



Explore the full richness of biological insights with single cell applications

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Science and Technology Advisor 10x Genomics,
CEE & Israel & Russia, Distributors

We are 10x Genomics



TheScientist TOP 10 INNOVATIONS

- 2020 Single Cell Multiome, Total Seq C
- 2019 Single Cell ATAC
- 2018 Single Cell Immune Profiling
- 2017 Single Cell Gene Expression
- 2015 Gemcode



2018
Single Cell Genomics



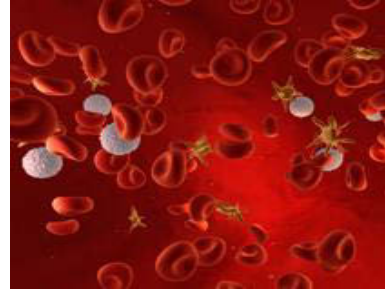
2019
Single Cell
Multimodal Omics

Mastering Biology to Advance Human Health

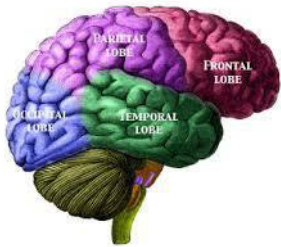
Tissues are Complex Mixtures of Distinct Cellular Populations



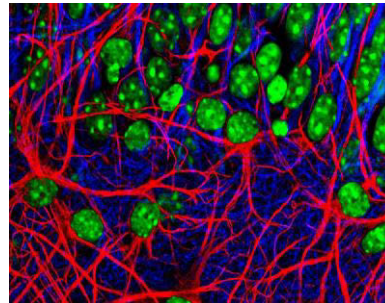
“Blood”



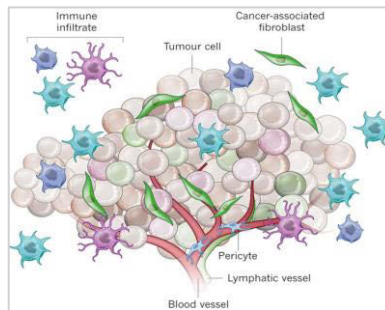
- Platelets and red blood cells
- Immune cells
- Circulating tumor cells



“Brain”

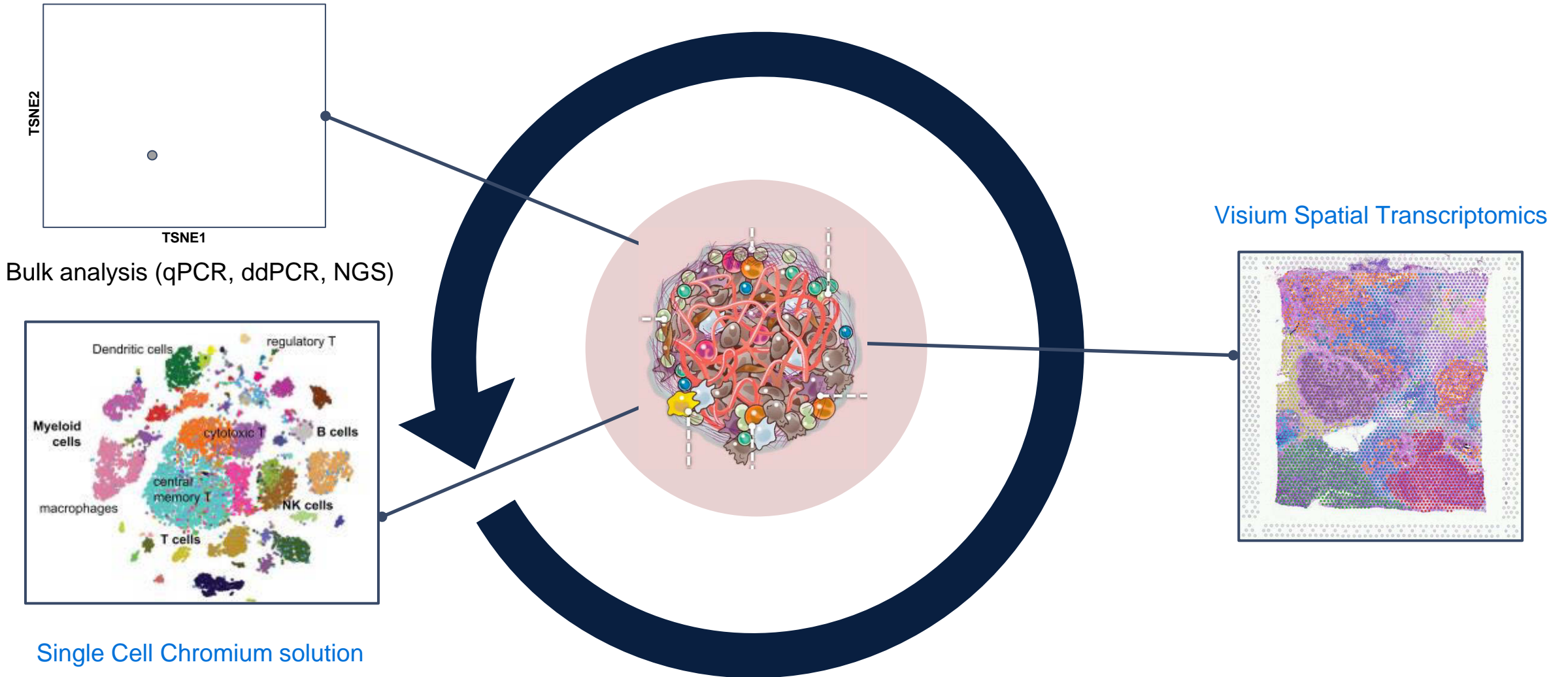


- Excitatory neurons
- Microglia
- Astrocytes



- Tumor infiltrating lymphocytes
- Stromal cells
- Tumor cells

Building your understanding of biological samples – cell by cell



Why Single-cell Analysis?



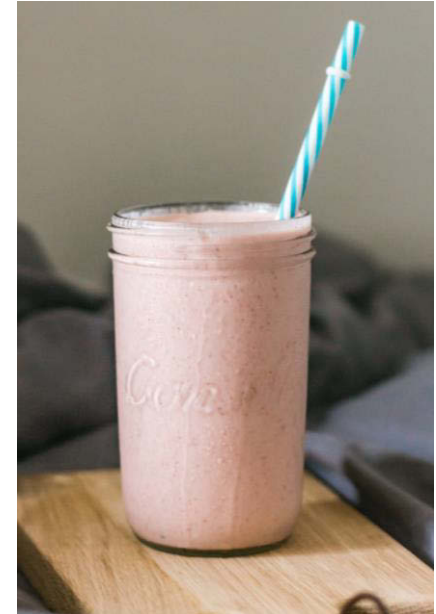
Single component analysis

Photo by [Tetiana Bykovets](#) on [Unsplash](#)



Your sample

Photo by [Blendtopia Smoothies](#) on [Unsplash](#)

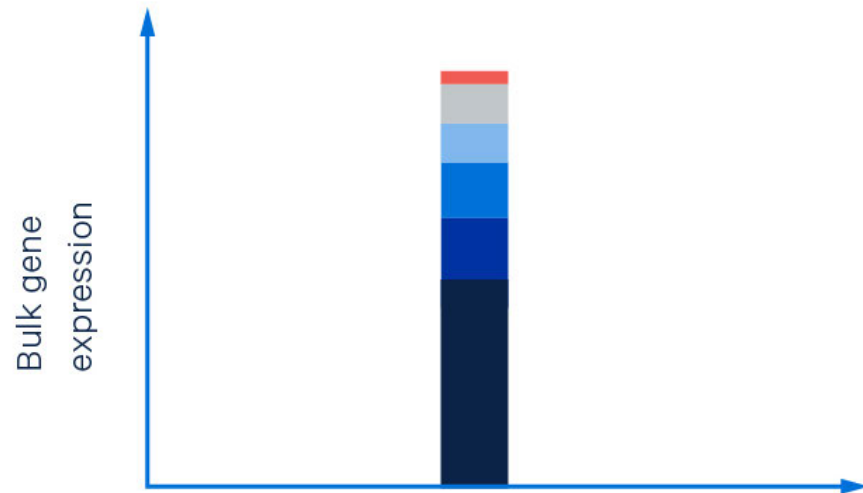


Bulk Analysis

Photo by [Sincerely Media](#) on [Unsplash](#)

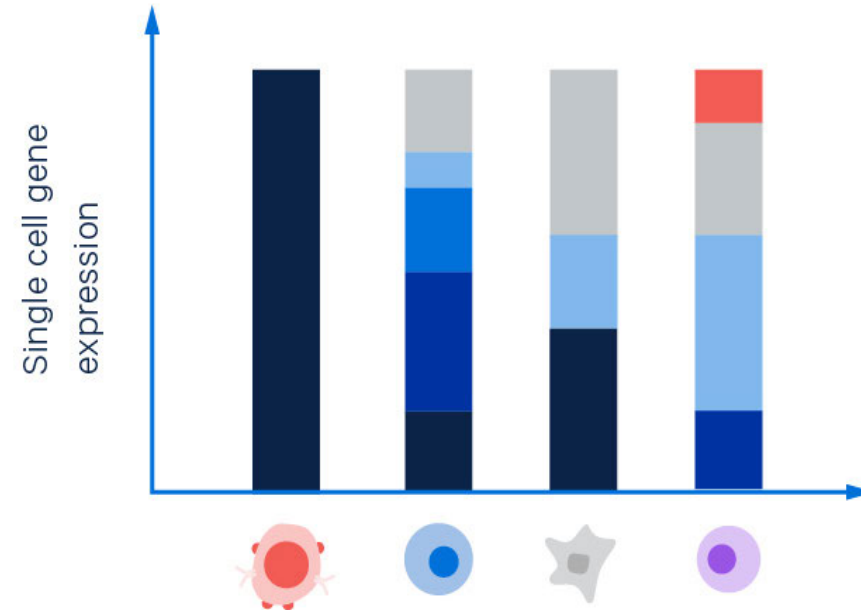
What is single cell sequencing?

Bulk RNA-sequencing



Average gene expression from all cells

Single cell resolution



Each cell type has a distinct expression profile

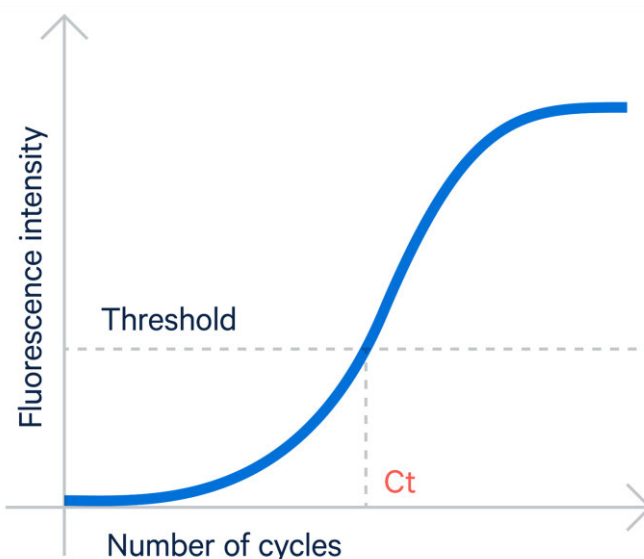
Bulk RNA seq



Single cell sequencing shares steps with other techniques

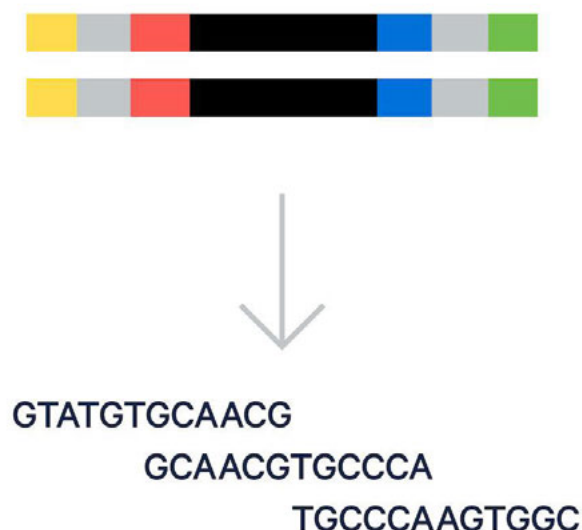
Familiar steps, with deeper insights

qPCR



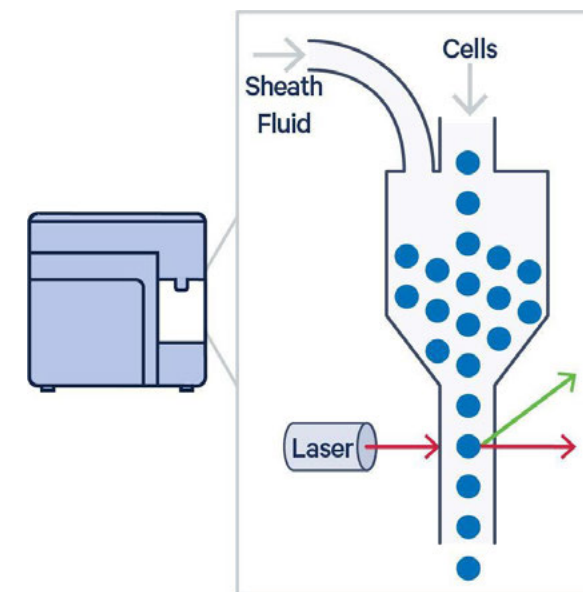
RNA released from lysed cells is converted to cDNA for analysis

Bulk RNA-seq



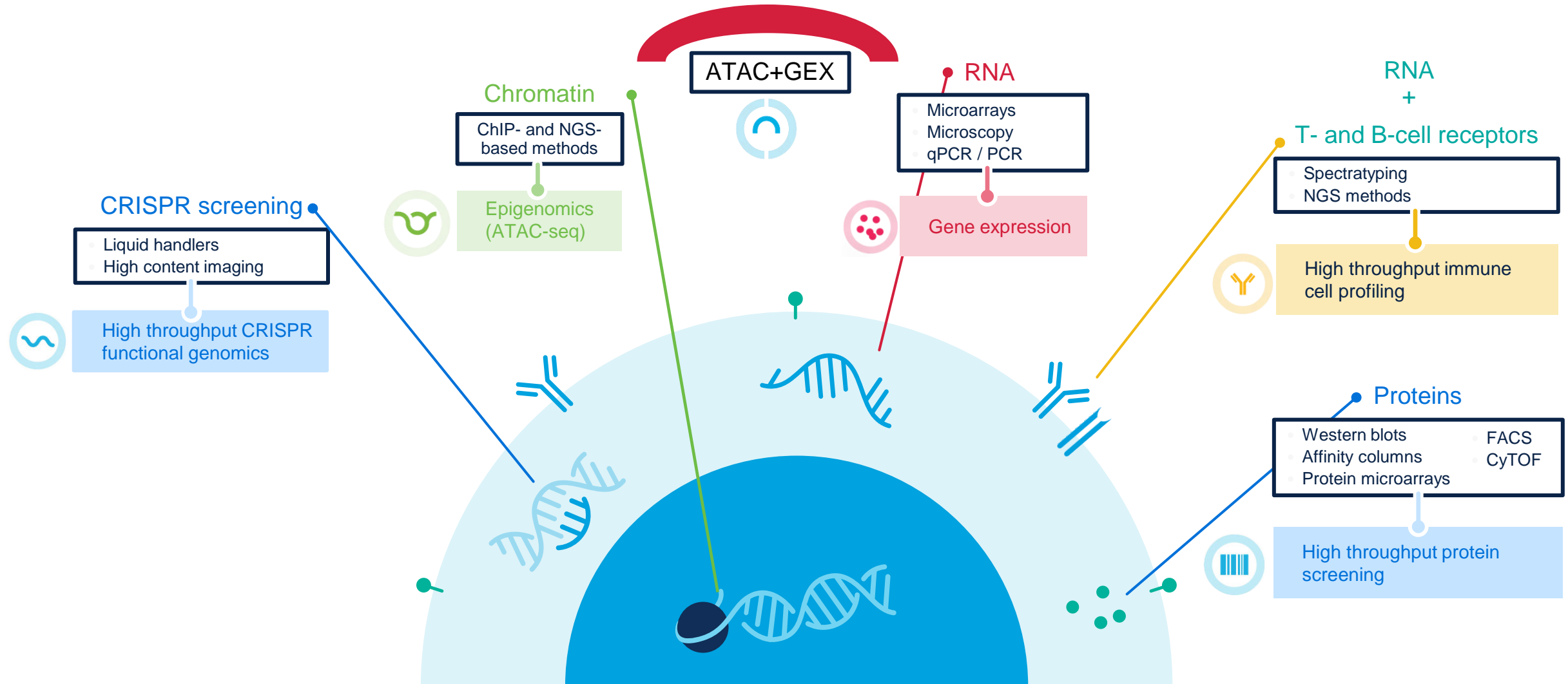
NGS library preparation and sequencing

Flow cytometry



Requires single cell suspensions as input

Next generation molecular profiling solutions



Single Cell RNA Sequencing

How does it work?

Chromium Single Cell Gene Expression Workflow

Input

Library Creation

Sequencing

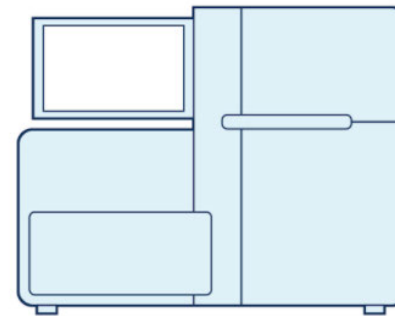
Data Analysis and
Visualization



Suspension of
dissociated single
cell/nuclei



**Cell partitioning and
molecular barcoding**



Sequencing

Analysis

Visualization

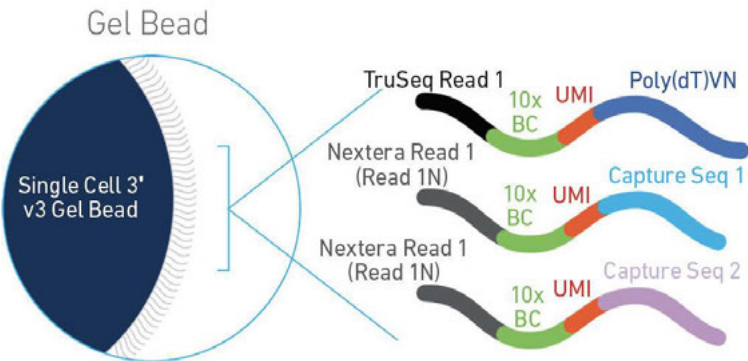
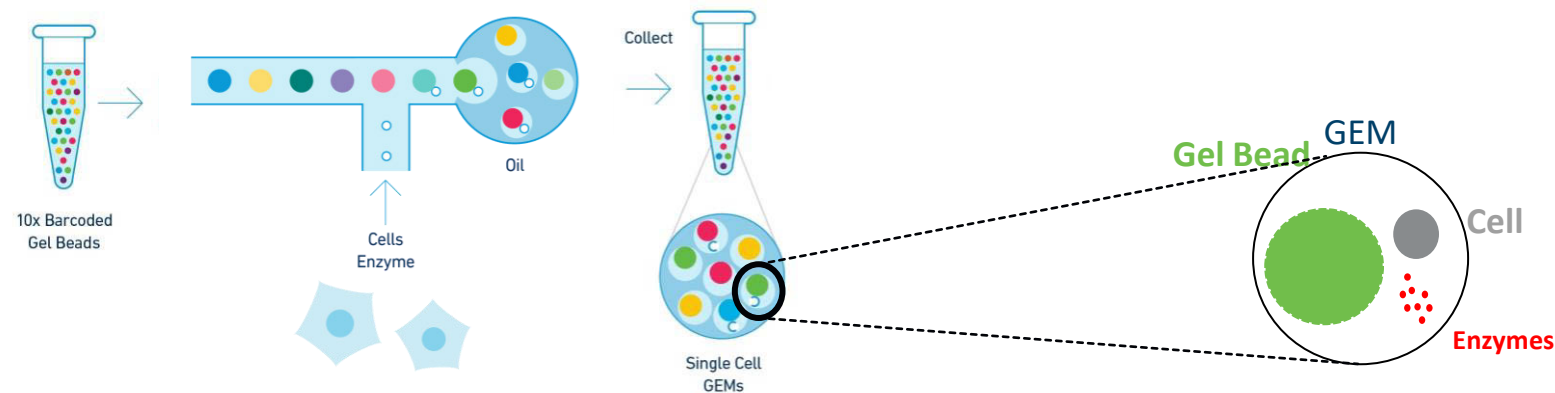


Community Analysis
Tools

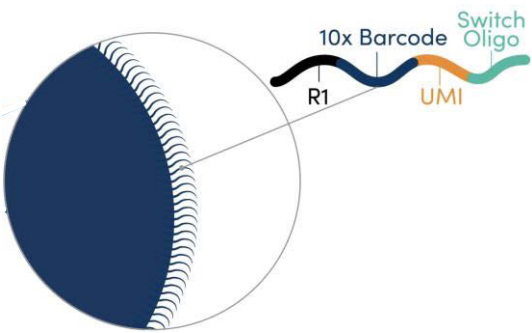
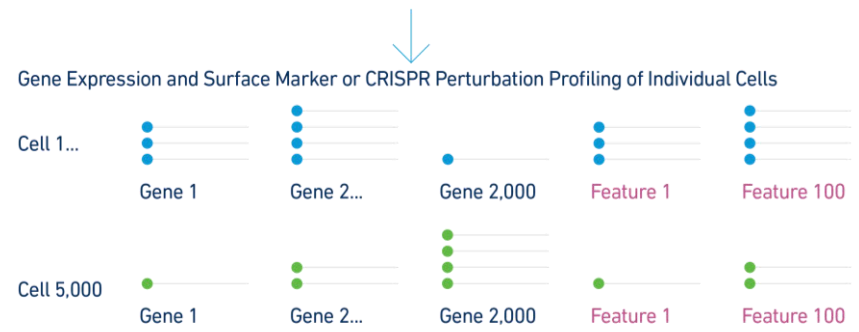
- 8 channels/chip
- 500-10 000 cells recovered per channel
- 40-65% cells recovered

Chromium Single Cell Gene Expression Workflow

Microfluidic system & Next GEM Technology



3' single cell gene expression



5' immune profiling

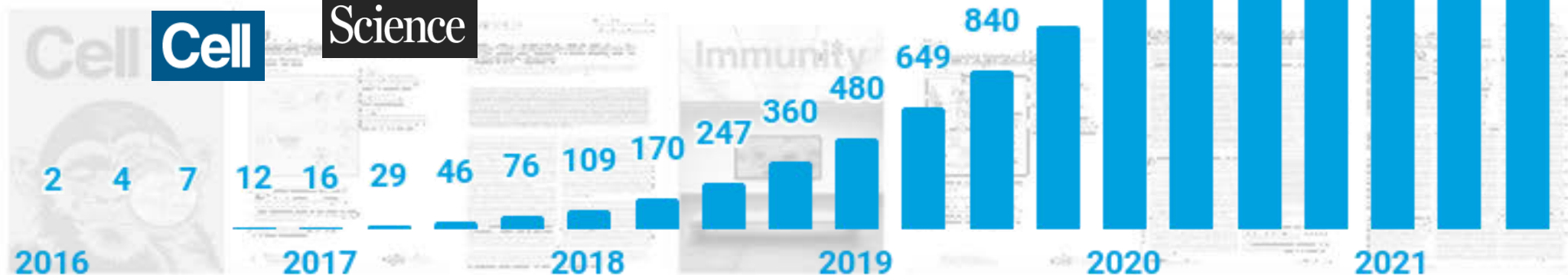
Customer publications citing 10x Genomics continue to rise

3,000+
Publications

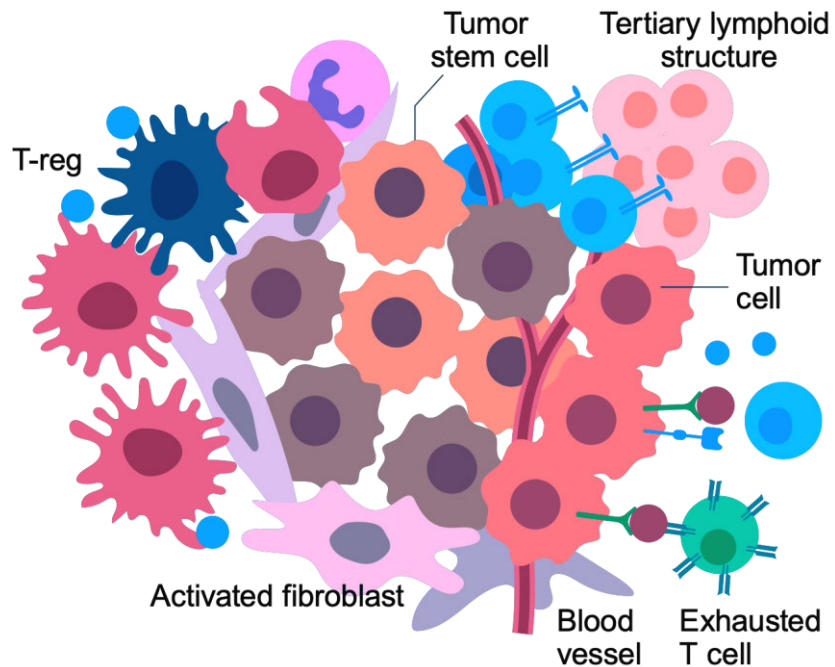
nature

Cell

Science



Understanding and Fighting Cancer



Cell

Article

Single-Cell Analyses Identify Brain Mural Cells Expressing CD19 as Potential Off-Tumor Targets for CAR-T Immunotherapies

nature
medicine

BRIEF COMMUNICATION

<https://doi.org/10.1038/s41591-021-01564-7>

Check for updates

Neurocognitive and hypokinetic movement disorder with features of parkinsonism after BCMA-targeting CAR-T cell therapy

CellPress

Cancer Cell

Letter

Expression of chimeric antigen receptor therapy targets detected by single-cell sequencing of normal cells may contribute to off-tumor toxicity

Cancer Cell

Article

Single-cell dissection of cellular components and interactions shaping the tumor immune phenotypes in ovarian cancer

Cancer Cell

Article

Tumor and immune reprogramming during immunotherapy in advanced renal cell carcinoma

nature
medicine

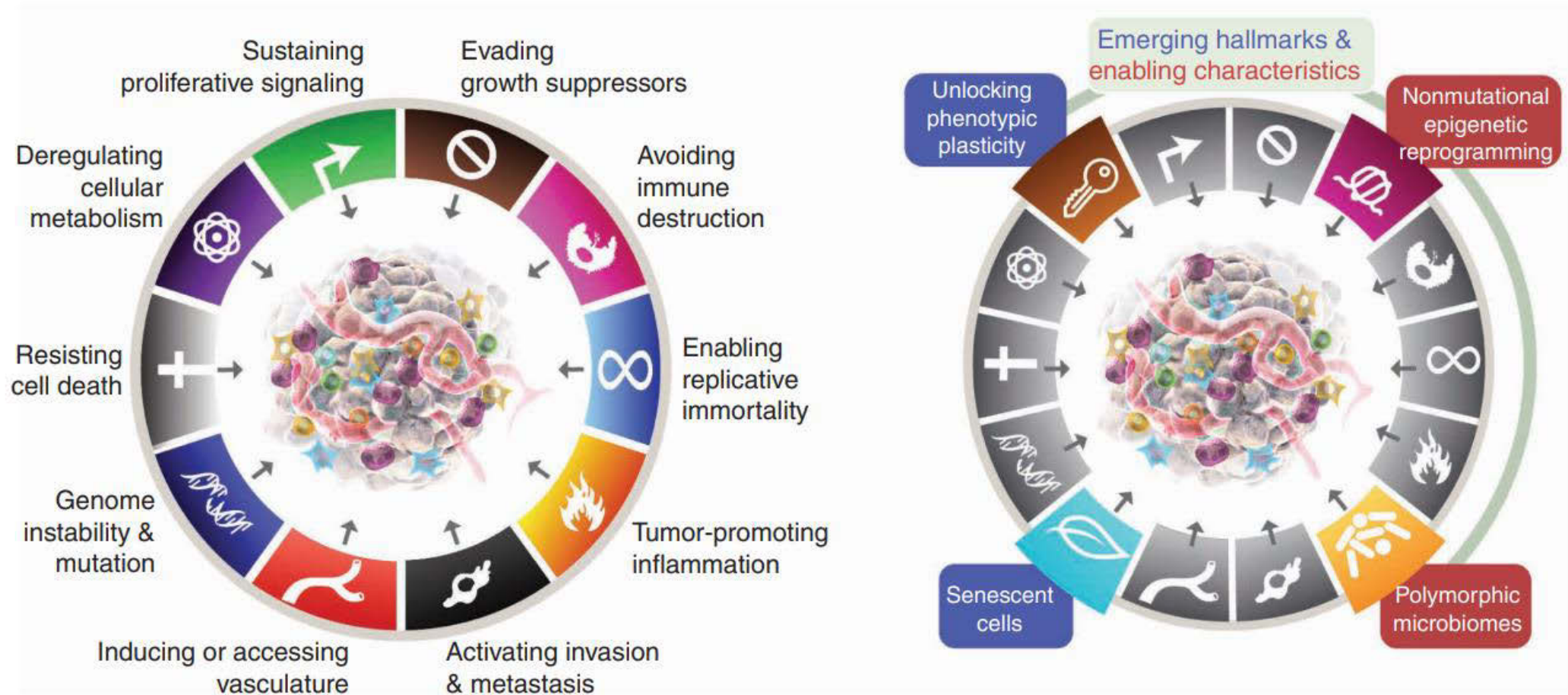
ARTICLES

<https://doi.org/10.1038/s41591-021-01323-8>

Check for updates

A single-cell map of intratumoral changes during anti-PD1 treatment of patients with breast cancer

Hallmarks of Cancer: New Dimensions



Understanding (Auto) immunity and (Auto)immune Therapies

Systemic sclerosis

TRANSLATIONAL SCIENCE

Single-cell transcriptome analysis identifies skin-specific T-cell responses in systemic sclerosis

Alyxandria M Gaydosik,¹ Tracy Tabib,¹ Robyn Domsic¹,¹ Dinesh Khanna²,² Robert Lafyatis,¹ Patrizia Fuschiotti¹

Article

Interpreting type 1 diabetes risk with genetics and single-cell epigenomics

<https://doi.org/10.1038/s41586-021-03552-w> Joshua Chiou^{1,2,3}, Ryan J. Geusz¹, Mei-Lin Okino¹, Jee Yun Han¹, Michael Miller¹, Rebecca Melton¹, Elisha Beebe¹, Paola Benaglio¹, Serina Huang¹, Katha Korgaonkar¹, Sandra Heller¹, Alexander Kleger¹, Sebastian Preiss¹, David U. Gorkin^{1,4}, Maik Sander^{1,4,5} & Kyle J. Gaulton^{1,6,7}

Received: 21 June 2020
Accepted: 14 April 2021
Published online: 19 May 2021

SCIENCE ADVANCES | RESEARCH ARTICLE

DISEASES AND DISORDERS

Gene expression signatures of target tissues in type 1 diabetes, lupus erythematosus, multiple sclerosis, and rheumatoid arthritis

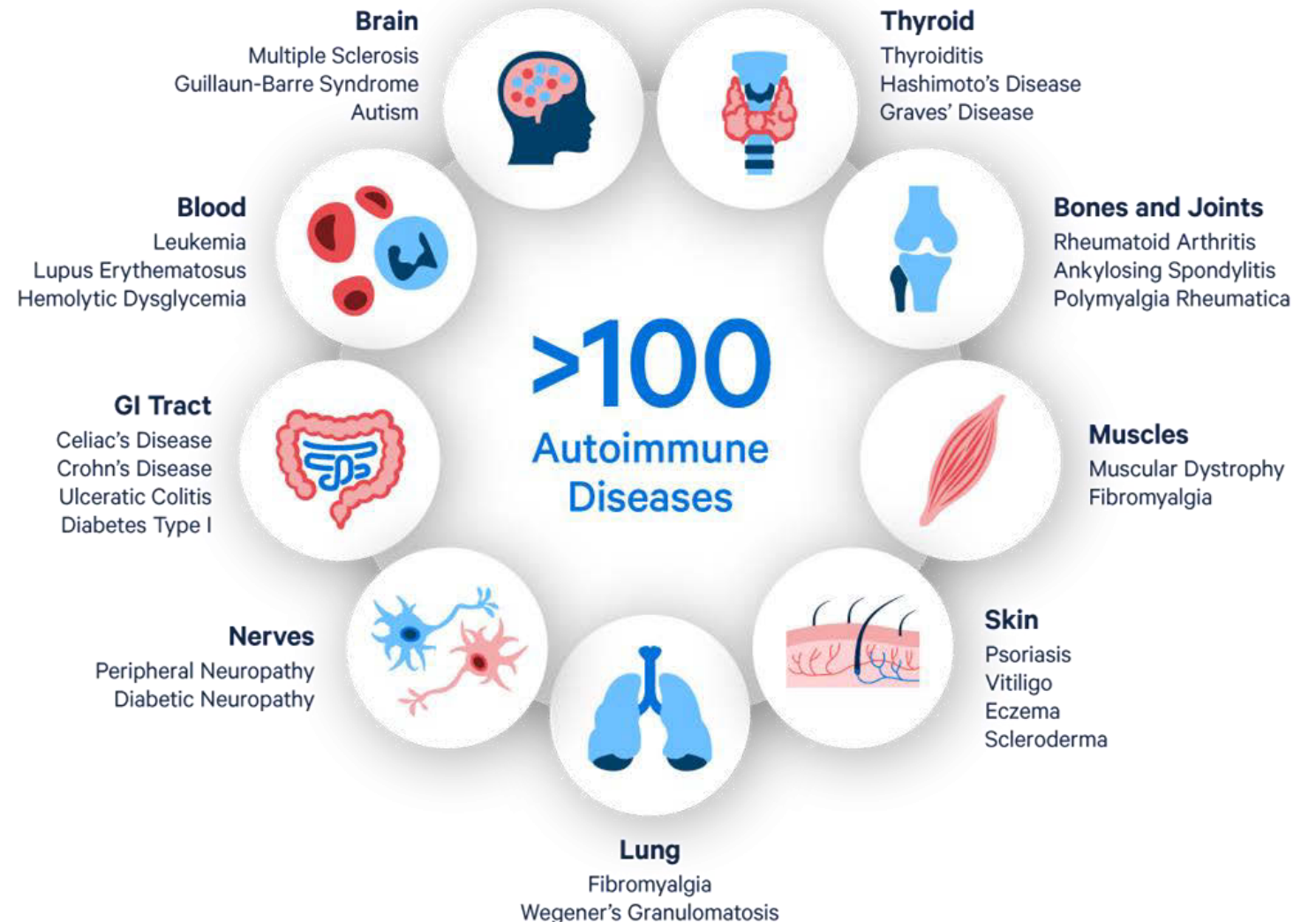
F. Szymczak^{1,2*}, M. L. Colli^{1,3*}, M. J. Mamula³, C. Evans-Molina⁴, D. L. Eizirik^{1,5†}

Article

A myeloid–stromal niche and gp130 rescue in NOD2-driven Crohn's disease

<https://doi.org/10.1038/s41586-021-03484-5> Shikha Nayar^{1,2}, Joshua K. Morrison², Mamta Giri¹, Kyle Gettler^{1,4}, Ling-shiang Chuang¹, Laura A. Walker¹, Hualin M. Ko^{1,2,3}, Ephraim Kenigsberg¹, Subra Kugathasan¹, Miriam Merad¹, Jaime Chu¹ & Judy H. Cho^{1,2}

Received: 8 May 2020
Accepted: 23 March 2021



Elucidating the Biology of Neurodegeneration



“We were blind to this complexity. Things looked simpler than they really are.”

—Richard Hodes
Director, National Institute of Aging

ARTICLES

<https://doi.org/10.1038/s41590-021-0093-5>

nature
immunology

Check for updates

Microglia use TAM receptors to detect and engulf amyloid β plaques

ARTICLES

<https://doi.org/10.1038/s41588-020-00721-x>

nature
genetics

Check for updates

Single-cell epigenomic analyses implicate candidate causal variants at inherited risk loci for Alzheimer's and Parkinson's diseases

Cell

Resource

Spatial Transcriptomics and *In Situ* Sequencing to Study Alzheimer's Disease

nature
genetics

ARTICLES

<https://doi.org/10.1038/s41588-021-00894-z>

Single-nucleus chromatin accessibility and transcriptomic characterization of Alzheimer's disease

Science

RESEARCH ARTICLES

Cite as: D. Gate *et al.*, *Science* 10.1126/science.abf7266 (2021).

CD4⁺ T cells contribute to neurodegeneration in Lewy body dementia

Acta Neuropathologica
<https://doi.org/10.1007/s00401-021-02263-w>

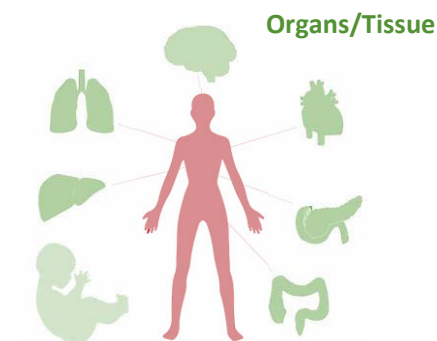
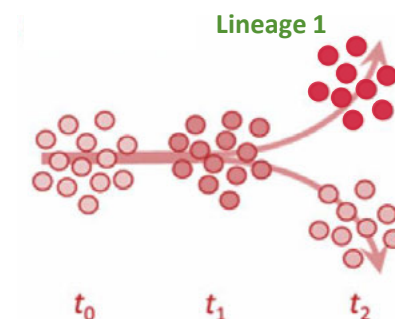
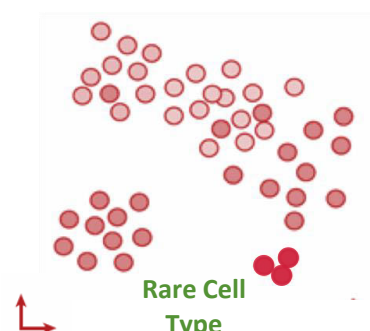
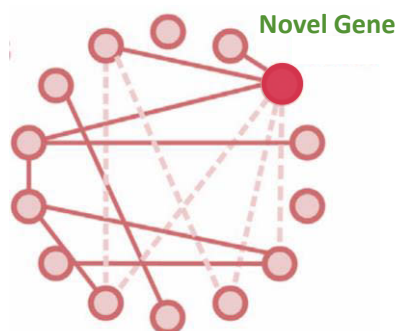
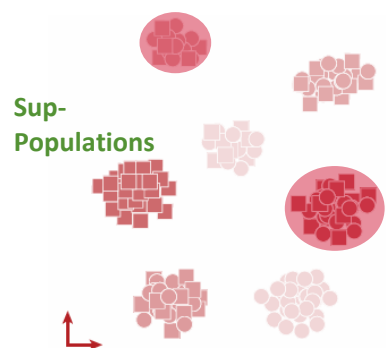
ORIGINAL PAPER



Distinct amyloid- β and tau-associated microglia profiles in Alzheimer's disease

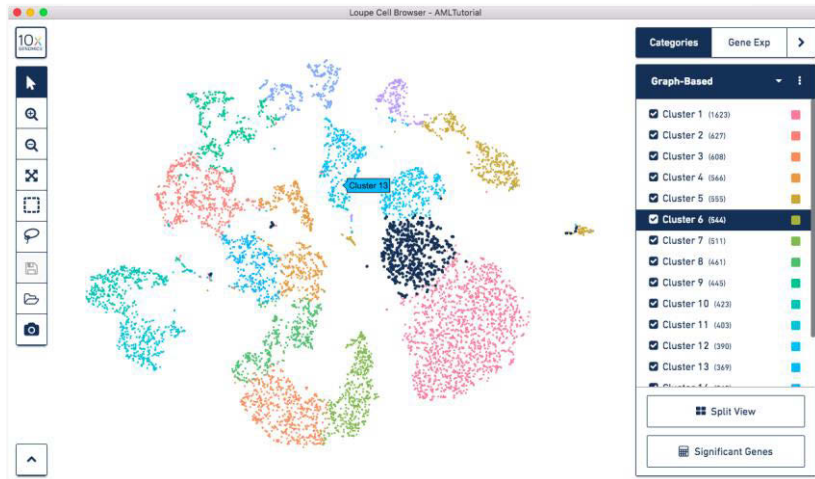
How can single cell data be applied to your research?

- Characterize & identify heterogeneous cell populations
- Discover new cell markers & regulatory pathways
- Uncover novel cell types, cell states & rare cell types
- Reconstruct developmental hierarchies and reveal lineage relationships
- Profiling healthy and diseased tissue and organs

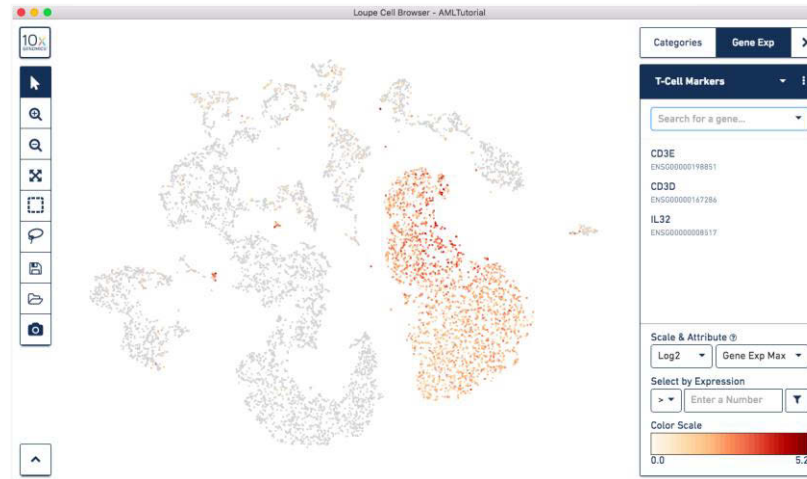


Loupe Cell Browser – Analysis for Everyone

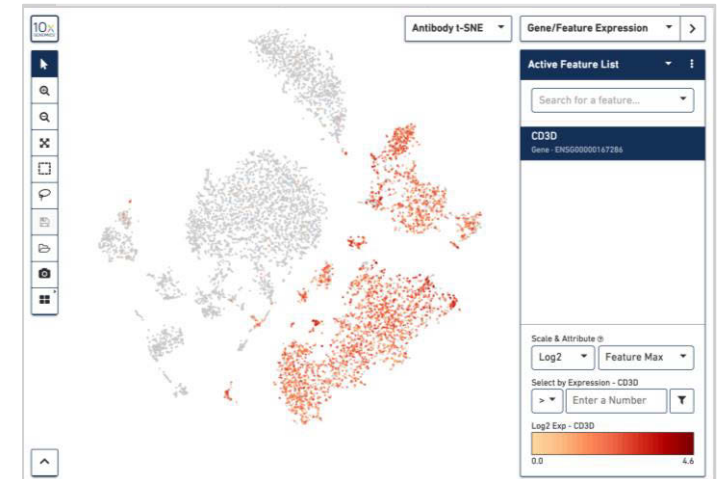
Precomputed GEX Clusters



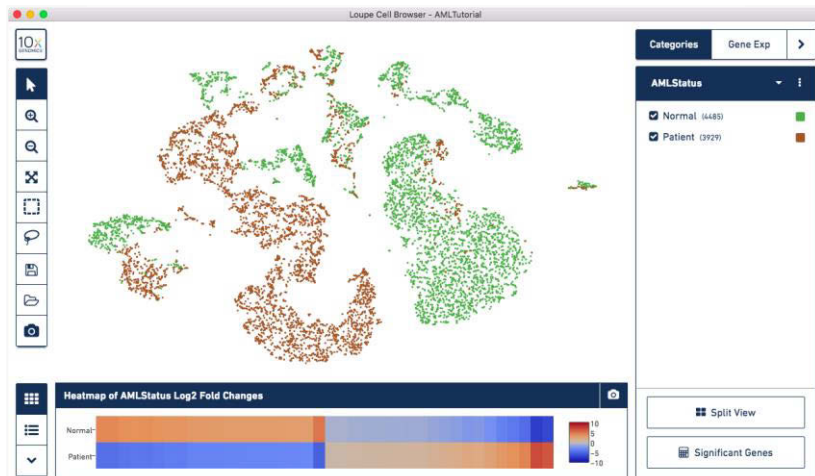
Gene Expression Level



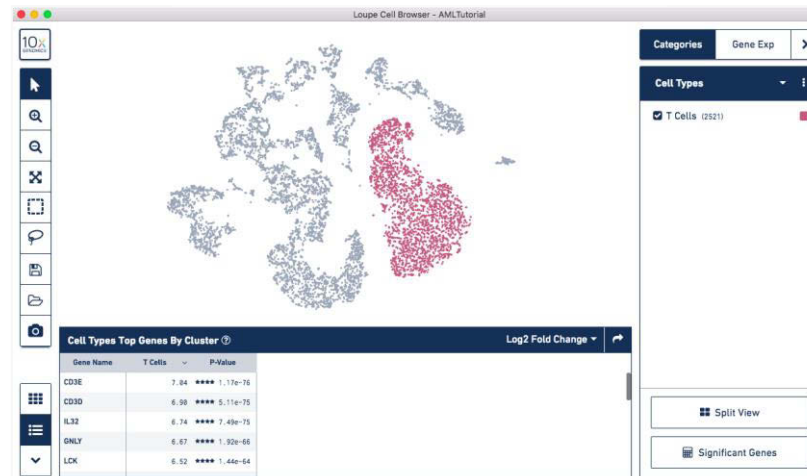
Protein Expression Level



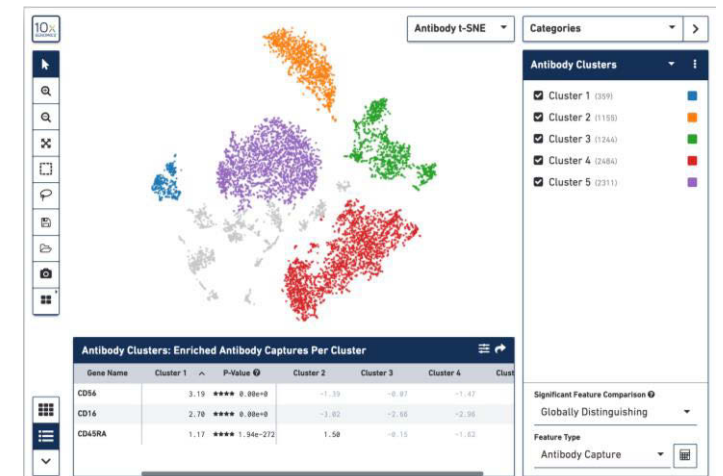
Experimental Conditions



Custom Groups

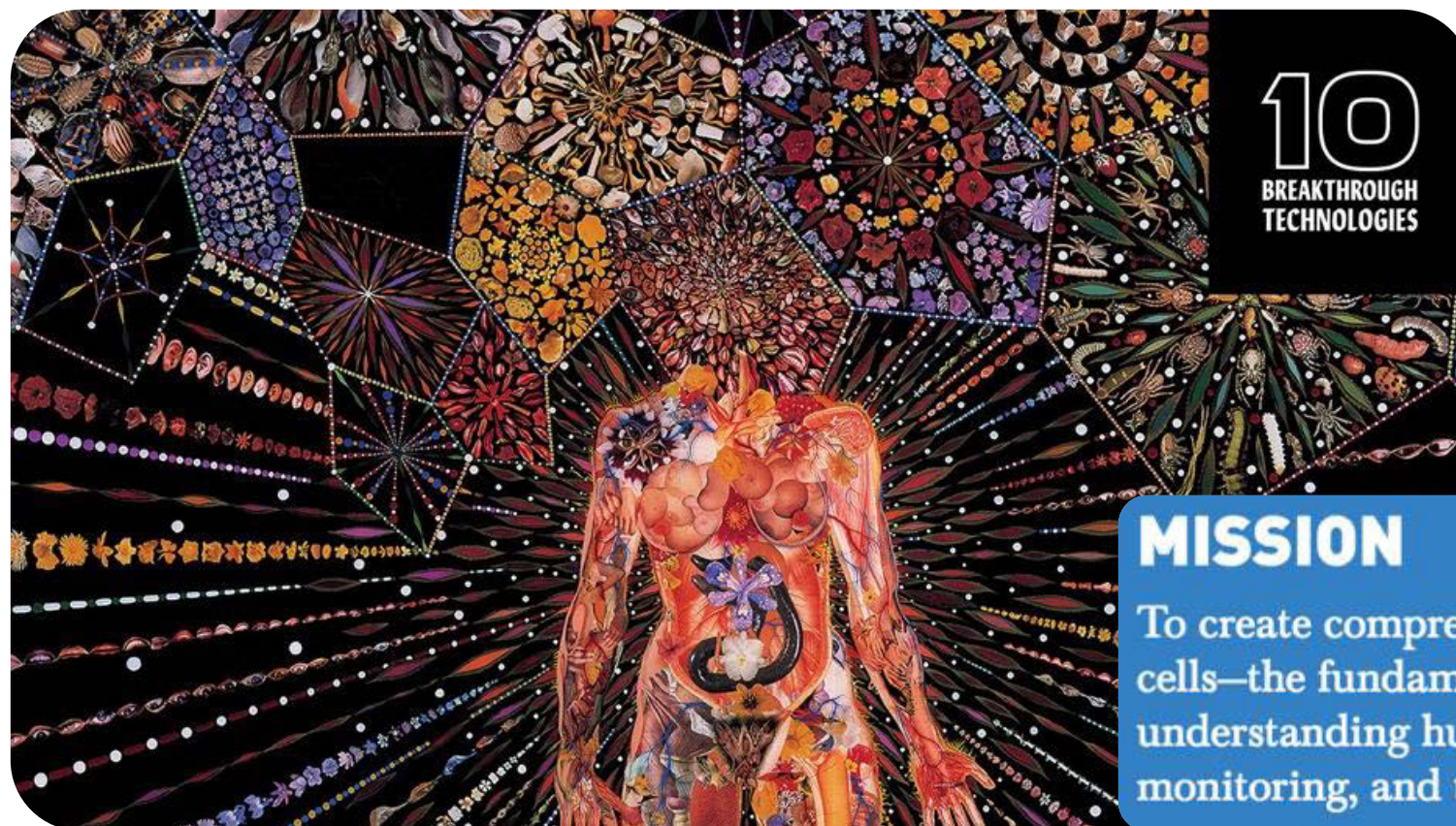


Precomputed Protein Clusters

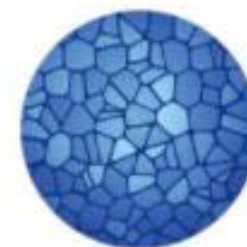


Chromium Single Cell Gene Expression Solution Chosen by the HCA

“Recent advances in single-cell technology have allowed us to look at cells with a clarity and depth of analysis that we have never been able to achieve before, making this ambitious project a reality within reach.” – Aviv Regev



10
BREAKTHROUGH
TECHNOLOGIES



HUMAN
CELL
ATLAS

MISSION

To create comprehensive reference maps of all human cells—the fundamental units of life—as a basis for both understanding human health and diagnosing, monitoring, and treating disease.

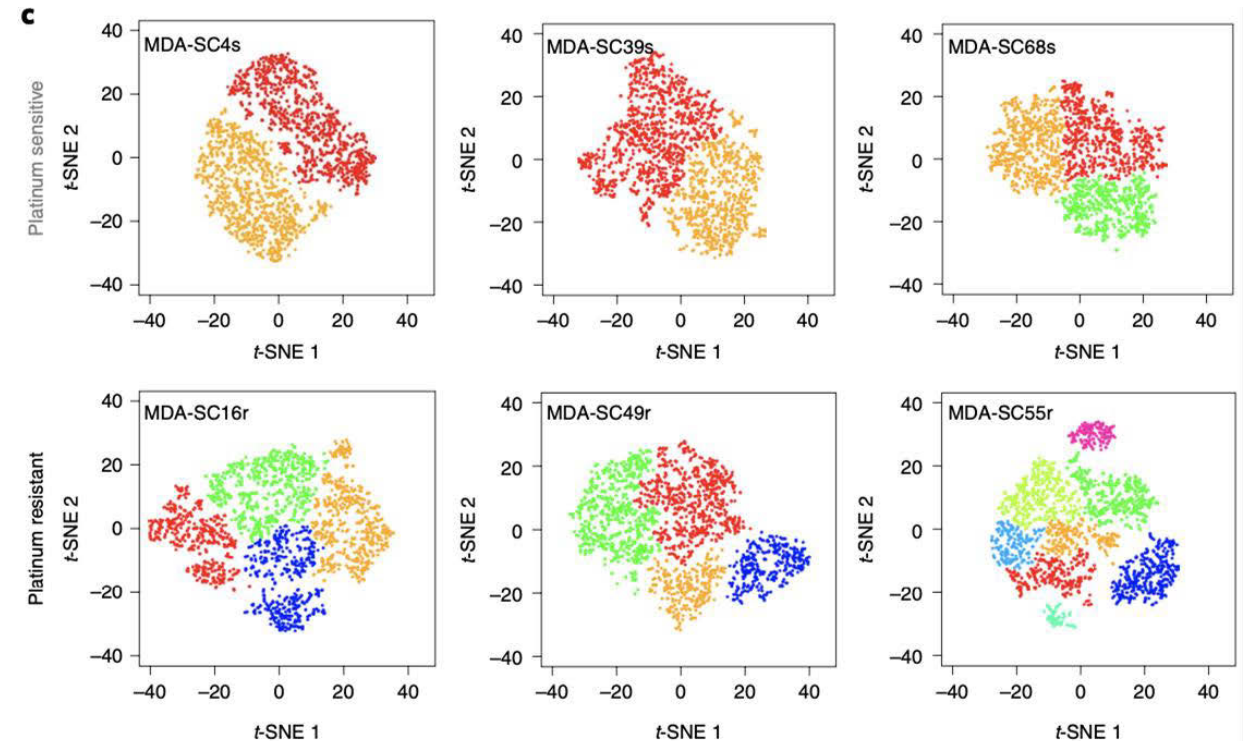
Single cell RNA seq –resolution at the scale



Increased intratumoral heterogeneity after onset of therapy resistance in small-cell lung cancer

Stewart et al. Nature Cancer 2020

- SCLC is known to have a robust response to treatment followed by a relapse
- Authors used single cell RNA sequencing to look at tumor heterogeneity after treatment
- Authors observed globally increased intratumoral heterogeneity (ITH) post-treatment, including expression of therapeutic targets and resistant pathways following treatment resistance



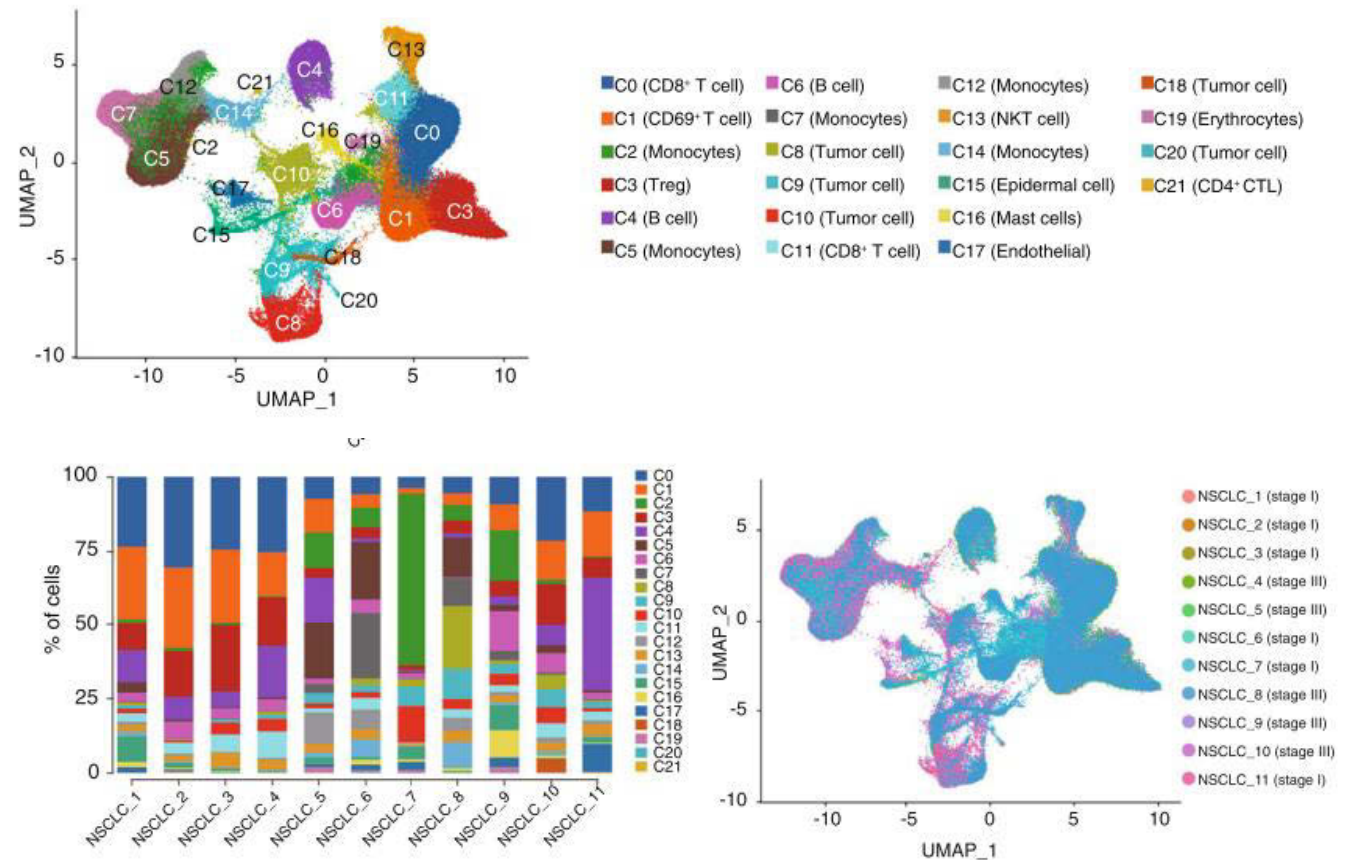
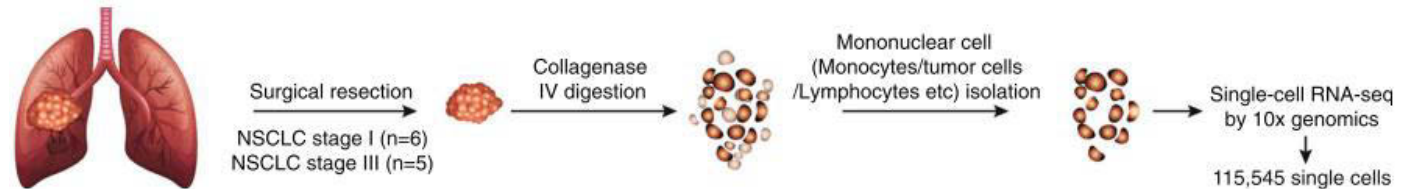
Single-cell transcriptome and antigen-immunoglobulin analysis reveals the diversity of B cells in non-small cell lung cancer

Malignant transformation and progression of cancer are driven by the coevolution of cancer cells and their dysregulated tumor microenvironment (TME).

Recent studies on immunotherapy demonstrate the efficacy in reverting the anti-tumoral function of T cells, highlighting the therapeutic potential in targeting certain cell types in TME. However, the functions of other immune cell types remain largely unexplored.

The naïve-like B cells are decreased in advanced NSCLC, and their lower level is associated with poor prognosis.

Plasma-like B cells in the advanced-stage NSCLC promote tumor cell proliferation while those in the early-stage NSCLC partially inhibit the proliferation.



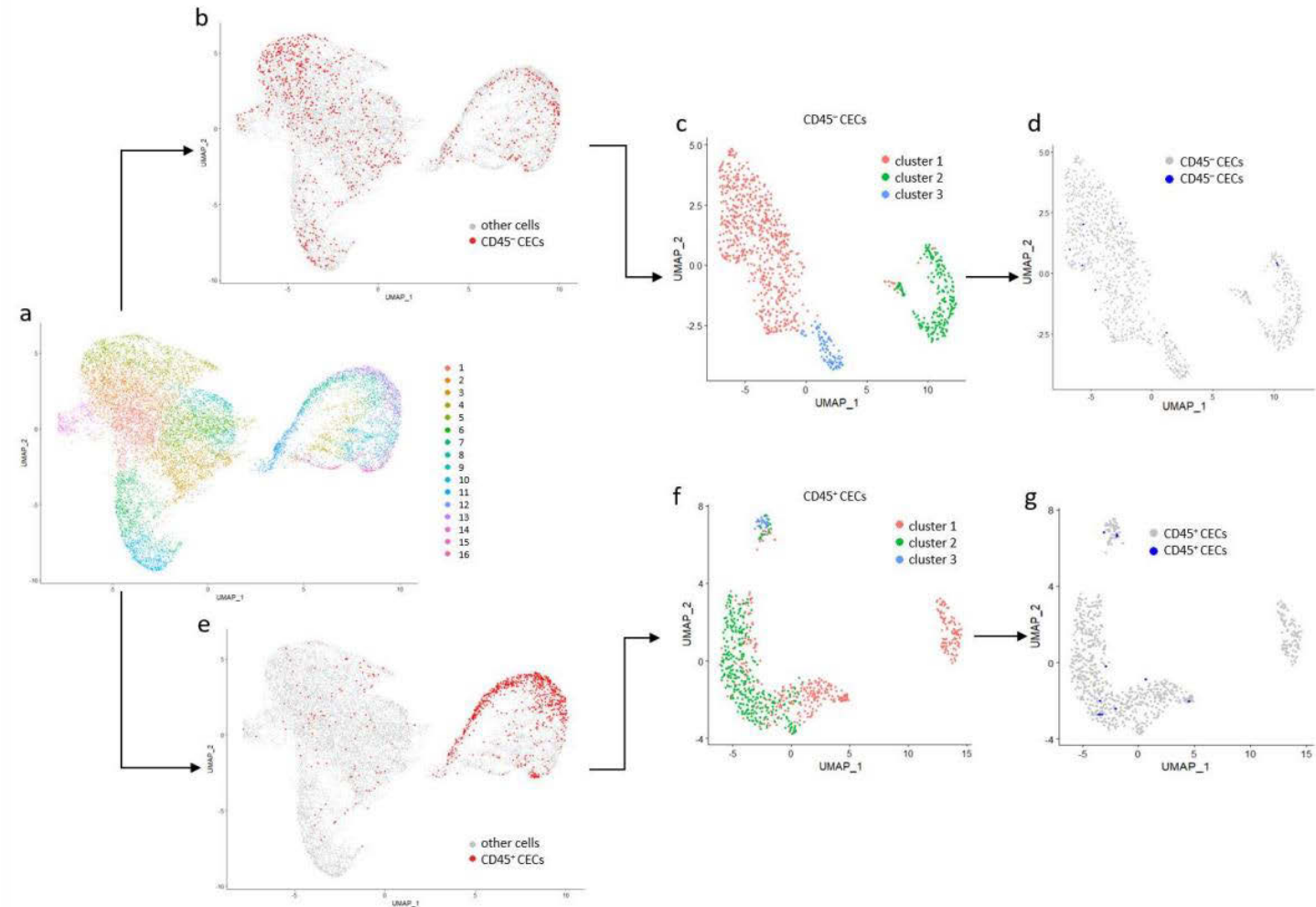
Deciphering heterogeneity of circulating epithelial cells in breast cancer patients

Circulating tumor cells (CTCs) and tumor hybrid cells, being the leading players in metastasis, have prognostic relevance and are potential antimetastatic targets.

CTCs are identified as epithelial-positive and CD45 (leukocyte)-negative cells, whereas tumor hybrid cells usually have epithelial and leukocyte components. However, epithelial and hybrid cells are also observed in healthy subjects that complicate the detection of CTCs and tumor hybrid cells in cancer patients.

This study evaluated the diversity of CD45-negative and CD45-positive circulating epithelial cells (CECs) in breast cancer patients (n=20) using single-cell RNA sequencing.

CD45⁻ and CD45⁺ CECs are highly heterogeneous in breast cancer patients and consist of transcriptionally-distinct cell populations

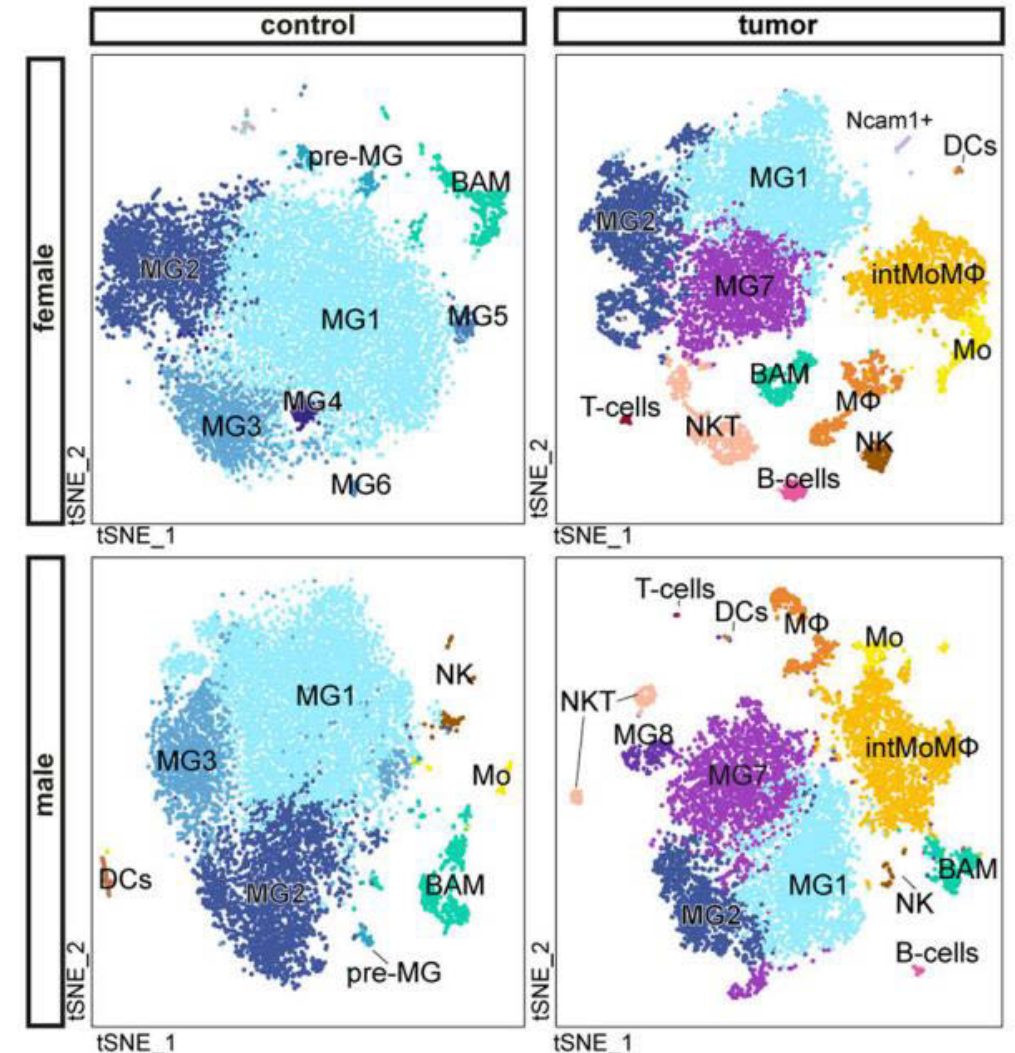


Single-cell RNA sequencing reveals functional heterogeneity of glioma-associated brain macrophages

Microglia and peripheral myeloid cells accumulate and adapt tumor supporting roles in human glioblastomas that show prevalence in men.

Cell heterogeneity and functional phenotypes of myeloid subpopulations in gliomas remain elusive. Here is shown scRNA-seq of CD11b⁺ myeloid cells in naïve and GL261 glioma-bearing mice that reveal distinct profiles of microglia, infiltrating monocytes/macrophages and CNS border-associated macrophages.

It is demonstrated an unforeseen molecular heterogeneity among myeloid cells in naïve and gliomabearing brains, validate selected marker proteins and show distinct spatial distribution of identified subsets in experimental gliomas.



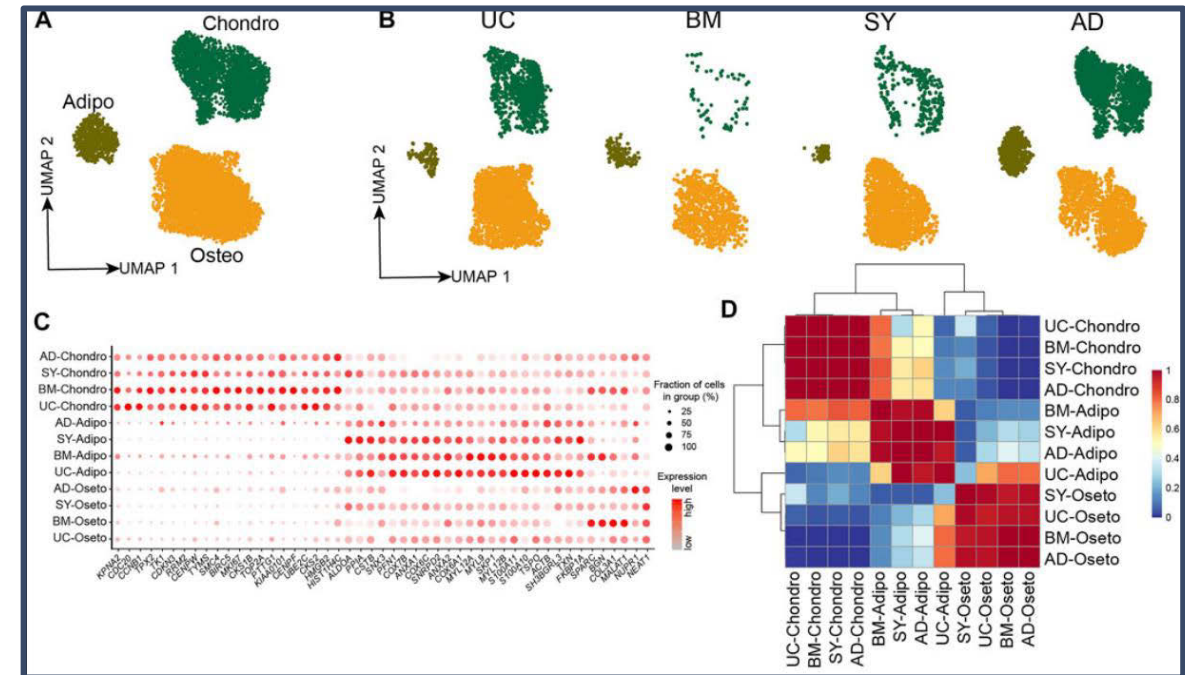
Cross-Tissue Characterization of Heterogeneities of Mesenchymal Stem Cells and Their Differentiation Potentials

Mesenchymal stem/stromal cells (MSCs) are multipotent stromal cells that can differentiate into a variety of cell types including chondrocytes, osteocytes, myocytes, and adipocytes *in vivo* and *in vitro*.

Mesenchymal stem/stromal cells (MSCs) are promising cell sources for regenerative medicine and the treatment of autoimmune disorders. Comparing MSCs from different tissues at the single-cell level is fundamental for optimizing clinical applications.

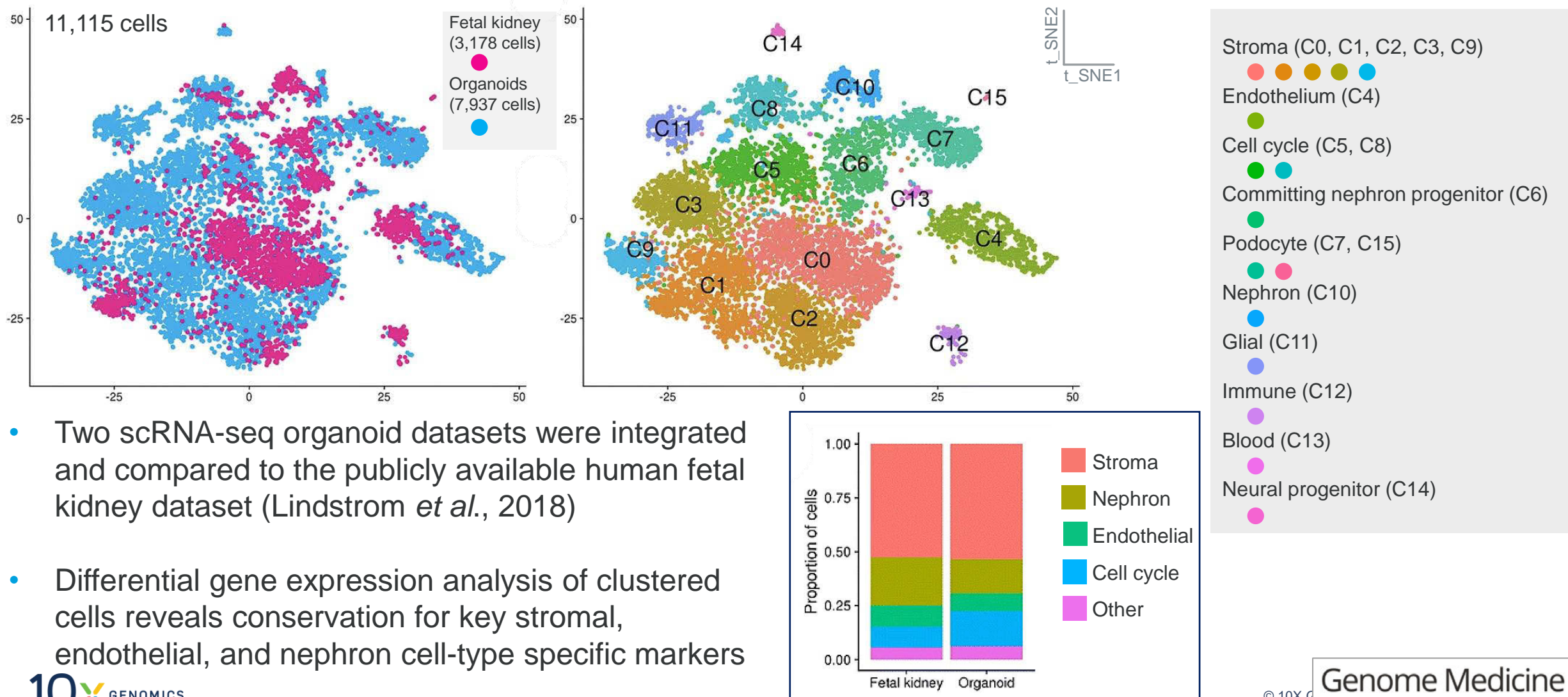
Here was analyzed scRNA-seq data of MSCs from four tissues, : **umbilical cord, bone marrow, synovial tissue, and adipose tissue**.

Was identified three major cell subpopulations, namely osteo-MSCs, chondro-MSCs, and adipo/myo-MSCs, across all MSC samples. MSCs from the umbilical cord exhibited the highest immunosuppression, potentially indicating it is the best immune modulator for autoimmune diseases



scRNA-Seq Showed Consistent MSC
Subpopulations Across Tissues

Cell Type Congruency in Organoids and Fetal Kidney Affirms the Fidelity of Human Kidney Organoids as Models



- Two scRNA-seq organoid datasets were integrated and compared to the publicly available human fetal kidney dataset (Lindstrom *et al.*, 2018)
- Differential gene expression analysis of clustered cells reveals conservation for key stromal, endothelial, and nephron cell-type specific markers

Single-Cell Analyses Identify Brain Mural Cells Expressing CD19 as Potential Off-Tumor Targets for CAR-T Immunotherapies

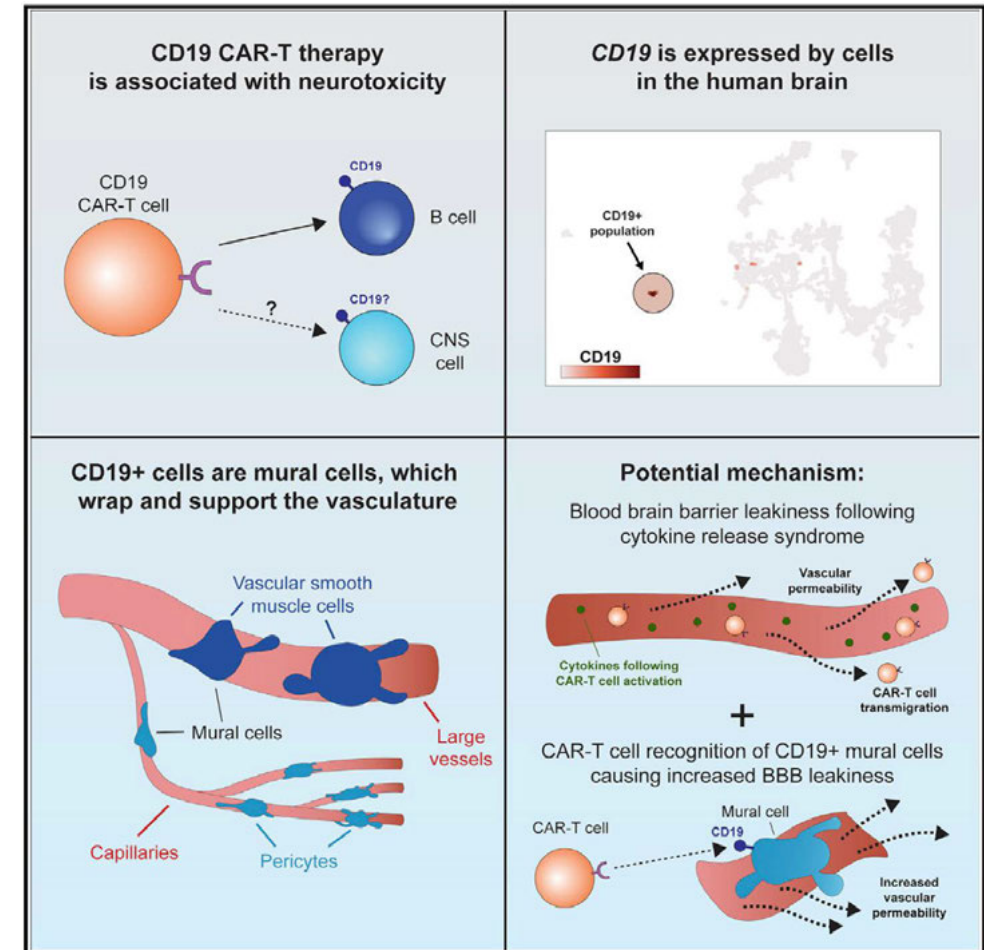
Cell

Chromium single-cell RNA sequencing analysis showed that CD19, primarily considered as a B cell-specific surface antigen, is expressed in human brain mural cells that are critical for blood-brain-barrier integrity.

This cell population may contribute to the neurotoxicity of CD19-directed immunotherapy including CAR-T.

Mouse mural cells demonstrated lower levels of Cd19 expression, suggesting limitations in preclinical animal models of neurotoxicity.

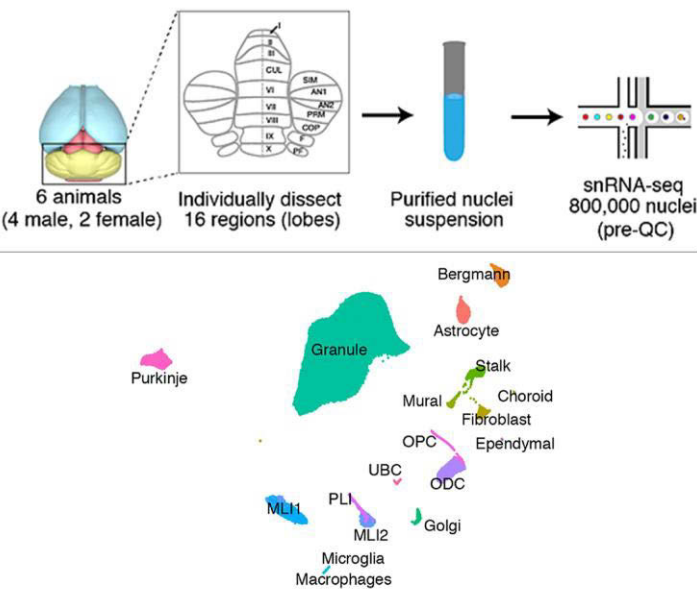
These data highlight the utility of human single-cell atlases for designing immunotherapies.



Detection of rare cell types/states

Neuroscience

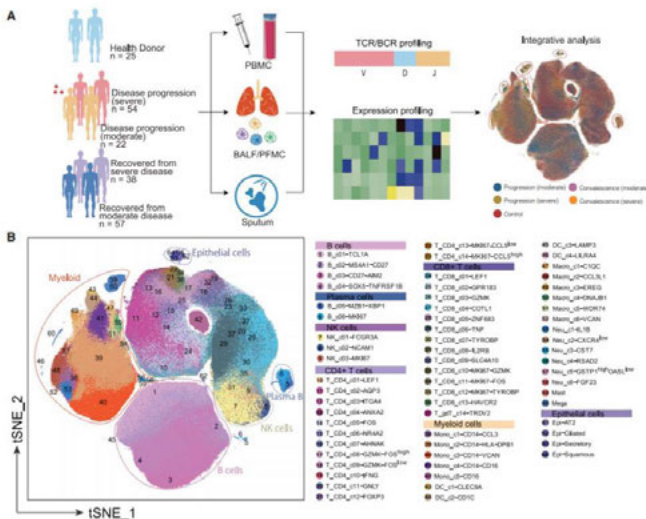
Kozareva *et al*, Nature, 2021



>600.000 nuclei profiled.
A transcriptomic atlas of the mouse cerebellum comprehensively defines cells types.

Infectious Disease

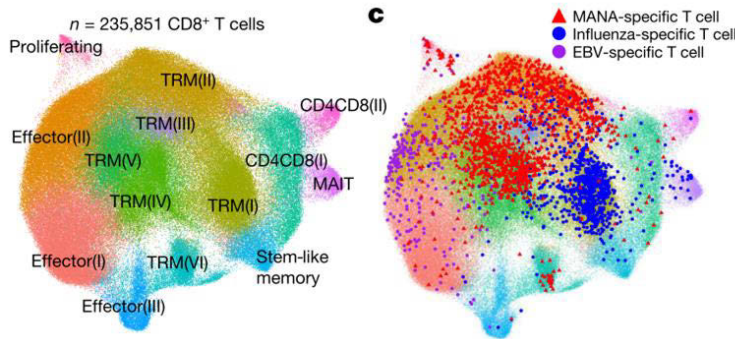
Ren *et al*, Cell, 2021



Detailed COVID-19 immune landscape depicted by integrated **1.46 million single cells**.
Association of immune subtypes with biological and clinical features.

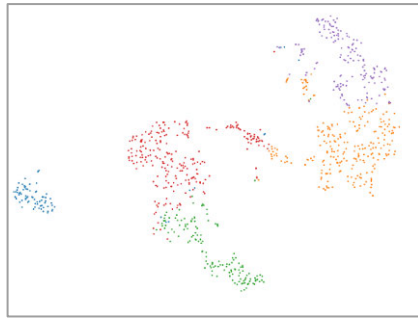
Immune-Oncology

Caushi *et al*, Nature, 2021



>560.000 cells profiled in total including 235.000 T cells.
1350 Mutation Associated NeoAntigens-specific TCR detected in anti-PD1-treated lung cancers.

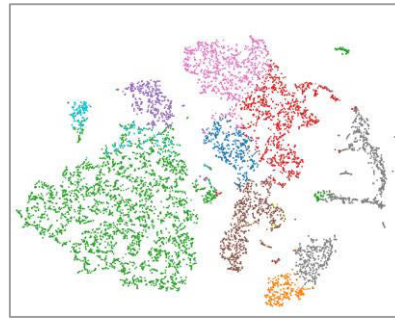
Match the scale you need for your Single Cell Gene Expression studies



Small scale

100-1,000 cells

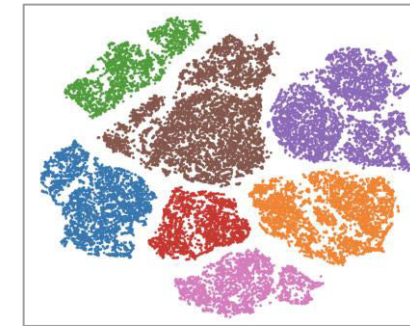
**Single Cell Gene
Expression LT**



Standard scale

500-10,000 cells

**Single Cell Gene
Expression**



Large scale

2,000-20,000 cells

**Single Cell Gene
Expression HT**

GROWING family of Chromium systems



Chromium Controller



Chromium iX

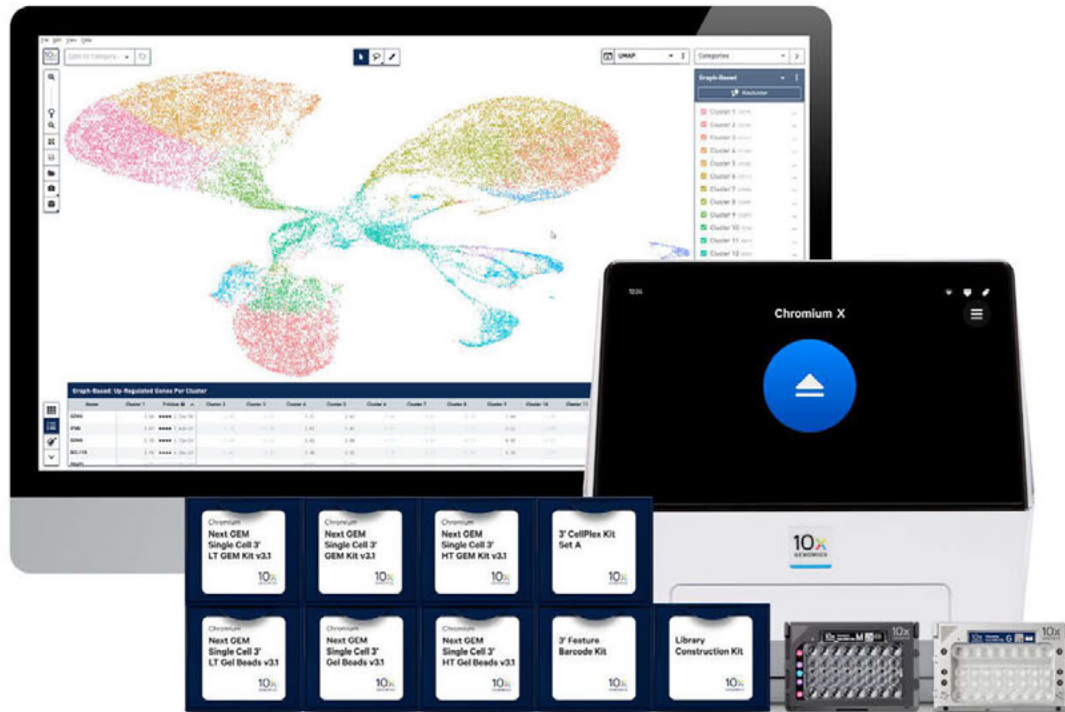


Chromium X



Chromium X Series

Enabling your most ambitious studies, affordably



Standard Throughput

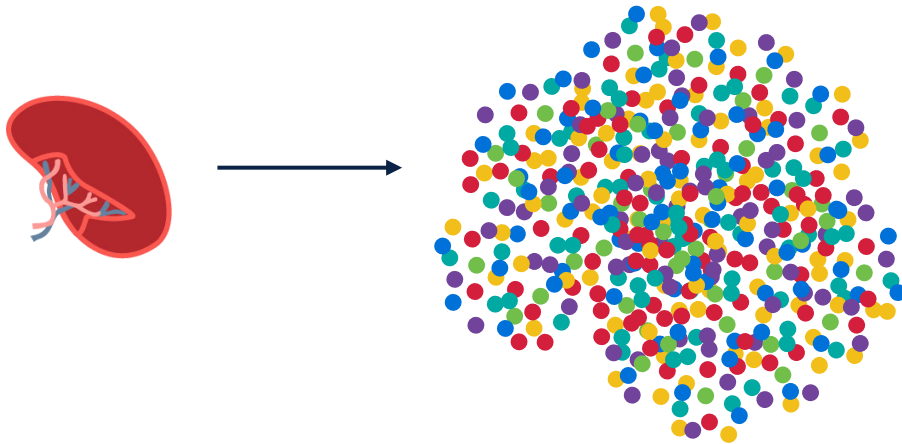


High Throughput



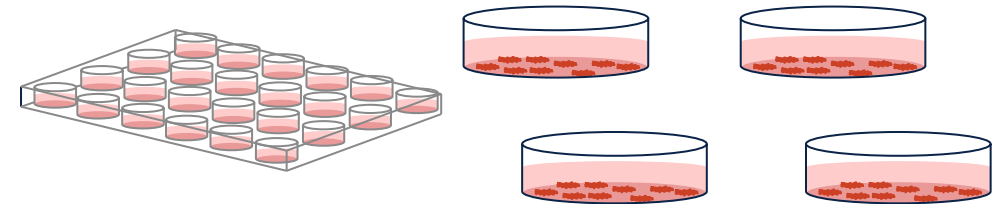
Two types of high throughput

Large number of cells



A few samples at a time, focus on high cell recovery

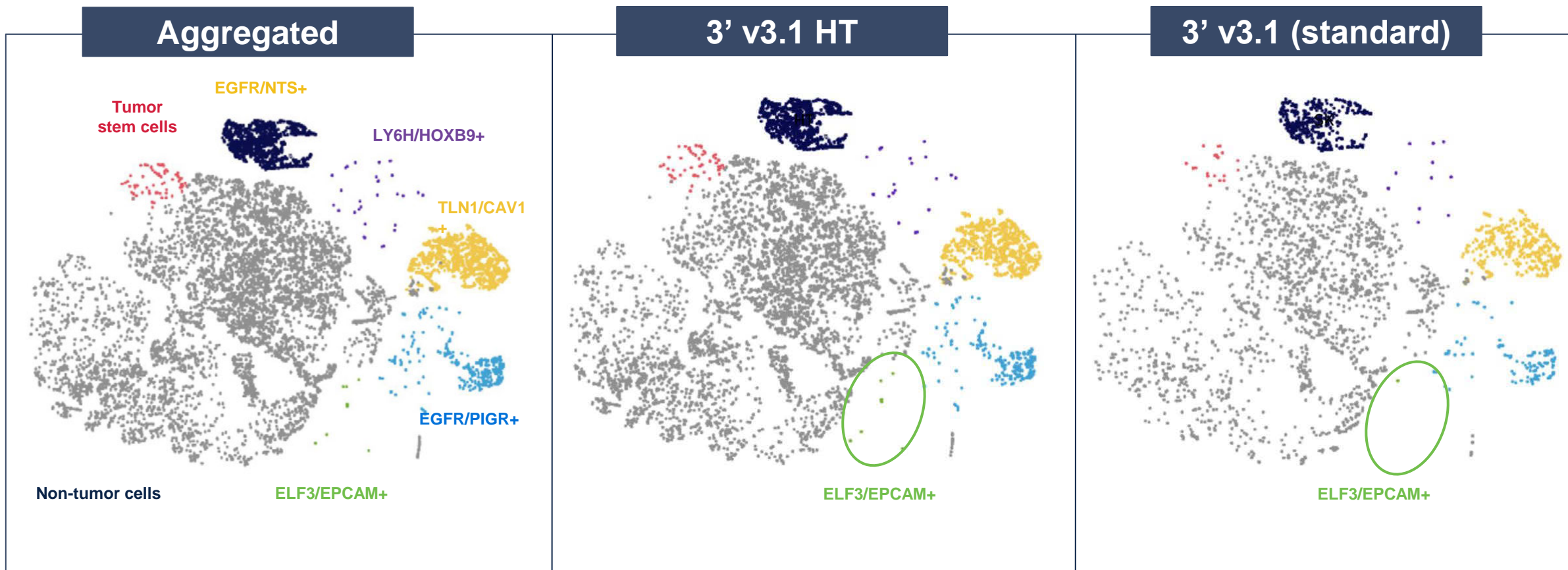
Large number of samples



More samples, easily prepared at the same time

HT enables identification of rare tumor clusters - NCSLC

Deeper look at donor 6



Powering Translational and Clinical Research

Genentech
A Member of the Roche Group

Article

Peripheral T cell expansion predicts tumour infiltration and clinical response

Thomas D. Wu^{1,2}, Shravan Madireddi³, Patricia E. de Almeida⁴, Romain Banchereau⁵, Ying-Jiun J. Chen⁶, Avantika S. Chitre⁷, Eugene Y. Chiang⁸, Hina Iftikhar⁹, William E. O'Gorman¹⁰, Amelia Au-Yeung¹¹, Chikara Takahashi¹², Leonard D. Goldstein¹³, Chungkee Poon¹⁴, Shilpa Keerthivasan¹⁵, Denise E. de Almeida Nagata¹⁶, Xiangnan Du¹⁷, Hyang-Mi Lee¹⁸, Karl L. Banta¹⁹, Sanjeev Mariathasan²⁰, Meghna Das Thakur²¹, Mahesh K. Murthy²², Thomas G. Schumacher²³, Robert S. Ochoa²⁴, Thomas M. Gajewski²⁵, Lellie



Single cell RNA and immune repertoire profiling of COVID-19 patients reveal novel neutralizing antibody

Fang Li¹, Meng Luo², Wenyang Zhou³, Jinliang Li⁴, Xiyun Jin⁵, Zhaochun Xu⁶, Liran Juan⁷, Zheng Zhang⁸, Yuou Li⁹, Renqiang Liu¹⁰, Yiqun Li¹¹, Chang Xu¹², Kexin Ma¹³, Huimin Cao¹⁴, Jingwei Wang¹⁵, Pingping Wang¹⁶, Zhigao Bu¹⁷, Qinghua Jiang^{18,19}

**UNIVERSITY of
WASHINGTON**

Article

Multi-Omics Resolves a Sharp Disease-State Shift between Mild and Moderate COVID-19

Yapeng Su¹, Daniel Chen², Dan Yuan^{3,4}, Christopher Lausted⁵, Jongchan Choi⁶, Chengzhen L. Dai⁷, Valentin Voillet^{8,9}, Venkata R. Duvvuri¹⁰, Kelsey Scherler¹¹, Pamela Troisch¹², Priyanka Baloni¹³, Guangrong Qin¹⁴, Brett Smith¹⁵, Sergey A. Kornilov¹⁶, Clifford Rostomily¹⁷, Alex Xu¹⁸, Jing Li¹⁹, Shen Dong²⁰, Alissa Rothchild²¹, Jing Zhou²², Kim Murray²³, Rick Edmark²⁴, Sunga Hong²⁵, John E. Heath²⁶, John Earls²⁷, Rongyu Zhang²⁸, Jingyi Xia²⁹, Sarah Li³⁰, Ryan Roper³¹, Lesley Jones³², Yong Zhou³³, Lee Rowen³⁴, Rachel Liu³⁵, Sean Mackay³⁶, D. Shane O'Mahony^{37,38}, Christopher R. Dale^{39,40}, Julie A. Wallick^{41,42}, Heather A. Algren^{43,44}, Michael A. Zager⁴⁵, the ISB-Swedish COVID19 Biobanking Unit, Wei Wei⁴⁶, Nathan D. Price⁴⁷, Sui Huang⁴⁸, Naeha Subramanian^{49,50}, Kai Wang⁵¹, Andrew T. Magis⁵², Jenn J. Hadlock⁵³, Leroy Hood⁵⁴, Alan Aderem⁵⁵, Jeffrey A. Bluestone⁵⁶, Lewis L. Lanier⁵⁷, Philip D. Greenberg^{58,59}, Raphael Gottardo^{60,61,62}, Mark M. Davis^{63,64,65}, Jason D. Goldman^{66,67,68}, and James R. Heath^{69,70}



GlaxoSmithKline



Original Article

Lymphocyte Activation Gene (LAG)-3 Is Associated With Mucosal Inflammation and Disease Activity in Ulcerative Colitis

Stephanie M. Slevin^{1,2}, Lucy C. Garner^{3,4,5}, Conor Lahiff^{6,7}, Malcolm Tan^{8,9}, Lai Mun Wang¹⁰, Helen Ferry¹¹, Borgel Greenaway¹², Kate Lynch¹³, Alessandra Geremia¹⁴, Stephen Hughes¹⁵, Karen Leavens¹⁶, David Krull¹⁷, Daniel J. B. Marks¹⁸, Katherine Nevin¹⁹, Kevin Page²⁰, Naren Srinivasan²¹, Ruth Tarzi²², Paul Klennerman²³, Simon Travis²⁴, Carolina V. Arancibia-Carcamo^{25,26}, Satish Keshav^{27,28}



TGFβ-blockade uncovers stromal plasticity in tumors by revealing the existence of a subset of interferon-licensed fibroblasts

Angelo L. Graue^{1,2}, Beverly Nguyen^{1,2}, David Ruddy³, Tyler Laszewski⁴, Stephanie Schwartz⁵, Jonathan Chang⁶, Julie Chen⁷, Michelle Piquet⁸, Marc Pelletier⁹, Zheng Yan¹⁰, Nathaniel D. Kirkpatrick¹¹, Jincheng Wu¹², Antoine deWeck¹³, Markus Riester¹⁴, Matt Hims¹⁵, Felipe Correa Geyer¹⁶, Joel Wagner¹⁷, Kenzie MacIsaac¹⁸, James Deeds¹⁹, Rohan Diwanji²⁰, Pushpa Jayaraman²¹, Yenyen Yu²², Quincey Simmons²³, Shaobu Weng²⁴, Alina Raza²⁵, Brian Minie²⁶, Mirek Dostalek²⁷, Pavitra Chikigowda²⁸, Vera Ruda²⁹, Oleg Iartchouk³⁰, Naiyan Chen³¹, Raphael Thierry³², Joseph Zhou³³, Iulian Pruteanu-Malinici³⁴, Claire Fabre³⁵, Jeffrey A. Engelman³⁶, Glenn Dranoff³⁷, Viviana Cremaschi³⁸



ARTICLE

Anti-human TREM2 induces microglia proliferation and reduces pathology in an Alzheimer's disease model



Sackler Faculty of Medicine
Tel Aviv University

RESEARCH ARTICLE

Multi-clonal SARS-CoV-2 neutralization by antibodies isolated from severe COVID-19 convalescent donors

Michael C. Lee¹, E. Clark Hila Shih², Sandra Moshe³, T. Freu⁴



RESEARCH ARTICLE SUMMARY

CLINICAL TRIALS

CRISPR-engineered T cells in patients with refractory cancer

Edward A. Stadtmauer^{1,2}, Joseph A. Fraietta³, Megan M. Davis⁴, Adam D. Cohen⁵, Kristy L. Weber⁶, Eric Lancaster⁷, Patricia A. Mangan⁸, Irina Kulikovskaya⁹, Minnal Gupta¹⁰, Fang Chen¹¹, Lifeng Tian¹², Vanessa E. Gonzalez¹³, Jun Xu¹⁴, In-young Jung¹⁵, J. Joseph Melenhorst¹⁶, Gabriela Plesa¹⁷, Joanne Shea¹⁸, Tina Matlawski¹⁹, Amanda Cervini²⁰, Avery L. Gaymon²¹, Stephanie Desjardins²², Anne Lamontagne²³, January Salas-McKee²⁴, Andrew Fesnak²⁵, Donald L. Siegel²⁶, Bruce L. Levine²⁷, Julie K. Jadowsky²⁸, Regina M. Young²⁹, Anne Chew³⁰, Wei-Ting Hwang³¹, Elizabeth O. Hexner³², Beatriz M. Carreno³³, Christopher L. Nobles³⁴, Frederic D. Bushman³⁵, Kevin R. Parker³⁶, Yanyan Qi³⁷, Ansuman T. Satpathy³⁸, Howard Y. Chang³⁹, Yangbing Zhao⁴⁰, Simon F. Lacey⁴¹, Carl H. June⁴²

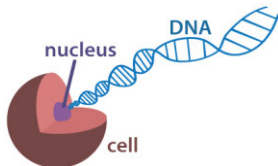
Chromium Single Cell Multiome ATAC + GEX

Two modalities, same cell

Mechanisms of transcriptional regulation

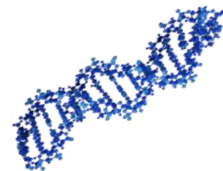
Chromatin remodeling

- **Chromatin structure**
- Chromatin exists in complex hierarchical structures within each cell.
- Rearrangement of chromatin from a condensed state (heterochromatin) to an accessible state (euchromatin) **determines DNA accessibility to DNA binding proteins**



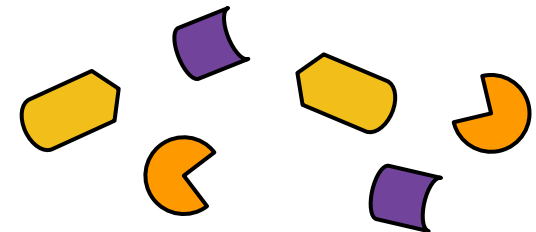
Epigenetic modifications

- Epigenetic modifications
- Modifications on histone proteins can change chromatin conformation.
- Modifications on DNA can prevent or promote protein binding.
- Epigenetic alterations are dynamic and potentially reversible, unlike genetic mutations.

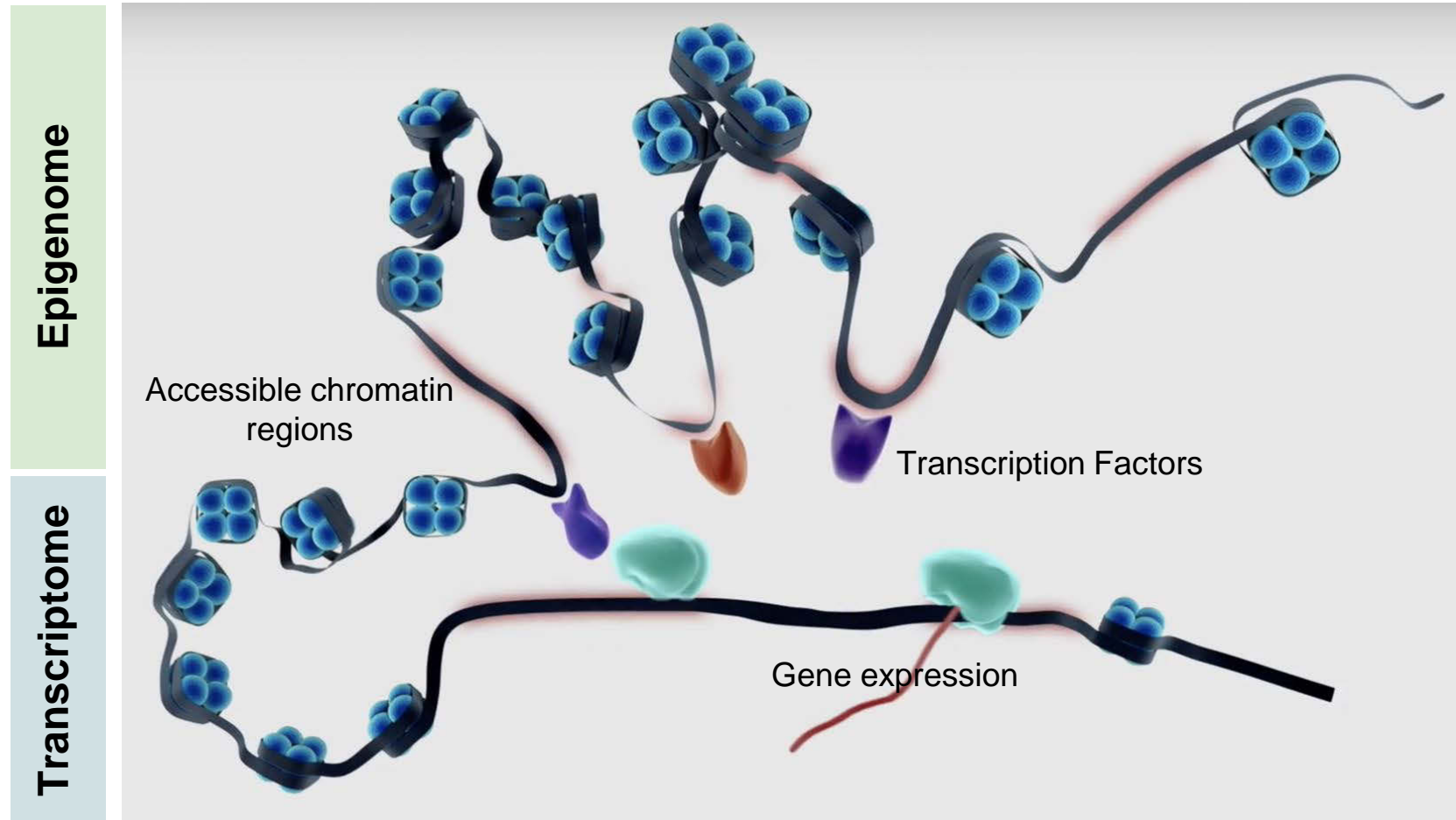


DNA binding proteins

- Transcription factors (TFs) modulate the process of transcription
- TFs can promote (activators), or block (repressors) the recruitment of RNA polymerase to the DNA site



Interplay Between Epigenetic Programs and Gene Expression - ATAC-seq + GEX



analyse
function of cells

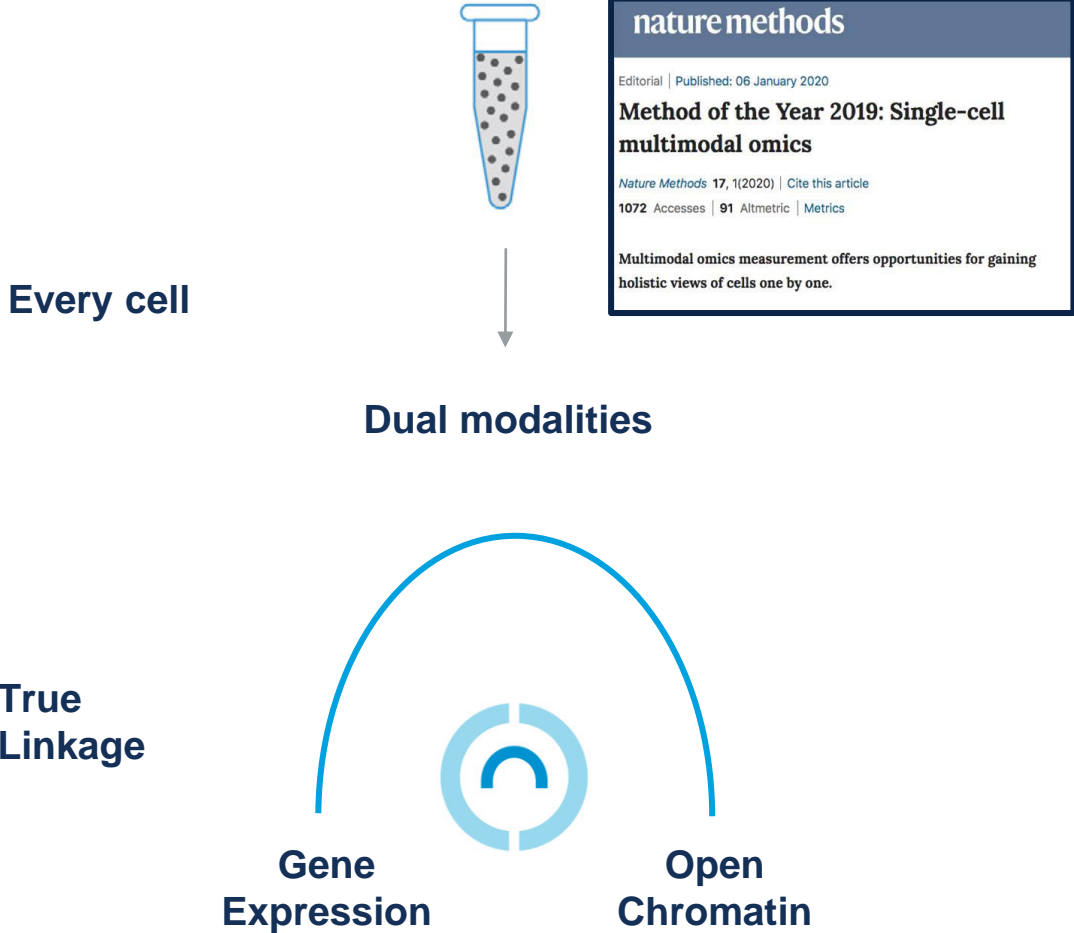
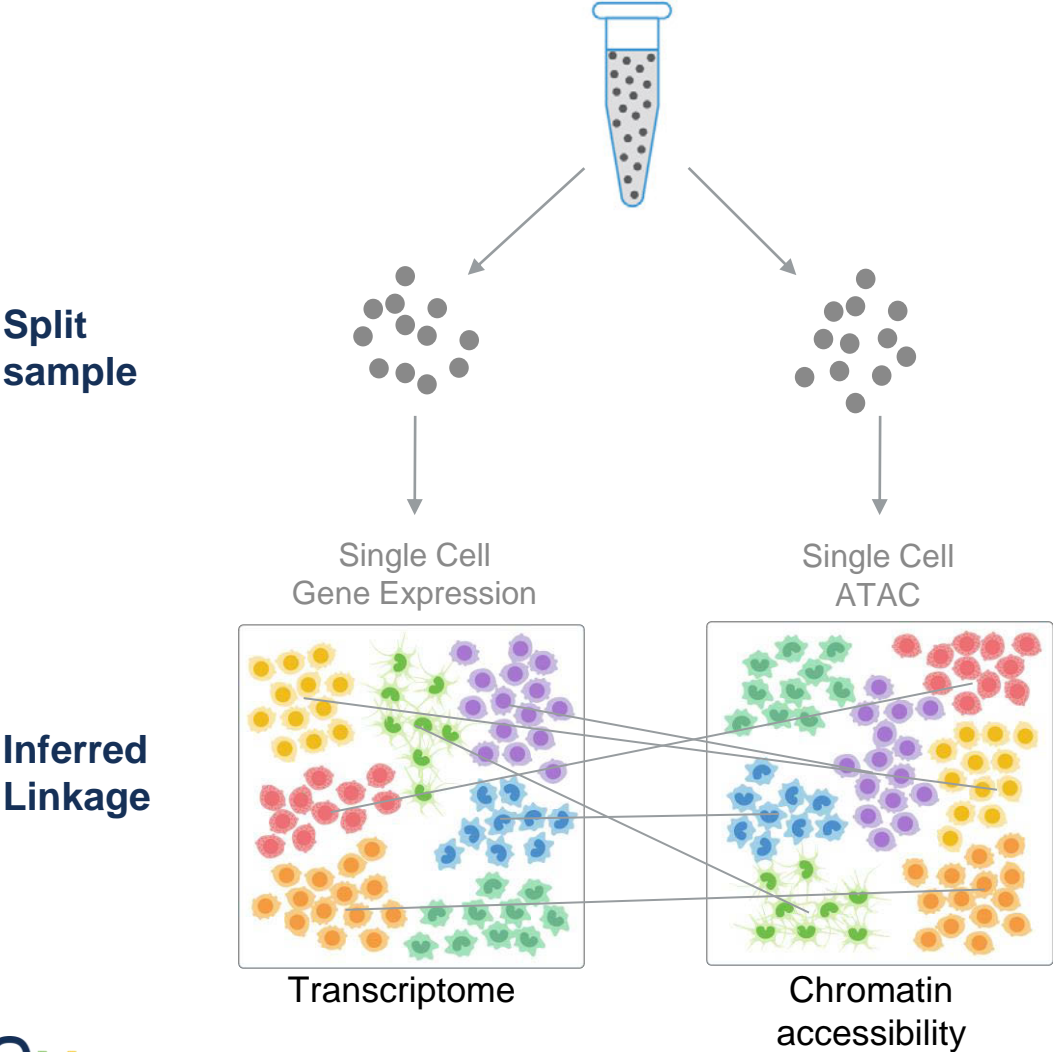
cell atlasing

discovery tool

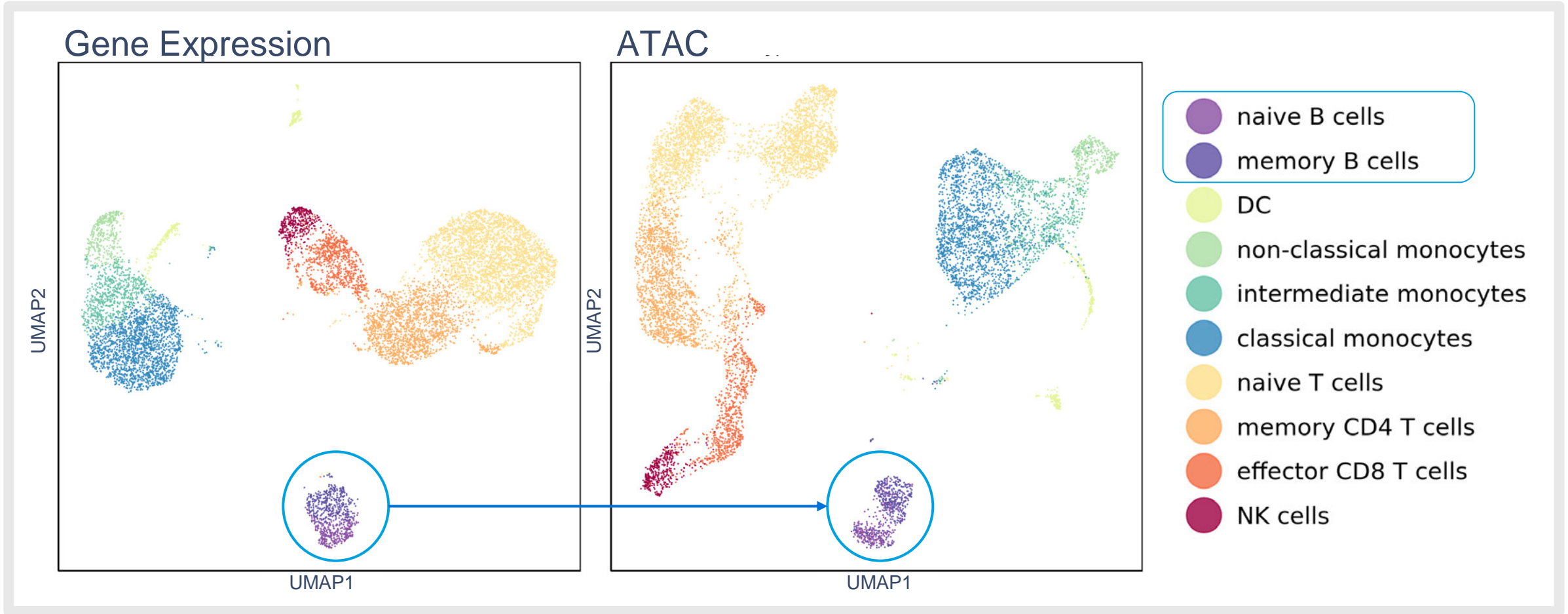
subpopulation
detection

understanding of
biological processes
and pathways

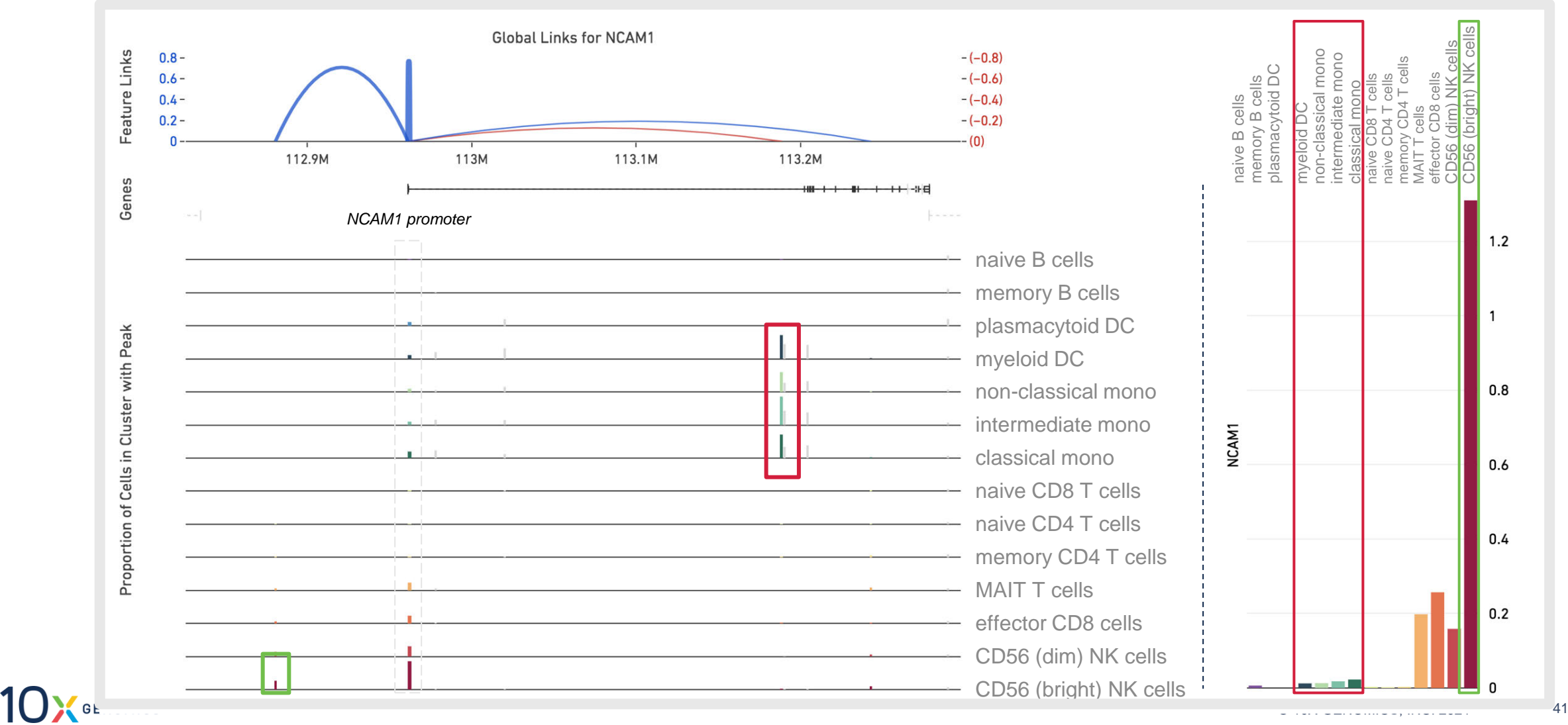
Multimodal approach



Better separate PBMC populations on ATAC space



Link putative regulatory elements to target genes (CD56)



Emergence of a High-Plasticity Cell State (HPCS) during Lung Cancer Evolution

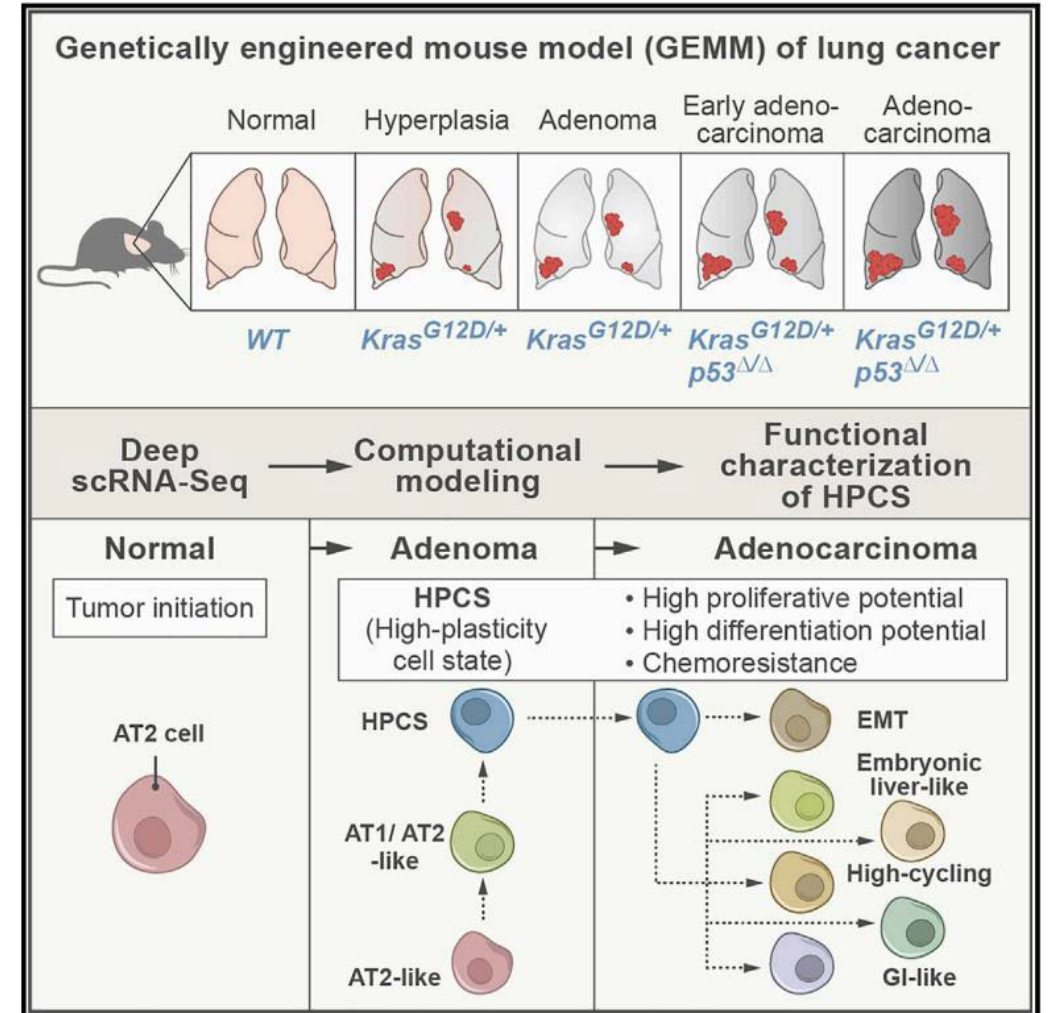
Authors used scRNA-seq to study cell state changes during tumor evolution in a mouse model of LUAD (lung adenocarcinoma) mimicking the oncogenic transformation processes observed in human disease.

Transcriptional heterogeneity grew dramatically during tumor progression.

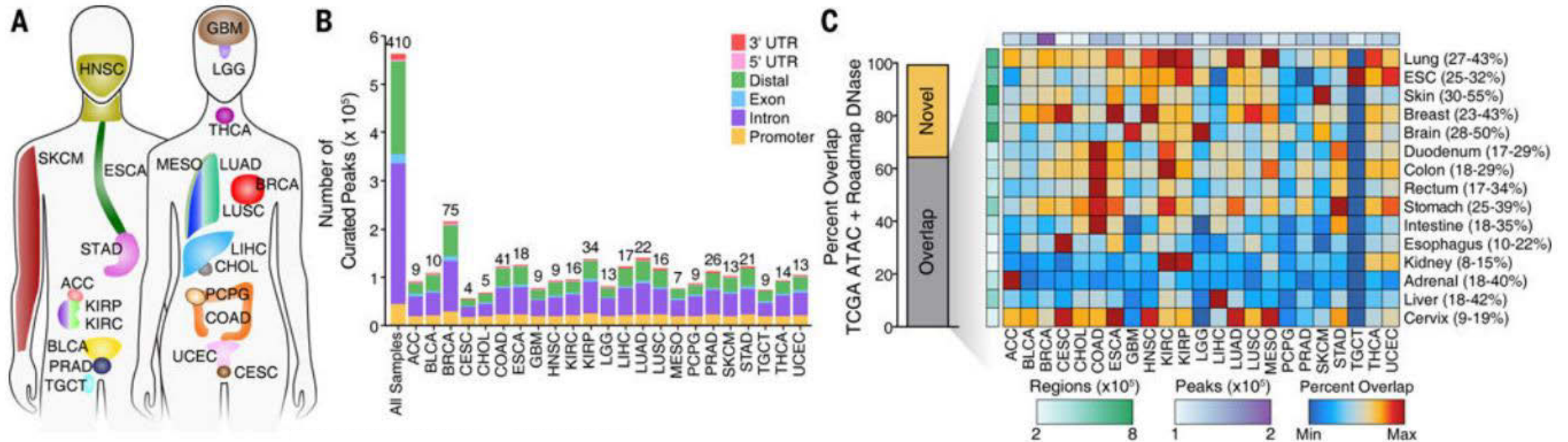
Identification of a Highly Plastic Cell State (HPCS) with a Distinct Chromatin Accessibility Profile (ATAC-seq)

State transitions occur via an HPCS harboring high differentiation and growth capacity

The HPCS is drug resistant and potends poor patient survival across all cancers. Targeting the HPCS may enable therapeutic strategies to suppress tumor heterogeneity and treatment resistance



The chromatin accessibility landscape of primary human cancers



Pan-cancer ATAC-seq of TCGA samples identifies diverse regulatory landscapes. Genome-wide chromatin accessibility profiles of 410 tumor samples spanning 23 cancer types from [The Cancer Genome Atlas \(TCGA\)](#). Were identified 562,709 transposase-accessible DNA elements that substantially extend the compendium of known cis-regulatory elements.

Assaying the molecular genetic regulation of immunology

Combined gene expression and chromatin accessibility measurements

Protein & Cell
https://doi.org/10.1007/s13238-020-00762-2

RESEARCH ARTICLE

A human circulating immune cell landscape in aging and COVID-19

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Received June 22, 2020 Accepted July 1, 2020

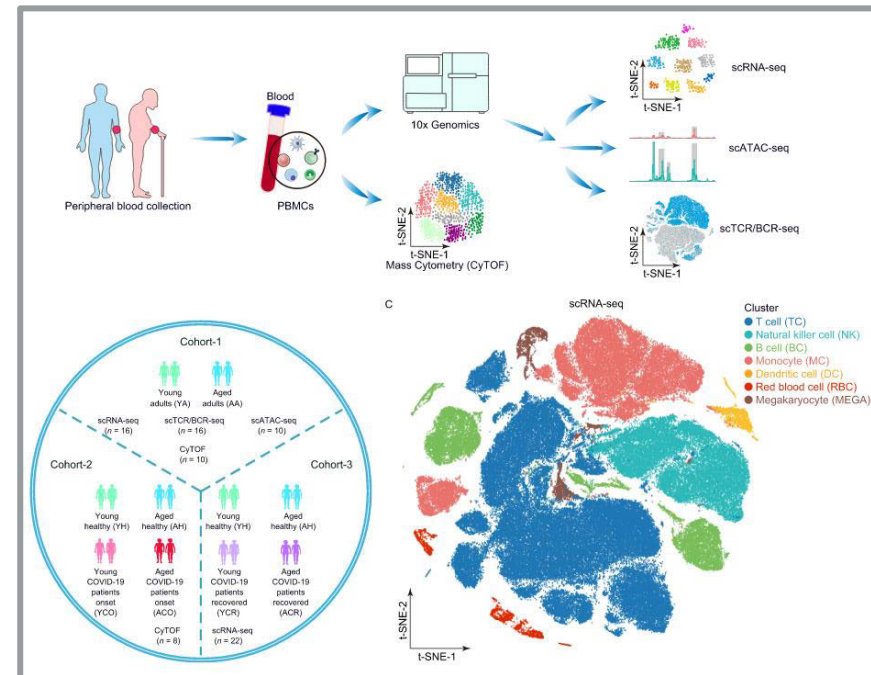
ABSTRACT
Age-associated changes in immune cells have been linked to an increased risk for infection. However, a global and detailed characterization of the changes that human circulating immune cells undergo with age is lacking. Here, we combined scRNA-seq, mass cytometry and scATAC-seq to compare immune cell types in peripheral blood collected from young and old subjects and patients with COVID-19. We found that the immune cell landscape was reprogrammed with age and was characterized by T cell polarization from naive and memory cells to effector, cytotoxic, exhausted and regulatory cells, along with increased late natural killer cells, age-associated B cells, inflammatory monocytes and age-associated dendritic cells. In addition, the expression of genes, which were implicated in coronavirus susceptibility, was upregulated in a cell subtype-specific manner with age. Notably, COVID-19 promoted age-induced immune cell polarization and gene expression related to inflammation and cellular senescence. Therefore, these findings suggest that a dysregulated immune system and increased gene expression associated with SARS-CoV-2 susceptibility may at least partially account for COVID-19 vulnerability in the elderly.

KEYWORDS aging, single-cell sequencing, blood, COVID-19, immune cells

Electronic supplementary material The online version of this article (https://doi.org/10.1007/s13238-020-00762-2) contains supplementary material, which is available to authorized users.

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Published online: 11 August 2020

Protein & Cell

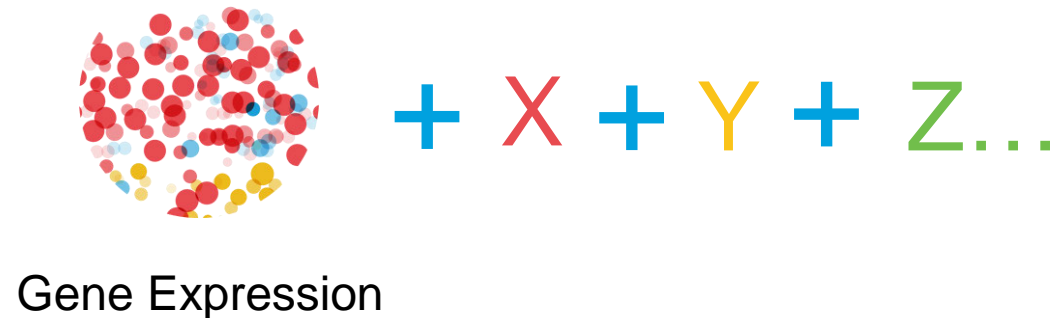
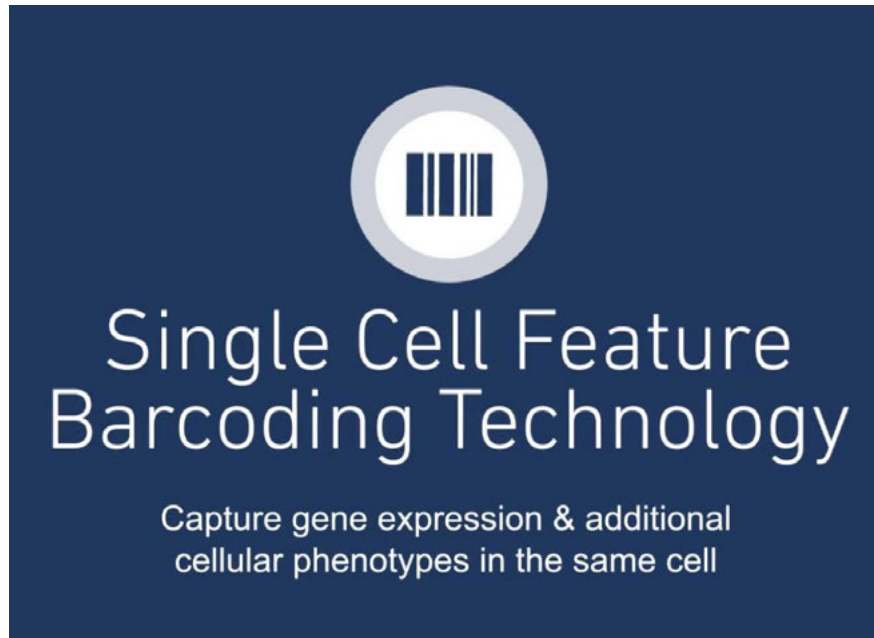


Single-cell chromosomal accessibility profiles of immune cells shows that the **AP-1 family TFs** are the most affected by ageing across all cell types and subtypes and are further upregulated in COVID-19

- Identification of aging-related cell-type-specific transcriptional and chromosomal accessibility changes
- Aging-associated heterogeneous changes in clonality and diversity of TCRs and BCRs
- Aging increases expression of COVID-19 susceptibility genes
- COVID-19 enhances upregulation of aging-induced inflammatory genes

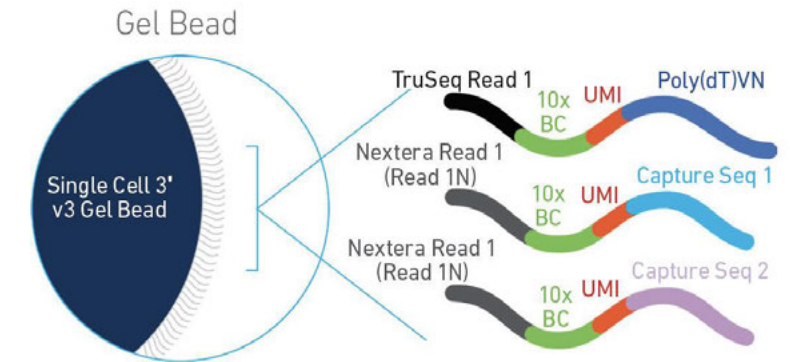
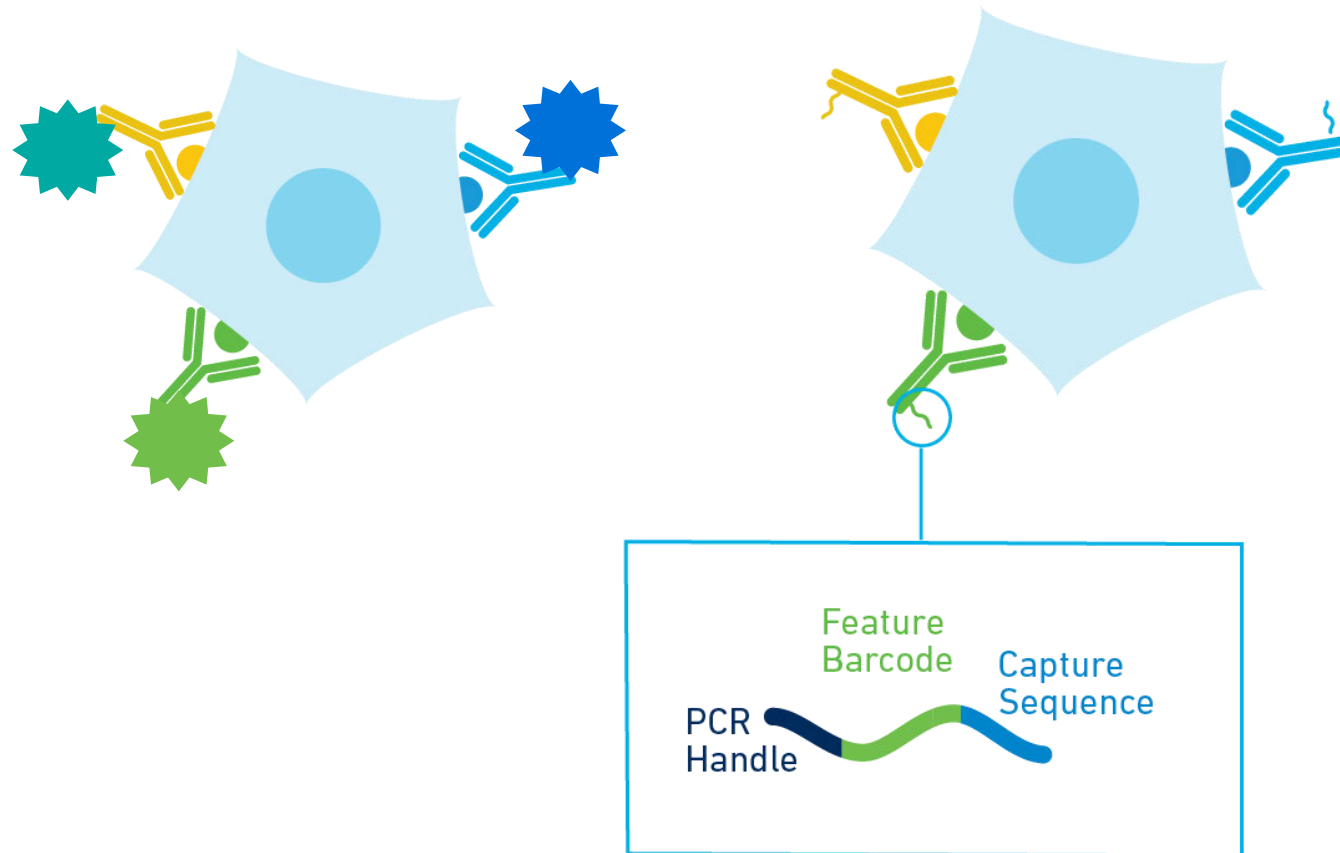
Feature Barcode Technology

Measuring multi-modalities from the same cell



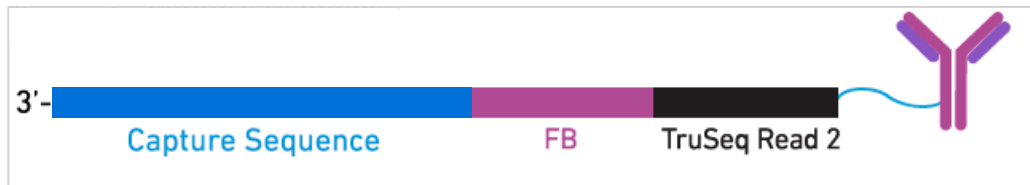
Moving beyond transcriptome ...

Feature Barcode Technology: Molecular (digital) cytometry

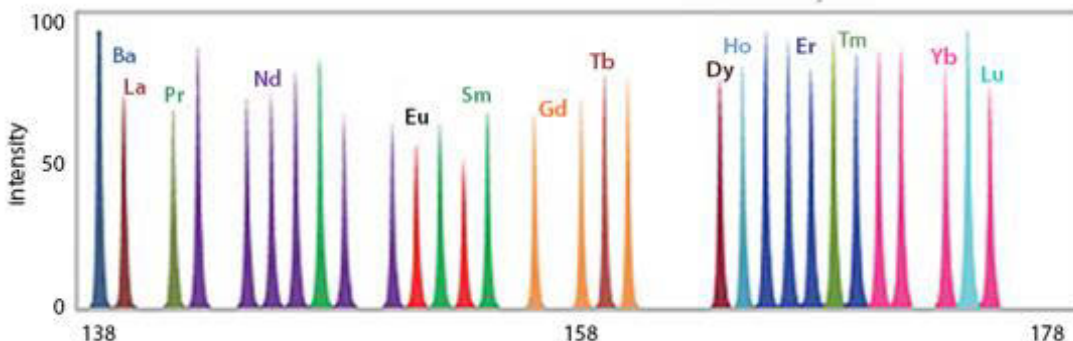


Potential thousands tags to determine cell surface proteins

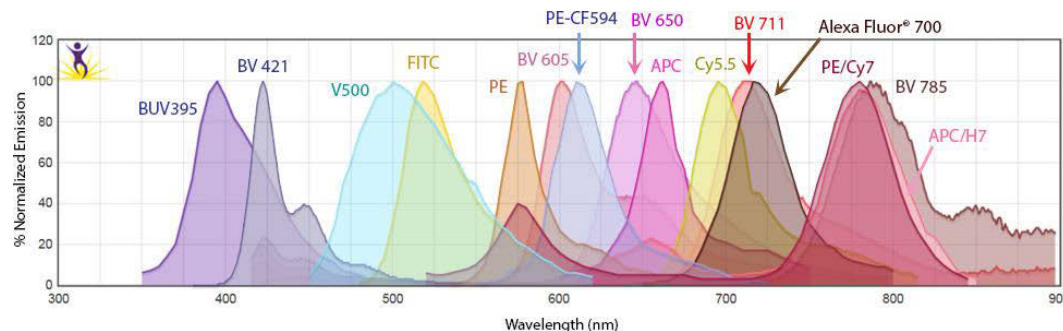
Feature Barcoding Technology
(Distinct 15nt Barcodes)



CyTOF
(Distinct Mass Isotopes)



Flow
(Distinct Emissions)



Scale

> 1 billion parameters (4^{15})

~70 parameters

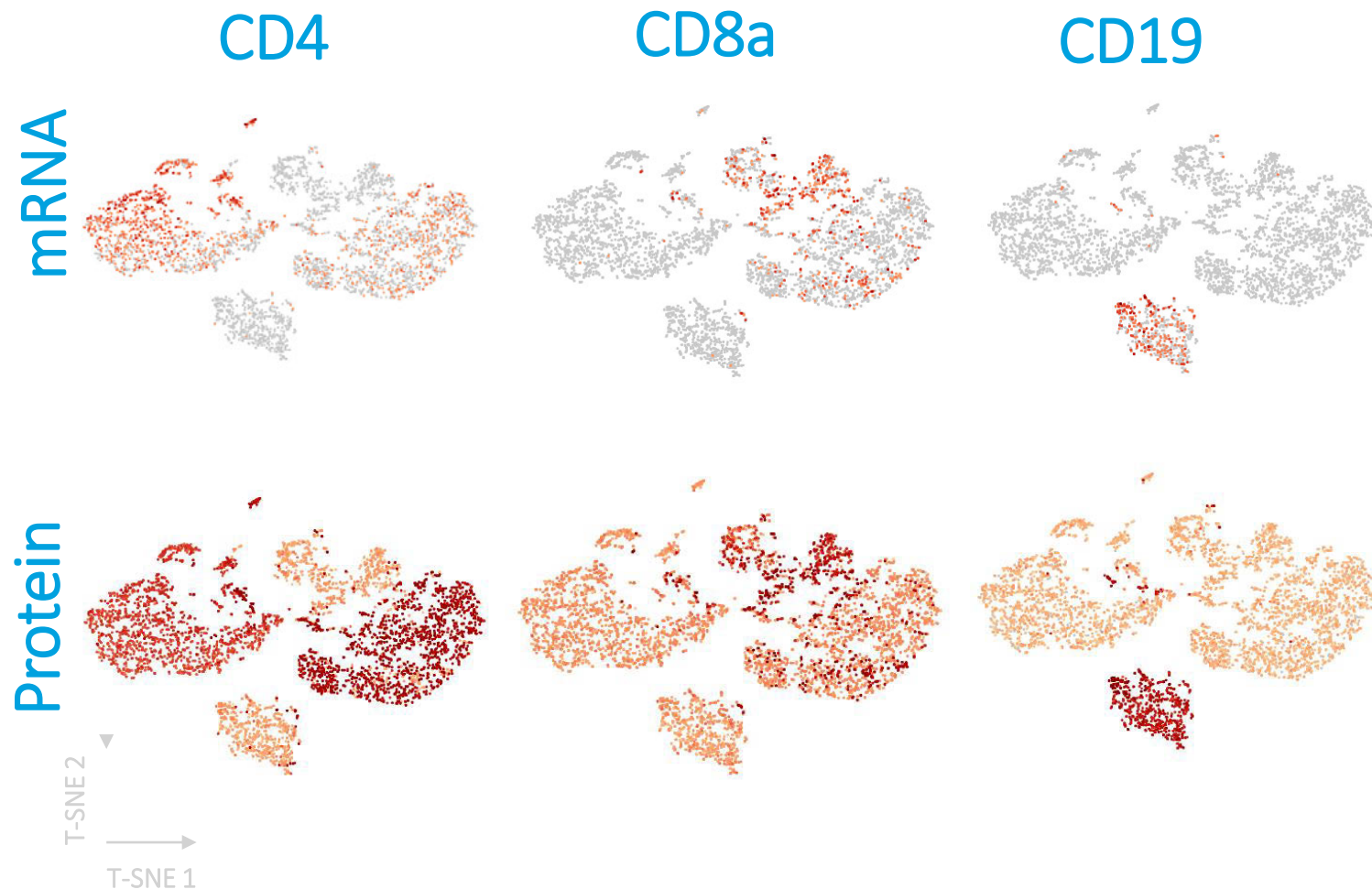
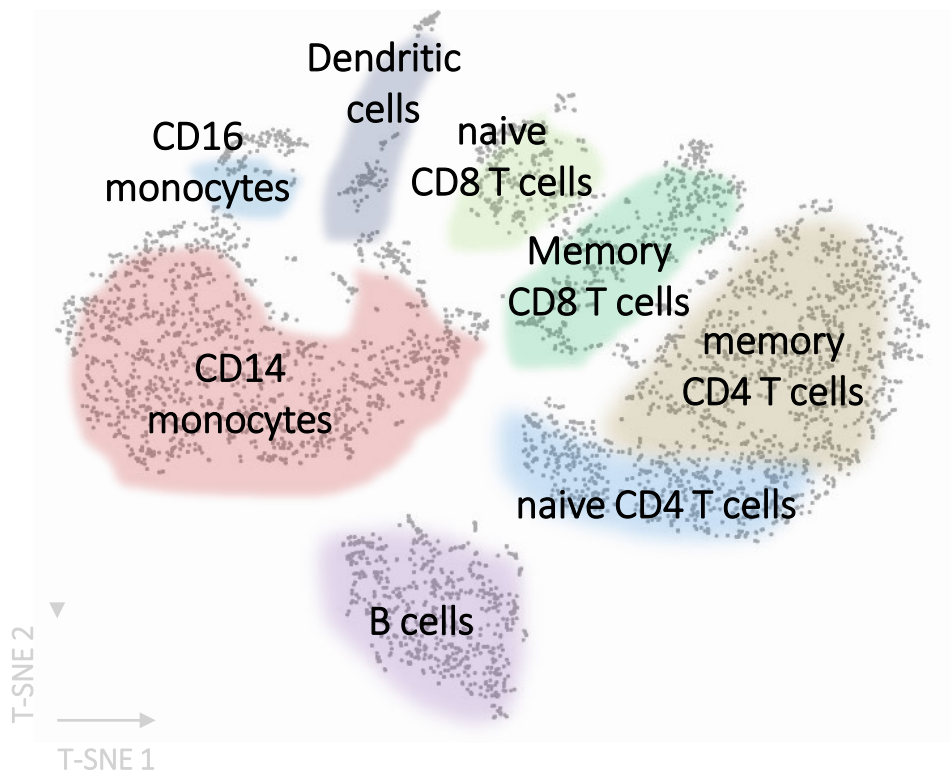
15-20 parameters

Single cell multi omics

Piqsels.com id fsjvu

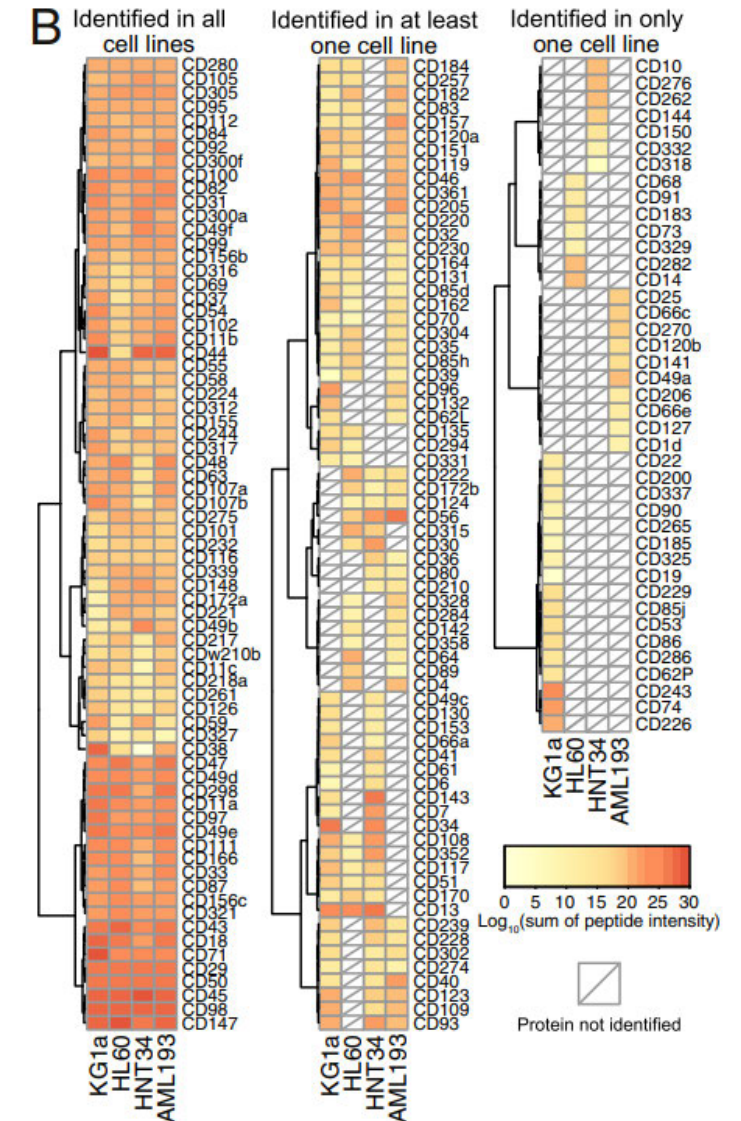


Detecting Low Expressed Biomarkers with Feature Barcoding



Multimomics of azacitidine-treated AML cells reveals variable and convergent targets that remodel the cell-surface proteome

- Acute myeloid leukemia (AML) is a disease of abnormal hematopoietic differentiation with aberrant epigenetic alterations.
- **Azacitidine (AZA) is a DNA methyltransferase inhibitor** widely used to treat MDS and AML
- Yet the impact of AZA on the cell-surface proteome has not been defined. **To identify potential therapeutic targets for use in combination with AZA** in AML patients, we investigated the effects of AZA treatment on four AML cell lines representing different stages of differentiation.
- One gene encoding a surface protein, **TRPM4**, was found to be **commonly up-regulated by AZA treatment** in all four cell lines and may represent a **potential novel therapeutic target for AML in combination with AZA**.



A pan-cancer blueprint of heterogeneous tumor microenvironment revealed by single-cell profiling

J Qian. *Cell Research*, 2020

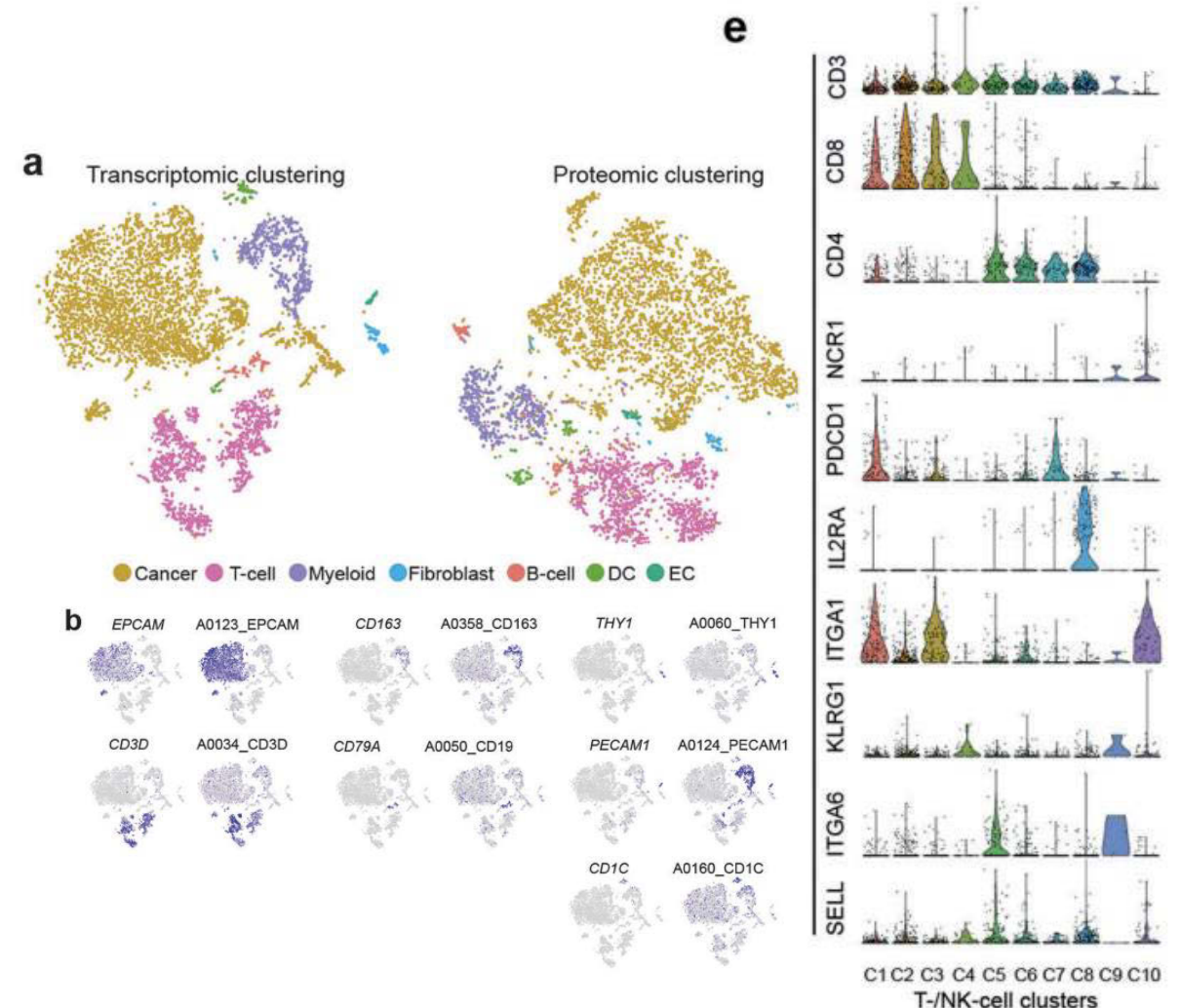
Profiled the transcriptomes of more than 233,000 cells cancer (n=36) (*lung, colorectal, ovary, breast*)

An outstanding question is to what extent this TME heterogeneity is similar between cancers affecting different organs.

Majority of cell phenotypes have previously not been characterized in detail at single-cell level.

Pooled T-/NK-cells with both RNA and protein data together. Selected marker genes amongst the **198 antibodies** and explored protein expression per cluster.

Were identified 68 stromal cell populations, of which 46 are shared between cancer types and 22 are unique.



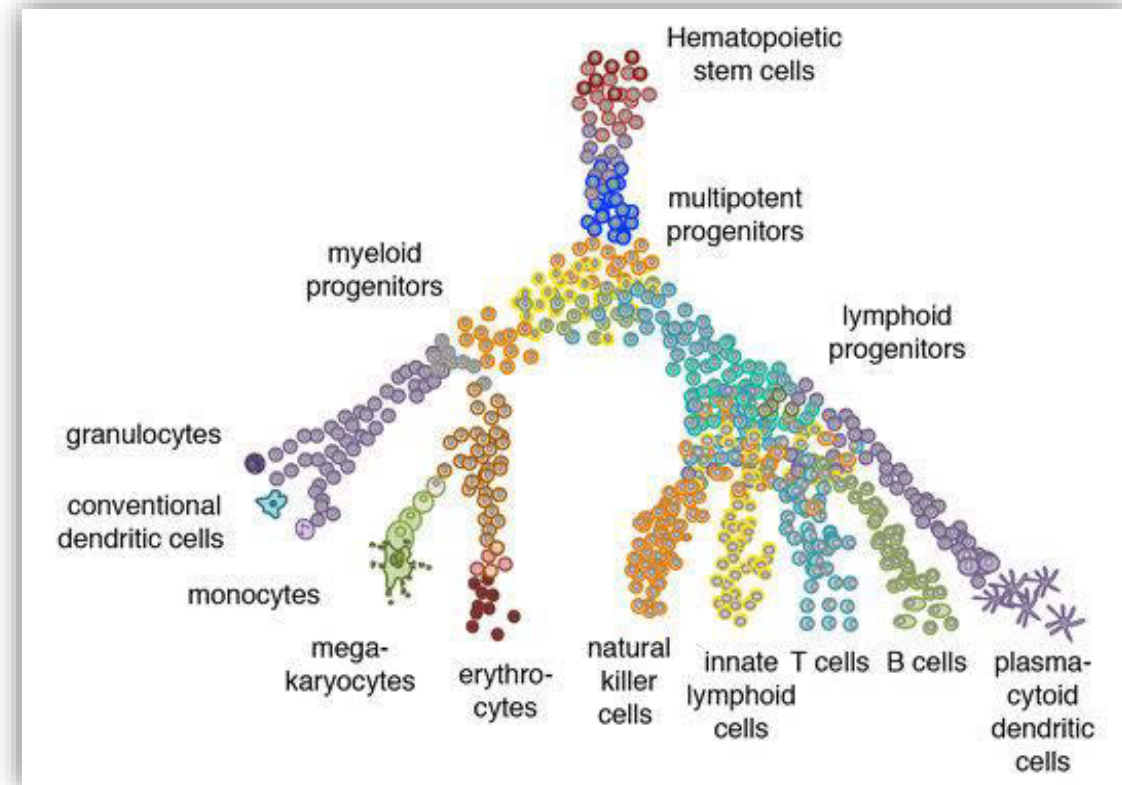
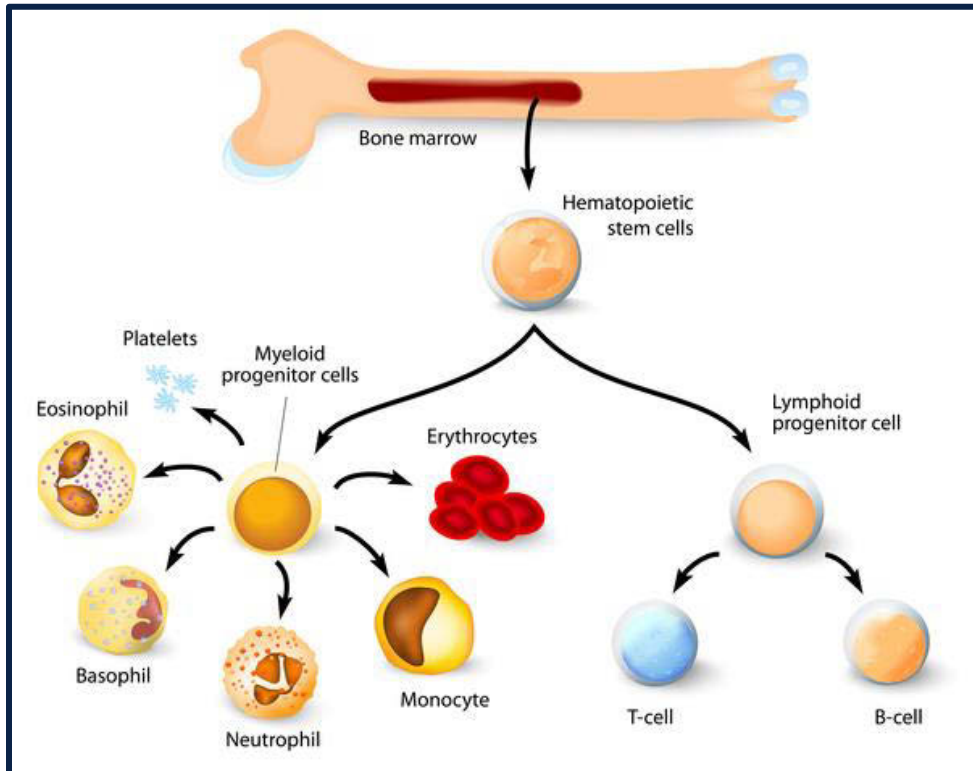
Single Cell Immune Profiling

A „Rosetta Stone” for immunology

Heterogeneity is a common factor in disease

- environment matters

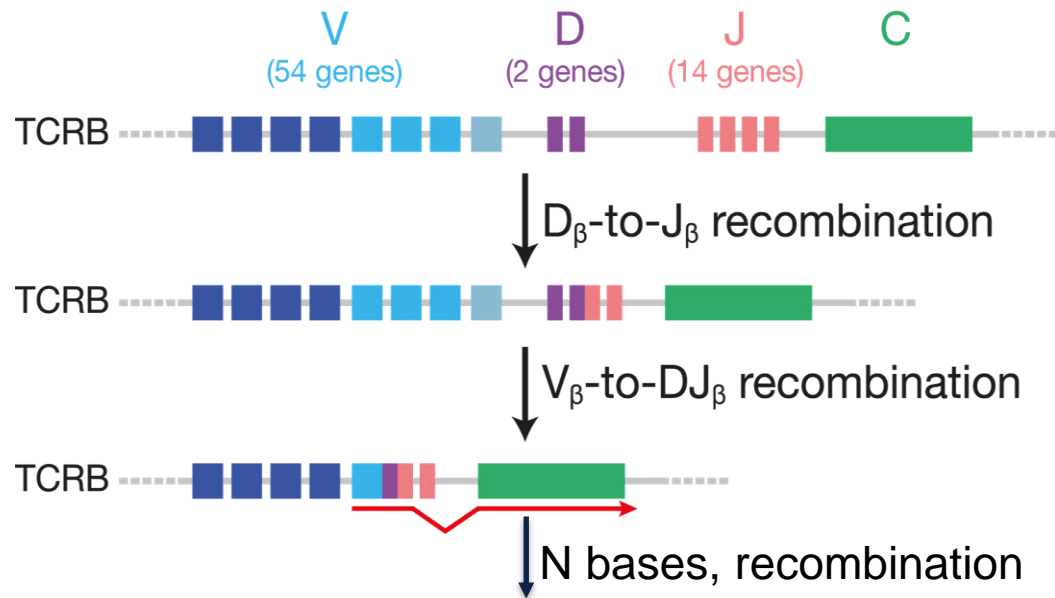
Evaluate Cell States and Progression



Adapted from © Dominic Grün.
https://www.ie-freiburg.mpg.de/4836851/immuncell_differenciation

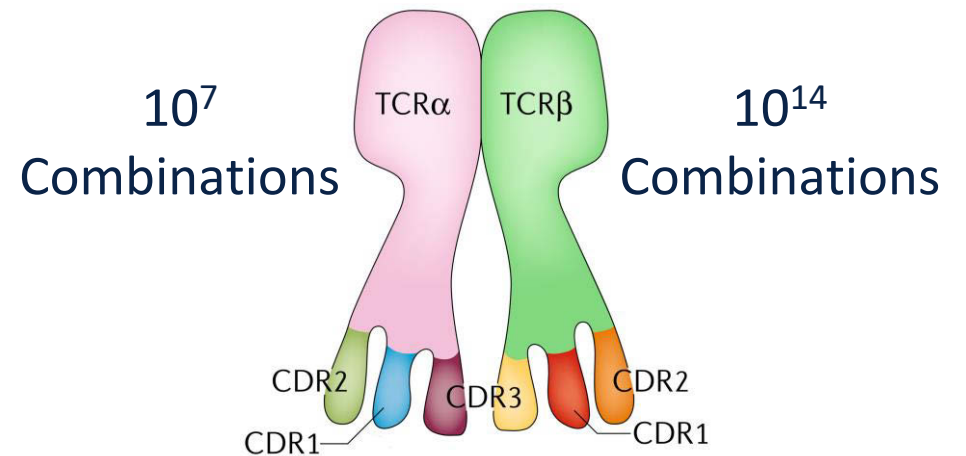
Generating Immune Repertoire Diversity (TCR/BCR)

V(D)J Recombination



10^{14} Combinations

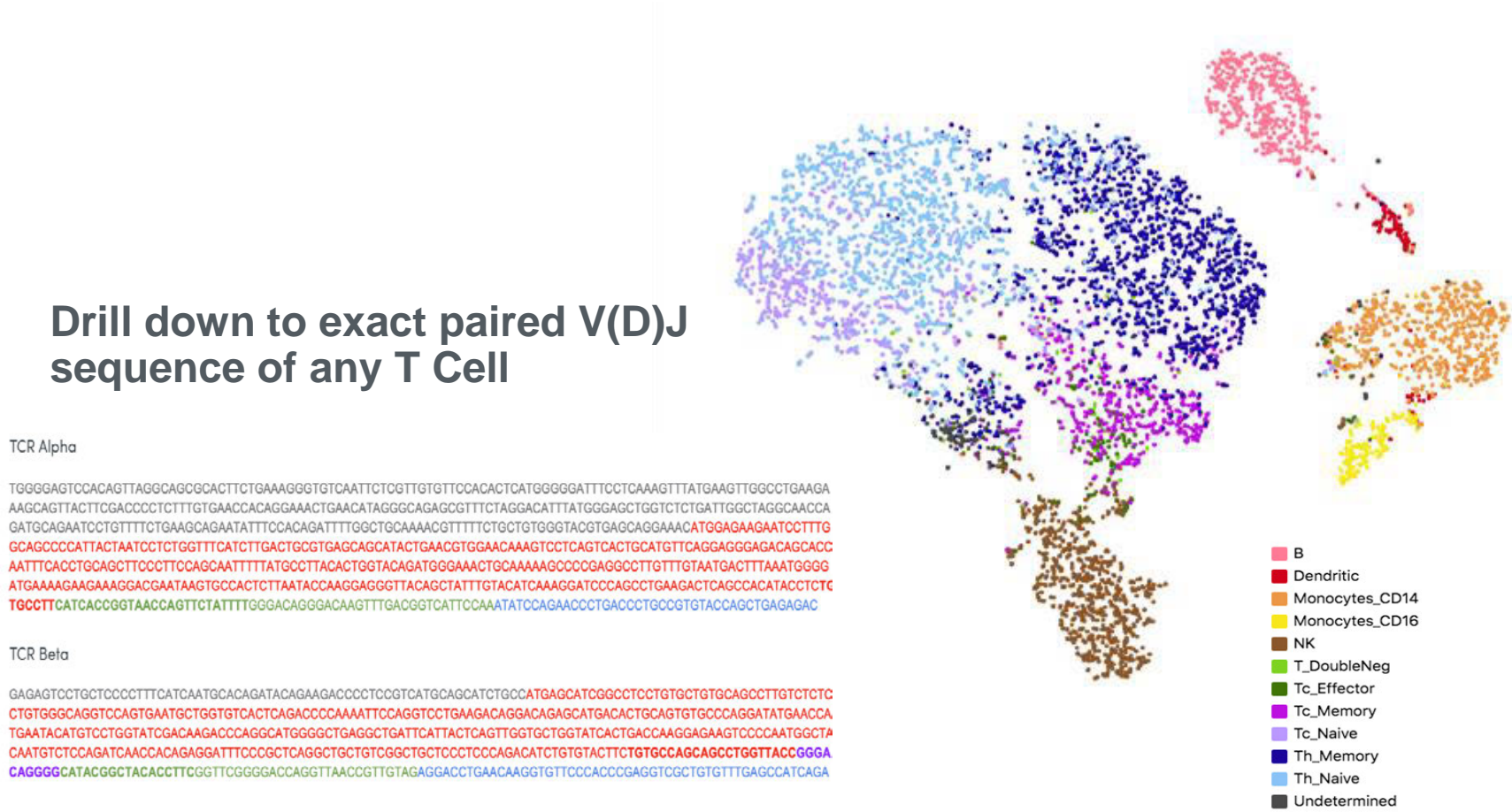
Paired rearranged TCRs



10^{21} Paired Possibilities

Paired V(D)J TCR sequences

Cells with TCR clonotypes identified



	V	D	J	C	#
α TRAV27			TRAJ36	TRAC]β
β TRBV12-3	TRBD1		TRBJ2-3	TRBC2	
α TRAV21			TRAJ17	TRAC]2
β TRBV6-3	TRBD1		TRBJ2-1	TRBC2	
α TRAV12-3			TRAJ24	TRAC]2
β TRBV15	TRBD1		TRBJ2-2	TRBC2	
α TRAV17			TRAJ29	TRAC]2
β TRBV14			TRBJ2-7	TRBC2	
α TRAV8-2			TRAJ23	TRAC]2
β TRBV12-3	TRBD1		TRBJ1-3	TRBC1	
α TRAV8-1			TRAJ27	TRAC]2
β TRBV12-3	TRBD2		TRBJ1-1	TRBC1	
α TRAV38-2...			TRAJ17	TRAC]2
β TRBV25-1			TRBJ1-1	TRBC1	
α TRAV8-2			TRAJ8	TRAC]2
β TRBV7-2	TRBD2		TRBJ2-7	TRBC2	
α TRAV38-2...			TRAJ53	TRAC]2
β TRBV6-6	TRBD2		TRBJ2-4	TRBC2	
α TRAV2			TRAJ8	TRAC]2
β TRBV4-3	TRBD1		TRBJ1-2	TRBC1	

Filter

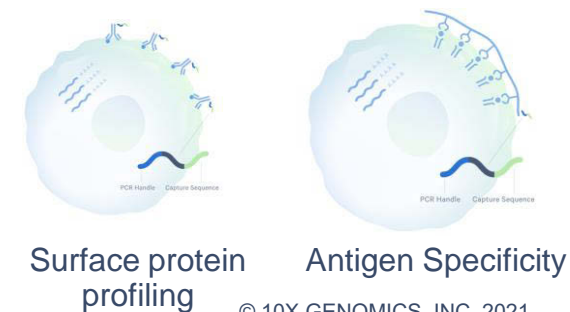
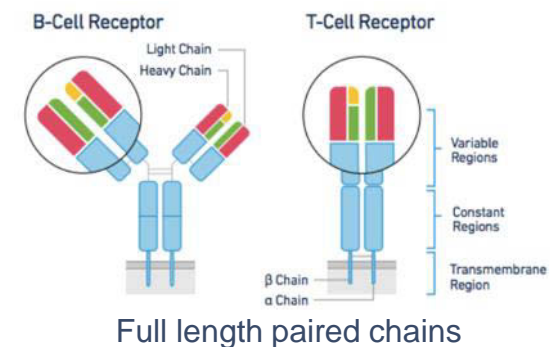
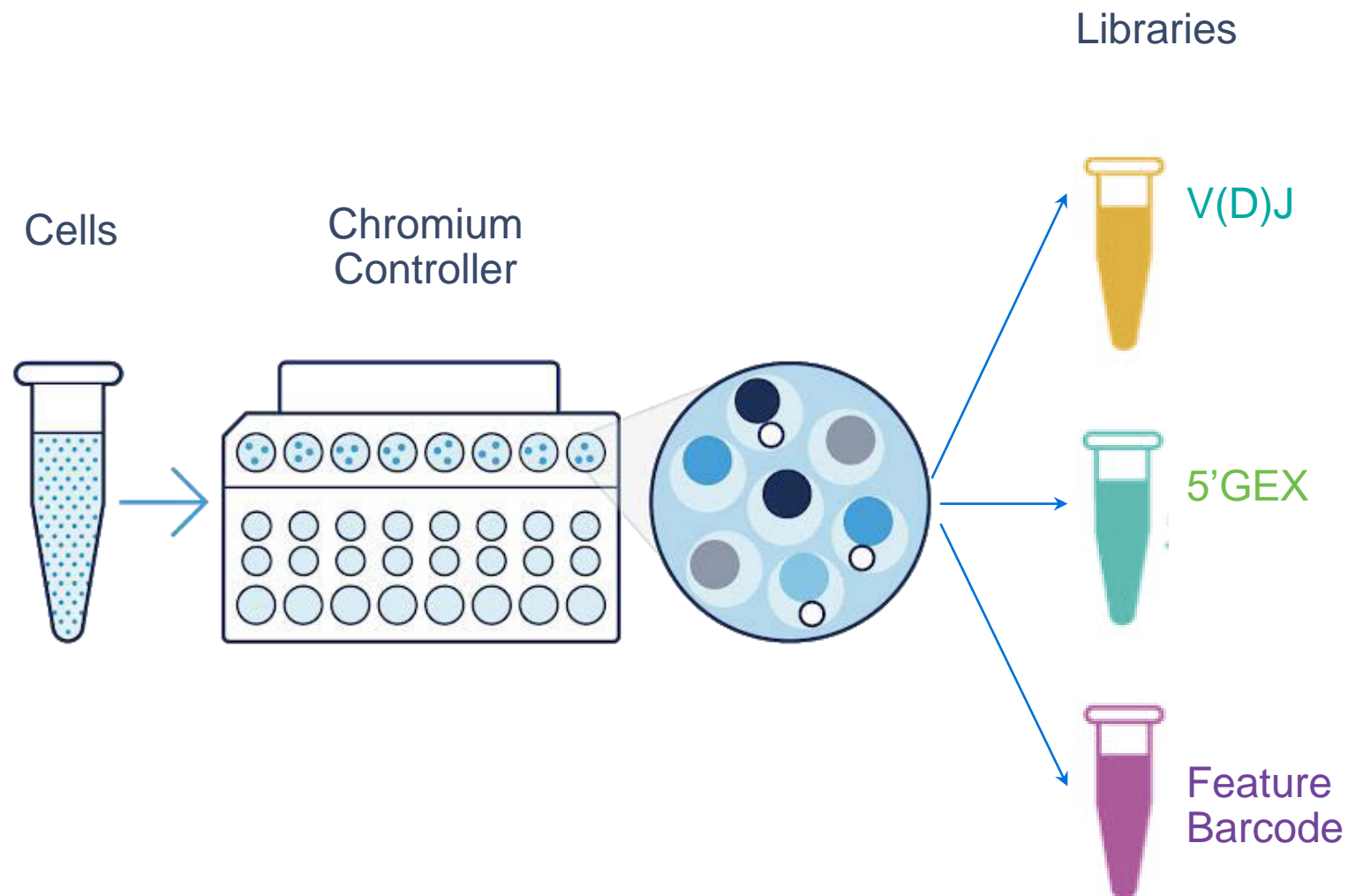
Cluster

Cluster

Identity: Th_Memory

✕

Single cell immune profiling library output



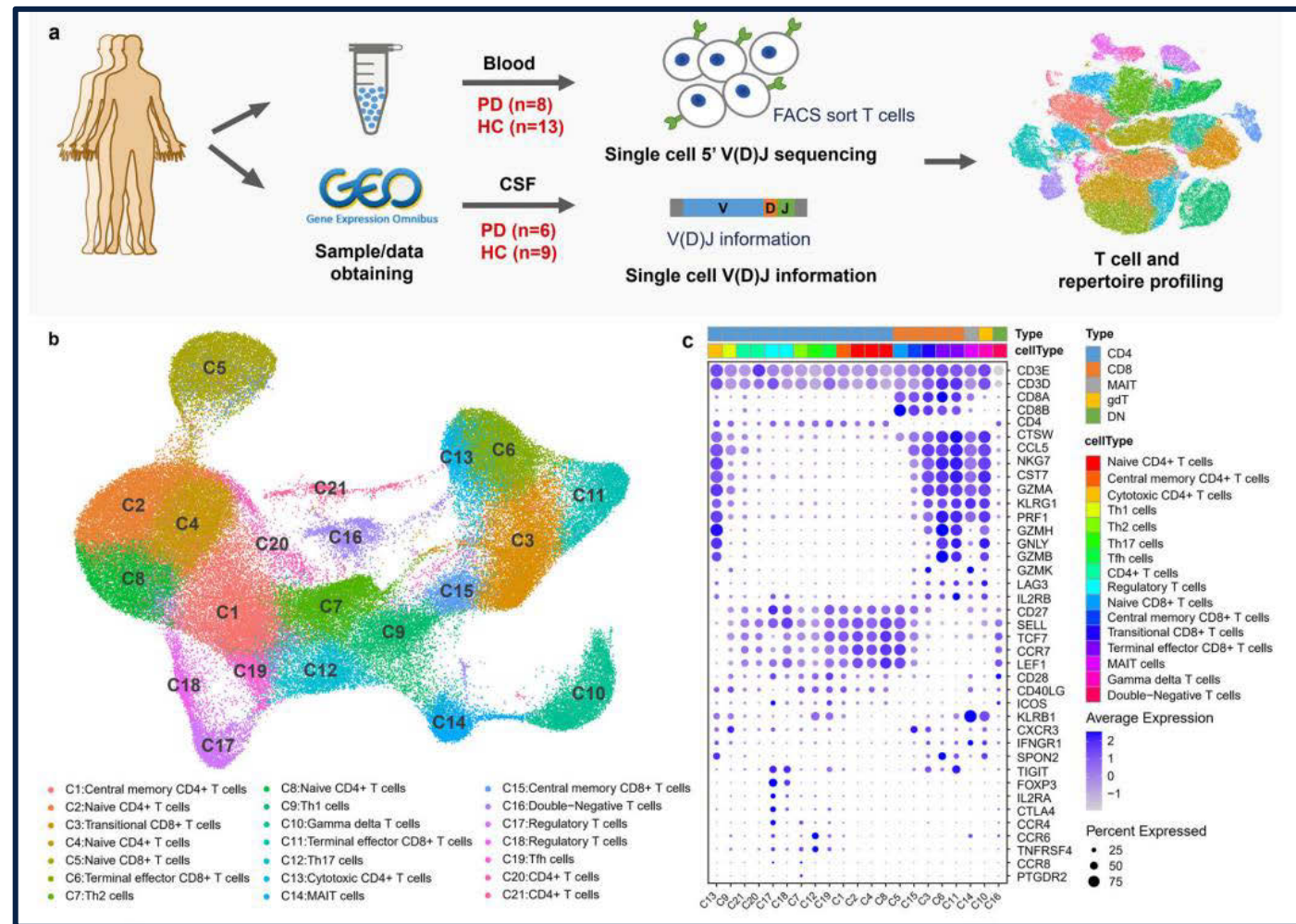
Single-cell transcriptome and TCR profiling reveal activated and expanded T cell populations in Parkinson's disease

Increasing studies suggest that immune system dysfunction plays important roles in the pathogenesis of PD.

Given the chronic inflammatory nature of Parkinson's disease (PD), T cell immunity may be important for disease onset.

It was performed single-cell transcriptome and TCR sequencing, and conducted integrative analyses to decode composition, function and lineage relationship of T cells in the **blood and cerebrospinal fluid of PD**.

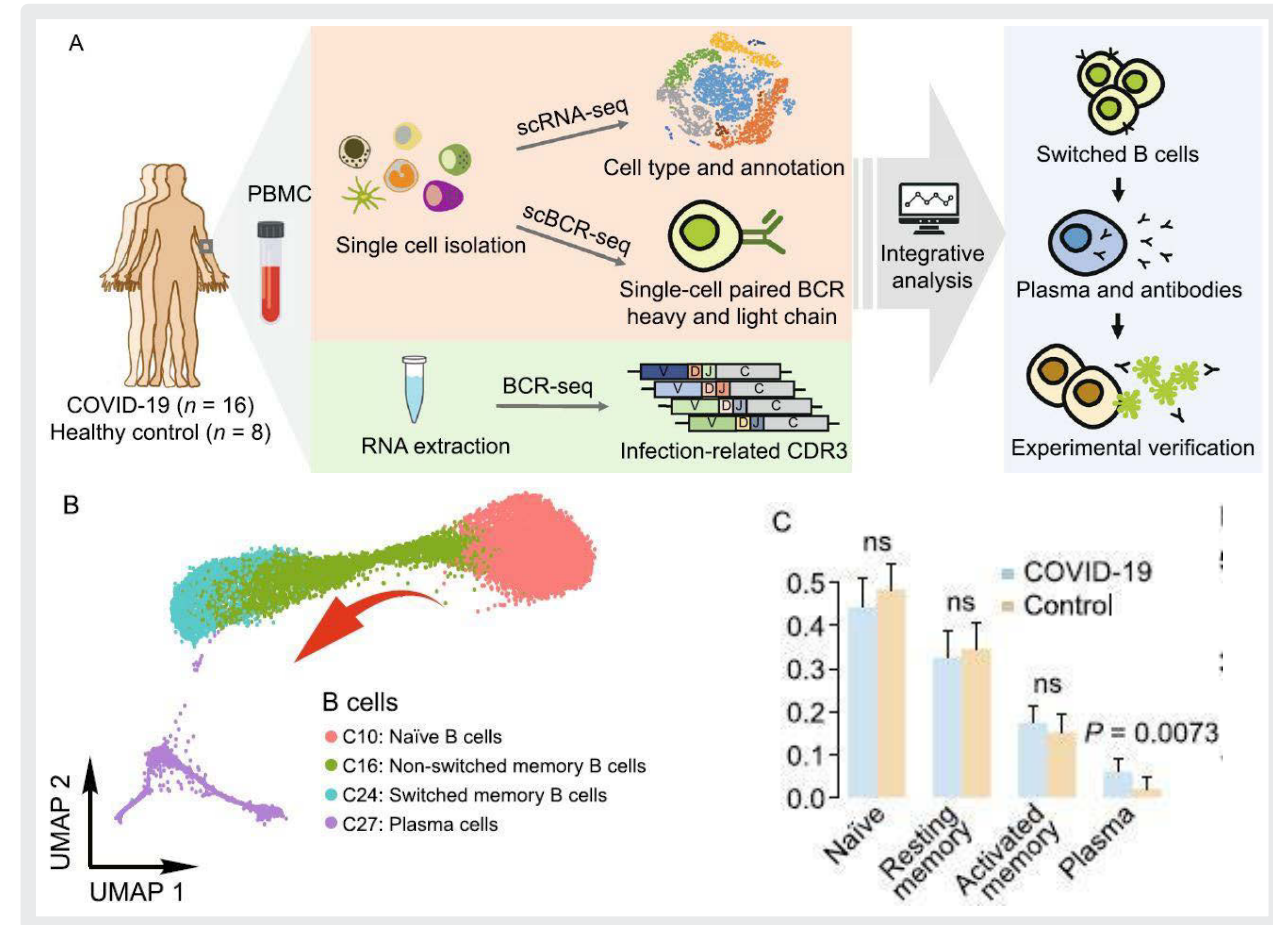
In total, **21 T cell subsets** with distinct functions were identified from 103,365 T cells. Integrative analyses of single-cell gene expression and TCRs revealed connectivity and potential differentiation trajectories of these subtypes and provided novel evidence of clonal expansion of T lymphocytes patrolling in the blood and cerebrospinal fluid of PD



Novel neutralizing antibody against SARS-CoV-2 revealed by single cell immune profiling

Li et al., 2020, Protein & Cell

- Chromium Single Cell Immune Profiling and BCR sequencing to **identify therapeutically relevant neutralizing antibody in early-stage recovered COVID-19 patients**
- Authors observed a **gradient of transcriptional states from naïve B cells to activated memory B cells then to plasma cells** (*fig. b*).
- COVID-19 patients showed lower BCR diversity, indicating **widespread clonal expansion upon likely antigen recognition**
- 347 BCR groups were selected for further study, of which 14 novel antibodies binding to S protein were identified (ELISA) and one (GD1-69) showed the highest neutralizing activity



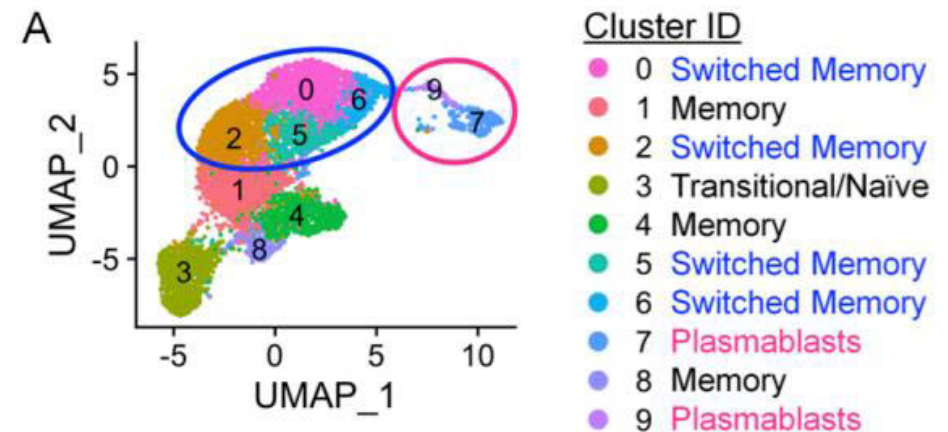
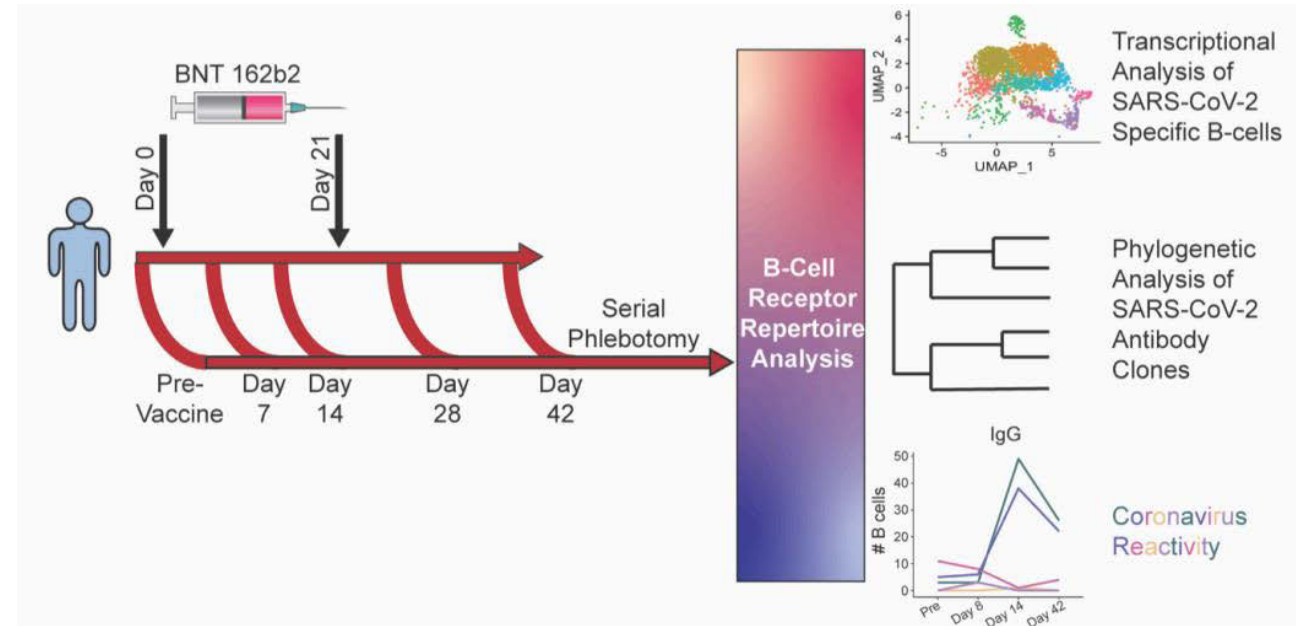
Single-Cell Profiling of the Antigen-Specific Response to BNT162b2 SARS-CoV-2 RNA Vaccine

RNA-based vaccines against SARS-CoV-2 are critical to limiting COVID-19 severity and spread.

Cellular mechanisms driving antigen-specific responses to these vaccines, however, remain uncertain.

B cells, T cells, and other leukocytes undergo significant shifts upon SARS-CoV-2 infection that may contribute to anti-viral immunity and protective antibodies

Here was used single-cell technologies to identify and characterized antigen-specific cells and antibody responses to the RNA vaccine BNT162b2 in longitudinal samples from a cohort of healthy donors.



Applying single cell sequencing to advance cell therapies

Leveraging the 10x toolkit across pre-clinical and translational clinical applications

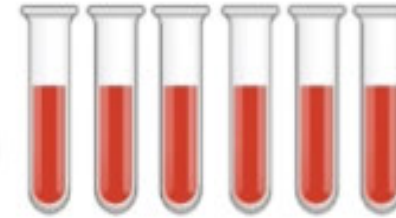
Target Discovery



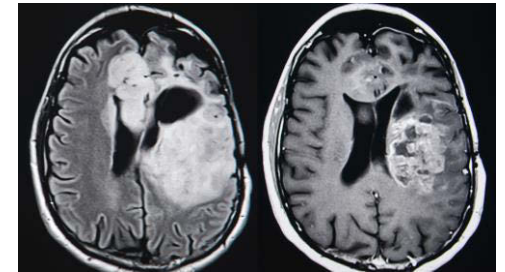
Characterization of Infusion Product



Response Monitoring



Resistance Profiling



Resource

Cell

Pooled Knockin Targeting for Genome Engineering of Cellular Immunotherapies

Theodore L. Roth,^{1,2,3,4,5,16*} P. Jonathan Li,^{2,4,5,16} Franziska Blaeschke,^{2,4,5,16} Jasper F. Nies,^{2,4,5,16} Ryan Apathy,^{2,4,5,16} Cody Mowery,^{2,3,4,5,16} Ruby Yu,^{2,4,5} Michelle L.T. Nguyen,^{2,4,5} Youjin Lee,^{2,4,5} Anna Truong,^{2,4,5} Joseph Hiatt,^{2,4,5,16} David Wu,^{1,2} David N. Nguyen,^{2,4,5,16} Daniel Goodman,^{2,4,5} Jeffrey A. Bluestone,^{2,4,5} Chun Jimmie Ye,^{2,4,5,16,17,18} Kole Roybal,^{2,3,4,14} Eric Shifrut,^{2,4,5} and Alexander Marson^{2,4,5,6,7,8,14,16,17,*}

nature
medicine

ARTICLES

Characteristics of anti-CD19 CAR T cell infusion products associated with efficacy and toxicity in patients with large B cell lymphomas

Qing Deng^{1,5}, Guangchun Han^{2,3}, Nahum Puebla-Osorio¹, Man Chun John Ma¹, Paolo Strati¹, Beth Chasen¹, Enyu Dai¹, Minghao Dang¹, Neeraj Jain¹, Haoping Yang¹, Yuanxin Wang¹, Shaojun Zhang¹, Ruiping Wang¹, Runzhe Chen¹, Jordan Showell¹, Sreejaye Ghosh¹, Sridevi Patchva¹, Qi Zhang¹, Ryan Sun¹, Frederick Hagemester¹, Luis Fayad¹, Felipe Samaniego¹, Hans C. Lee¹, Loretta J. Nastoupil¹, Nathan Fowler¹, R. Eric Davis¹, Jason Westin¹, Sattva S. Neelapu^{1,15}, Linghua Wang^{1,16} and Michael R. Green^{1,16}

nature
COMMUNICATIONS

ARTICLE

Clonal kinetics and single-cell transcriptional profiling of CAR-T cells in patients undergoing CD19 CAR-T immunotherapy

Alyssa Sheh^{1,5}, Valentin Voillet^{2,5}, Laila-Alcha Hanafi^{1,5}, Hannah A. DeBerg³, Masanao Yajima⁴, Reed Hawkins¹, Vivian Gersuk¹, Stanley R. Riddell^{1,5,6}, David G. Maloney^{1,5,6}, Martin E. Wohlschlag¹, Dnyanada Pande¹, Mark R. Enstrom¹, Hans-Peter Kiem^{1,5,7}, Jennifer E. Adair^{1,5,6}, Raphael Gottardo^{2,5,8}, Peter S. Linsley³ and Cameron J. Turtle^{1,5,14}

Article

c-Jun overexpression in CAR T cells induces exhaustion resistance

Rachel C. Lynn^{1,2}, Evan W. Weber^{1,2}, Elena Scillito^{1,2}, David Gennert¹, Peng Xu¹, Zinaida Good^{1,2}, Hima Anbumathan¹, John Lattin¹, Robert Jones¹, Victor Tieu¹, Surya Nagaraja¹, Jeffrey Grange¹, Charles F. A. de Bourcy^{1,2}, Robbie Majzner¹, Ansuman T. Satpathy¹, Stephen R. Quake^{1,2}, Michelle Monje^{1,2}, Howard Y. Chang^{1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,*} and Crystal L. Mackall^{1,2,10,11,12,13,14,15}

Cell

CellPress

Single-Cell Analyses Identify Brain Mural Cells Expressing CD19 as Potential Off-Tumor Targets for CAR-T Immunotherapies

Kevin R. Parker^{1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38,39,40,41,42,43,44,45,46,47,48,49,50,51,52,53,54,55,56,57,58,59,60,61,62,63,64,65,66,67,68,69,70,71,72,73,74,75,76,77,78,79,80,81,82,83,84,85,86,87,88,89,90,91,92,93,94,95,96,97,98,99,100,101,102,103,104,105,106,107,108,109,110,111,112,113,114,115,116,117,118,119,120,121,122,123,124,125,126,127,128,129,130,131,132,133,134,135,136,137,138,139,140,141,142,143,144,145,146,147,148,149,150,151,152,153,154,155,156,157,158,159,160,161,162,163,164,165,166,167,168,169,170,171,172,173,174,175,176,177,178,179,180,181,182,183,184,185,186,187,188,189,190,191,192,193,194,195,196,197,198,199,200,201,202,203,204,205,206,207,208,209,210,211,212,213,214,215,216,217,218,219,220,221,222,223,224,225,226,227,228,229,230,231,232,233,234,235,236,237,238,239,240,241,242,243,244,245,246,247,248,249,250,251,252,253,254,255,256,257,258,259,260,261,262,263,264,265,266,267,268,269,270,271,272,273,274,275,276,277,278,279,280,281,282,283,284,285,286,287,288,289,290,291,292,293,294,295,296,297,298,299,300,301,302,303,304,305,306,307,308,309,310,311,312,313,314,315,316,317,318,319,320,321,322,323,324,325,326,327,328,329,330,331,332,333,334,335,336,337,338,339,340,341,342,343,344,345,346,347,348,349,350,351,352,353,354,355,356,357,358,359,360,361,362,363,364,365,366,367,368,369,370,371,372,373,374,375,376,377,378,379,380,381,382,383,384,385,386,387,388,389,390,391,392,393,394,395,396,397,398,399,400,401,402,403,404,405,406,407,408,409,410,411,412,413,414,415,416,417,418,419,420,421,422,423,424,425,426,427,428,429,430,431,432,433,434,435,436,437,438,439,440,441,442,443,444,445,446,447,448,449,450,451,452,453,454,455,456,457,458,459,460,461,462,463,464,465,466,467,468,469,470,471,472,473,474,475,476,477,478,479,480,481,482,483,484,485,486,487,488,489,490,491,492,493,494,495,496,497,498,499,500,501,502,503,504,505,506,507,508,509,510,511,512,513,514,515,516,517,518,519,520,521,522,523,524,525,526,527,528,529,530,531,532,533,534,535,536,537,538,539,540,541,542,543,544,545,546,547,548,549,550,551,552,553,554,555,556,557,558,559,560,561,562,563,564,565,566,567,568,569,570,571,572,573,574,575,576,577,578,579,580,581,582,583,584,585,586,587,588,589,590,591,592,593,594,595,596,597,598,599,600,601,602,603,604,605,606,607,608,609,610,611,612,613,614,615,616,617,618,619,620,621,622,623,624,625,626,627,628,629,630,631,632,633,634,635,636,637,638,639,640,641,642,643,644,645,646,647,648,649,650,651,652,653,654,655,656,657,658,659,660,661,662,663,664,665,666,667,668,669,670,671,672,673,674,675,676,677,678,679,680,681,682,683,684,685,686,687,688,689,690,691,692,693,694,695,696,697,698,699,700,701,702,703,704,705,706,707,708,709,710,711,712,713,714,715,716,717,718,719,720,721,722,723,724,725,726,727,728,729,730,731,732,733,734,735,736,737,738,739,740,741,742,743,744,745,746,747,748,749,750,751,752,753,754,755,756,757,758,759,760,761,762,763,764,765,766,767,768,769,770,771,772,773,774,775,776,777,778,779,780,781,782,783,784,785,786,787,788,789,790,791,792,793,794,795,796,797,798,799,800,801,802,803,804,805,806,807,808,809,810,811,812,813,814,815,816,817,818,819,820,821,822,823,824,825,826,827,828,829,830,831,832,833,834,835,836,837,838,839,840,841,842,843,844,845,846,847,848,849,850,851,852,853,854,855,856,857,858,859,860,861,862,863,864,865,866,867,868,869,870,871,872,873,874,875,876,877,878,879,880,881,882,883,884,885,886,887,888,889,890,891,892,893,894,895,896,897,898,899,900,901,902,903,904,905,906,907,908,909,910,911,912,913,914,915,916,917,918,919,920,921,922,923,924,925,926,927,928,929,930,931,932,933,934,935,936,937,938,939,940,941,942,943,944,945,946,947,948,949,950,951,952,953,954,955,956,957,958,959,960,961,962,963,964,965,966,967,968,969,970,971,972,973,974,975,976,977,978,979,980,981,982,983,984,985,986,987,988,989,990,991,992,993,994,995,996,997,998,999,1000}

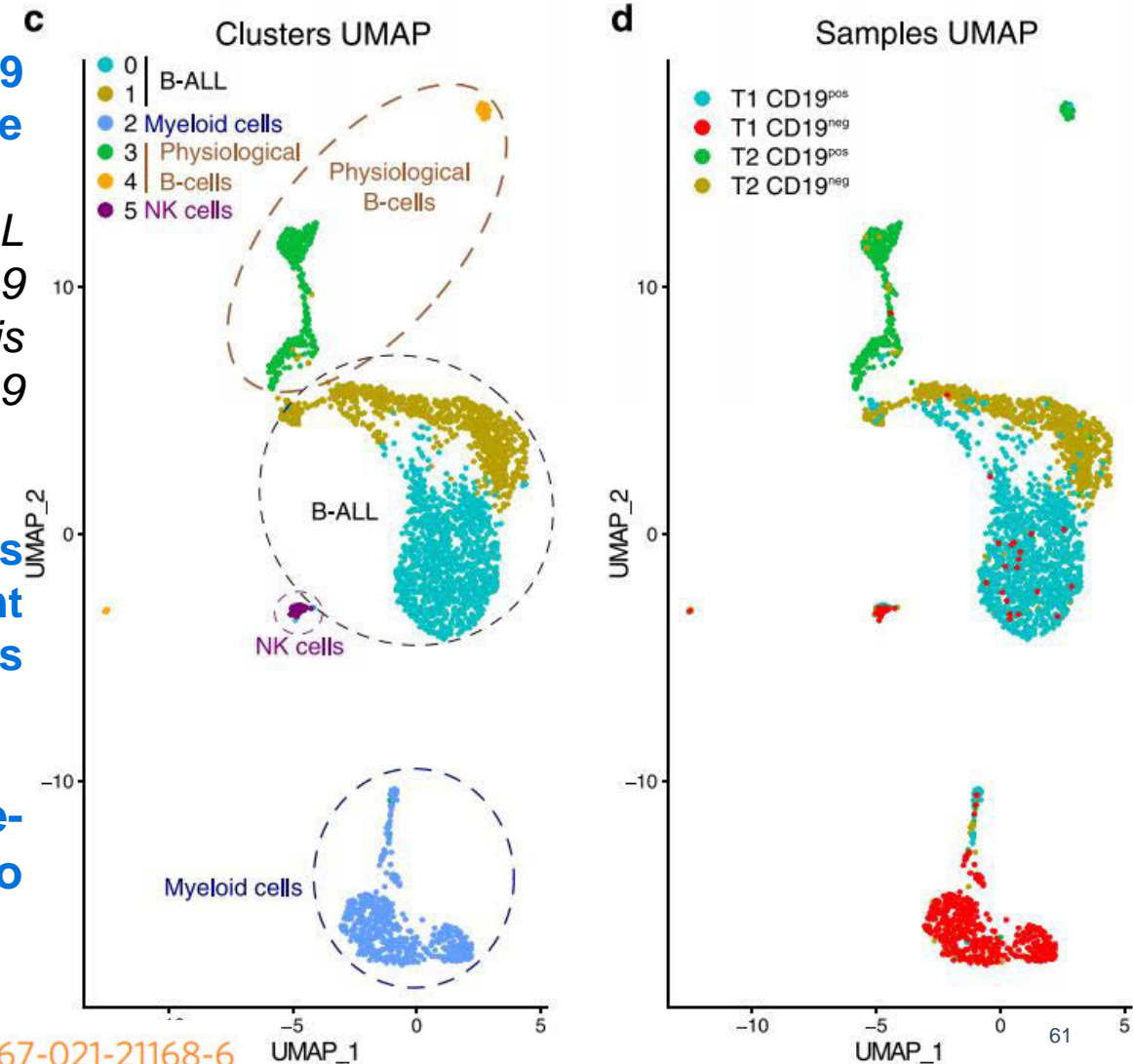
Single-cell profiling identifies pre-existing CD19-negative subclones in a B-ALL patient with CD19-negative relapse after CAR-T therapy

Around half of relapsing CD19 CAR-T patients develop CD19 negative (CD19neg) B-ALL allowing leukemic cells to evade CD19-targeted therapy.

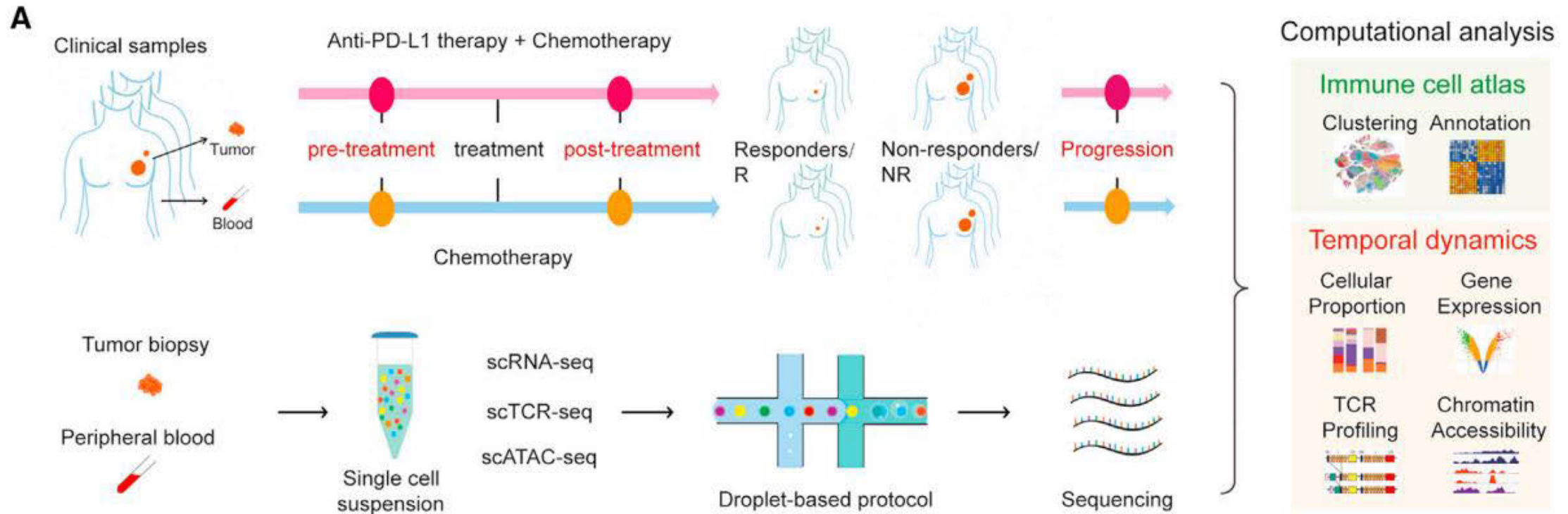
Herin, were investigated leukemic cells of a relapsing B-ALL patient, at two-time points: before (T1) and after (T2) anti-CD19 CAR-T treatment. It was shown that at T2, the B-ALL relapse is CD19 negative due to the expression of a non-functional CD19 transcript retaining intron 2.

Using single-cell RNA sequencing (scRNAseq) approach, was demonstrated that CD19neg leukemic cells were present before CAR-T cell therapy and thus that the relapse results from the selection of these rare CD19neg B-ALL clones.

Study shows that scRNAseq profiling can reveal pre-existing CD19neg subclones, raising the possibility to assess the risk of targeted therapy failure.

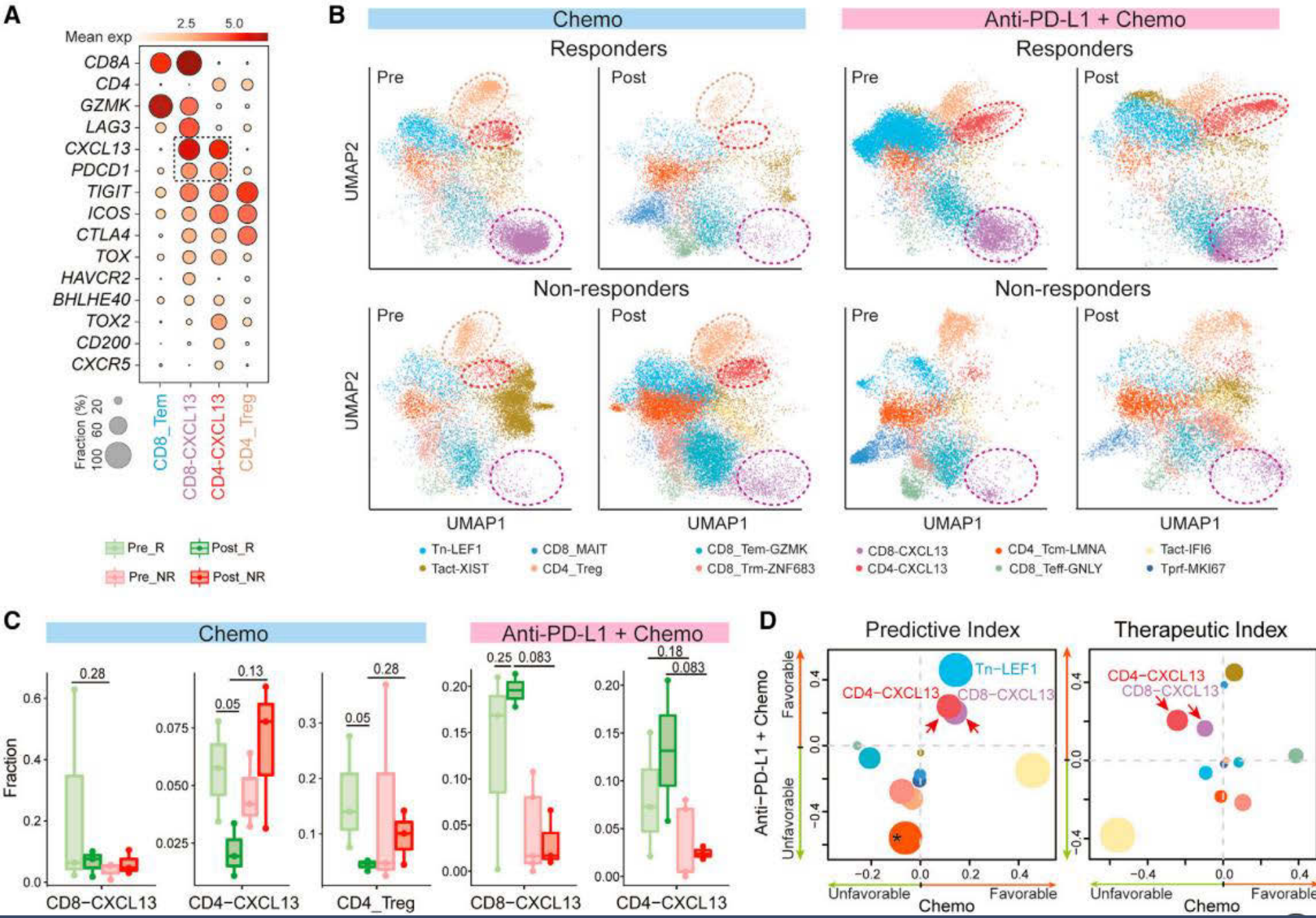


Single-cell analyses reveal key immune cell subsets associated with response to PD-L1 blockade in triple-negative breast cancer



Zhang et al. combine **single-cell RNAseq**, **TCR-seq**, and **ATAC-seq** to investigate immune cell dynamics in the tumor microenvironment and peripheral blood of patients with TNBC treated with paclitaxel or paclitaxel plus atezolizumab, revealing immune features of responders and nonresponders, the mechanisms and intertwined effects of paclitaxel and atezolizumab in TNBC treatment.

Temporal dynamics of tumor-infiltrating T cell subsets

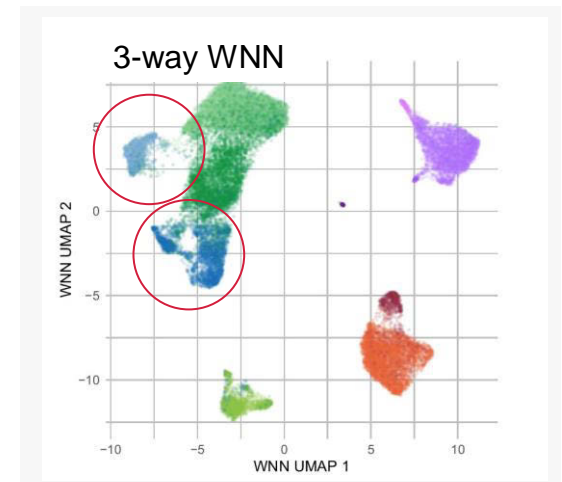
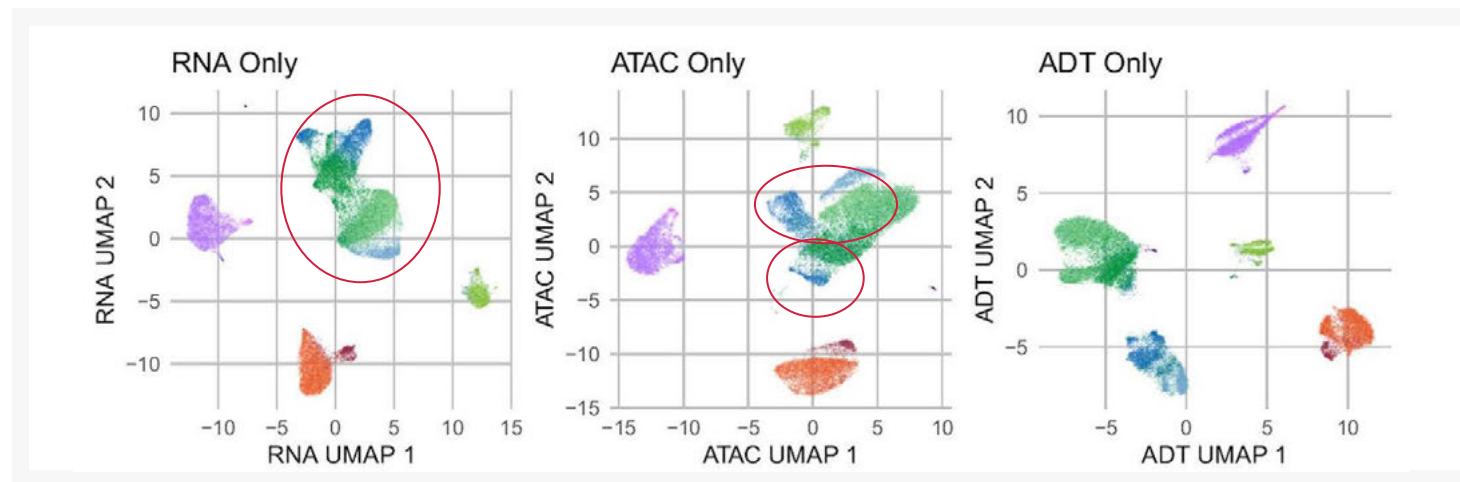
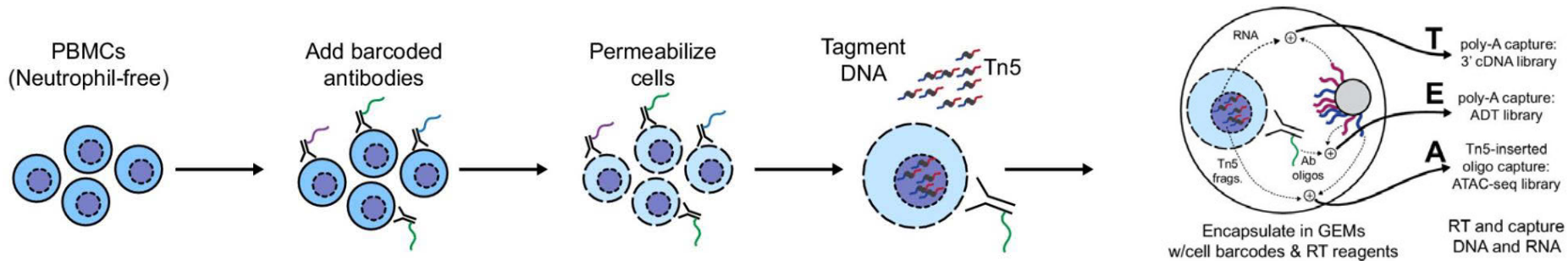


<https://doi.org/10.1016/j.ccell.2021.09.010>

What's more?

Measuring additional modalities increases power to separate cell states

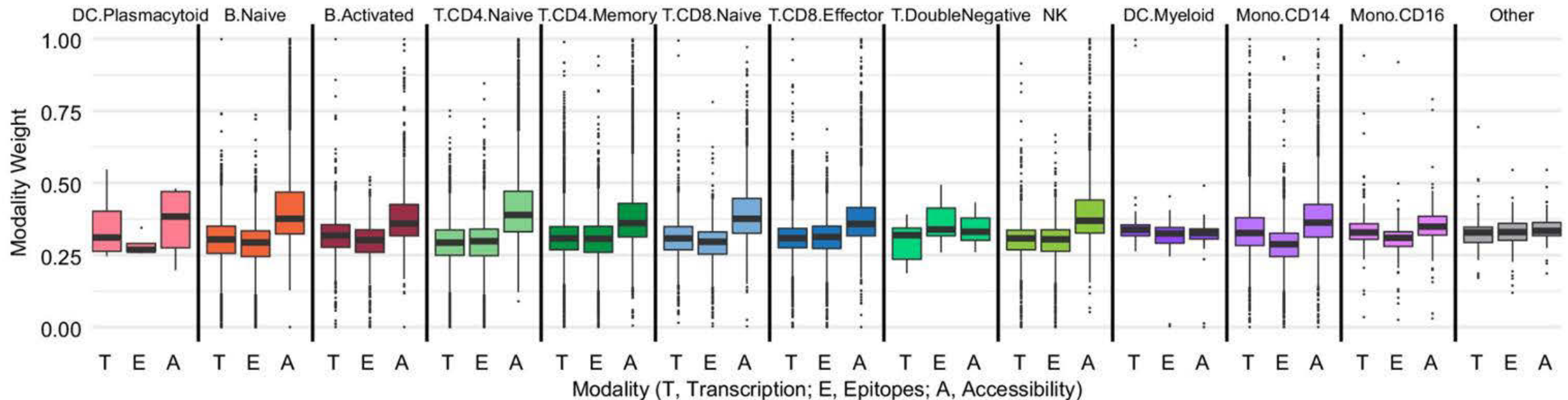
TEA-seq: Simultaneous detection of transcriptome, epitopes, and epigenome



Improved resolution of CD4/CD8 T cell states when using RNA, ATAC, and protein data

For many cell types, scATAC-seq largest contributor to improved cell state resolution

Using MULTIOME 10x



Simultaneous trimodal single-cell measurement of transcripts, epitopes, and chromatin accessibility using TEA-seq


Learn more about TEA-seq from the authors

“Deep immune profiling at scale: Efficient pipelines and TEA-seq for simultaneous trimodal measurements”

Symposium Series

Next Generation Multiomics Symposium

Reveal the full spectrum of biological complexity



Peter Skene
Allen Institute for Immunology
USA



Lucas Graybuck, PhD
Allen Institute for Immunology
USA

Watch this webinar on-demand:



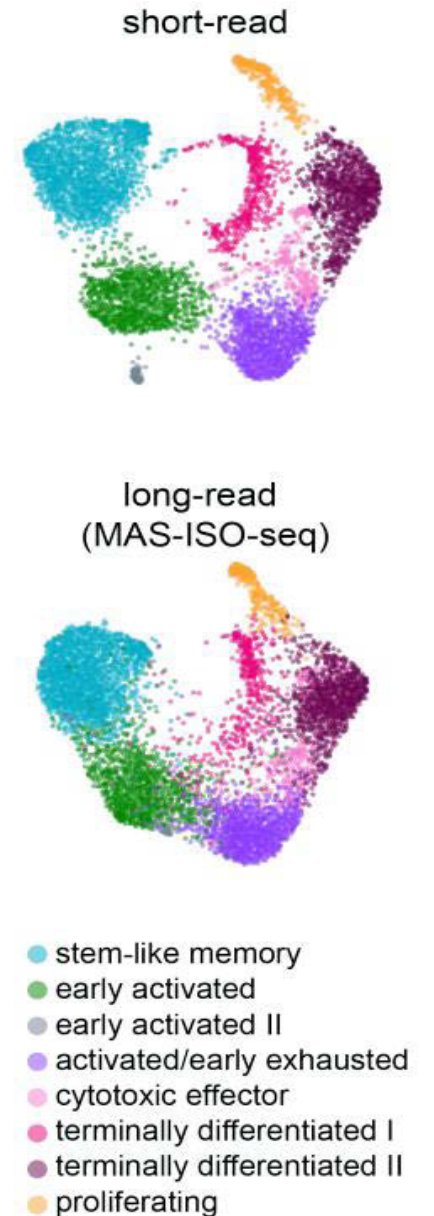
<https://pages.10xgenomics.com/2021-multiomics-week.html>

High-throughput RNA isoform sequencing using programmable cDNA concatenation

Alternative splicing is a core biological process that enables profound and essential diversification of gene function. Short-read RNA sequencing approaches fail to resolve RNA isoforms and therefore primarily enable gene expression measurements - an isoform unaware representation of the transcriptome.

Conversely, full-length RNA sequencing using long-read technologies are able to capture complete transcript isoforms, but their utility is deeply constrained due to throughput limitations. Here it's introduced **MAS-ISO-seq**.

This approach was used for single-cell RNA sequencing analysis of tumor-infiltrating T cells (based on 5'GEX and CSP analysis). PacBio sequencing was used.





Coming soon

ATAC v2

- Measure chromatin accessibility at single cell resolution
- Improved signal to noise ratio
- Reduced sequencing costs

Expected **Mid 2022**



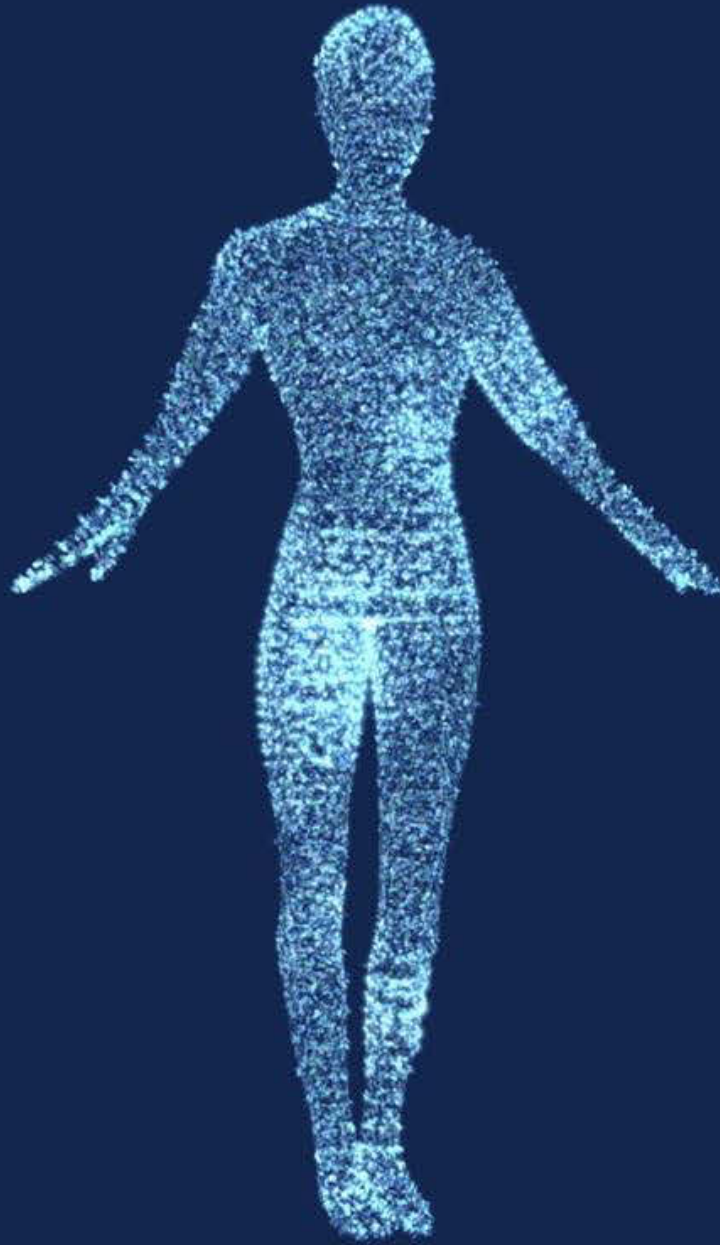
5' CRISPR

- Measure perturbation effects with multiomic readouts
- Increased flexibility for functional genomics studies
- Rapidly deploy existing Cas9 RNA libraries

Expected **Early 2022**

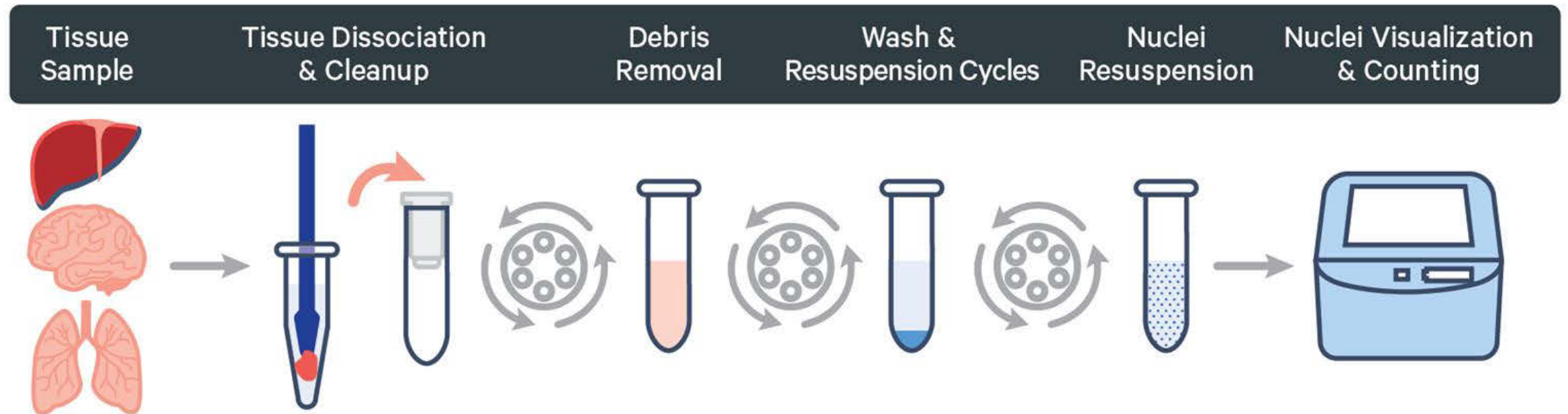


Nuclei Isolation



Nuclei Isolation Kit

Streamlined sample preparation workflow



All you need is an hour of lab time, a benchtop centrifuge, and an interesting frozen sample!

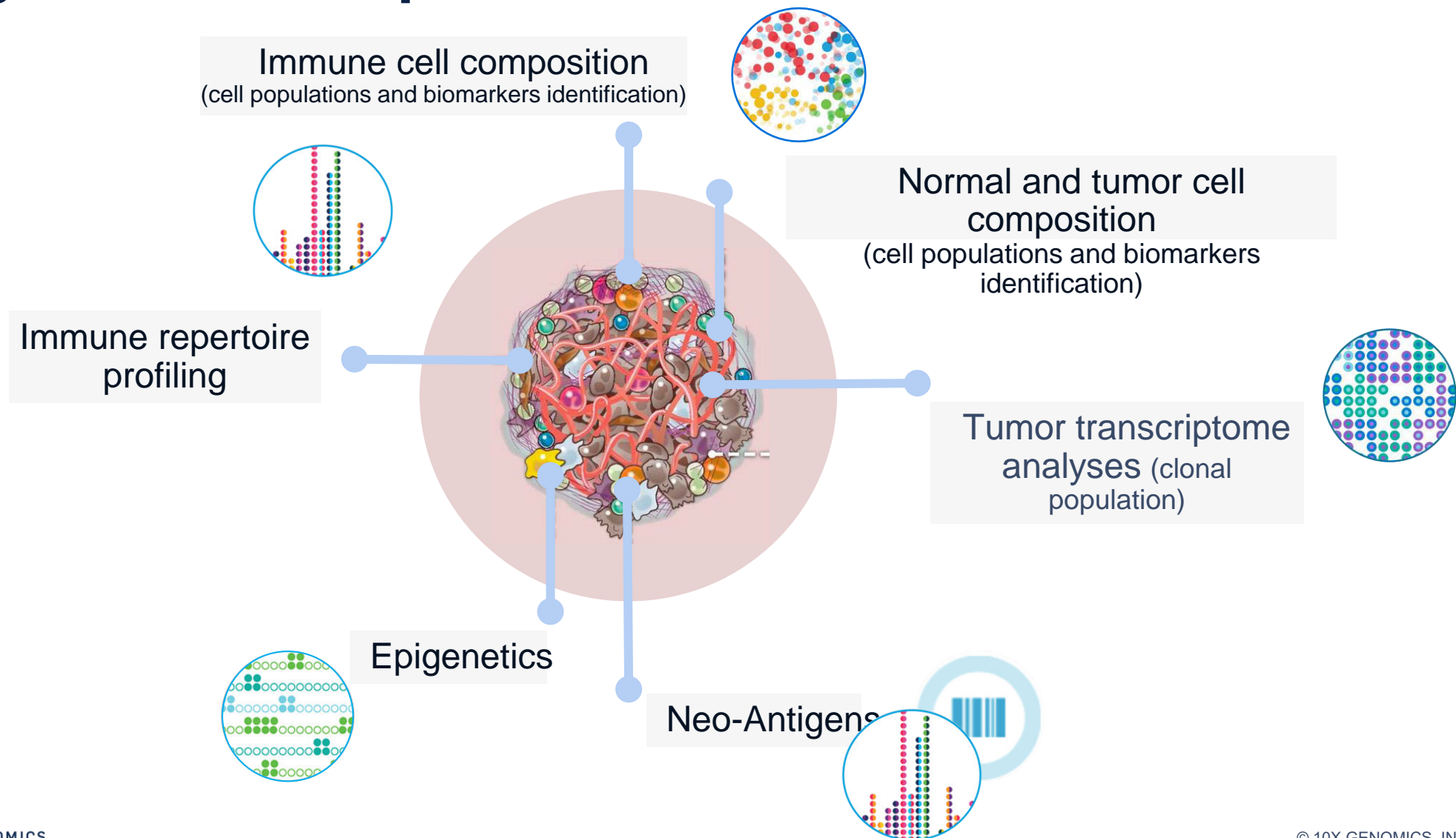
Become one of single cell Ninjas 😊



The Single-Cell *Ninjas*

Summary

Integration of sample Information with 10x Genomics



Thank You from the 10x Team & our Collaborators

Agnieszka Ciesielska

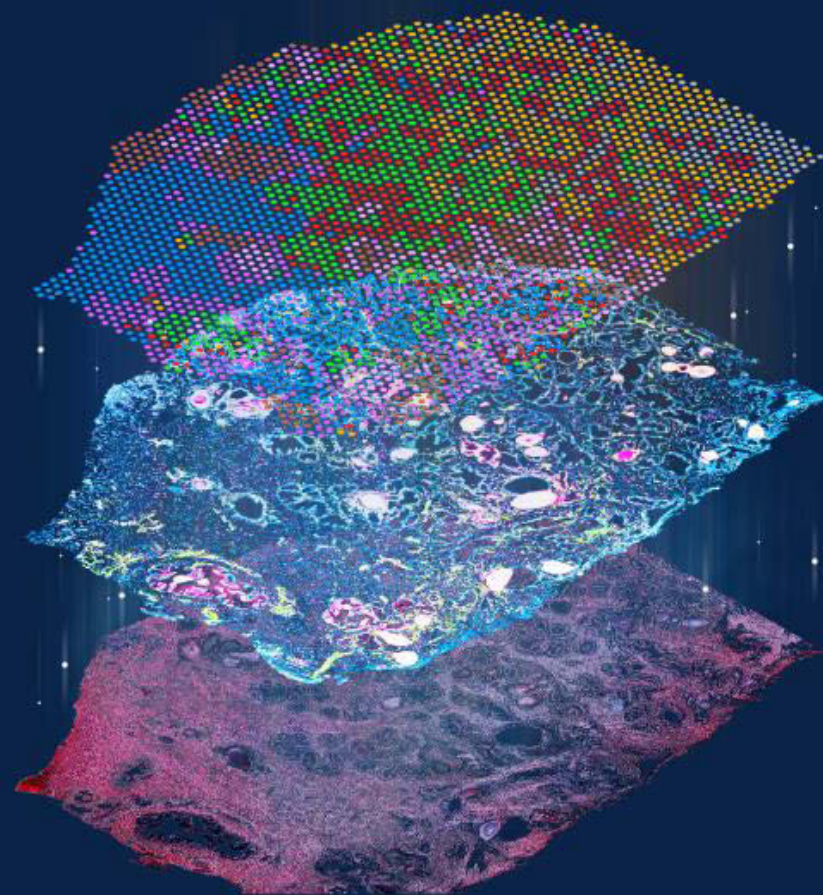
Science & Technology Advisor



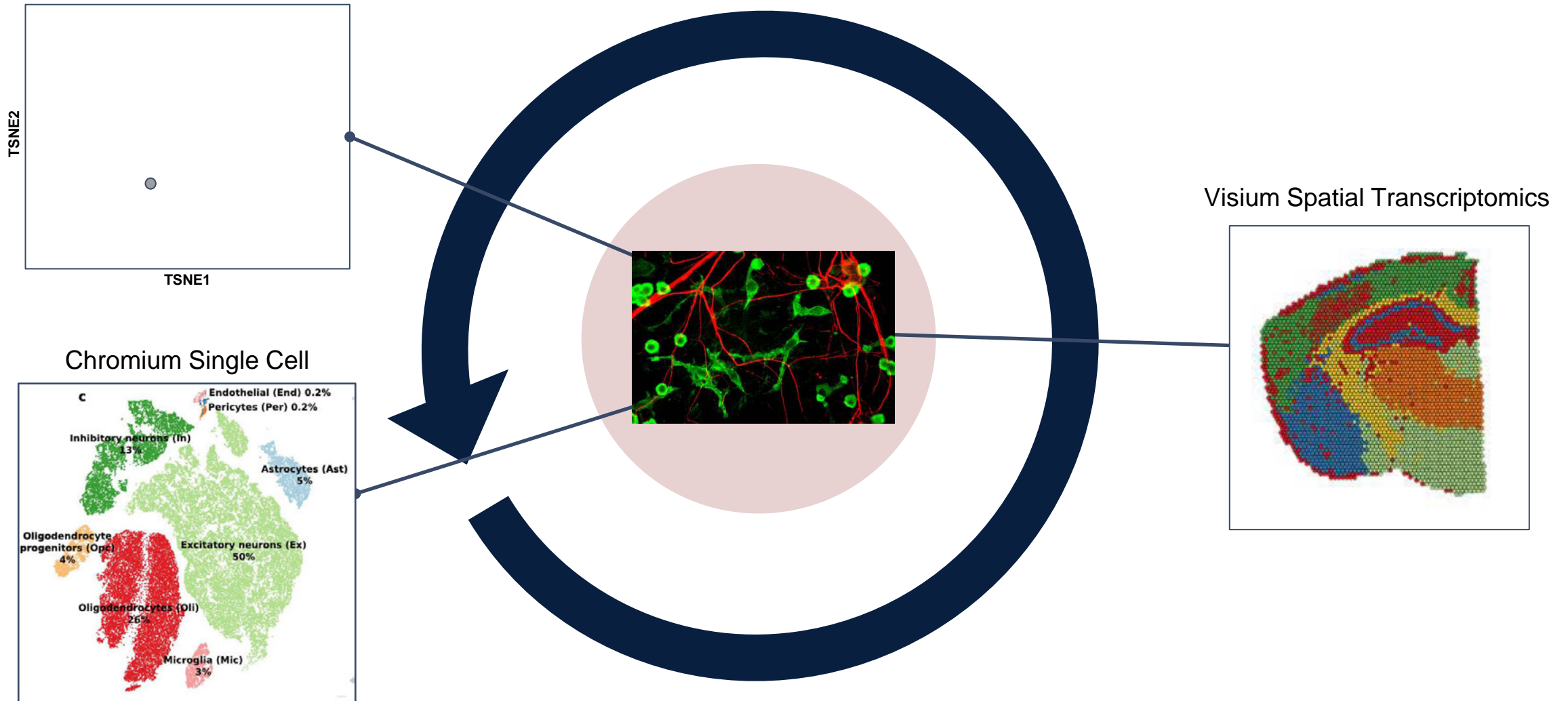
Enter the next level of complexity – spatial transcriptomics approaches

Agnieszka Ciesielska PhD

Science and Technology Advisor 10x Genomics,
CEE & Israel & Russia, Distributors

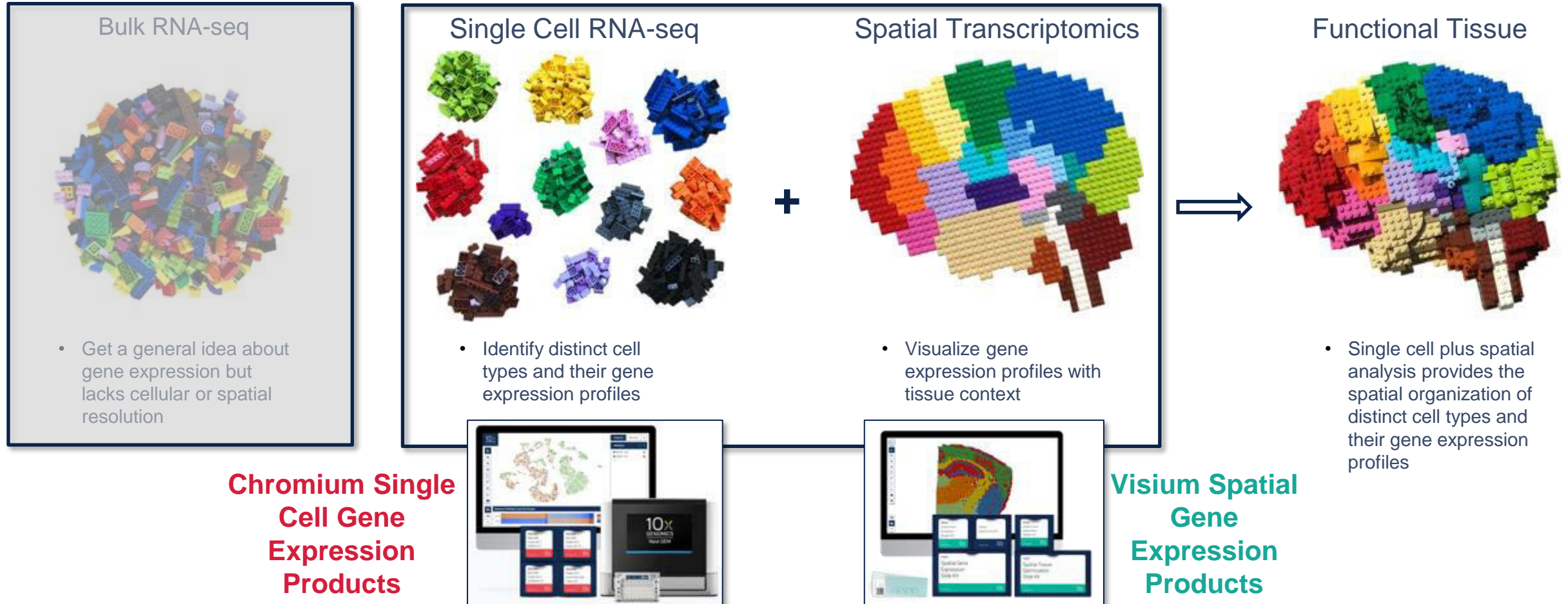


Building your understanding of biological system – slice by slice



Gain a Complete View of Biology with Single Cell and Spatial Analysis

- Complementarity of single cell and spatial methods from 10x Genomics



Spatial

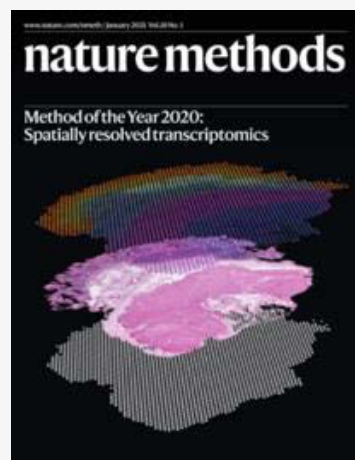


Entering digital histology era. TOP 10 Innovations

TheScientist
TOP 10
INNOVATIONS

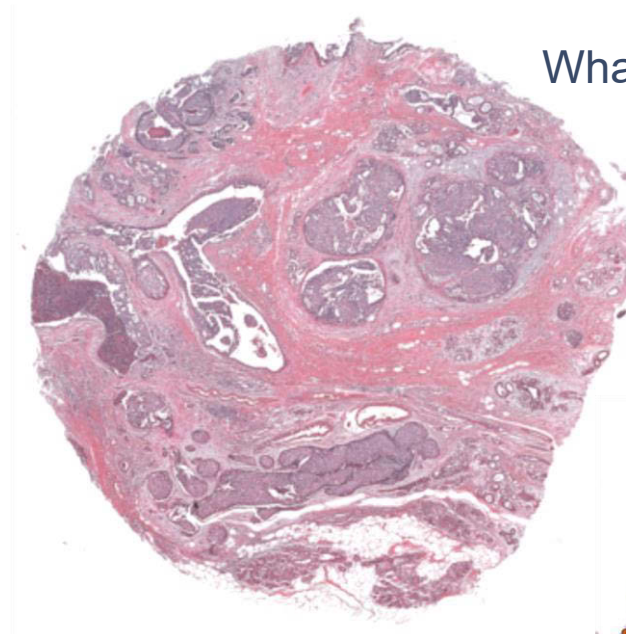
2020

Visium Spatial Gene Expression

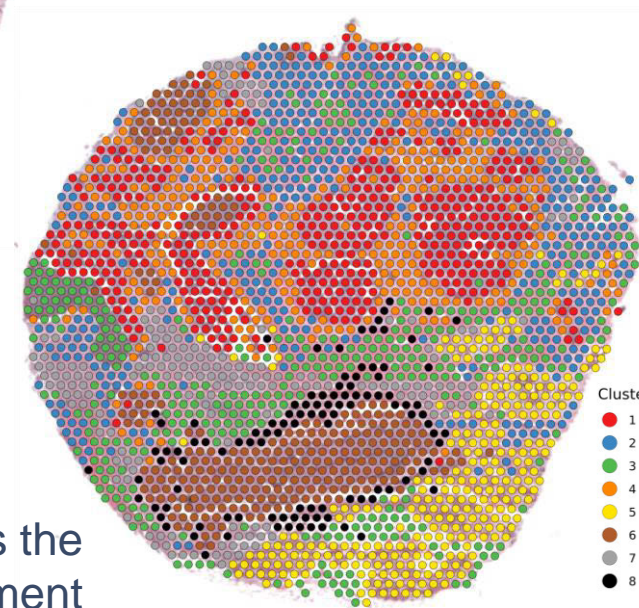


2020

Spatially Resolved Transcriptomics



What's hiding in this tissue?



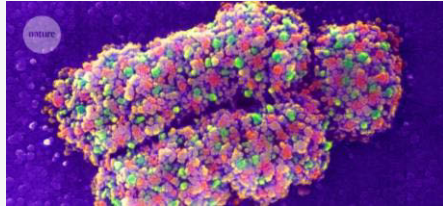
Visium reveals the
complex environment

Advancing spatial biology

nature

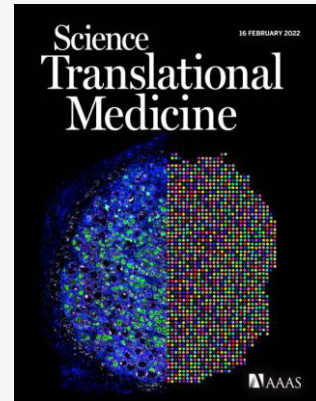
TECHNOLOGY FEATURE | 25 January 2022

Seven technologies to watch in 2022



2022

Spatial Multiomics



2022

1st Visium Cover



2020

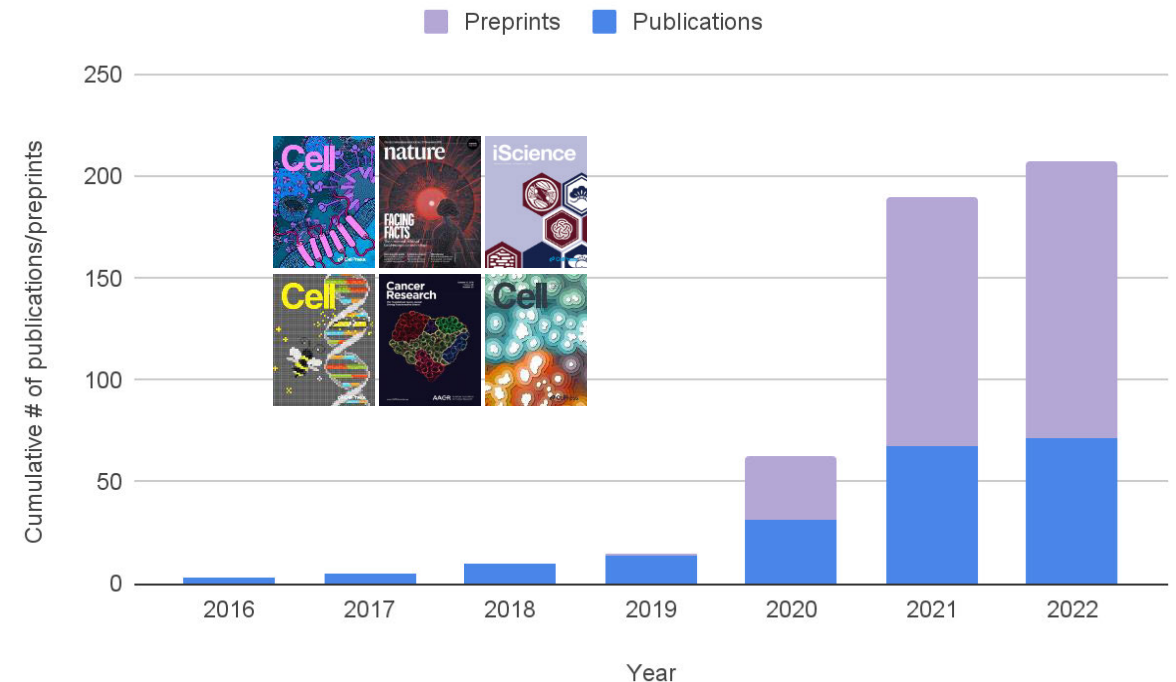
Spatially Resolved
Transcriptomics

TheScientist
TOP 10
INNOVATIONS

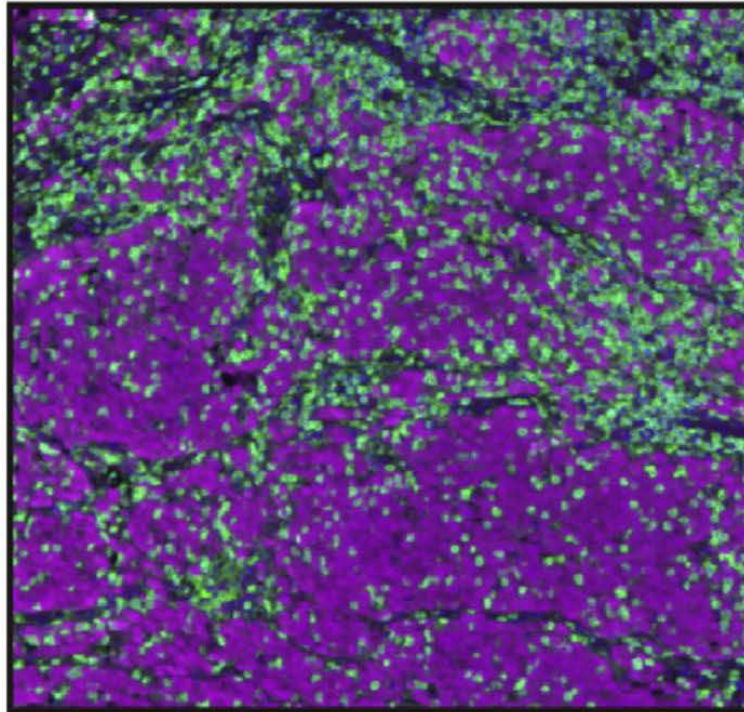
2020

Visium Spatial Gene
Expression

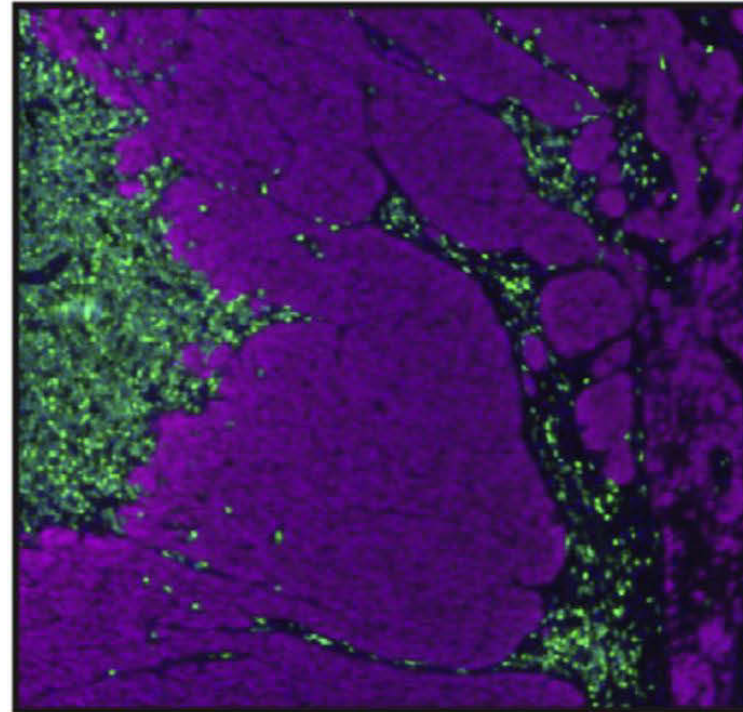
200+ Visium Publications and Preprints



Why Spatial Analysis? Location Matters!



“Hot tumor”
Lymphocytes infiltrating tumor



“Cold tumor”
Lymphocytes stopped at tumor boundary. “non-inflamed”, „non-immunogenic” tumors

Van der Woude *et al.*, 2017

Tumor cell
Immune cell

The level of infiltration of tumors by lymphocytes can be a prognostic factor.

Visium Spatial Gene Expression



Whole Transcriptome Analysis in FFPE and Fresh Frozen Tissues

Spatially resolve mRNA across FFPE and fresh frozen tissue sections

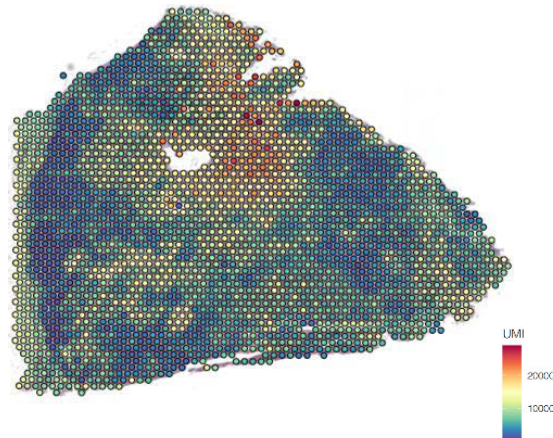
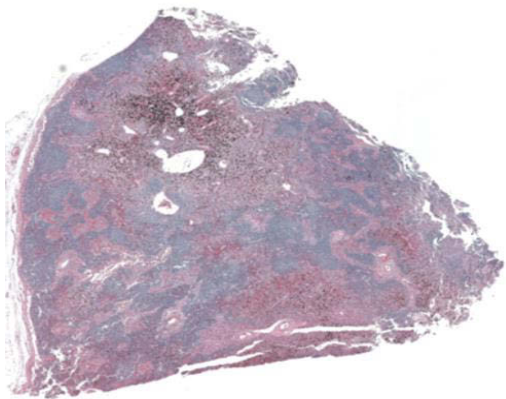
FFPE tissues

Human lymph node



~18,000 genes targeted in human

~20,000 genes targeted in mouse

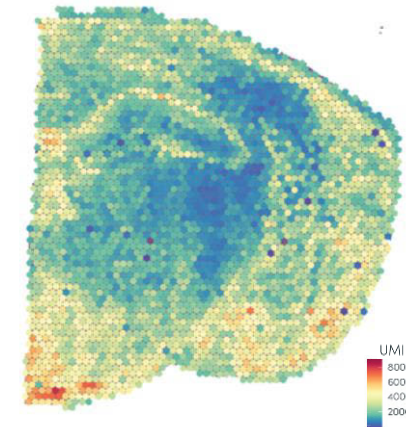
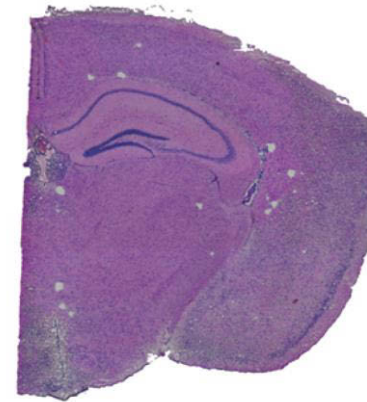


Fresh frozen tissues

Mouse brain



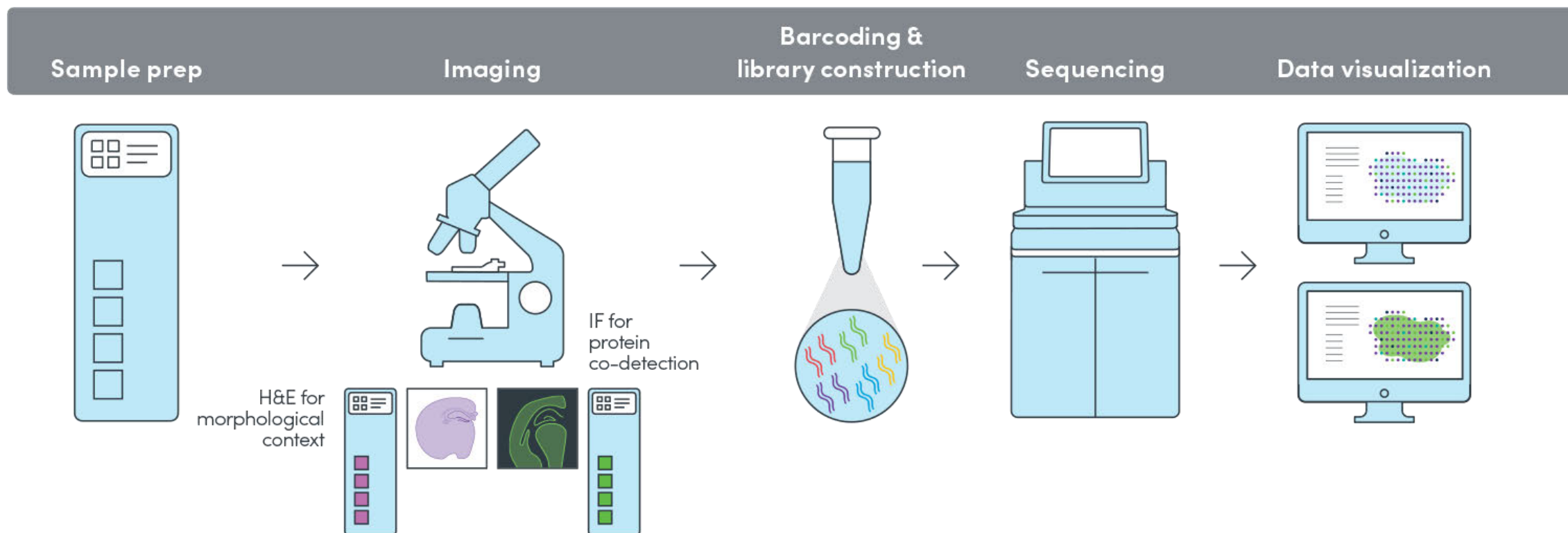
Applicable for a variety of species



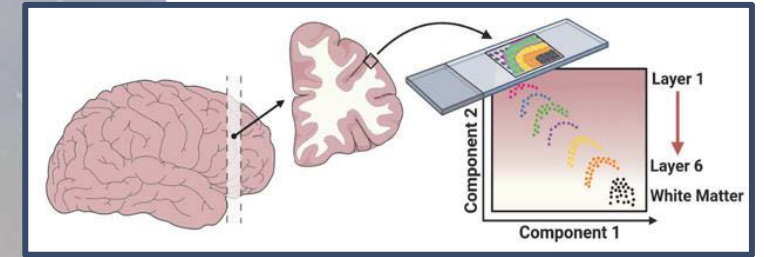
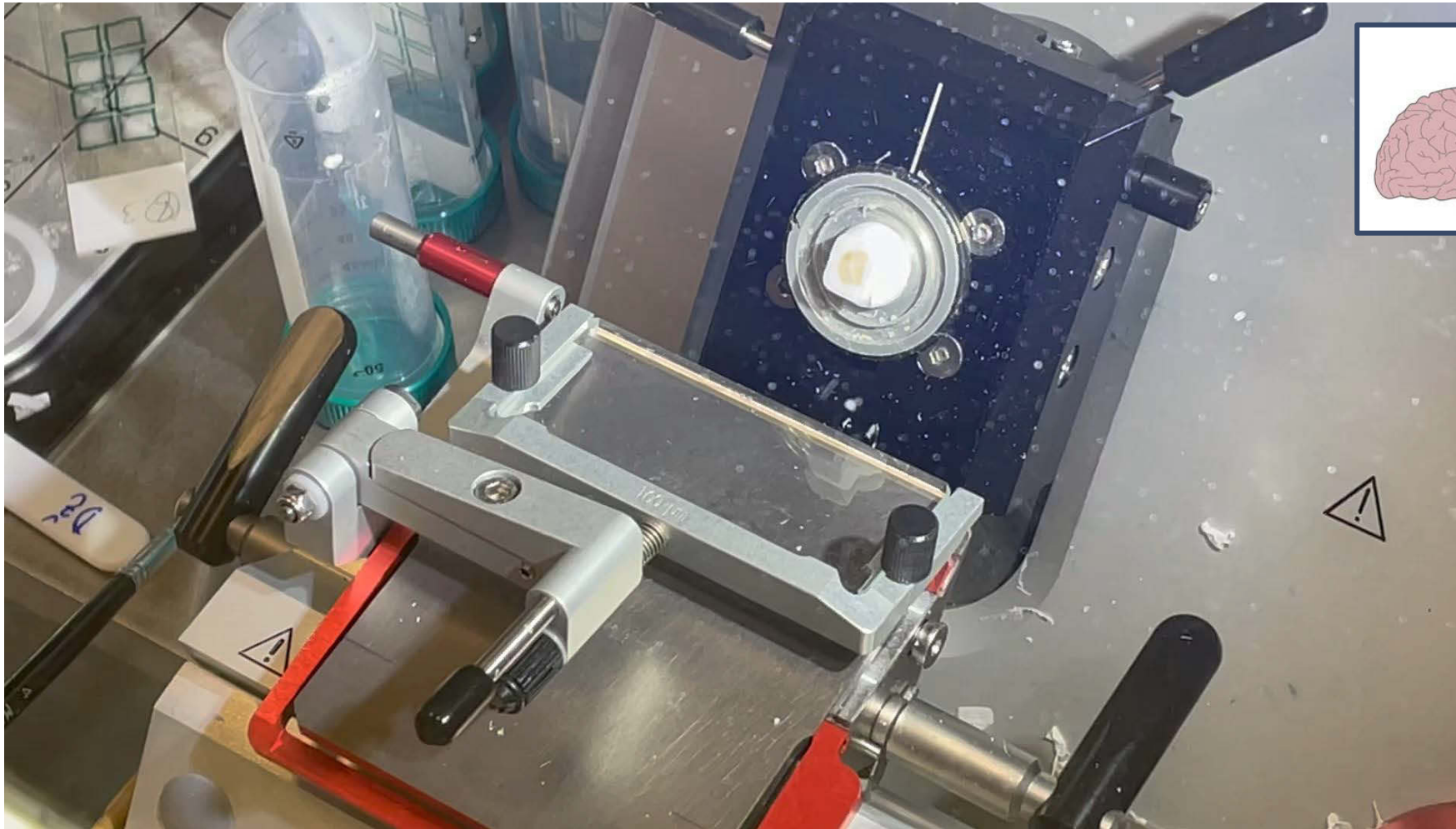
Visium Spatial Gene Expression workflow

Streamlined workflow with the choice of H&E or Immunofluorescence

Provides whole transcriptome or targeted gene expression with protein co-detection

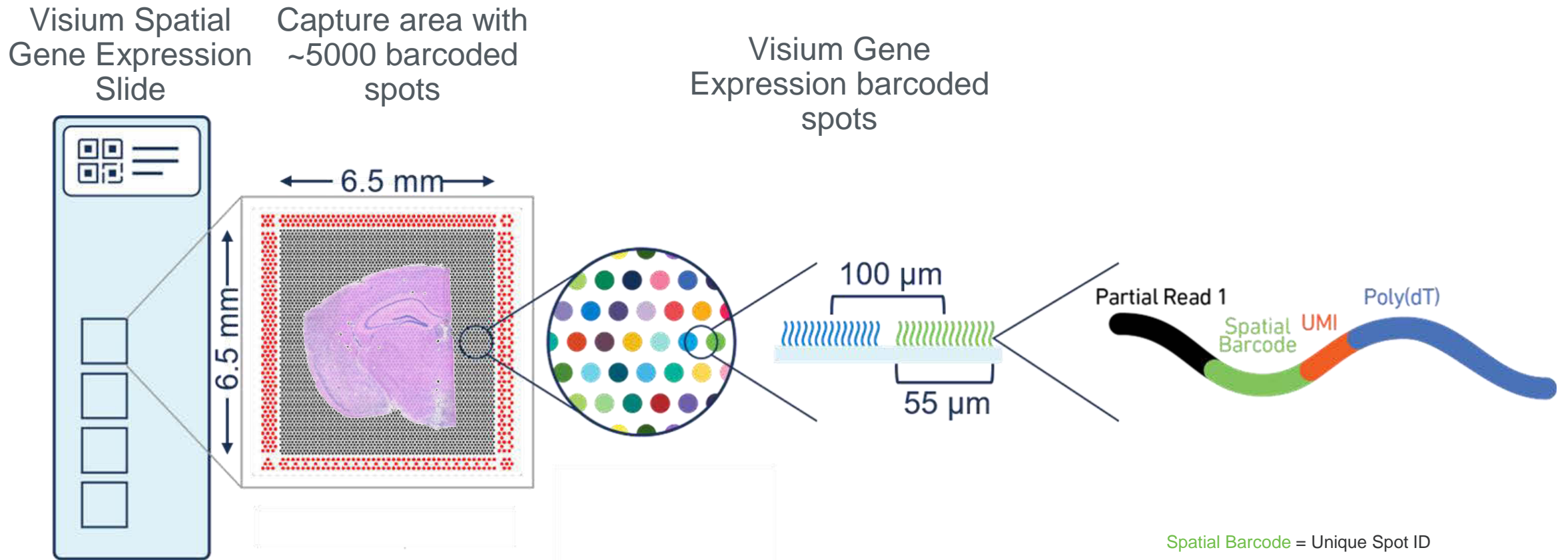


Visium Fresh - Frozen samples hands on procedure



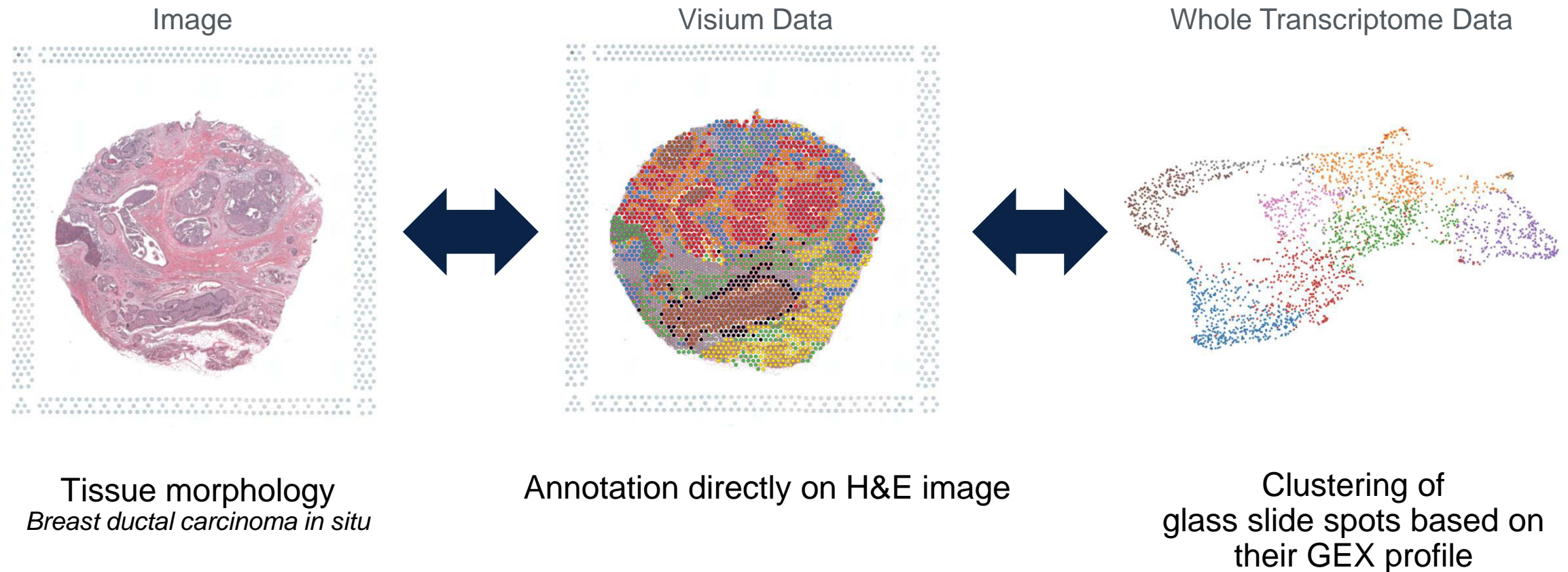
Unbiased gene expression at high spatial resolution

Utilizing Poly-A Capture and unique spatial barcodes



Cluster or image driven analysis of spatial data

Start with the gene expression data or microscopy images of the same section

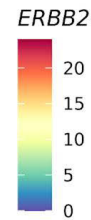
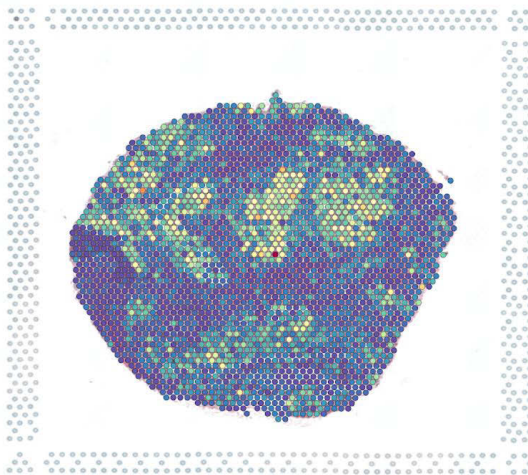


Explore Ductal Carcinoma In Situ of the Breast Cancer

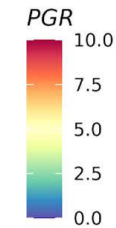
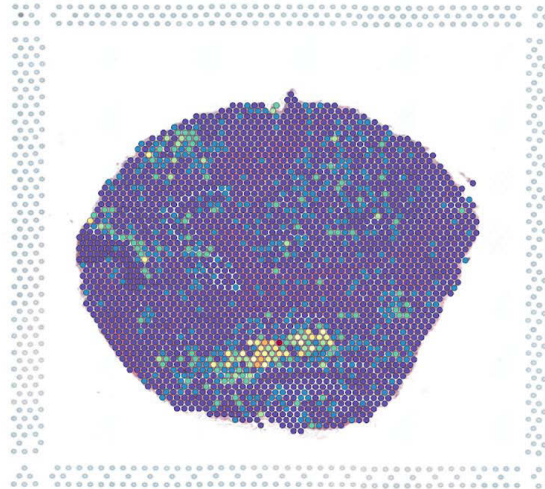
Start with gene expression

Key breast cancer biomarkers

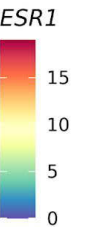
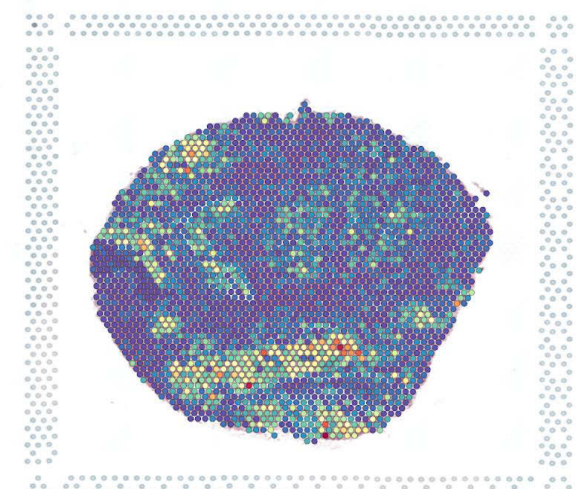
ERBB2



PGR



ESR1



Formalin is most common sample preservation method

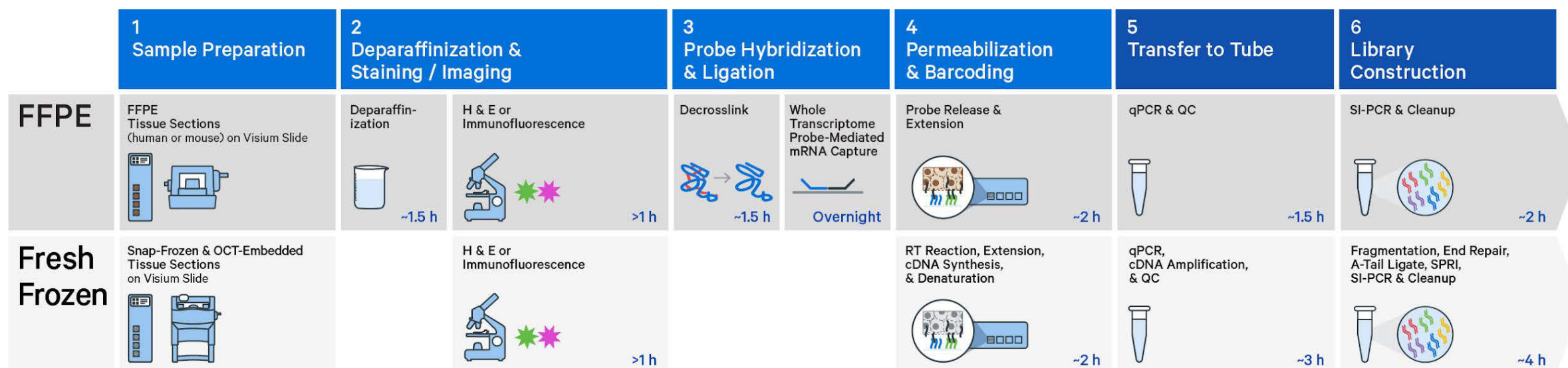


- Fresh tissue with methanol fixation is the only processing currently compatible with Visium
- Formalin Fixation and Paraffin Embedding (FFPE) is the primary method of clinical sample preservation
 - Excellent for preservation and tissue stability

Current problem:

- **FFPE leads to the sequestration of analytes (due to crosslinking) and nucleic acid degradation**
 - Makes Next Generation Sequencing (NGS) analysis particularly challenging

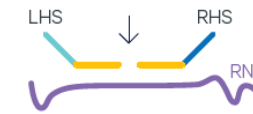
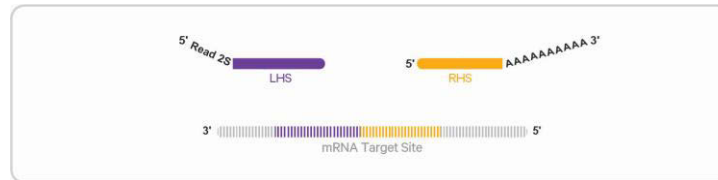
Fresh- Frozen and FFPE samples workflow comparison



Visium for FFPE utilizes new chemistry to detect mRNA

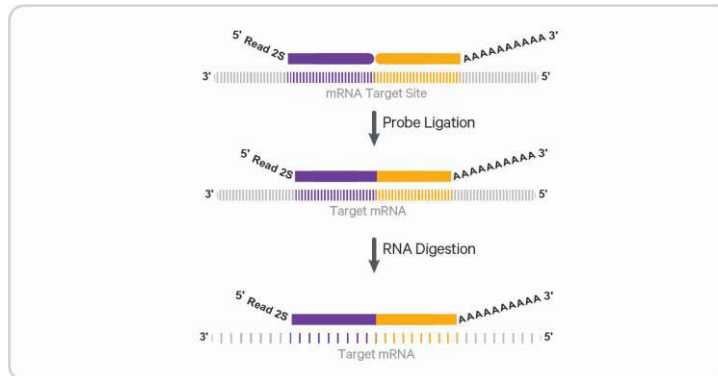
RNA Templated Ligation (RTL) for sensitive, specific RNA detection in FFPE samples

Probe pairs designed against the protein-coding transcriptome, one pair per gene



Split two probe chemistry reduces nonspecific signal

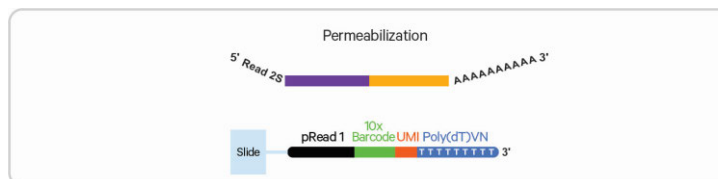
Probe hybridization and ligation



~18,000 genes targeted in human

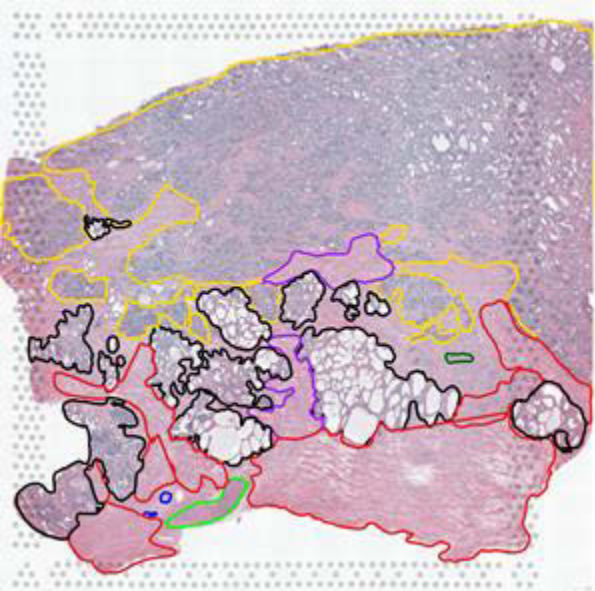
Ligated probe pair released

Tissue permeabilization and ligated probe pair capture



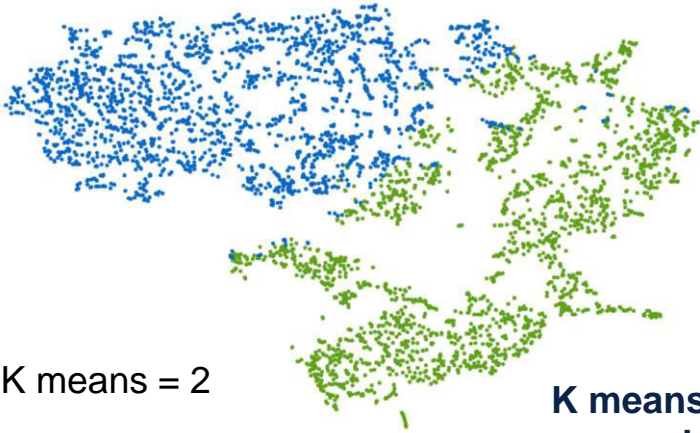
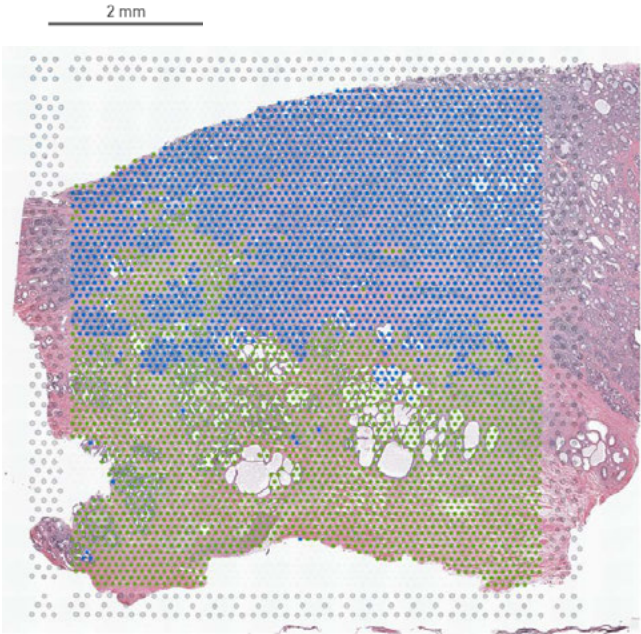
~20,000 genes targeted in mouse

Prostate Tumor and Markers

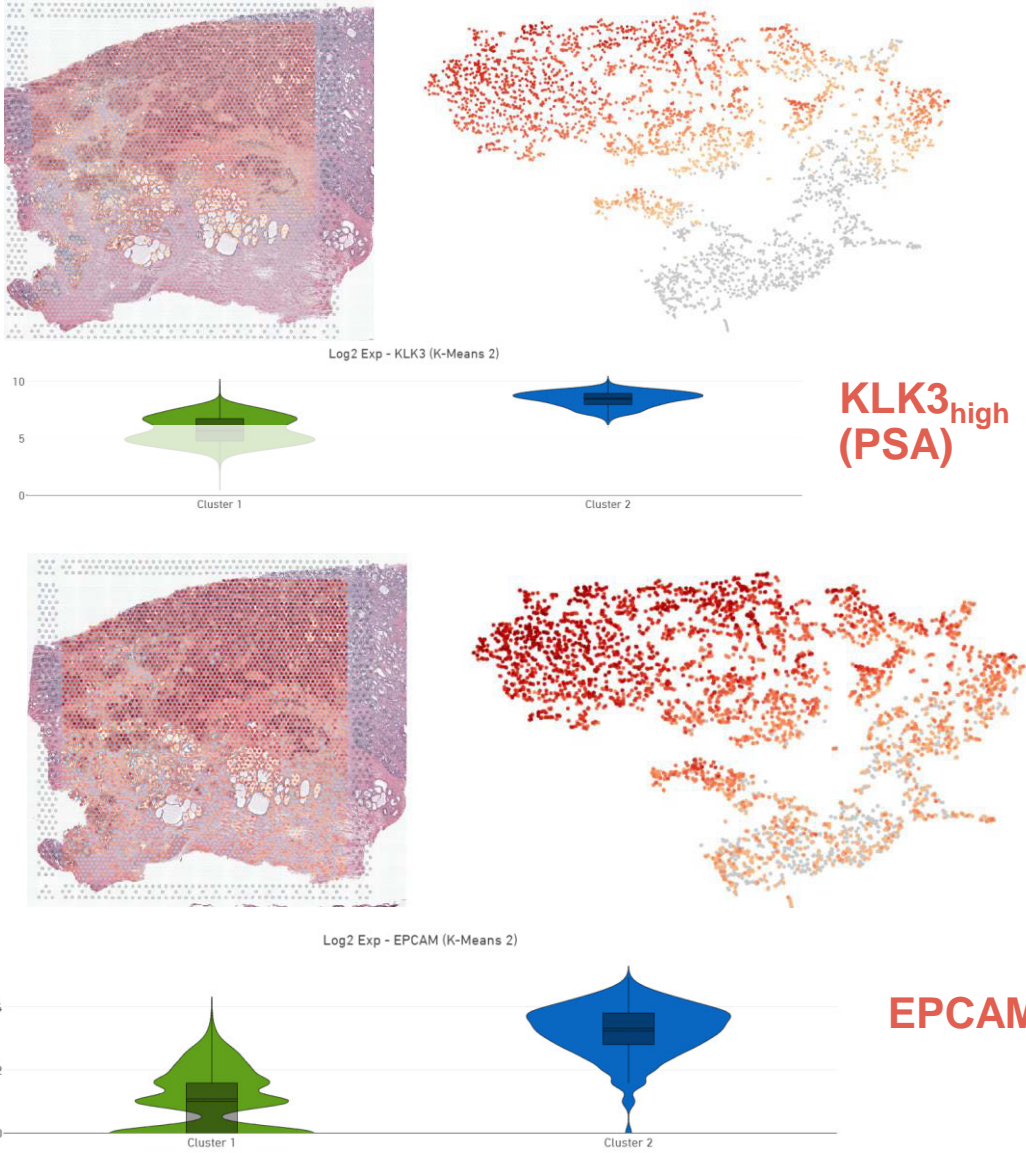


Classification

- Blood Vessel
- Fibro-Muscular Tissue
- Fibrous Tissue
- Immune Cells
- Invasive Carcinoma
- Nerve
- Normal Gland



K means = 2

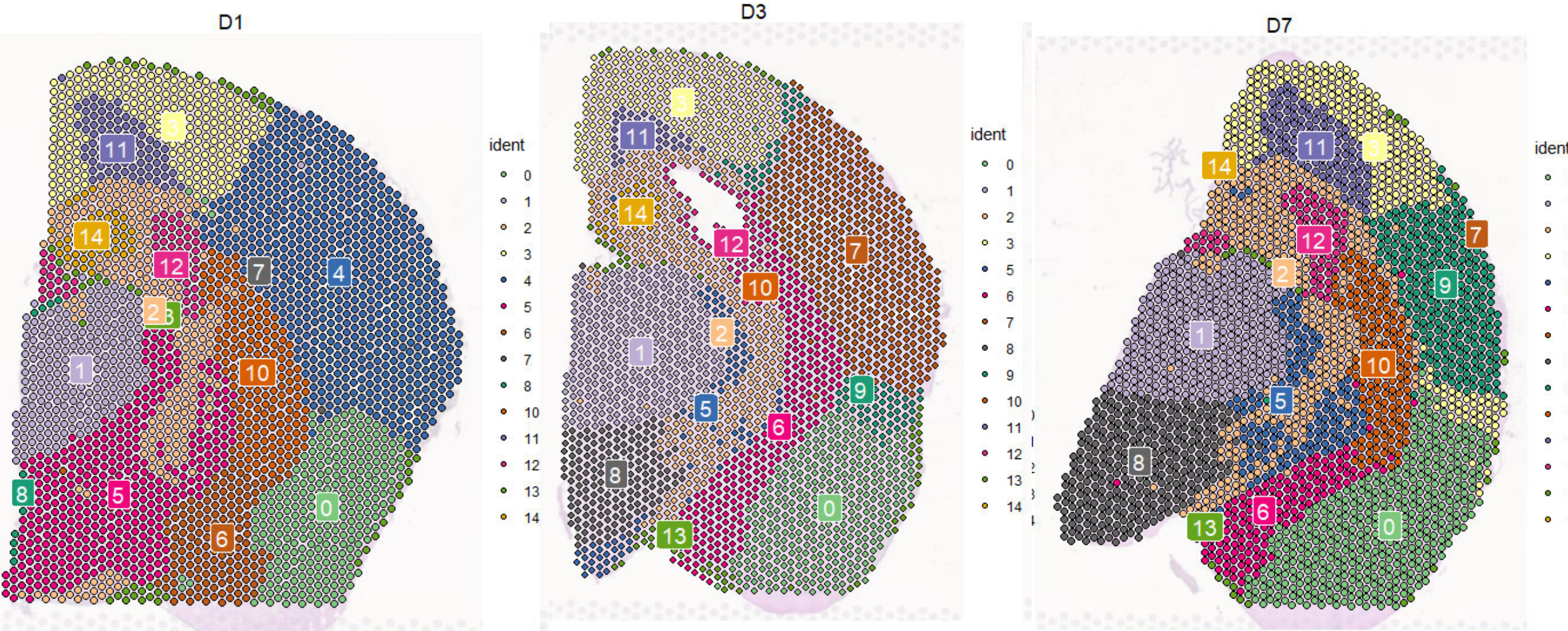


KLK3_{high}
(PSA)

EPCAM

K means clustering finds strongest pattern in cell clusters.
eg carcinoma vs noncarcinoma if we define to find 2 strongest clusters

Inschemia in colour, by Daniel Zucha, Institute of Biotechnology, CAS



Day 1

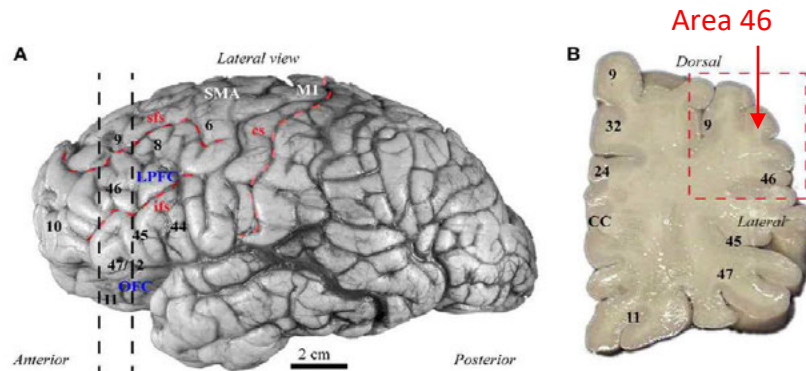
Day 3

Day 7

Spatial Gene Expression in Human Prefrontal Cortex

- Visium platform was used to define the spatial topography of gene expression in the **six-layered human dorsolateral prefrontal cortex**

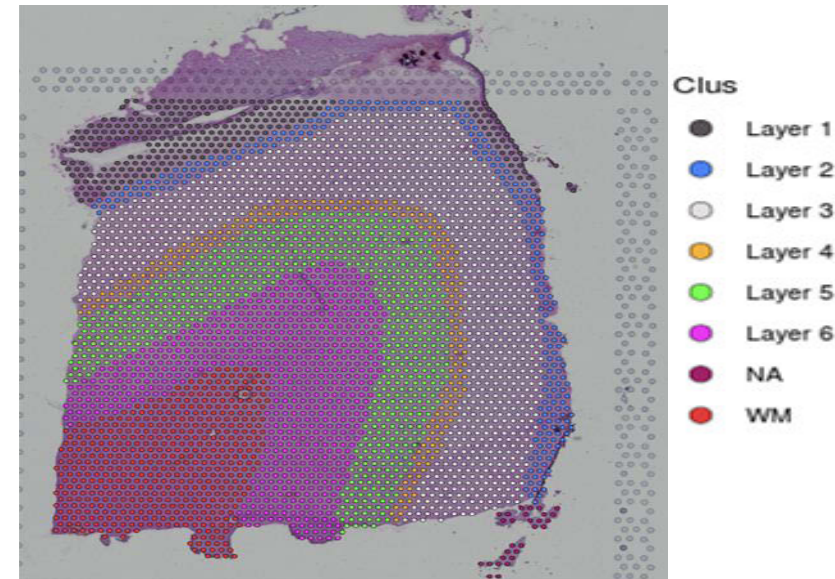
Human brain prefrontal cortex



Zikopoulos and Barbas, 2013

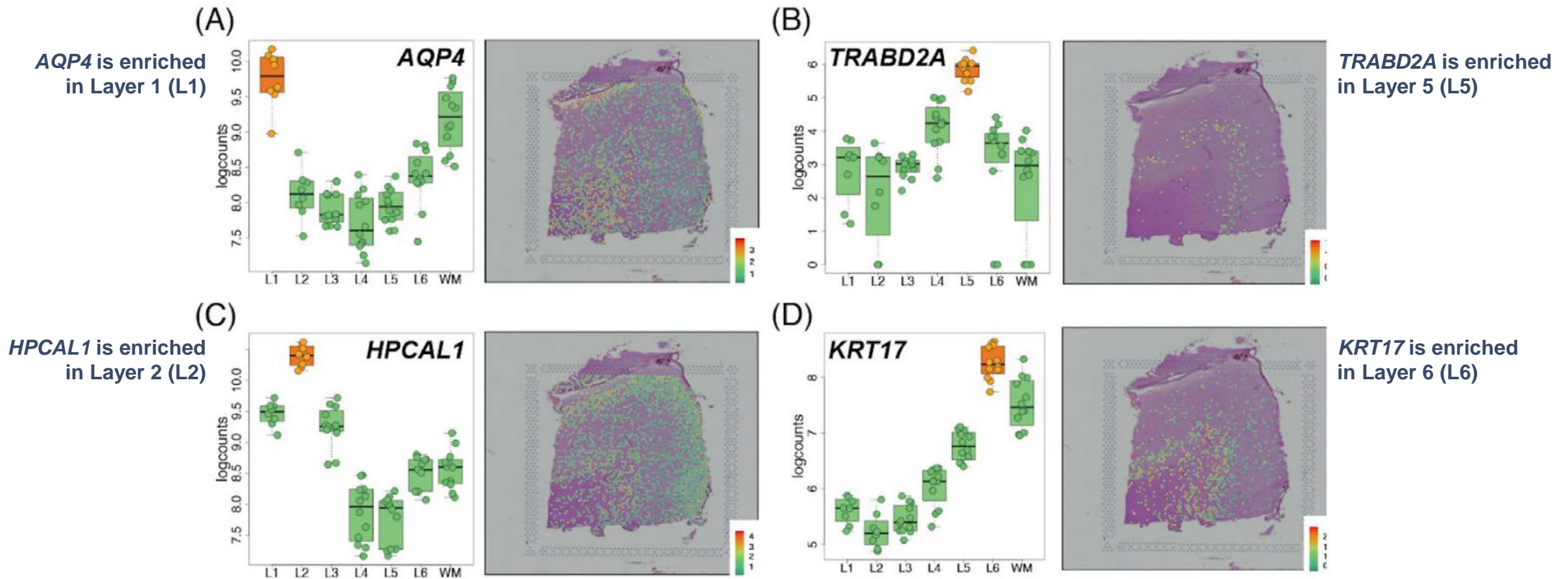
nature
neuroscience

Cortical layers revealed by Visium



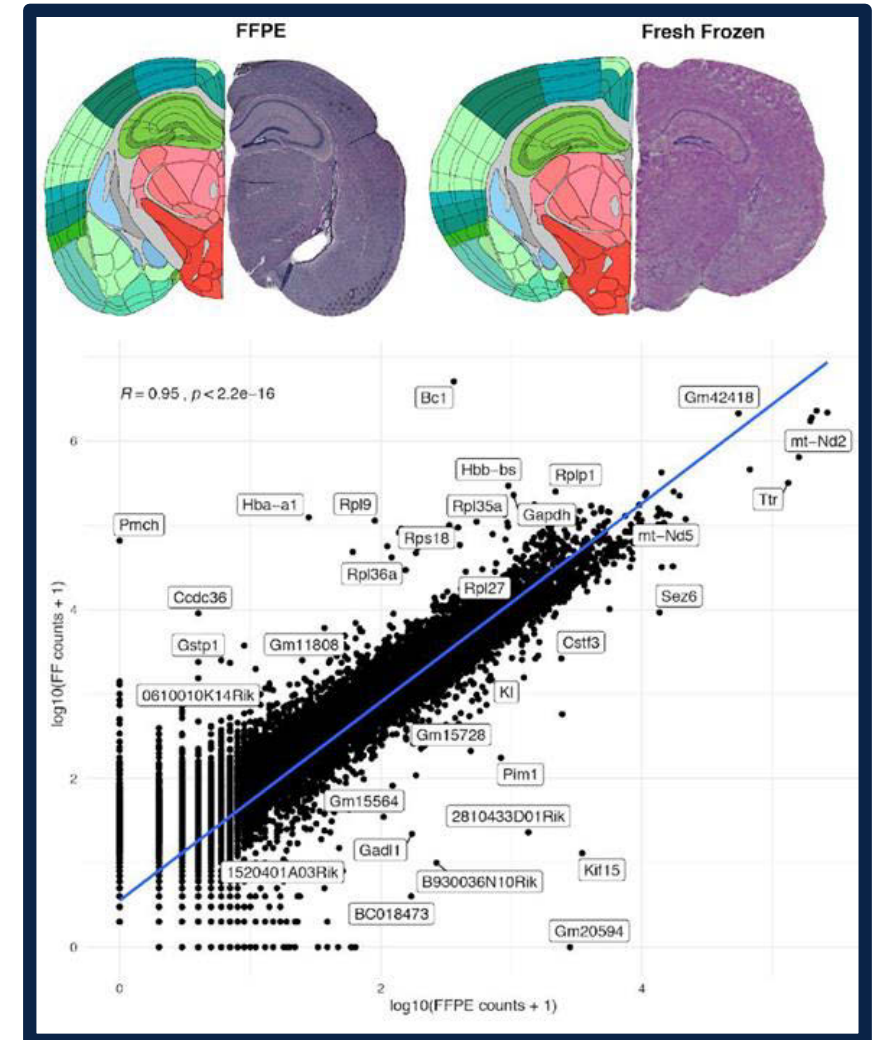
Spatial Gene Expression in Human Prefrontal Cortex

- Visium discovers novel cortical layer-enriched genes, associated with schizophrenia and autism spectrum disorder, highlighting the clinical relevance of spatially defined expression.



Genome-wide Spatial Expression Profiling in FFPE Tissues

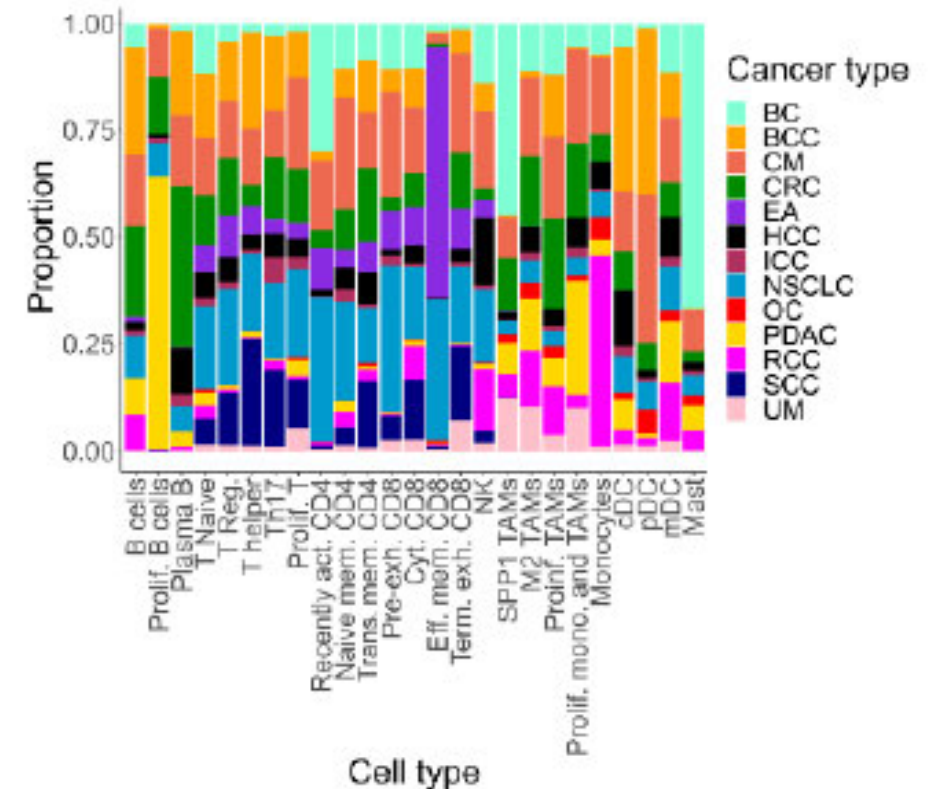
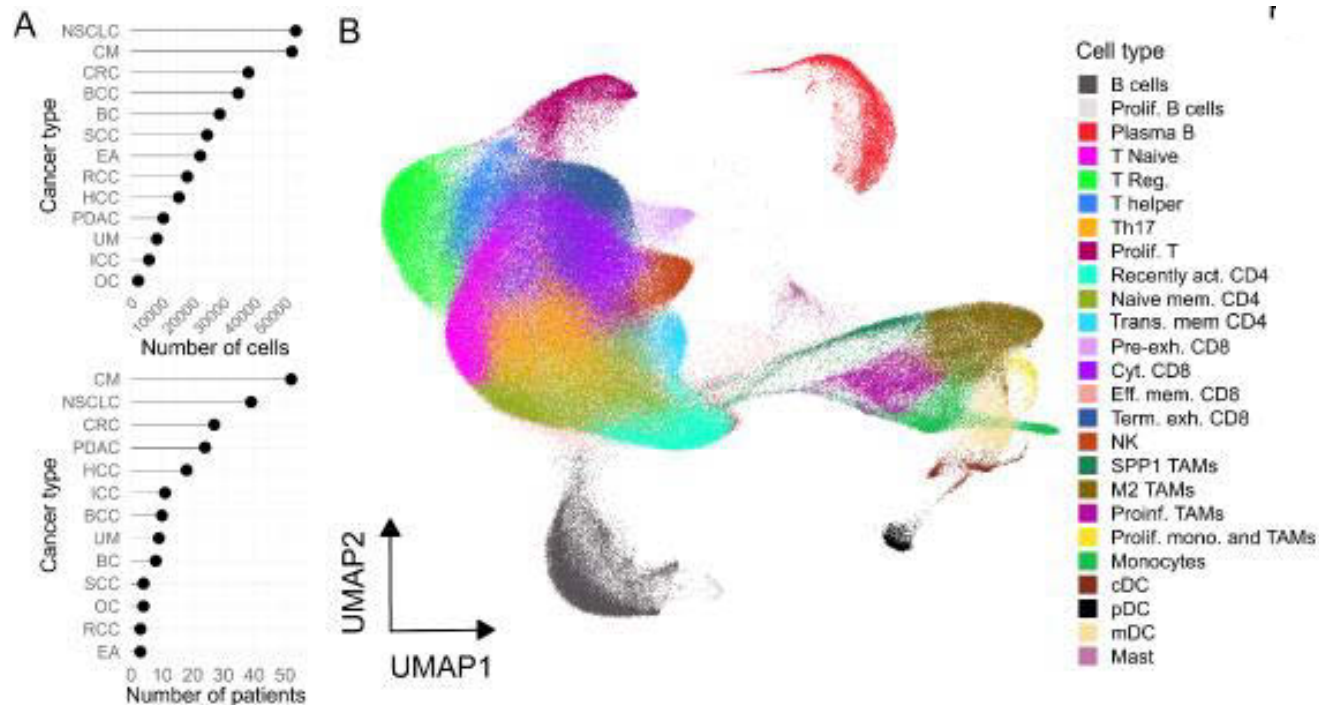
- Authors present a procedure to perform genome-wide spatial analysis of mRNA in FFPE tissue sections using *Visium Spatial*.
- Authors conducted expression profiling and cell type mapping in **coronal sections from the mouse brain** to demonstrate the method's capability to delineate anatomical regions from a molecular perspective.
- They further explored the spatial composition of transcriptomic signatures in **ovarian carcinosarcoma** samples using data driven analysis methods, exemplifying the method's **potential to elucidate molecular mechanisms in heterogeneous clinical samples**.
- Comparison with data from Allen Brain Atlas



A single-cell tumor immune atlas for precision oncology

Built and tested a cancer immune atlas for patient stratification using single cell and spatial gene expression data

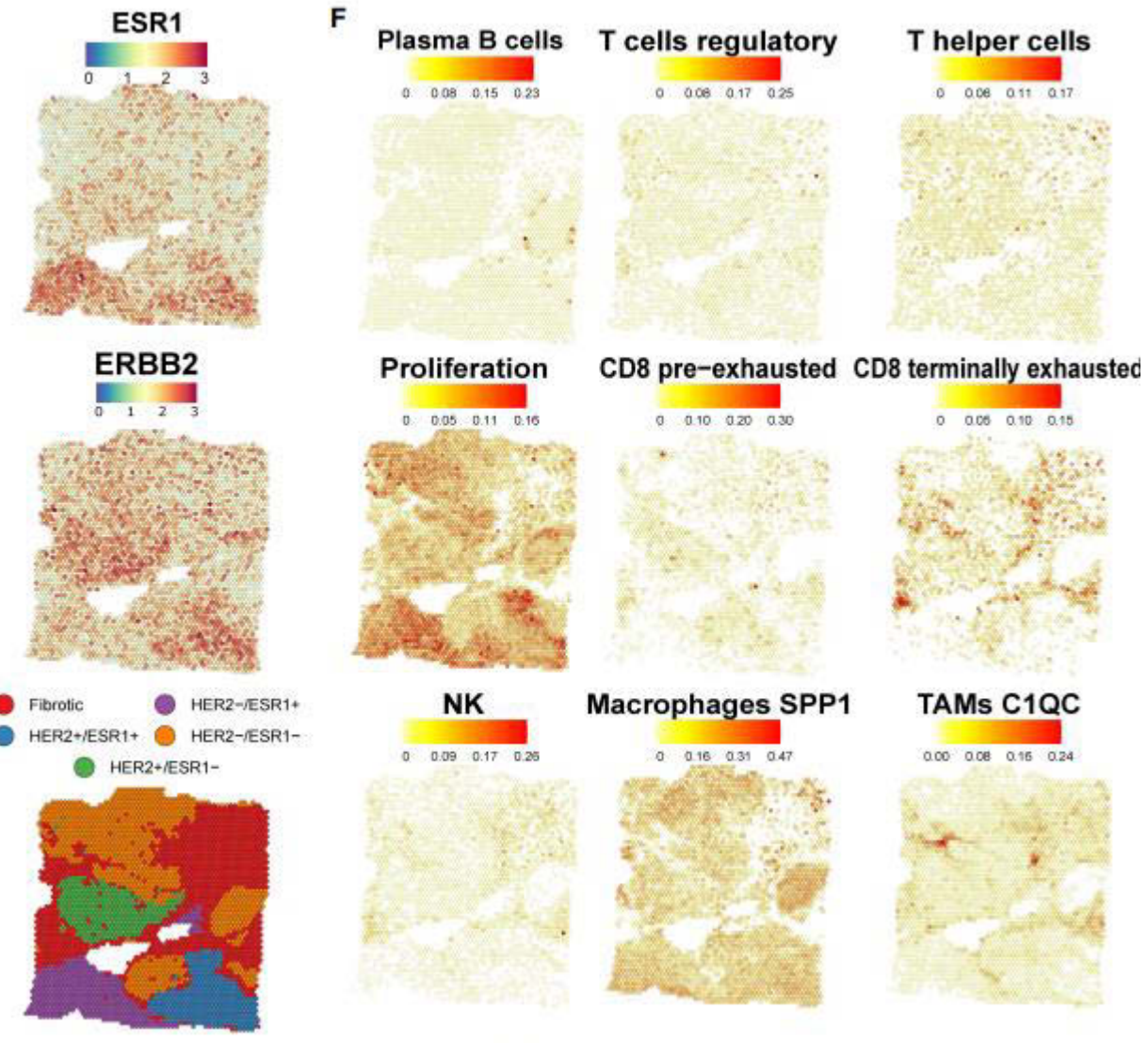
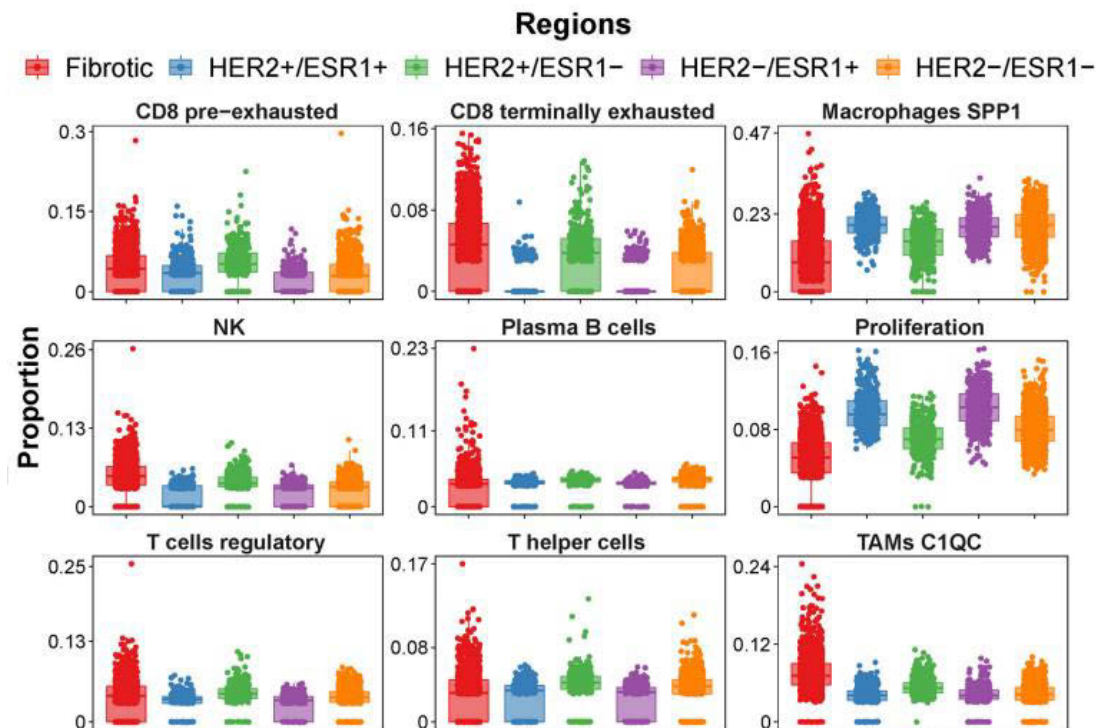
Identified shared immune signatures across patients and cancer types



<https://www.genome.org/cgi/doi/10.1101/gr.273300.120>

A single-cell tumor immune atlas for precision oncology

- Integrated single cell and spatial data to explore tumor immune composition across cancer types
- Found differences in immune architecture across tumor types which could contribute to unraveling immuno-therapy responses



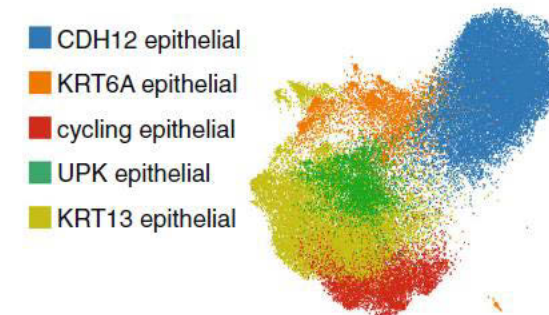
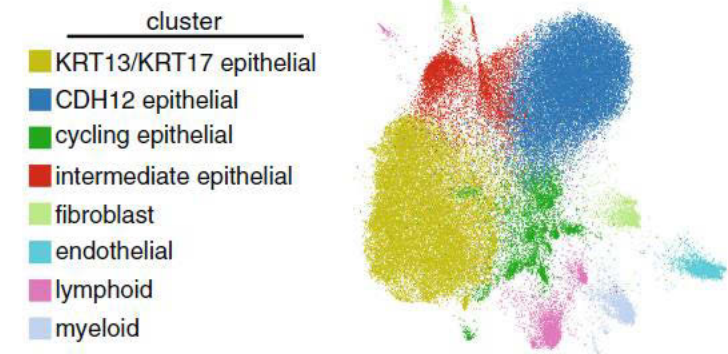
An N-Cadherin 2 expressing epithelial cell subpopulation predicts response to surgery, chemotherapy and immunotherapy in bladder cancer

The authors combined *Chromium* single nuclei RNA sequencing with *Visium* spatial transcriptomics and single-cell resolution spatial proteomic analysis of human bladder cancer to identify an epithelial subpopulation with therapeutic response prediction ability.

These cells expressed Cadherin 12 (CDH12, N-Cadherin 2), catenins, and other epithelial markers. CDH12-enriched tumors defined patients with poor outcome following surgery with or without neoadjuvant chemotherapy.

In contrast, CDH12-enriched tumors exhibited superior response to immune checkpoint therapy.

MIBC urothelial histology
fresh-frozen
(N = 25 patients) → nuclei isolation → snSeq (10x 3')



Gouin et al., 2021, Nature Communications, <https://doi.org/10.1038/s41467-021-25103-7>

Single-cell ATAC and RNA sequencing reveal pre-existing and persistent subpopulations of cells associated with relapse of prostate cancer

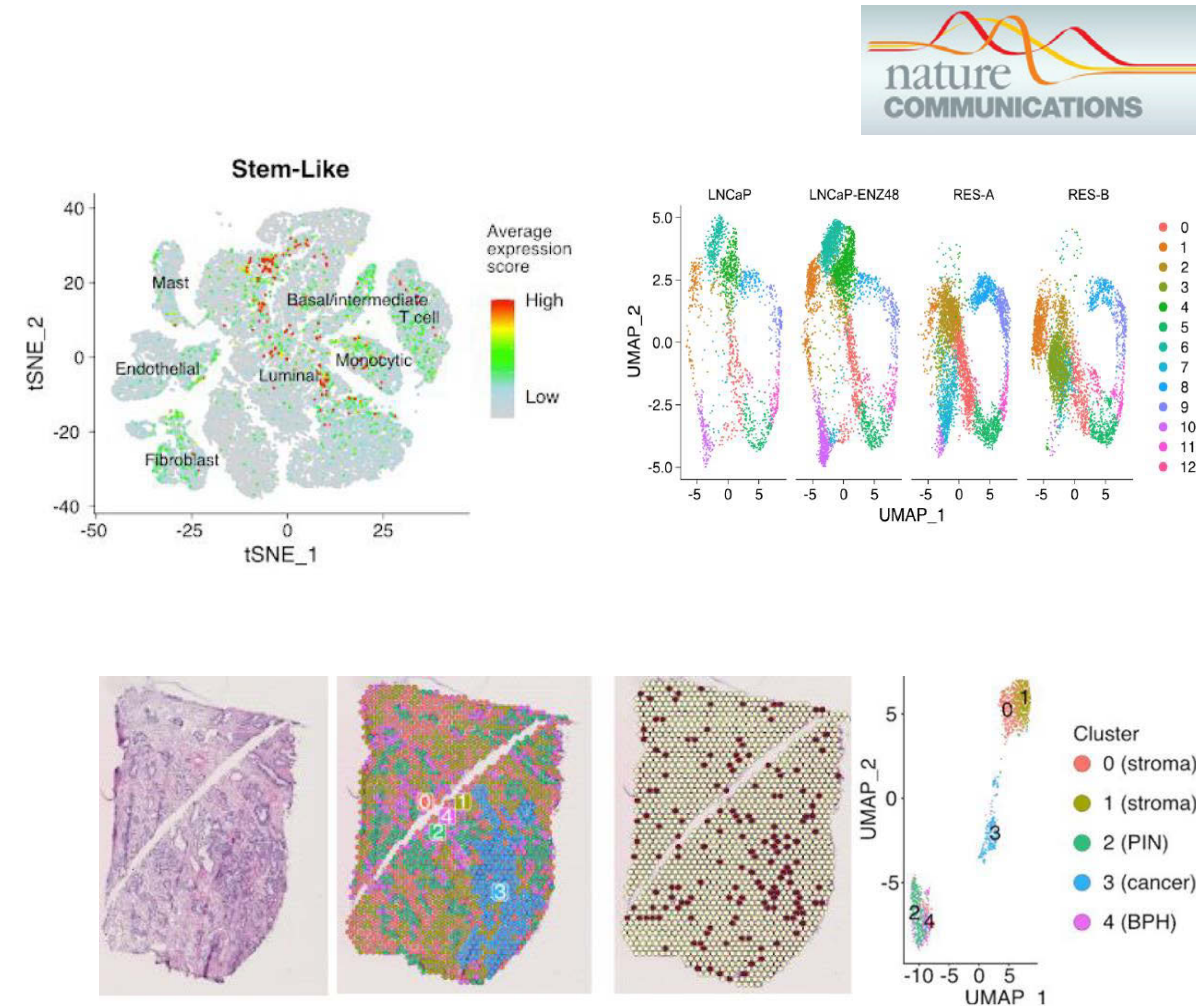
The authors employed *Chromium* single-cell assay for transposase-accessible chromatin (ATAC) and RNA sequencing in prostate cancer models of early treatment response and resistance to enzalutamide (ENZ).

They identified pre-existing and treatment-persistent cell subpopulations that possess transcriptional stem-like features and regenerative potential.

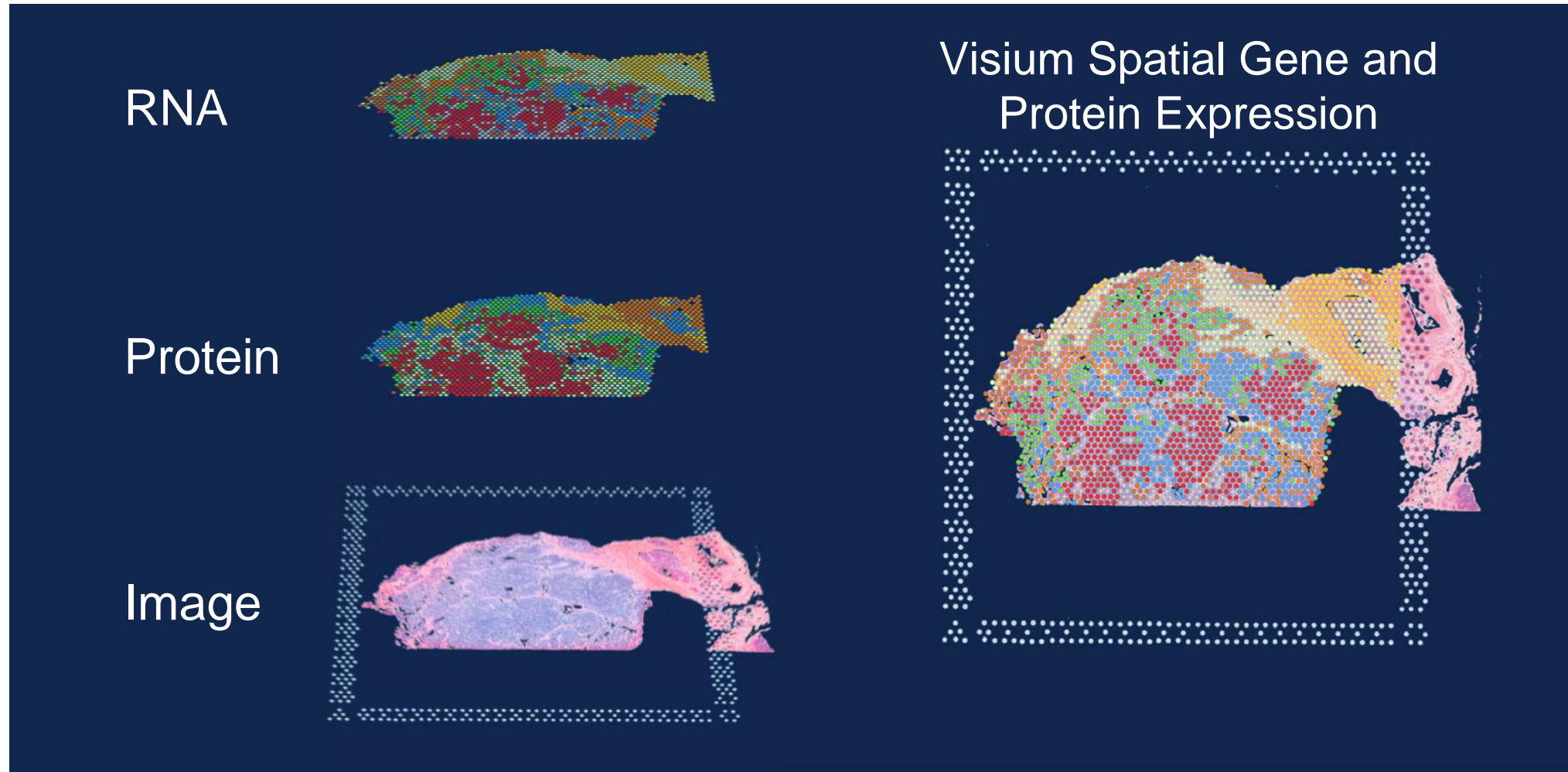
Distinct chromatin landscapes were associated with ENZ treatment and resistance.

Transcriptional profiles characteristic of persistent stem-like cells were able to stratify the treatment response of patients.

Using *Visium Spatial*, the presence and location of the signatures of the stem-like cells within two sections of primary untreated prostate cancer were assessed.



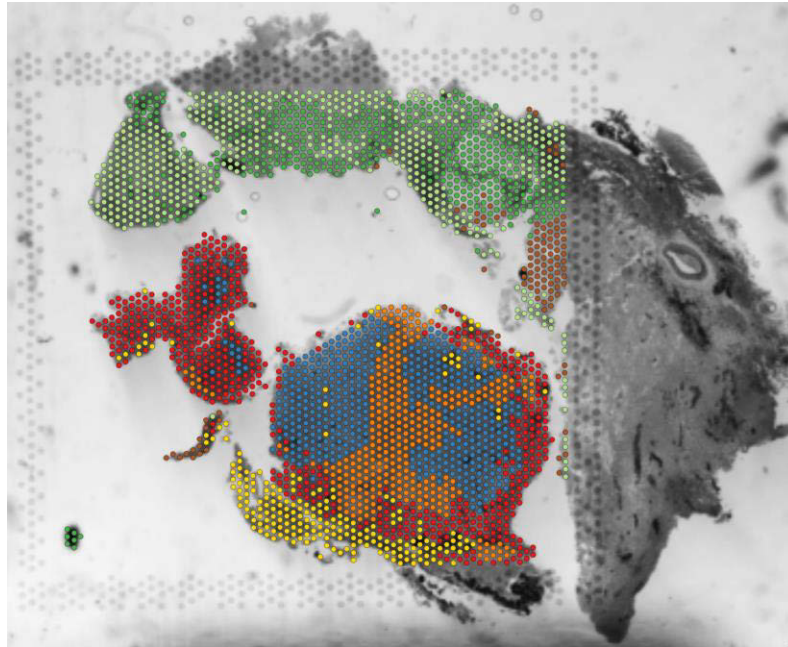
Simultaneous RNA and Protein detection for deeper insights



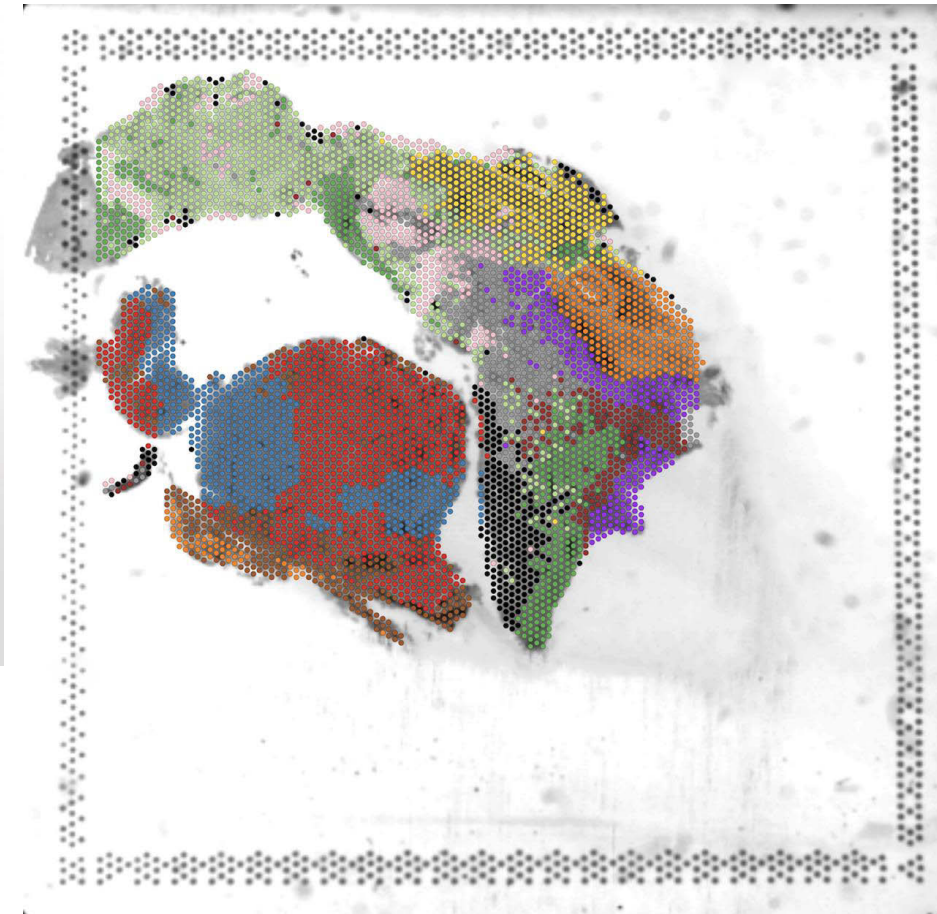
Visualize larger tissue sections without compromising block integrity

FFPE Human Brain Cancer

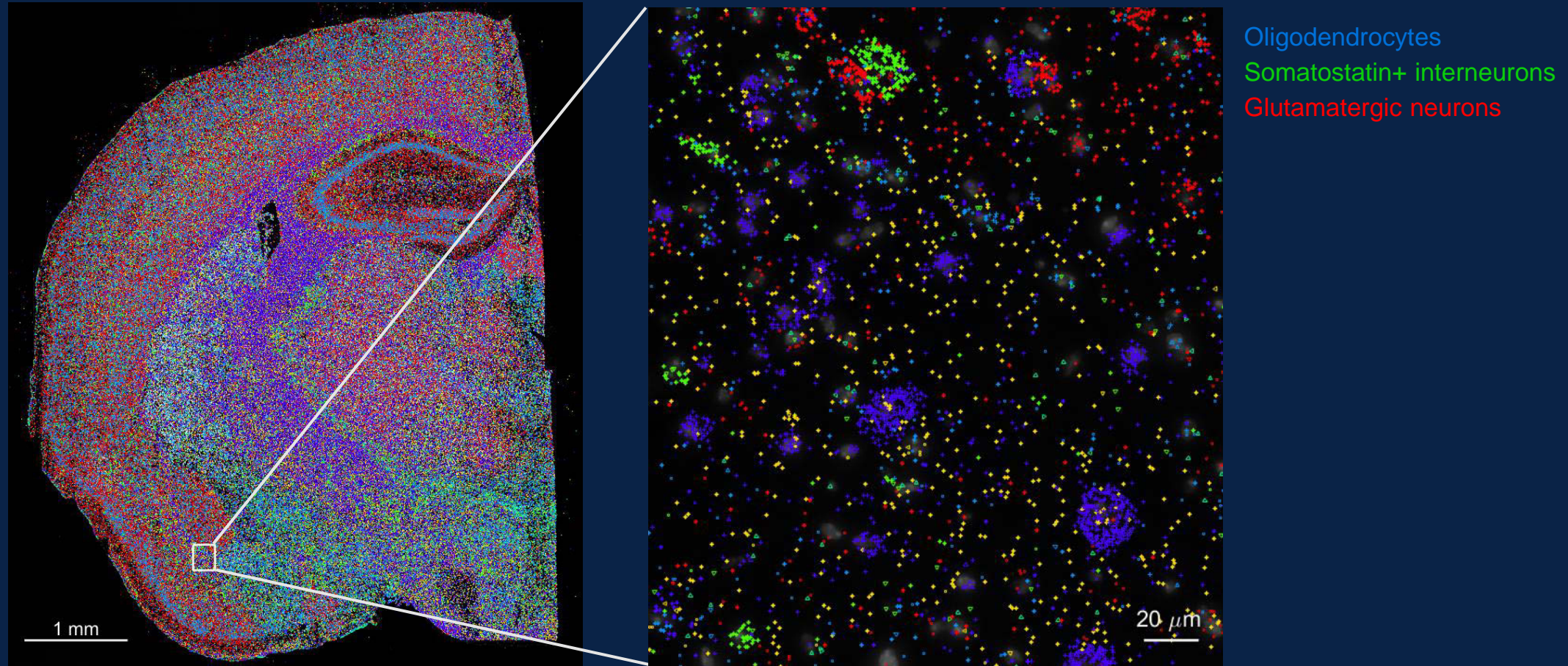
Visium GEx slide, 6.5mm



Visium GEx slide, 11mm



Profile an entire tissue section with subcellular resolution – In Situ solution



Xenium

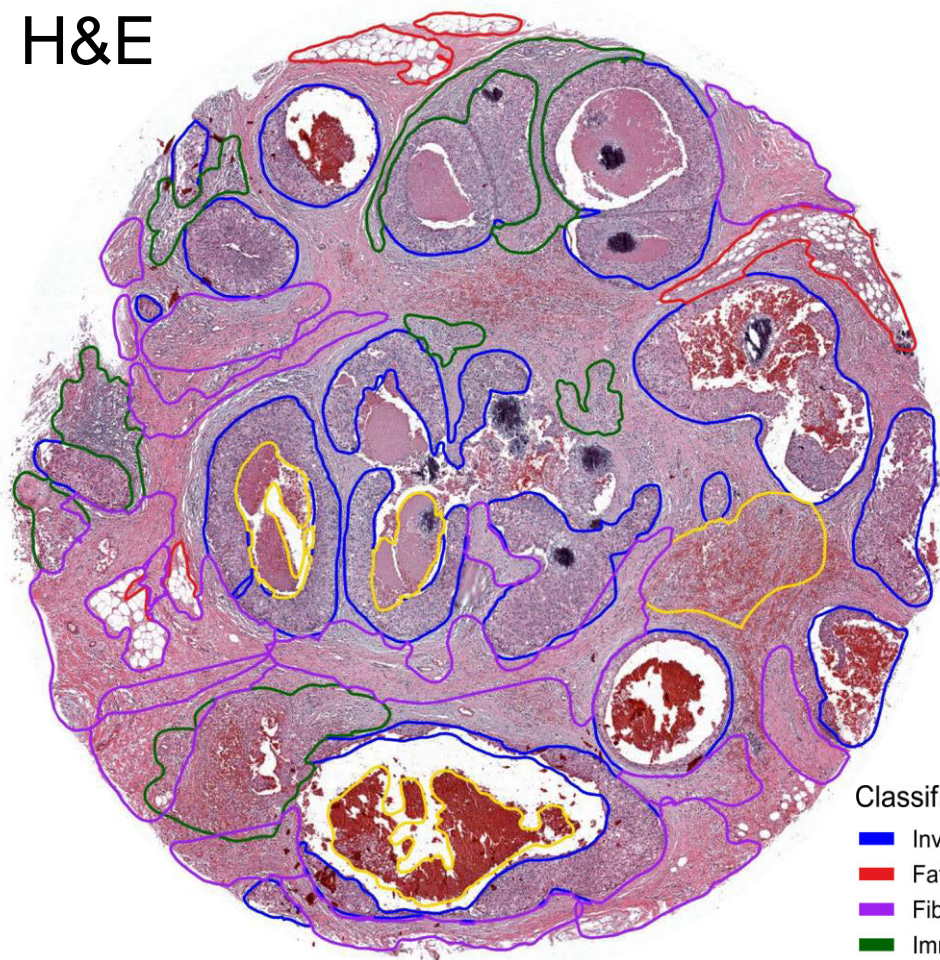
In Situ Platform

- Hundreds of gene targets
- Subcellular resolution
- Microscopy based read-out
- Fresh Frozen and FFPE
- Simultaneous RNA and proteins
- Throughput for larger cohorts



In situ analysis of human FFPE breast cancer

H&E



Classification

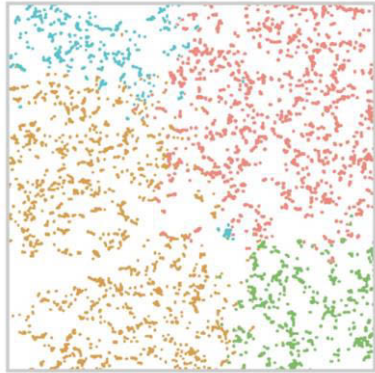
- Invasive Carcinoma
- Fat
- Fibrous Tissue
- Immune Cells
- Necrosis

Xenium
In Situ

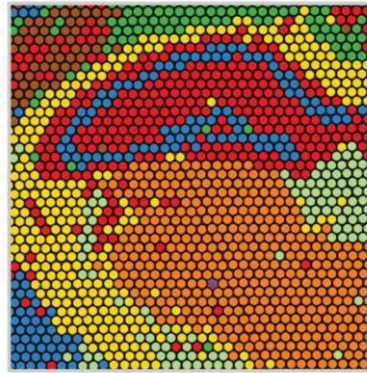


From discovery to clinical with three complementary workflows

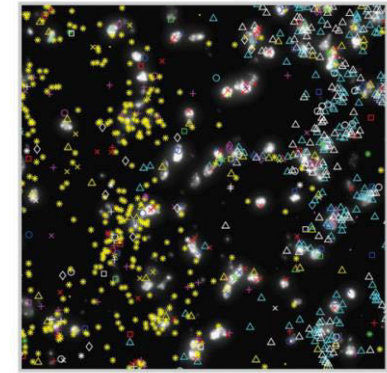
Chromium Single Cell



Visium Spatial



In Situ



Discovery

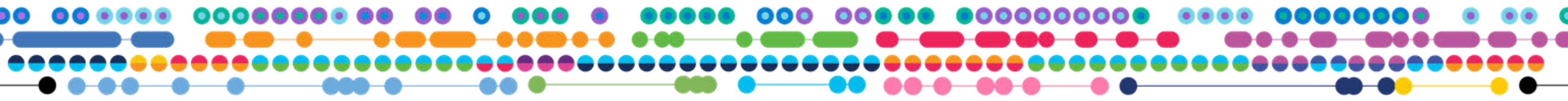
Translational

Clinical

Sample Preparation for 10x Genomics

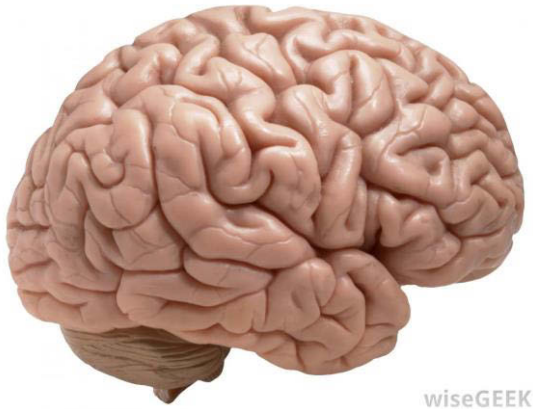
A How To Guide: Considerations and Best Practices

Agnieszka Ciesielska PhD, STA 10x Genomics

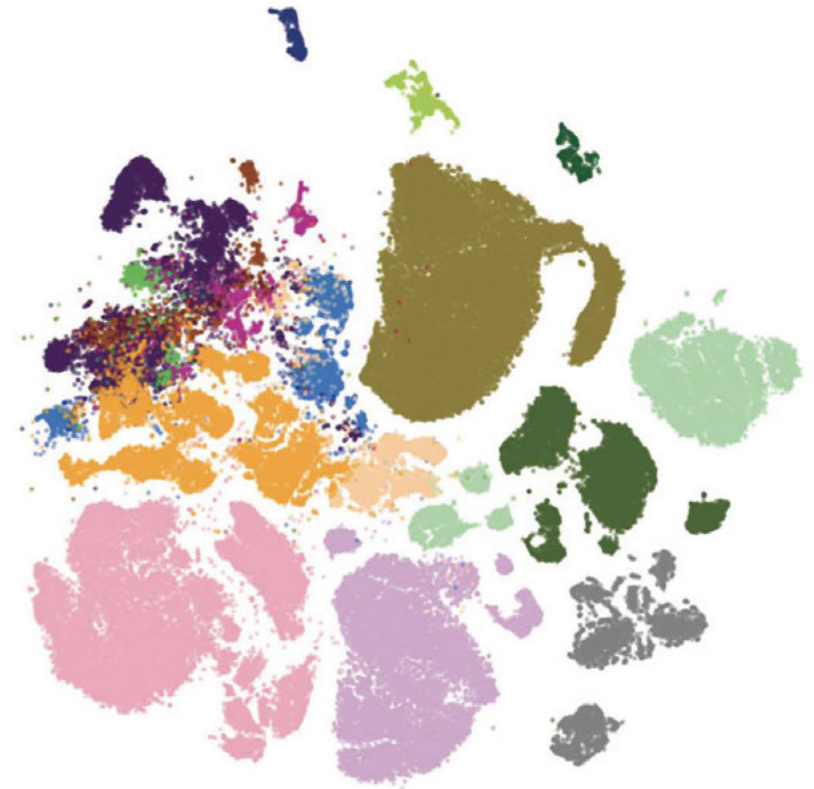


From averages to high resolution

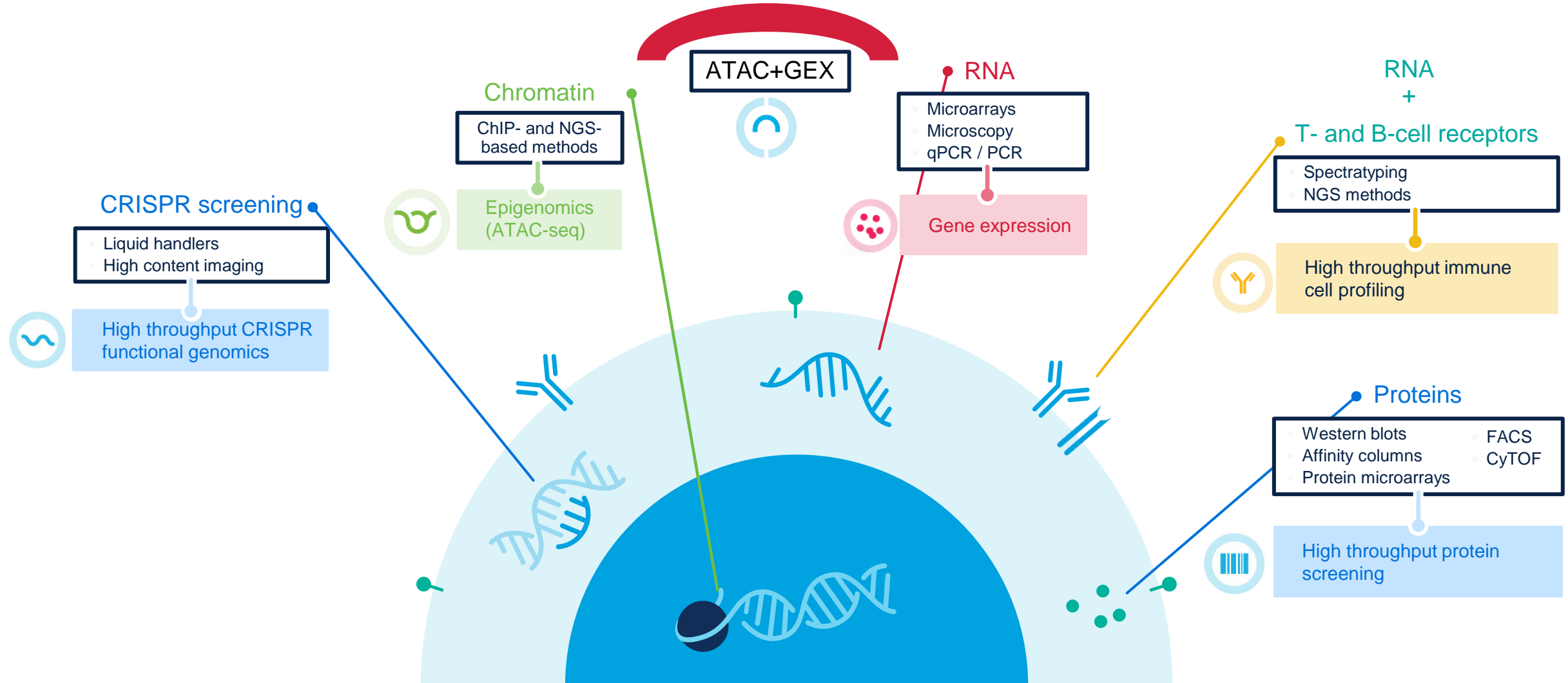
Whole Tissue/Organs
(Genetic) Disease Model



Resolve Cell Type-Specific
Data



Next generation molecular profiling solutions



How We Think About Sample Preparation

It's what you bring to the experiment.

It's a workflow. A set of decisions.

Quality is critical.

Chromium Single Cell Gene Expression Workflow

Input

Library Creation

Sequencing

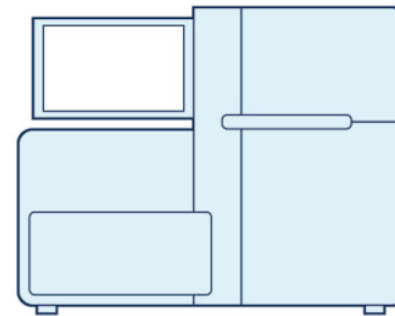
Data Analysis and
Visualization



Suspension of
dissociated single
cell/nuclei



**Cell partitioning and
molecular barcoding**



Sequencing

Analysis

Visualization

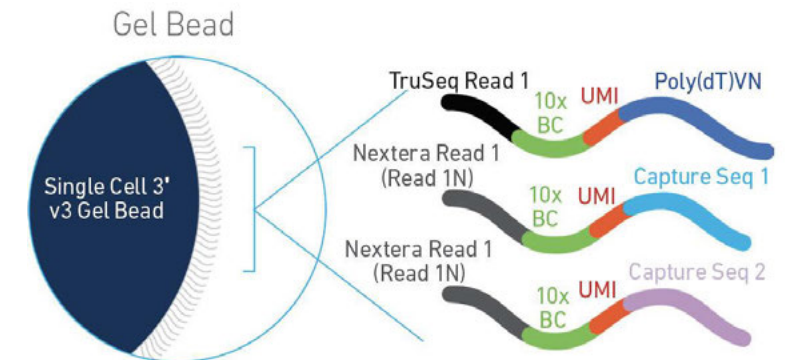
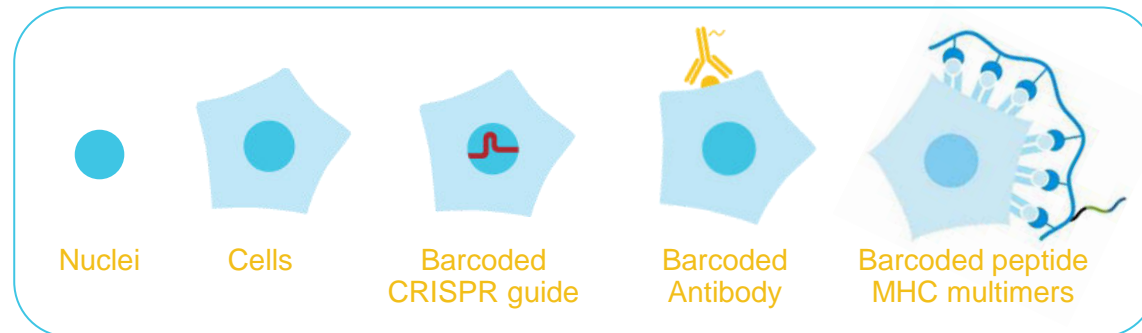
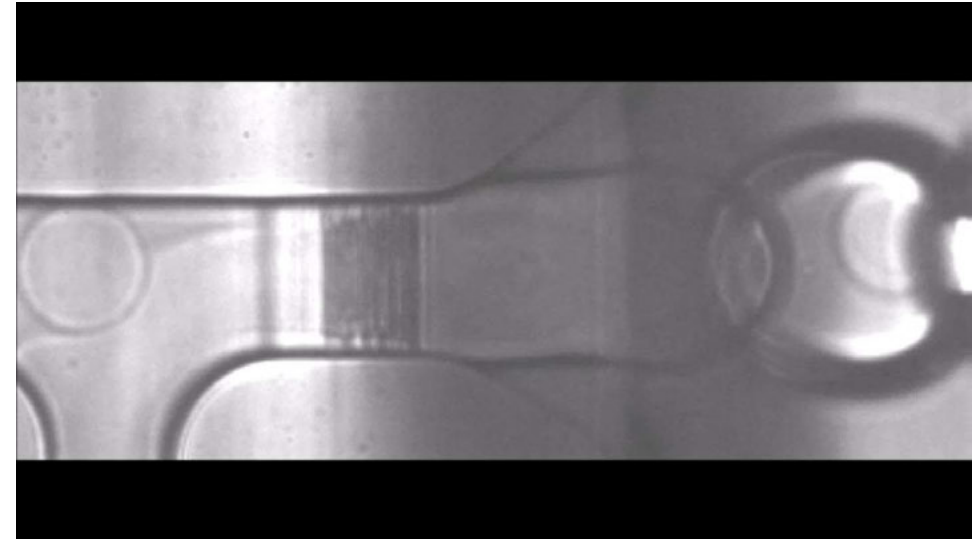
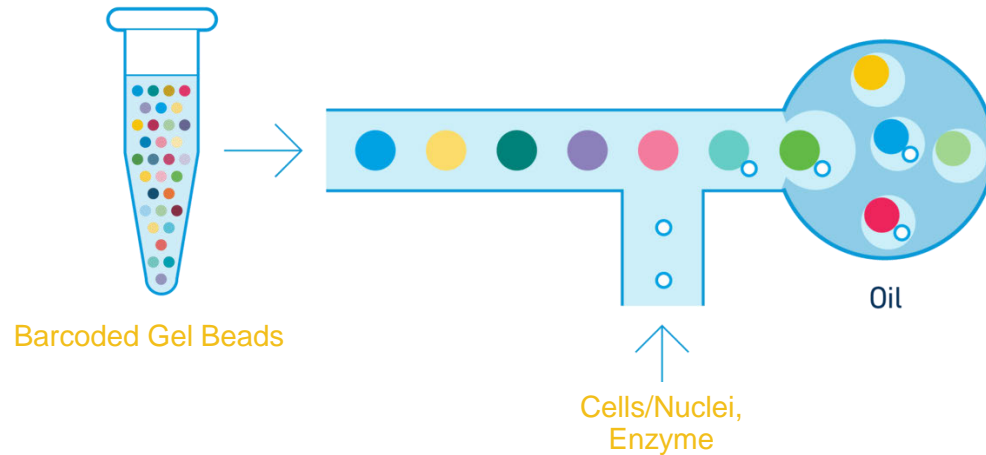


Community Analysis
Tools

- 8 channels/chip
- 500-10 000 cells recovered per channel
- 40-65% cells recovered

Technology

Partitioning and molecular barcoding millions of parallel reactions



Single cell sample prep resources from 10x Genomics

- <https://support.10xgenomics.com/>
- Protocols are free to download

General sample preparation guidelines

- Guidelines for optimal sample preparation
- Guidelines for accurate target cell counts
- General cell preparation guide
- Preparation of single cell suspensions from cultured cell lines
- Isolation of nuclei

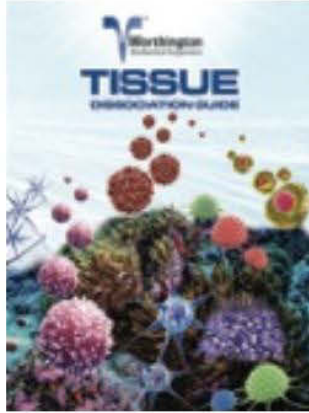
Preparation of specific sample types

- Fresh frozen human-mouse cell line mixtures
- Fresh frozen human peripheral blood mononuclear cells
- Dissociation of mouse embryonic neural tissue
- Tumor dissociation
- Methanol fixation of cells
- Moss protoplast suspensions

Sample improvement

- Enrichment of CD3+ T cells from dissociated tissues
- Removal of dead cells from single cell suspensions

General Cell Handling Recommendations



Worthington Tissue Dissociation Guide

Introduction

Tissue dissociation/primary cell isolation and cell harvesting are principal applications for enzymes in tissue culture research and cell biology studies. Despite the widespread use of enzymes for these applications over the years, their mechanisms of action in dissociation and harvesting are not well understood. As a result, the choice of one technique over another is often arbitrary and based more on past experience than on an understanding of why the method works and what modifications could lead to even better results.

Tissue Tables (references, grouped by tissue type and species)

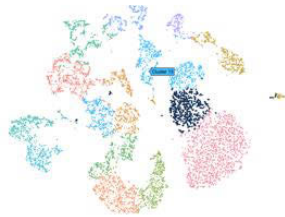
Adipose/Fat	Adrenal	Bone	Brain
Cartilage	Colon	Endothelial	Epithelial
Eye	Heart	Intestine	Kidney
Liver	Lung	Lymph nodes	Mammary
Miscellaneous	Muscle	Neural	Pancreas
Parotid	Pituitary	Prostate	Reproductive
Scales	Skin	Spleen	Stem
Thymus	Thyroid/Parathyroid	Tonsil	Tumor

<https://www.worthington-biochem.com/tissuedissociation/>

General Cell Handling Recommendations

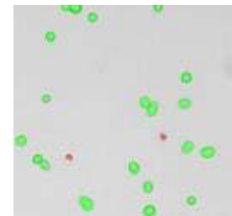
Analysis of Single Cell Transcriptomes

- Requires a fully dissociated, **single cell suspension**.
- Minimizing the presence of cellular aggregates, dead cells, non-cellular nucleic acids and potential inhibitors of reverse transcription is critical to obtaining high quality data.
- Suspension cell lines, bead-enriched and flow-sorted cells can be used directly after washing



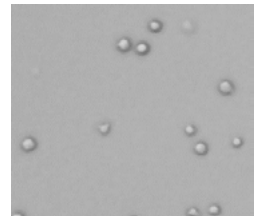
Importance of Input Cell Quality

- Ideally, input cell suspensions should contain more than **90% viable cells**.
- **The presence of a high fraction of non-viable or dying cells may decrease recovery.**
 - **The presence of ambient RNA and cellular debris may impact application performance and negatively impact quality metrics reported by Cell Ranger.**



Cell Handling

- It is important to **treat cells *gently*** to minimize cell lysis and loss:
 - **When cells lyse, the released ambient mRNA will contaminate other GEMs**
 - Wash cells twice using a wide-bore pipette tip to remove ambient RNA and contaminants.
 - Wash and resuspend in PBS + 0.04% non-acetylated BSA to minimize cell loss during handling.



General Cell Handling Recommendations

Debris/Aggregate Removal

- Use a cell strainer to remove aggregates or debris from washed cells
- The presence of cell aggregates, debris and/or fibers can result in inaccurate cell counts
- GEM generation occurs in microfluidic channels that are narrower than the typical human hair (i.e. $< 100\ \mu\text{m}$) and the presence of cell debris or large aggregates may clog or wet the chip



Cell Counting

- Quantitate cells accurately before loading into the system
 - Approximately 65% loaded cells will be recovered
 - To maximize the likelihood of achieving the desired recovery target, the optimal input cell concentration is 700-1200 cells/ μl
 - Recommended range: 500 to 10,000 recovered cells
 - Under- or over-loading may impact application performance

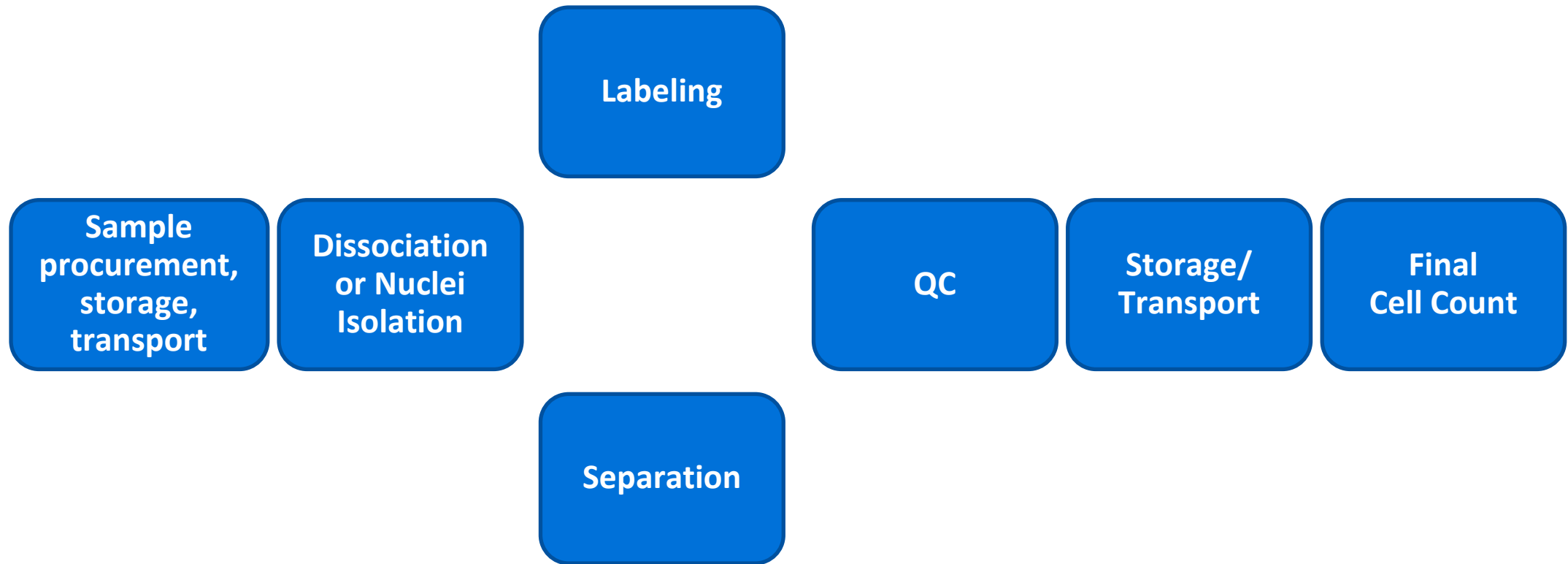


Storage of Single Cell Suspensions

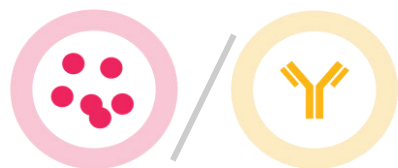
- Cell suspensions should always be kept on ice and where possible proceed with cell loading immediately after sample preparation
 - Ideally incubation time should be kept to a minimum ($< 30\ \text{min}$)
- Some cell types are more fragile and cell viability may decrease significantly if not processed and loaded immediately



It's a Workflow. A Set of Decisions.



Choosing a Single Cell Assay



Gene Expression

Provides **transcriptome +/- immune receptor profiling**

Considerations:

- Interested in **cellular mRNA**?
- Interested in **feature barcode** or **cellplexing**?
- Interested in **targeted gene** expression?
- Interested in **automation**?
- High sensitivity



Multiome ATAC + Gene Expression

Provides nuclear transcriptome with **paired** chromatin accessibility profiling

Considerations:

- Interested in multimodal cell phenotyping?
- Pair gene expression with regulatory activity?
- **Limited sample** type?
- Interested in **nuclear mRNA** only?



ATAC

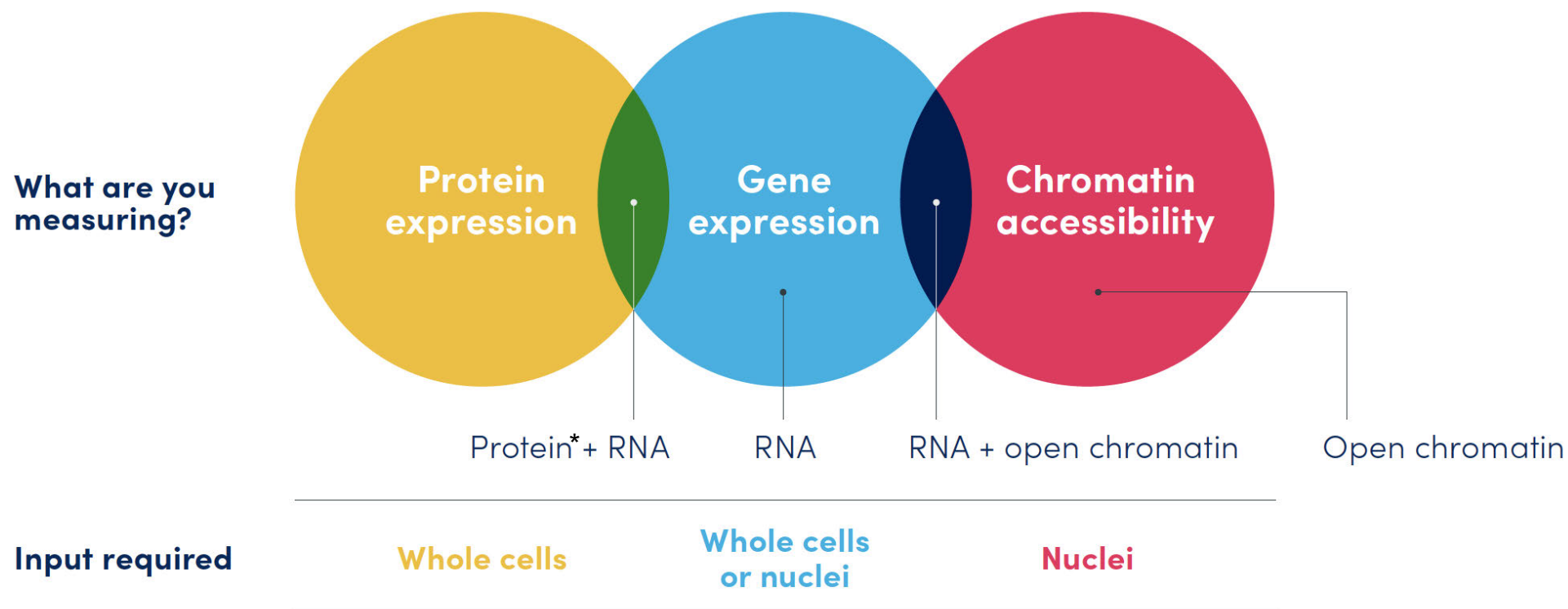
Provides **chromatin** accessibility profiling

Considerations:

- Interested in **open chromatin & TF binding** only?
- Sample types with **unknown or low mRNA integrity**?
- Cost sensitive?

Integrate data with third party tools using Single Cell Multiome ATAC + Gene Expression as bridge

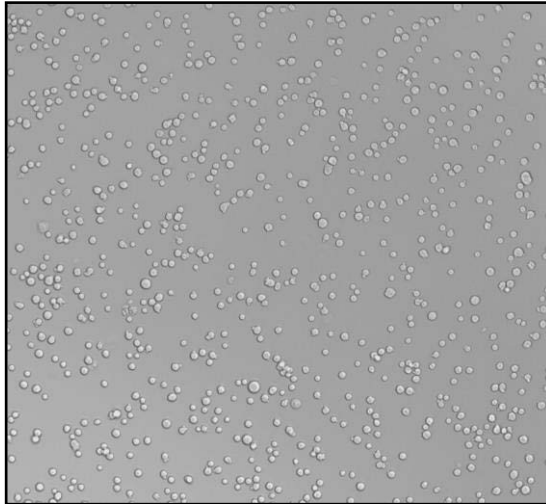
Different assays require different input materials



**Cell Surface protein*

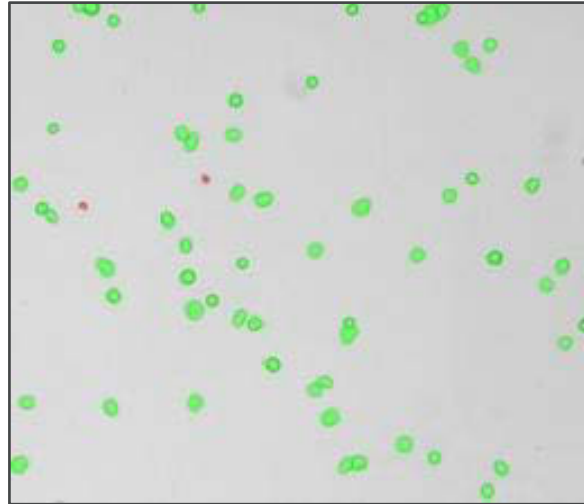
Quality is Critical

Clean



- Aggregates/clumps
- Subcellular debris
- Free-floating RNA/DNA

Healthy



- Biological decomposition
- RNA leakage (background)
- RNA degradation (signal)

Intact



- Physical decomposition
- RNA leakage (background)
- RNA degradation (signal)

Goal is to minimize

Sample Procurement, Storage, and Transport


SCIENTIFIC REPORTS

RESEARCH


Open Access



Cryopreservation of human cancers conserves tumour heterogeneity for single-cell multi-omics analysis

Sunny Z. Wu^{1,2} , Daniel L. Roden^{1,2}, Ghamdan Al-Eryani^{1,2}, Nenad Bartonicek^{1,2}, Kate Harvey¹, Aurélie S. Cazet^{1,2}, Chia-Ling Chan^{1,3}, Simon Junankar^{1,2}, Mun N. Hui^{1,4}, Ewan A. Millar^{5,6,7}, Julia Beretov^{5,8}, Lisa Horvath^{1,4,9}, Anthony M. Joshua^{1,10}, Phillip Stricker¹⁰, James S. Wilmott^{11,12}, Camelia Quek^{11,12}, Georgina V. Long^{11,12,13}, Richard A. Scolyer^{11,12,14} , Bertrand Z. Yeung¹⁵, Davendra Segara¹⁰, Cindy Mak⁴, Sanjay Warrier^{16,17}, Joseph E. Powell^{3,18}, Sandra O'Toole^{1,2}, Elgene Lim^{1,2,10} and Alexander Swarbrick^{1,2*} 

Elena Der

Olivier Clement¹, Rebecca K. Simmons^{1,2}, Ryan Lister^{1,2} and Alistair R. R. Forrest¹ 

Received: 1 March 2020

Accepted: 4 July 2020

Published online: 23 July 2020

RESE

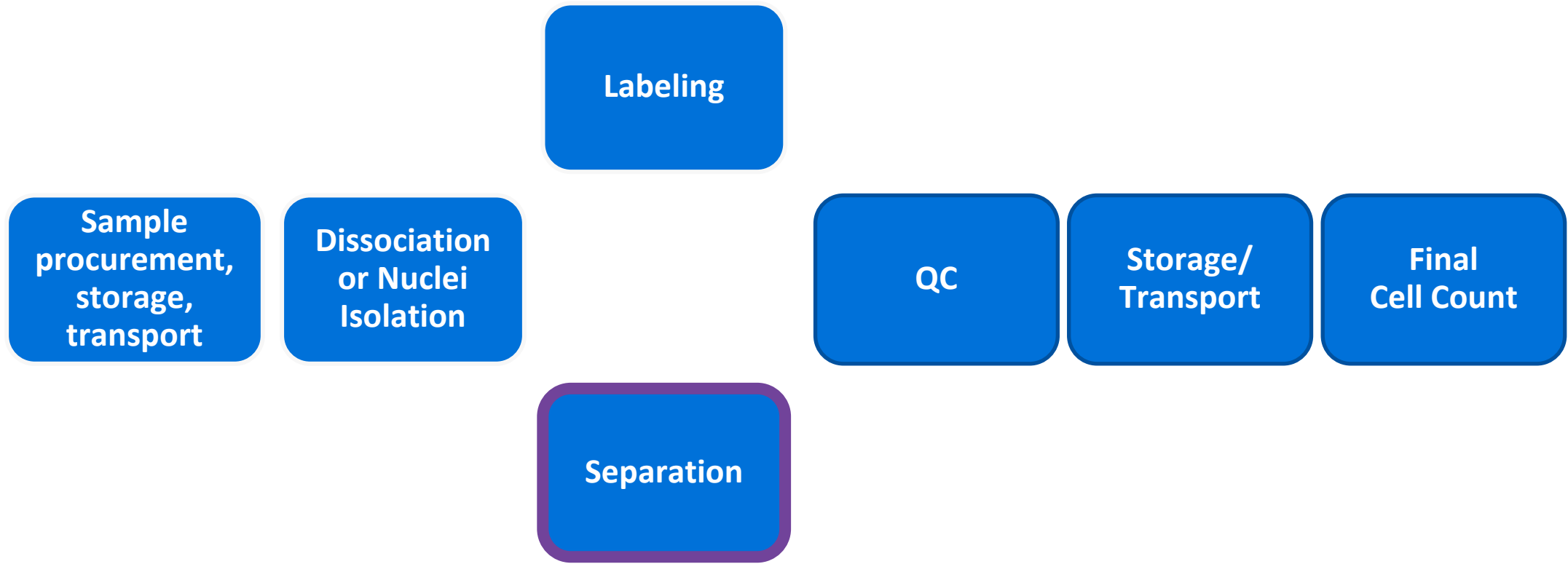
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Detailed decisions: Single Cell GEX sample separation



Sample Separation

Separate intact cells and nuclei from

- Aggregates/clumps
- Debris
- Free-floating mRNA
- Dead Cells
- Enrichment/Depletion

Challenges with separation

- Samples are fragile
- Physical stress
- Buffers
- Time
- Yield

Want the minimum handling necessary. Maintain sample integrity.

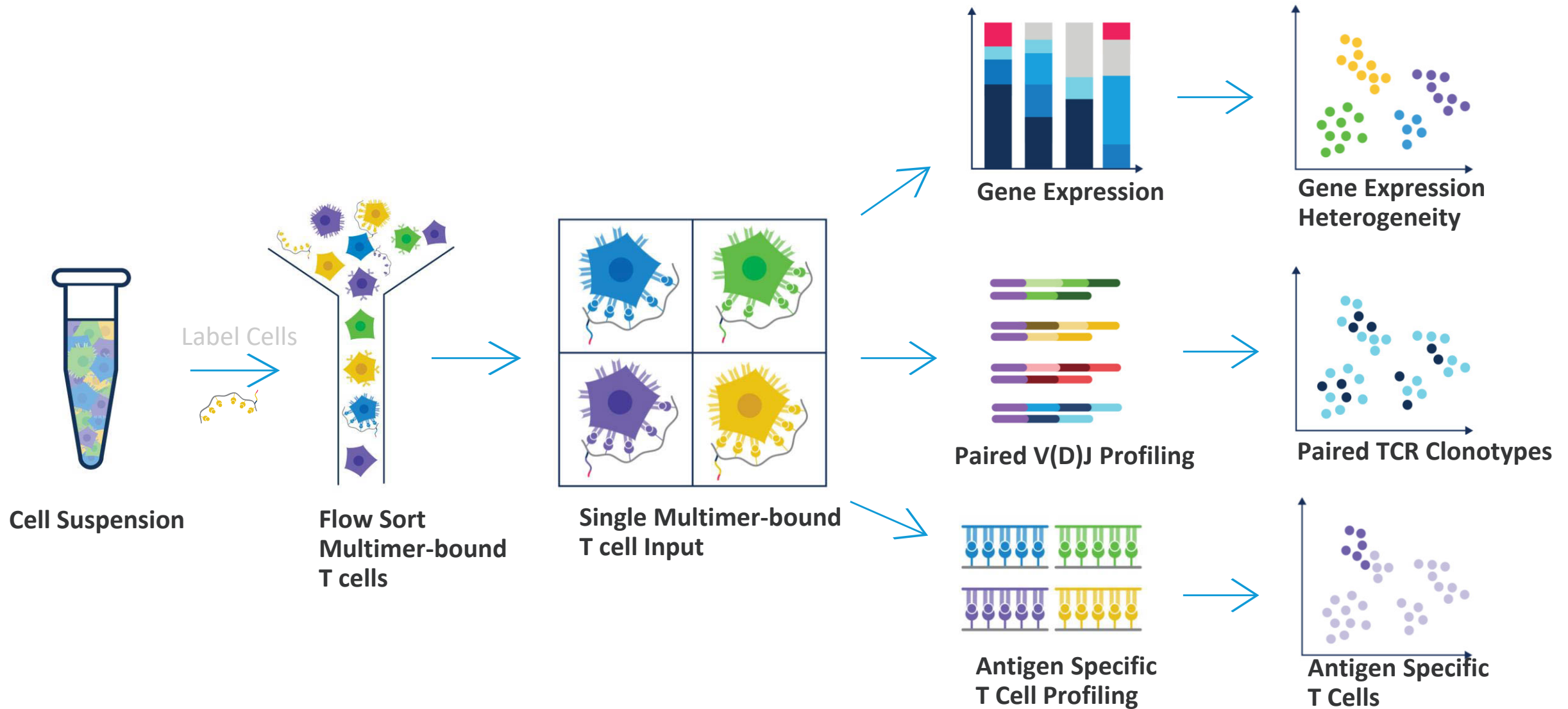


Separation

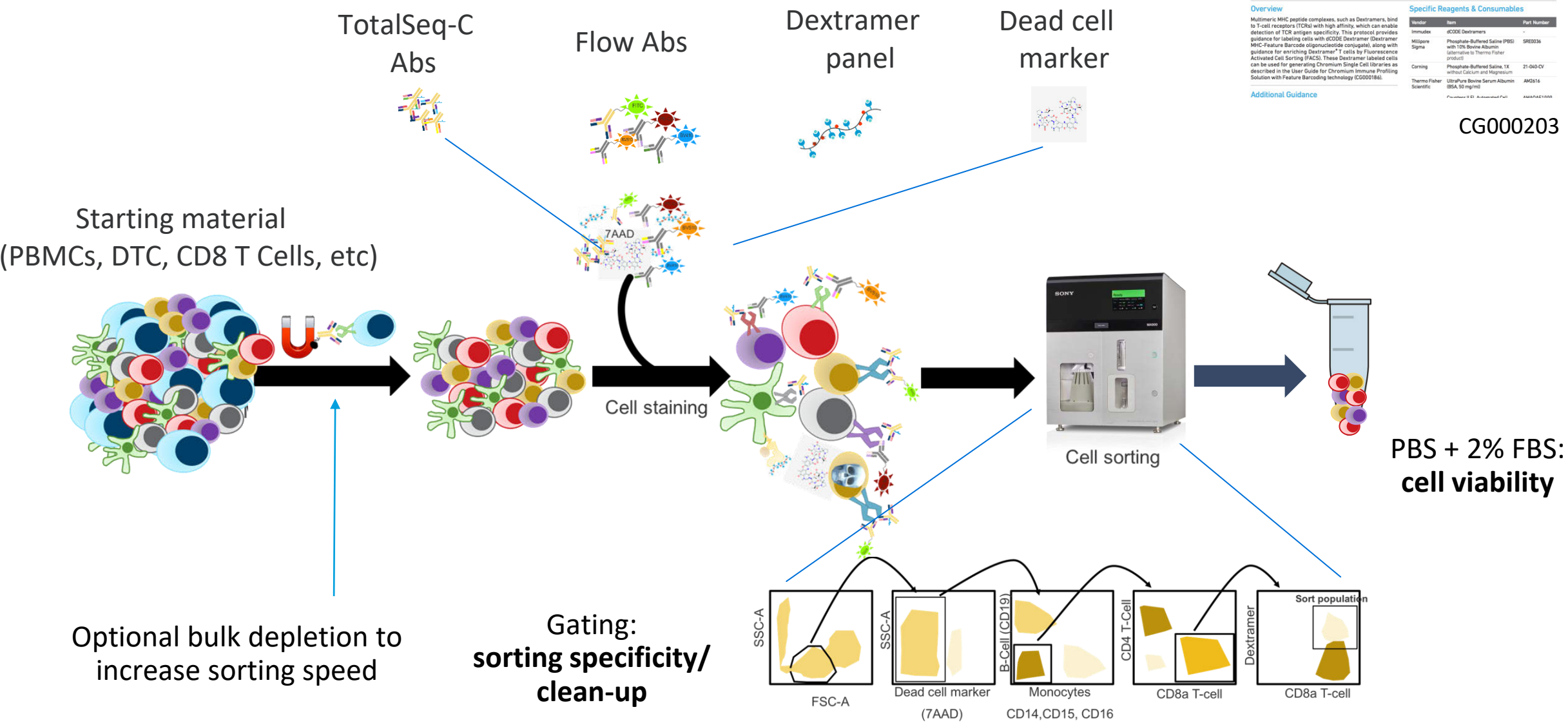
Basic Methods for Sample Separation

Method	Thorough centrifugation (e.g. 3x with PBS + 0.04% BSA)	Gentle centrifugation (e.g. 1x with media)	Magnetic beads	Density Gradient	FACS
10x Protocol Example	PBMC (CG000039)	Cell Prep Guide (CG000053)	Dead Cell Removal (CG000093)	Nuclei Isolation (CG000124)	Customer Developed Protocol (Martelotto)
Sample Size	Abundant	Limited	Abundant	Abundant	Limited
Benefits	Thorough	Gentle	Specific, easily accessible, scalable	Removes Debris	Versatile, quick
Possible Challenges	Yield, Harsh	Less thorough	Yield	Yield, Harsh, Time	Expensive, Harsh

Reveal Antigen Specificity with Feature Barcoding Technology



Sample Prep Workflow: Rare Cell Population



DEMONSTRATED PROTOCOL

CG000203 • Rev A

Cell Labeling with Dextramers for Single Cell RNA Sequencing Protocols with Feature Barcoding technology

Overview

Multimeric MHC peptide complexes, such as Dextramers, bind to T-cell receptors (TCRs) with high affinity, which can enable detection of TCR antigen specificity. This protocol provides guidance for labeling cells with dCODE Dextramer (Dextramer MHC-Feature Barcode oligonucleotide conjugate), along with guidance for enriching Dextramer⁺ T cells by Fluorescence Activated Cell Sorting (FACS). These Dextramer labeled cells can be used for generating Chromium Single Cell libraries as described in the User Guide for Chromium Immune Profiling Solution with Feature Barcoding technology (CG000186).

Specific Reagents & Consumables

Vendor	Item	Part Number
Immudex	dCODE Dextramers	-
Millipore Sigma	Phosphate-Buffered Saline (PBS) with 10% Bovine Albumin (alternative to Thermo Fisher product)	SRE0036
Corning	Phosphate-Buffered Saline, 1X without Calcium and Magnesium	21-040-CV
Thermo Fisher Scientific	UltraPure Bovine Serum Albumin (BSA, 50 mg/ml)	AM0216

Additional Guidance

For more information, see the User Guide for Chromium Immune Profiling Solution with Feature Barcoding technology (CG000186).

CG000203

Summary of Key Lessons

When our standard guidance isn't applicable:

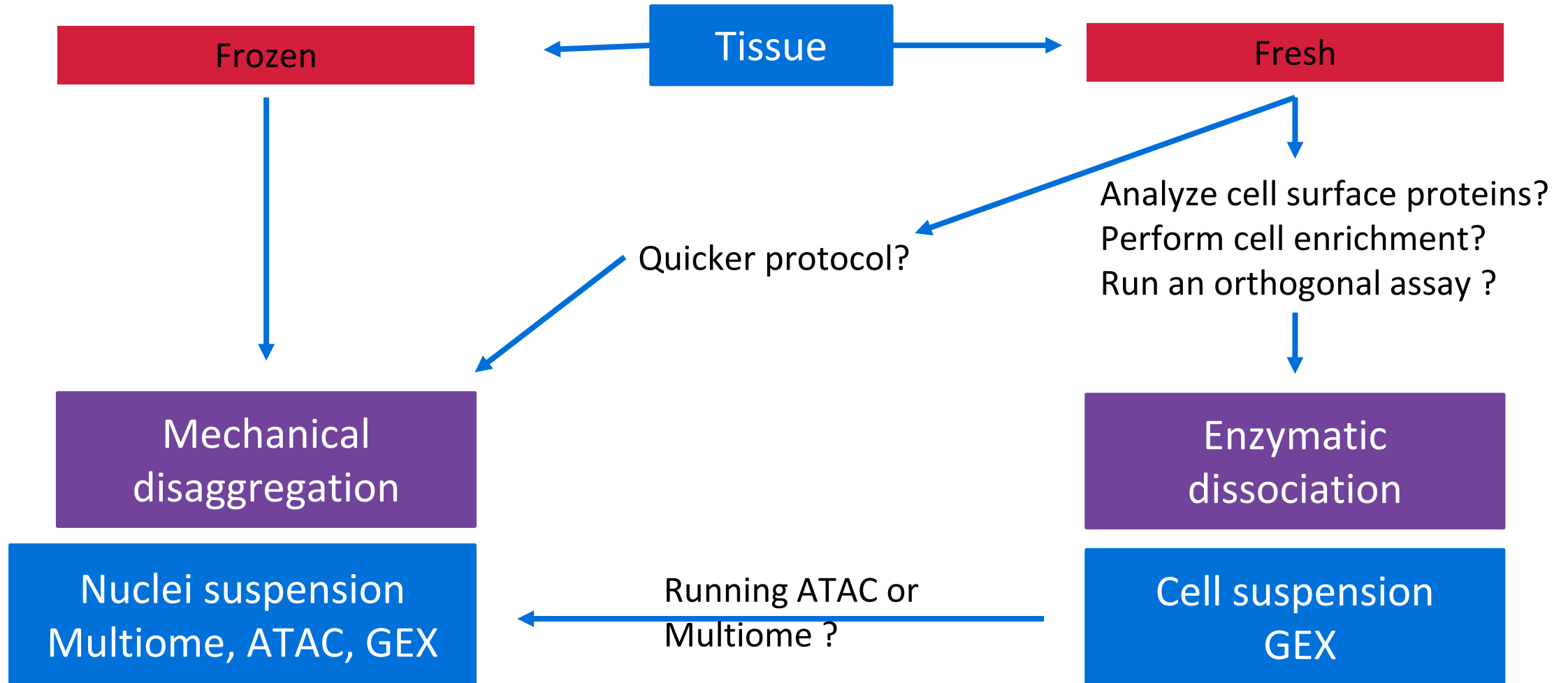
- Treat cells gently and minimize decomposition
 - Use gentl(er) lysis conditions
 - Reduce wash steps*
 - Use a swinging bucket centrifuge
 - Keep cells in media + FBS instead of PBS**
- Work quickly
 - Consider sorting, it is a versatile tool for sample prep
 - Minimize unnecessary handling steps
- *Consider the benefits and drawbacks of every different technique*

*Cell surface protein analysis requires thorough washing

**ATAC has specific buffer formulation

Isolation of Nuclei for Single Cell Sequencing

Prepare a Cell or Nuclei Suspension



Why Use Nuclei?

A clean, viable single cell suspension is necessary for optimal results in scRNA sequencing. However, there are times when getting a good cell suspension is difficult and nuclei is an alternative option.

- **When cells are large and exceed the limits for the microfluidic chip**
 - Hepatocytes
 - Neurons with significant extensions
- **When cells are of a challenging shape**
 - Cardiomyocytes
- **When cells are difficult to get into a single cell suspension**
 - Sample contains a lot of debris
 - Neurons are highly interconnected and may not efficiently dissociate into single cells after enzymatic treatment
 - Dissociation-resistant tissue samples such as complex tissues/ organs where nuclei (but not whole intact cells) can be isolated

Why Use Nuclei?

- ***Possible* solution for archival (cryopreserved) or damaged samples in which the cell wall is breaking down**
 - Laser capture microdissection will physically damage whole cells (cell wall)
 - Nuclei isolation will not rescue damaged cells that are already dying or undergoing apoptosis
- ***Possible* solution for experiments aiming to reveal molecular genetic regulatory mechanisms specific to the nucleus**
- **Sample types that have a cell wall that does not lyse in our assay**
 - Various plants, yeast
- **For ATAC and Multiome**

General Handling Recommendations

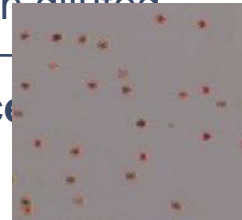
Starting Sample Requirements

- Tissues or cell suspensions
- Dissociate tissues when possible, some tissues will require going straight into nuclei isolation
- If starting with low viability cell suspension, sorting prior to nuclei isolation may help reduce ambient DNA and cellular debris
 - **Sorting after nuclei isolation is not recommended as it may damage nuclear membrane**



Nuclei Isolation

- Refer to Demonstrated Protocols for Nuclei Isolation for Single Cell ATAC Sequencing:
 - Nuclei Isolation from mouse brain tissue
 - Nuclei Isolation from cell lines and PBMCs
 - **Isolation of Nuclei for Single Cell RNA Sequencing demonstrated protocol decreases single cell ATAC assay performance**
- Resuspend nuclei in diluted Nuclei Buffer (1X)—**assay performance**



Nuclei Isolation

- It is important to **treat nuclei gently** to minimize lysis and loss
- Count Nuclei using Countess and trypan blue, ideal viability should be <5% live
 - Counting may also be done using ethidium homodimer and fluorescence microscope or Countess II FL.
- Visualization under the microscope may give further indications of nuclear membrane quality



Demonstrated Protocol Available from 10x Genomics

Nuclei Isolation for Single Cell ATAC Sequencing (From Cell Lines and PBMCs)

- Demonstrated protocol includes recommendations/tips for preparing nuclei from PBMCs and from cell lines (GM12878:EL4 mix), fresh and cryopreserved
- Low sample input protocol in appendix for limited samples
- Protocol can be adapted for other cell types with optimization
- Demonstrated Protocol is available on 10x support website
- **Note: Not all demonstrated protocols on our website will be compatible with the Chromium Single Cell ATAC Solution**

CG000169 • Rev C

DEMONSTRATED PROTOCOL

Nuclei Isolation for Single Cell ATAC Sequencing

Overview

This protocol outlines how to isolate, wash, and count nuclei suspensions for use with the Chromium Single Cell ATAC Solution. Cryopreserved primary cells (PBMCs) and cell lines (GM12878 cells; EL4 cells) were used to develop this protocol. PBMCs were cryopreserved in IMDM + 40% FBS + 15% DMSO. Cell lines were cryopreserved in RPMI + 15% FBS + 5% DMSO. Optimization of some protocol steps (e.g. lysis time, centrifugation speed/time and filtration steps) may be needed based on cell type.

⚠️ The recommended buffer compositions, final nuclei suspension concentration, and the wash step guidelines presented in this protocol for nuclei sample preparation are critical for optimal Chromium Single Cell ATAC Solution performance. Failure to adhere to these guidelines may result in compromised microfluidics chip operation.

Additional Guidance

Consult Demonstrated Protocol Cell Preparation Guide (Document CG000053) for Tips & Best Practices.

Cells carry potentially hazardous pathogens. Follow material supplier recommendations and local laboratory procedures and regulations for the safe handling, storage and disposal of biological materials.

Cell Sourcing

Cell Type	Species	Supplier
GM12878	Human	Coriell Institute
EL4	Mouse	ATCC
Normal Peripheral Blood MNC (PBMC)	Human	AlloCells

Preparation – Buffers

Diluted Nuclei Buffer	Stock	Final	1 ml
Nuclei Buffer (20X) (10x Genomics, PN-2000153*/ 2000207*)	20X	1X	50 µl
Nuclease-free Water	-	-	950 µl

See Appendix for DNase Treatment specific reagents & buffers

Wash Buffer

Prepare fresh, maintain at 4°C

	Stock	Final	2 ml
Tris-HCl (pH 7.4)	1M	10 mM	20 µl
NaCl	5M	10 mM	4 µl
MgCl ₂	1M	3 mM	6 µl
BSA	10%	1%	200 µl
Tween-20	10%	0.1%	20 µl
Nuclease-free Water	-	-	1.75 ml

Lysis Buffer

Prepare fresh, maintain at 4°C

	Stock	Final	2 ml
Tris-HCl (pH 7.4)	1M	10 mM	20 µl
NaCl	5M	10 mM	4 µl
MgCl ₂	1M	3 mM	3 µl
Tween-20	10%	0.1%	20 µl
Nonidet P40 Substitute (if using Sigma (74385) 100% solution, prepare a 10% stock)	10%	0.1%	20 µl
Digitonin (insoluble at 45°C to dissolve precipitate before use)	5%	0.01%	4 µl
BSA	10%	1%	200 µl
Nuclease-free Water	-	-	1.729 ml

Additional Buffers

RPMI + 10% FBS (maintain at 4°C, pre-warm at 37°C before use)

PBS + 0.04% BSA (maintain at 4°C)

Specific Reagents & Consumables

Vendor	Item	Part Number
10x Genomics	Nuclei Buffer*/20X Nuclei Buffer*	2000153/ 2000207
Thermo Fisher Scientific	Digitonin Tubes, 0.2 ml, flat cap tube**	BN20206 AB0620
Fisher Scientific	Sorvall Microtube Adapters**	76003750
Millipore- Sigma	Trizma Hydrochloride Solution, pH 7.4 Sodium Chloride Solution, 5M Magnesium Chloride Solution, 1M Nonidet P40 Substitute	T2194 59222C M1028 74385
Milleniy Biotech	MACS BSA Stock Solution	130-091-376
Bel-Art	Flowmi Cell Strainer, 40 µm	H13680-040

*Included in the Single Cell ATAC Library Kits

**ONLY for Low Cell Input Nuclei Isolation protocol

10x GENOMICS

10xGenomics.com Demonstrated Protocol – Nuclei Isolation for Single Cell ATAC Sequencing • Rev C

Nuclei Isolation for Single Cell ATAC Sequencing

Required Reagents and Buffer Composition


Reagents

Vendor	Item	Part Number
10x Genomics	Nuclei Buffer*/20X Nuclei Buffer*	2000153/ 2000207
Thermo Fisher Scientific	Digitonin Tubes, 0.2 ml, flat cap tube**	BN2006 AB0620
Fisher Scientific	Sorvall Microtube Adapters**	76003750
Millipore- Sigma	Trizma Hydrochloride Solution, pH 7.4 Sodium Chloride Solution, 5M Magnesium Chloride Solution, 1M Nonidet P40 Substitute	T2194 59222C M1028 74385
Miltenyi Biotec	MACS BSA Stock Solution	130-091-376
Bel-Art	Flowmi Cell Strainer, 40 µm	H13680-0040

*Included in the Single Cell ATAC Library Kits

**ONLY for the Low Cell Input Nuclei Isolation protocol

Buffers

Diluted Nuclei Buffer Maintain at 4°C	Stock	Final	1 ml
 Nuclei Buffer (20X) (10x Genomics, PN-2000153*/ 2000207*)	20X	1X	50 µl
Nuclease-free Water	-	-	950 µl
Wash Buffer Prepare fresh, maintain at 4°C	Stock	Final	2 ml
Tris-HCl (pH 7.4)	1M	10 mM	20 µl
NaCl	5M	10 mM	4 µl
MgCl ₂	1M	3 mM	6 µl
BSA	10%	1%	200 µl
Tween-20	10%	0.1%	20 µl
Nuclease-free Water	-	-	1.75 ml
Lysis Buffer Prepare fresh, maintain at 4°C	Stock	Final	2 ml
Tris-HCl (pH 7.4)	1 M	10 mM	20 µl
NaCl	5 M	10 mM	4 µl
MgCl ₂	1 M	3 mM	3 µl
Tween-20	10%	0.1%	20 µl
Nonidet P40 Substitute (if using Sigma (74385) 100% solution, prepare a 10% stock)	10%	0.1%	20 µl
Digitonin (incubate at 65°C to dissolve precipitate before use)	5%	0.01%	4 µl
BSA	10%	1%	200 µl
Nuclease-free Water	-	-	1.729 ml
Additional Buffers			
RPMI + 10% FBS (maintain at 4°C, pre-warm at 37°C before use)			
PBS + 0.04% BSA (maintain at 4°C)			

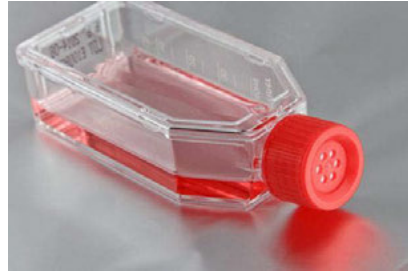


Supplied in Chromium Single Cell ATAC Library Kits. The tube contains enough to make ~32 ml of working dilution for final nuclei resuspension. Typical usage is up to 1 ml per sample.

Once prepared, maintain diluted Nuclei Buffer on ice while isolating nuclei.

Validated with Nuclei Isolated from Multiple Sample Types

Cell Lines



- Suspension: GM12878, A20, EL4, K562
- Adherent: A549

Primary Immune Cells



- Human Peripheral Blood Mononuclear Cells
- Human Bone Marrow Mononuclear Cells

Dissociated Primary Tissues



- Embryonic Mouse Brain Tissue
- Adult Mouse Brain Tissue
- Mouse Splenocytes

Nuclei isolation for Single Cell Multiome ATAC + Gene Expression sequencing

Lysis can be assessed using a cell viability stain

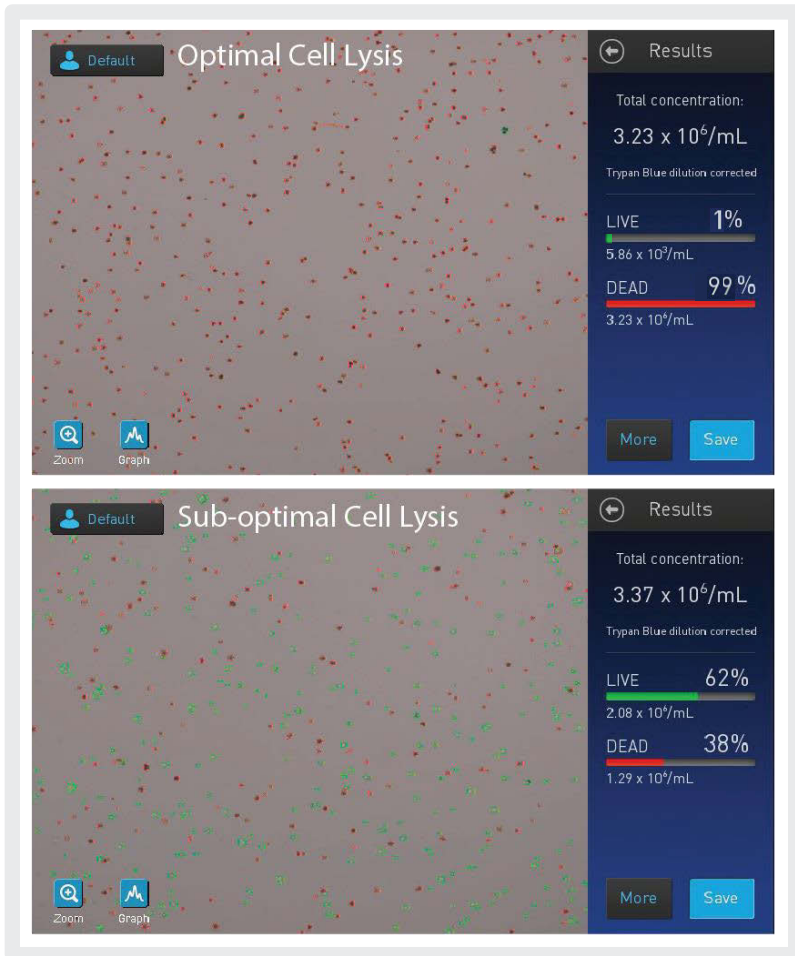
When counting look for:

- <5% Live Cells (or >95% dead)
 - Nuclei will stain as dead
 - Lysis time will be cell-type dependent
 - Lysis time course may be required to determine optimal lysis time
- Clean, clump free nuclei
 - Filtering may help break clumps and remove debris

This provides a yes/no answer as to whether the cell membrane was lysed. Resolution is enough to assess clumping and debris but may not be enough to evaluate nuclear membrane integrity

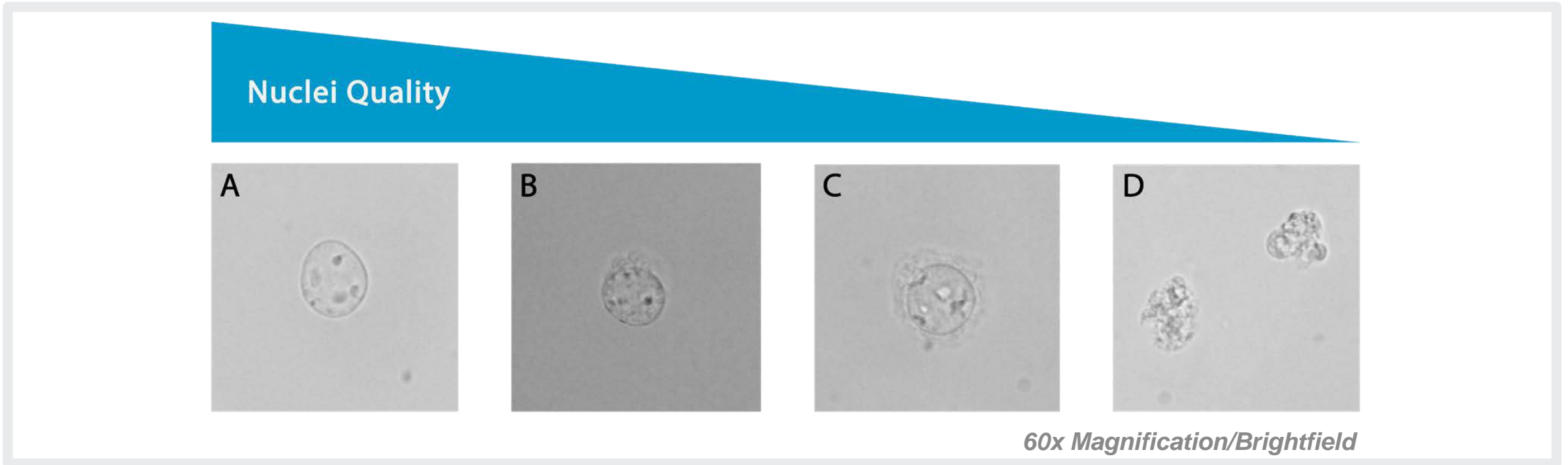


For hard to count cells (small size, lots of debris), use a fluorescent stain like Ethidium Homodimer-1. An automated counter with fluorescent capability is needed.



Nuclei isolation for Single Cell ATAC sequencing

Nuclear morphology can indicate nuclei quality



A: High-quality nuclei have well-resolved edges. Optimal quality for single cell ATAC libraries.

B: Mostly intact nuclei with minor evidence of blebbing. Quality single cell ATAC libraries can still be produced.

C: Nuclei with strong evidence of blebbing. **Proceed at your own risk.**

D: Nuclei are no longer intact. **Do not proceed!**

Data Review – Cells vs Nuclei

Gene Expression Levels Are Well Correlated Between Cells and Nuclei

nature
biotechnology

ANALYSIS

<https://doi.org/10.1038/s41587-020-0465-8>



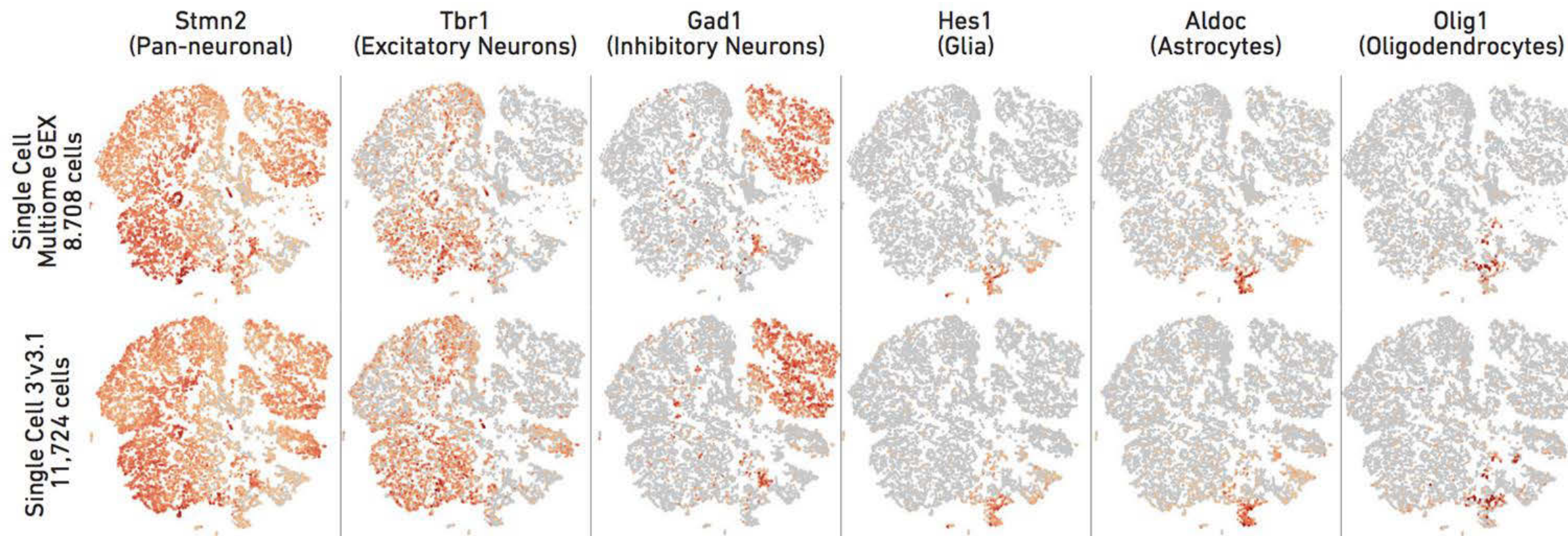
Systematic comparison of single-cell and single-nucleus RNA-sequencing methods

Jiarui Ding¹, Xian Adiconis^{1,9}, Sean K. Simmons^{1,9}, Monika S. Kowalczyk¹, Cynthia C. Hession¹, Nemanja D. Marjanovic¹, Travis K. Hughes^{1,2,3,4}, Marc H. Wadsworth^{1,2,3,4}, Tyler Burks¹, Lan T. Nguyen¹, John Y. H. Kwon¹, Boaz Barak⁵, William Ge ¹, Amanda J. Kedaigle ¹, Shaina Carroll^{1,2,3,4}, Shuqiang Li¹, Nir Hacohen^{1,6}, Orit Rozenblatt-Rosen¹, Alex K. Shalek ^{1,2,3,4}, Alexandra-Chloé Villani^{1,6,7}, Aviv Regev ^{1,4,8} and Joshua Z. Levin ¹ 

Cell type specific markers are conserved

Between Single Cell Multiome Gene Expression and 3' v3.1

Gene Expression

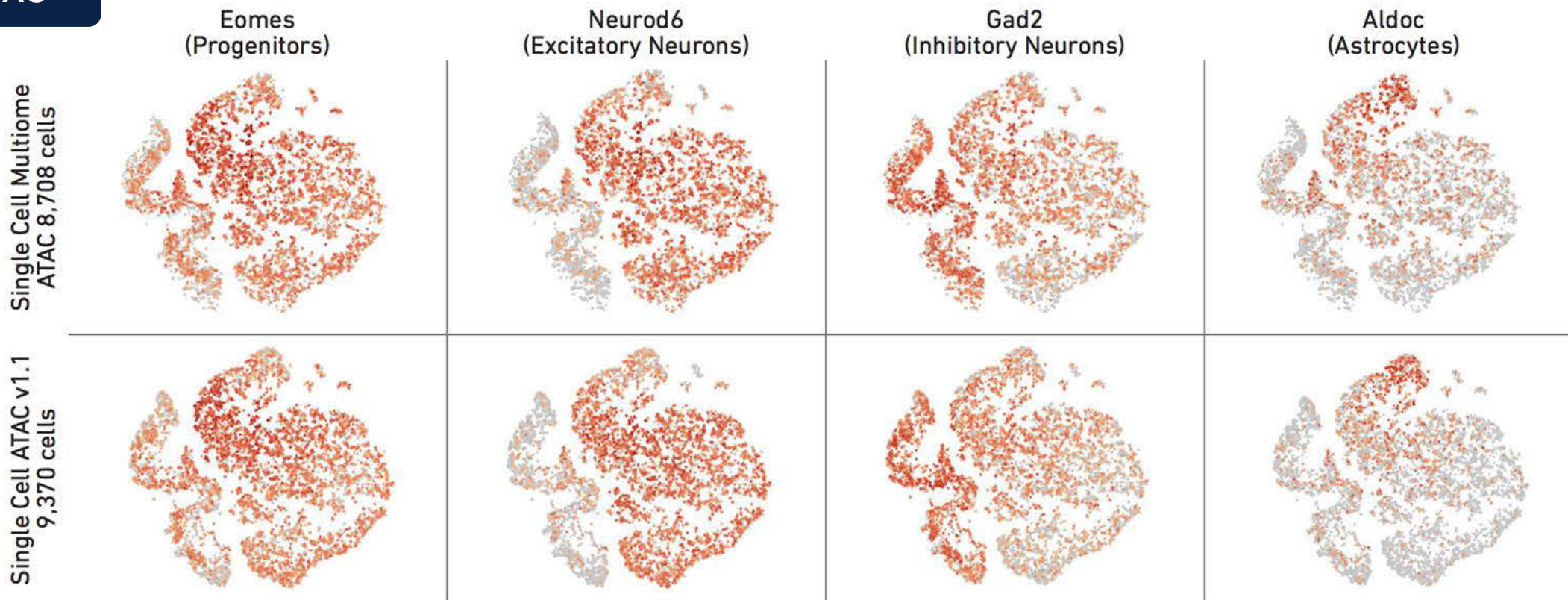


Mouse embryonic E18.5 brain nuclei

Cell type specific markers are conserved

Between Single Cell Multiome ATAC and ATAC v1.1

ATAC

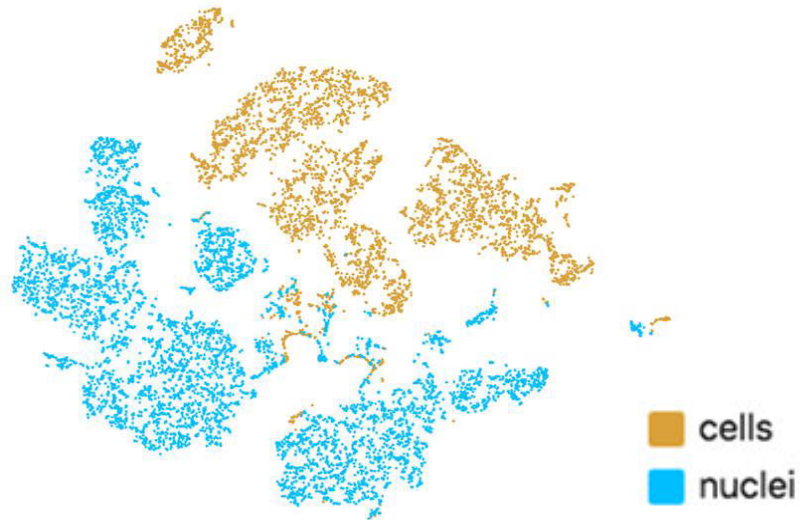


Mouse embryonic E18.5 brain nuclei

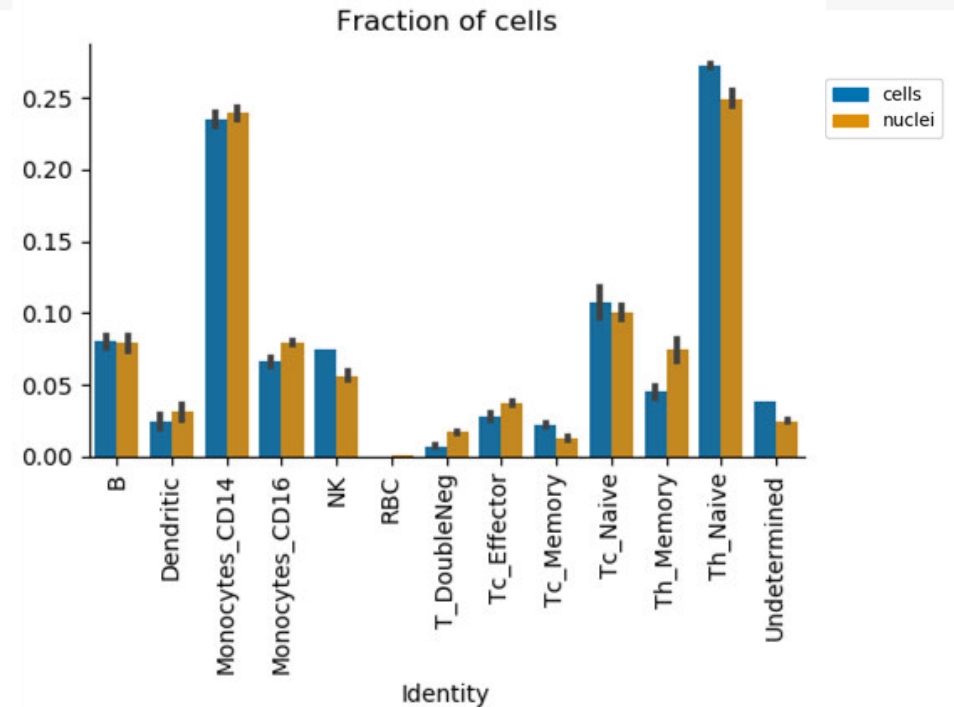
How does information from nuclei compare to cells?

Clusters do not overlap but biological information is conserved

PBMCs



- Cells run on SC3'v3.1 capture mostly mRNA
- Nuclei run on Multiome ATAC+GEX capture mostly pre-mRNA (unspliced mRNA)



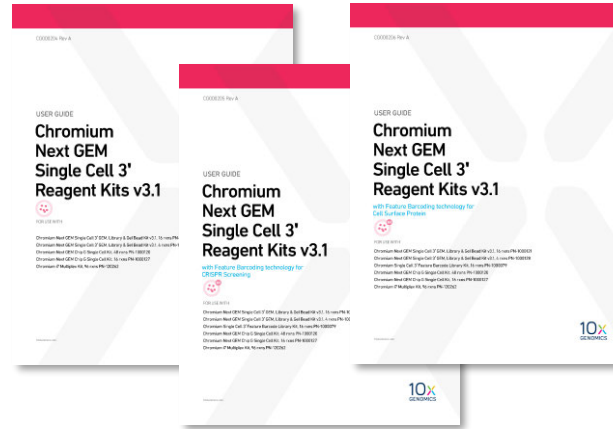
- Cell populations can still be identified

Resources

Demonstrated Protocols



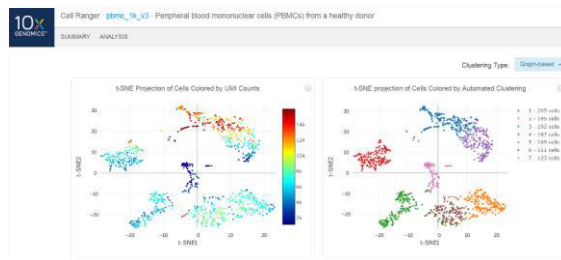
User Guides



Application Notes



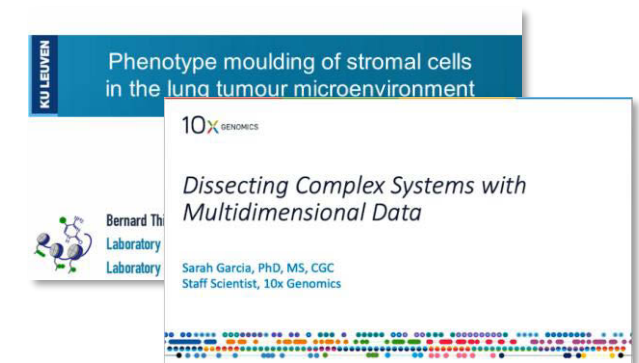
Public Data Sets



How-To Videos



Scientific Seminars



<https://support.10xgenomics.com/>
<https://www.10xgenomics.com/10x-university/>

Customer Developed Protocols

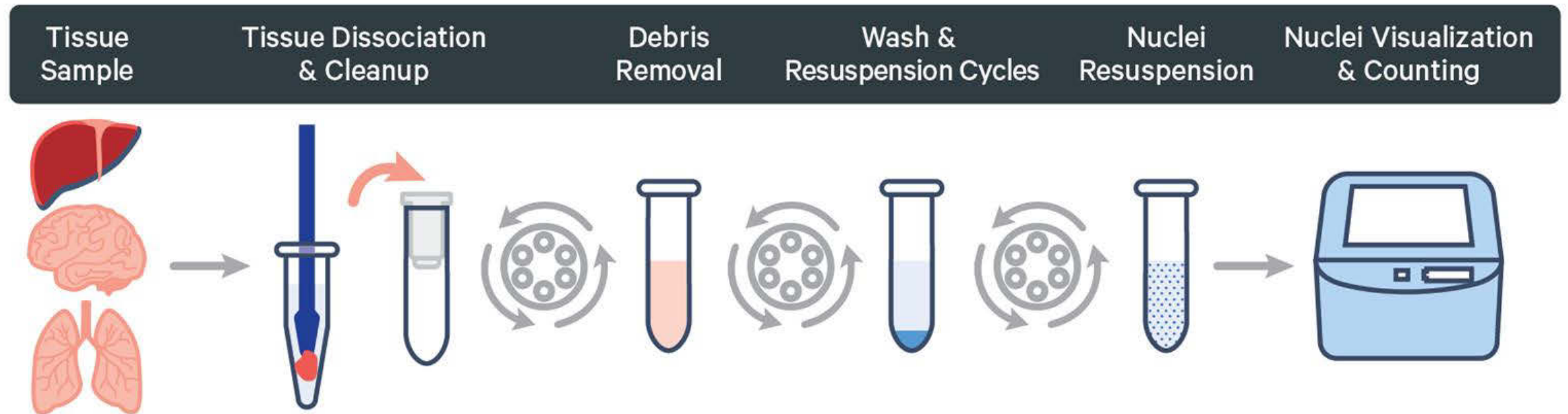
<https://community.10xgenomics.com/t5/Customer-Developed-Protocols/ct-p/customer-protocols>

1. [High Molecular Weight Genomic DNA Extraction from Grape Leaves](#) Contributed by: Xia Xu, Lance Cadle-Davidson
USDA-ARS, GGRU
2. [CTAB Protocol for Isolating DNA from Plant Tissue](#) Contributed by: Allen Van Deynze, Van Deynze Lab, UC Davis
3. [Cell dissociation and crypt isolation of the mouse small intestine](#) Contributed by: Aviv Regev, Regev Lab, Broad Institute
4. [Tissue dissociation and single cell preparation of breast cancer patient-derived xenografts](#) Contributed by: Ioannis Ragoussis and Morag Park
5. [Isolation of single cell suspensions from epidermis](#) Contributed by: Samuel Lukowski
6. [Generation of single cell suspension from E8.25 mouse embryos](#) Contributed by: Bertie Gottgens
7. [Preparation of non-myocyte cardiac single cell suspensions](#) Contributed by: Galen Squiers & Alex Pinto, Pinto Lab, The Jackson Laboratory, Bar Harbor
8. ['Frankenstein' protocol for nuclei isolation from fresh and frozen tissue](#) Contributed by: Luciano Martelotto, Ph.D., Melbourne, Centre for Cancer Research, Victorian Comprehensive Cancer Centre

Nuclei Isolation Kit



Streamlined sample preparation workflow



All you need is an hour of lab time, a benchtop centrifuge, and an interesting frozen sample!

Thank You from the 10x Team & our Collaborators

Agnieszka Ciesielska

Science & Technology Advisor

