

Single cell and single nucleus RNA sequencing experiments: a practical guide to design and analysis

Omics bootcamp I & II: Thursday 5/11/23 and Friday 5/12/23

Amelia Weber Hall, PhD

Gene Regulation Observatory

The Broad Institute of MIT and Harvard



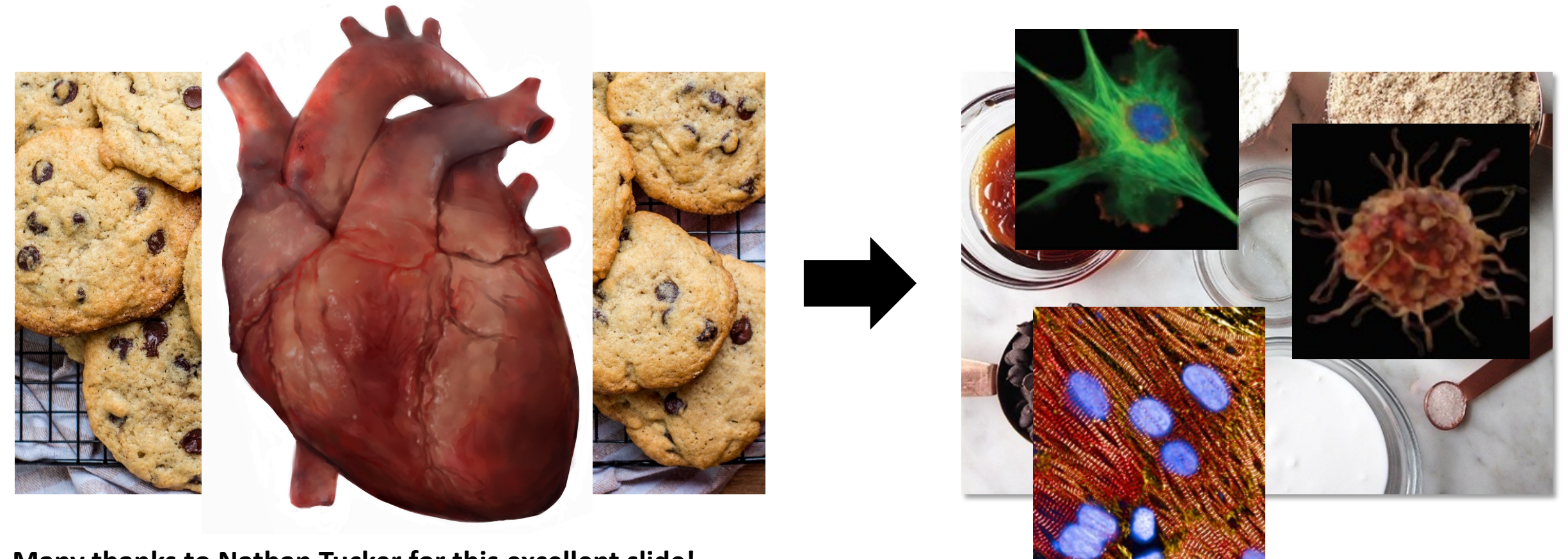
Vascular Discovery:
From Genes to Medicine 2023

Outline

- Single cell transcriptomics: a history
- A little intro to ATAC-seq and multi-omic assays
- Specific single cell studies of note in the cardiovascular system
- Experimental design best practices
 - Nuclei isolation methods
 - Tissue storage and validation of single cell findings
- Computational Analysis best practices
- Useful resources
- Limitations
- Brief preview of tomorrow's coding bootcamp

First, a question: what can we learn from single cell?

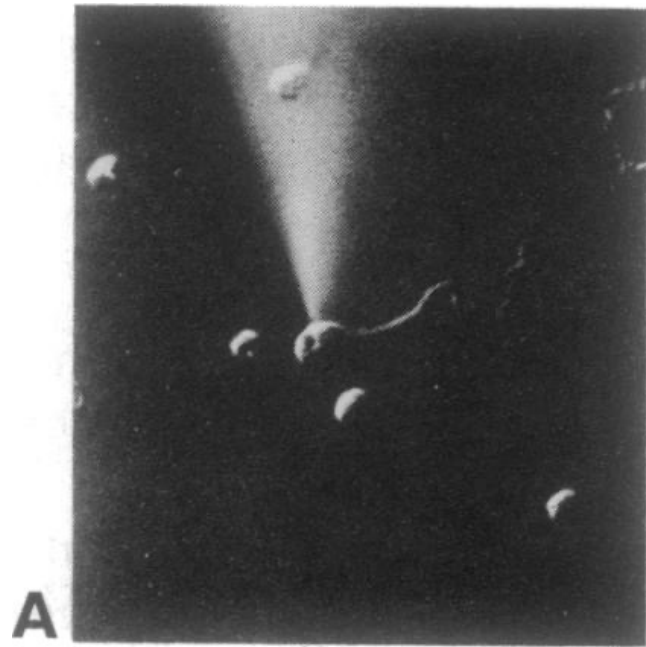
- **We are deconvoluting tissue into cellular subtypes**
- The goal is to learn more about what subtypes make up a complex and heterogeneous tissue



Many thanks to Nathan Tucker for this excellent slide!

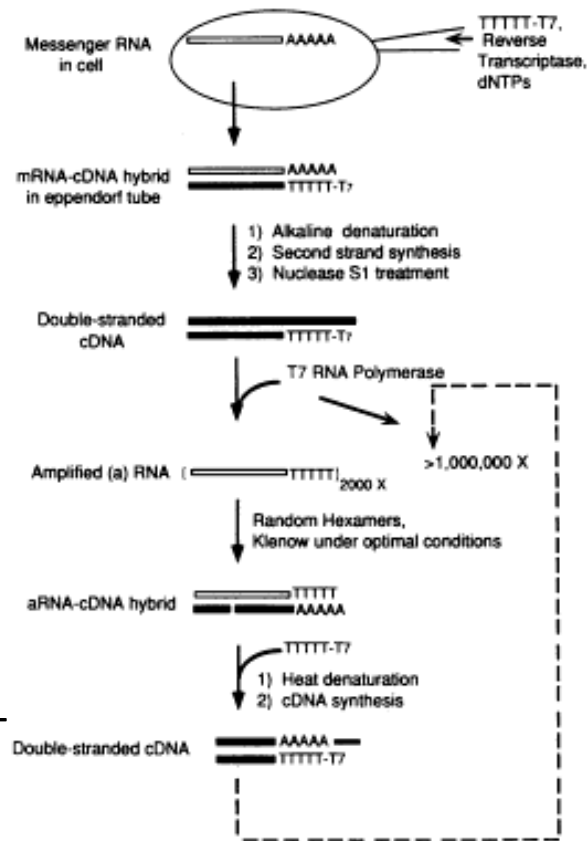
Single cell transcriptomics: a History

- Eberwine (1992, neurons) and Iscove (1990, hematopoietic cells)



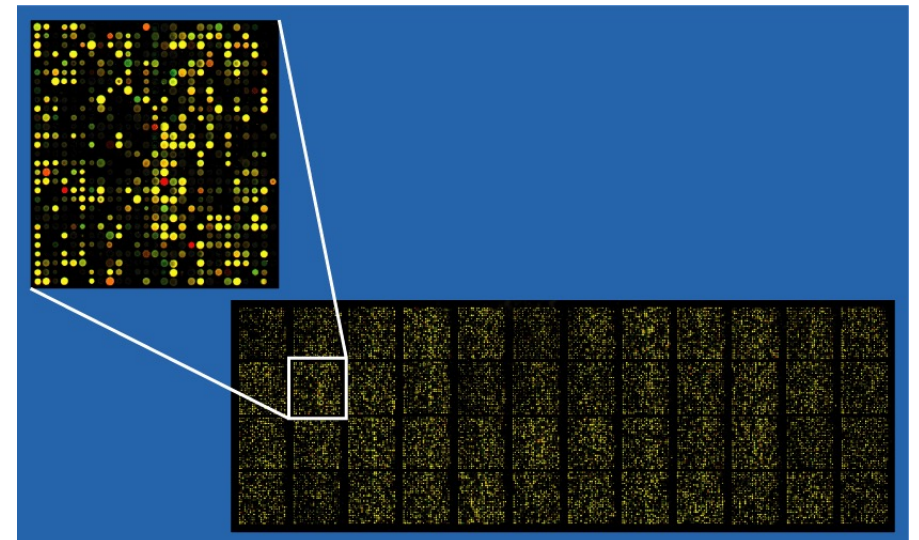
A Patch-clamp of a rat hippocampal pyramidal neuron to inject an oligo-dT Nucleotide, dNTPs and RT enzyme

Proc. Natl. Acad. Sci. USA 89 (1992)



Initially, cDNAs were cloned for further study.

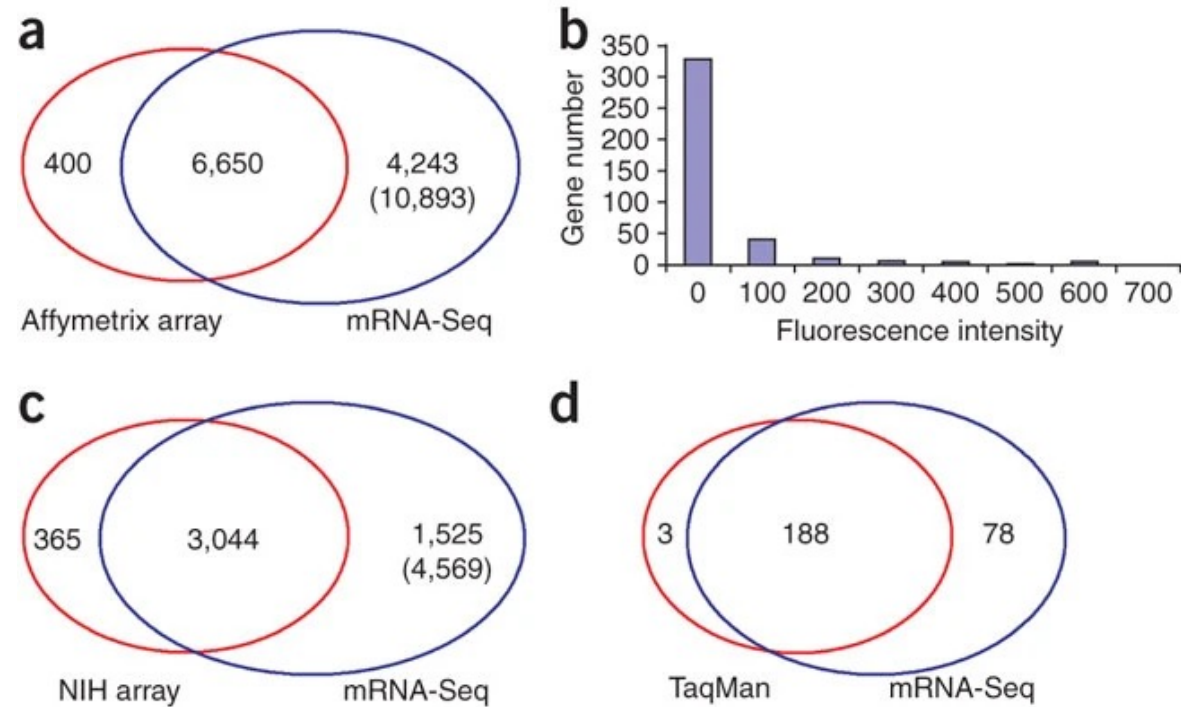
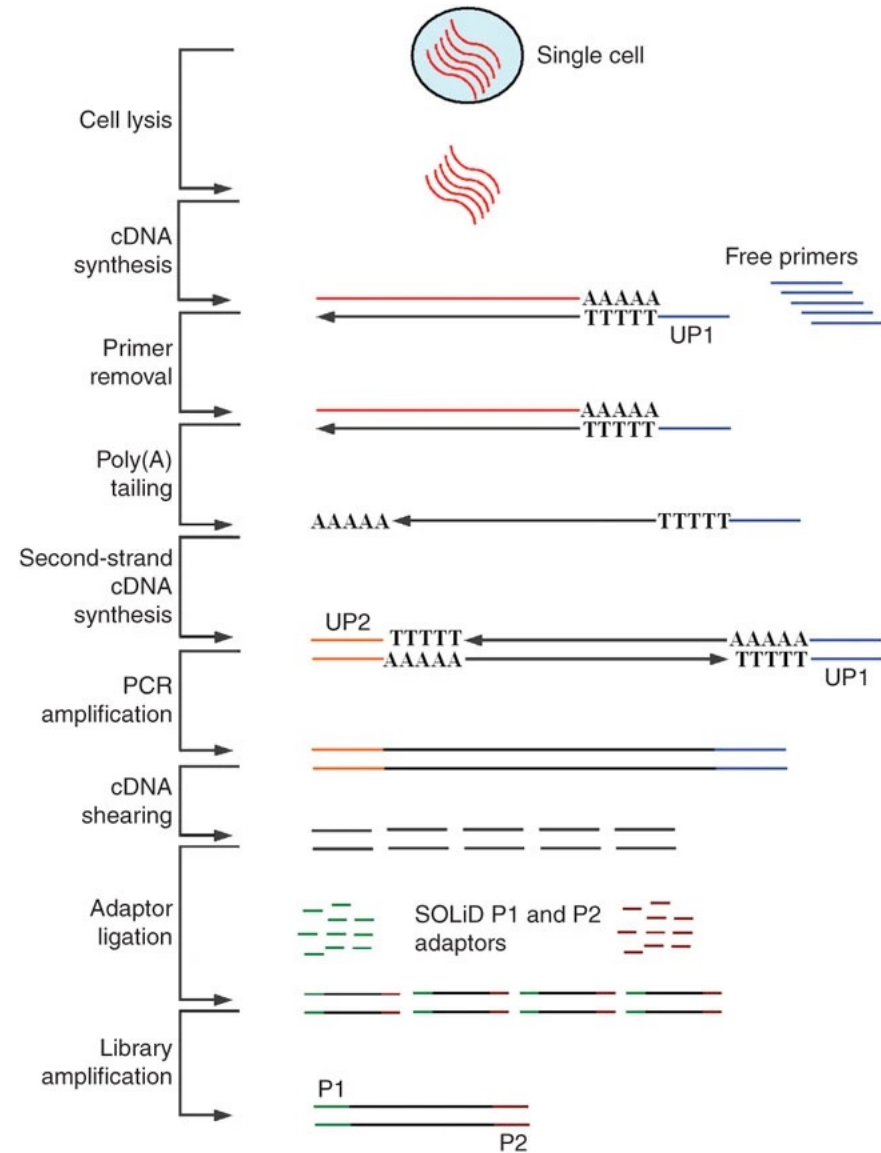
Starting in 1995, the advent of microarrays allowed for chip based study of the transcriptome using hybridization to an array of probes.



By Paphrag at English Wikipedia - Transferred from en.wikipedia to Commons., Public Domain,
<https://commons.wikimedia.org/w/index.php?curid=1612185>

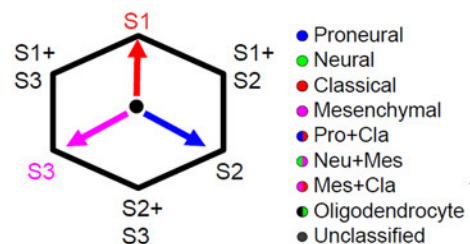
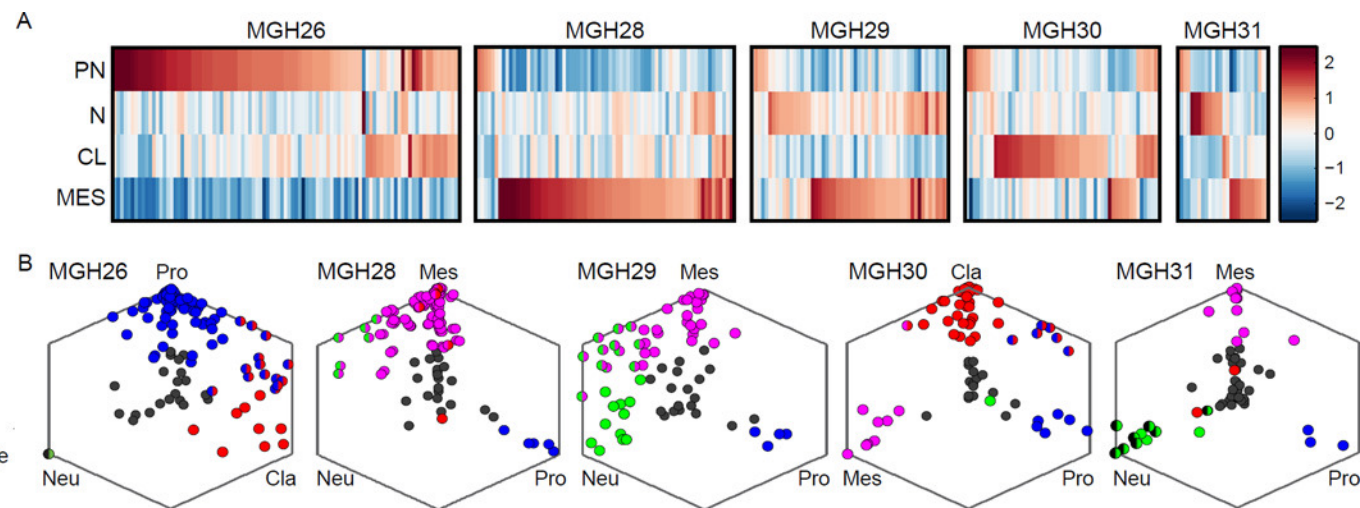
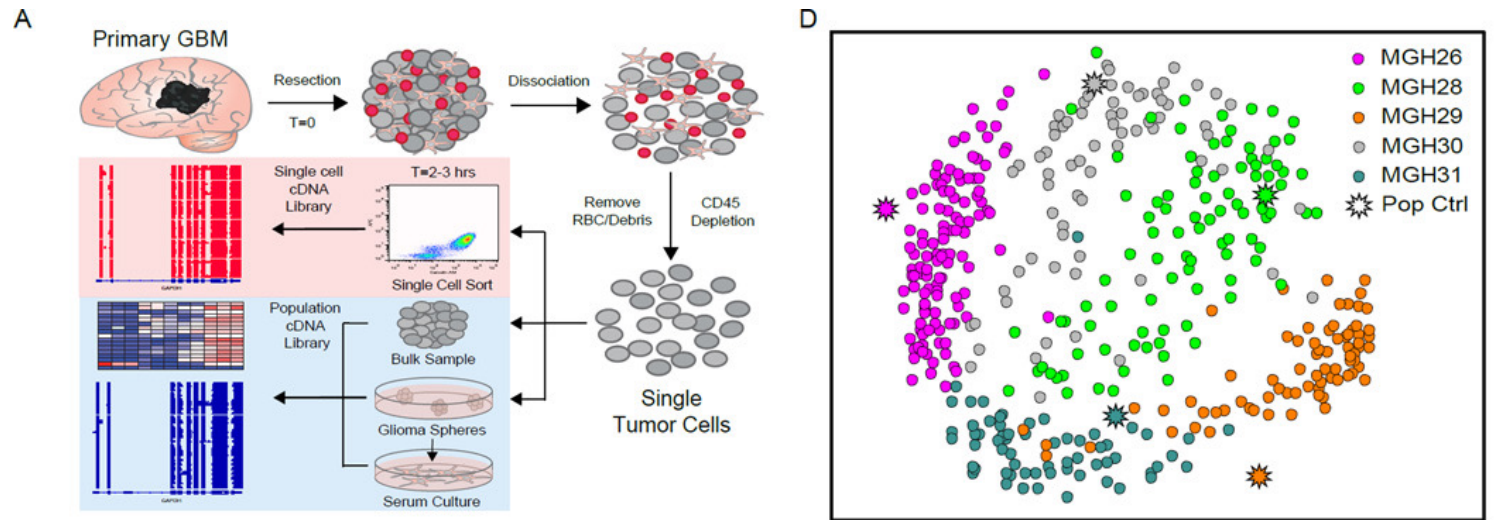
Single cell transcriptomics: a History

- Tang et al in 2009 were the first to use next generation sequencing on a single cell
 - Sequenced a single mouse blastomere
 - Can detect many more transcripts than using a microarray



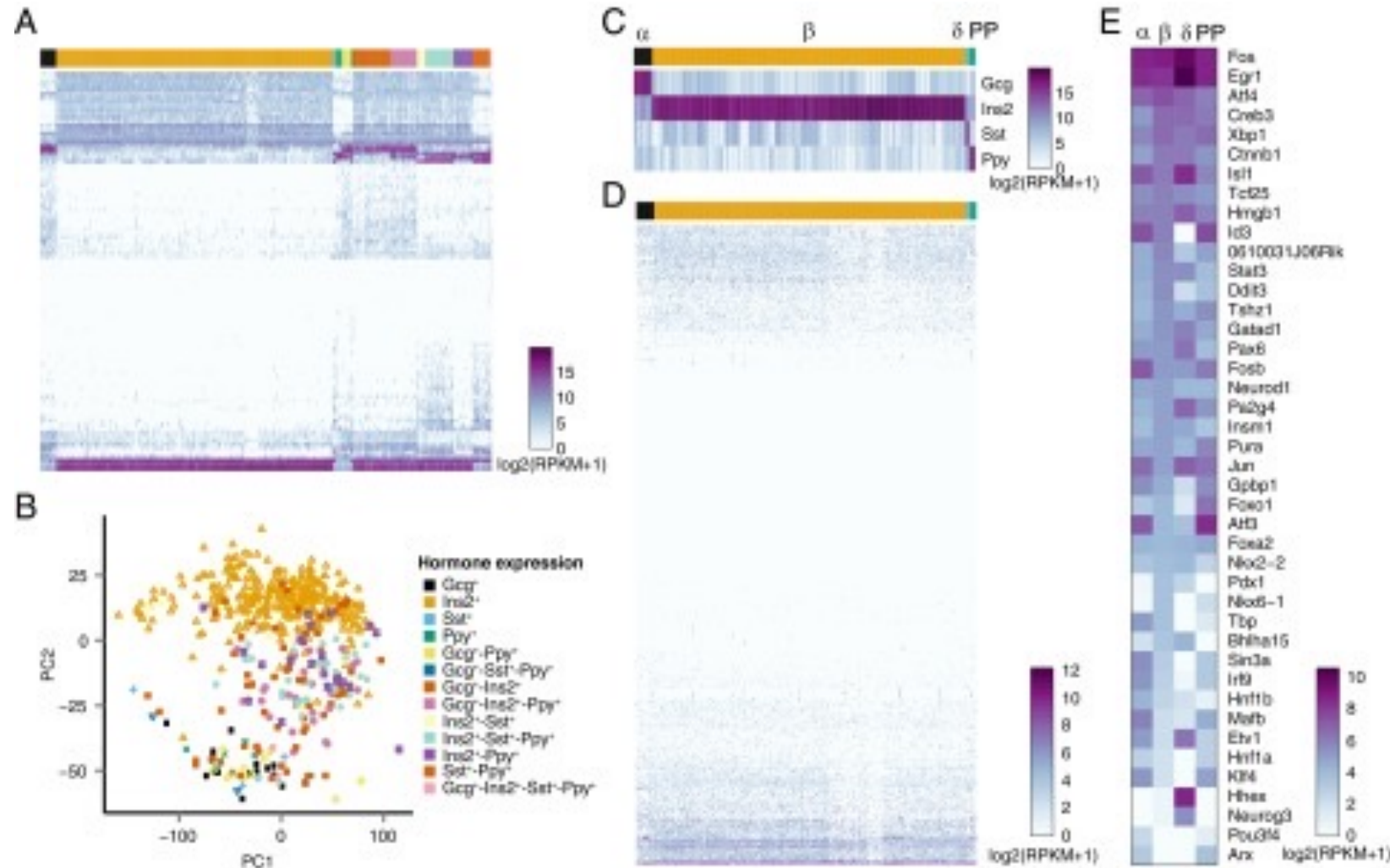
Single cell transcriptomics: a History

- Patel et al 2014 – sequenced 430 single cells from 5 primary human glioblastoma tumors
 - Glioblastoma has 3 major subtypes, defined by transcription
 - But each tumor contained cells that resembled all 3 subtypes (independent from bulk transcriptional analysis)
- SMART-seq to amplify mRNA



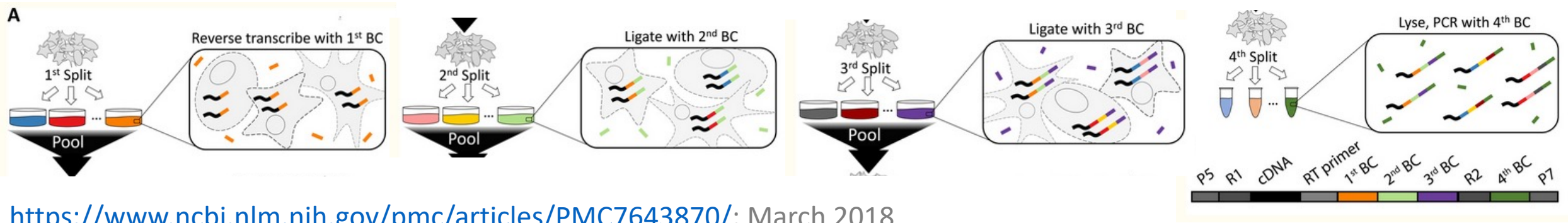
Single cell transcriptomics: a History

- Xin et al 2016 – sequencing of mouse pancreatic islets
- Used the Fluidigm C1 platform to isolate cells for single cell RNA-seq
- Sequenced 622 cells, lost close to 50% due to QC



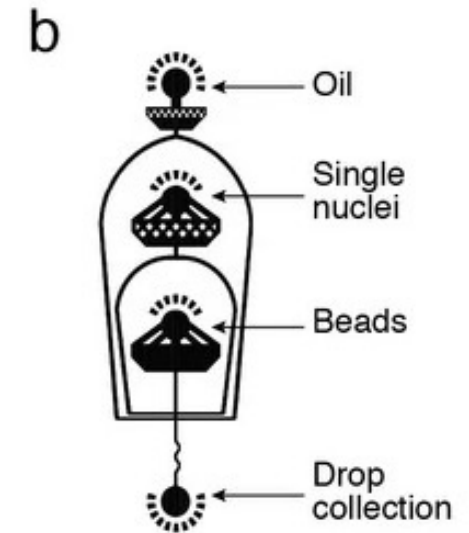
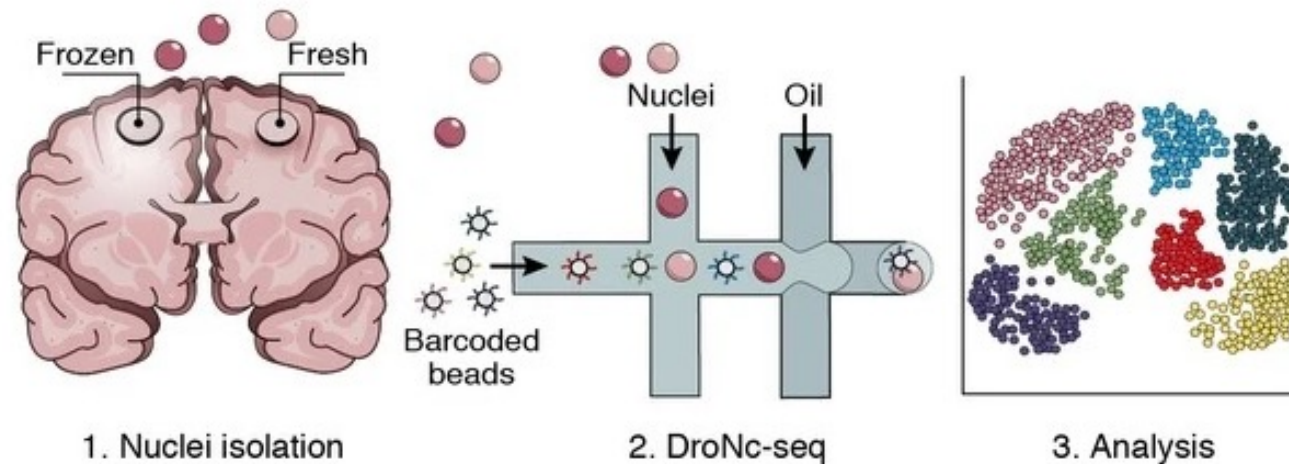
Single cell transcriptomics: a History

SPLiT-seq



<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7643870/>; March 2018

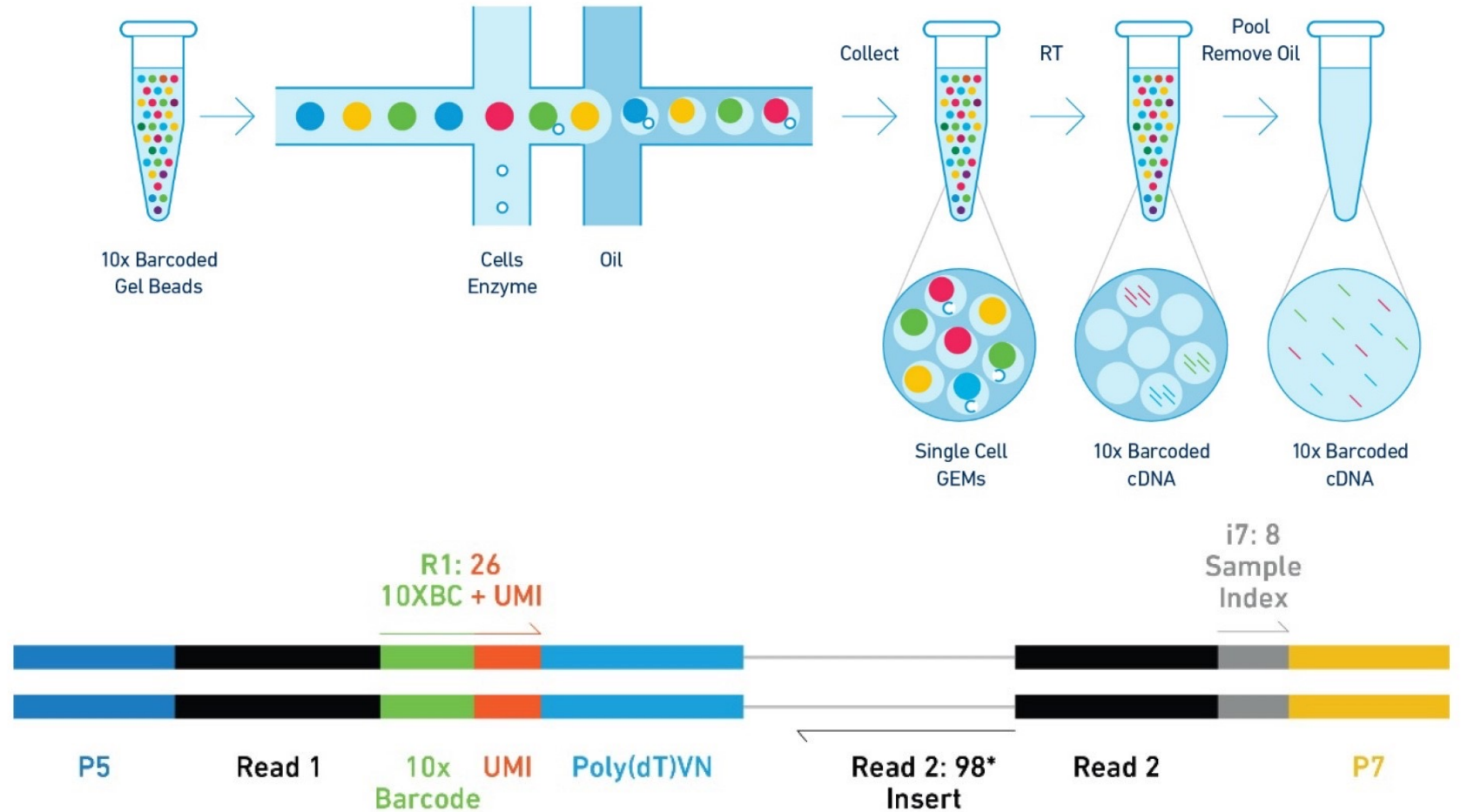
DroNc-seq



<https://www.nature.com/articles/nmeth.4407>; August 2017

Single cell transcriptomics: a History

- 10x genomics is effectively the commercial form of droNc-seq
- Uses 3' capture of transcripts
- Multi-omics assays are being developed



Single cell multi-omic methods: ATAC-seq

- Generally transcriptome and chromatin accessibility (ATAC-seq)
- 10x offers a kit that does this!
- Several other methods: SHARE-seq (SpliT-seq derived), ISSAAC-seq (Xu et al., 2022)

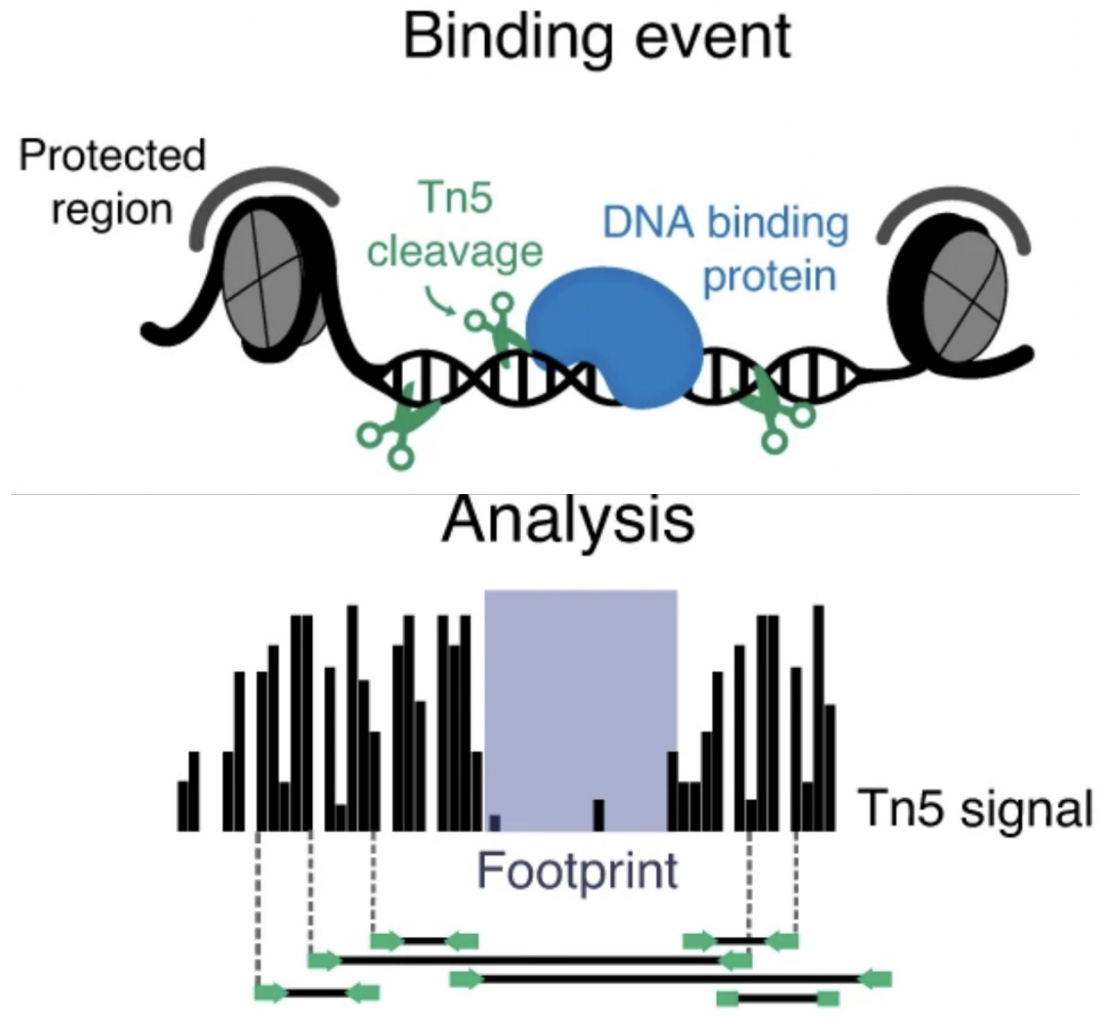
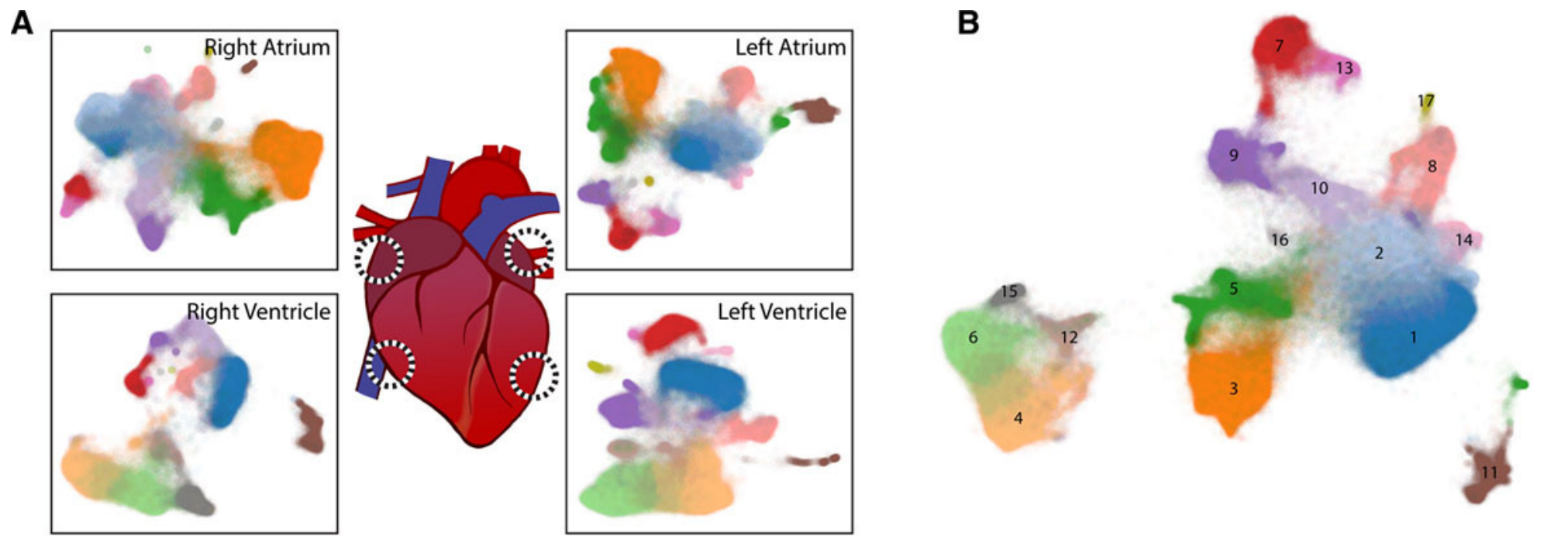


Image credit: Bentsen, Mette, et al. *Nature communications* (2020)

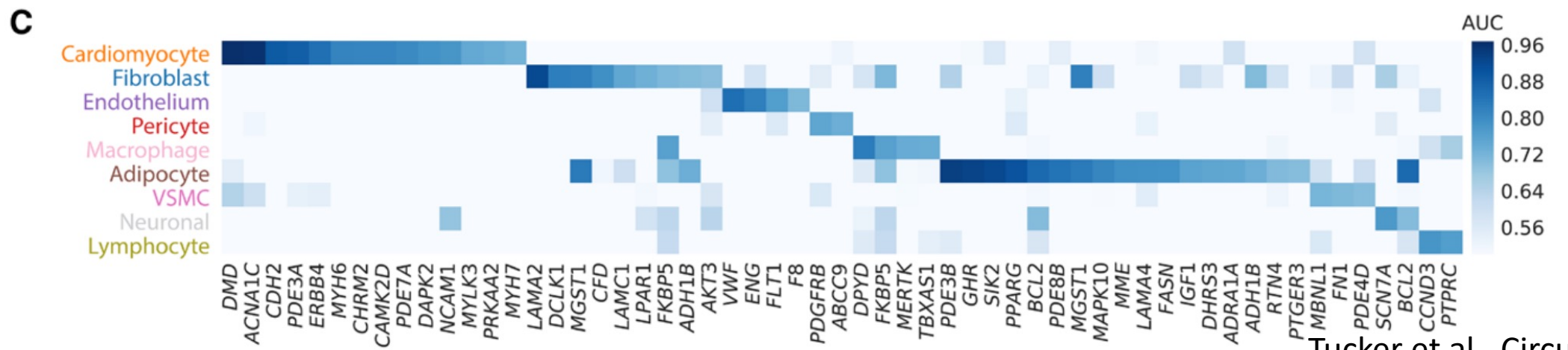
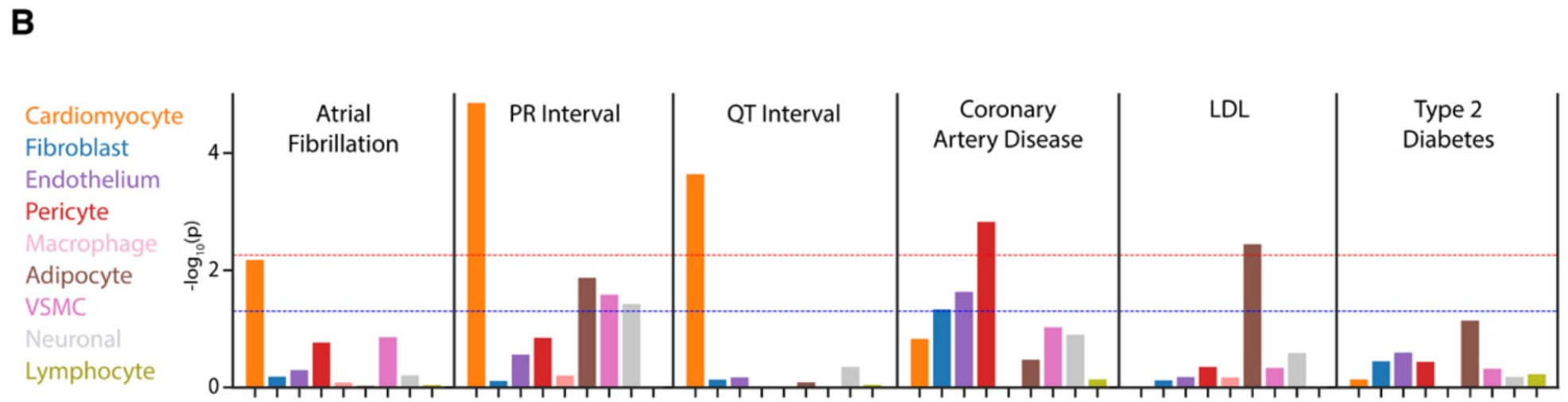
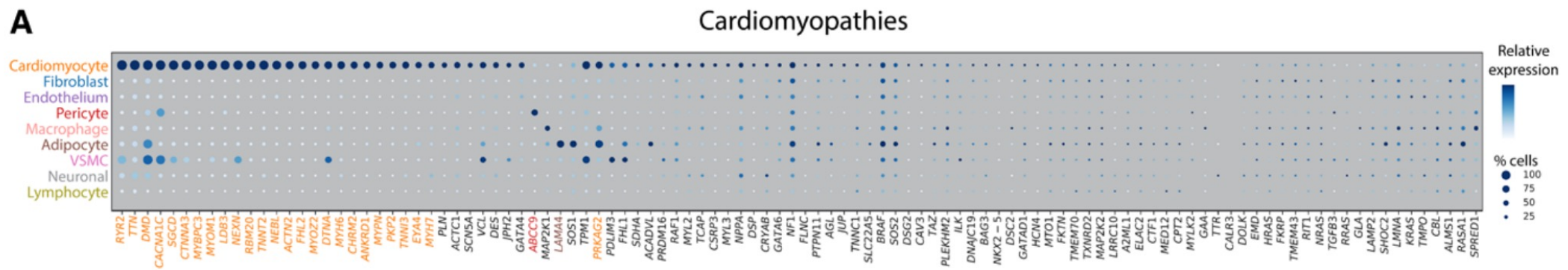
Many thanks to Chuck Epstein and Jason Buenrostro for this slide

ORIGINAL RESEARCH ARTICLE

Transcriptional and Cellular Diversity of the Human Heart



- 1 Fibroblast I
- 2 Fibroblast II
- 3 Atrial Cardiomyocyte
- 4 Ventricular Cardiomyocyte I
- 5 Cytoplasmic Cardiomyocyte I
- 6 Ventricular Cardiomyocyte II
- 7 Pericyte
- 8 Macrophage
- 9 Endothelium I
- 10 Endothelium II
- 11 Adipocyte
- 12 Cytoplasmic Cardiomyocyte II
- 13 Vascular Smooth Muscle
- 14 Fibroblast III
- 15 Ventricular Cardiomyocyte III
- 16 Neuronal
- 17 Lymphocyte



Cells of the adult human heart

<https://doi.org/10.1038/s41586-020-2797-4>

Received: 10 February 2020

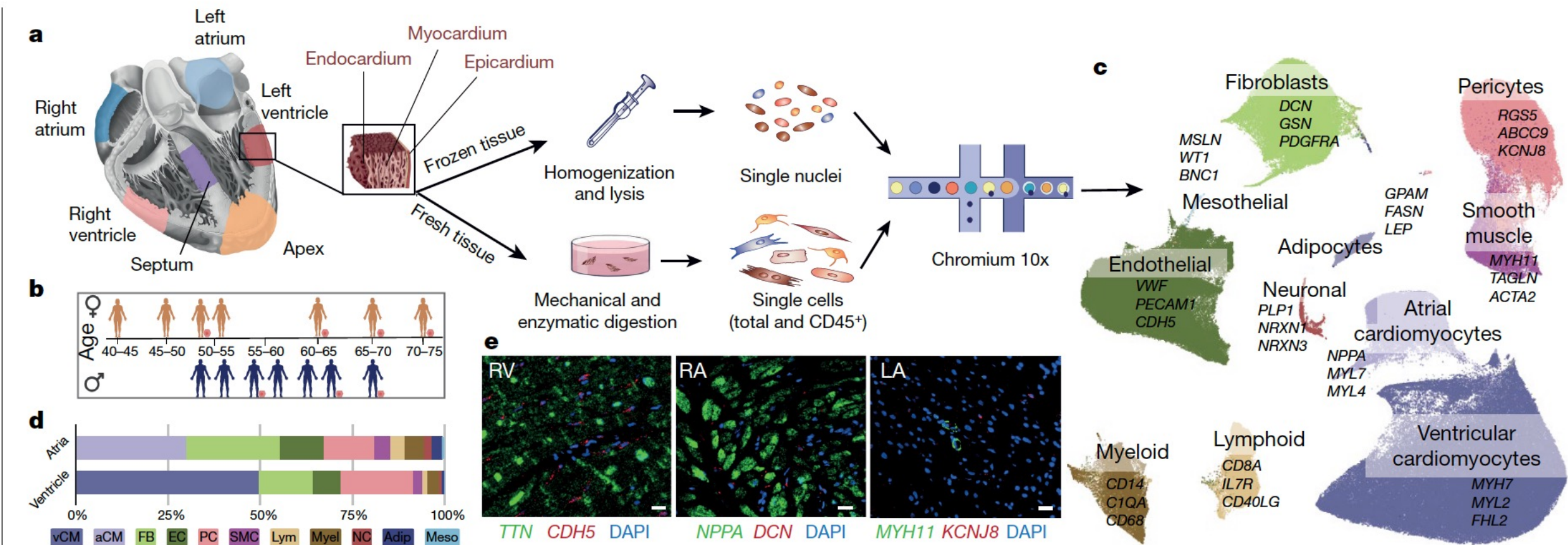
Accepted: 18 September 2020

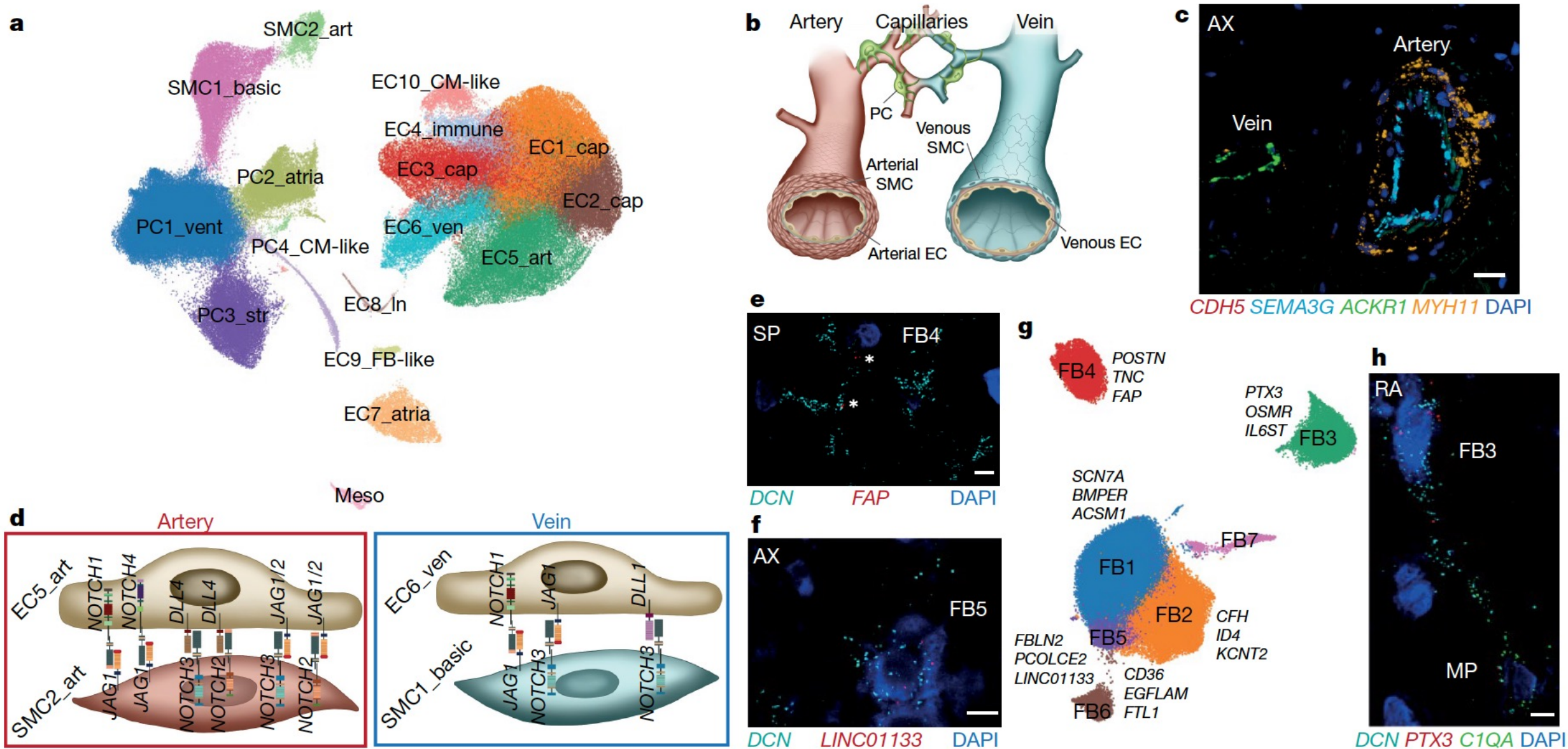
Published online: 24 September 2020

Open access

Check for updates

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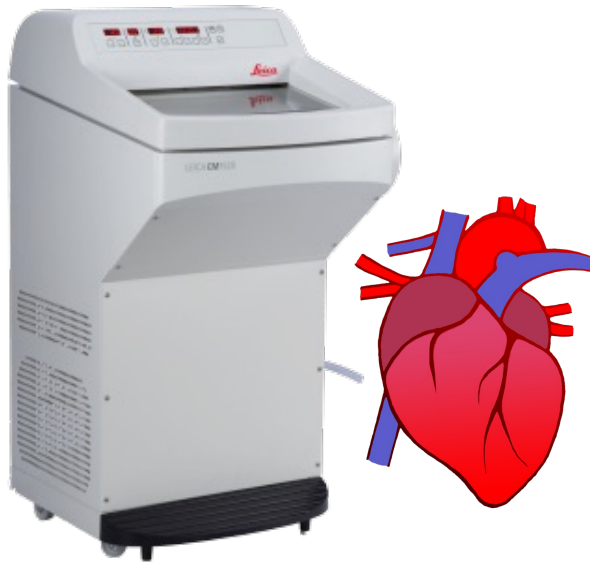


10x genomics single cell RNA-seq overview

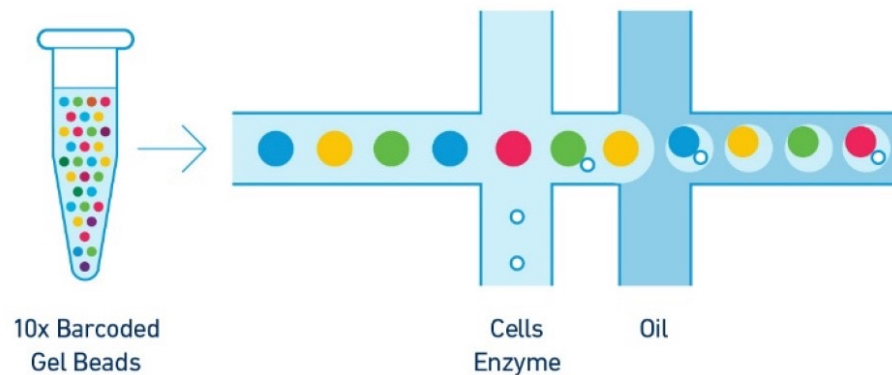
10X genomics protocol control – not much to change

Most user control

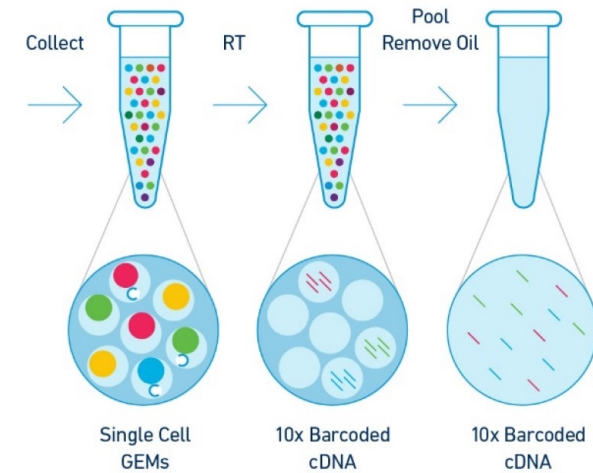
Prepare sample
(isolate cells: fresh tissue,
isolate nuclei: frozen tissue)



Run sample through chromium controller



Perform RT reaction,
break emulsion



**Output
standardized**

Sequence

some
user control

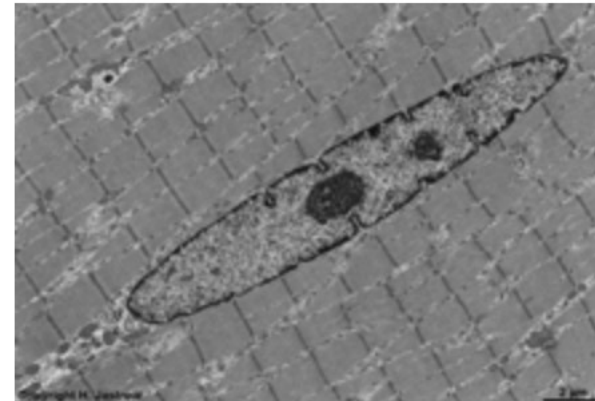
Analysis
pipeline

Experimental design best practices: nuclear isolation

- Tissue preparation and nuclear isolation are where you have the most control in this protocol!
 - The Chromium B chip can handle **30um** cells at MOST
- Human nuclei are typically about **10um** in diameter, so nuclear preps always fit
 - And good from frozen tissue
- Debris can clog the chip too, important to have clean nuclear preparations.

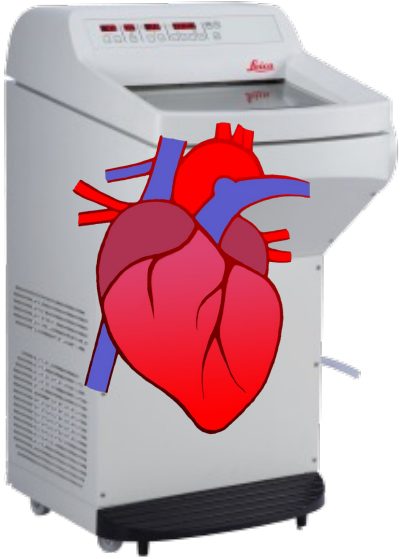
Cell type	diameter
PBMC	10um
HeLa	20um
hepatocyte	25-30um
Endothelial cells	50-70um x 10-30um
cardiomyocyte	60-140um x 17-25um
fibroblast	Up to 100um (moving)
nucleus	10um

Skeletal muscle nucleus



Experimental design best practices: nuclear isolation

100um sections



Dounce:
First loose,
then tight pestle



Spin at 40g:
pellet heavy debris



40um filter



10um filter



Ultracentrifuge through 2M sucrose gradient
(great for acellular/diseased samples):

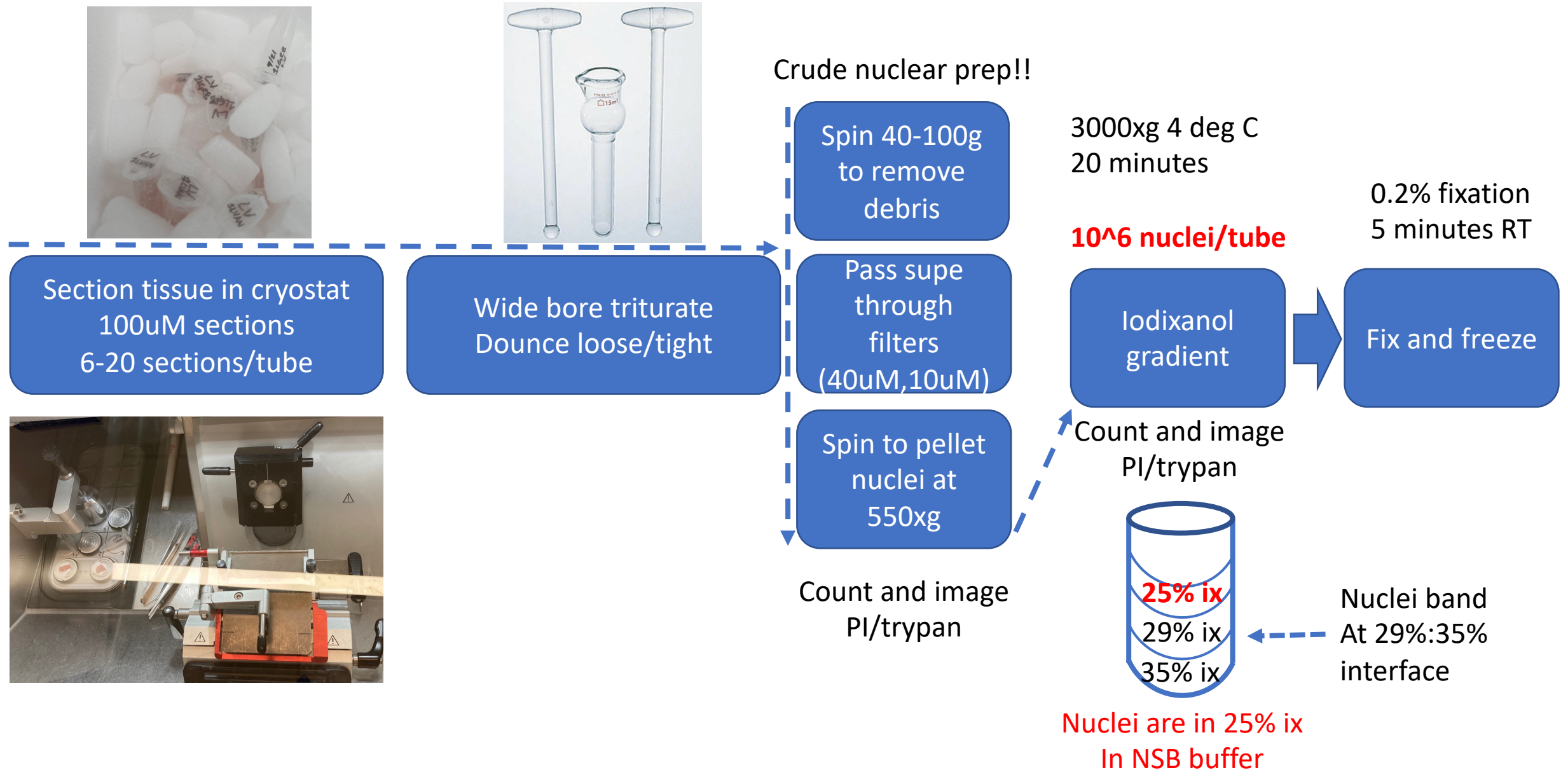


45min, 30000xg, 4 deg C

• Sample preparation

- Fresh tissue? Might be able to isolate cells and freeze viable
- Frozen tissue? Need to isolate nuclei
 - With vascular/CV derived tissue, sectioning on a cryotome is key
 - **Diseased tissue is often acellular**

Overall workflow of this nuclear isolation protocol



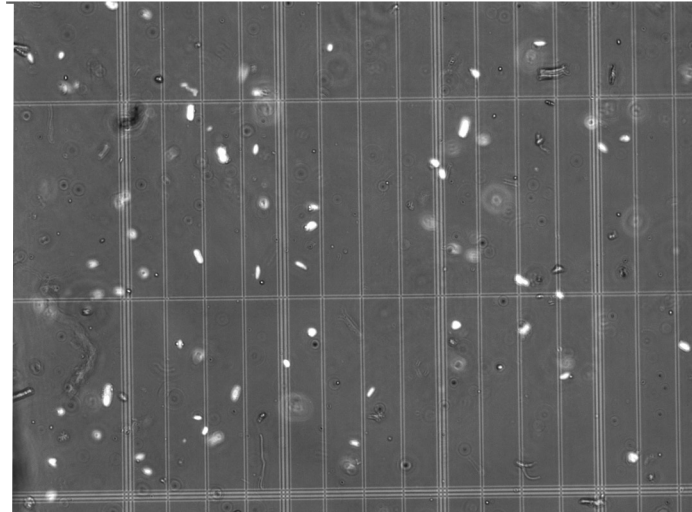
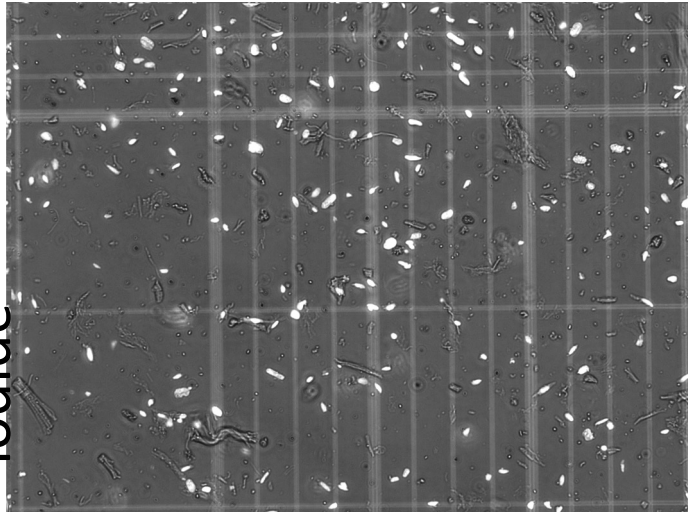
Heart Left Ventricle Nuclear Prep

10x

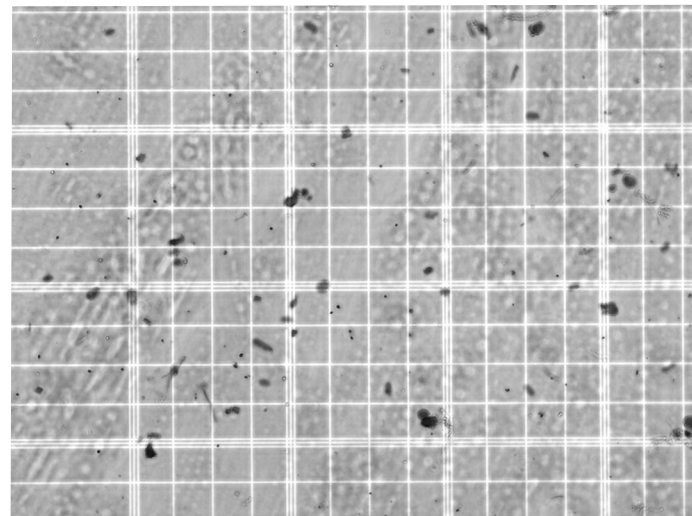
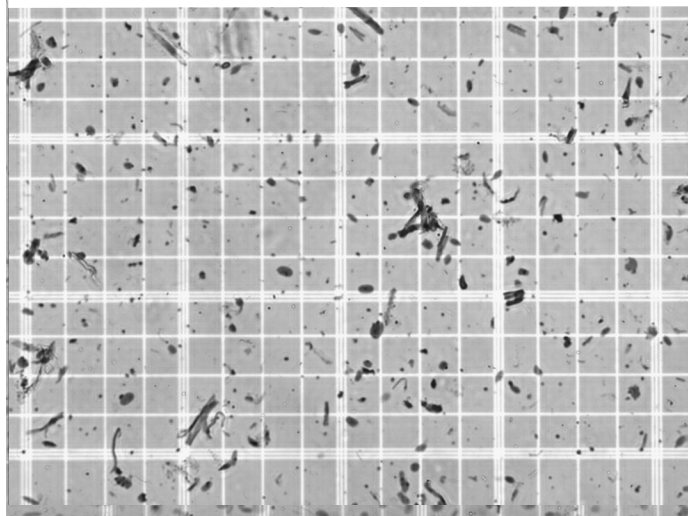
pre-iodixanol

post-iodixanol

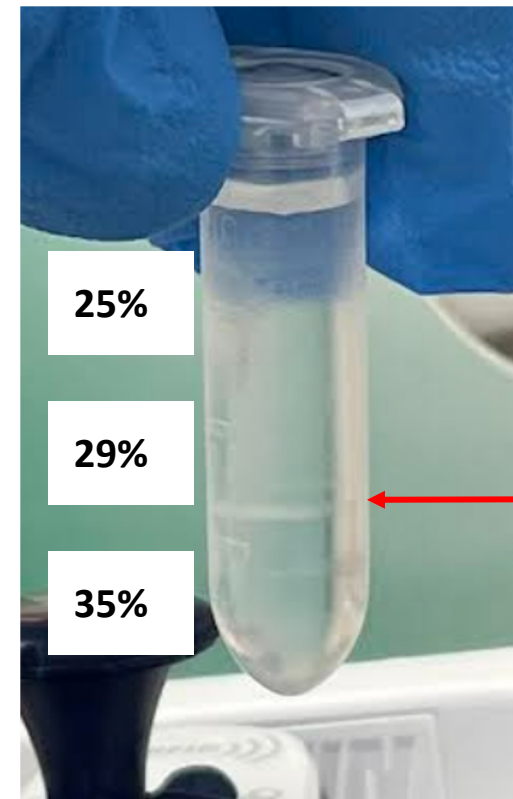
+propidium
iodide



+trypan



Pre-iodixanol gradient: 1.2M nuclei
Post-iodixanol gradient: .94M nuclei



IX gradient from a colon prep:
shown as an example

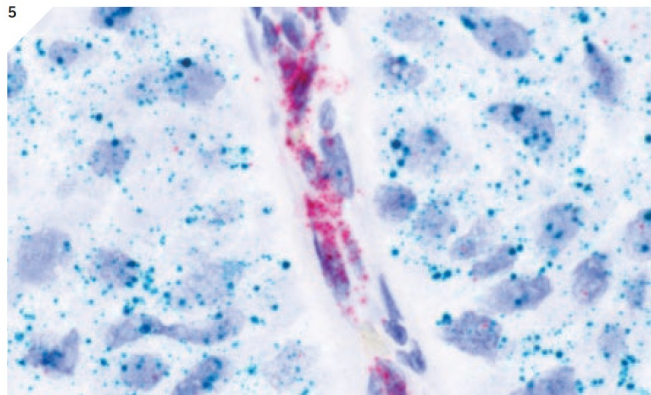
Experimental design best practices: tissue validation

- Identify a novel marker gene using computational analyses?
 - You need to be prepared to validate!!

RNA ISH (RNAscope)

Oligo + fluorophore staining

Breast cancer slide

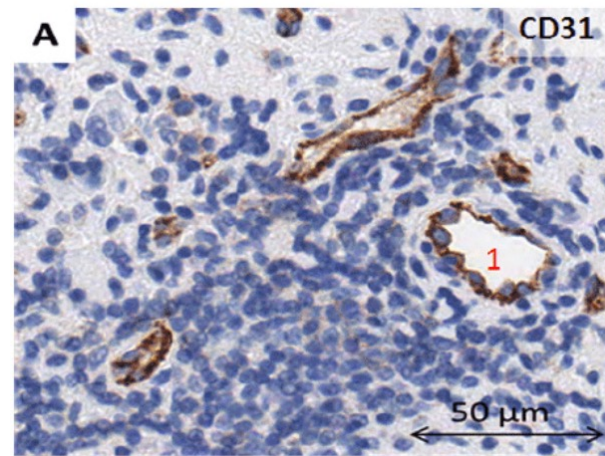


EPCAM in red
EGFR in green

IHC

immunohistochemistry

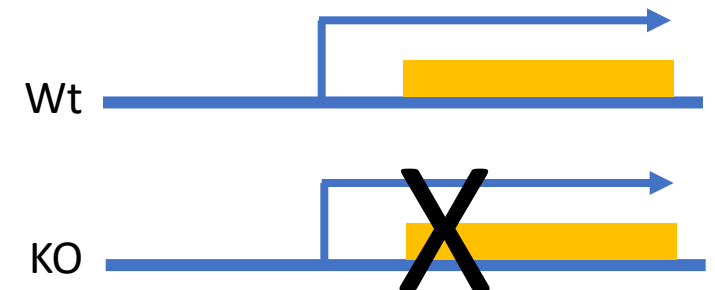
Neuroblastoma slide



CD31 in brown
Counterstain in blue

In vitro models

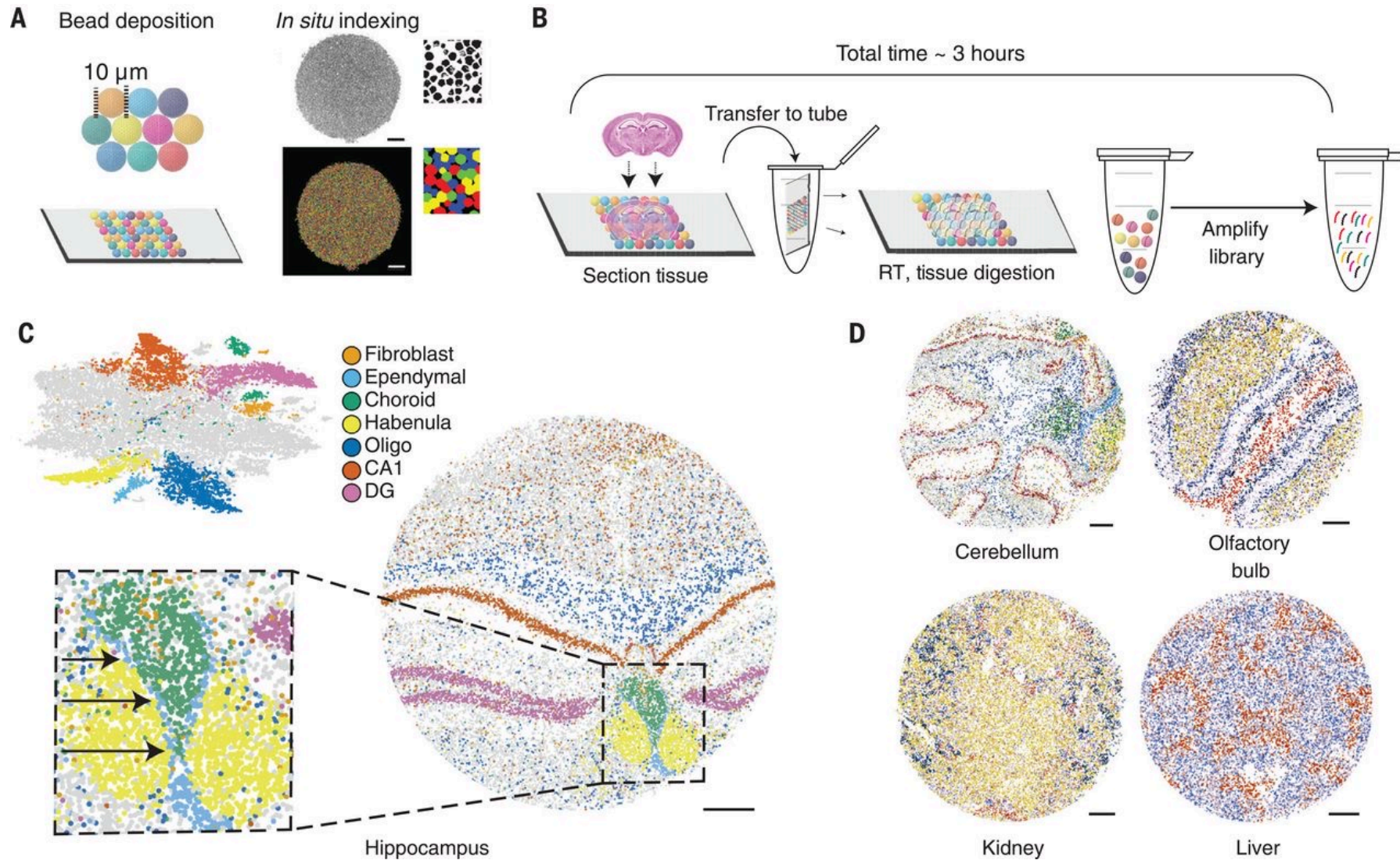
CRISPR KO (cells)
Plus imaging



Need a good *in vitro* cell line
iPSC cells from patients possible

Can also do EP to show effect in cell types

Experimental design best practices: tissue validation



SLIDE-seq

Also see:

MERFISH

Visium

Sci-space

Sequencing structure for 10x multiome

R1: 28-50bp
i7: 8-10bp
i5: 10-24bp
R2: 49-90bp

mRNA

Dual Indexed Sequencing Run: Single Cell Multiome Gene Expression libraries are dual-indexed. This means both i5 and i7 reads are used for demultiplexing. We do not recommend sequencing 10x Single Cell Multiome Gene Expression dual index libraries with a single-index configuration.

Read	Read 1	i7 Index	i5 Index	Read 2
Purpose	Cell barcode & UMI	Sample Index	Sample Index	Insert
Length**	28	10	10	90

ATAC

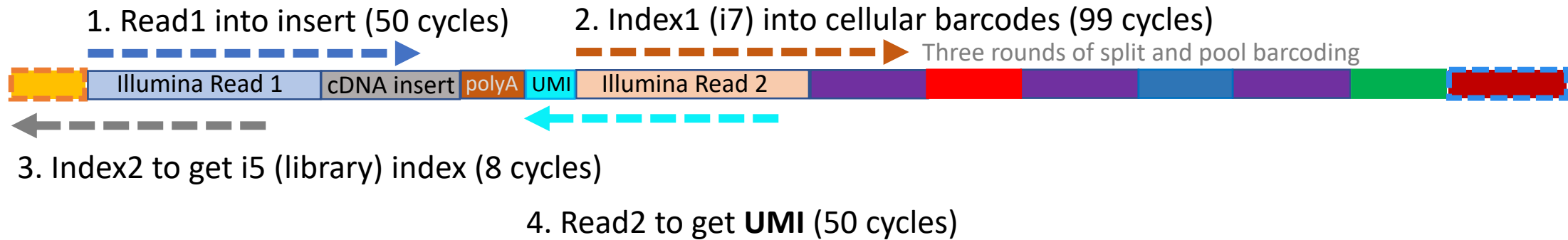
Read	Read 1	i7 Index	i5 Index	Read 2
Purpose	Transposed DNA	Sample Index	10x Barcode	Transposed DNA
Length	50 **	8	24 ⁺	49 **

Don't combine ATAC and mRNA on the same **flowcell!!**

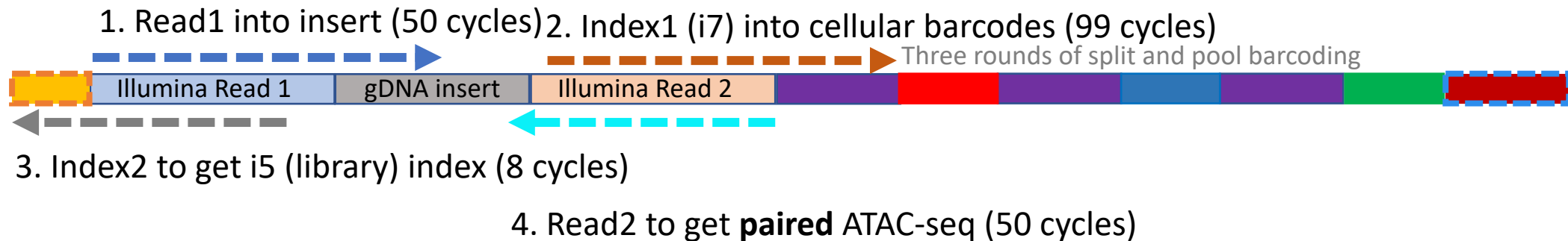
Sequencing structure for SHARE

R1: 50bp
i7: 99bp
i5: 8bp
R2: 50bp

mRNA

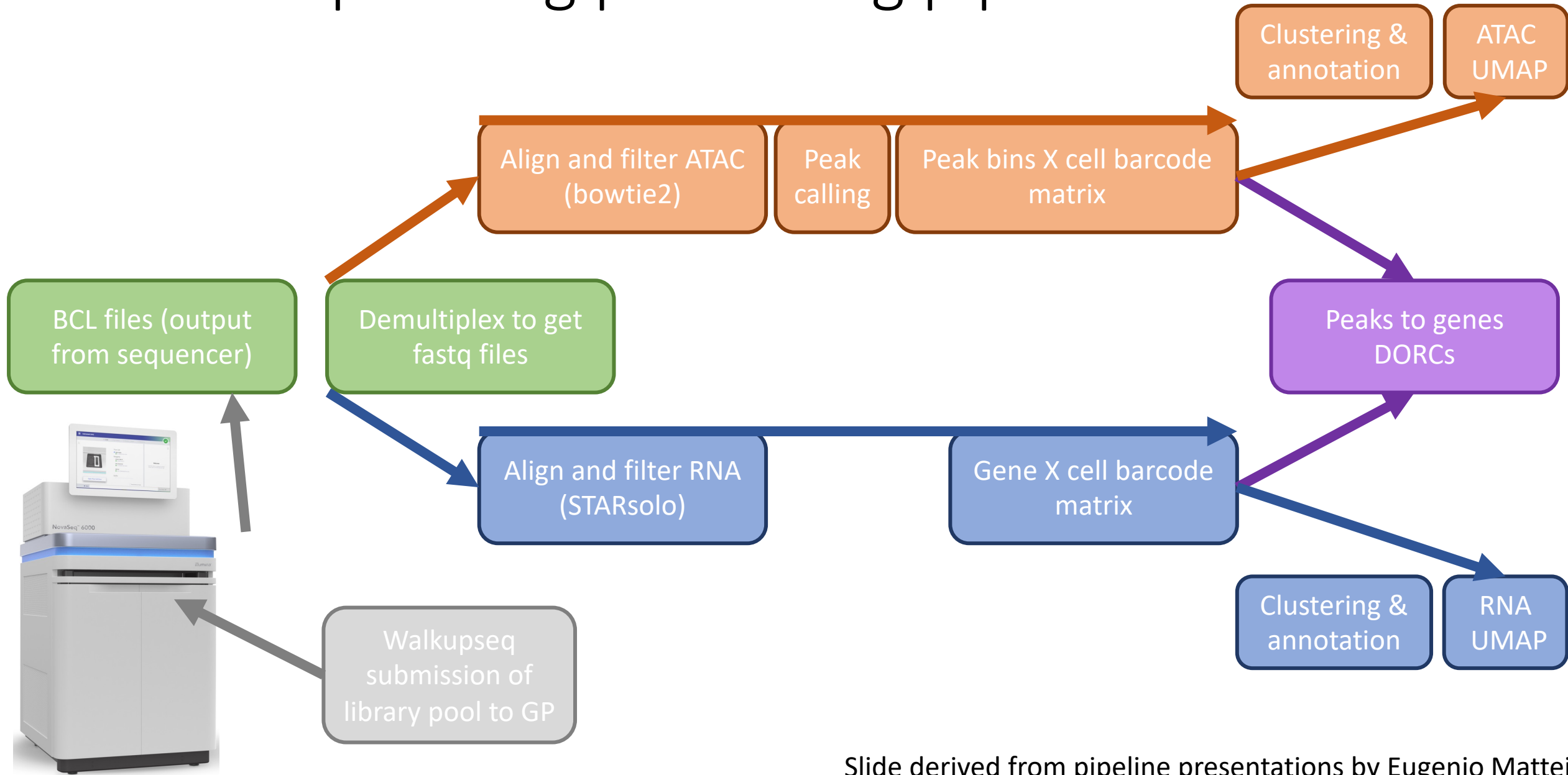


ATAC



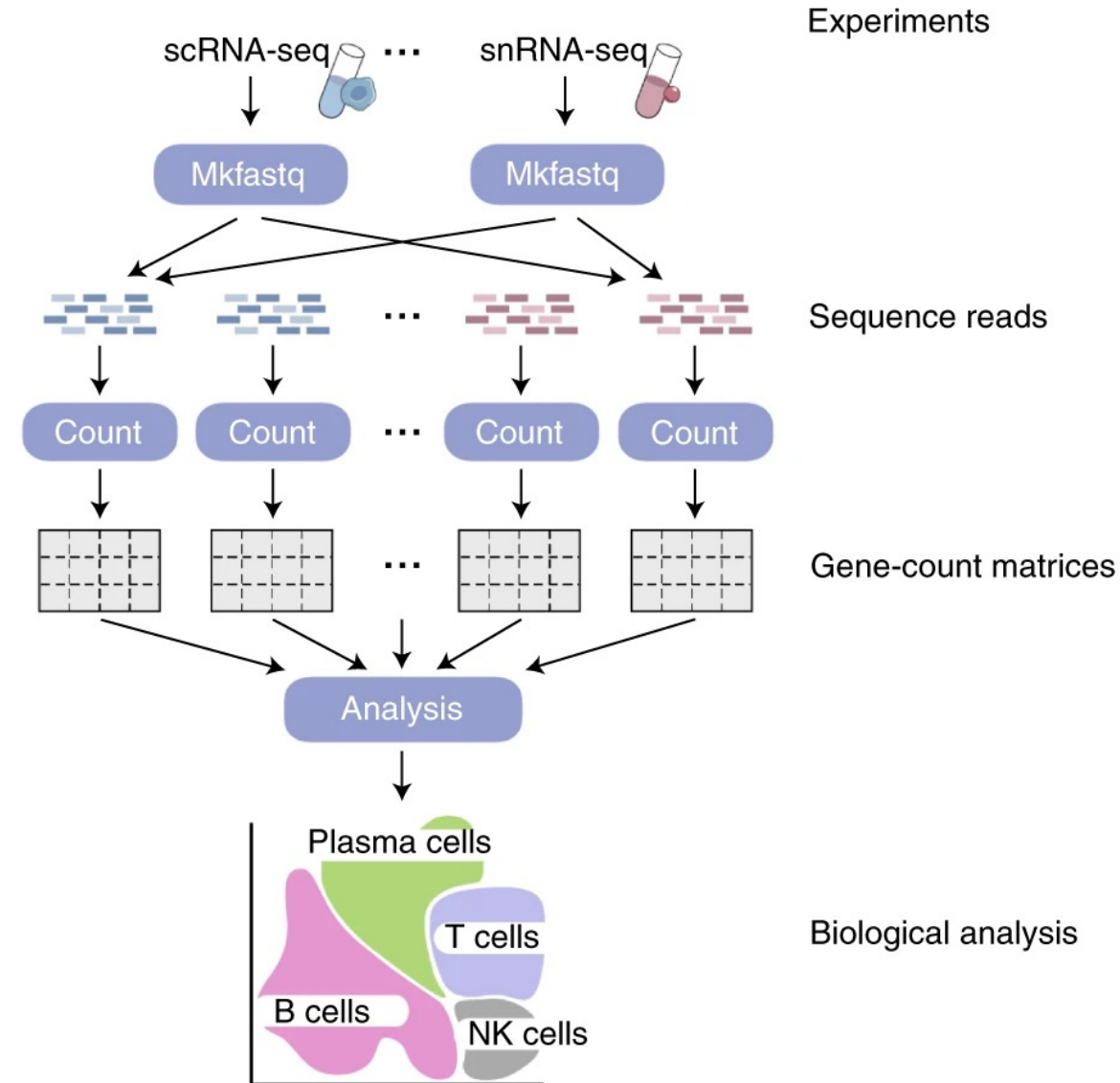
Don't combine ATAC and mRNA on the same lane!!

In brief: sequencing processing pipeline



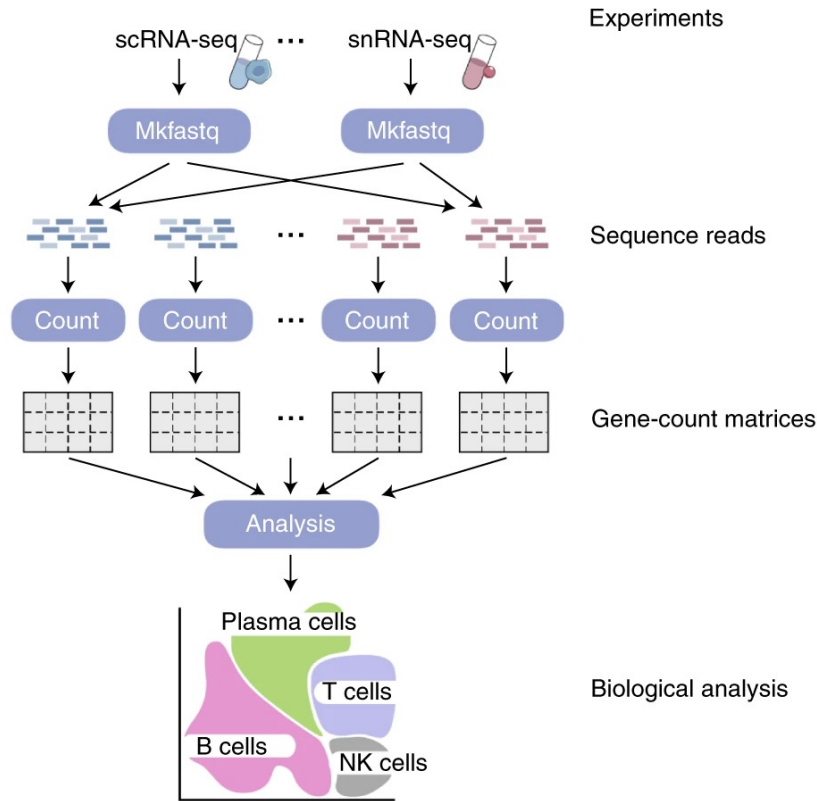
Computational analysis best practices: pipelines

a

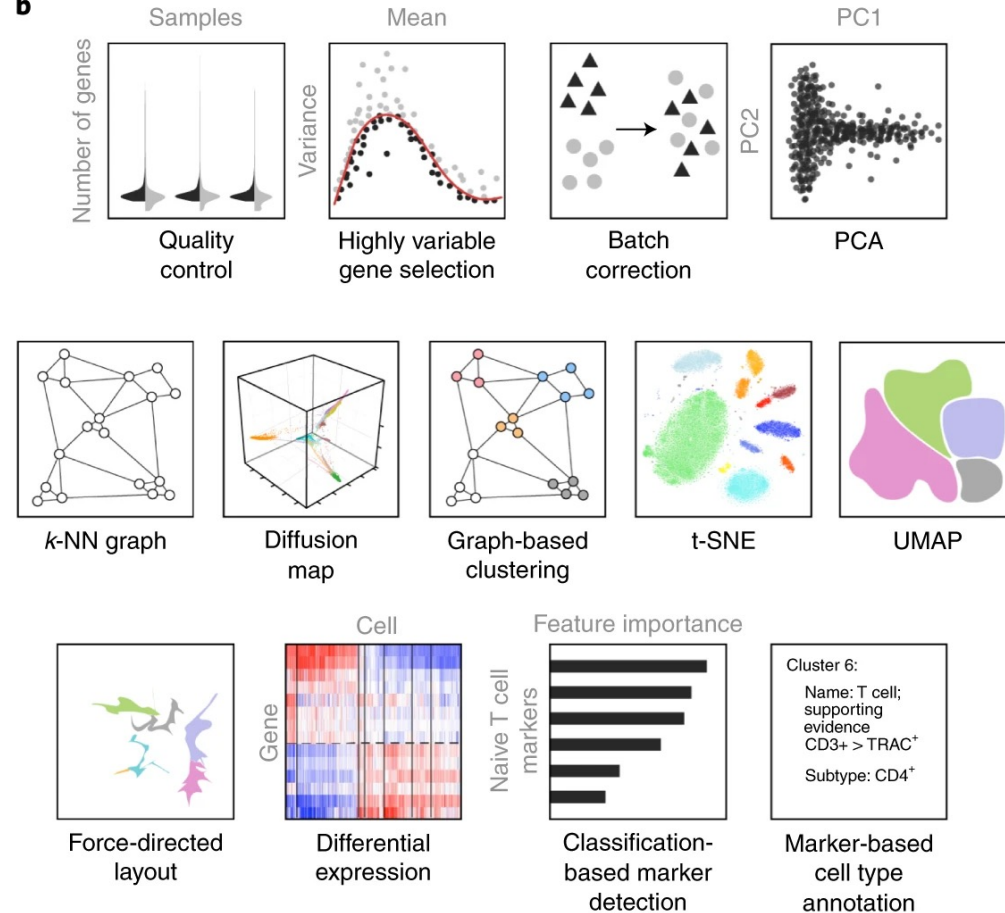


Computational analysis best practices: pipelines

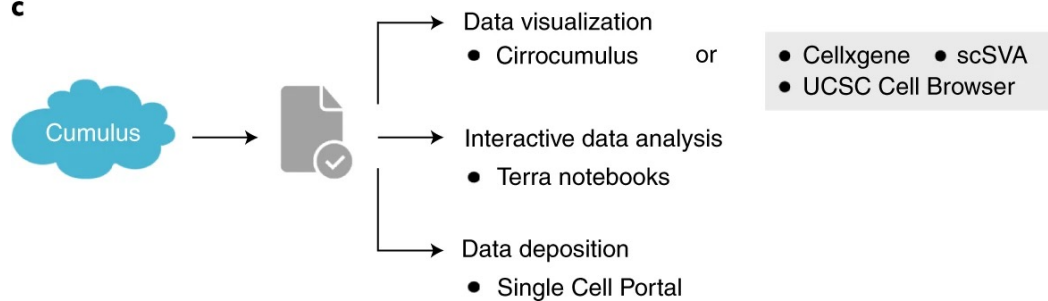
a



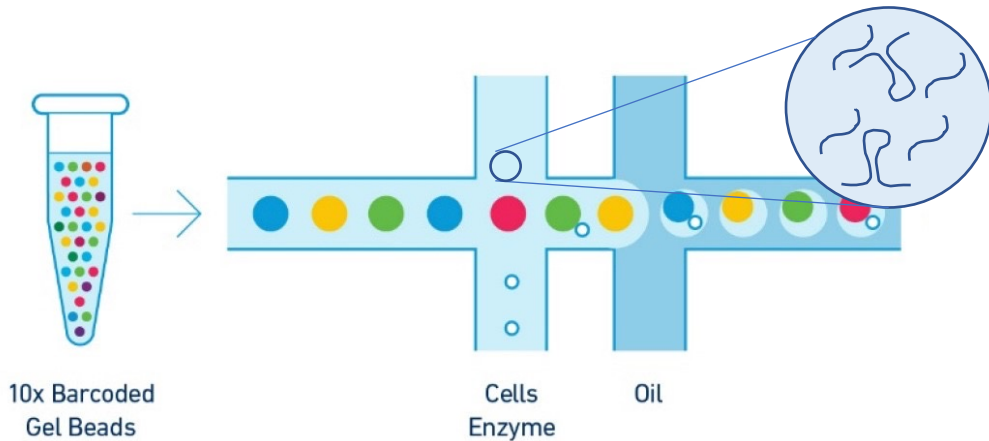
b



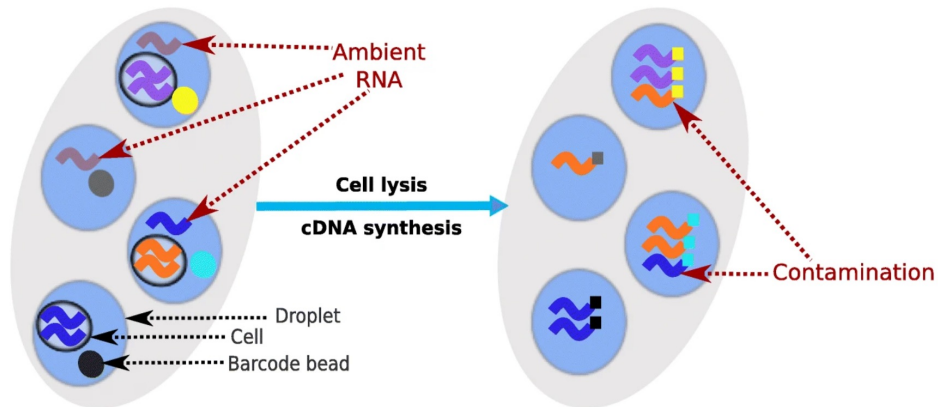
c



Computational analysis best practices: ambient RNA



- VERY important for single nucleus RNA-seq
 - Some cytoplasmic material can remain, or organelles attached to the nucleus
 - This leads to some ambient RNA contaminating the aqueous solution around the nucleus + gel bead



Many methods available to handle this:

Cellbender: <https://github.com/broadinstitute/CellBender>

SoupX: <https://github.com/constantAmateur/SoupX>

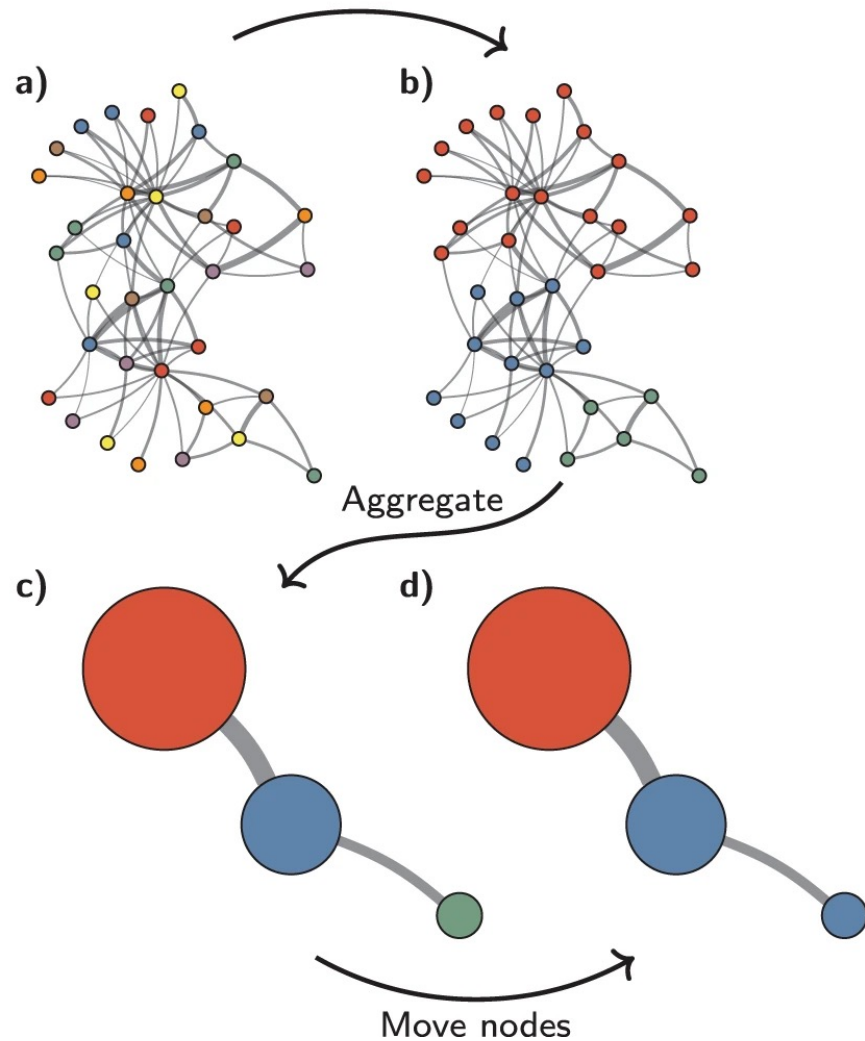
ScVI: <https://docs.scvi-tools.org/en/stable/>

DeconX: <https://github.com/campbio/celda>

Computational analysis best practices: clustering

Louvain

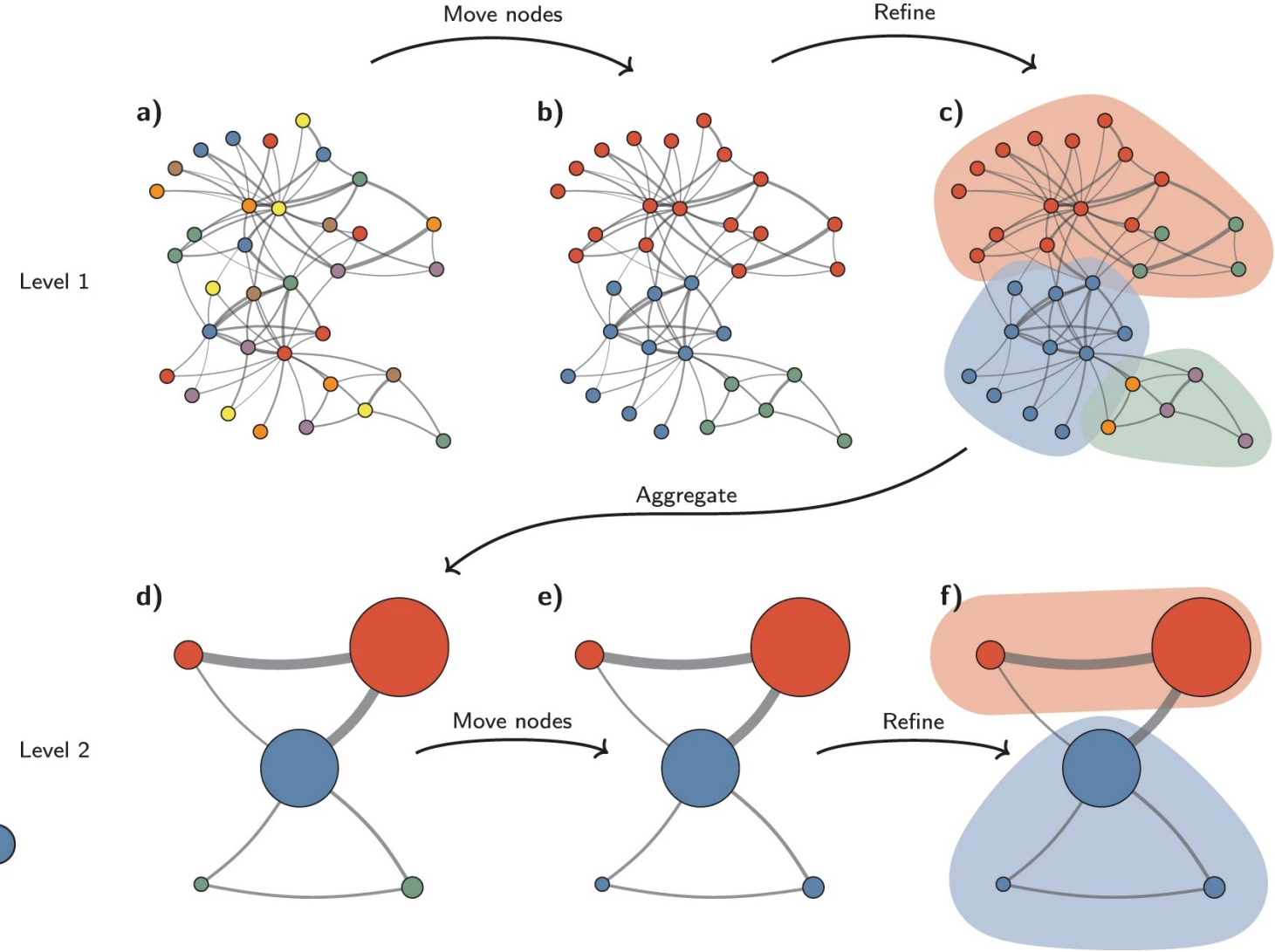
Move nodes



Leiden

Move nodes

Refine



Resources: The single cell portal at the Broad Institute

The screenshot shows the Single Cell Portal interface. At the top, there is a navigation bar with the logo, 'Single Cell BETA PORTAL', and links for 'Help & Resources', 'Create a Study', and 'Sign In'. Below the navigation bar is a large banner with the 'Single Cell BETA PORTAL' logo and the tagline 'Reducing barriers and accelerating single-cell research'. A circular graphic on the right side of the banner displays 'Featuring 372 studies' and '13,731,607 cells'. The main content area features a search bar with 'heart' entered, and filters for 'cell type', 'organ', 'species', and 'disease'. A 'COVID-19 Studies' button is visible on the right. Below the search bar, it indicates '15 total studies found' and shows the first two results. The first result is 'Single-nucleus transcriptomic survey of cell diversity and functional maturation in postnatal mammalian hearts' with 12320 cells. The second result is 'Transcriptional and Cellular Diversity of the Human heart' with 287269 cells.

Single Cell BETA PORTAL

Help & Resources + Create a Study Sign In

Single Cell BETA PORTAL

Reducing barriers and accelerating single-cell research

Featuring 372 studies 13,731,607 cells

Search Studies Search Genes

heart cell type organ species disease More Facets

COVID-19 Studies Download

Q: Text contains (heart) Clear All

15 total studies found Page 1 of 3

Single-nucleus transcriptomic survey of cell diversity and functional maturation in postnatal mammalian hearts
12320 Cells
Hu, P., Liu, J., Zhao, J., Wilkins, B. J., Lupino, K., Wu, H., and Pei, L. (2018). Single-nucleus transcriptomic survey of cell diversity and functional maturation in postnatal mammalian hearts. Genes & development, 32(19-20), 1344-1357. Scripts for data analysis are available on GitHub repository (https://github.com/wulabupenn/Hu_GenesDev_2018) A fundamental challenge in understanding cardiac biology and disease is that the remarkable heterogeneity in cell type composition and functional states have not been well characterized at single-cell resolution in maturing and diseased mammalian hearts. Massively parallel single-nucleus RNA sequencing (snRNA-seq) has emerged a ...*(continued)*

Transcriptional and Cellular Diversity of the Human heart
287269 Cells
Transcriptional and Cellular Diversity of the Human heart Nathan R. Tucker,1,2,# Mark Chaffin,1,# Stephen J. Fleming,1,3 ... The Broad Institute of MIT and Harvard, Cambridge, MA, USA 02142# These authors contributed equally Abstract Background: The human heart requires a complex ensemble of specialized cell types to perform its essential function. A greater knowledge of the intricate cellular milieu of the heart is critical to increase our understanding of cardiac homeostasis and pathology. As recent advances in low input RNA-sequencing have allowed definitions of cellular transcriptomes at single cell resolution at scale, here we have applied these approaches to assess the cellular and transcriptional diversity of the non-failing human heart. Methods: Microfluidic encapsulation and barcoding was used to perform single nuclear RNA sequencing with samples from seven human donors, selected for ...*(continued)*

https://singlecell.broadinstitute.org/single_cell

Resources: Azimuth (Satija lab)

HuBMAP

[Home](#) [References](#)

Azimuth

App for reference-based single-cell analysis

Azimuth is a web application that uses an annotated reference dataset to automate the processing, analysis, and interpretation of a new single-cell RNA-seq experiment. Azimuth leverages a 'reference-based mapping' pipeline that inputs a counts matrix of gene expression in single cells, and performs normalization, visualization, cell annotation, and differential expression (biomarker discovery). All results can be explored within the app, and easily downloaded for additional downstream analysis.

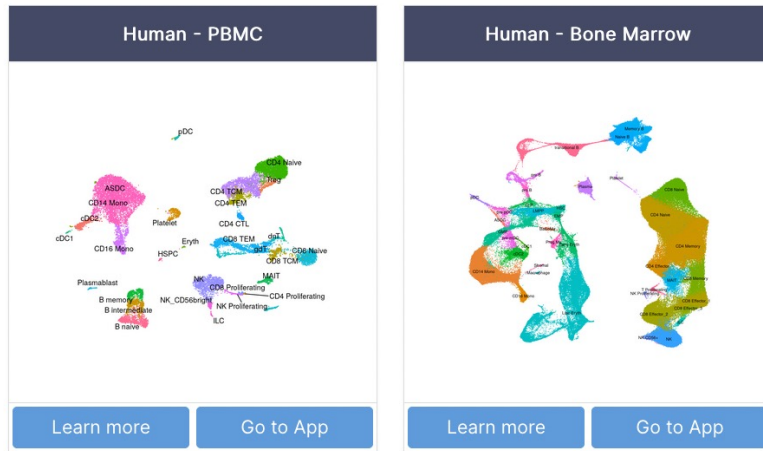
The development of Azimuth is led by the New York Genome Center Mapping Component as part of the [NIH Human Biomolecular Atlas Project \(HuBMAP\)](#). Seven molecular reference maps are currently available, with more coming soon.

NOW WITH SINGLE CELL ATAC-seq DATA!!!

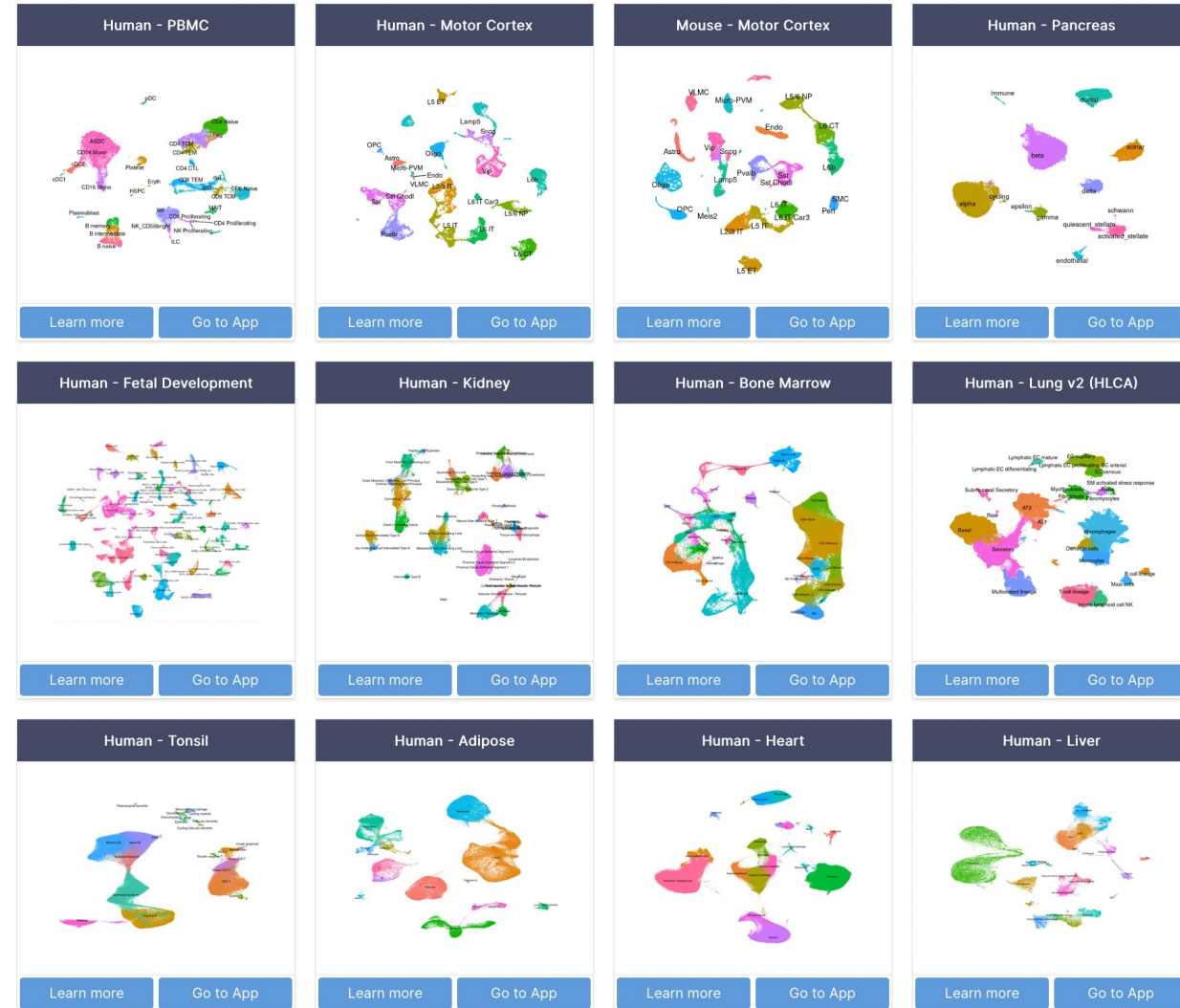
<https://azimuth.hubmapconsortium.org/>

Resources: Azimuth (Satija lab)

References for scATAC-seq Queries



References for scRNA-seq Queries



NOW WITH SINGLE CELL ATAC-seq DATA!!!

<https://azimuth.hubmapconsortium.org/>

Resources: Hubmap! (validated cell type markers!)

ASCT+B REPORTER

v2.5

The CCF ASCT+B Reporter is a visualization tool for displaying anatomical structures and cell types for different human organs. The tables are used to develop a common codebook for the Hubmap Consortium website.

Go to Visualization

Launch

Select one or more ASCT+B Tables

<input type="checkbox"/>	Organ Name	Ver
<input type="checkbox"/>	All by CCF-HRA release	v
<input type="checkbox"/>	Blood	v1.3
<input checked="" type="checkbox"/>	Blood Vasculature	v1.3
<input type="checkbox"/>	Bone Marrow	v1.3

Cancel

Submit

- Anatomical Structures
- Cell Types
- Gene Biomarkers
- AS-AS, AS-CT, CT-BM Paths

Cell Types

A-Z Alphabetically

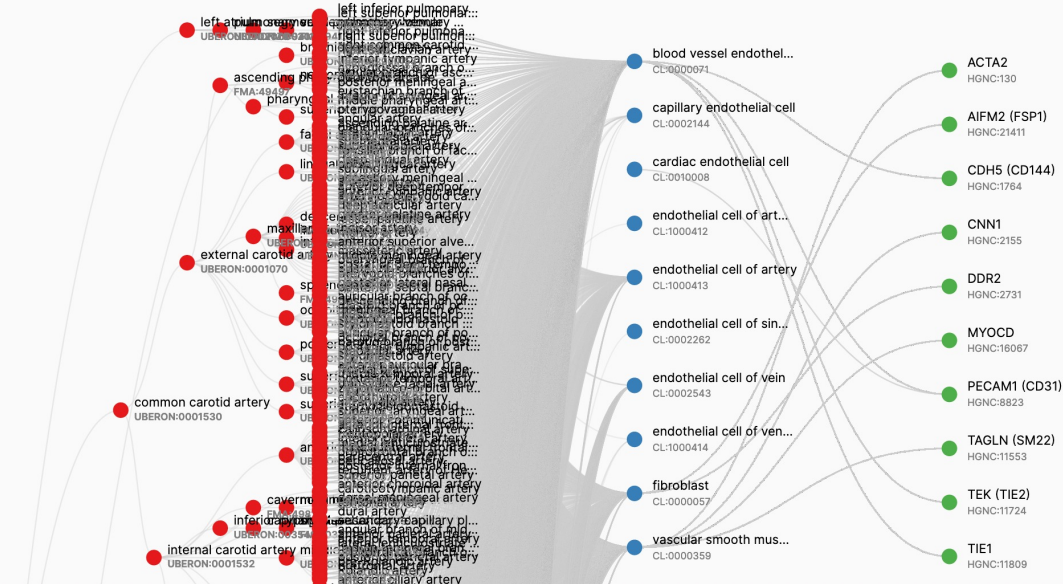
None

Biomarkers

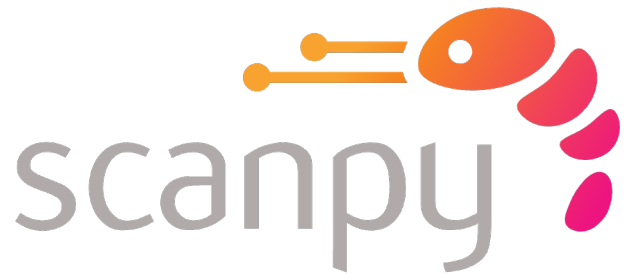
A-Z Alphabetically

None

All

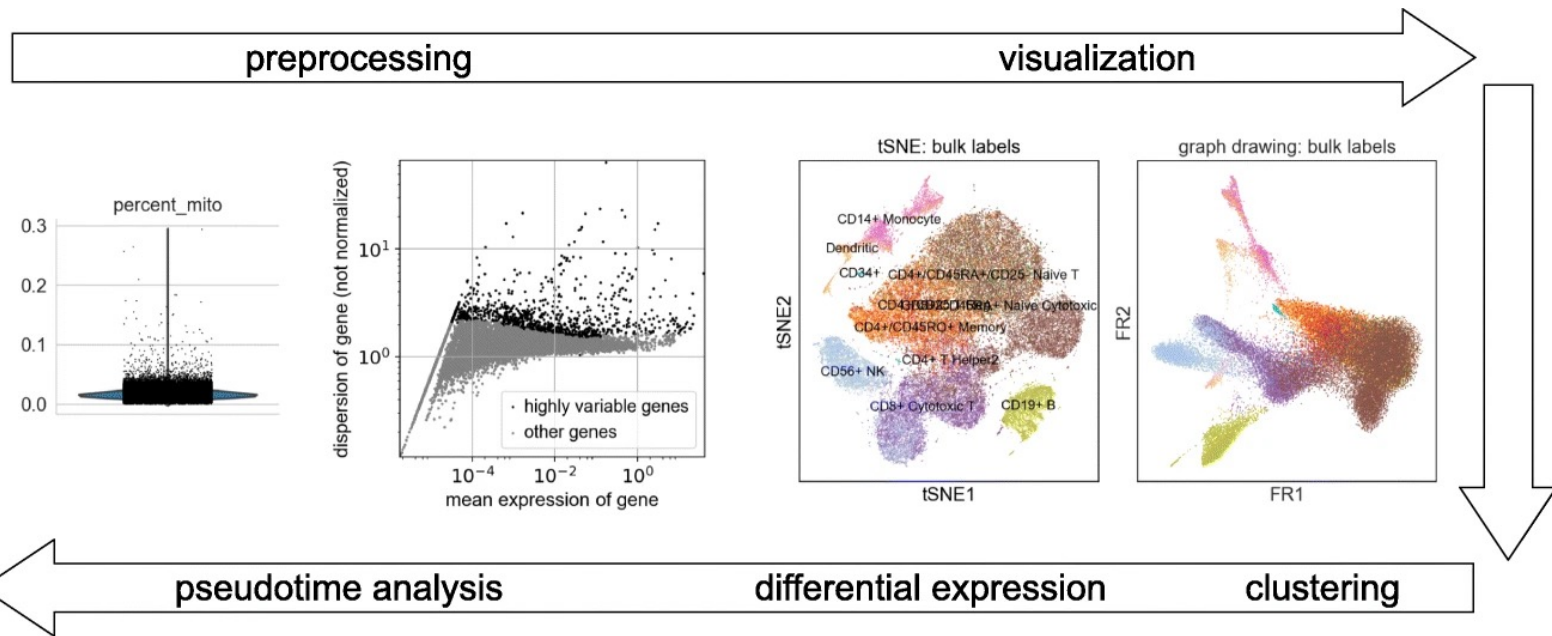


<https://hubmapconsortium.github.io/ccf-asct-reporter/>



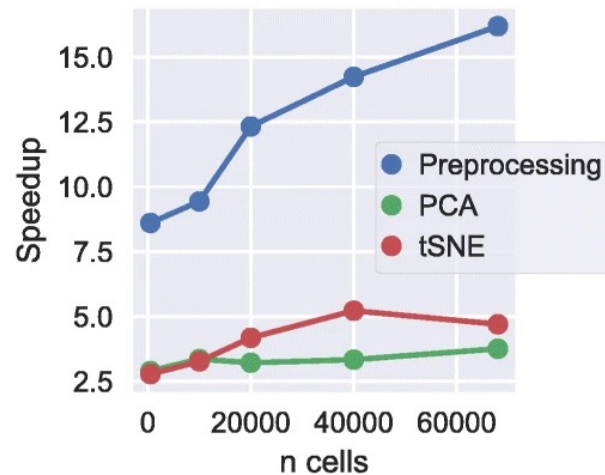
Packages: Scanpy (python)

a



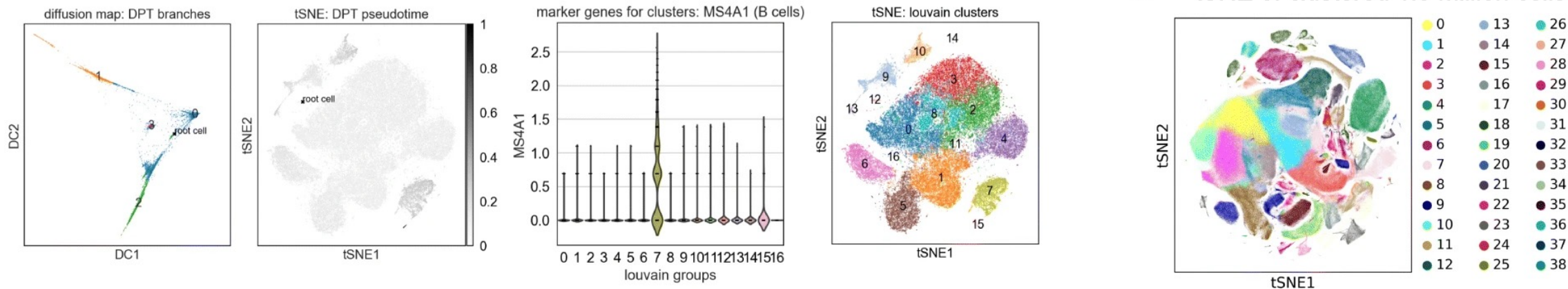
b

Speedup: Scanpy vs. Cell Ranger R



c

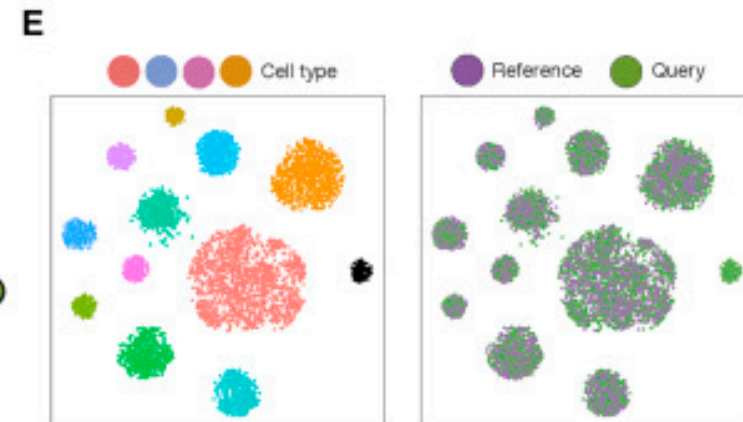
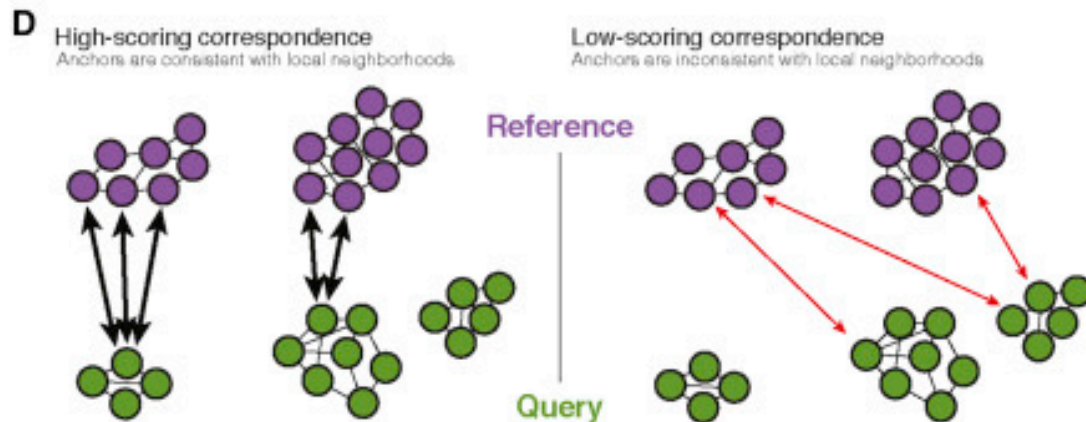
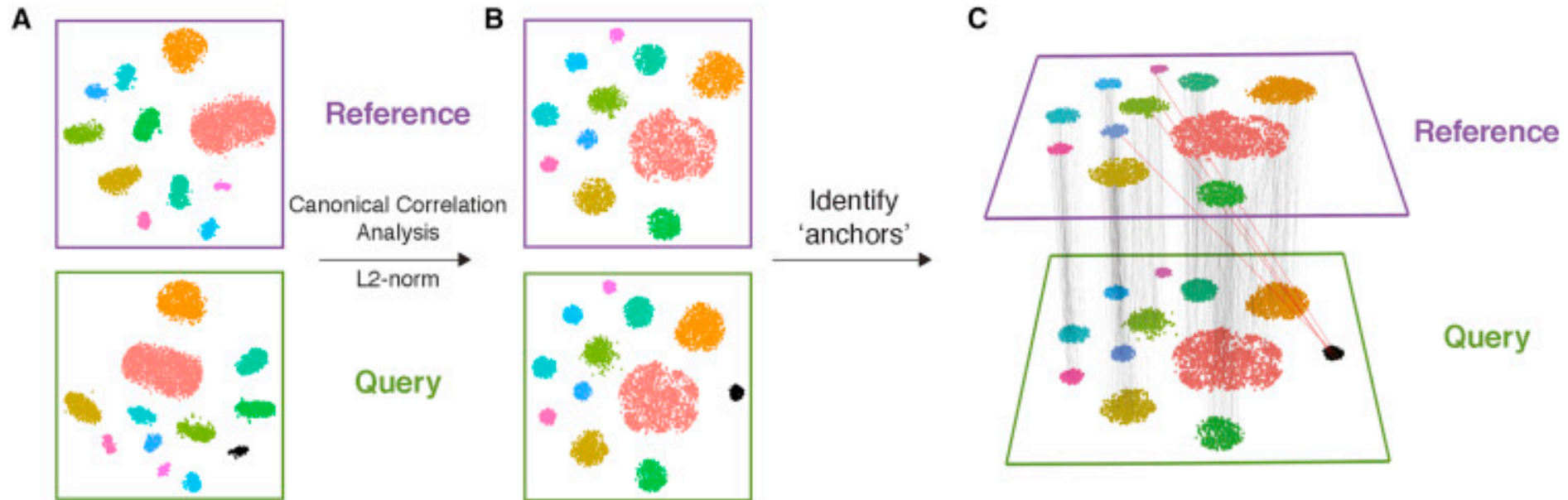
tSNE of clustered 1.3 million cells



SEURAT

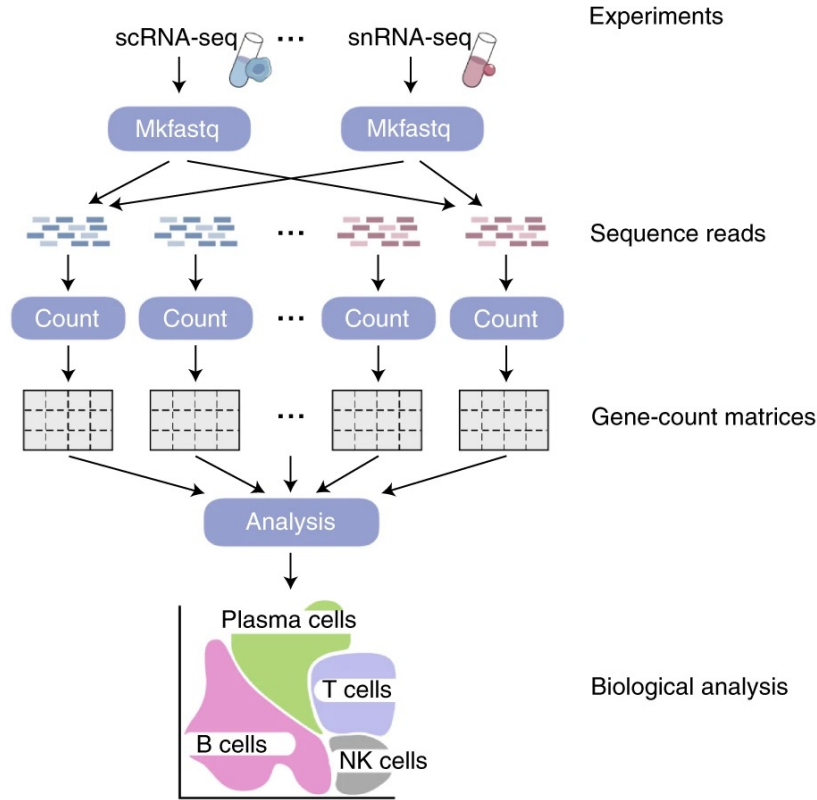
R toolkit for single cell genomics

Packages: Seurat (R)

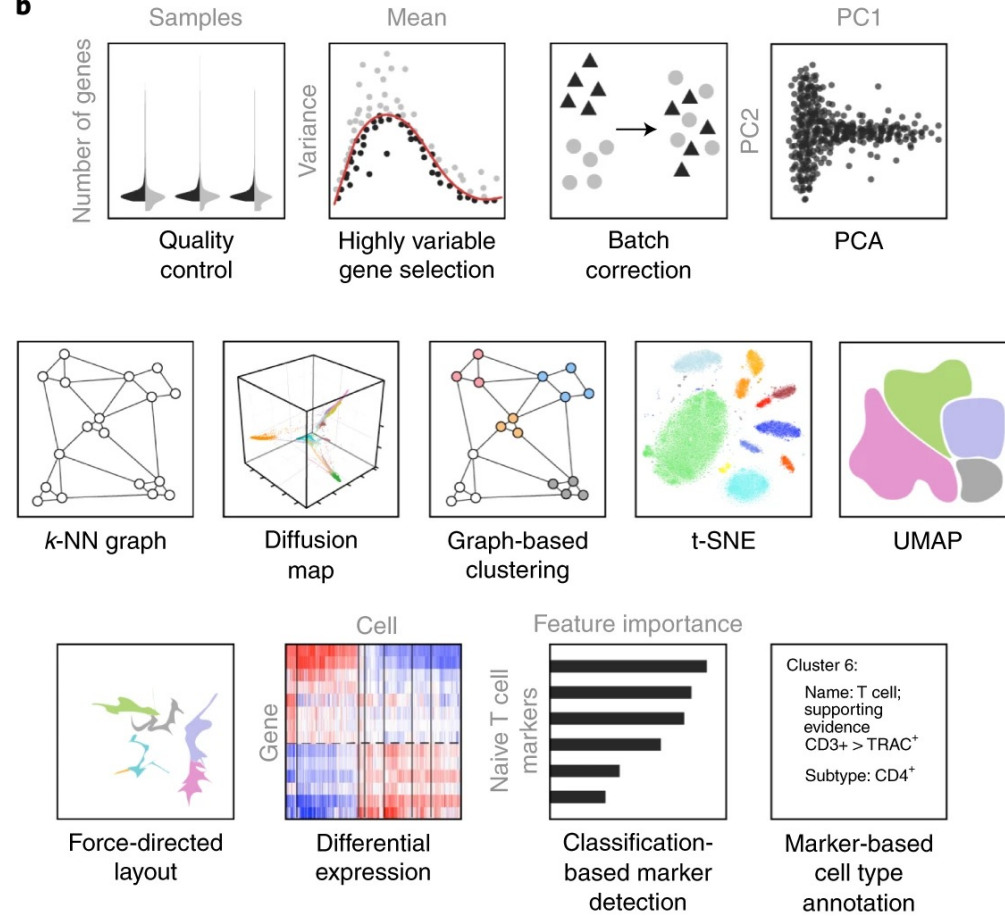


Pipelines: Cumulus (Terra workflow)

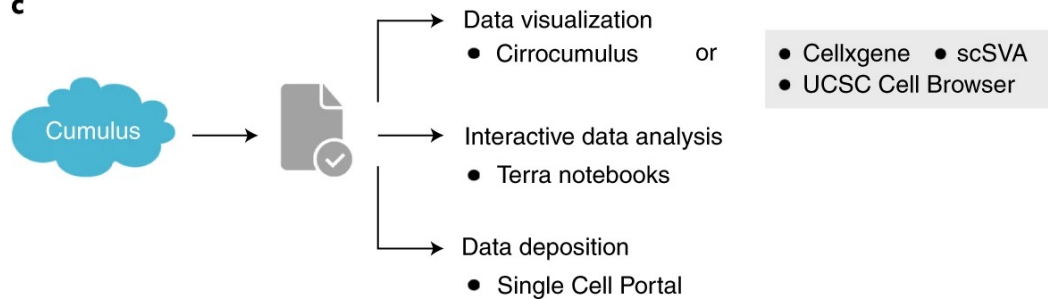
a



b



c



<https://cumulus.readthedocs.io/en/stable/index.html>

Li et al., Nature Methods, 2020

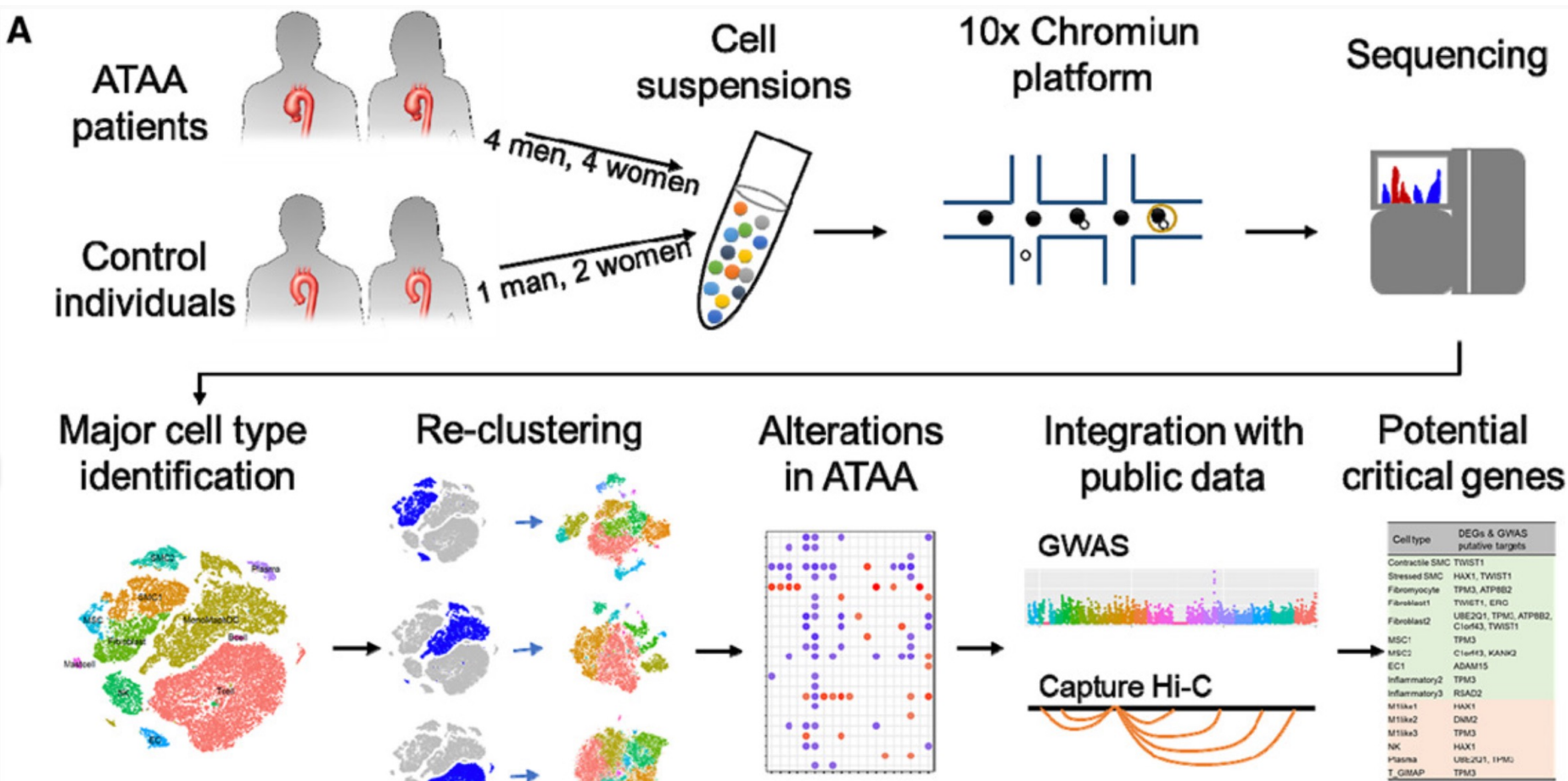
Limitations of 10x/single nucleus RNA-seq

- mRNA is nuclear, so intermediate in spliceosome processing
 - No isoform determination
 - Must map to entire gene body
- RNA is specifically nascent, non-cytoplasmic
 - Limiting factor for some trajectory analysis
- Need good quality isolated nuclei to avoid clogging 10x chip
 - For some tissues, debris removal is necessary to isolate intact nuclei
 - Iodixanol, ultracentrifugation
- For multi-omic assays, it can be hard to balance ATAC-seq quality with RNA-seq quality

Single-Cell Transcriptome Analysis Reveals Dynamic Cell Populations and Differential Gene Expression Patterns in Control and Aneurysmal Human Aortic Tissue

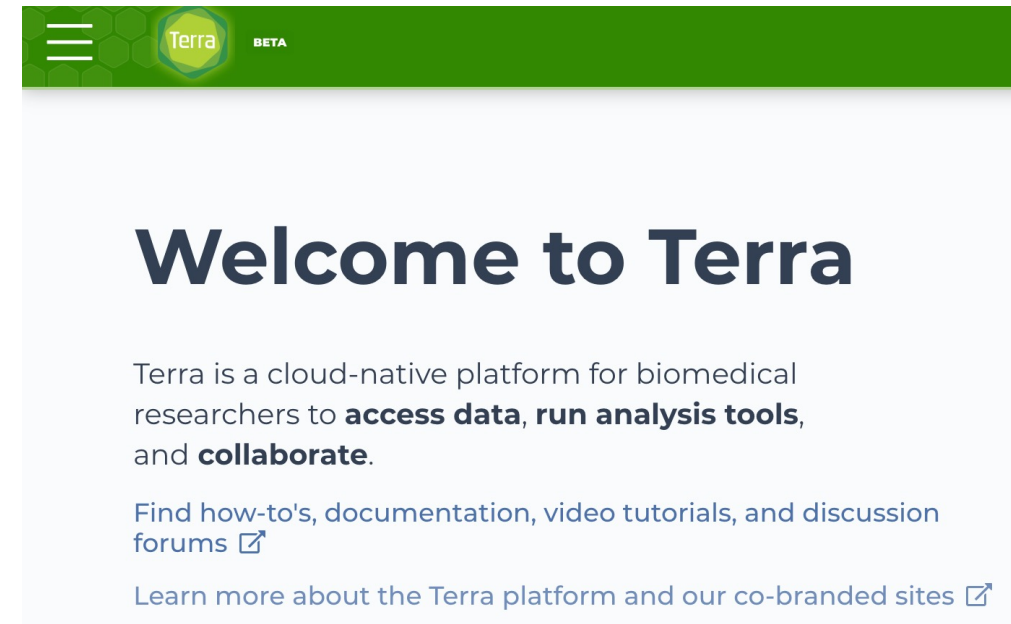
Yanming Li, Pingping Ren, Ashley Dawson, Hernan G. Vasquez, Waleed Ageedi, Chen Zhang, Wei Luo, Rui Chen, Yumei Li, Sangbae Kim, Hong S. Lu, Lisa A. Cassis, Joseph S. Coselli, Alan Daugherty, Ying H. Shen , Scott A. LeMaire 

Originally published 5 Oct 2020 | <https://doi.org/10.1161/CIRCULATIONAHA.120.046528> | Circulation. 2020;142:1374–1388






Preview/overview of the coding session tomorrow

- We will be using a workspace on Terra →
 - https://app.terra.bio/#workspaces/gro-share-seq-computational/useful_notebooks_for_teaching_public_analysis/launch/2021_09_human_aorta_single_cell_cumulus_scanpy.ipynb
 - We will be examining the human aorta dataset on human aorta reprocessed using the pipeline Cumulus
- We also have a single cell ArchR workspace for ATAC analysis
 - https://app.terra.bio/#workspaces/gro-share-seq-computational/useful_notebooks_for_teaching_public_analysis/launch/2022_02_25_ArchR_tutorial_atac.ipynb



Article | Published: 27 July 2020

Cumulus provides cloud-based data analysis for large-scale single-cell and single-nucleus RNA-seq

Bo Li , Joshua Gould, Yiming Yang, Siranush Sarkizova, Marcin Tabaka, Orr Ashenberg, Yanay Rosen, Michal Slyper, Monika S. Kowalczyk, Alexandra-Chloé Villani, Timothy Tickle, Nir Hacohen, Orit Rozenblatt-Rosen  & Aviv Regev 

Nature Methods 17, 793–798 (2020) | [Cite this article](#)

7876 Accesses | 13 Citations | 74 Altmetric | [Metrics](#)

Thank you! And see you tomorrow!

- Many thanks to:
 - The AHA GPM Council
 - **Lisa Wilsbacher**
 - **Matt Naylor**
 - Charles Epstein, for letting me take time to teach



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Vascular Discovery: From Genes to Medicine 2023