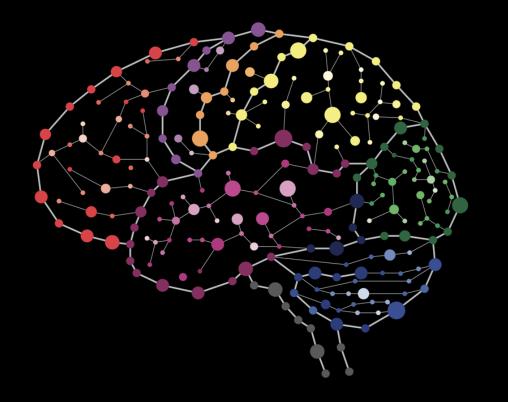


# Krembil Centre for Neuroinformatics

Using big data, artificial intelligence and brain modelling to

fundamentally change our understanding of mental illness.



#### **SUMMER SCHOOL 2020**

Day 3

Gene expression: DNA to cells

Part 2: single-cell gene expression and integration with bulk

# Today's Agenda





**Day 3**Genetics and

Transcriptomics

9:00 am -10:30 am Basics of Genotype, Central Dogma, GWAS, and Polygenic Risk Scores

Dan Felsky

10:45 am - 12:15 pm

Transcriptomics at the Single-Cell Level
Shreejoy Tripathy

1:00 pm -2:30 pm WORKSHOP: Intro to Transcriptomic Data Types, Including Single-Cell and Bulk RNAseq Sonny Chen

2:45 pm -4:15 pm WORKSHOP: Explore Cellular Changes in Major Depression Using Bulk and Single-Cell RNAseq Data

Keon Arbabi

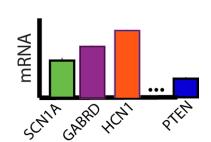
# Bridging brain scales



Genetics



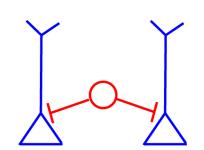
**Gene Expression** 



Cellular Physiology



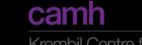
**Neural Circuits** 



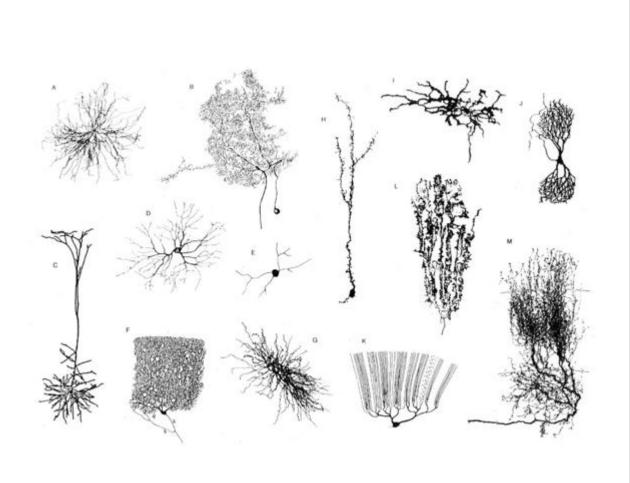
**Traits & Behaviors** 

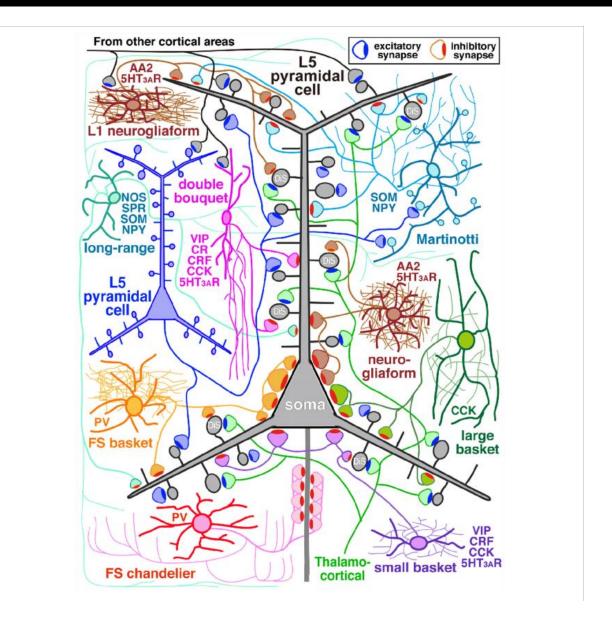


# There are many types of cells in the brain



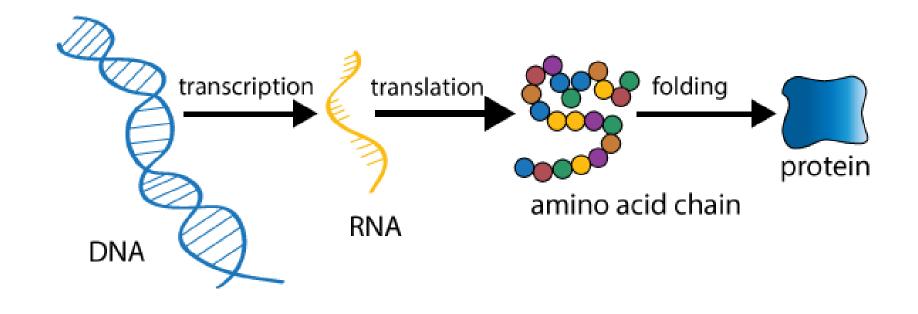
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# Central dogma of molecular biology



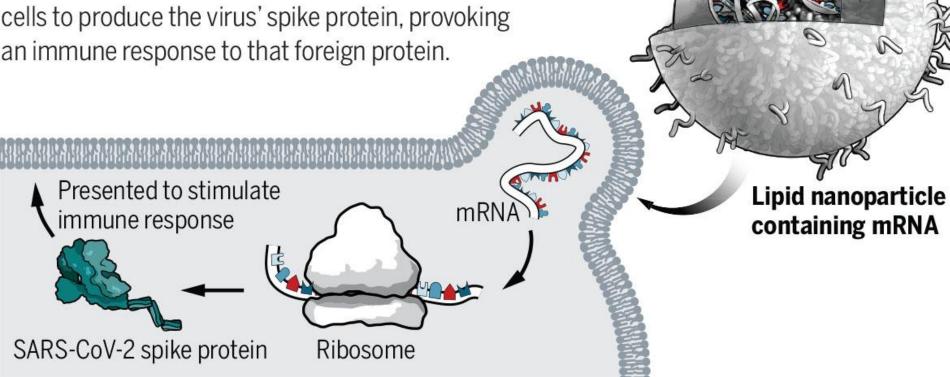


# mRNA vaccines illustrate the magic of RNA

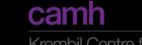


### **Special delivery**

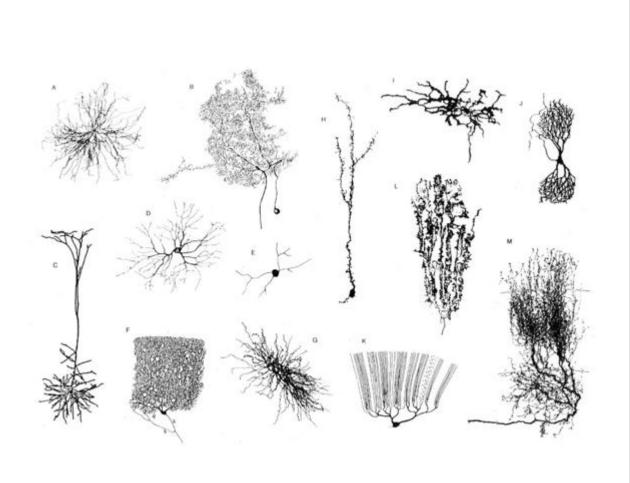
Two apparently successful coronavirus vaccines use fat bubbles called lipid nanoparticles to deliver messenger RNA (mRNA) to cells. Once there, the mRNA directs cells to produce the virus' spike protein, provoking an immune response to that foreign protein.

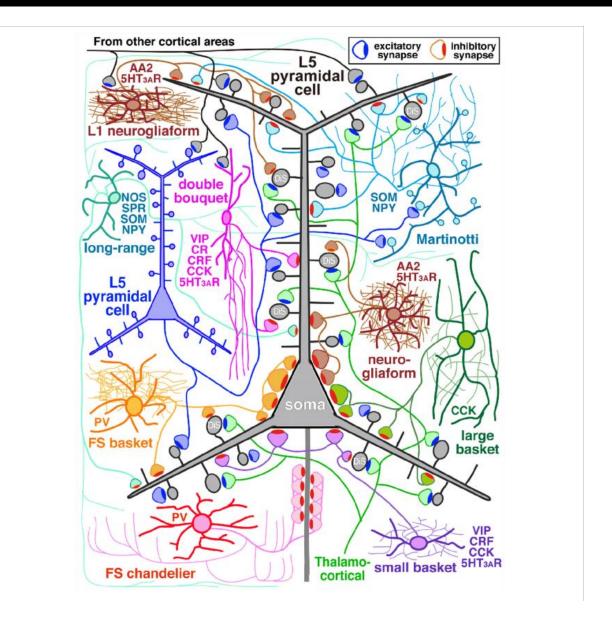


# There are many types of cells in the brain



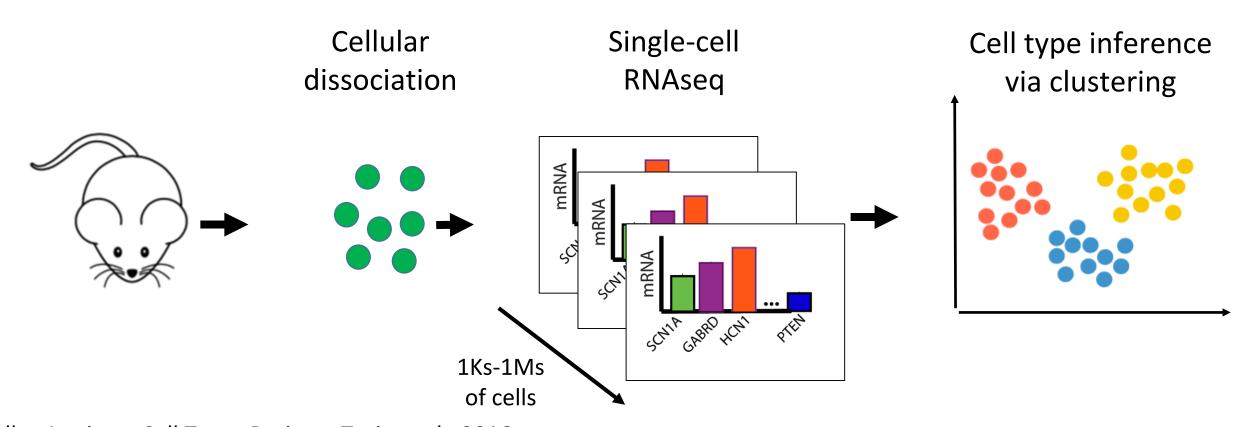
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# Single-cell RNA-sequencing





Allen Institute Cell Types Project; Tasic et al., 2016; 2018

Mousebrain.org; Zeisel et al., 2018

Tabula Muris Consortium, many others

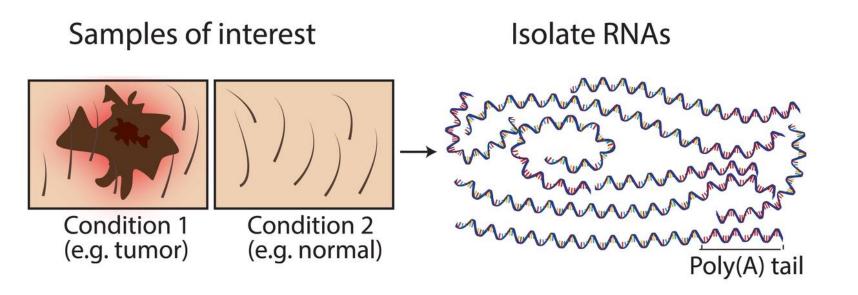
# RNA sequencing in a nutshell



RNA-Seq (named as an abbreviation of "RNA sequencing") is a
particular technology-based sequencing technique that reveals the
presence and quantity of RNA in a biological sample at a given
moment, analyzing the continuously changing cellular
transcriptome.

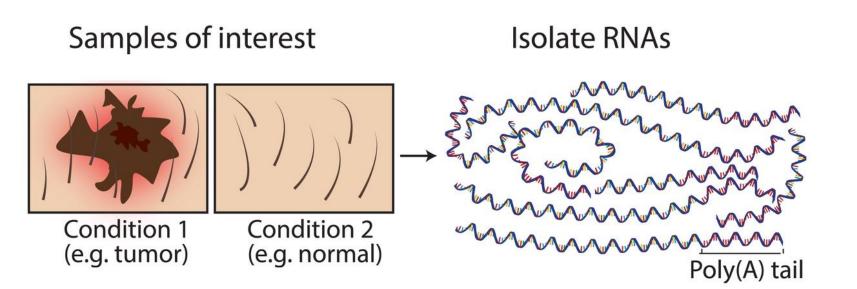
# A brief overview of RNA sequencing



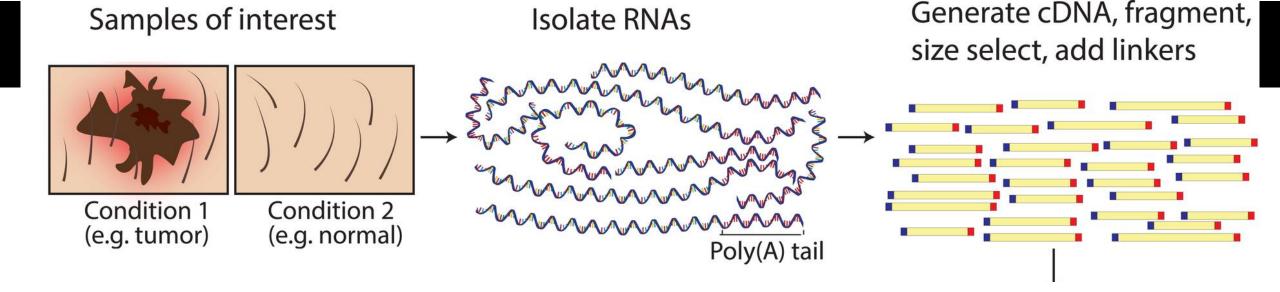


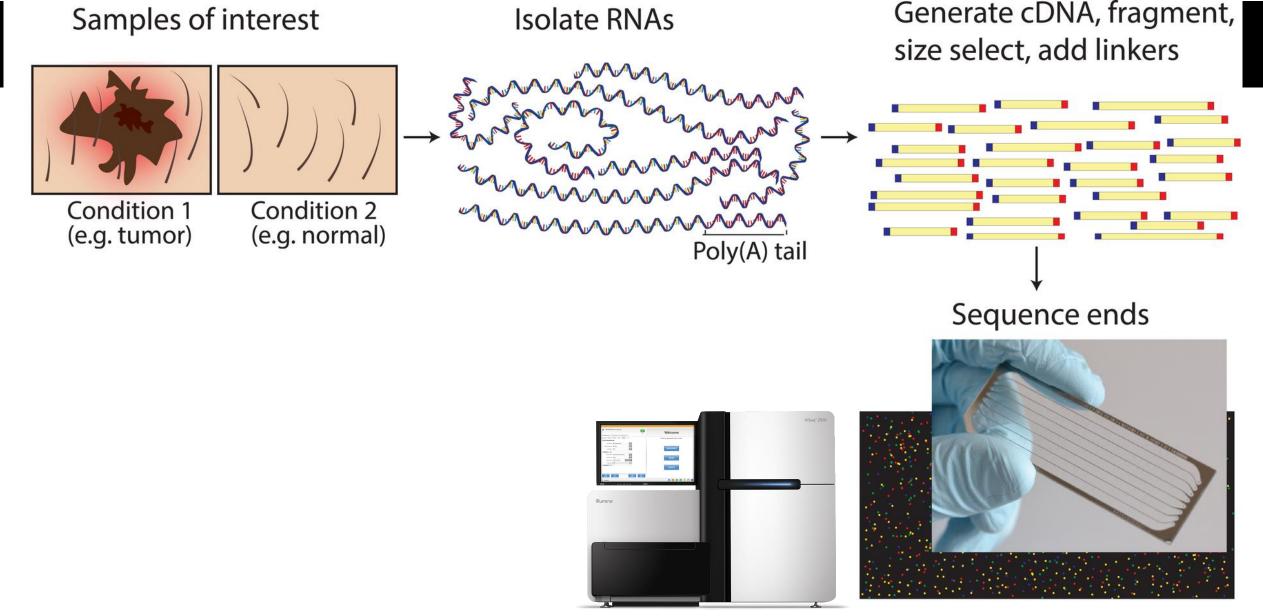
# A brief overview of RNA sequencing







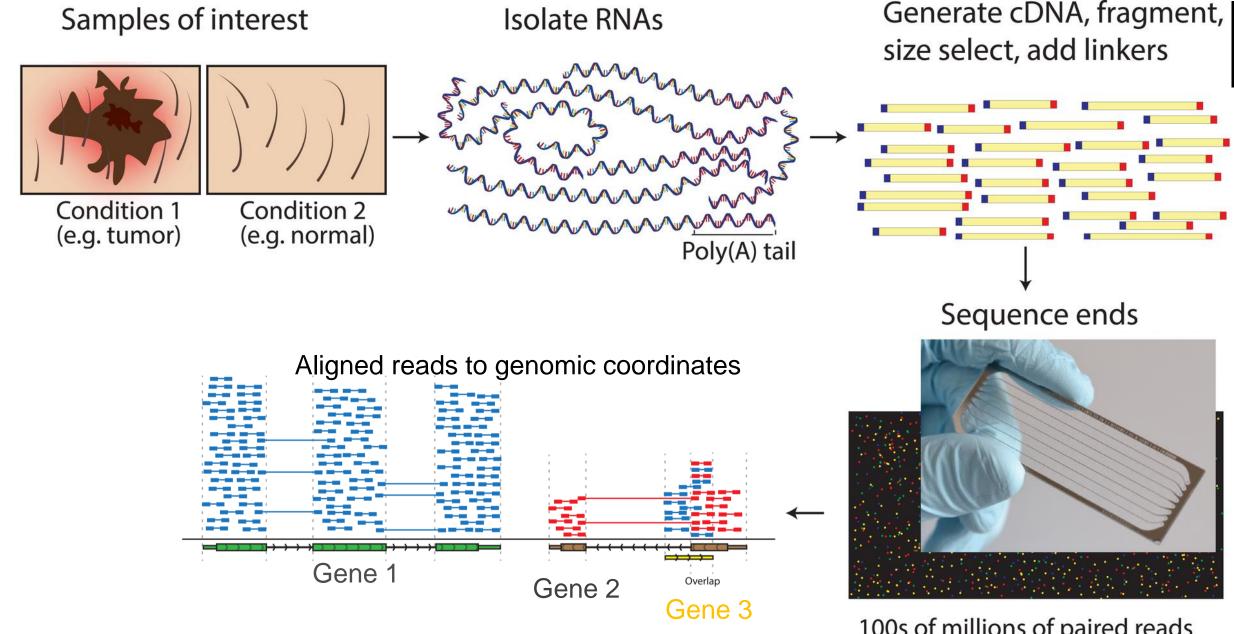




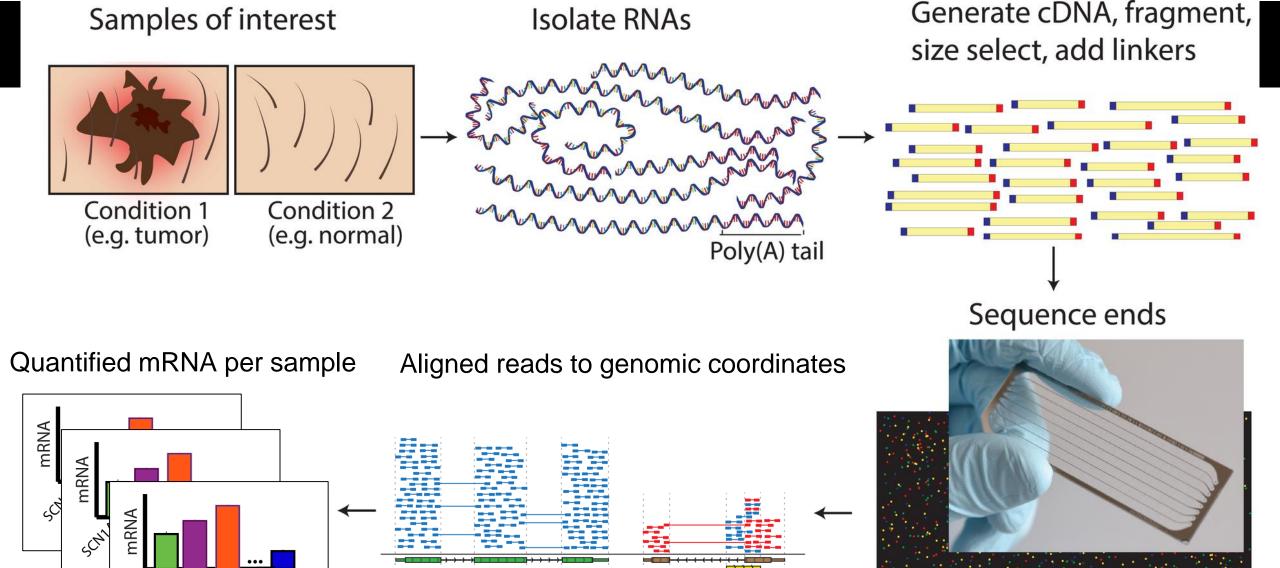
100s of millions of paired reads 10s of billions bases of sequence

# Youtube video illustrating DNA sequencing





100s of millions of paired reads 10s of billions bases of sequence



Gene 2

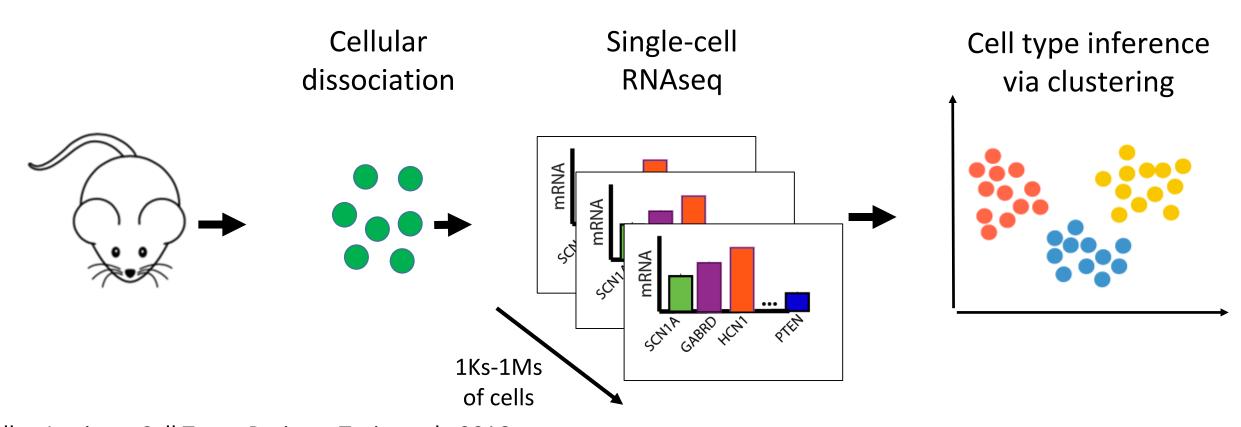
Gene 3

Gene 1

100s of millions of paired reads 10s of billions bases of sequence

# Single-cell RNA-sequencing





Allen Institute Cell Types Project; Tasic et al., 2016; 2018

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Tabula Muris Consortium, many others

# Gene expression count matrix and metadata rembil Centre for Neuroinformatics

Gene expression "count matrix"

	Cell1	Cell2	 CellN
Gene1	3	2	13
Gene2	2	3	1
Gene3	1	14	18
• • •			
GeneM	25	0	0

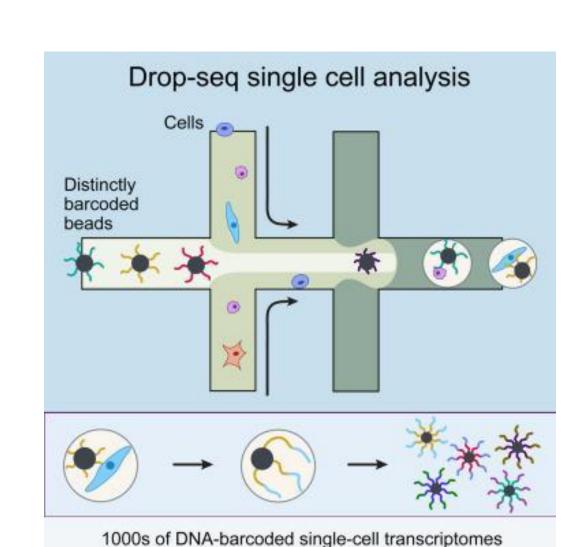
#### Sample metadata matrix

Cell ID	Subject ID	Expt group	Cell Type	
Cell 1	Patient 1	Control	Astrocyte	
Cell 2	Patient 1	Control	Neuron	
Cell N	Patient N	Case	Neuron	



Neuroinformatics

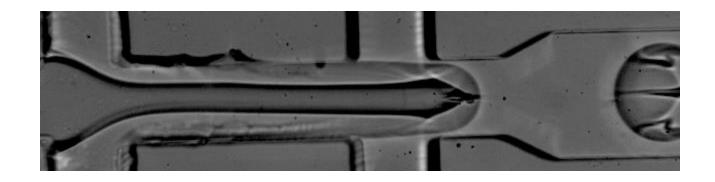
# Microfluidics enables single-cell omics





Neuroinformatics

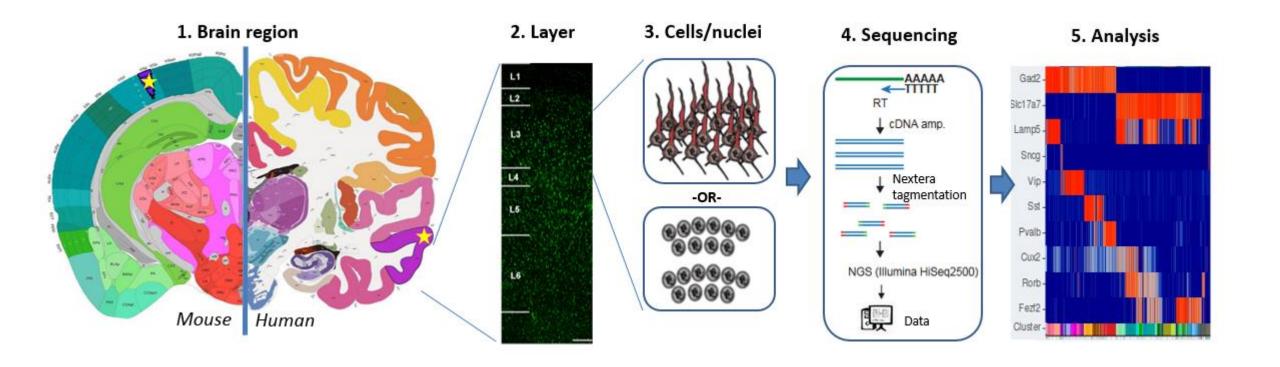
# Microfluidics enables single-cell omics



# scRNAseq rules of thumb



- Single-cell technologies are sampling the transcriptome
  - Most datasets are only sampling a fraction of the most highly expressed genes
- Nuclei are not cells
  - Especially important for large cells, like neurons
- Garbage in, garbage out
  - scRNAseq is less forgiving of degraded mRNA than bulk RNAseq methods

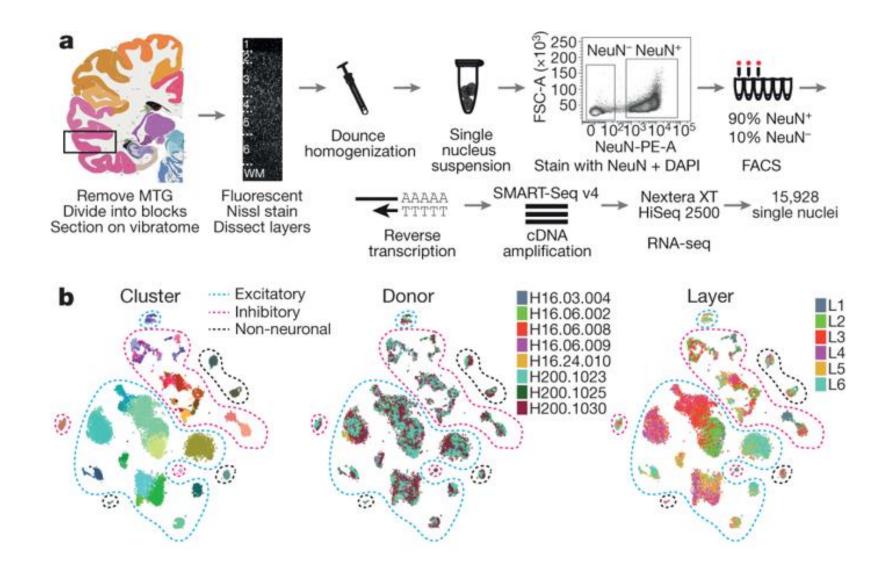




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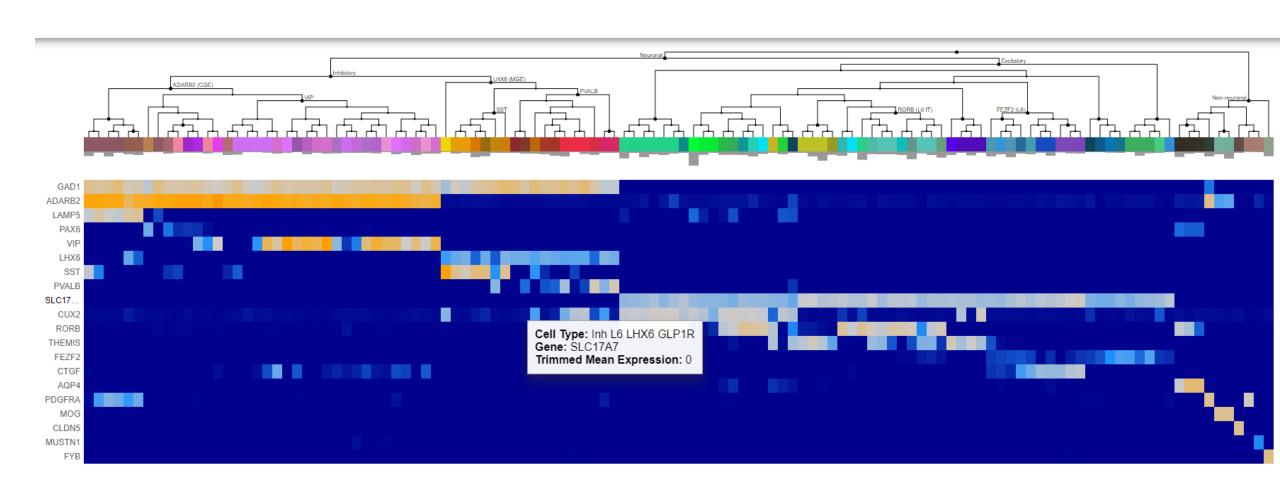
# Human cortex single-nucleus RNAseq

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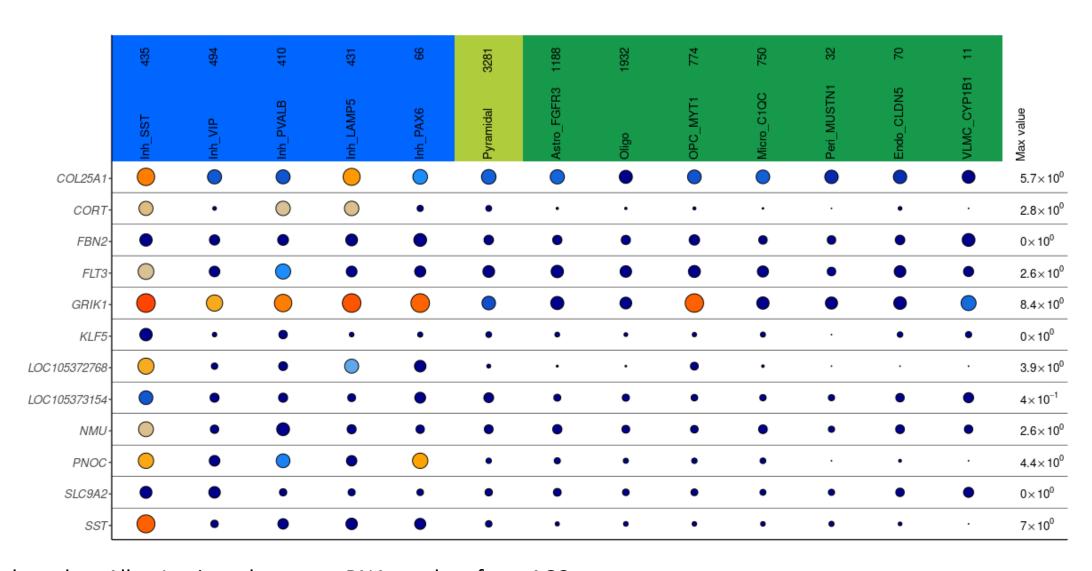
# Allen Institute Cell Types Resources

http://celltypes.brain-map.org/



# SST cell type markers based on snRNAseq





Markers based on Allen Institute human snRNAseq data from ACC reanalyzed by Sonny Chen, post-doc in Tripathy Lab

# Tutorial: Applications of single-cell RNAseq



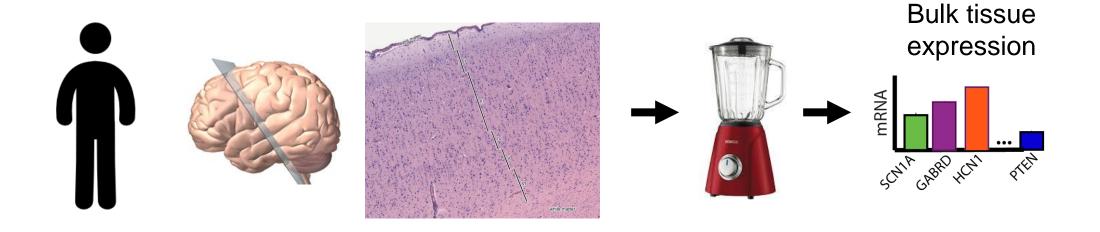
- Scientific concepts:
  - Become familiar with brain gene expression and brain cell types
  - Learn about cell types and concept of marker genes specific to cell types
- Specific learning objectives
  - Gain familiarity with scRNAseq data, preprocessing, and analysis
  - Gain familiarity with dimensionality reduction

# Break for questions



# Studying the post-mortem human brain

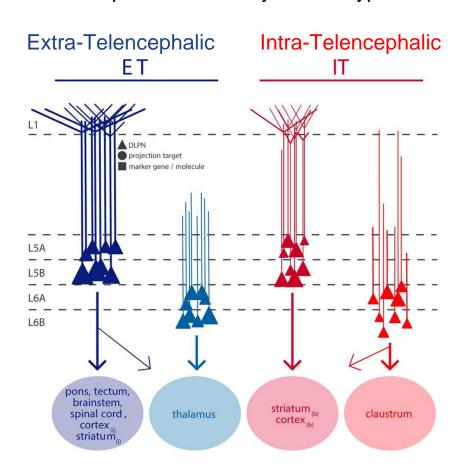




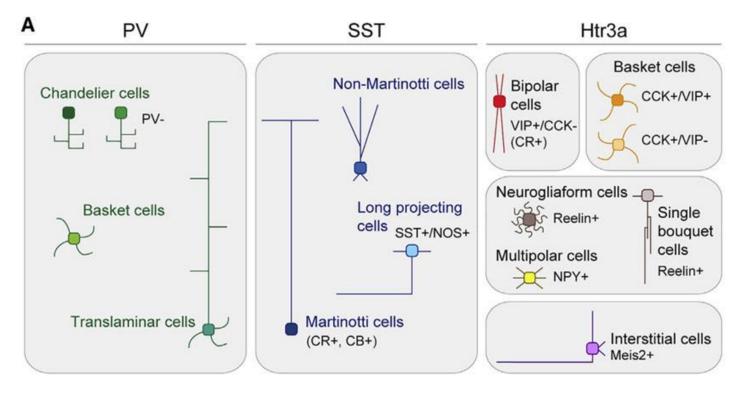
# Neocortical cell type diversity



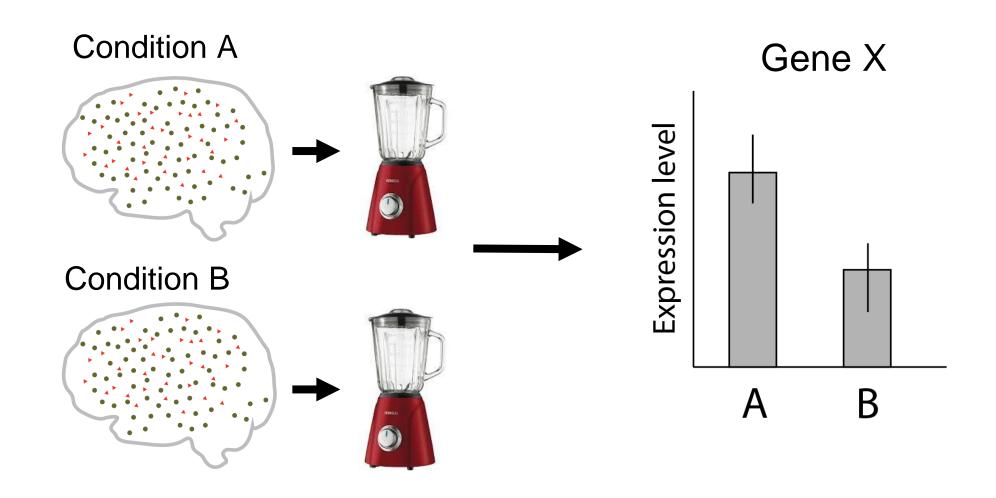
#### Examples of excitatory neuron types



#### Diversity of major inhibitory neuron types



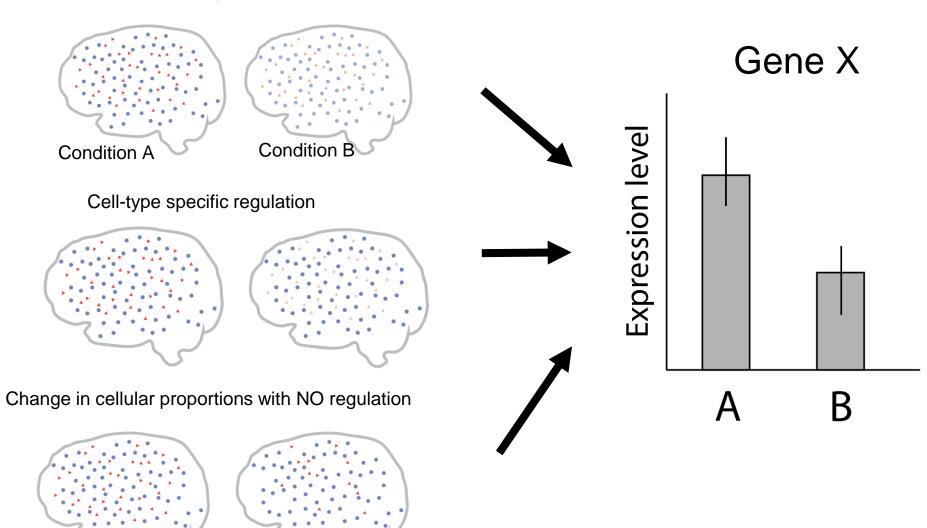
# Bulk tissue transcriptomes complicate interpretation



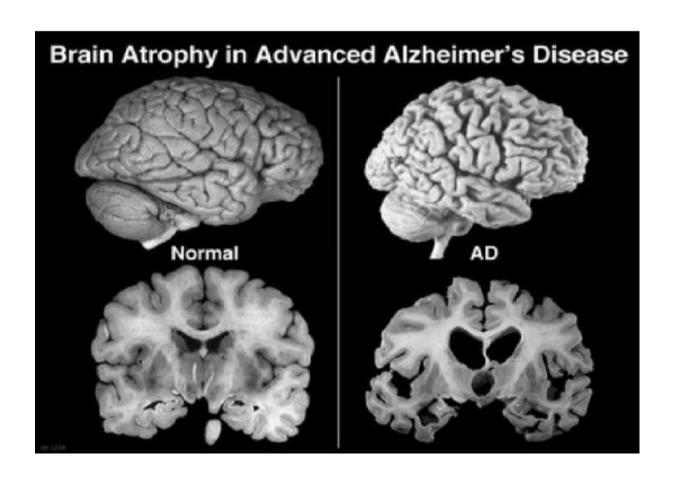


## Bulk tissue transcriptomes complicate interpretation

#### **Broad regulation**



# Cell type changes in Alzheimer's disease



 How does AD pathology impact specific cell types in the neocortex?

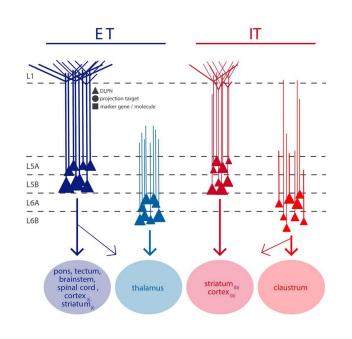


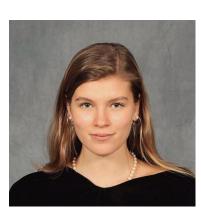


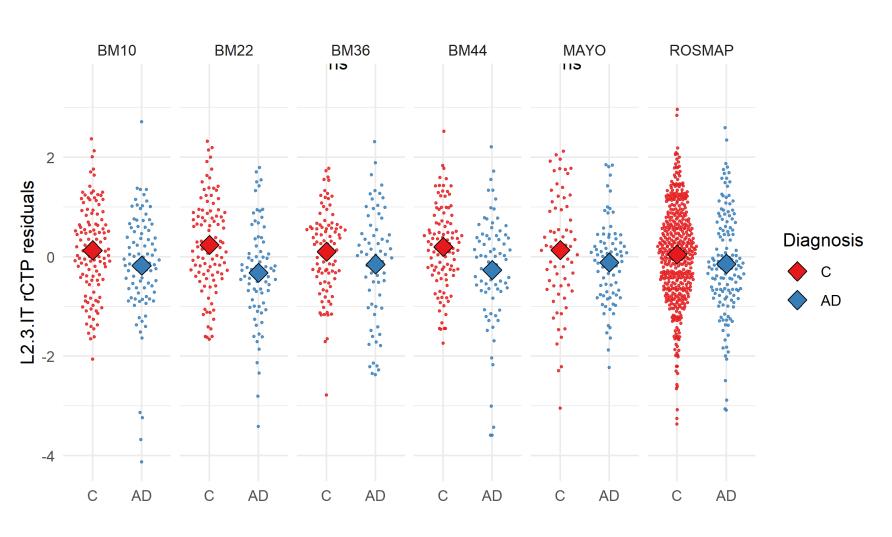
Micaela Consens (co-advised with Dan Felsky) 4<sup>th</sup> year Undergrad in Bioinformatics and Computer Science

# Cell type changes in Alzheimer's disease





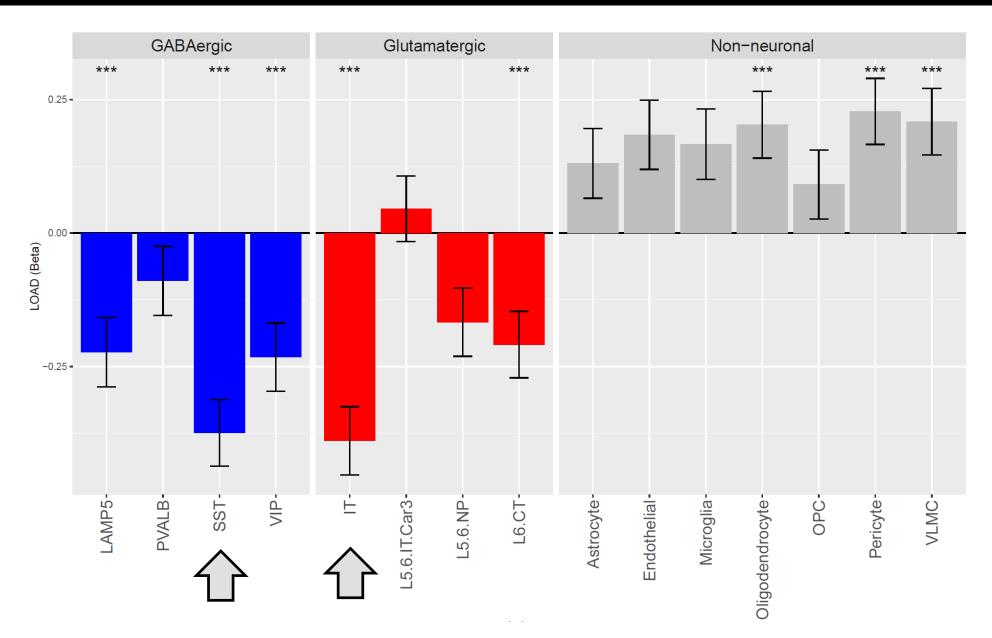




Micaela Consens (co-advised with Dan Felsky)
4th year Bioinformatics and Computer Science Ugrad

# Cell type changes in Alzheimer's disease

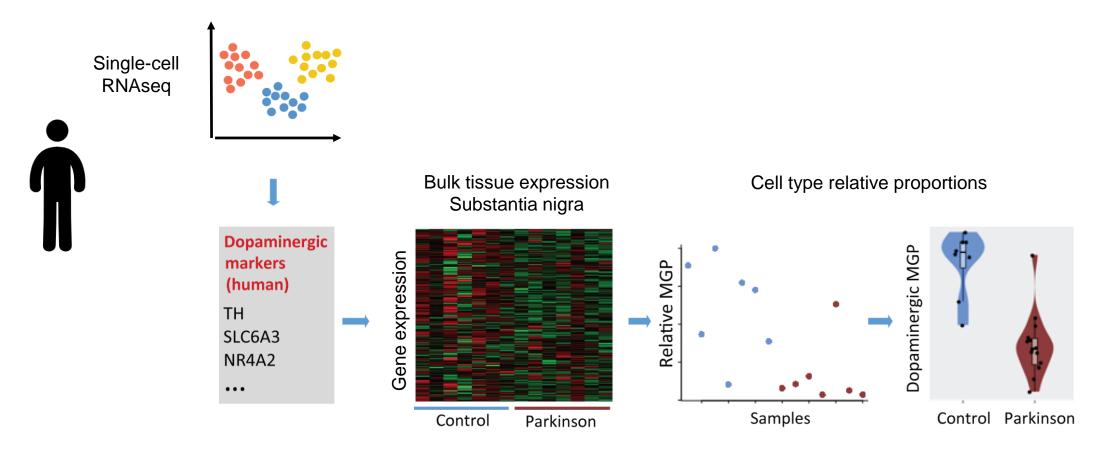






- Scientific concepts:
  - Learn how scientists use post-mortem brain gene expression to infer cellular changes in psychiatric conditions
  - Learn about cell type proportions and how these might differ in neuropsychiatric disease
- Specific Learning objectives:
  - Learn how to analyze brain bulk gene expression data
  - Learn about cell types and concept of marker genes specific to cell types

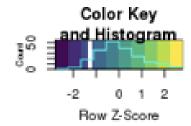
# Cellular abundance estimation using Marker Gene Profiles

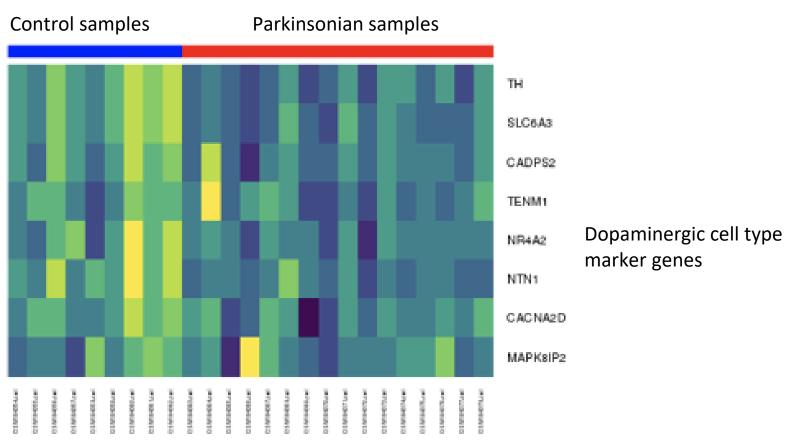


Methodology: Mancarci et al., *eNeuro*, 2017 Application to SCZ and BD: Toker et al, *Biol. Psych., 2018* 

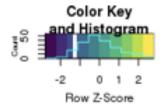
### Marker gene profile estimation

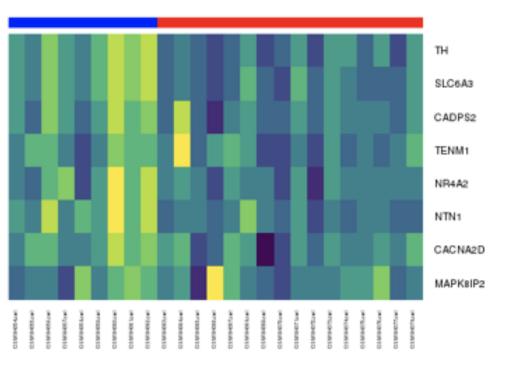


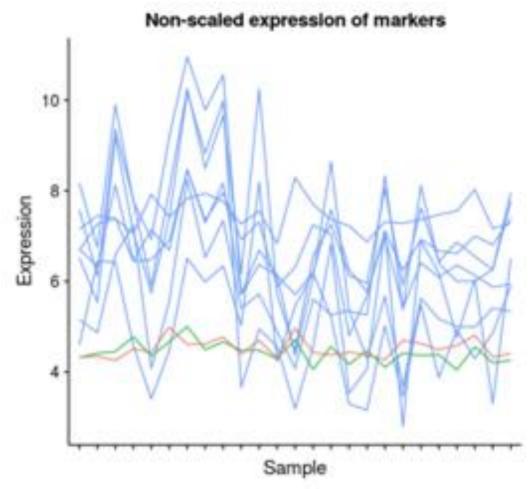




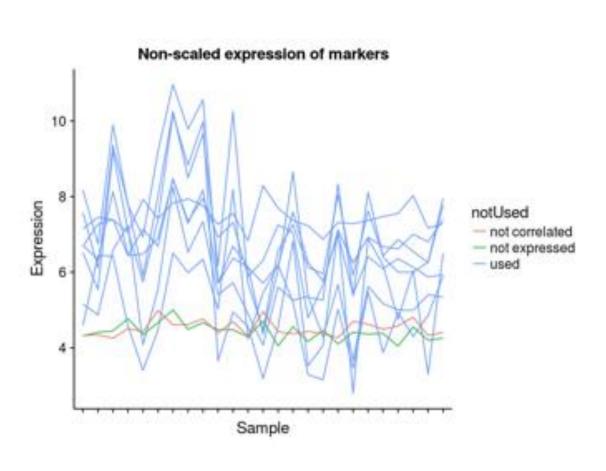
### Cell type markers are correlated in bulk samples

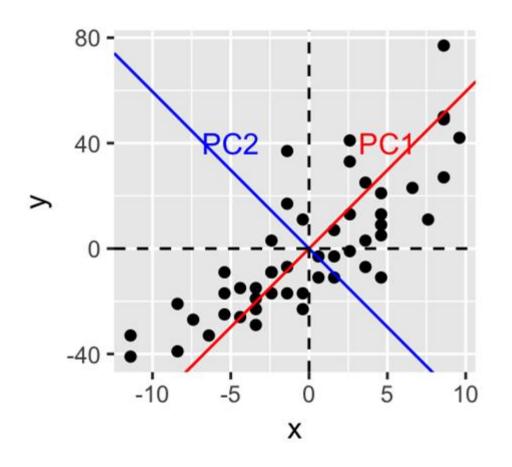






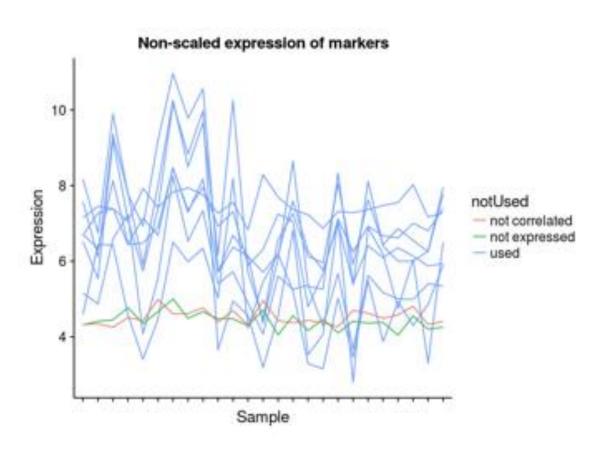


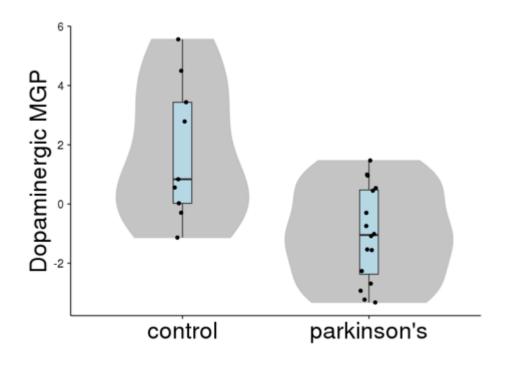




### Marker gene profile estimation







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### Break for questions



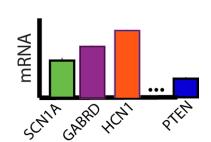
### Bridging brain scales



Genetics



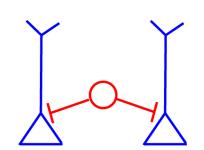
**Gene Expression** 



Cellular Physiology



**Neural Circuits** 

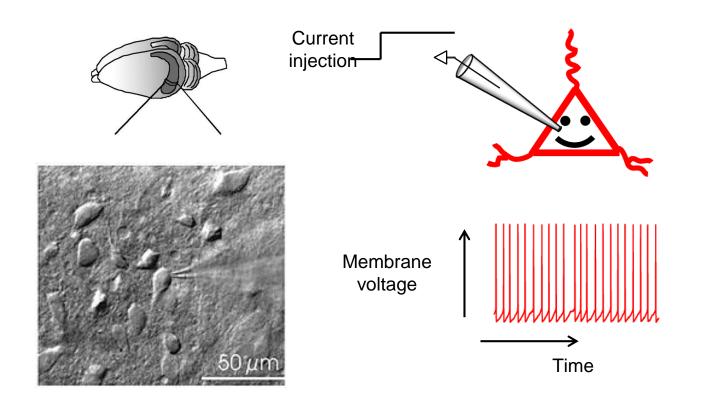


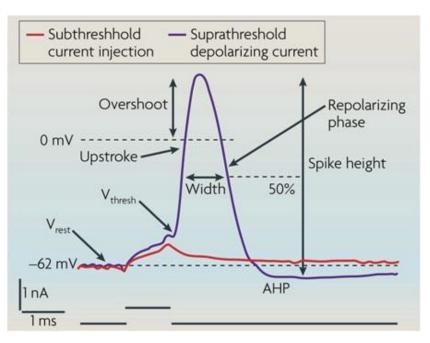
**Traits & Behaviors** 



### Intracellular electrophysiology



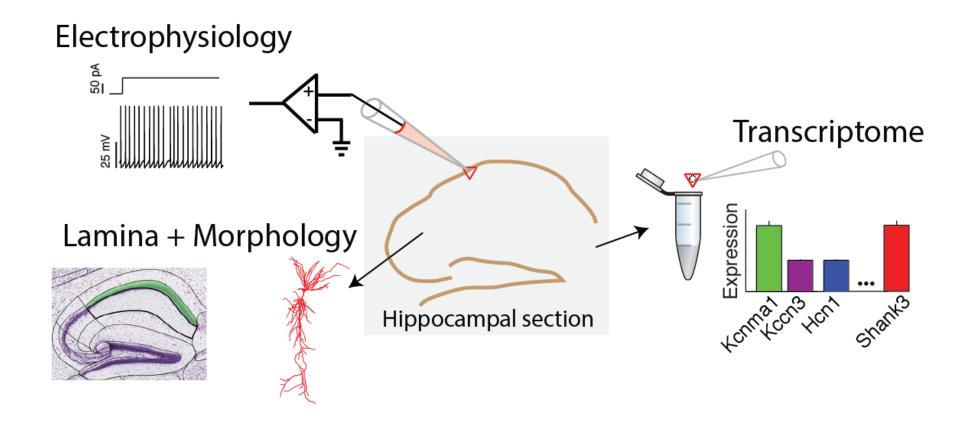




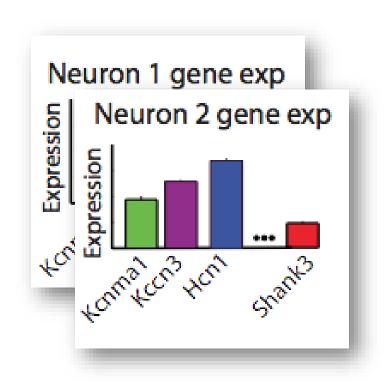
Nature Reviews | Neuroscience

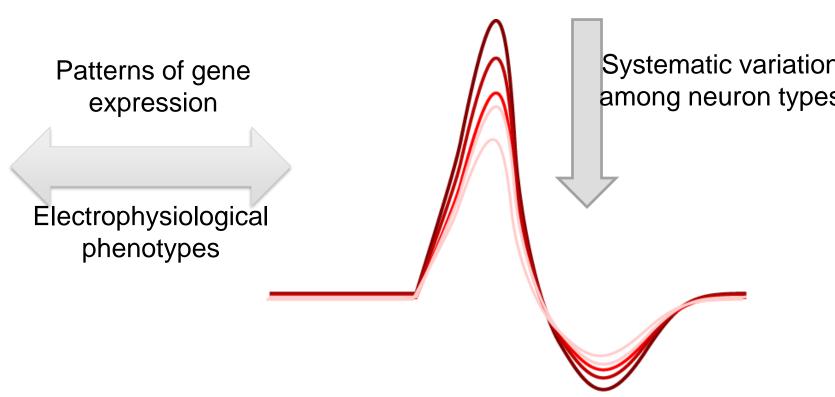
### Patch-seq combines scRNAseq, ephys, and morphology





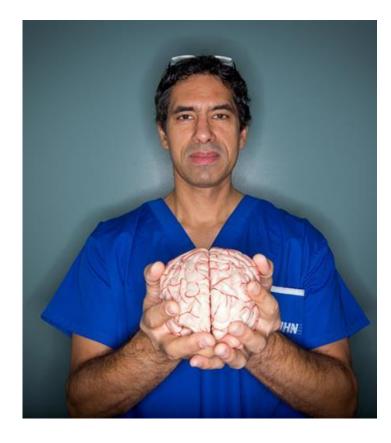
# Identifying genes underlying diversity in cell phenotypes bil Centre for





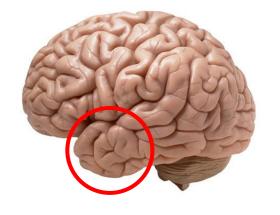
#### Neurosurgery enables access to cells in the human brain





Dr. Taufik Valiante
University Health Network and
Krembil Research Institute



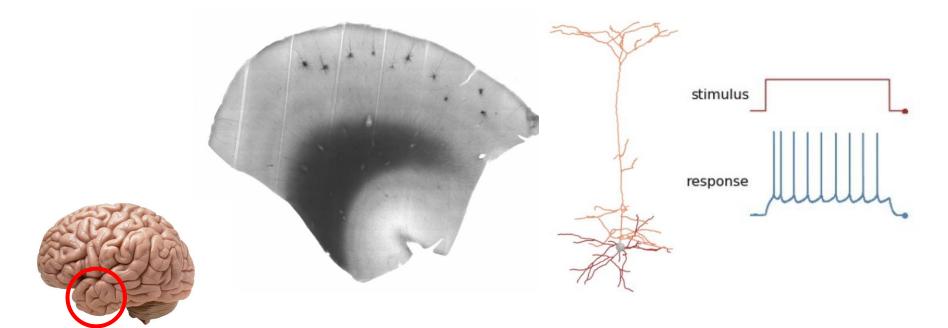




### Exploring neuron diversity in the human brain



#### Human neurosurgical tissue



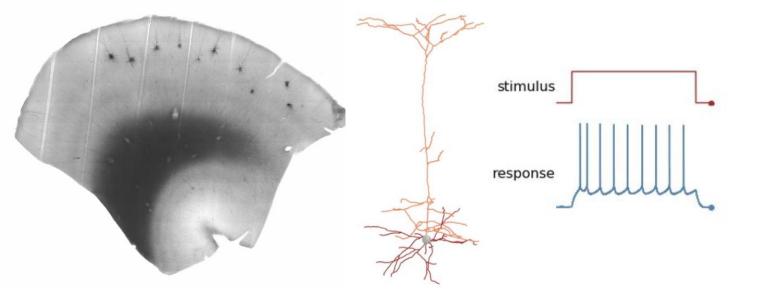


Homeira Moradi, Valiante Lab Moradi et al, *Nat Comm.* 2021

### Exploring neuron diversity in the human brain



#### Human neurosurgical tissue

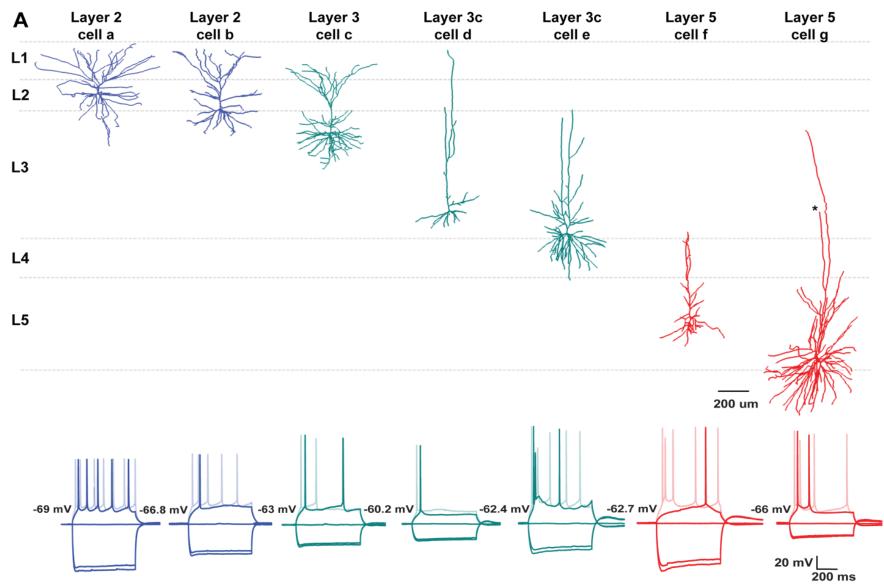


#### **Major Questions**

- How are human neurons different...
  - from one another?
  - from those in other species?
  - in aging?
  - because of genetics?
- Can we use these data to build "template" circuits of the human brain in health and in disease?
  - In collaboration with Hay Lab and KCNI

### Exploring neuron diversity in the human brain





### Towards personalized mental health

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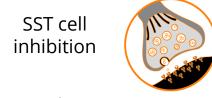
Clinical populations



Genomics sampling (DNA, RNA, etc.)

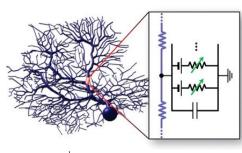


Inference of activity in brain cells





## Personalized brain models



 $\frac{dm}{dt} = \alpha_m(V_m)(1 - m) - \beta_m(V_m)m$ 

 $\frac{dh}{dt} = \alpha_h(V_m)(1 - h) - \beta_h(V_m)h$ 

### Acknowledgements

camh

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