Exploring Omics with CAT: A Gateway to Genomic and Spatial Insight

Decoding the Spatial Dimension of Gene Expression:

A Comprehensive Introduction to Spatial Transcriptomics Platforms, Bioinformatics Analysis, and Their Applications in Biological Research.

Bo Young Ahn and **Heewon Seo** Applied Spatial Omics Centre Cumming School of Medicine University of Calgary



June 17, 2025

Envision New Dimension: Spatial Omics

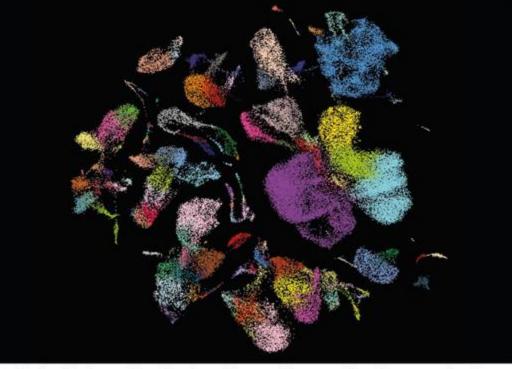
nature

TECHNOLOGY FEATURE | 22 January 2024

Seven technologies to watch in 2024

Advances in artificial intelligence are at the heart of many of this year's most exciting areas of technological innovation

By Michael Eisenstein



A cell atlas of the human lung describes different cell types and how they are regulated.

Cell atlases

If you're looking for a convenient cafe, Google Maps can find nearby options and tell you how to get there. There's no equivalent for navigating the much more complex landscape of the human body, but ongoing progress from various cell-atlas initiatives – powered by advances in single-cell analysis and 'spatial omics' methods – could soon deliver the tissue-wide cellular maps that biologists crave.

"Maybe it's not the end for the spatial resolution limits. It might get better."





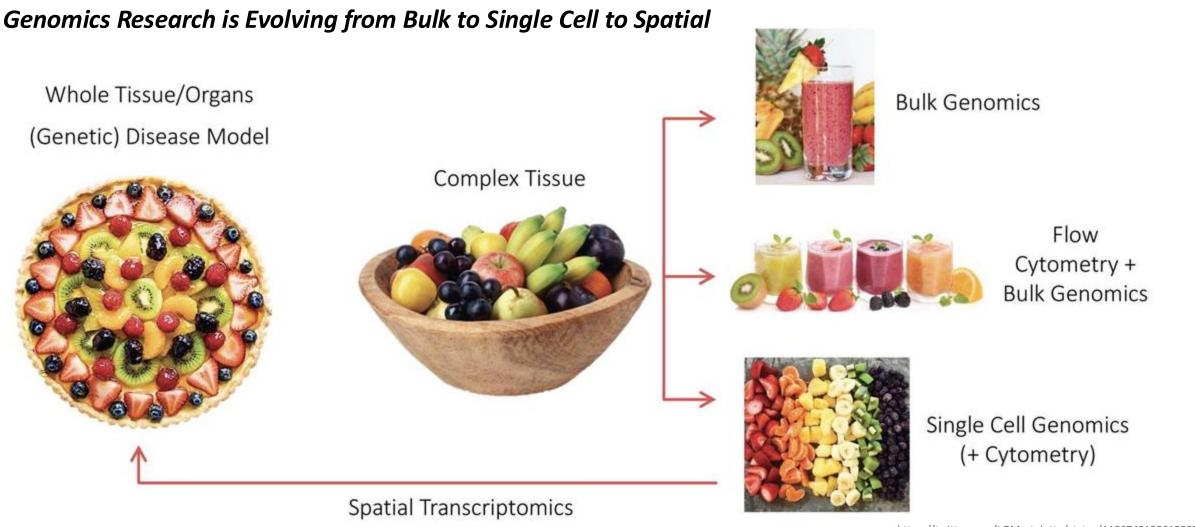
What is Spatial Omics?

- Spatial Omics includes:
 - Spatial Transcriptomics
 - Spatial Proteomics

 Spatial Omics is an overarching term for all methods that assign transcriptomics/proteomics data to the original location within the tissue.



Spatial Transcriptomics in Layman's Term I



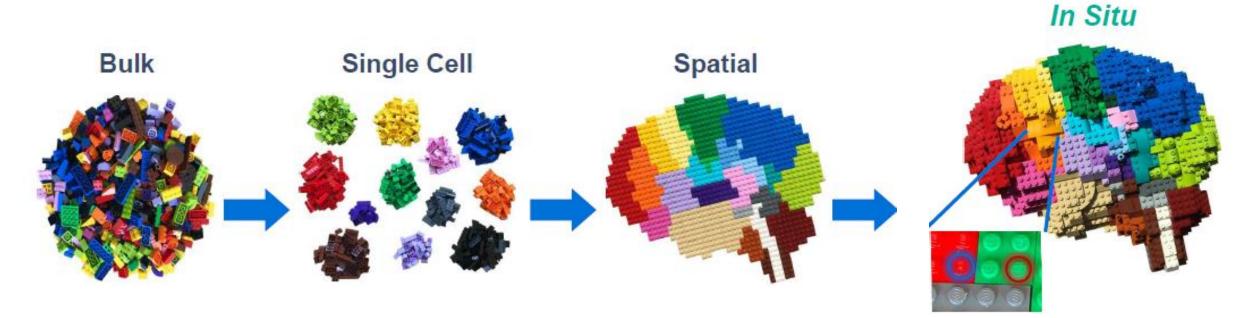
What is Spatial Omics

https://twitter.com/LGMartelotto/status/1186745128615985152



Spatial Transcriptomics in Layman's Term II

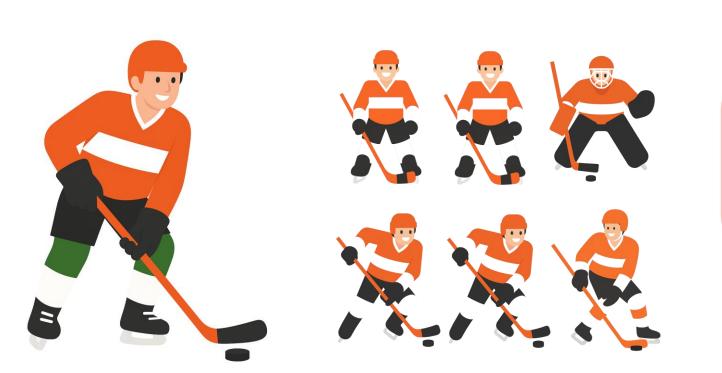
Getting the Best View of Biology and Disease Overview of legacy and current "omic" methods



https://twitter.com/bioxia7/status/1261464021322137600?s=12



Spatial Transcriptomics in Canadian's Term





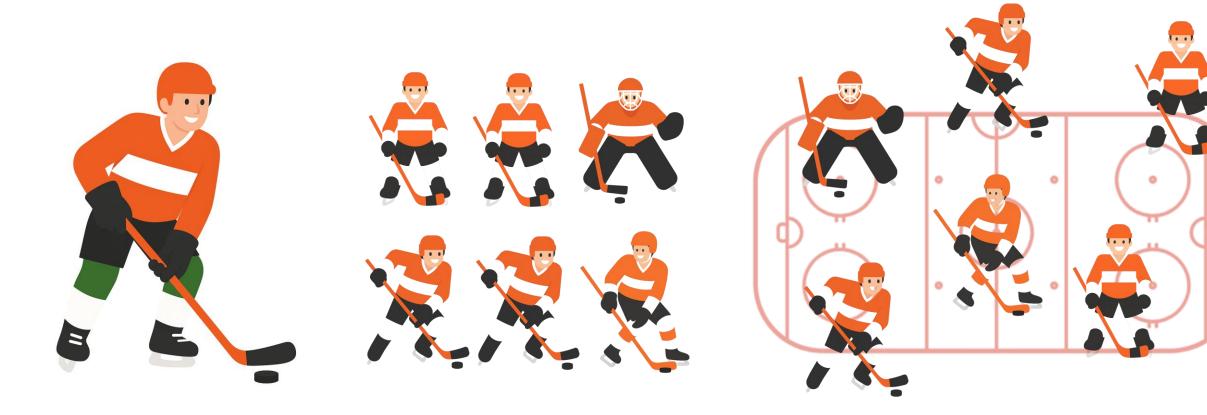
Bulk-Player Information

Single-Player

Spatial-Players



Spatial Transcriptomics in Canadian's Term

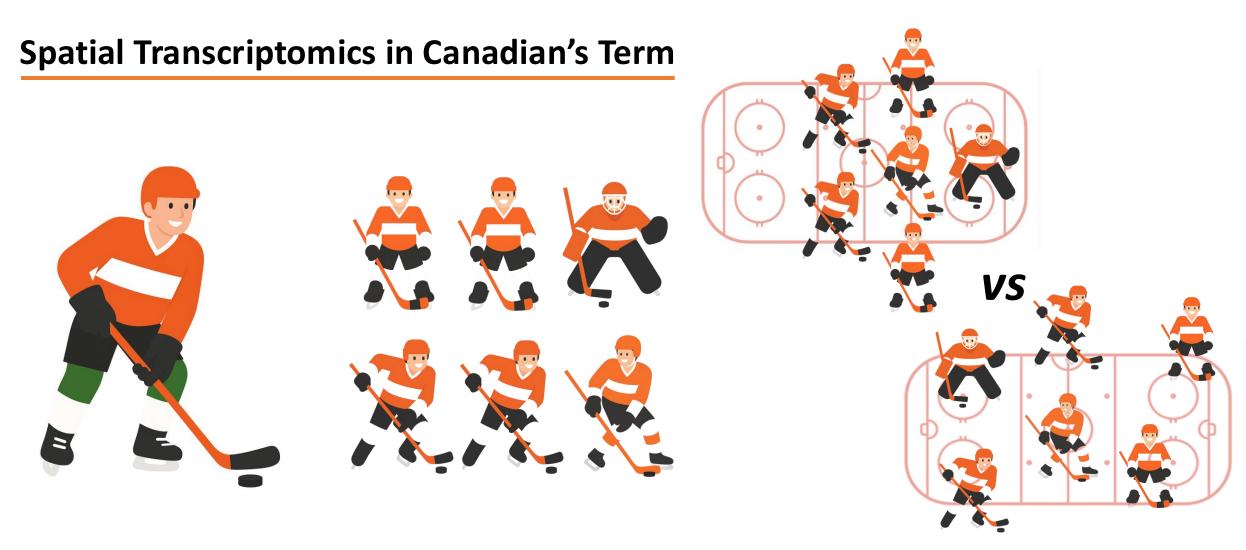


Bulk-Player Information

Single-Player

Spatial-Players



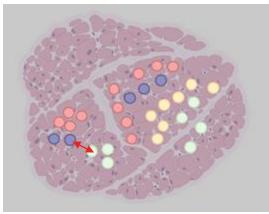


Bulk-Player Information Bulk Genomics Single-Player Single Cell Genomics Spatial-Players Spatial Genomics

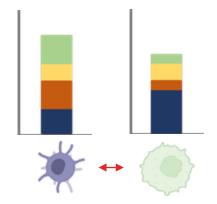


Advances in Spatial Transcriptomics

Spatial Transcriptomics



Map the locations of cell types across a tissue

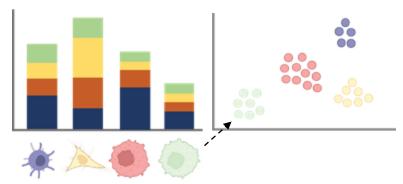


Cell-cell interaction

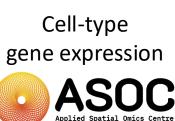
Property	Bulk	Single-cell	Spatial
Cell-type	Х	0	0
Location	Х	Х	0

Bulk Genomics

Single-cell Genomics



Average gene expression



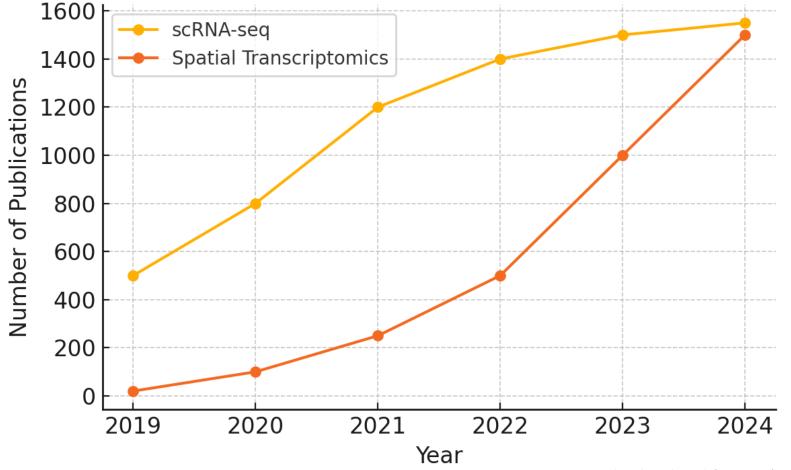
Cell-type clusters

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Why it's important

Publications Trends (2019-2024)



PubMed search trends (2019–2024)



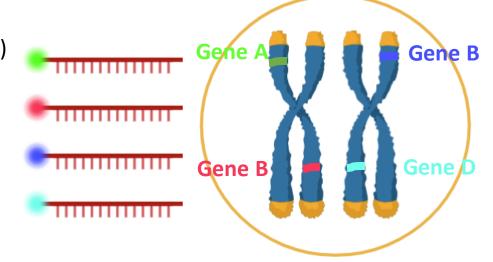
Why it's important

History of Spatial Transcriptomics

• Spatial transcriptomics isn't brand new;

It began in the 1970s but could only handle one gene at a time.

- Dr. Gall lab reported a DNA-DNA hybridization method (*in situ* hybridization) using radioactive labeling for detecting the cellular localization of DNA sequences (*PNAS* 1969).
- The recent rise in spatial transcriptomics comes from the ability to measure hundreds, thousands, or even whole transcriptome at once!



Fluorescence in-situ Hybridization (FISH)



Four Main Key Spatial Transcriptomics Technologies

Sequencing-based method:

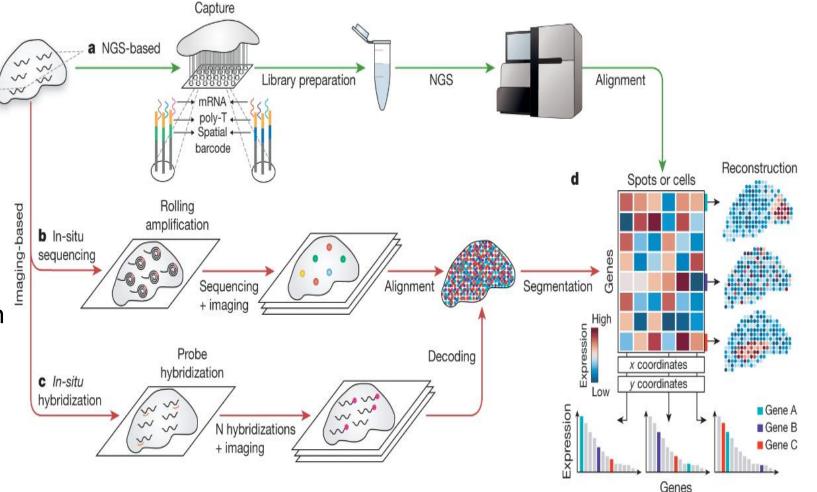
- *in situ* capture ex, Visium
- Region of interest (ROI)-based ex, GeoMx

Imaging-based method:

 Fluorescence in situ hybridization (fISH)

ex, MERFISH, CosMx

in situ sequencing (ISS)
 ex, Xenium



Rao et al., Nature (2021)



Introduction to ST technologies

Four Main Key Spatial Transcriptomics Technologies

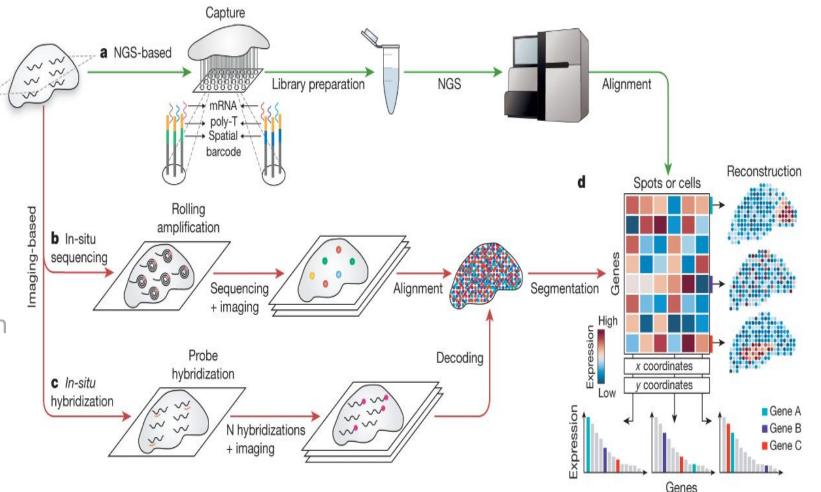
Sequencing-based method:

- Need Sequencing
- Region of interest (ROI)-based ex, GeoMx

Imaging-based method:

• No Need Sequencing Longer Scanning Time

• *in situ* sequencing (ISS) ex, Xenium

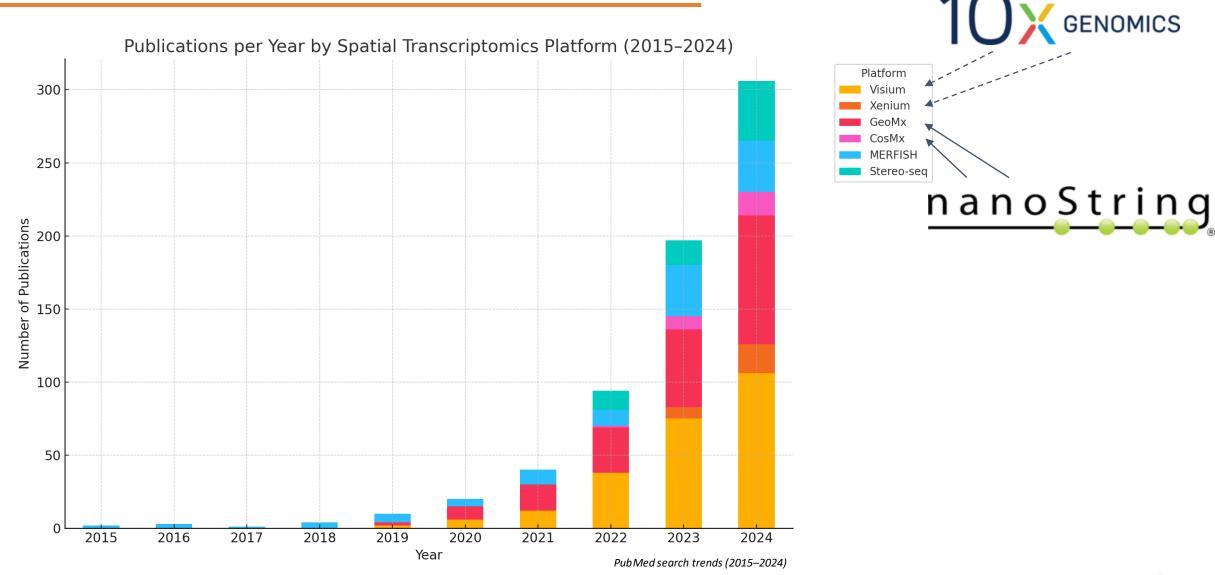


Rao et al., Nature (2021)



Introduction to ST technologies

Publications in Spatial Omics (2015-2024)



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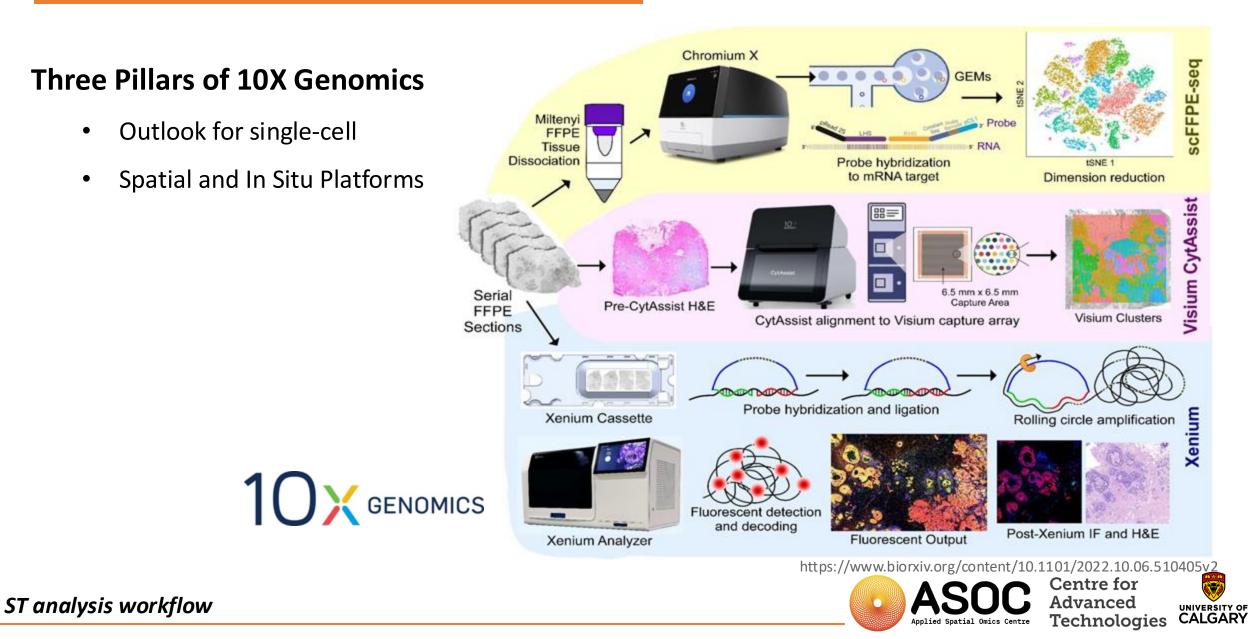
Advanced

Technologies

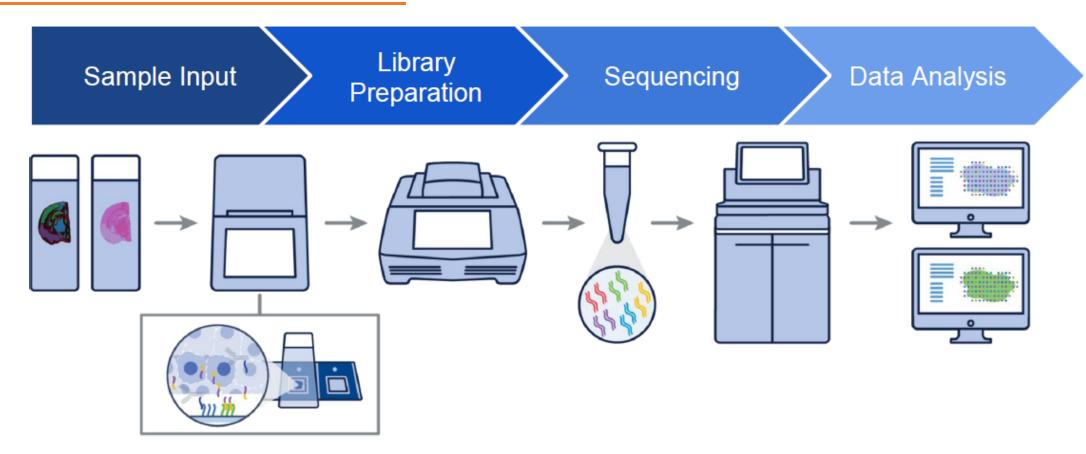
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Overview of key methods

Spatial Omics Platforms: *10X Genomics*



Visium Spatial Gene Expression



Sample Preparation

User-supplied human samples Freshly Placed FFPE Sections Archived H&E Slides

Library Preparation

Visium CytAssist 10x Genomics Kits User-supplied third-party reagents

Sequencing

10x Genomics library-compatible sequencer (e.g., Illumina sequencer)

Analysis

Space Ranger analysis pipeline Loupe Browser visualization tool

Source: 10X Genomics





ST analysis platform – 10X Genomics: Visium

Visium Spatial Gene Expression

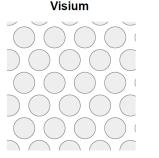
First-Generation assays (55 um barcoded spot)

- v1 3' Gene Expression: 3' ploy(A) capture-based chemistry for fresh frozen tissues from diverse species Resolution of 1–10 cells per 55 µm barcoded spot
- v2 WT Panel Gene Expression: Probe-based chemistry for human and mouse FFPE, fresh frozen, and fixed frozen tissues Resolution of 1–10 cells per 55 μm barcoded spot

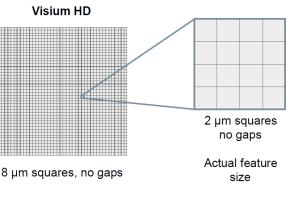
High-Definition assay (2 um barcoded spot) – Single-Cell Resolution

- HD 3' Gene Expression: 3' ploy(A) capture-based chemistry for fresh frozen tissues from diverse species Resolution of single cell per 2 μm barcoded spot
- HD WT Panel Gene Expression: Probe-based chemistry for human and mouse FFPE, fresh frozen, and fixed frozen tissues Resolution of single cell per 2 μm barcoded spot





55 μm spots, hexagonally arranged



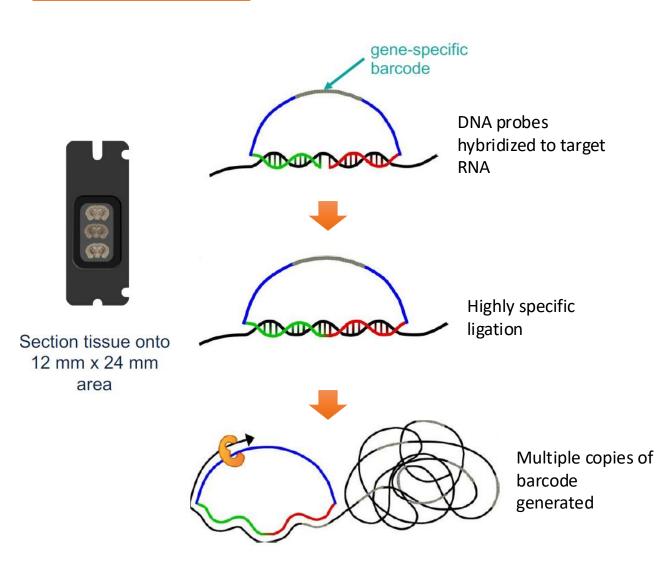
Source: 10X Genomics





ST analysis platform – 10X Genomics: Visium

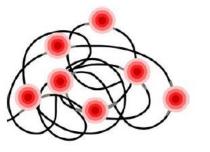
Xenium *in situ*



ST analysis platform – 10X Genomics: Xenium

Chemistry and Imaging





Fluorescent oligos hybridize to amplified probes

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Multiple cycles of probe binding and stripping

Onboard analysis decodes fluorescent signal across cycles to transcripts

10X Genomics



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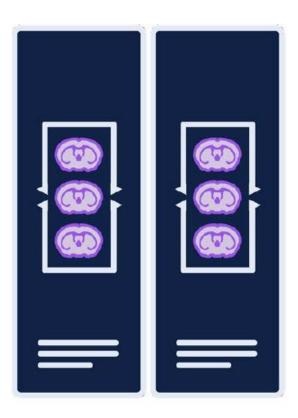


Xenium in situ



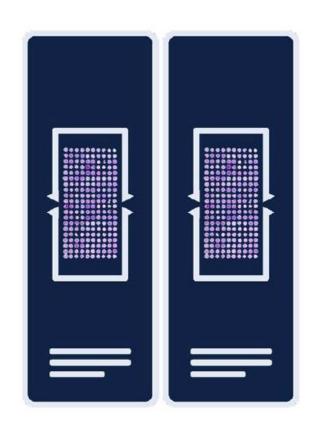
Two 10.5 x 22.5mm sections

~2M cells



Six 6 x 10mm mouse brain sections

~1M cells



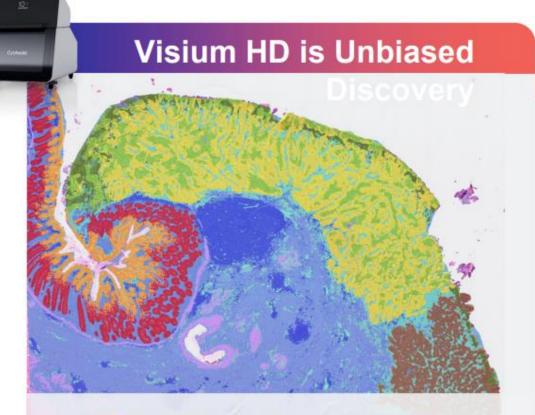
~Four hundred 1 x 0.6mm biopsies in TMAs

~1M cells

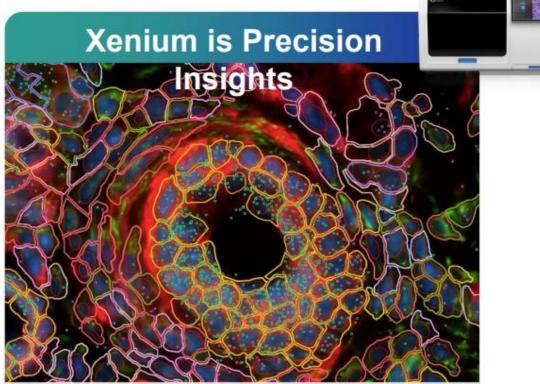




ST analysis platform – 10X Genomics: Xenium



- Whole transcriptome
- Sequencing based
- Transcripts assigned to multi-micron areas



5000s • 100s–1000s of transcripts • High-resolution imaging based • Transcripts assigned to cells

10X Genomics

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ST analysis platform – 10X Genomics: Visium HD vs Xenium

Spatial Omics Platforms: *Nanostring*



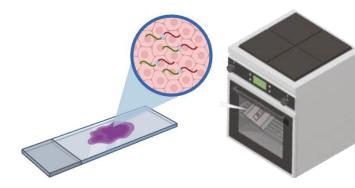


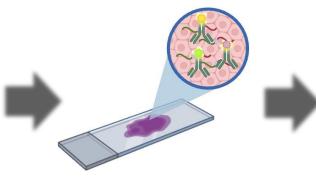
ST analysis platform: Nanostring



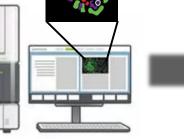


GeoMx Digital Spatial Profiler (DSP)





GeoMx 🔛	tal Bal Profiler
satabilita	

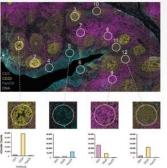


Hybridization with barcoded RNA probes

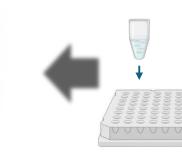
Stain with morphology markers

Repeat for each ROI

Scan and Define Region of Interest (ROI)







Dispense the UV-cleaved oligos into a 96-well plate

Probes release from each area-of illumination (AOI) with UV light

0,00

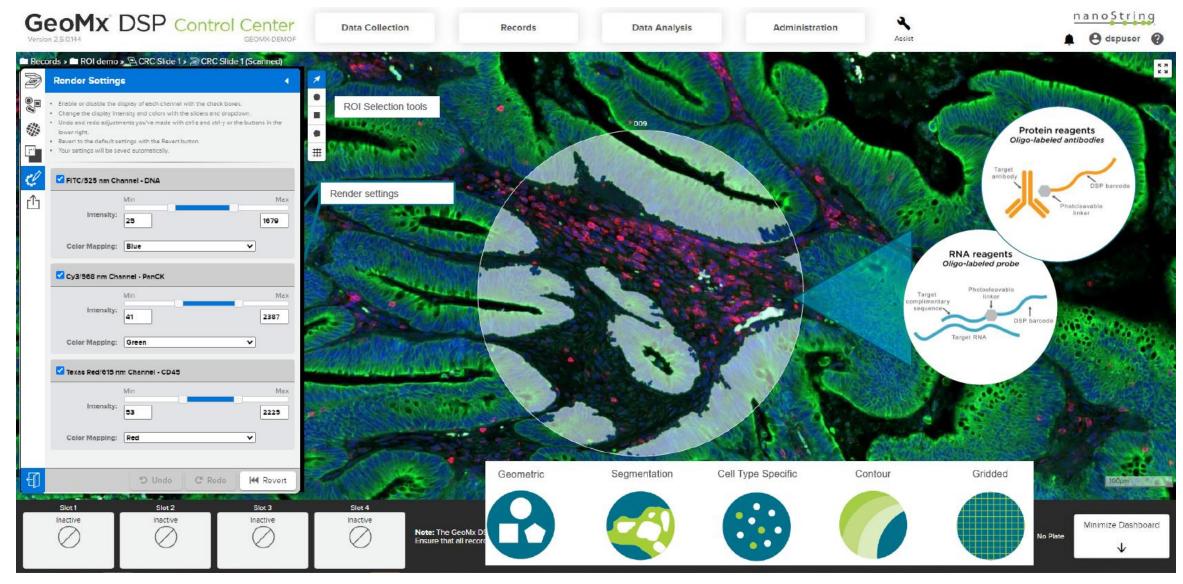




Each AOI is uniquely indexed during library preparation for NGS readout

ST analysis platform – Nanostring: GeoMx

GeoMx Region of Interest (ROI) Selection and Segmentation



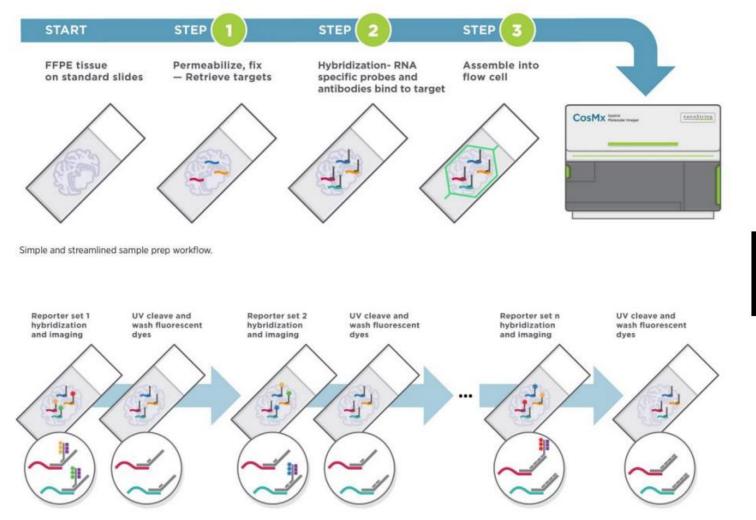
ST analysis platform – Nanostring: GeoMx



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CosMx Spatial Molecular Imager



ST analysis platform – Nanostring: CosMx

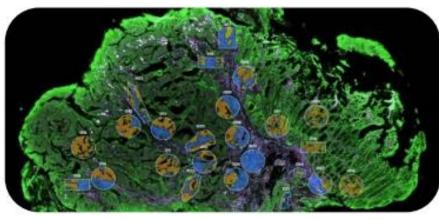


Nanostring

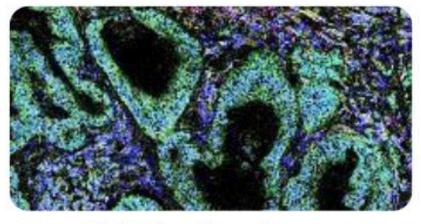




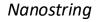




- High-plex, whole transcriptome RNA, 570+ protein (multiomics)
- Fast and flexible platform
- High sensitivity with userdefined tissue sampling
- ex situ sequencing readout



- High Plex detection 6K→18K RNA, 72 proteins
- Accurate cell segmentation
- Comprehensive data analysis with AtoMx[™] SIP
- In situ imager





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ST analysis platform – Nanostring: GeoMx vs CosMx

How Many

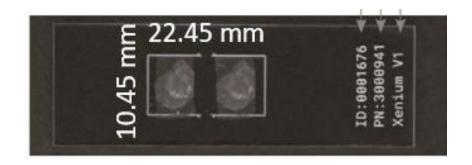
• Visium: 6.5 mm or 11 mm

2/3 whole mouse brains or mouse embryo per capture area Tissue microarray (TMA) is available for Visium CytAssist



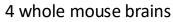
• Xenium: 10.45 mm X 22.45 mm

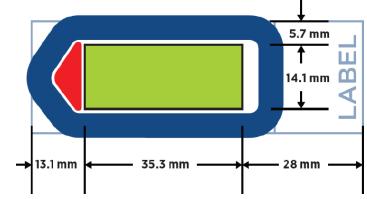
3 whole mouse brains



ST analysis workflow

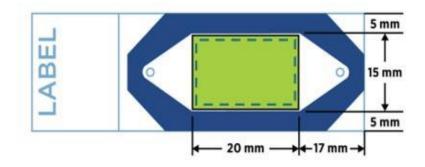
• GeoMx: 35.3 mm X 14.1 mm





• CosMx: 20 mm X 15 mm

3 whole mouse brains



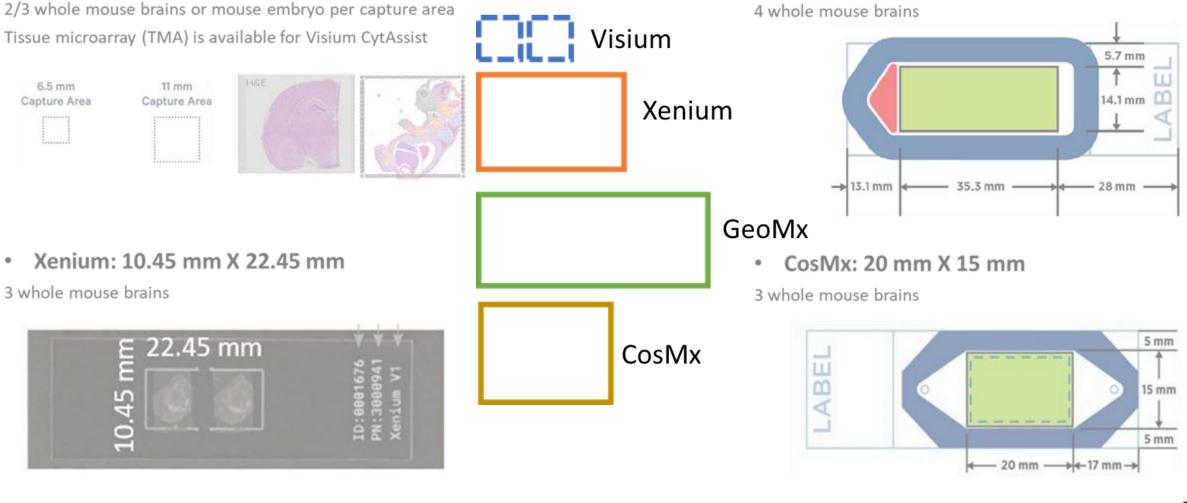




Adapted from 10X Genomics

How Many

Visium: 6.5 mm or 11 mm ٠



ST analysis workflow

Adapted from 10X Genomics

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А Applied Spatial Omics Centre

GeoMx: 35.3 mm X 14.1 mm



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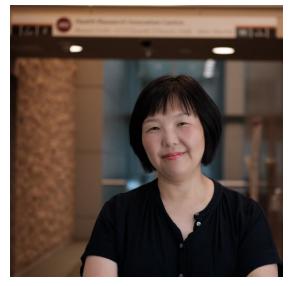
Let me introduce Applied Spatial Omics Centre (ASOC)



Who We Are

ASOC, Applied Spatial Omics Centre

ASOC provides comprehensive services in histology, spatial transcriptomics, and bioinformatic analysis, empowering researchers with the tools and expertise needed to drive innovations in biomedical research



Bo Young Ahn, PhD Senior Spatial Biology Specialist



Shiying Liu, MSc Cell & Molecular Biology Specialist With CGE



Kim Goring, MSc Histology Specialist



Heewon Seo, PhD Lead Bioinformatician

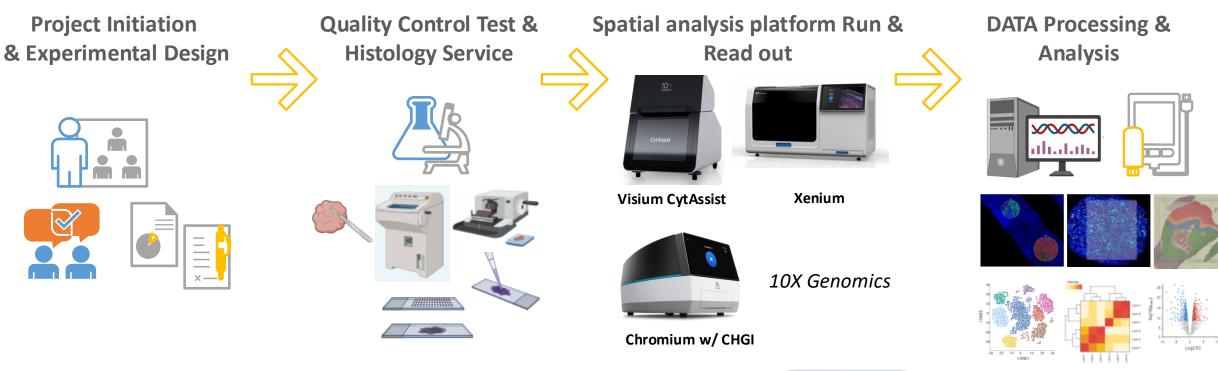
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Our Team member

Who We Are

End-to-end Spatial Omics Service



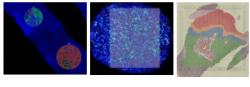
- Project Review and Platform & Service Overview
- Comprehensive RNA & DNA Purification for Genomic Research
- Core Histology Services for Spatial Transcriptomics
- Hands-On Histology Training for Trainees Launching Soon!

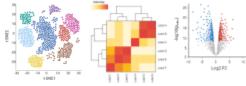


CosMx

GeoMx

Nanostring





- Custom Bioinformatics **Analysis Services** - One-on-One Workshops in Spatial Data Analysis



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How We Help

Obstacles of Spatial Omics Analysis for individual lab

• High cost, expertise, and/or need for specialized equipment

• A wide range of technologies to choose from in situ hybridization, in situ sequencing, and in situ capturing technologies *etc*

• Tissue of interest may require additional optimization











Why Work with Us

Affordable



- Services at cost, non-profit
- Centralization ensures top-tier equipment access



Expertise

- Specialized team dedicated to research support
- Seasoned staff with continuous, open

communication channels

- End-to-end service

Why Work with Us

Adaptable



- Customizable options and rates
- Diverse services for all tech needs
- Open to faculty and external clients

Free consultation!!

https://asoc.ucalgary.ca/

ASOC@ucalgry.ca



What is the goal of the spatial experiment?

- 1. What species are the samples from?
- 2. How are the tissue samples prepared?
- 3. What is the stage of the project?
- 4. For the experiment, how many control and test samples will be included?
- 5. How many sections of each sample type can fit onto

a slide with the chosen platform

Hypothesis Generating

Profile many genes Determine larger area Visium GeoMx

Hypothesis Validating

Profile fewer genes Higher resolution Xenium CosMx





How It Works

Contact us!

We'll collaborate with you to tailor the project and LOWER the costs



How It Works

• High-dimensional complexity

- Spatial omics datasets are massive, thousands of genes across thousands of spatial locations
- Requires sophisticated computational tools to preprocess, normalize, and reduce noise

Integration of modalities

- Spatial omics often combines transcriptomic, proteomic, and imaging data
- Enables the integration and comparison of these layers for a systems-level view

Interpretation at scale

- Manually analyzing these data is infeasible
- Scales insights through automated workflows, machine learning, and statistical modeling



Bioinformatics Applications in Spatial Omics

1. Data preprocessing and quality control

- Spot/barcode filtering, image registration, spatial resolution adjustment
- 2. Spatial clustering and cell type/state annotation
 - Identifying regions with distinct molecular signatures.
- 3. Spatially variable gene detection
 - Identifying genes with expression patterns that correlate with tissue structure
- 4. Cell-cell communication and neighborhood analysis
 - Inferring interactions between nearby cells based on ligand-receptor expression



Hypothesis generation

- Reveal unexpected spatial patterns that can lead to novel biological questions
- Data-driven validation
 - Support or refine experimental findings with quantitative spatial evidence
- Interdisciplinary insights
 - Enable collaborations between biologists and clinicians by transforming complex spatial data into interpretable results
- Scalable and reproducible research
 - Bioinformatics pipelines enable reproducible, transparent, and sharable analyses across teams and projects



What to Ask When Visiting a Core Facility

- Focus on *experience*, not just *information*
 - Don't settle for answers that can be easily found in protocols or online resources
- Ask about practical expertise
 - Inquire about how they've handled similar data, experiments, or projects
- Evaluate problem-solving skills
 - Look for evidence of how the team responds to unexpected challenges and improves outcomes over time

• Assess the team's real-world experience!



• Committed team focused on enhancing data quality

1. Have you assessed your data quality, and can you demonstrate improvement over time?

• Hands-on experts who actively analyze data

2. Do you actively engage with your data to generate insights?

• Research-driven collaborators who align with your goals and contribute to publications

3. Do you have the expertise to understand the current landscape and align with the goals of the field?

• Experience matters when adapting protocols to specific needs!



1. Quality Control and Data Integrity

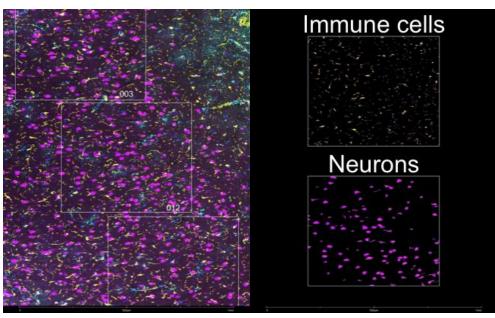
Have you assessed your data quality, and can you demonstrate improvement over time?

- Spatial omics platforms produce a variety of metrics due to the complexity of the processes involved
 - Collaboration between wet- and dry-lab is crucial
- Establishing robust quality control workflows
 - **1.** Collect quality metrics at both the pre-analysis (*run-level*) and post-analysis (*interpretation-level*) stages
 - 2. Identify trends that impact analytical outcomes, e.g., the number of genes retained in downstream analysis
 - **3.** Adjust operational strategies based on QC insights to optimize the performance of spatial omics platforms
 - 4. Review and compare QC metrics across runs to ensure consistency and drive continuous improvement



- In case of *GeoMx*:
 - Users often aim to include as large an area as possible to
 - capture more data for analysis
 - Unaware of the trade-offs associated with larger ROI
 - Larger ROIs tend to accumulate higher background signal
 - Non-specific negative control probes that do not target mRNA
 - True biological signals must compete with elevated background noise, impacting downstream interpretation
- Revealed a clear trend:
 - Larger ROIs are associated with a higher number of genes

being excluded during downstream processing

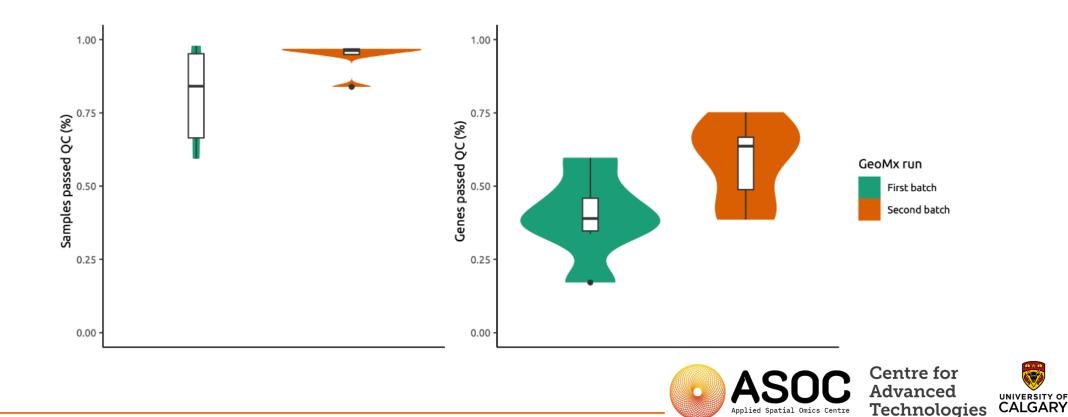


One of the brain ROIs from Nanostring.com



Improving ROI Selection and Data Quality in GeoMx

- Gained expertise in strategic ROI selection to help users obtain the highest-quality data
 - Provided guidance to customers on maximizing signal quality while minimizing background interference
 - Violin plots: first batch (*before* Summer 2024) and second batch (*after* Summer 2024)



2. Encountered Technical Biases

Do you actively engage with your data to generate insights?

- Spatial transcriptomics technologies still face limitations in
 - Sensitivity and specificity
 - Technical variability
- Recurrent observation: consistent high or low gene coverage
 regardless of tissue type
 - Evaluated pre-designed panels using public datasets
 - Seven Xenium In Situ datasets with the Xenium Prime 5k panel
 - 16 GeoMx DSP datasets with the Human Whole Transcriptome Atlas (WTA)

Materials

A. [ST] 10X Genomics Xenium Prime 5K (Human)

1. Breast Cancer (FFPE) 2. Cervical Cancer (FFPE)

- 3. Lymph Node Reactive (FFPE)
- 4. Ovarian Cancer (FFPE)
- 5. Prostate Adenocarcinoma (FFPE)
- 6. Skin Primary Dermal Melanoma (FFPE)
- 7. Ovarian Adenocarcinoma (FF)

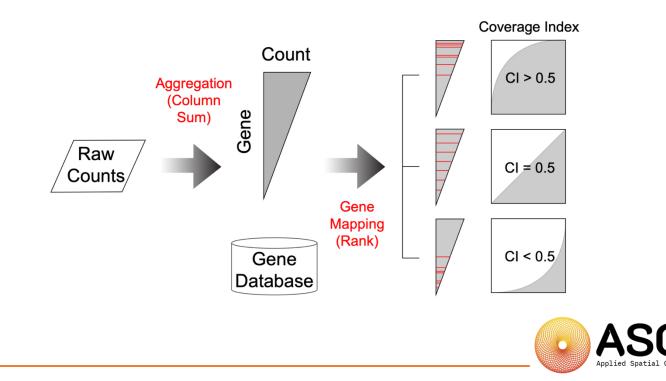
B. [ST] Nanostring GeoMx DSP WTA (Human)

1. SOA-Brain	8. GSE208747
2. SOA-Colon	9. GSE244117
3. SOA-Kidney	10. GSE254145
4. SOA-Liver	11. GSE263897
5. SOA-Lymph Node	12. GSE272995
6. SOA-Pancreas	13. GSE274938
7. TAP-Kidney	14. GSE275677
	15. GSE278670



16. GSE281193

- Coverage Index (CI)
 - Quantitative metric to assess the representation of established gene signatures
 - Higher CI values denote greater coverage among gene members within a representative gene list, while lower CI values indicate reduced coverage

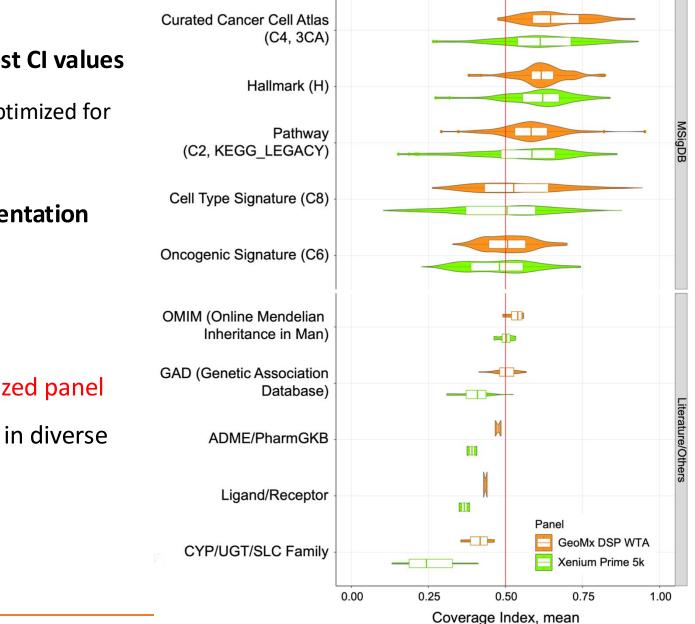


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Evaluating Technical Variability in Spatial Transcriptomics



- Cancer-related gene lists exhibit the highest CI values
 - Suggesting that pre-designed panels are optimized for oncology research
- Ligand/receptor genes show lower representation
 - Indicating potential gaps in panel design

 The findings highlight the need for customized panel designs to ensure adequate gene coverage in diverse biological and clinical applications

3. Translating Data into Biological Understanding

Do you have the expertise to understand the current landscape and align with the goals of the field?

- Linked spatial expression patterns to clinical entities or anatomical contexts
- Supported scientists in **hypothesis generation** by:
 - Generating genome-wide profiles
 - Identifying spatially variable genes
 - Mapping cell-cell communication and microenvironment structures
- Data-driven validation by:
 - Executing a targeted panel assay
 - Collaborating in downstream integrative analyses
 - Combining spatial data with omics/pathology data



- Conducting genome-wide exploratory data analysis to uncover underlying patterns and trends
- Identifying biologically relevant features through statistical and computational approaches
- Mapping extracted features to established biological pathways and disease mechanisms
- Performing clustering analyses to reveal sample subgroups or functional modules
- Integrating prior knowledge using biomedical ontologies and curated databases
- Formulating biologically plausible and testable hypotheses based on data-driven insights
- Ensuring analytical reproducibility and comprehensive documentation to support future validation efforts



- Comparing findings with established benchmarks or gold-standard datasets to confirm biological relevance, e.g., cell-type markers
- Designing robust computational workflows to evaluate generated hypotheses using independent datasets
- Incorporating technical and biological replicates to assess consistency, reproducibility, and biological variability in validation outcomes
- Visualizing validation results through interpretable plots and metrics
- **Documenting validation strategies and outcomes** to ensure transparency and reproducibility of analytical conclusions



Harvesting Insights: Validated Findings and Scholarly Contributions

- Multiscale spatial transcriptomics for *hypothesis generation* and *validation*
 - GeoMx enabled a comprehensive regional overview by identifying broad spatial transcriptomic patterns across tissue sections
 - Xenium provided **high-resolution insights at single-cell and subcellular levels**, allowing for precise mapping of gene expression within individual cells
- The integration of GeoMx and Xenium bridged the gap between tissue-wide transcriptomic profiling and single-cell resolution data

> Sci Adv. 2025 May 2;11(18):eadt8976. doi: 10.1126/sciadv.adt8976. Epub 2025 Apr 30.

Impaired parvalbumin interneurons in the retrosplenial cortex as the cause of sex-dependent vulnerability in Alzheimer's disease

Dylan J Terstege ^{1 2}, Yi Ren ^{1 2}, Bo Young Ahn ³, Heewon Seo ³, Kabirat Adigun ^{1 2}; Alzheimer's Disease Neuroimaging Initiative; Liisa A M Galea ^{4 5 6 7}, Derya Sargin ^{2 8 9}, Jonathan R Epp ^{1 2}

Affiliations + expand

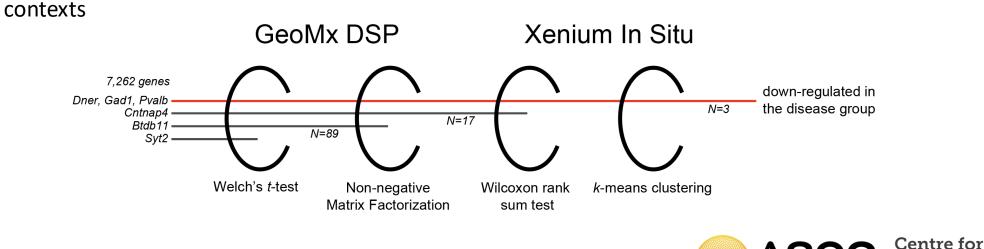
PMID: 40305608 PMCID: PMC12042879 DOI: 10.1126/sciadv.adt8976





Companion Paper Under Review

- Developed **robust computational workflows** to evaluate data-driven hypotheses using multiple • spatial transcriptomics datasets
 - **Integrated two complementary spatial transcriptomics platforms** to comprehensively characterize cells • of interest
 - Leveraged bulk RNA-seq data from the GeoMx platform (*broader coverage*) to classify 10x Xenium-• derived cell types (*higher resolution*), addressing the limitation of canonical cell type markers in disease





Advanced

Advancing Spatial Omics Through Bioinformatics at ASOC

- 1. Quality control and data integrity
- 2. Encountered technical biases
- 3. Translating data into biological understanding (publications)
- ASOC Bioinformatics team aims to accelerate spatial omics research and its applications in clinical practice and provide customized and case-by-case analysis services for the life science research community
- 4. Pipeline development and workflow standardization
- 5. Scalability, reproducibility and data sharing



4. Pipeline Development and Workflow Standardization

- Spatial omics is still emerging standardized workflows are lacking
 - Vendor tools and third-party software cover only a small portion of the full analysis workflow
 - Visium Spatial/HD and Xenium In Situ from 10X Genomics
 - GeoMx DSP and CosMx SMI from Nanostring
- Workflow standardization and best practices
 - Developed and maintained analysis pipelines for all four major spatial omics platforms
 - Created modular and scalable pipelines adaptable to diverse project needs
 - Ensured timely deployment of pipelines to match rising demand
 - Hosted training workshops and created documentation to support researchers



> Nat Methods. 2025 Apr;22(4):813-823. doi: 10.1038/s41592-025-02617-2. Epub 2025 Mar 13.

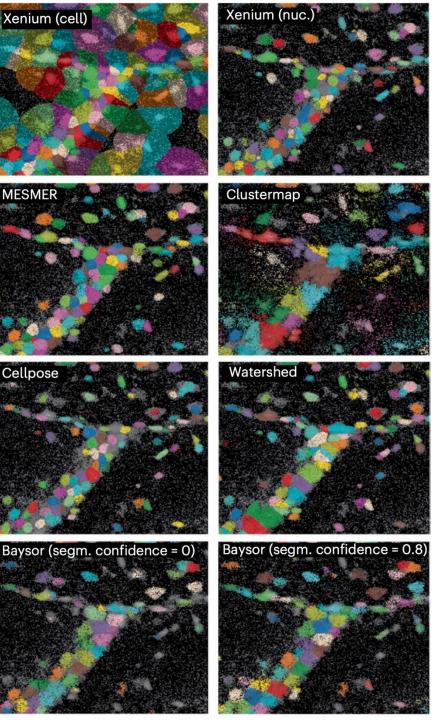
Optimizing Xenium In Situ data utility by quality assessment and best-practice analysis workflows

Sergio Marco Salas ¹ ², Louis B Kuemmerle ^{# 3} ⁴, Christoffer Mattsson-Langseth ^{# 5}, Sebastian Tismeyer ^{# 6}, Christophe Avenel ^{# 7}, Taobo Hu ^{# 5}, Habib Rehman ^{# 3 8}, Marco Grillo ^{# 5}, Paulo Czarnewski ^{# 9 10}, Saga Helgadottir ^{# 5}, Katarina Tiklova ⁵, Axel Andersson ⁷, Nima Rafati ¹¹, Maria Chatzinikolaou ⁵, Fabian J Theis ³, Malte D Luecken ^{3 12}, Carolina Wählby ⁷, Naveed Ishaque ⁶, Mats Nilsson ¹³

Affiliations + expand

PMID: 40082609 PMCID: PMC11978515 DOI: 10.1038/s41592-025-02617-2

- Current consensus on data analysis for Xenium In Situ.
 - Comparison of cell identification using various segmentation algorithms (see right)
 - Utilized <u>a default parameter set</u> to ensure a fair comparison

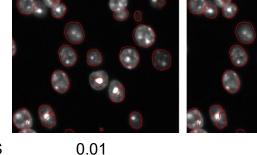


Xenium In Situ Workflows at ASOC

- Employ a selection of consensus methods, i.e., **StarDist**
- Parallelize computational processes with <u>a diverse array of</u>

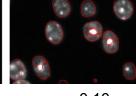
parameters

• e.g., healthy mouse brain (100 μm)



NMS



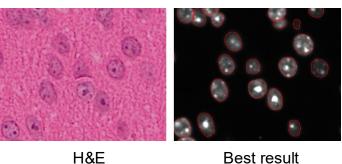


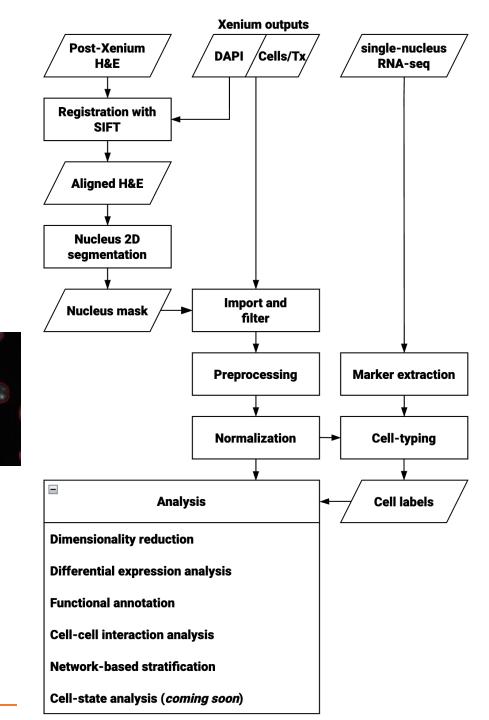


Best result

0.25

Optimize output to address the biological questions •





5. Scalability, Reproducibility and Data Sharing

- Built reproducible analysis environments with version-controlled pipelines
- Ensured compatibility with downstream analysis platforms
- Supported collaborative data sharing through structured results, interactive visualizations, and documentation





- **Designed the workflows to be scalable** across different computing infrastructures, from local machines to high-performance clusters
- Packaged all computational workflows into a Docker application to ensure reproducibility across environments
- Provided full source code and documentation publicly via GitHub for transparency and community use
- Ensured that analyses can be reliably reproduced and audited by peers or reviewers
- Facilitated easy deployment of the analysis environment without manual setup or dependency issues



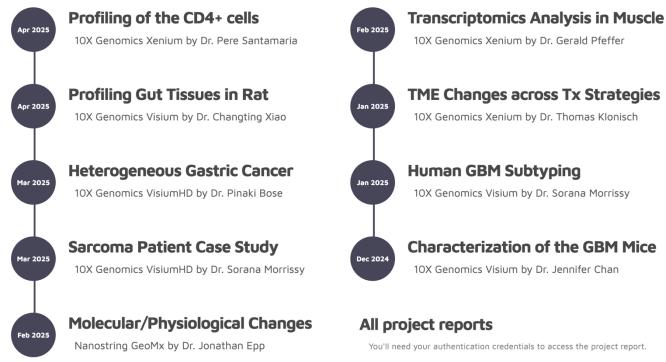


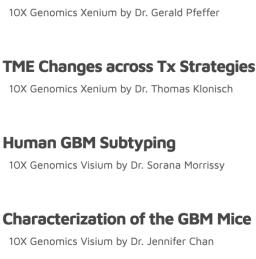


Project Reports and Data Sharing

• ASOC portal (ASOC.ucalgary.ca) provides *one-click* access to project reports—no downloads or client-side backups required

Latest Project Reports





You'll need your authentication credentials to access the project report.







Educational Activities and Beyond

- Hosted a hands-on bioinformatics workshop focused on best practices for GeoMx data analysis
- **Provided step-by-step guidance on running Docker-based workflows** in participants' own working environments, e.g., laptops
- Offered comprehensive training materials and reusable code examples to support independent learning and application
 Hands-on Bioinformatics Workshop
- Presented real-world case studies to demonstrate end-to-end reproducibility and equip participants to analyze their own data effectively



Series 1 Nanostring GeoMx DSP (Digital Spatial Profiler)

Host Applied Spatial Omics Centre, University of Calgary

Date Friday, October 04, 2024

Location Health Science Centre (HSC) 1504

URL https://ASOC.UCalgary.ca/HBW

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