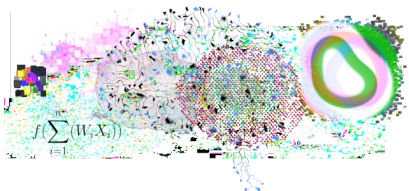


# Advances in Spatial Transcriptomics and Computational Approaches

Guillaume MARCY

*AI / ML for the analysis of single-cell spatial transcriptomics (UCBL Lyon) – 15<sup>th</sup> October 2025*



Université Claude Bernard

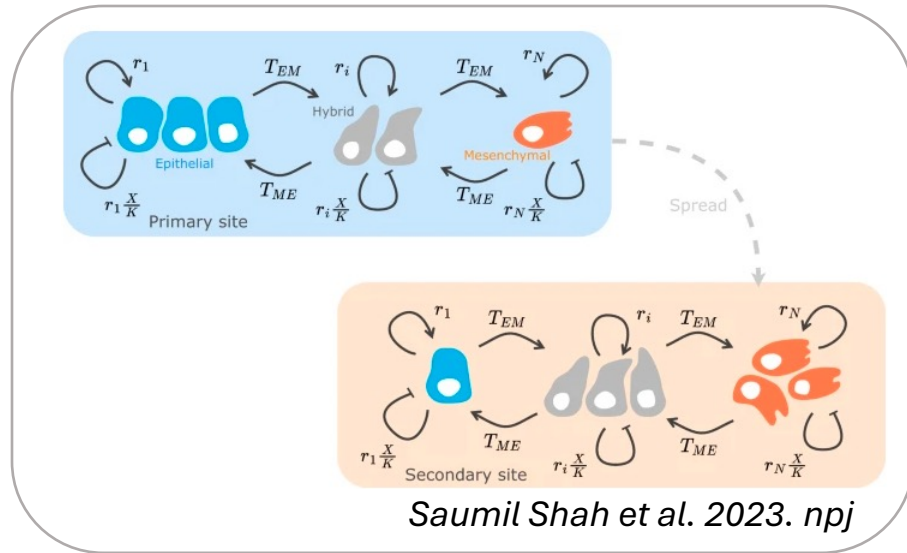


Lyon 1



**LABEX  
CORTEX**  
UNIVERSITÉ DE LYON

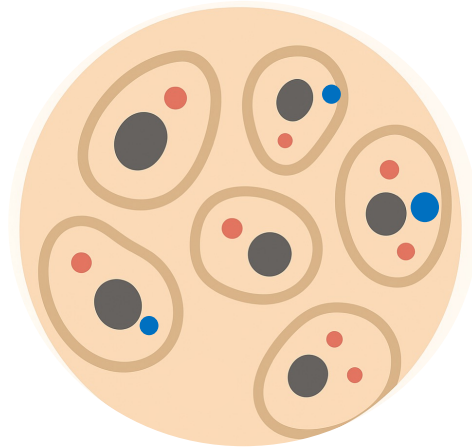
# Why is Spatial Transcriptomics (ST) Important ?



★★ Position of a cell relative to its cellular or non-cellular environment influences its exposition to signals

➔ Cellular phenotype

➔ Function of the tissue



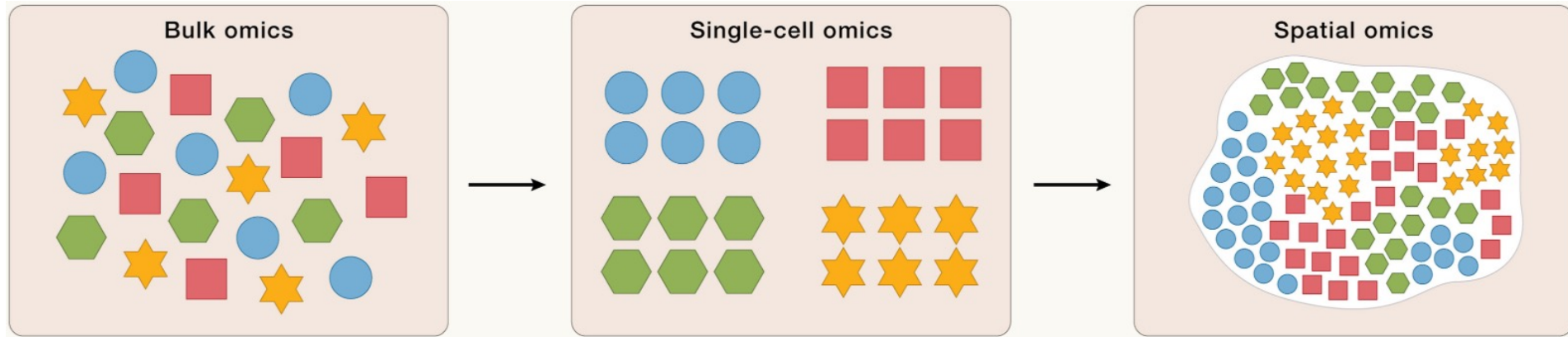
★★ Sub-cellular localization of mRNAs can vary depending on the function of the gene

➔ Target proteins to a specific location

**ST aims to count the number of transcripts of each gene at distinct spatial locations in a tissue**

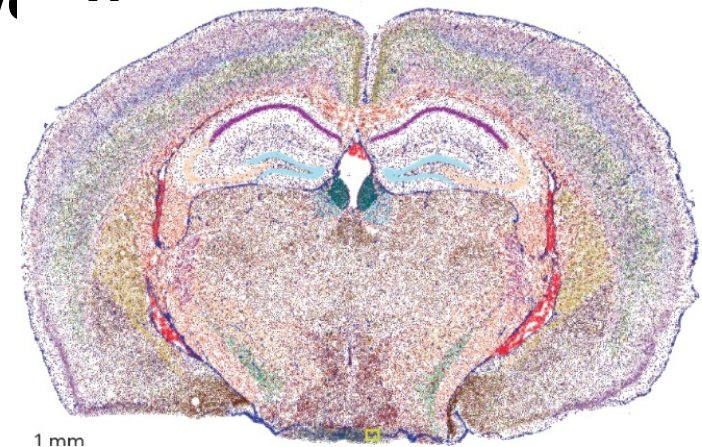
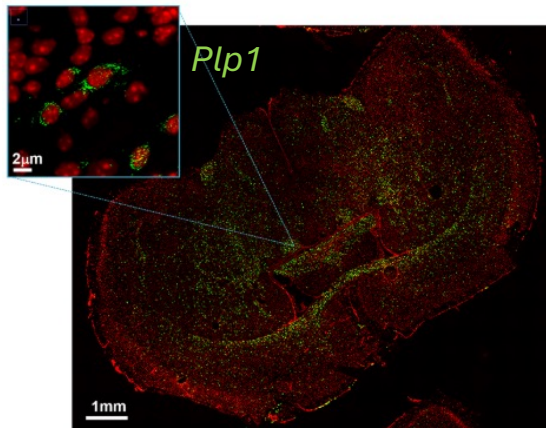
# How Does it Works ?

## 1 Sequencing-based ST technologies – From high throughput sequencing world



*Liu et al. 2024. Cell Review. Spatiotemporal omics for biology and medicine.*

## 2 Imaging-based ST technologies – From imaging with



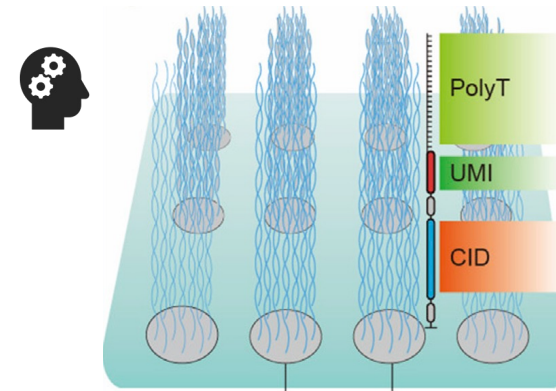
*Salas et al. 2025. Xenium*



# How Does it Works ?

## 1 Sequencing-based ST technologies – From high throughput sequencing world

- ★★ **Extract** mRNAs from tissue while keeping the spatial information
- ★★ mRNAs are profiled using **sequencing**



## 2 Imaging-based ST technologies – From imaging world

- ★★ **Imaging** mRNAs directly within tissue via microscopy
  - ★★ Hybridization of mRNAs to fluorescently labelled probes
  - ★★ Direct sequencing of barcodes associated to amplified mRNAs



Merscope by Vizgen



Xenium by 10X Genomics 3/18

# Different Approaches for Different Answers

**? How many genes ?**

**? What gene capture efficiency ?**

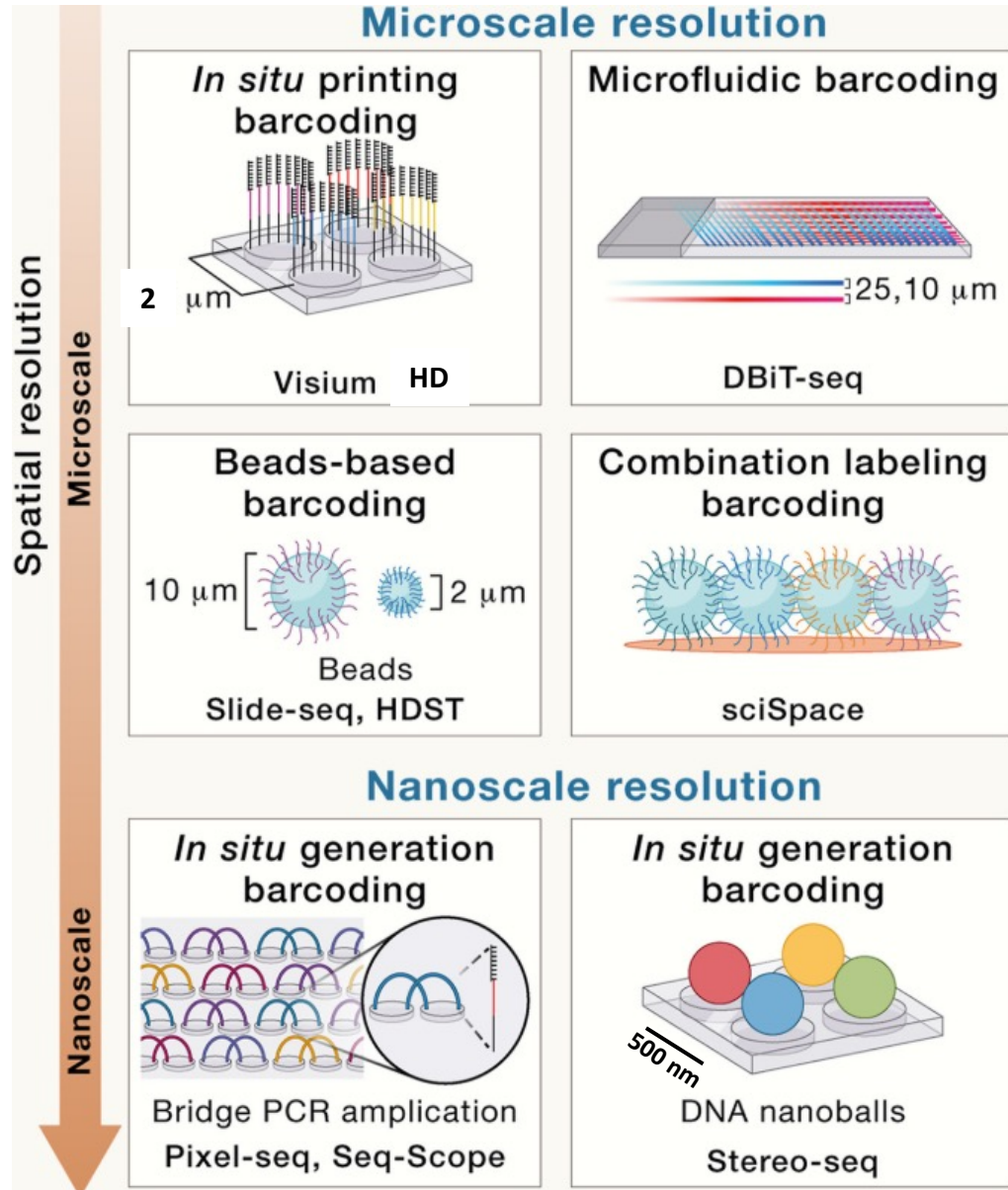
**? Which kind of genes ?** ➡ PolyA mRNAs, microRNAs, lncRNAs ...

**? What spatial resolution ?** ➡ Single cells versus anatomical regions/domains

**? What is the size of the tissue ?** ➡ Full embryo versus dissected regions

**? Which kind of samples ?** ➡ Fresh Frozen (FF), Bioarchives/biopsies (FFPE)

# Sequencing-based Spatial Transcriptomics



## Common steps

★★ concept of scRNA-seq library + **spatial barcode**

★★

spatially barcoded mRNAs are **sequenced**

★★

reads contain spatial barcode + gene sequence



**Spatial location x gene matrix**



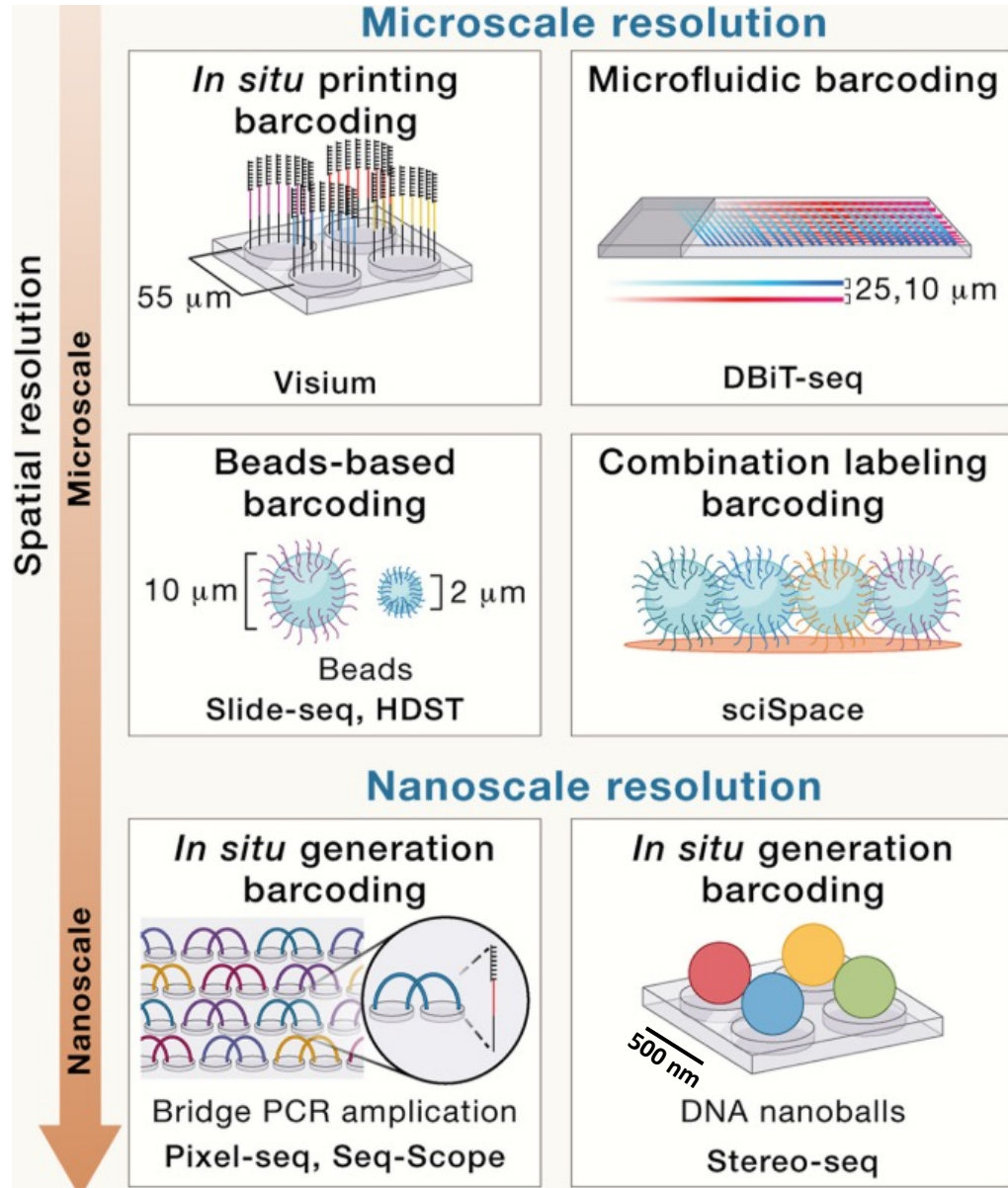
★★ Full transcriptome



★★ Low capture efficiency

★★ Single cell Resolution

# Sequencing-based Spatial Transcriptomics

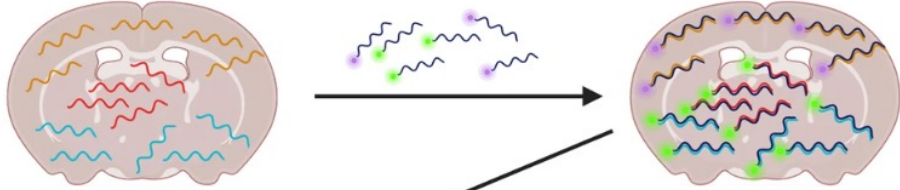


	VisiumHD	Stereo-seq	Slide-seq (Curio seeker)
<b>Company</b>	10X Genomics	STOMICS	Takara/Curio
<b>Sample type</b>	FF & FFPE	FF & FFPE	FF
<b>Species</b>	Agnostic (FF) Hu & Mu (FFPE)	Agnostic	Agnostic
<b>Area</b>	6.5 x 6.5 mm	10 x 10 mm + Larger & custom	10 x 10 mm
<b>Resolution</b>	2 $\mu\text{m}$	0.5 $\mu\text{m}$	10 $\mu\text{m}$
<b>Bin size</b>	8 and 16 $\mu\text{m}^2$ & SC	Custom>SC	10 $\mu\text{m}^2$
<b>Capture efficiency /10<math>\mu\text{m}^2</math></b>	250 genes	300 genes (Up to 1500 genes/Neuron)	250 genes

# Imaging-based Spatial Transcriptomics

## ISH

1. Hybridise labelled probes to target mRNA



3. Image fluorophore locations

4. Repeat  $n$  times to generate gene-specific fluorophore barcode

## ISS

1. Rolling circle amplification of target transcripts

2. Hybridise short, labelled probes to determine 1-2nt of transcript's sequence



3. Image fluorophore at each location

4. Repeat as necessary to determine mRNAs' sequences

## Common steps

- ★★ **design** oligonucleotide probes targeting predefined set of genes through hybridization
- ★★ quantification of transcripts via **microscopy imaging**



**Spatial location x gene matrix**



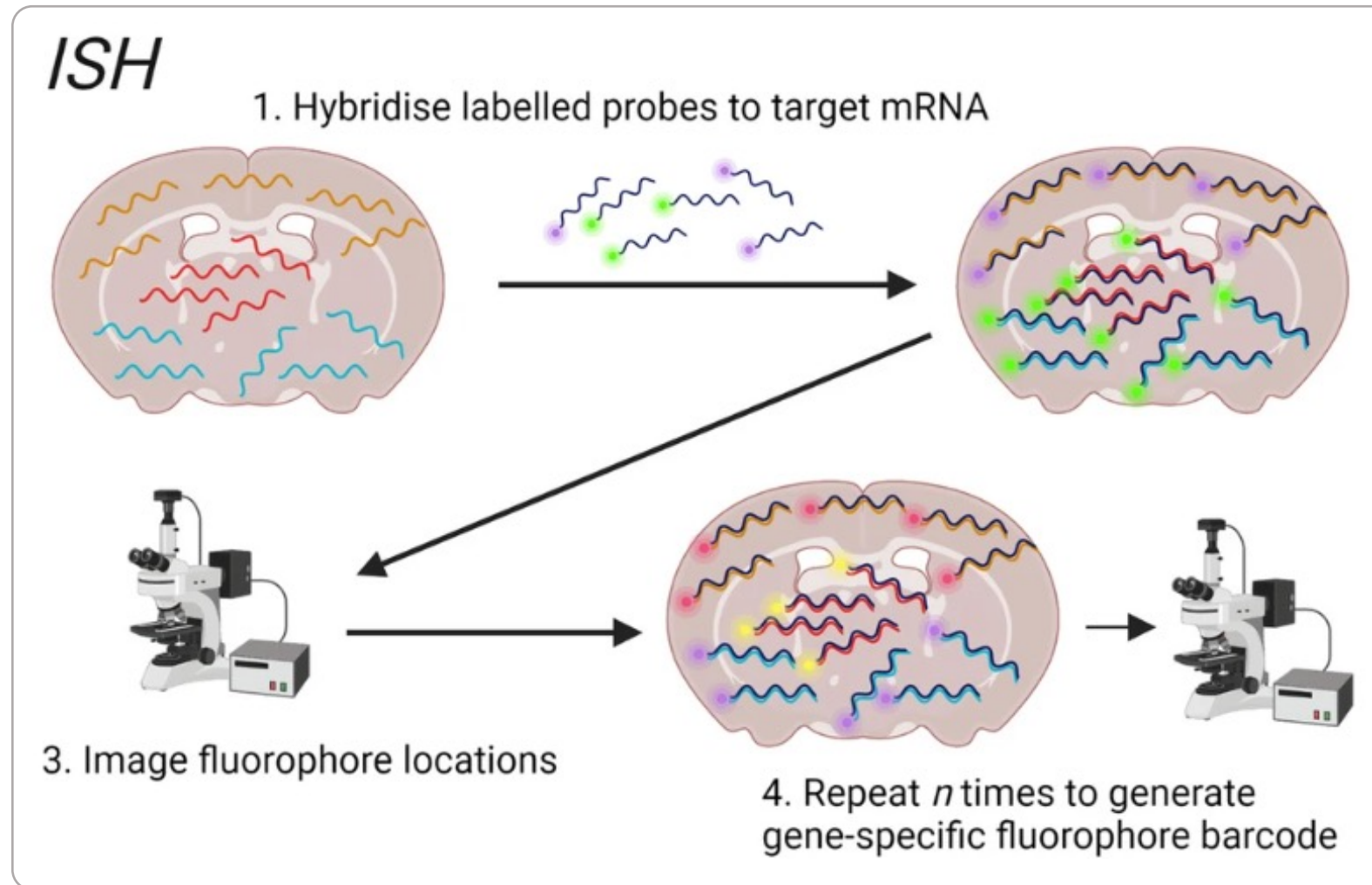
- ★★ All RNAs, not only mRNA
- ★★ Subcellular resolution



- ★★ Targeted panel of genes



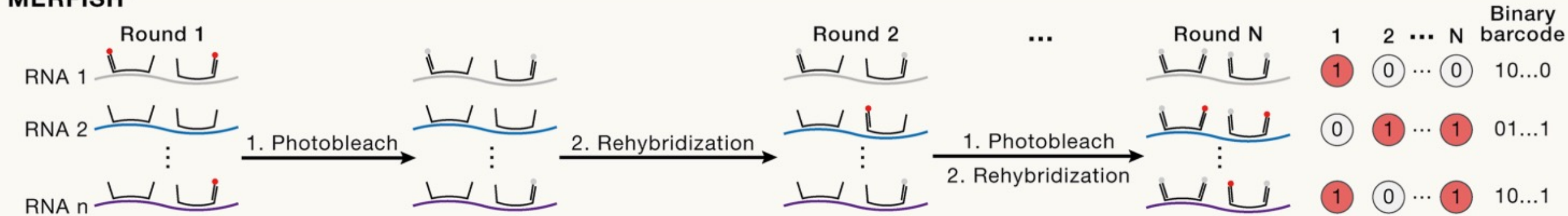
# Imaging-based Spatial Transcriptomics – ISH technique



**Hybridize RNAs to fluorescently labelled gene-specific probes**

# Imaging-based Spatial Transcriptomics – ISH Technique

## MERFISH



Liu et al. 2024. Cell Review. Spatiotemporal omics for biology and medicine.

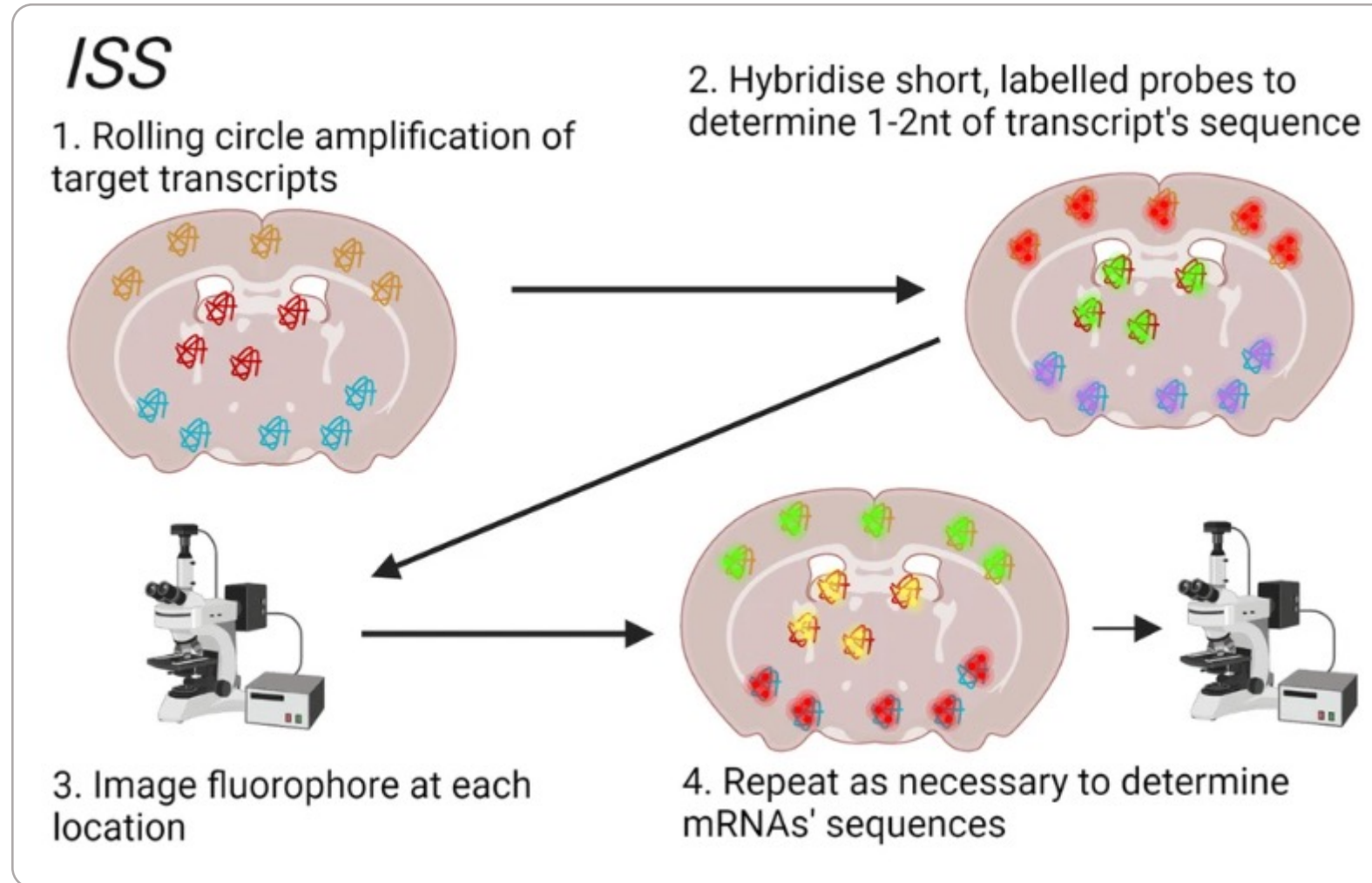
1. Each RNA is assigned a unique binary barcode e.g. Gene A : 0110100101011
2. Each RNA is targeted by several designed probes that contains 2 regions :
  - ★★ complementary RNA sequence
  - ★★ readout barcode sequences
3. Barcodes are read using sequential FISH rounds



- ★★ All RNAs, not only mRNAs
- ★★ Subcellular resolution

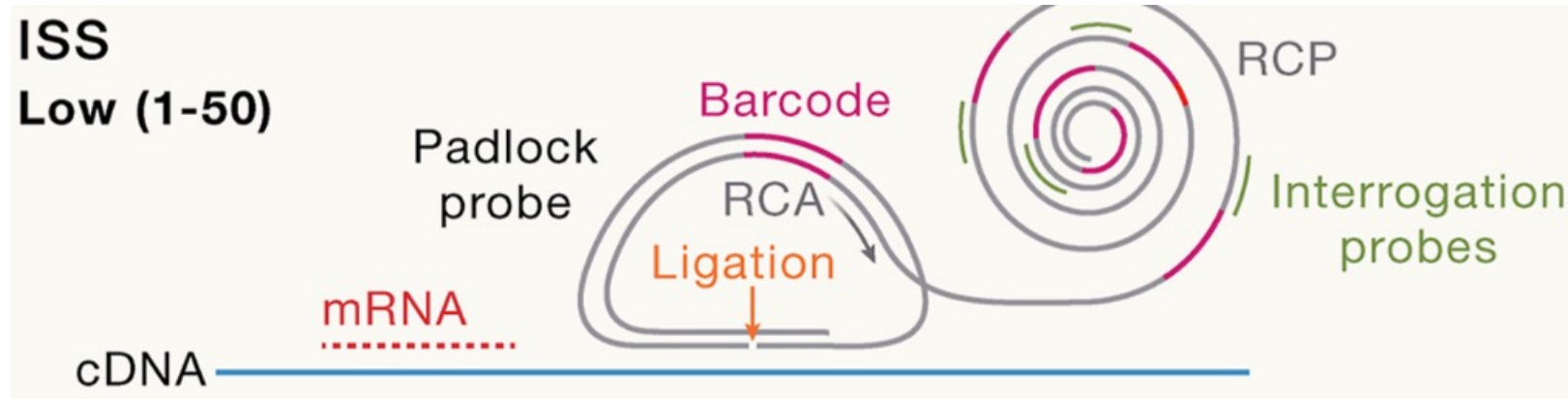
- ★★ Targeted, panel of genes

# Imaging-based Spatial Transcriptomics – ISS Technique



**Directly sequence amplified mRNAs inside the tissue section by sequencing by ligation technique**

# Imaging-based Spatial Transcriptomics – ISS Technique



*Liu et al. 2024. Cell Review. Spatiotemporal omics for biology and medicine.*

1. reverse transcription of the RNA
2. hybridization of padlock probes to the cDNA, containing a unique corresponding barcode
3. Rolling circle amplification (RCA) to produce many copies of the barcode
4. Read the barcode through imaging the sequence of the barcode with fluorescently labelled probes targeting nucleotides



★★ All RNAs, not only mRNA  
★★ Subcellular resolution

★★ Targeted, panel of genes



# Beyond Spatial Transcriptomics

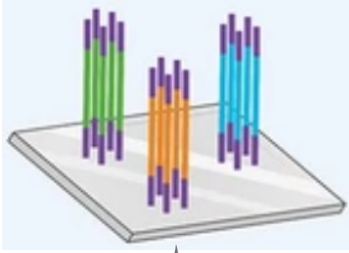
- ★★★ Spatial ATAC-Seq (*Deng et al. 2022*) → Chromatin accessibility
- ★★★ Spatial Cut&Tag (*Deng et al. 2022*) → Epigenetic landscape
- ★★★ Spatial Proteomics → Macsima (Miltenyi), PhenoCycler (Akoya)...
- ★★★ Spatial Metabolomics (*Ganesh et al. 2021*) → Metabolites quantification (mass spectrometry)

## Multi-modalities

- ★★★ Spatial ATAC-Seq + RNA expression (*Zhang et al. 2023*)
- ★★★ Spatial Cut&Tag + RNA expression (*Zhang et al. 2023*)
- ★★★ Spatial CITE-seq : Proteomics + RNA expression (*Liu et al. 2023*)

# ST Data Analysis

## 1 Sequencing-based ST technologies



Alignment and Mapping

## 2 Imaging-based ST technologies

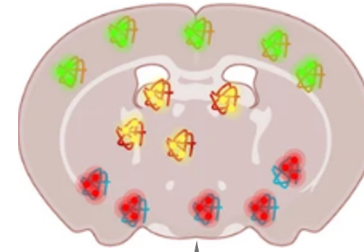


Image processing and RNA identification

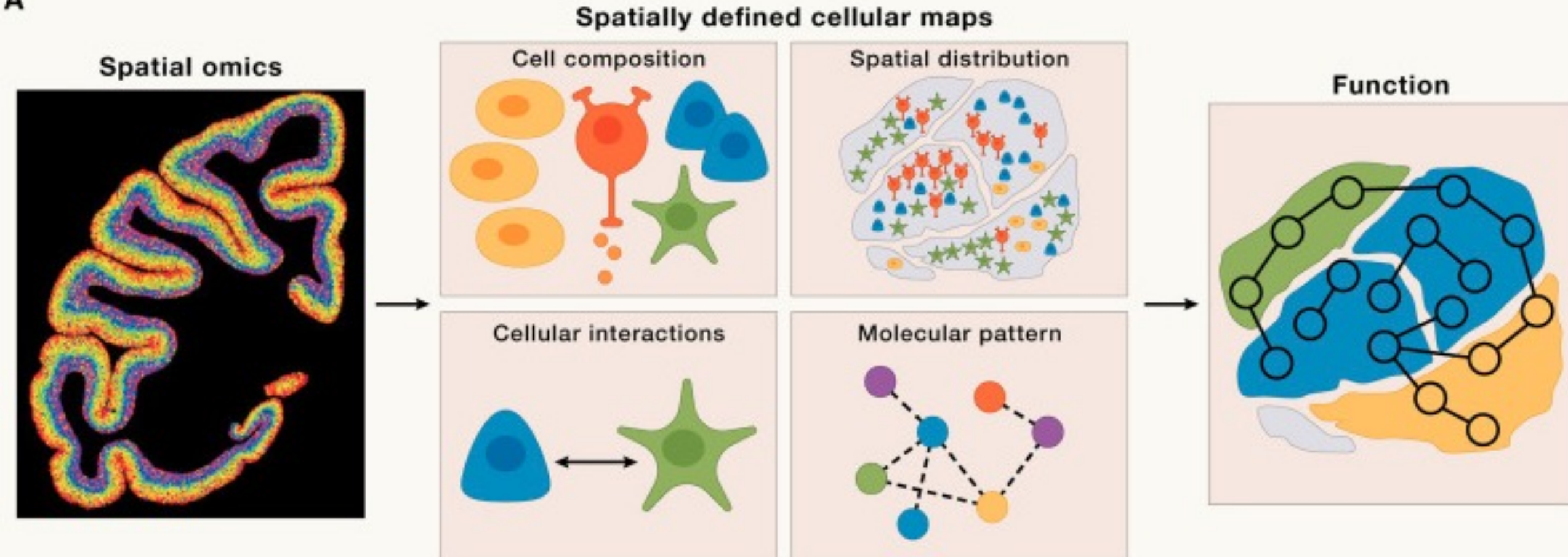
Spatial location x Gene matrix (.h5 file)

	spot1	spot2	spotn
Gene1	3	2	13
Gene2	2	3	1
Gene3	1	14	18
...	.	.	.
...	.	.	.
...	.	.	.
GeneM	25	0	0

1. QC filtering
2. Normalization (log1p, scTransform, SpaNorm...)

# What Do you Do with ST Data Analysis ?

A



*Liu et al. 2024. Cell Review. Spatiotemporal omics for biology and medicine.*

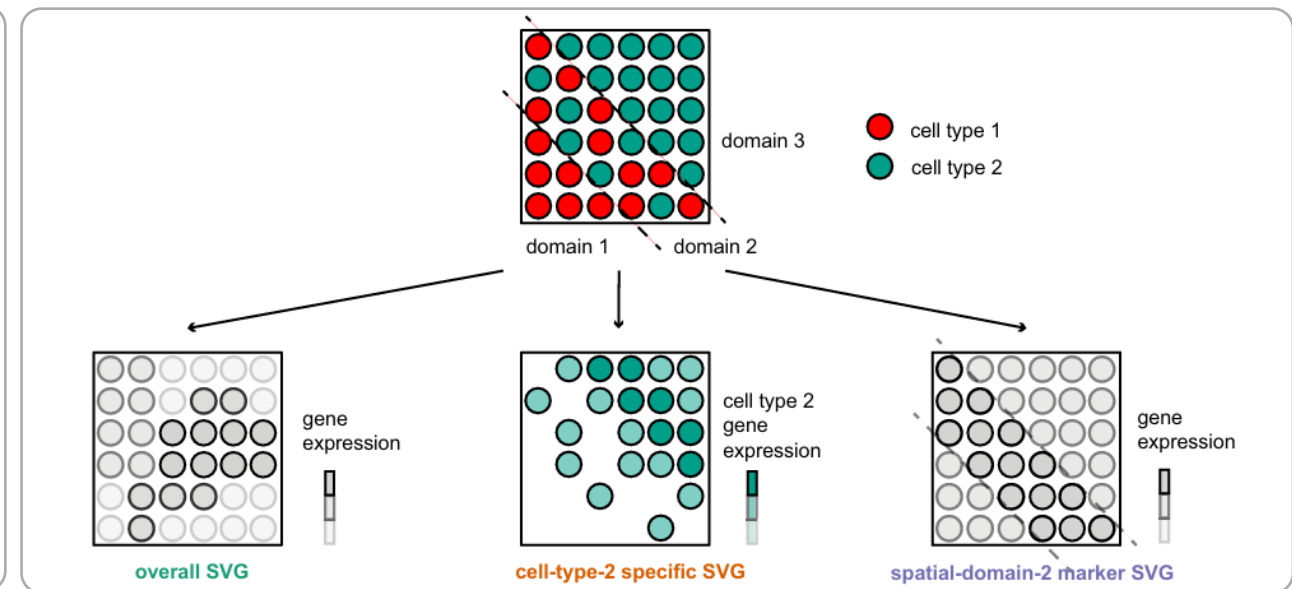
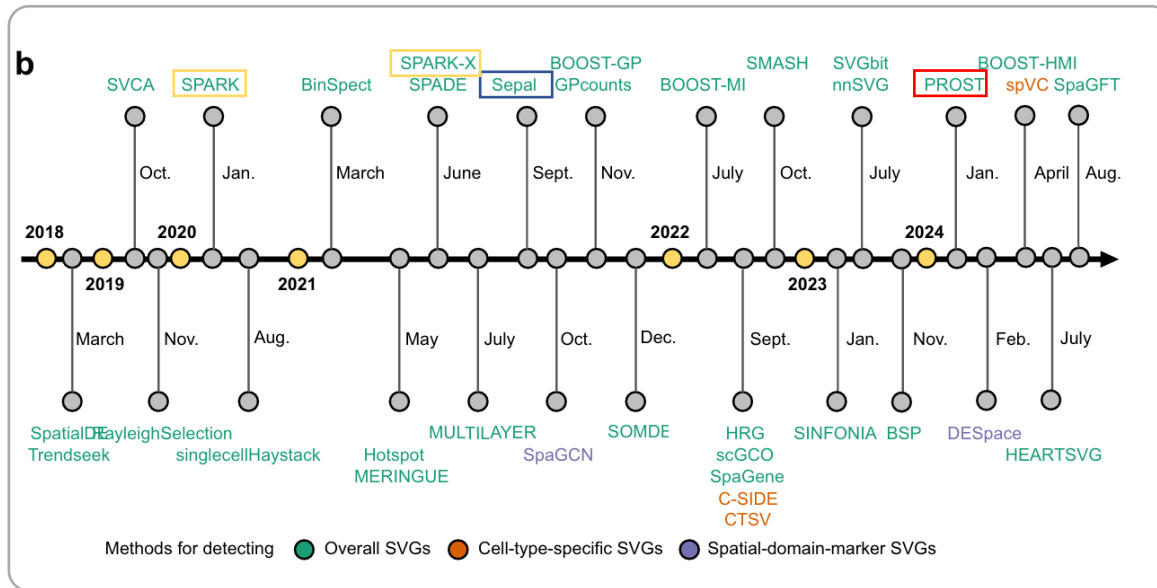
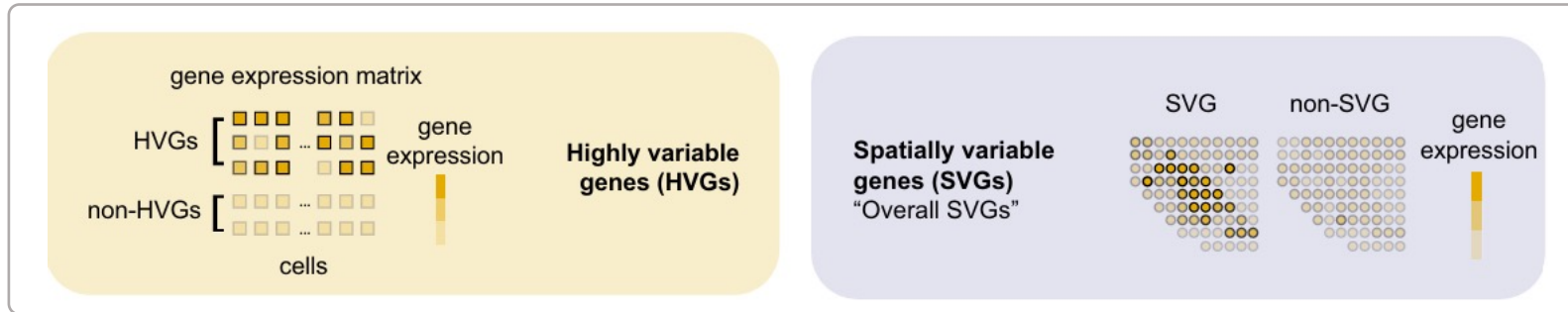
**Tissue/Region  
Level**

**Cellular Level**

**Molecular Level**

# Spatial Variable Genes Detection

Molecular Level




Guanao Yan et al. 2025. Review Nature Communication




Imaging-based technologies




Imaging-based ST technologies achieve a limited number of genes from the entire transcriptome. This restricted gene coverage limits the comprehensive understanding of the molecular landscape of the tissue.

★★★ mapping single-cells onto the ST data  **Integration**

★★★ imputing spatial gene expression from scRNA-seq data  **Prediction**

★★★ **Mapping tools** : Seurat, LIGER, Harmony, Tangram, ENGEF, Tissue, ENGEF etc.

 lack specific optimizations for spatial gene expression

★★★ **Imputing tools** : SpaGE, stPlus, novoSpaRc, stDiff, SpatialScope, iss\_patcher, SPRITE, **SpaIM**, **TransImpLR**, **stAI**, **TransImpute**, **SpotDiff**, **spRefine**

 Actively developing area

# Cellular Segmentation

Cellular Level

★ Detected mRNAs (by sequencing or imaging) need to be attributed to individual cells

➡ Cellular segmentation on staining image

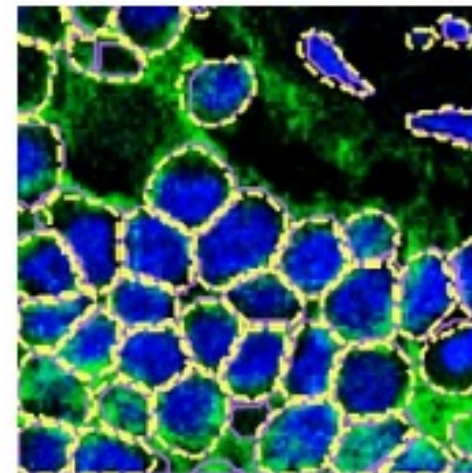
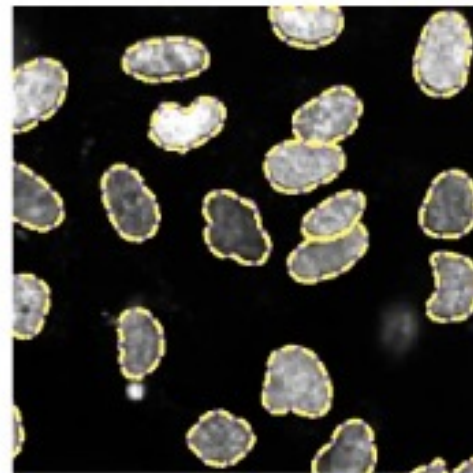
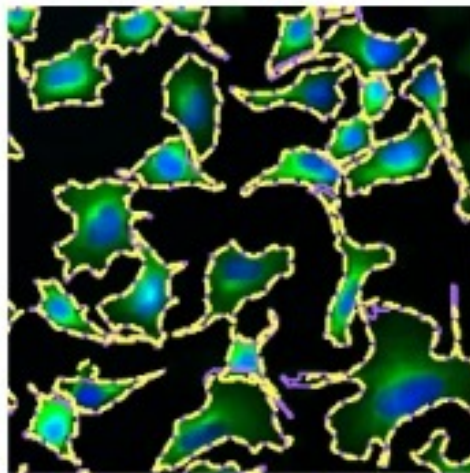
★★ some commercially available ST platforms implemented cell segmentation in their pipeline (VisiumHD, Stereo-seq, Merscope, Xenium) through H&E, DAPI or other stainings

★★ segmentation tools : **CellPose**, **StarDist**

**Baysor**, **SAINSC**

➡ **Staining and Deep Learning**

➡ **Staining free**



*Tringer et al. Cellpose3. 2025.*

★ Microscale ST signals need to decompose cell types mixture for each spot

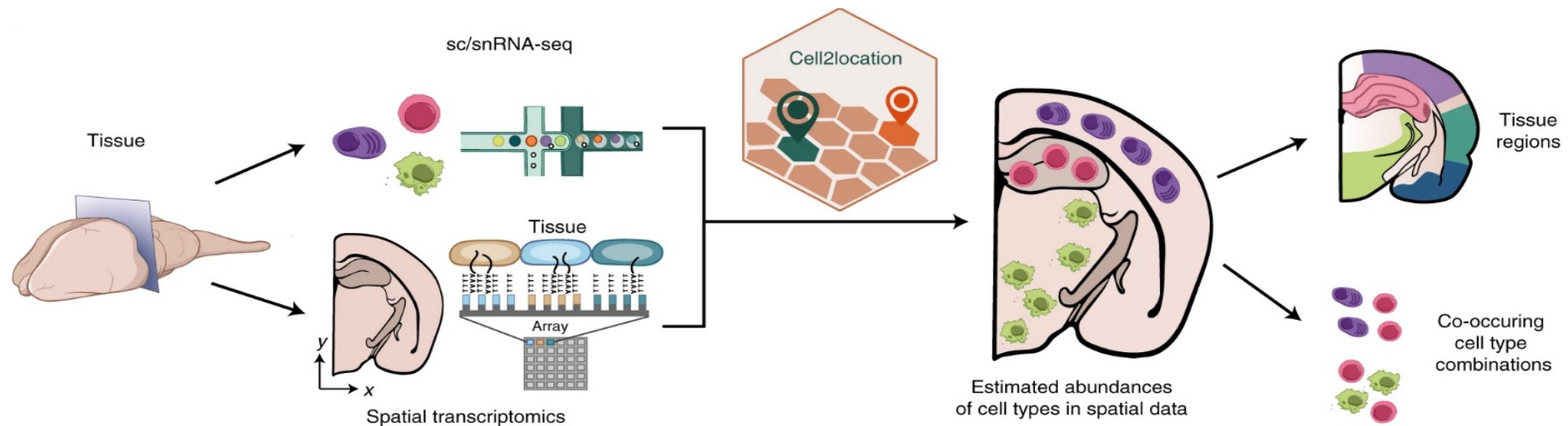
➔ **Methods depend on associated single scRNA-seq**

## ★★ **Cell2location**

1. Estimates cell types reference signatures from scRNA-seq datasets
2. Decomposes the mixture of counts from the reference cell types signature , estimating the cell type abundance at each spatial location

😊 Nanoscale platform

😞 Tissue Thickness



## Region Level

- ➡ **algorithms that look for neighborhoods in both gene expression and physical spaces**

- a**
- Cells in physical space
- Neighbor-augmented expression matrix
- Cells
- Cell  $i$
- Neighbors
- $\sqrt{1-\lambda}$
- $\sqrt{\frac{2\lambda}{3}}$
- $\text{mean}(G(r) \cdot \text{expression})$
- $\sqrt{\frac{\lambda}{3}}$
- $| \text{mean}(e^{i\phi} \cdot G(r) \cdot \text{expression}) |$
- $g_1$   
 $g_2$   
 $g_m$   
 $g_1^{\text{nbr}}$   
 $g_2^{\text{nbr}}$   
 $g_m^{\text{nbr}}$   
 $g_1^{\text{nbr}}$   
 $g_2^{\text{nbr}}$   
 $g_m^{\text{nbr}}$
- Cells' own expression
- Mean expression in local neighborhood
- Magnitude of the expression gradient
- $2\pi$   
 $\phi$   
 $0$
- $y$   
 $x$
- 

16/18



★★★ Unravelling complex cell-cell communications in spatial microenvironments



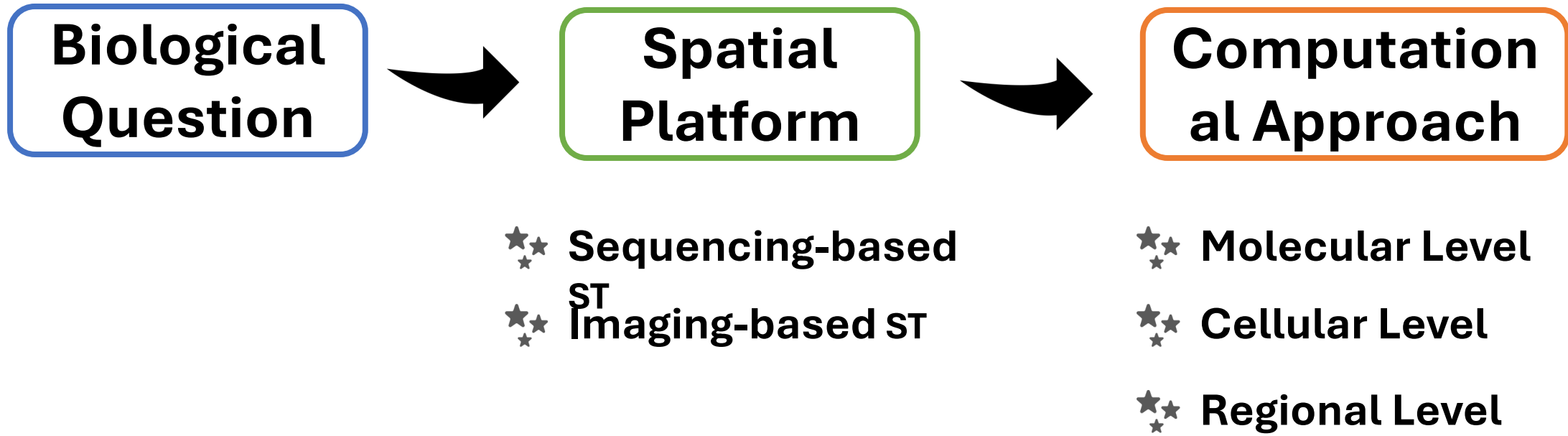
**A lot of tools are available !!**

★★★ tools employ a range of distinct strategies (e.g. statistics-based, expression-based, co-expression-based, correlation-based, network-based, and ML-based)

★★★ popular tools are **NicheNet**, CellPhoneDB, CellChat

★★★ CCC-Catalog (Cesaro et al. 2025) can be found at : <https://sysbiobig.gitlab.io/ccc-catalog/>

# Conclusions



# Acknowledgments



**Maxime Lepetit**



**Cyril Dégletagne  
Carole Audouyraud  
Thibault**

**Laurie Tonon  
Raphael Schneider**



**Severine Croze  
Joel Lachuer**

**Inserm**

La science pour la santé  
From science to health

**Olivier Raineteau (SBRI/INSERM)  
Hugues Berry (Astrocytes/INRIA)  
Sergio Sarnataro (Spatial-Cell-ID)**



*Inria*