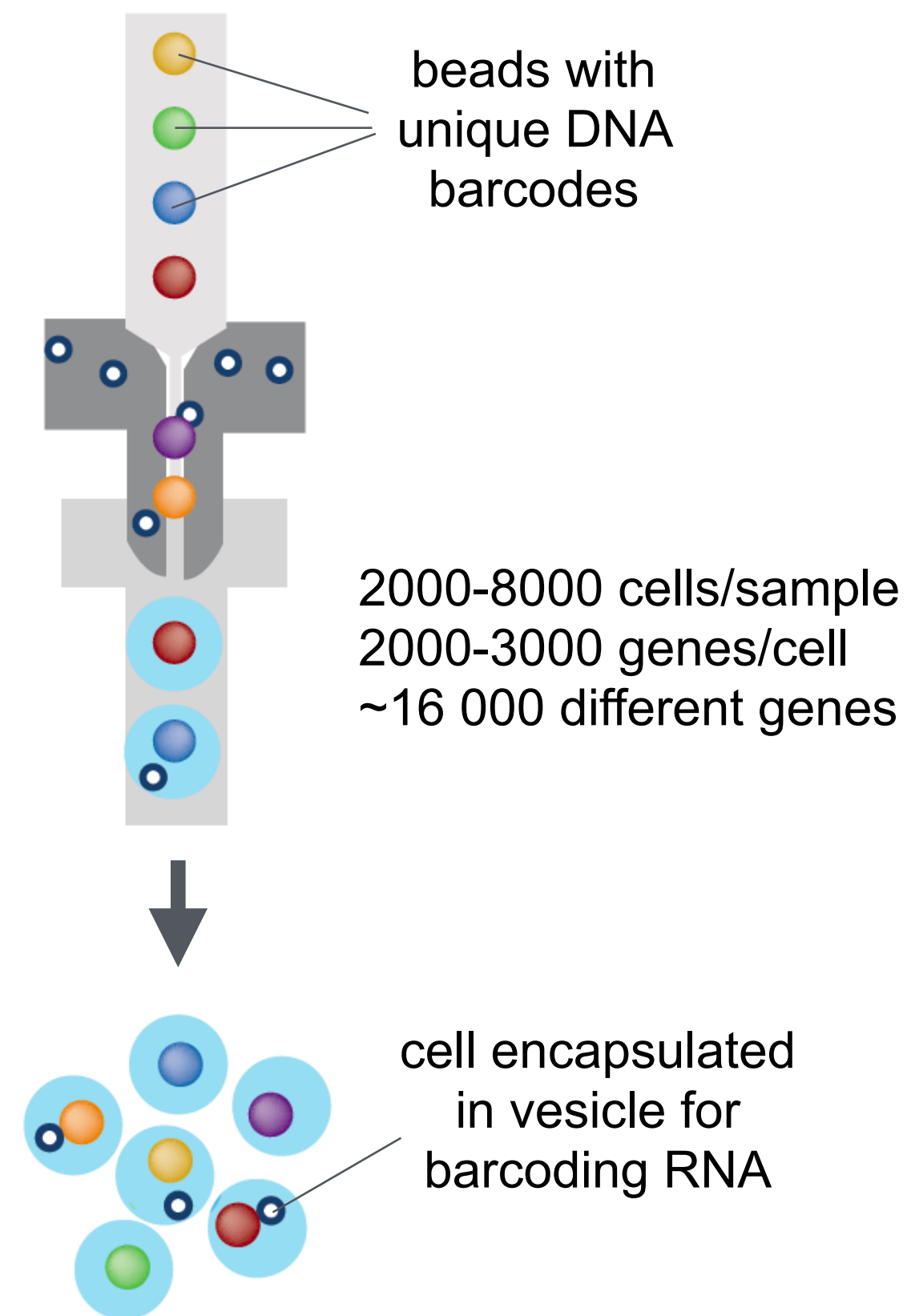


Single cell sequencing to dissect tumor heterogeneity by transcriptome and (epi)genome analysis

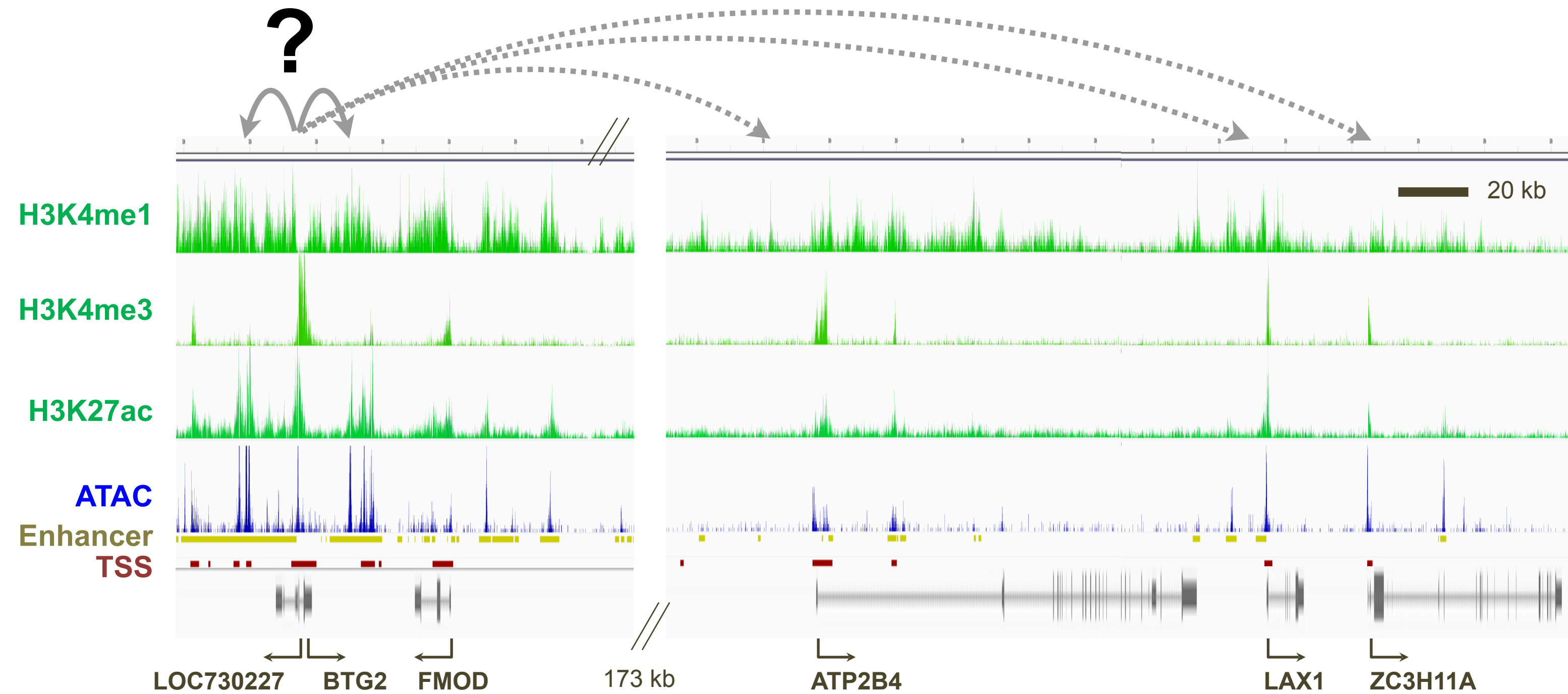
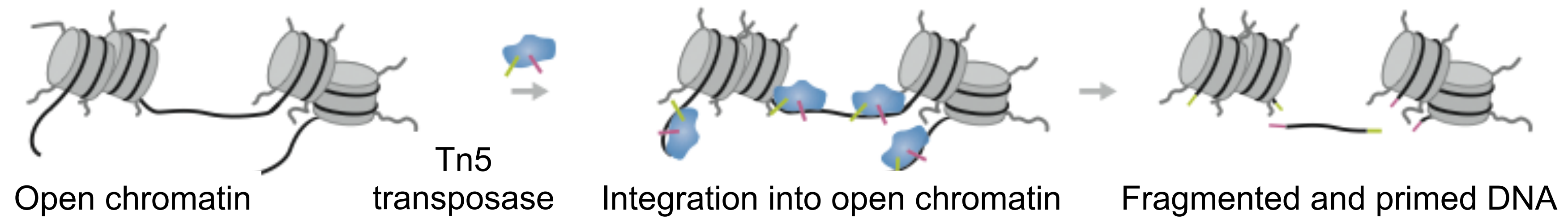


Single cell RNA sequencing of cell suspensions (bone marrow aspirate from multiple myeloma patients)

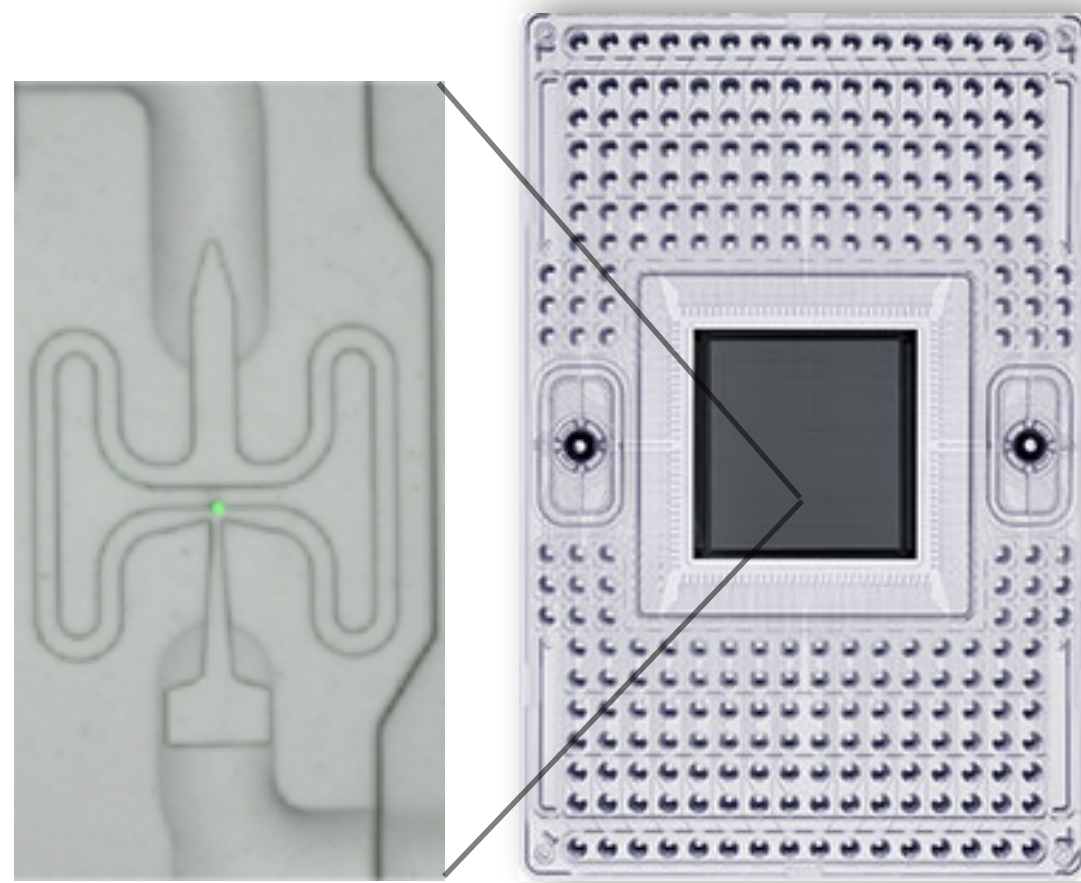
single cell RNA-seq
("drop-seq", 10x Genomics)



Mapping active enhancer and promoters by ATAC-seq

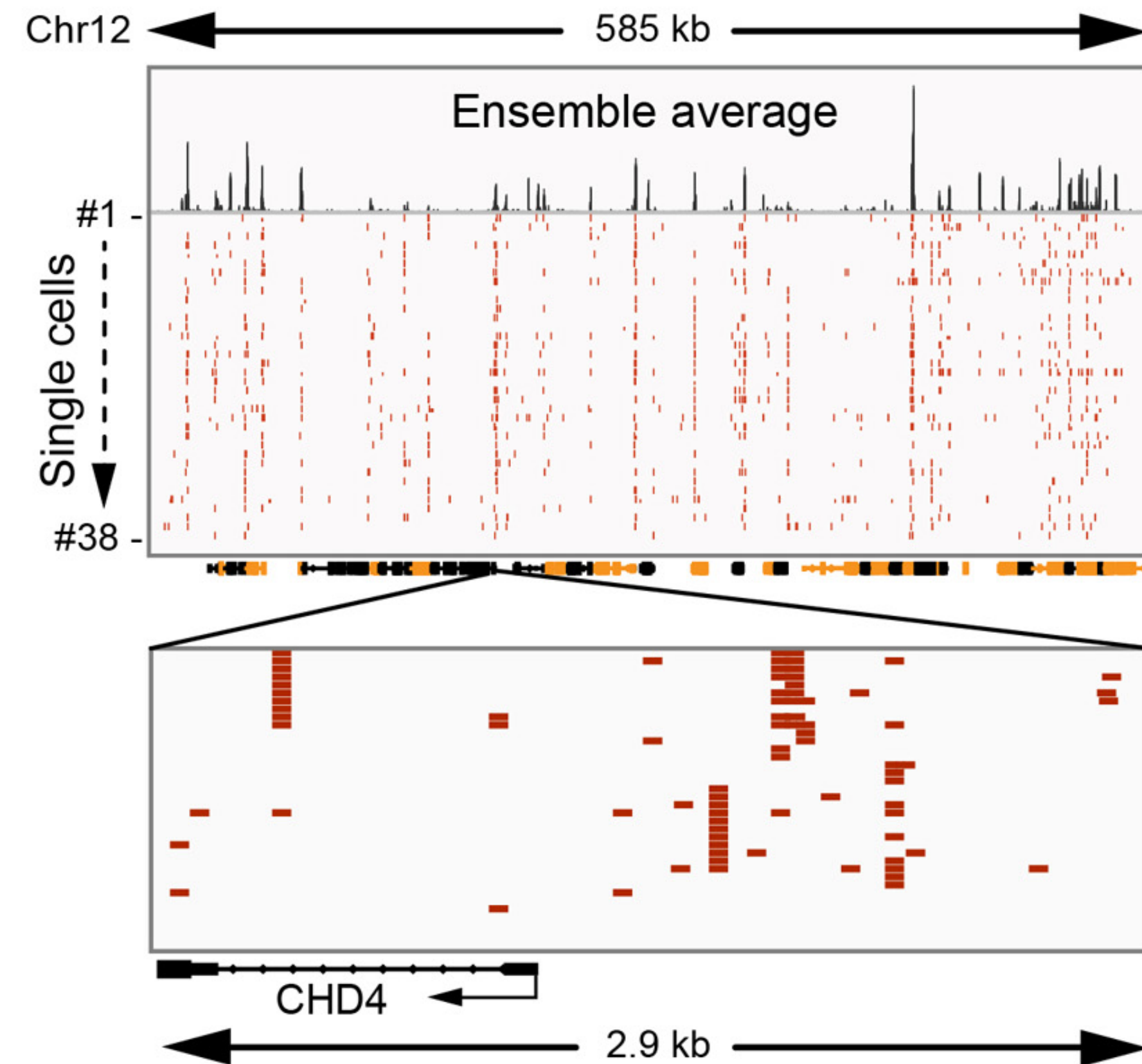


scATAC-seq analysis in chronic lymphocytic leukemia (CLL)



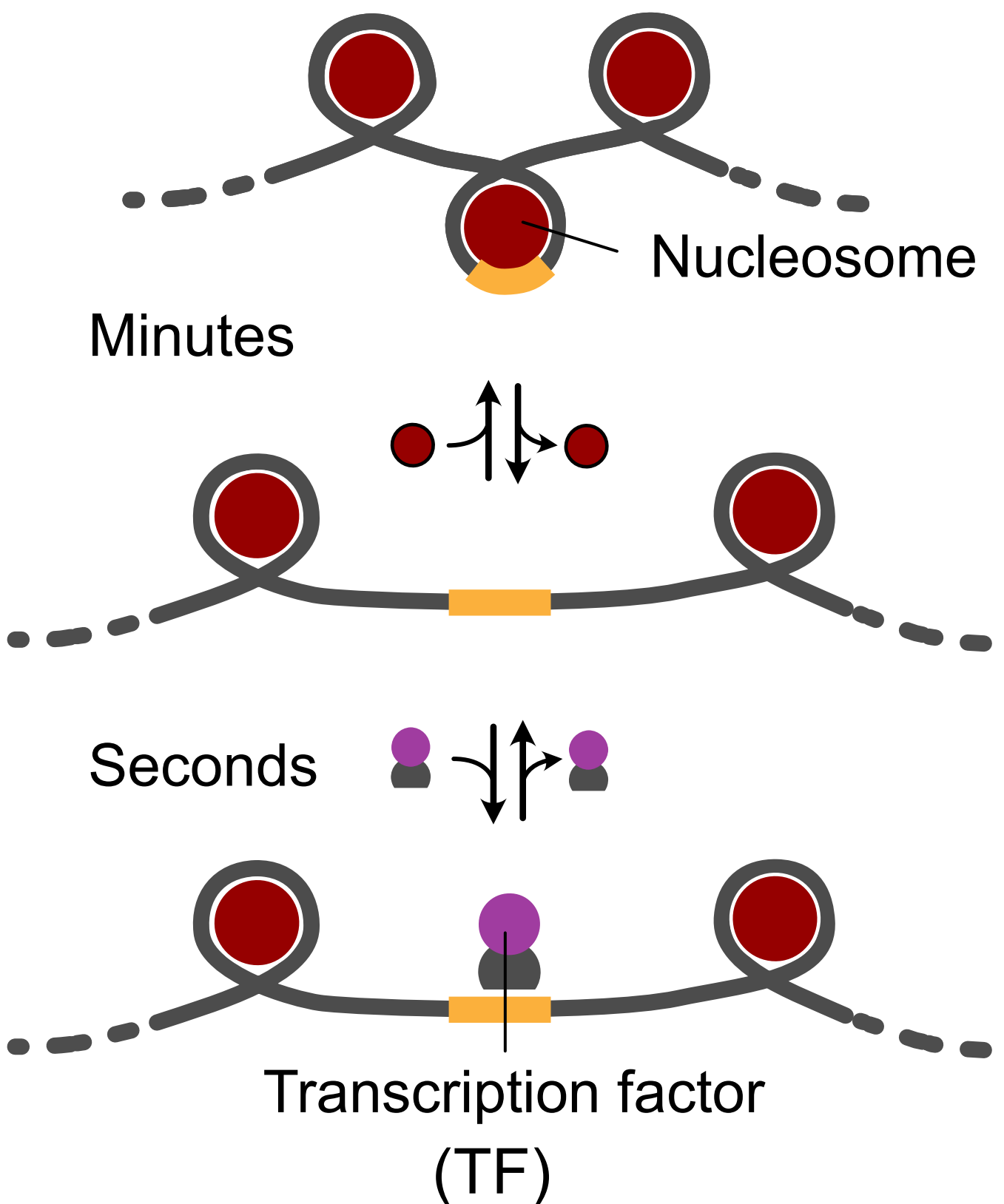
scATAC-seq (Fluidigm C1)

- C19+ B-cells, up to 96 single cells/run
- 40 000-130 000 integrations/cell
- 500-1500 open enhancer/promoters/cell
- 343 non-malignant and 494 CLL
- High-throughput flow cell (2 x 400 cells)

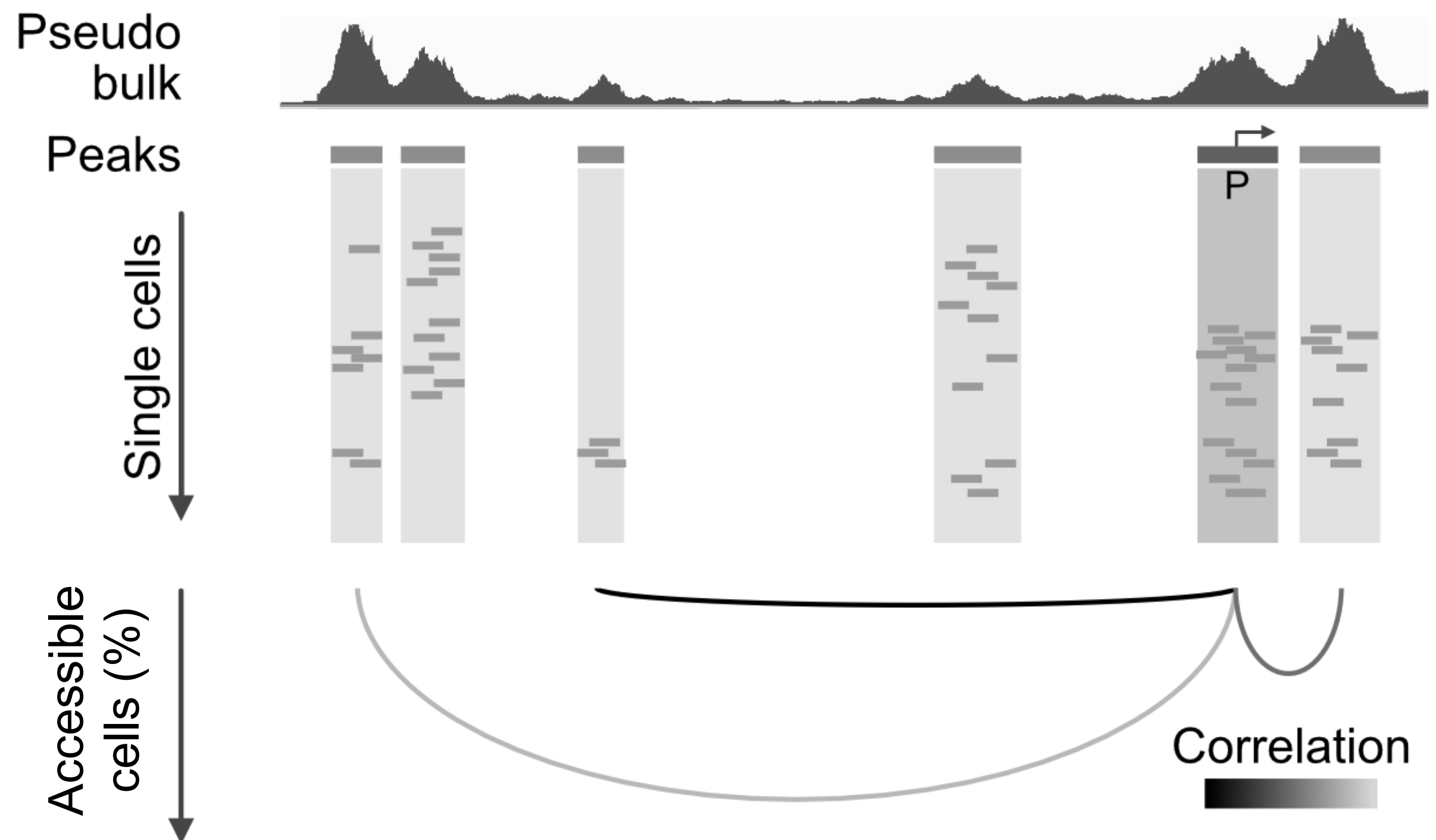


Single-cell ATAC-seq provides information on correlated accessibility changes that reflect transcription activity

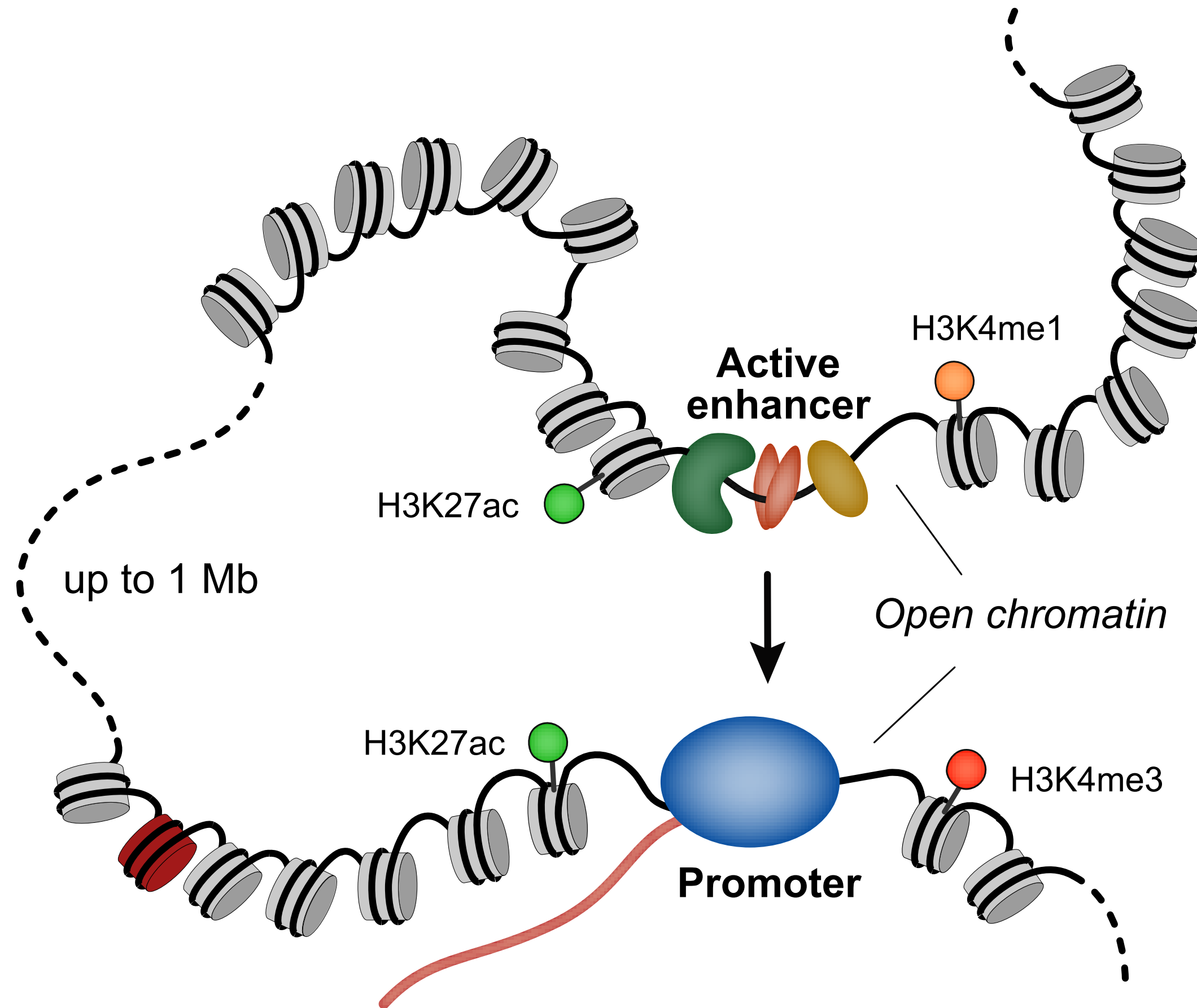
Accessibility changes



Accessibility correlations



Changes of chromatin features at promoters and enhancers

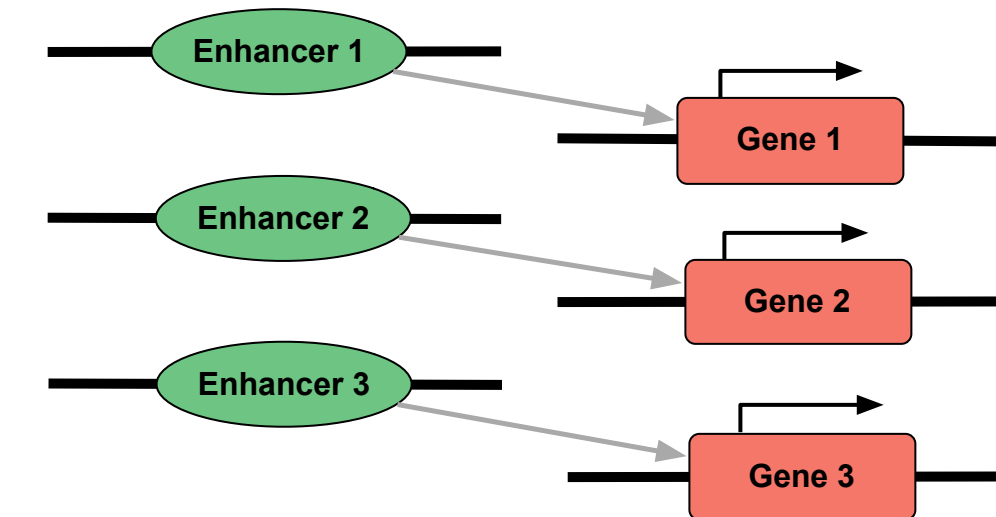


Cell type specific gene regulation in humans

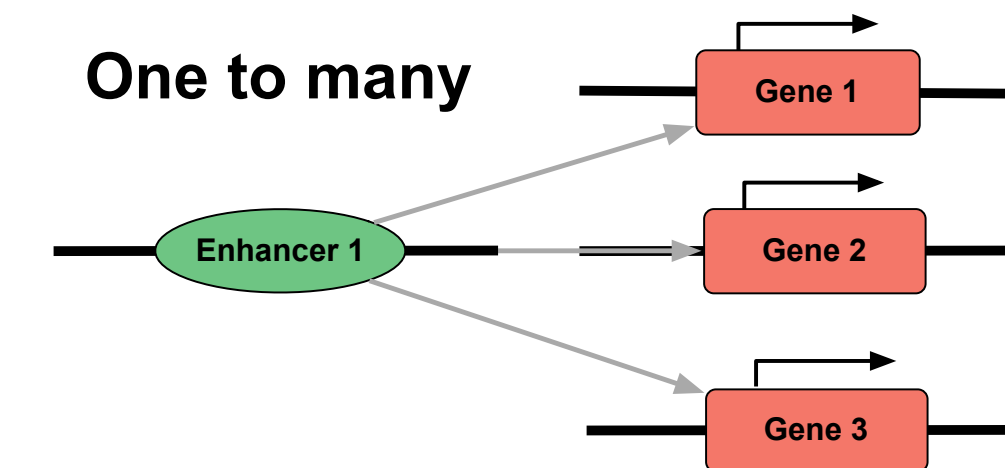
- ~230 000 potential enhancers (ChromHMM)
- ~60 000 promoters (RefSeq)

Enhancer-promoter topology

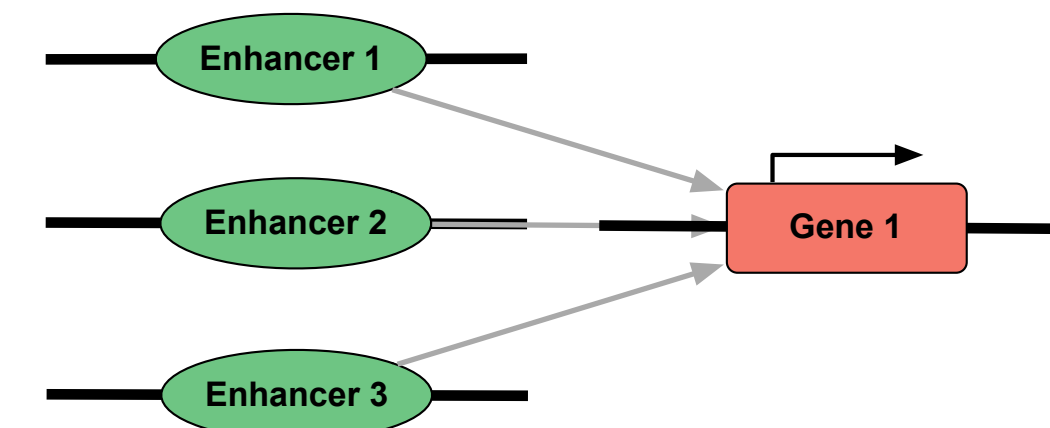
One to one



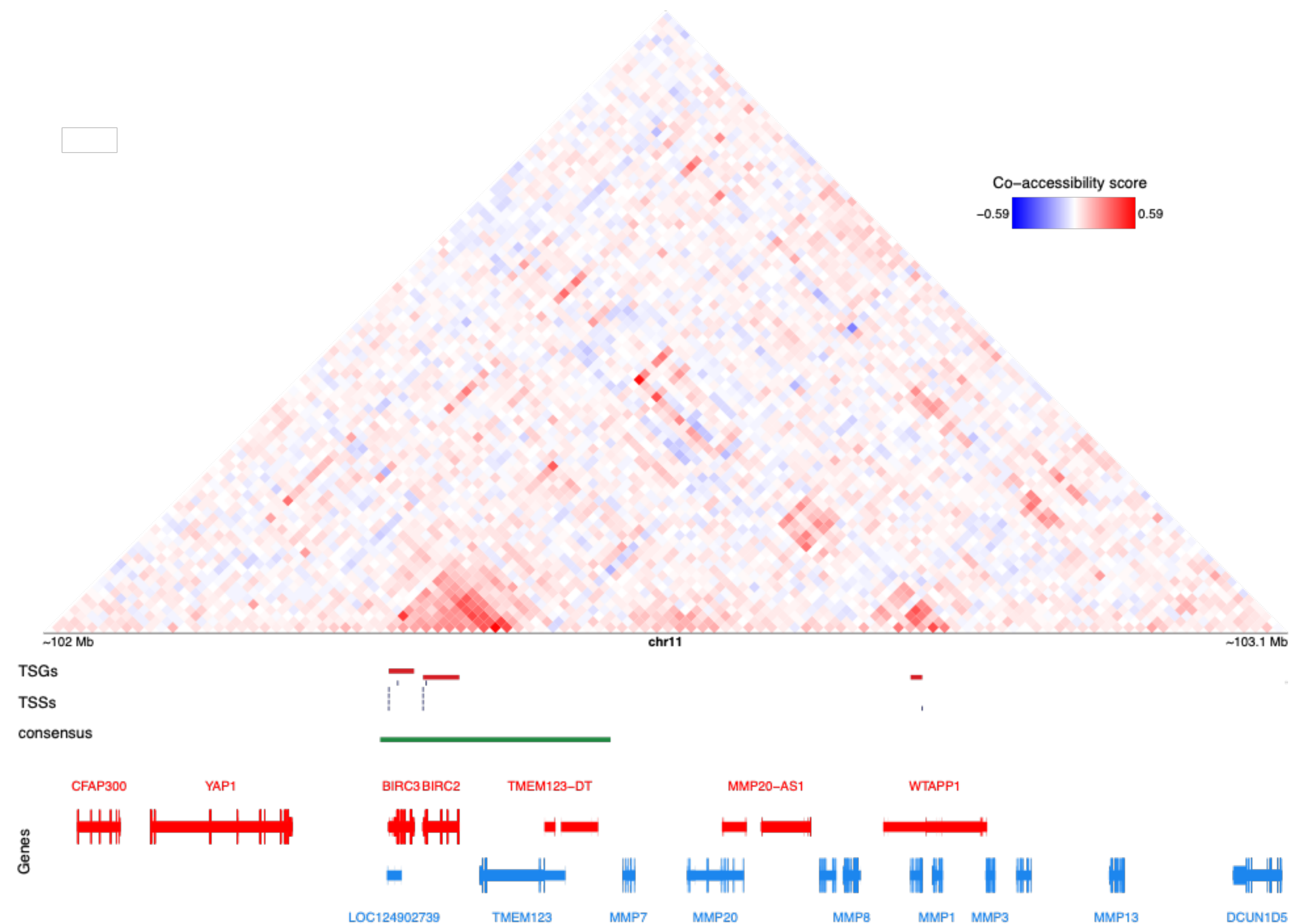
One to many



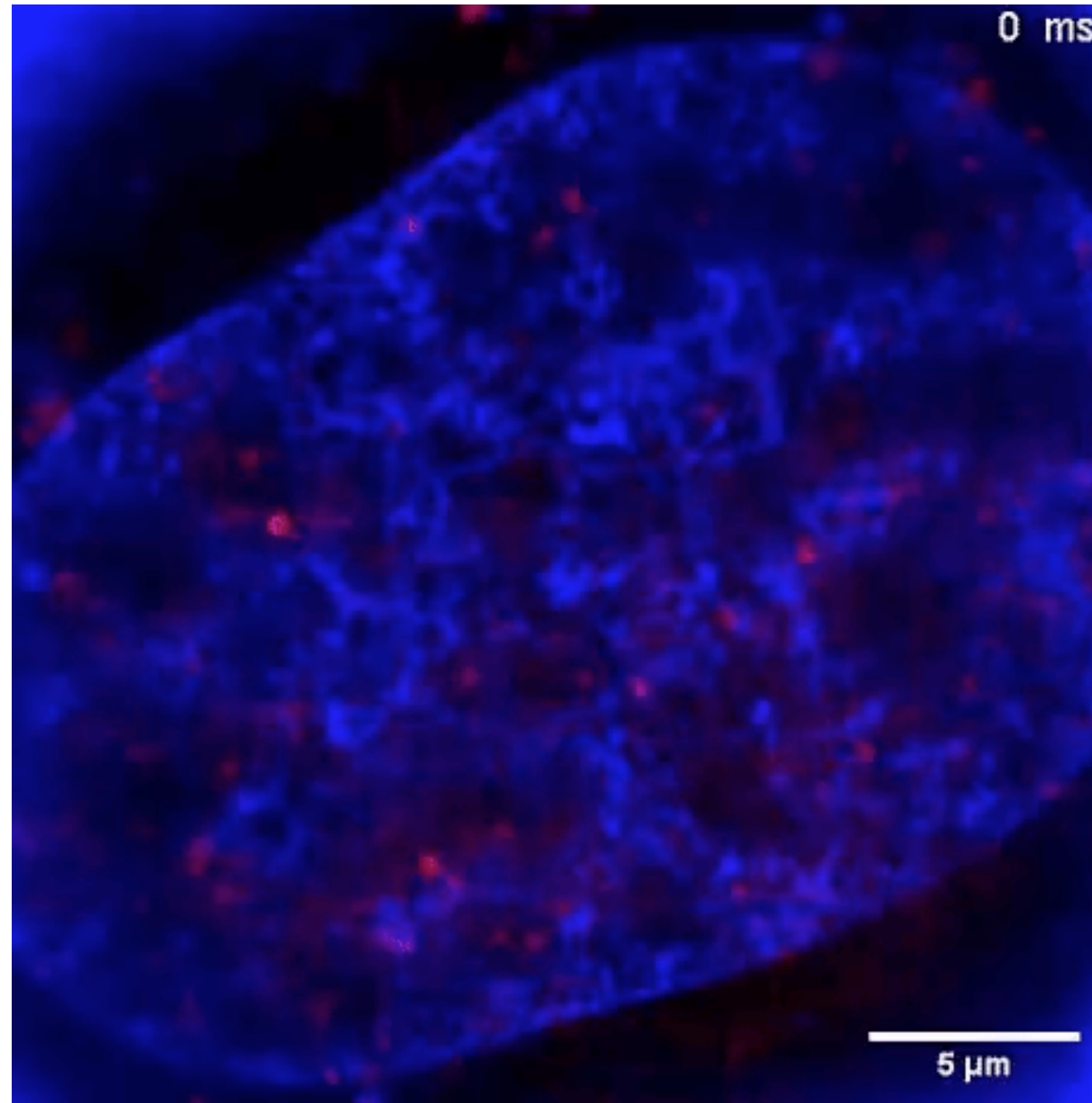
Many to one



Single-cell ATAC-seq to identify domains with increased co-accessibility score and transcription factor binding activity



Locally enriched activity of NF- κ B stimulated with TNF α in the cell nucleus



Tissue context information is lost in sequencing analysis

“Tissue” (fruit pieces = cells)



Spatially resolved transcriptomics

- Link histological and molecular markers
- Resolve tumor heterogeneity in tissue
- Interplay of cancer cells and microenvironment
- Features of tumor boarder lines/invasive fronts

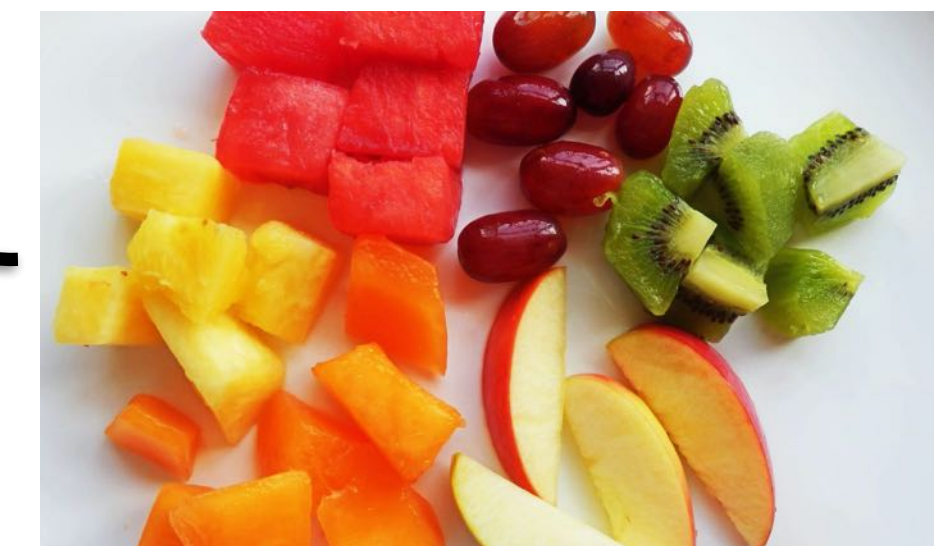
Bulk sequencing



Bulk sequencing after sorting



Single cell sequencing

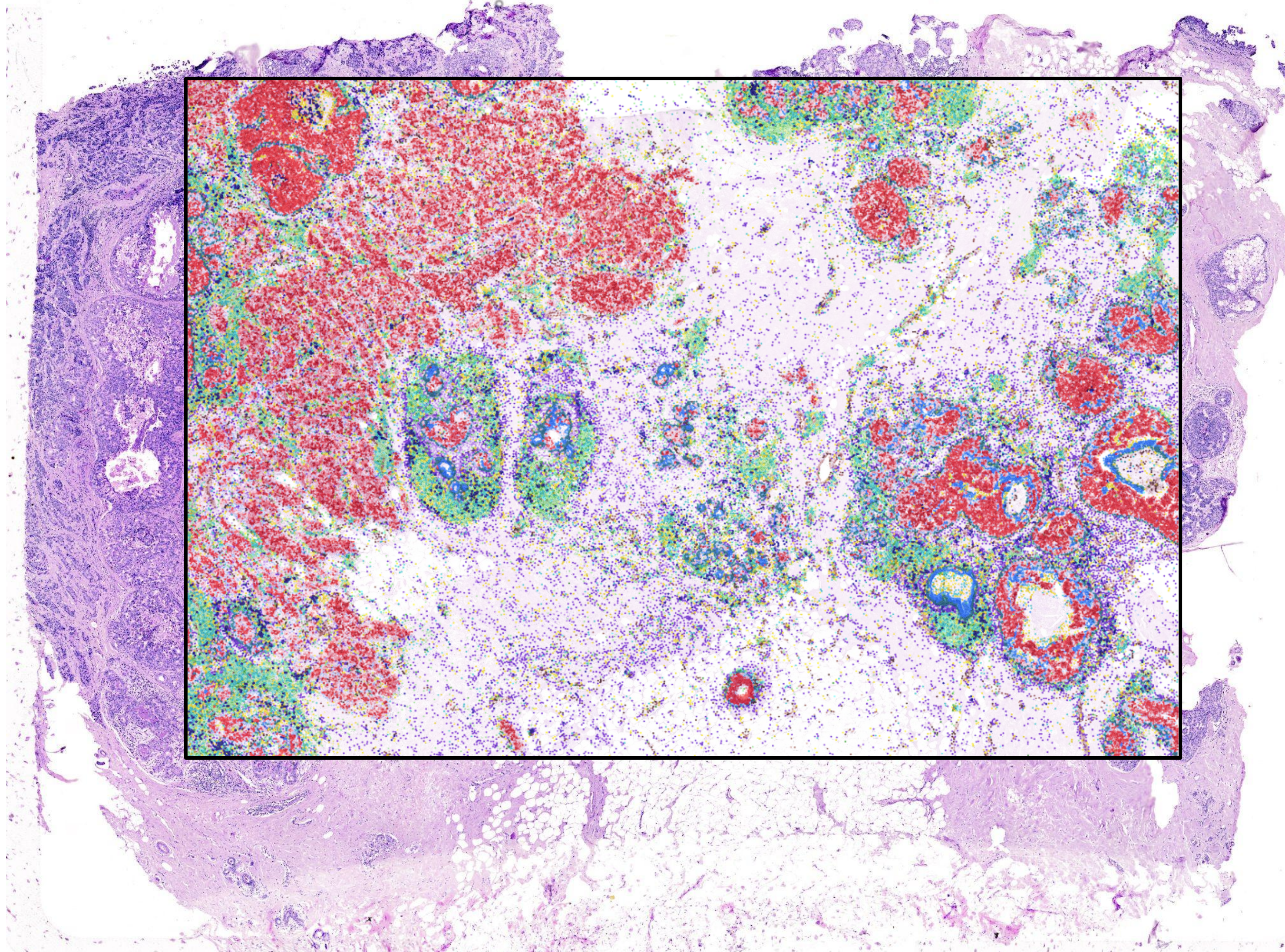


Spatial transcriptomics of a breast cancer tissue section

Hematoxylin & eosin staining (H&E): 2 colors

- pink: protein (any)
- purple: DNA

Morphology
= cell type



Spatial omics: 50-1,000 markers

- specific proteins
- specific transcripts

Cell type

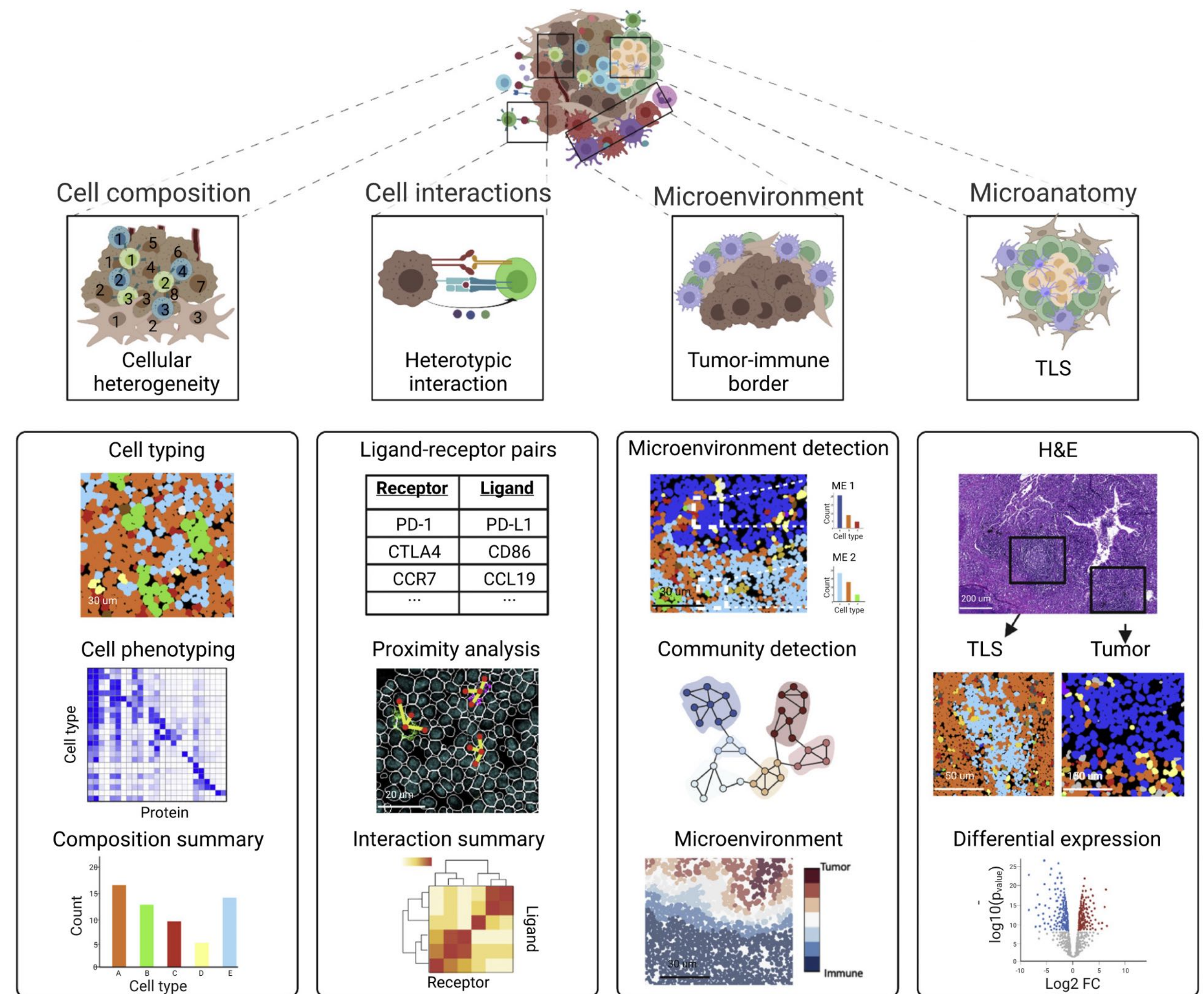
Endothelial RGS5	NKT cells
Endothelial Lymphatic LYVE1	DCs
Endothelial ACKR1	Cycling_Myeloid
Endothelial CXCL12	Macrophage
CAFs MSC iCAF-like	Monocyte
CAFs myCAF-like	Mature Luminal
PVL Immature	Myoepithelial
Cycling PVL	Luminal Progenitors
PVL Differentiated	Plasmablasts
B cells Memory	Cancer Basal SC
B cells Naive	Cancer Cycling
T cells CD8+	Cancer Her2 SC
T cells CD4+	Cancer LumB SC
Cycling T-cells	Cancer LumA SC
NK cells	

Cell types & states defined
by molecular characteristics

Use AI to characterise
1,000,000 cells/slide

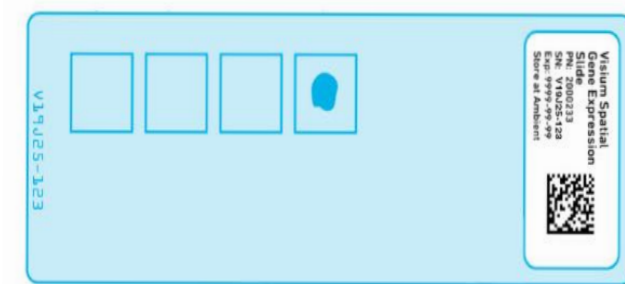
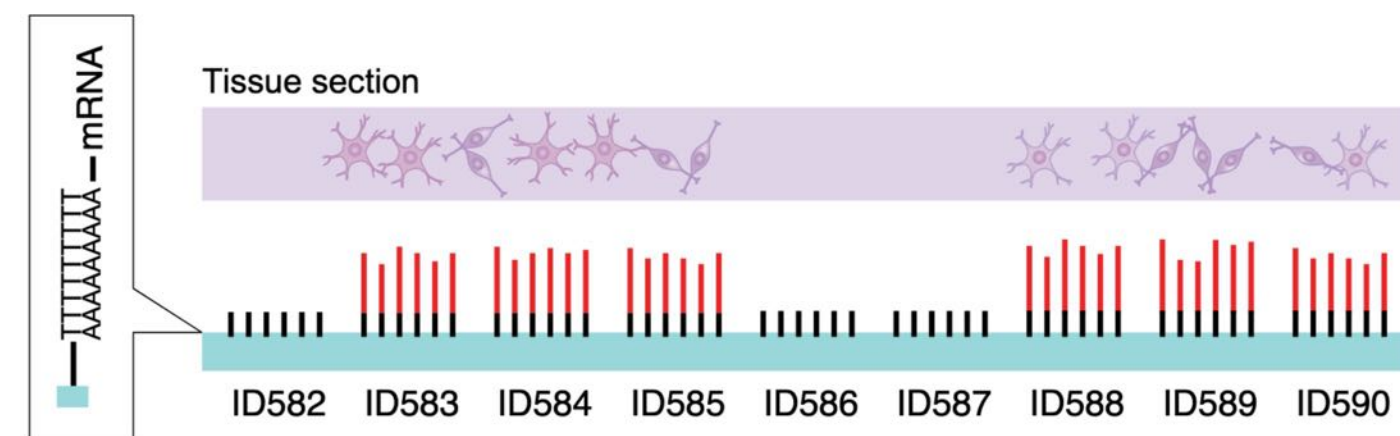
Spatial cancer biology and oncology

- **Cell composition** of the tumor microenvironment:
→ ~50% of cells are normal
- **Resolve intra-tumor heterogeneity & subclones**
→ Mechanisms of progression and resistance
- **Interactions with immune and other normal cells:**
→ Identify drug targets, eg immuno-oncology
- **Characterize tumor microanatomy:**
→ Role of tissue structure

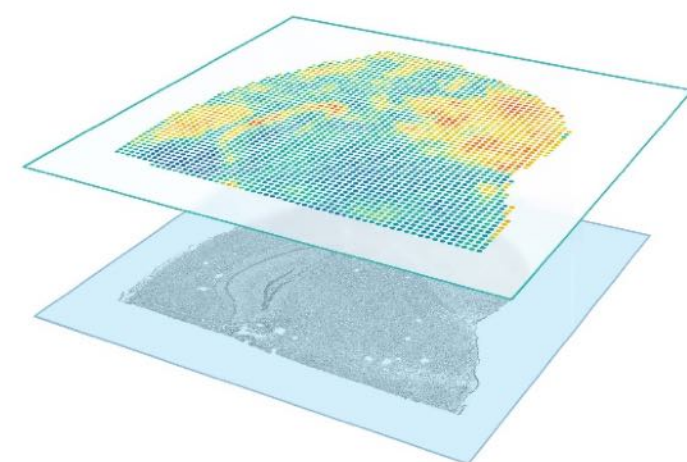


Sequencing and hybridization-based methods for spatial transcriptomics

Sequencing-based methods (unbiased)

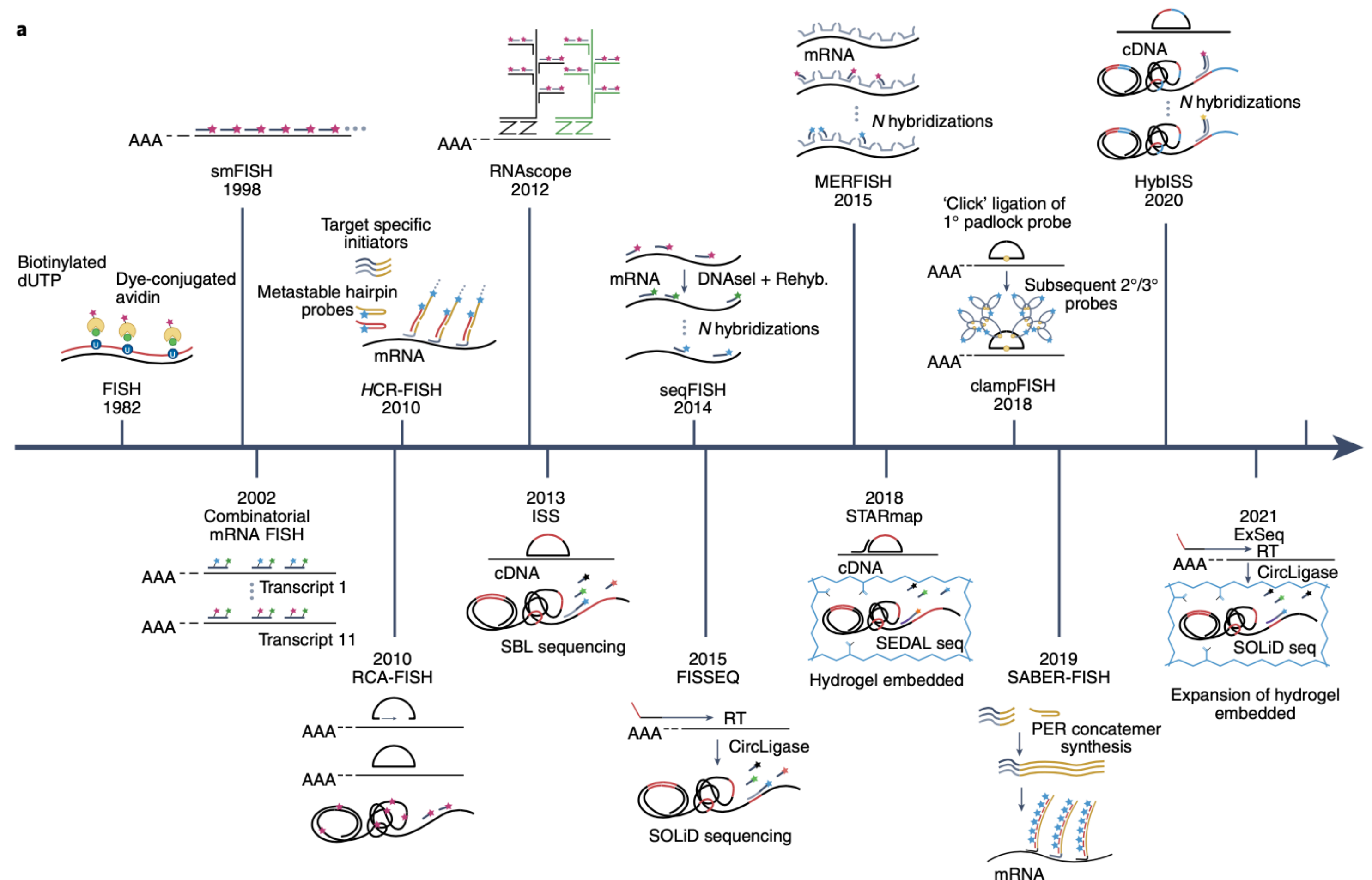


Sequencing of captured RNAs

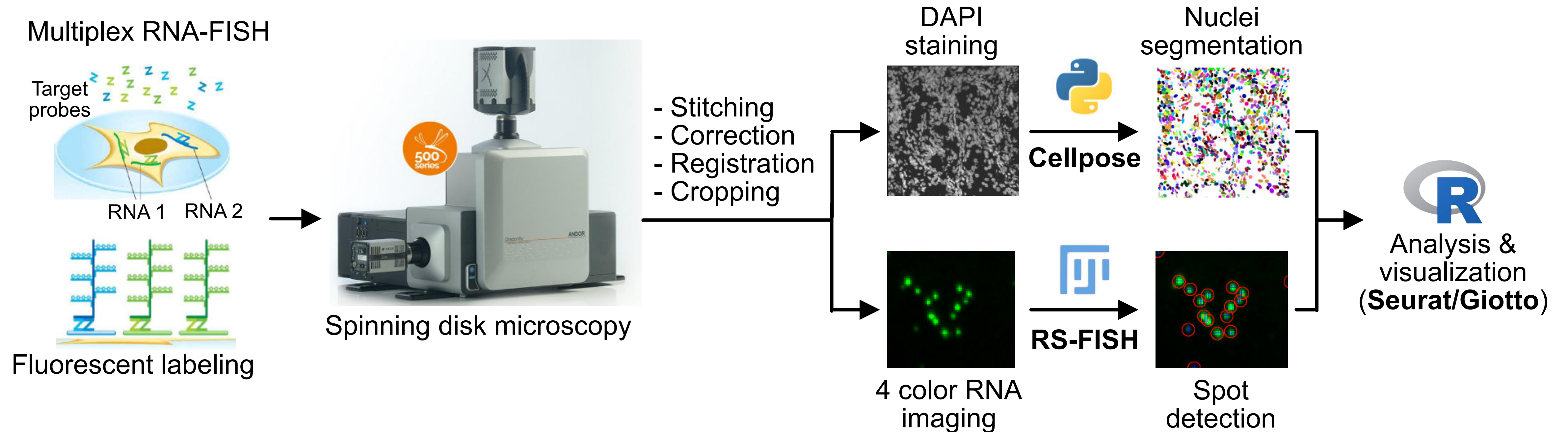


- Visium
- Seq-Scope
- Stereo-seq

Hybridization-based methods (targeted)

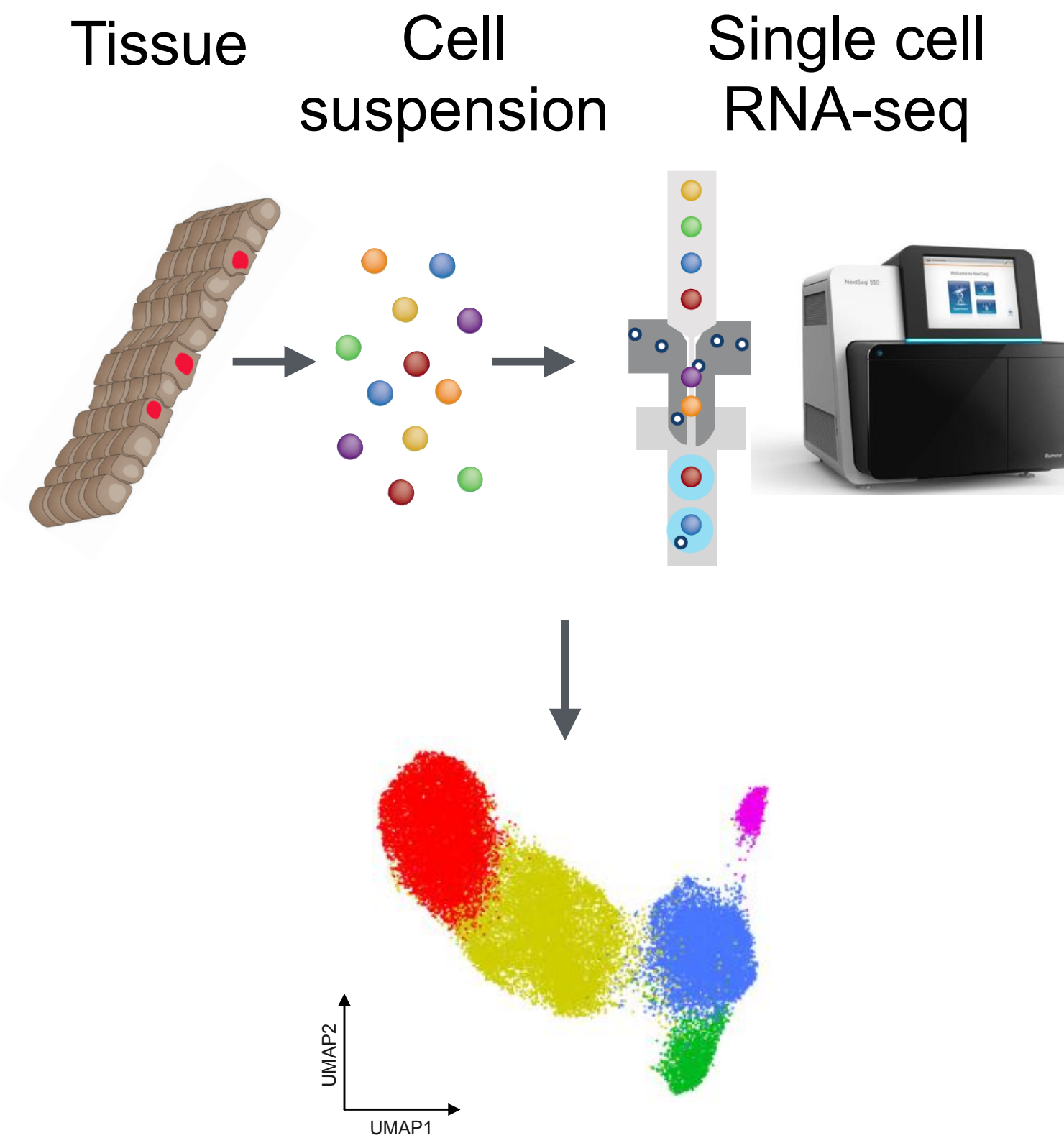
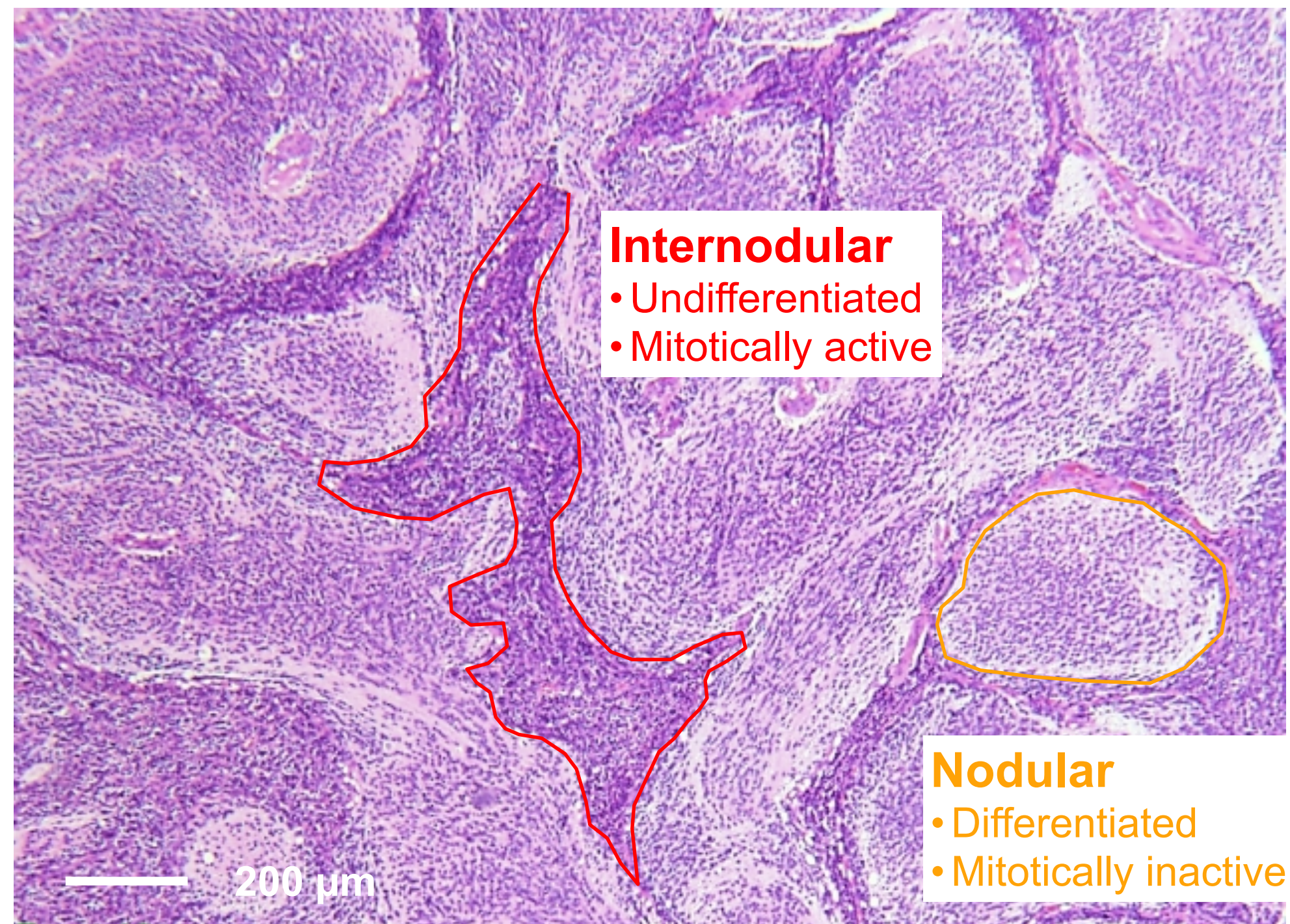


RNAScope with 12-48 probes provides subcellular resolution



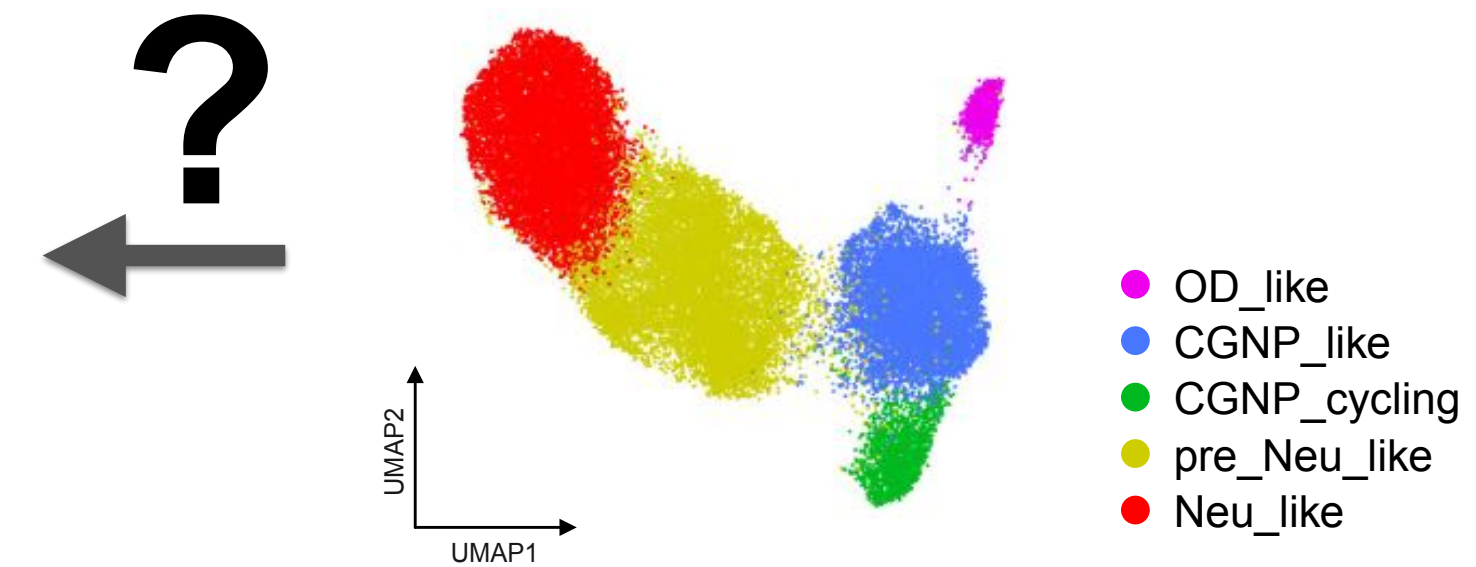
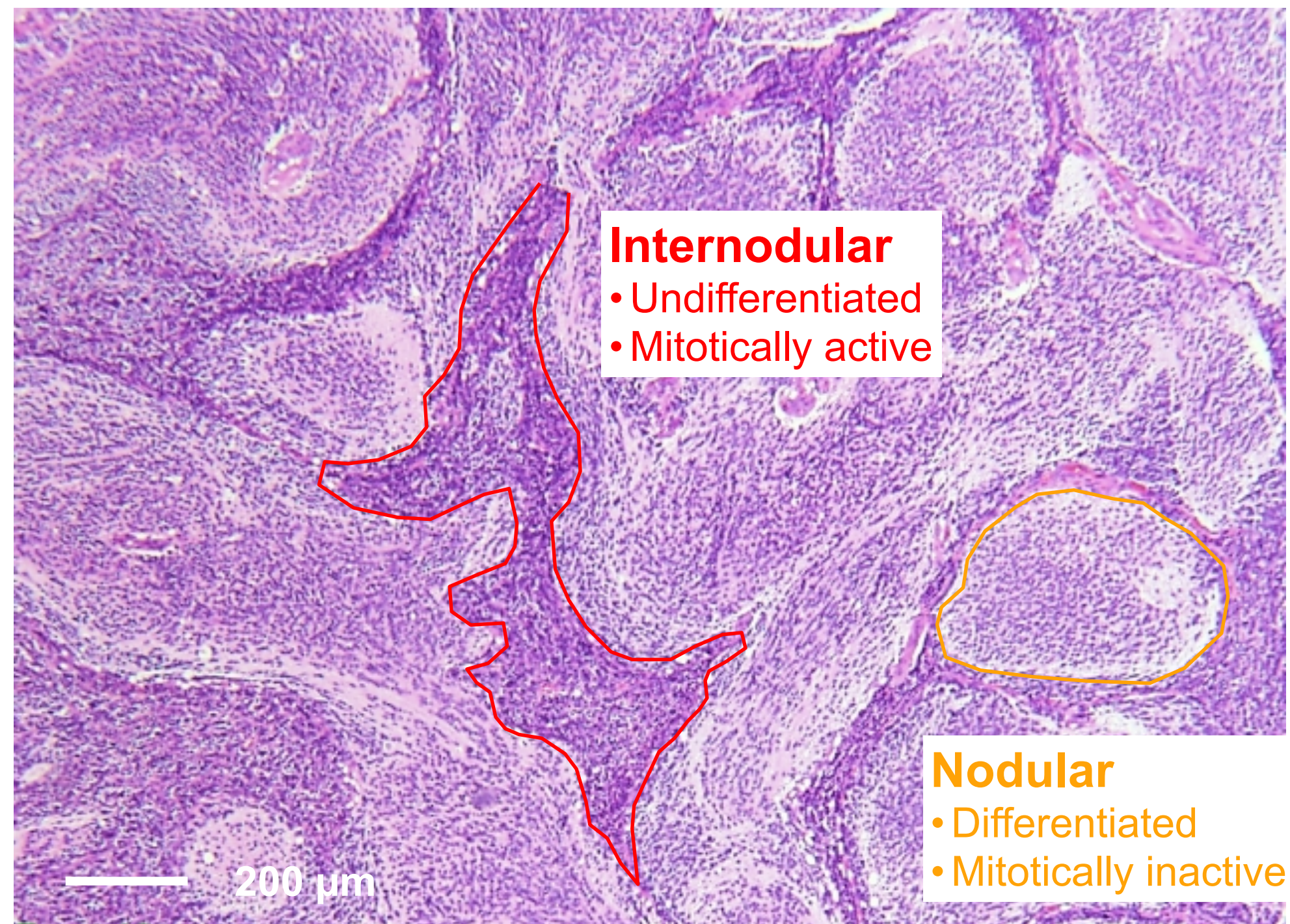
Spatially transcriptomics (ST) to study medulloblastoma with extensive nodularity (MBEN)

Medulloblastoma with extensive nodularity (MBEN)



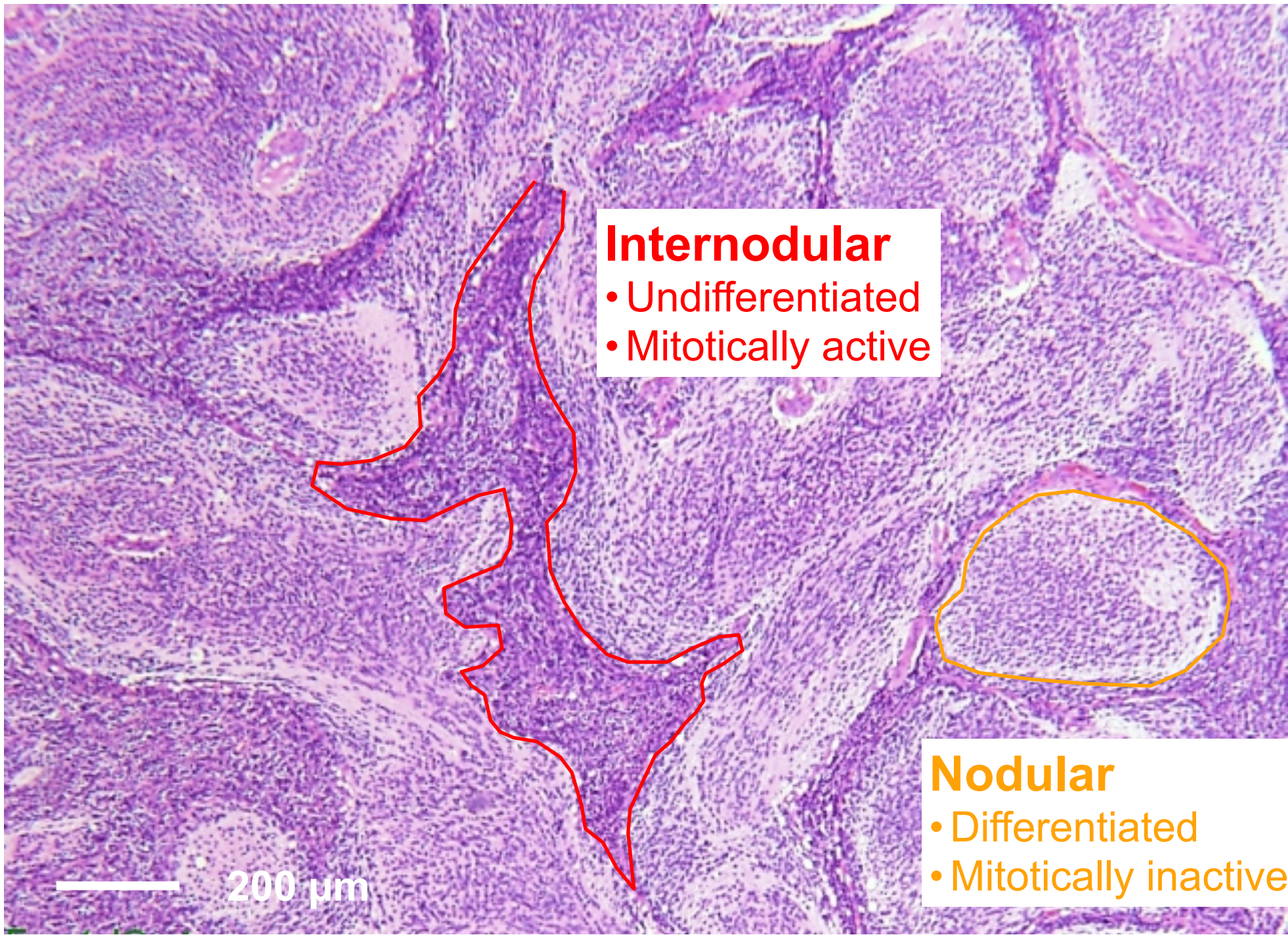
Assignment of scRNA-seq reference data from cell suspension

Medulloblastoma with extensive nodularity
(MBEN)

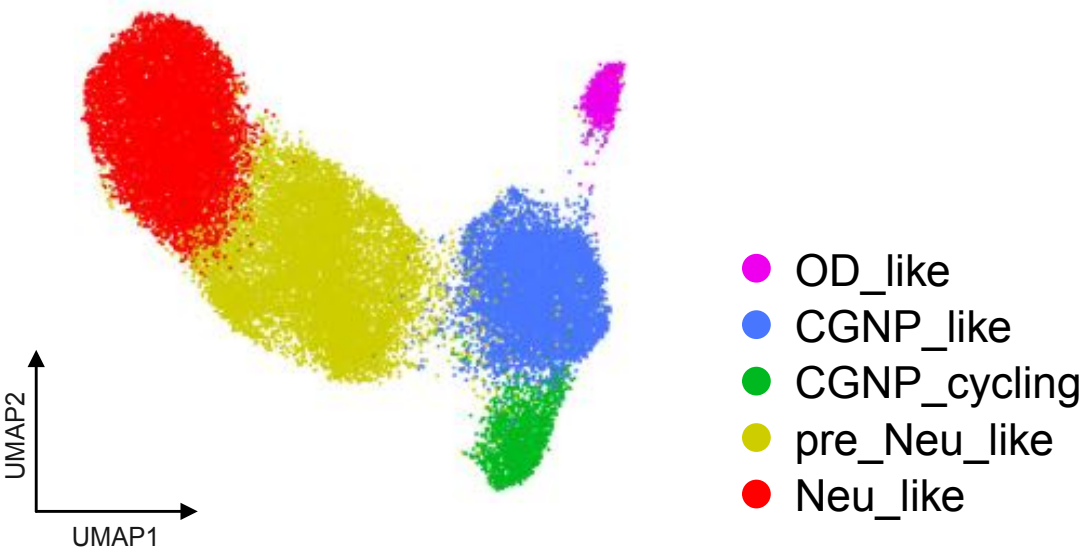


Selection of marker RNAs

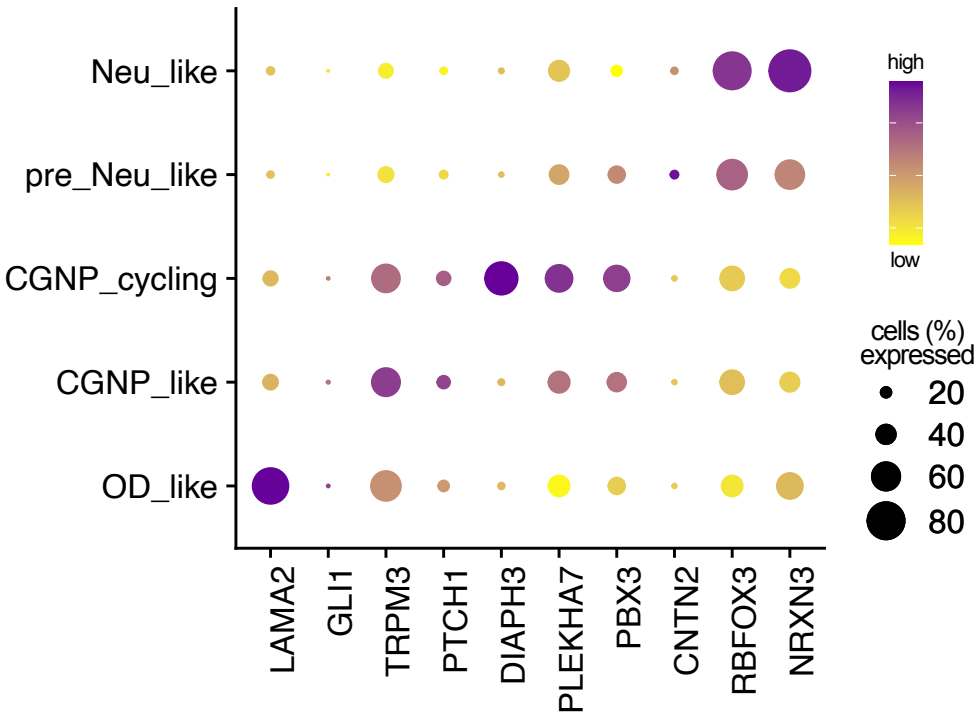
Medulloblastoma with extensive nodularity
(MBEN)



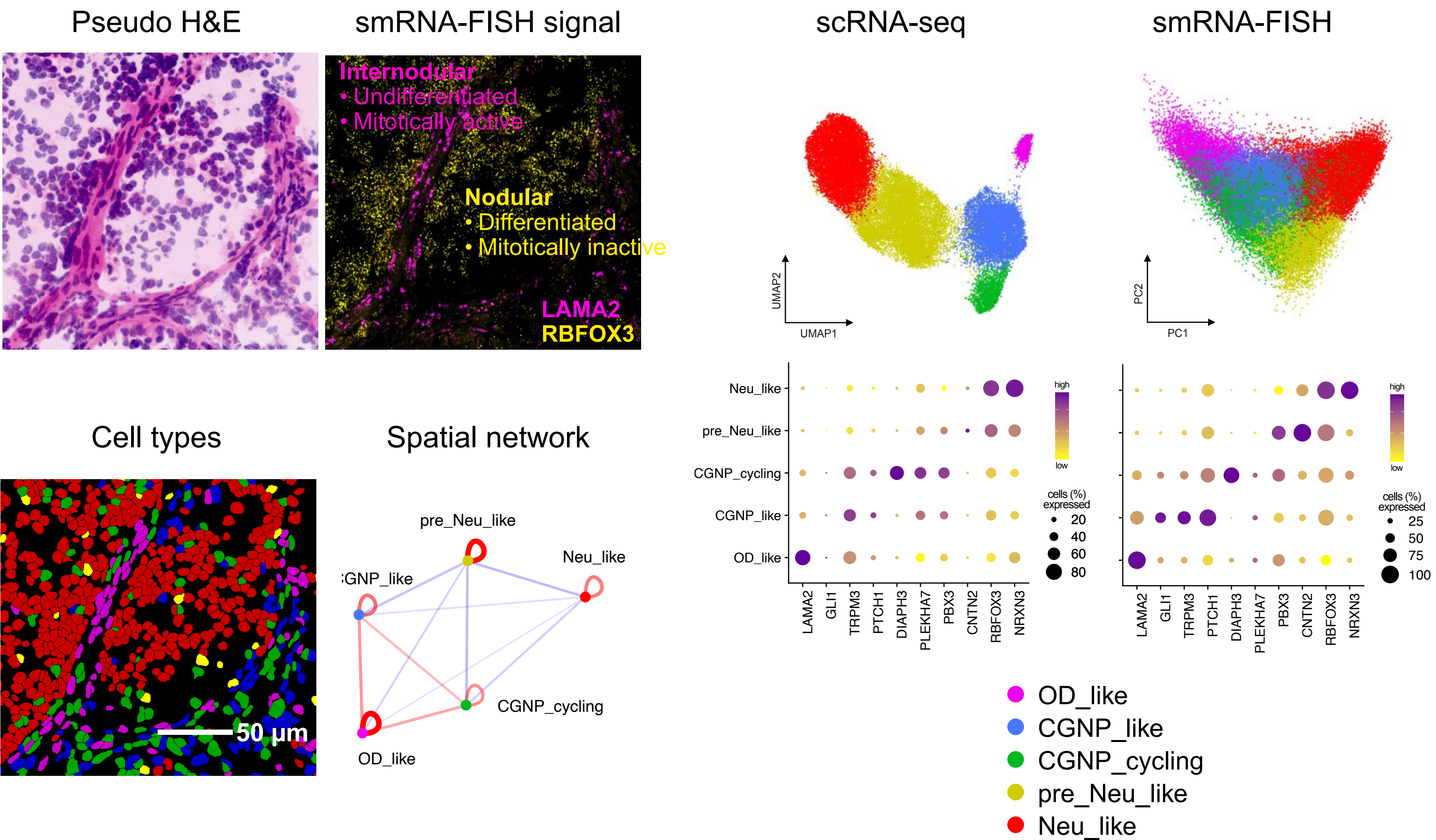
scRNA-seq



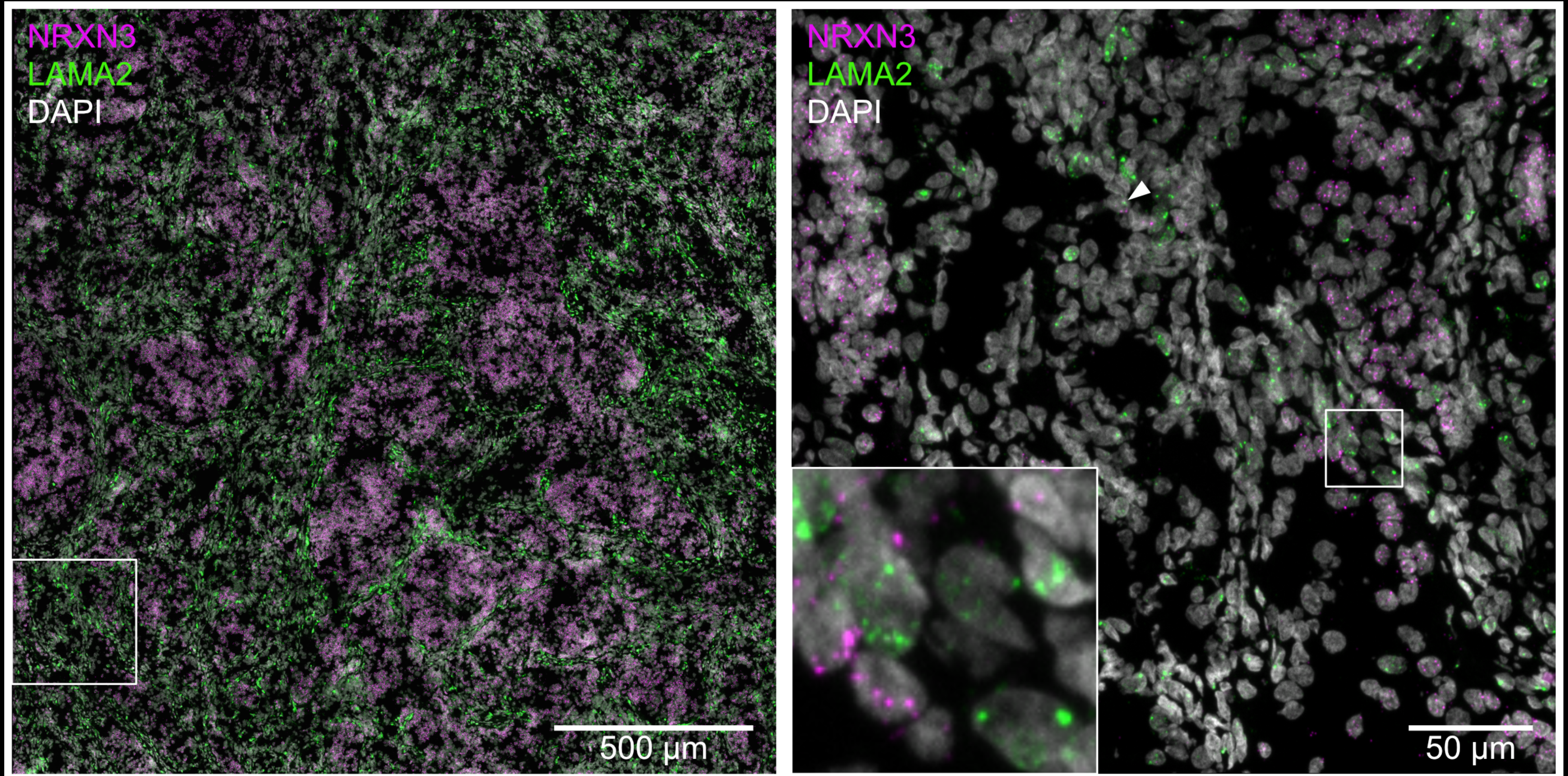
Identification of marker RNAs



Cell type identification and spatial relations in MBEN

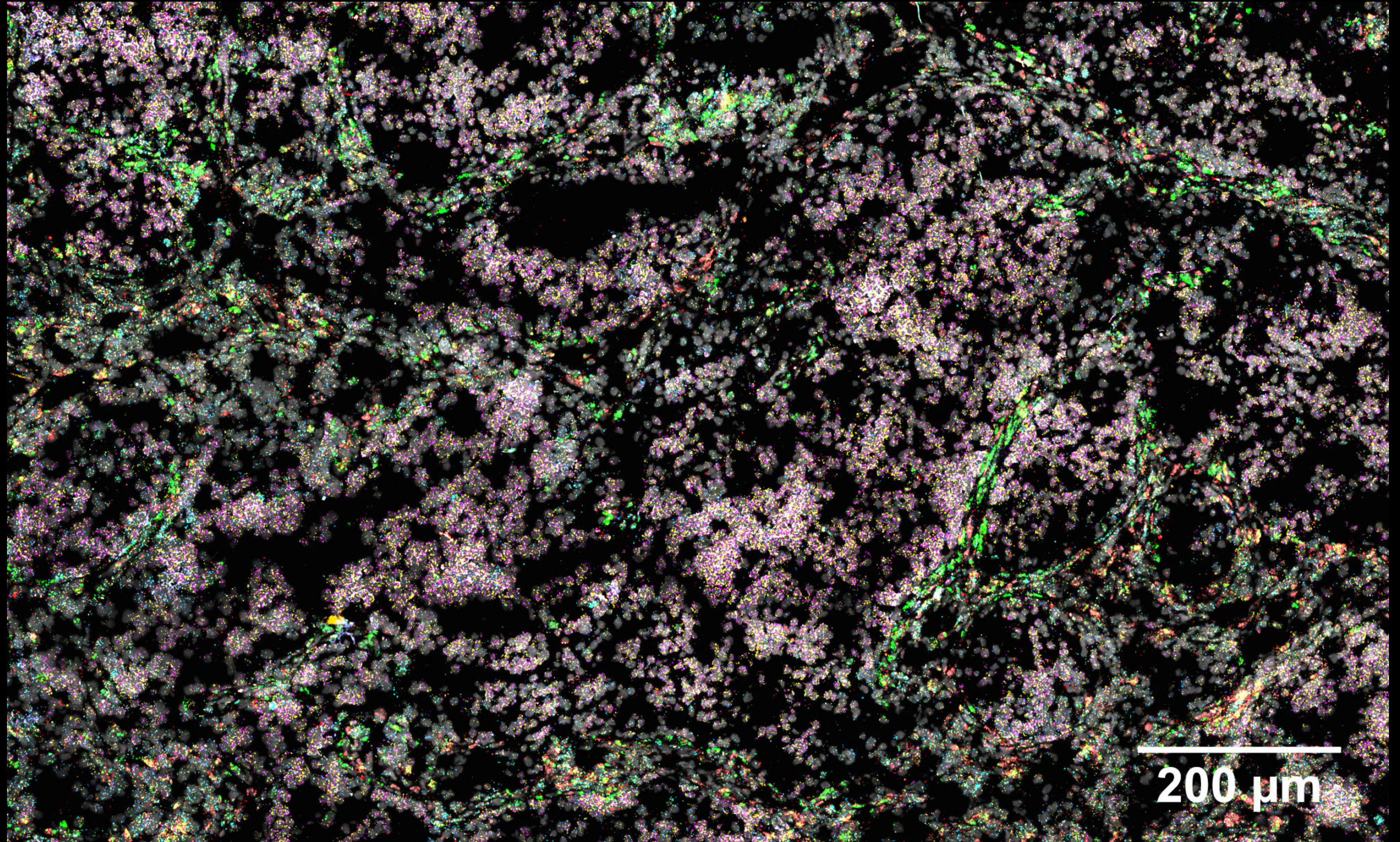


MBEN RNAScope analysis of transcripts provides subcellular resolution



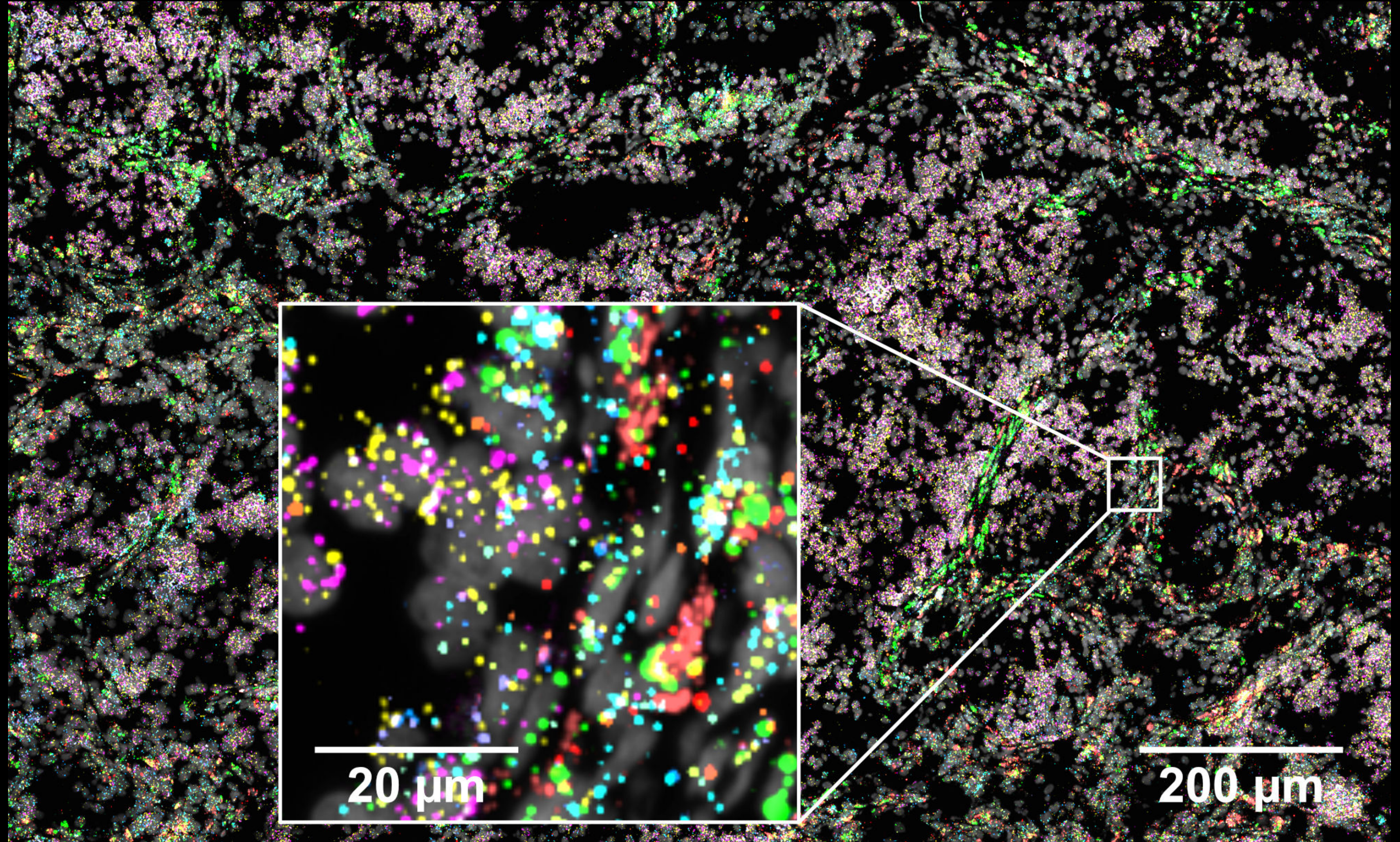
Spatial transcriptomics analysis of MBENs with 12 marker RNAs

DIAPH3
GL1
RBFOX3
PTCH1
NRXN3
CNTN2
PLEKHA7
PBX3
LRRTM4
FOS
LAMA2
TRPM3
DAPI



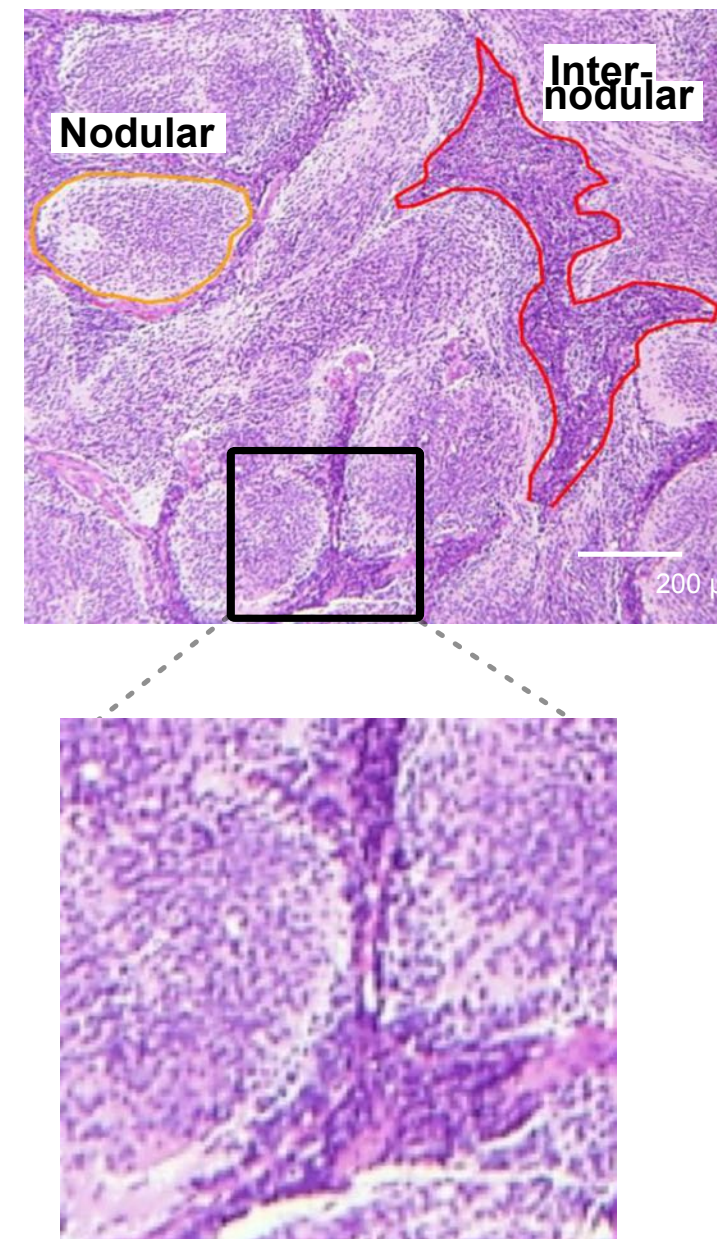
Spatial transcriptomics analysis of MBENs with 12 marker RNAs

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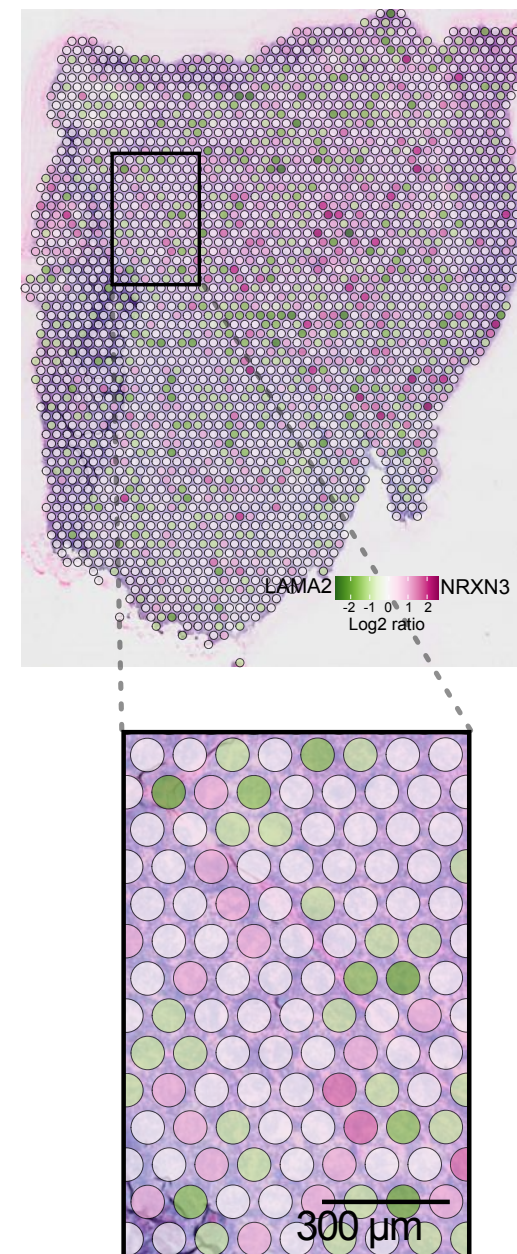
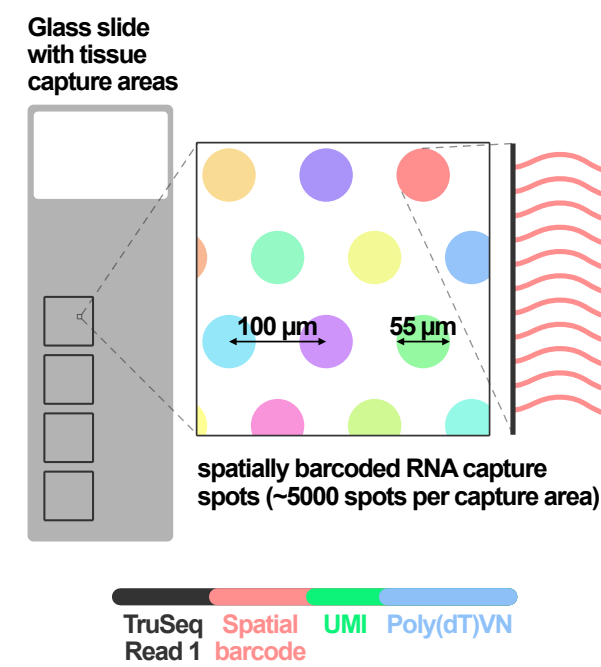


Comparison of different methods for spatial transcriptomics

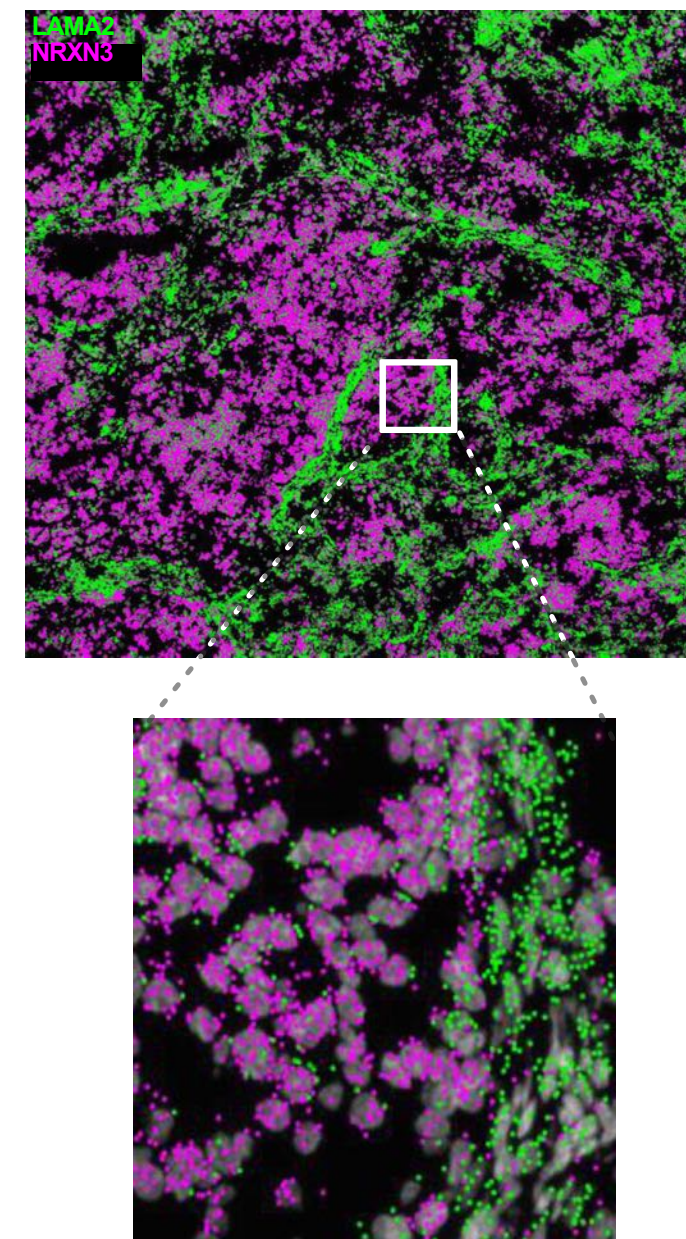
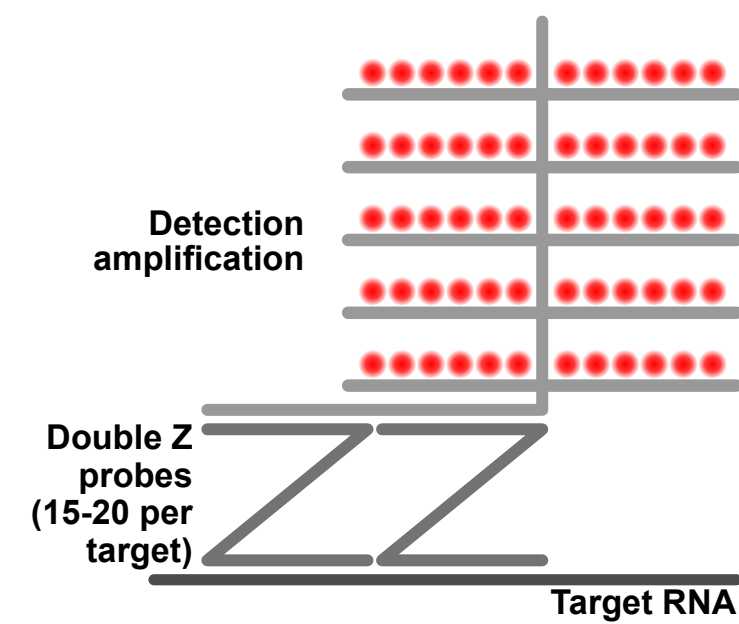
H&E
reference



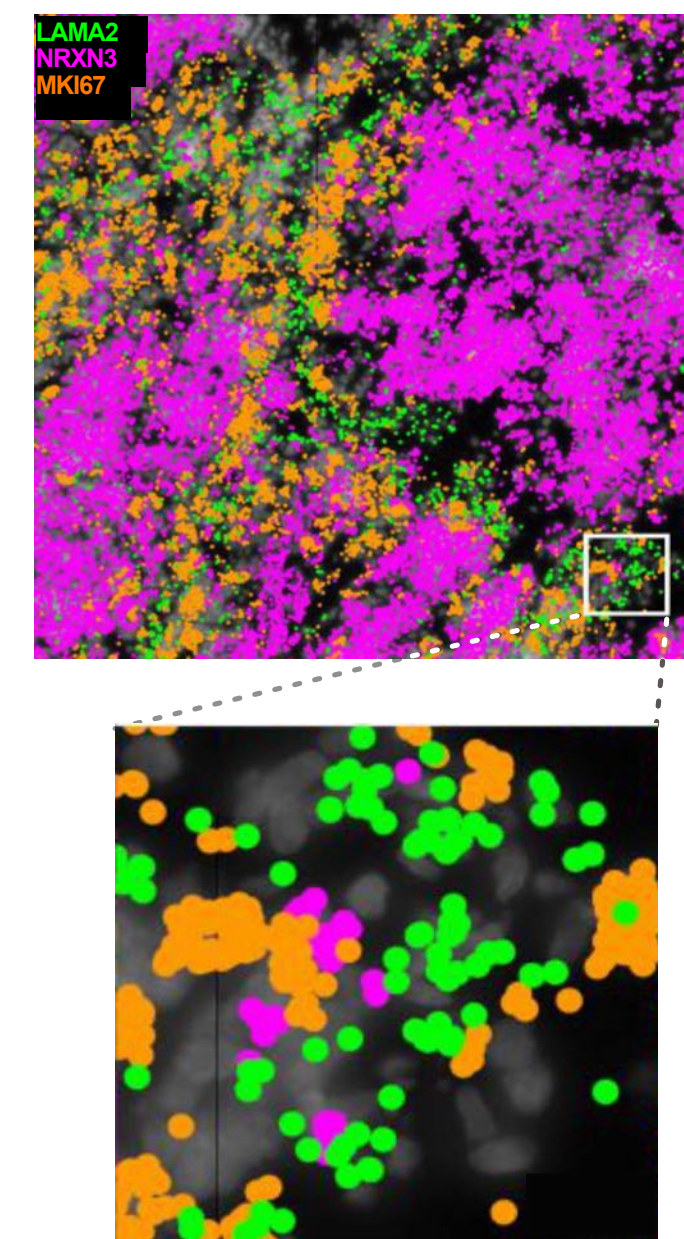
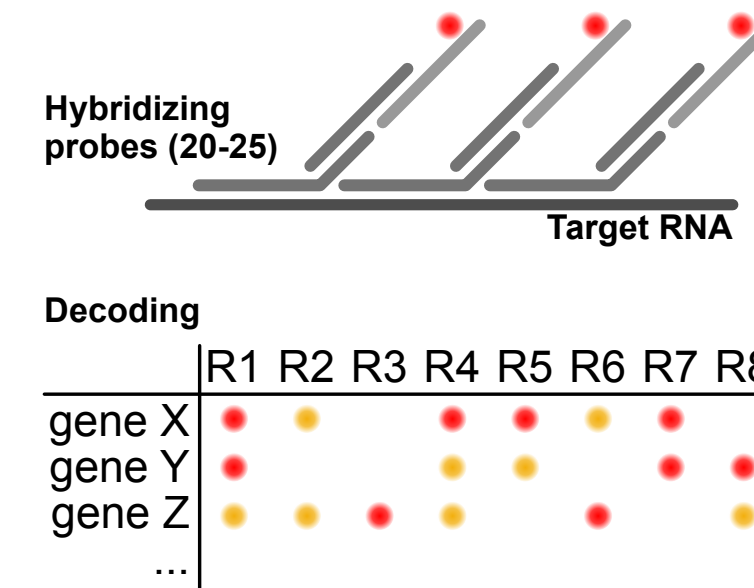
Visium



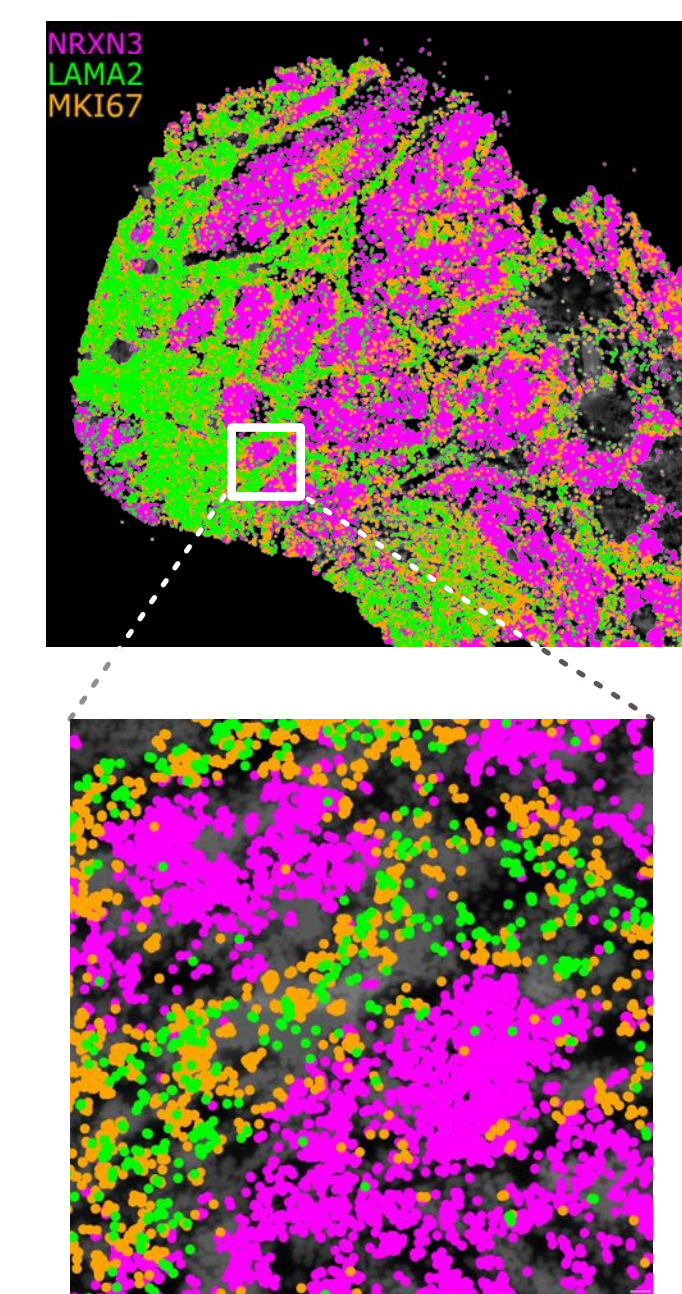
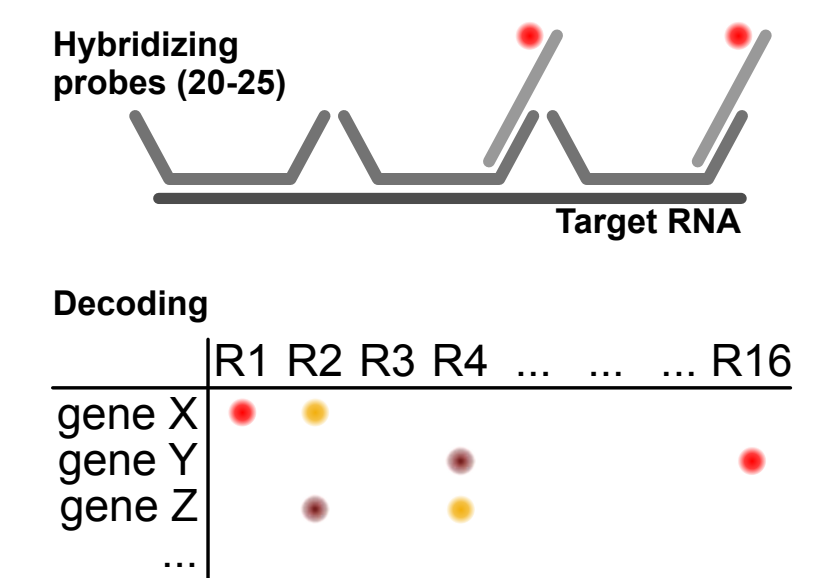
RNAscope
HiPlex



Molecular
Cartography



MERFISH
(Merscope)

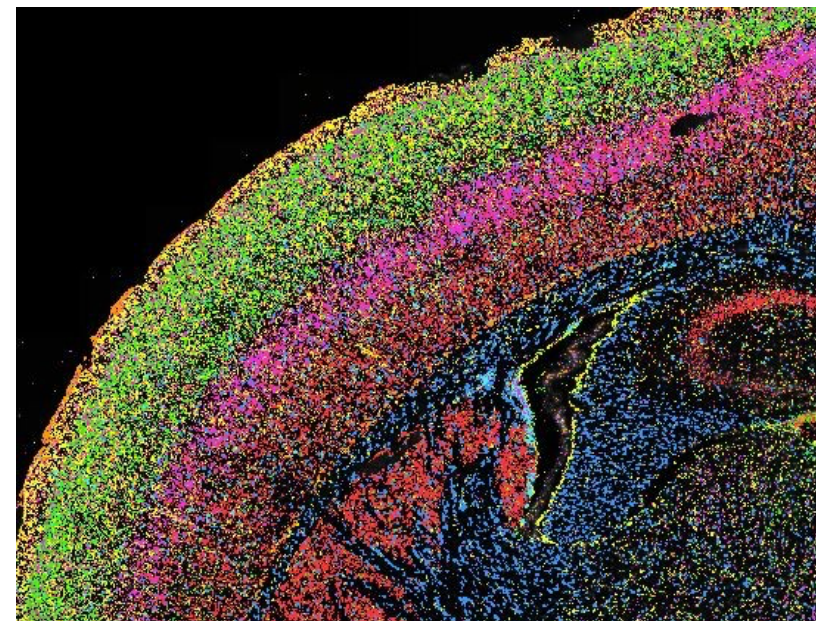


Automated targeted spatial transcriptomics and proteomics

Hybridization-based transcriptomics

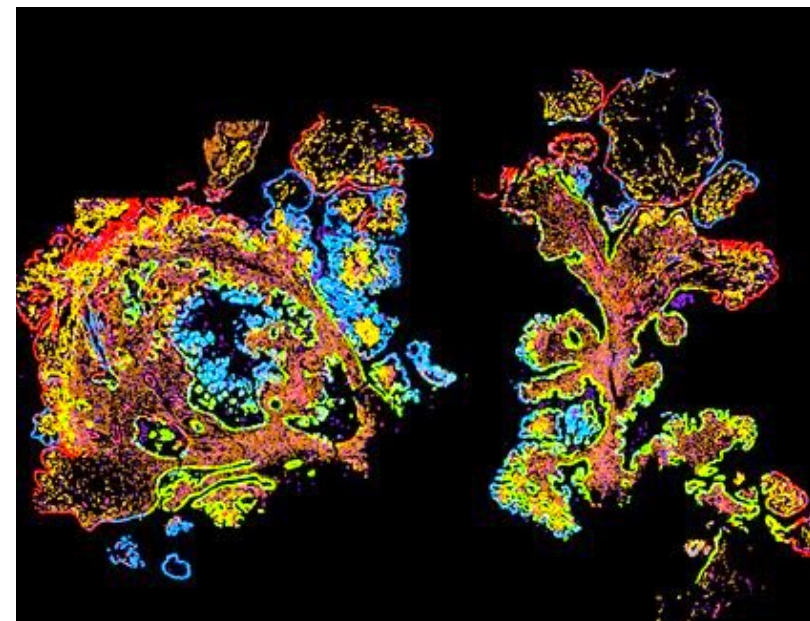
Molecular cartography (Resolve Biosciences)

- smFISH
- 100 targets (200 soon)
- custom gene set
- fresh frozen (FFPE)



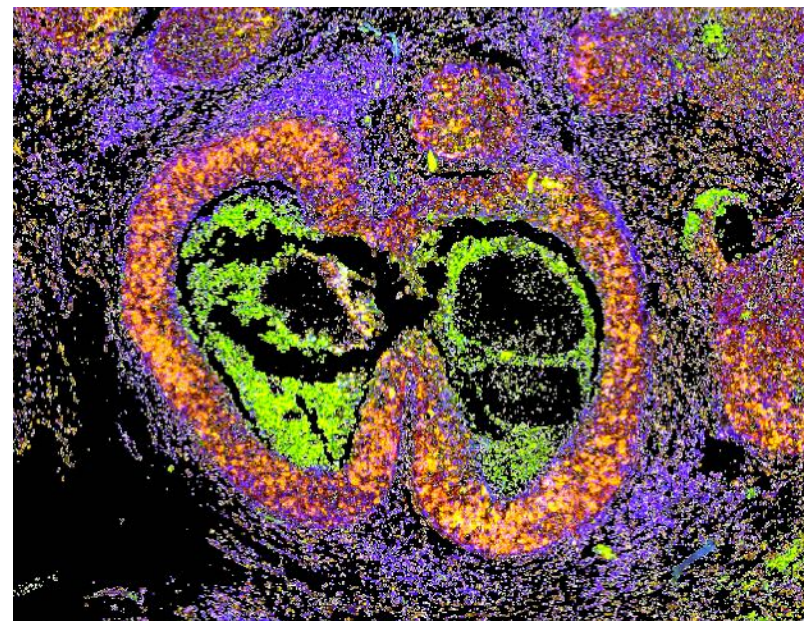
Merscope (Vizgen)

- smFISH
- 140, 300, 500 targets
- custom gene set
- tissue clearing
- fresh frozen & FFPE



Xenium (10x Genomics)

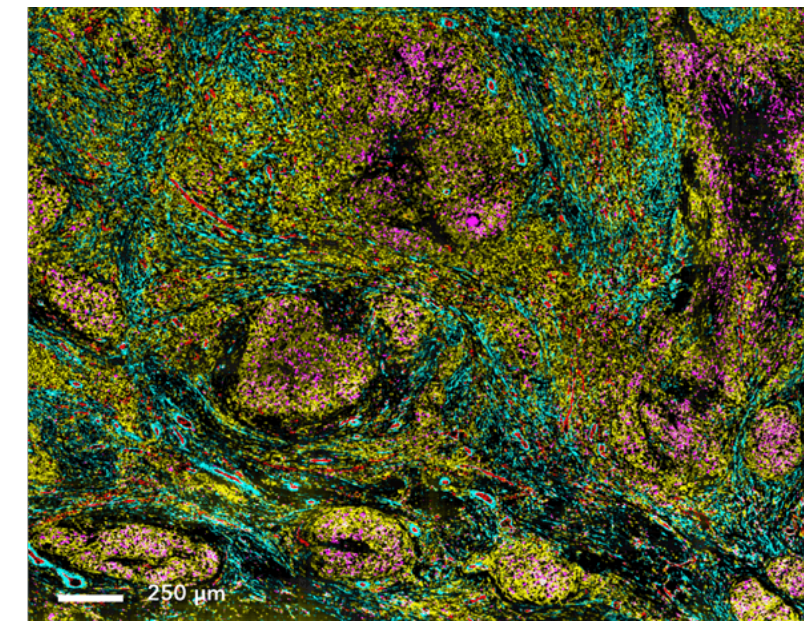
- RNA "in situ sequencing" (Cartana) technology
- 200-400 targets
- fixed/custom gene set
- FFPE & fresh frozen



Antibody-based proteomics

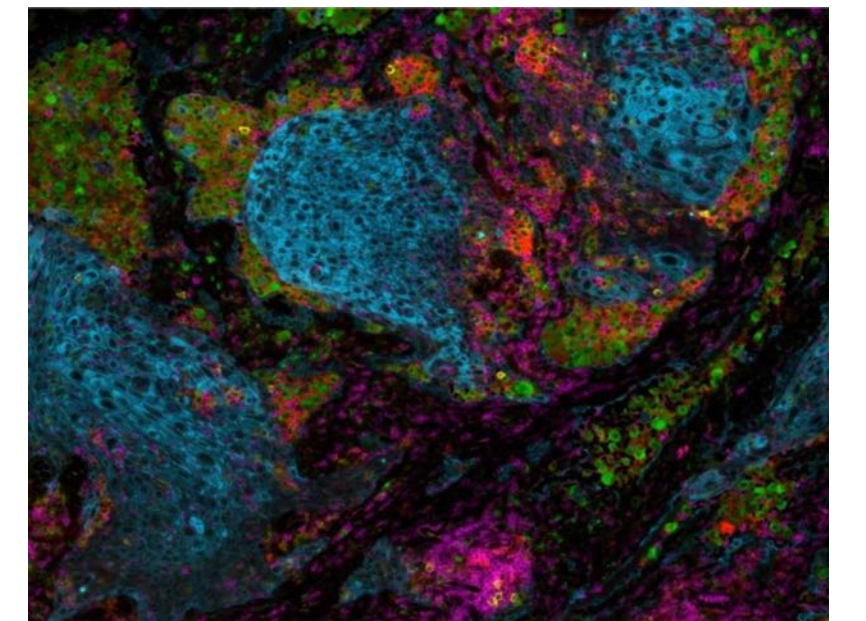
Mibiscope (Ionpath)

- Protein (via mass-spec imaging of Ab)
- ~40 targets
- custom Ab panel
- FFPE & fresh frozen

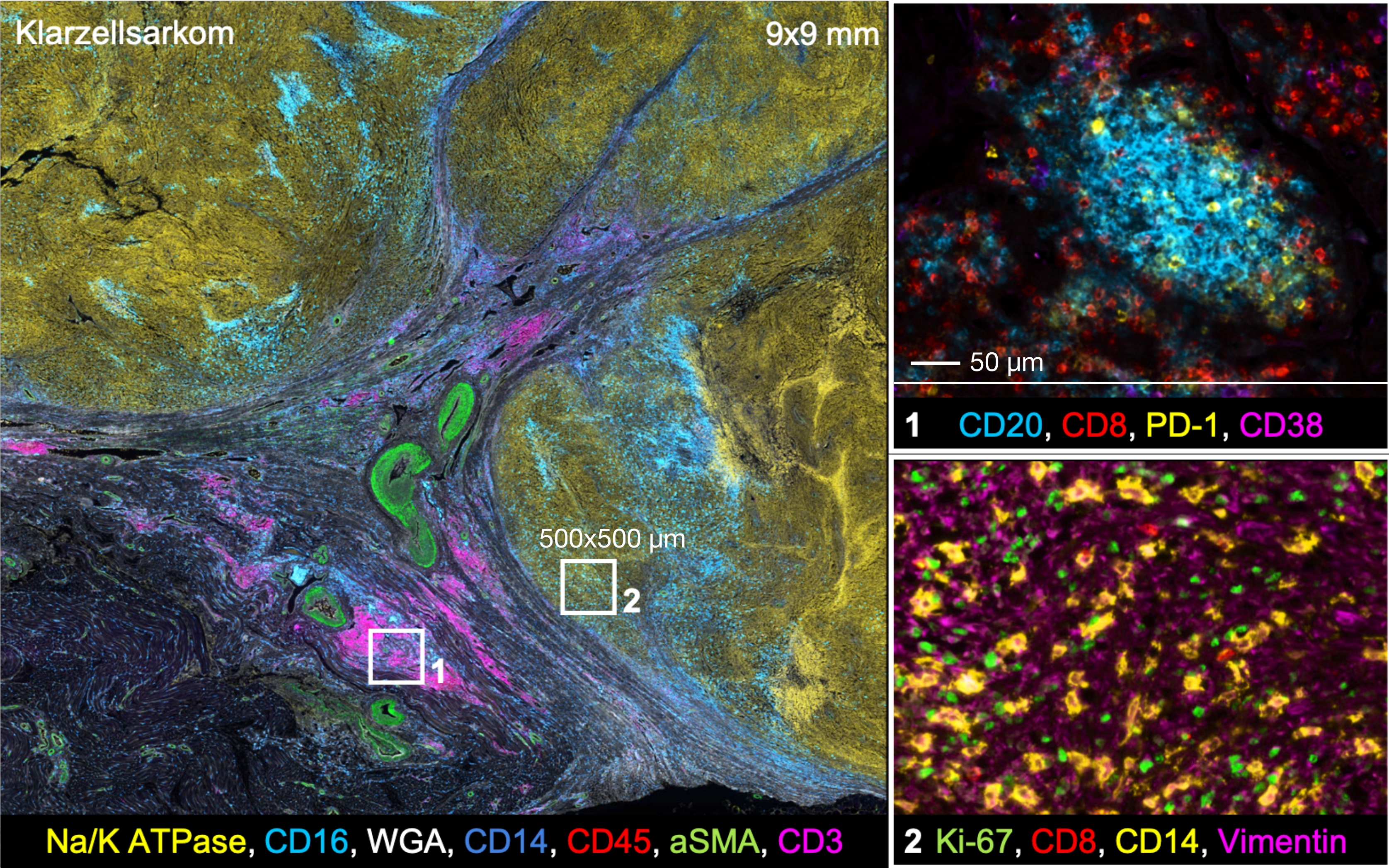
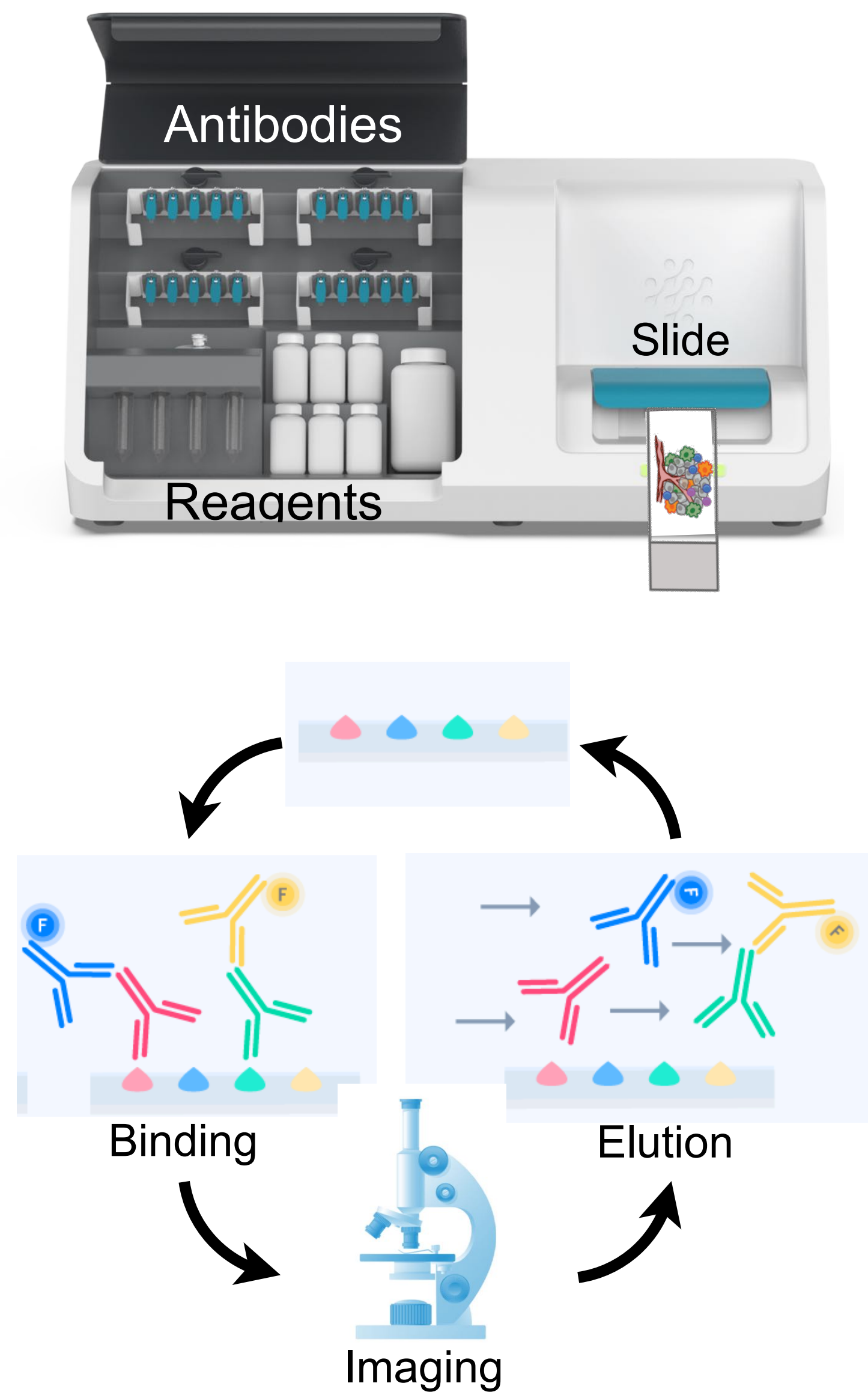


Comet (Lunaphore)

- Protein (via fluorescence imaging of Ab)
- ~40 targets
- custom Ab panel
- FFPE & fresh frozen

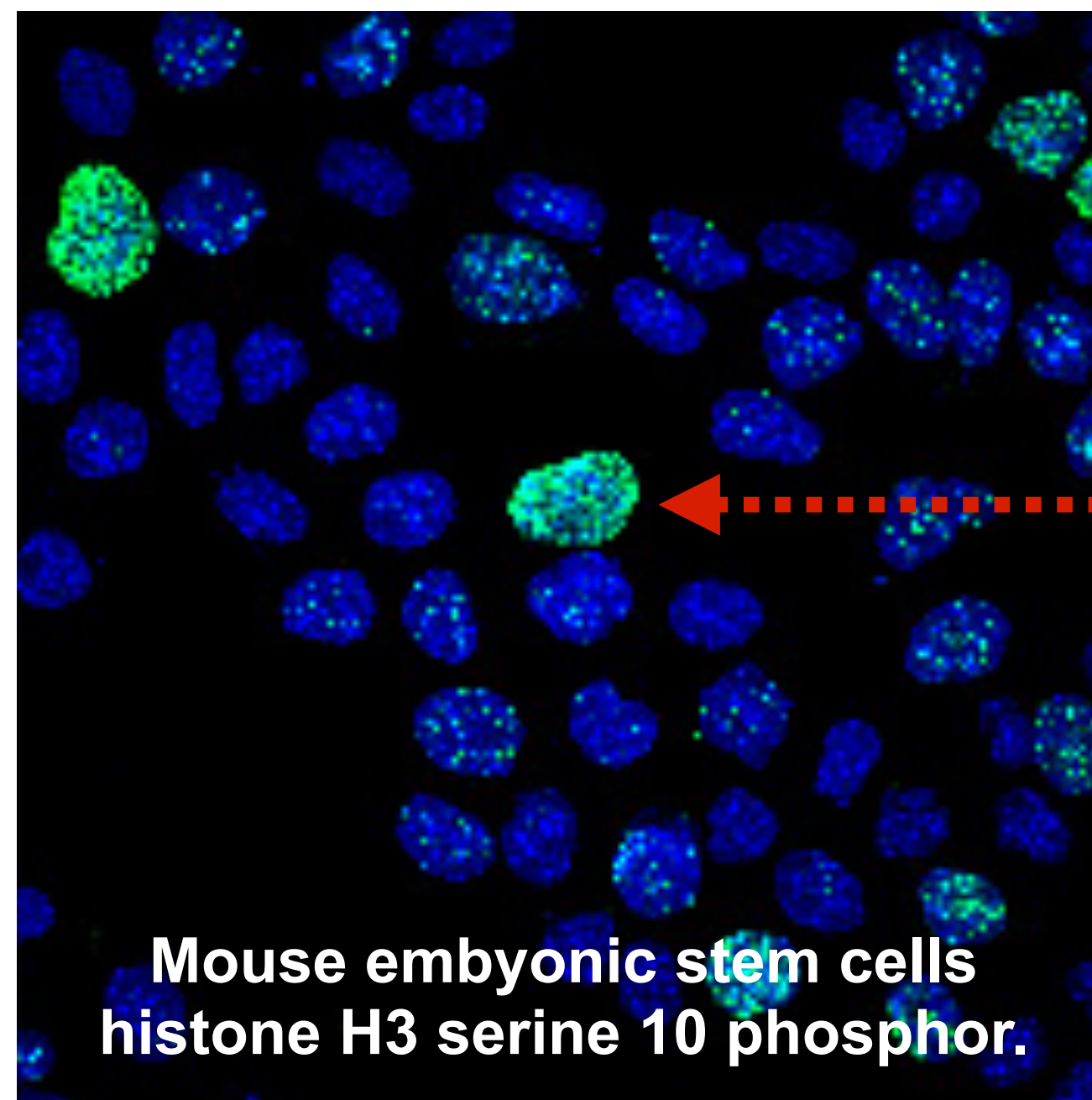


Sarcoma immunoenvironment (FFPE) analysis on Lunaphore Comet system

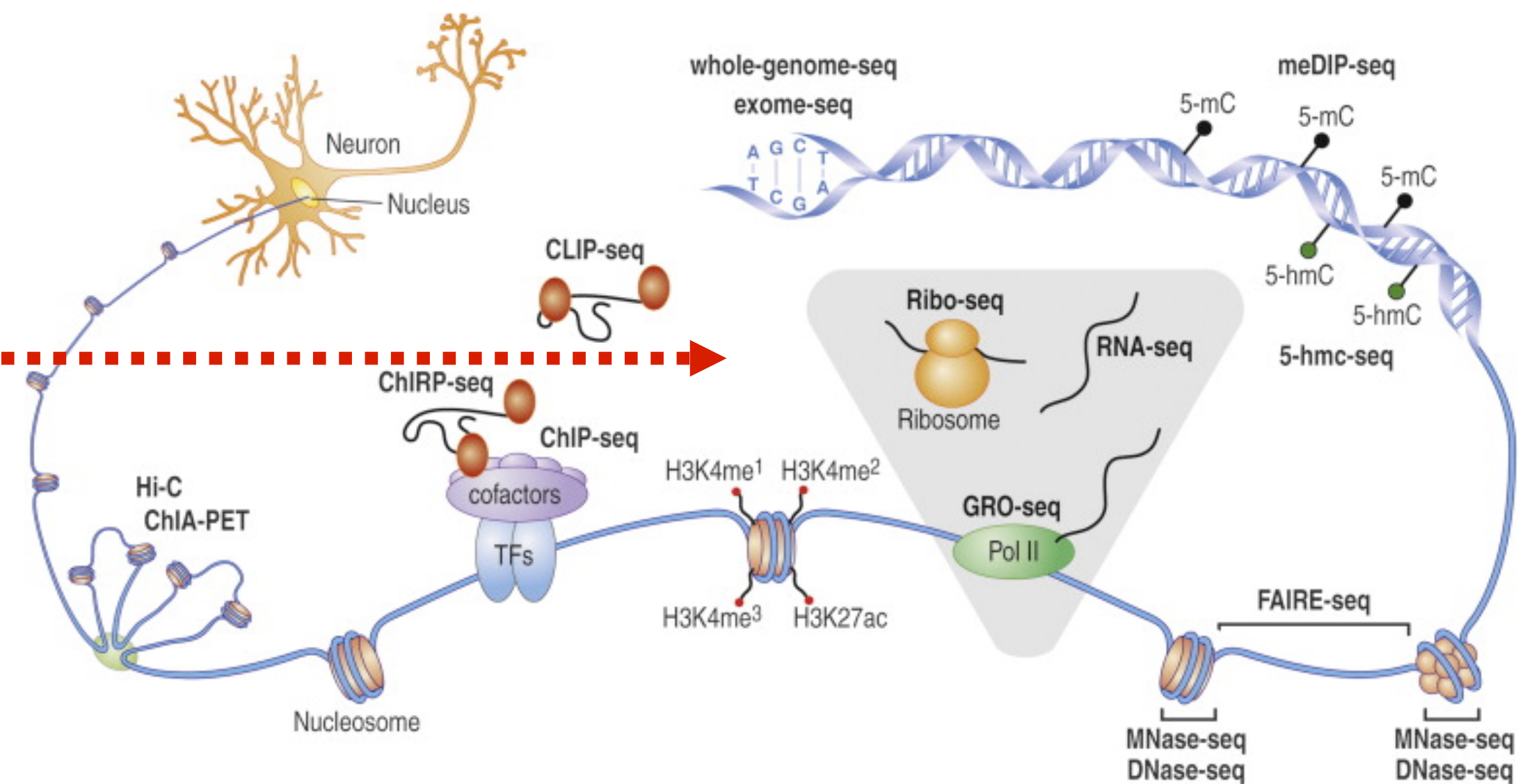


Spatial omics integrates imaging and sequencing based analysis

Fluorescence microscopy



Sequencing



The future of studying genome structure-function relationships

