Bringing single cell genomics closer to the clinic for patients with leukemia

Scott Furlan, MD (he/his) - 7-29-22 Assistant Professor, FHCRC



Overview and Disclosures

- Introduce the concept of measurable residual disease (MRD) as told from the of one patient
- Provide rationale for the use of single cell genomics to potentially improve MRD assessment after transplant
- Share our preliminary data using single cell RNA seq in patients with relapsed leukemia
 - Highlight novel molecular and computational approaches
- Broader applicability to understanding the biology of acute leukemia and mechanisms of relapse

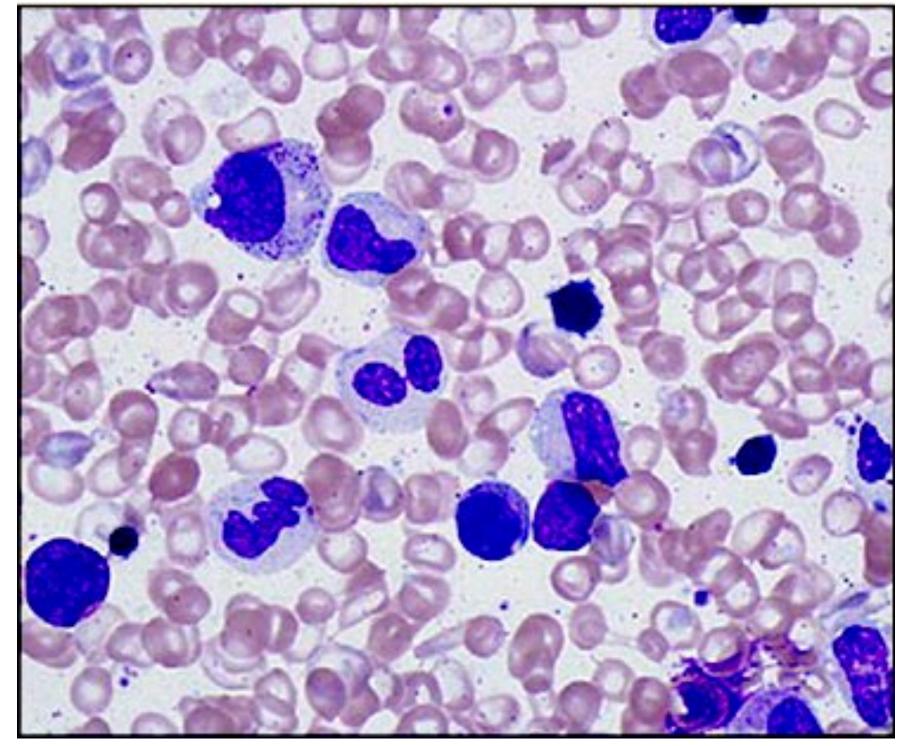


No Conflicts of Interest or relevant disclosures!



Thank you

- 12 yo with a history of Myelodysplastic Syndrome diagnosed in late 2018
- Evolved to Acute Myeloid Leukemia (AML) shortly thereafter
- Matched Unrelated Bone Marrow Transplant in April 2019
- Two years later (April 2021), patient developed low blood counts on routine monitoring -> AML

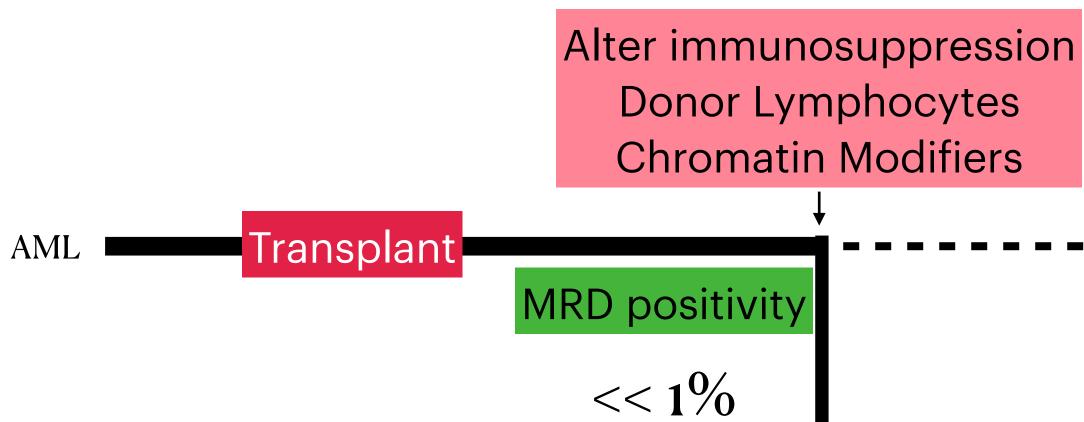


Nguyen P.L. Hematol Oncol Clin North Am. 2009; 23: 675-691

- considered for a second transplant using cord blood.
- Flow cytometry 1-2%
- But his TP53mut (R248Q), IDH1mut (R132C) > 10% VAF.

After relapse, underwent reinduction chemotherapy and was being

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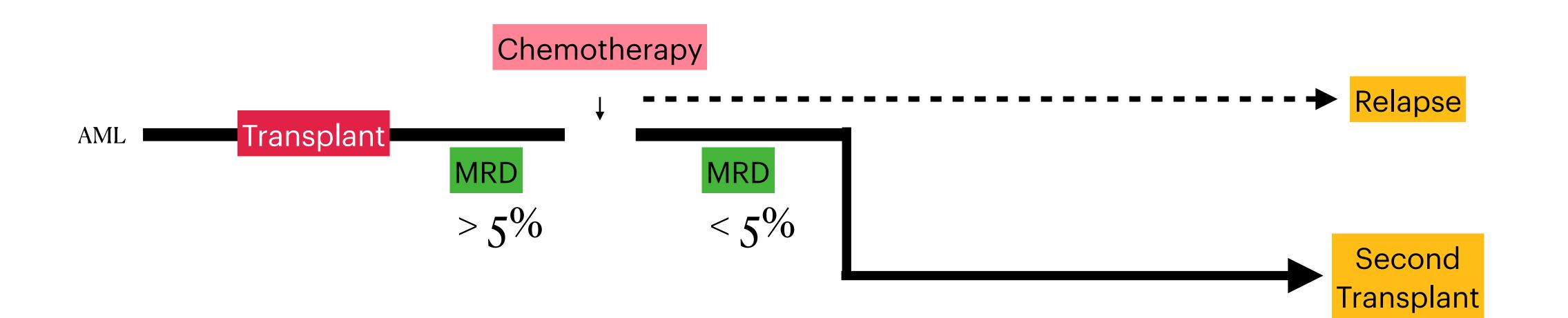


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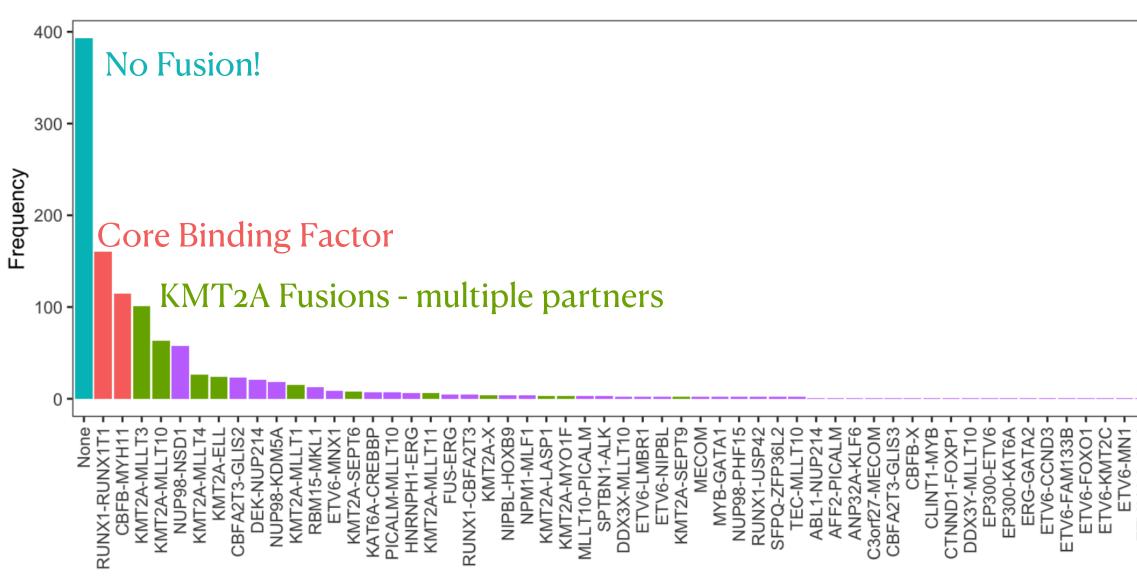


• PCR based tests: Fusion transcripts

- High sensitivity, but are limited in their applicability (A priori)

The 'long tail' of acute leukemia is a problem for molecular assays

PCR • PCR based tests: Fusion transcripts



High sensitivity, but are limited in their applicability (A priori)

1210 Pediatric AML patients COG Trial - AAML1031 Cytogenetics + Bulk RNA sequencing

E I VO-MINI - E I VO-MINI - E I VO-MINI - E EWSR1-FEV - EVSR1-FEV - EWSR1-FEV - EVSR1-FEV - EWSR1-FEV - EVSR1-FEV - EWSR1-FEV





The 'long tail' of acute leukemia is a problem for molecular assays

- PCR
 PCR based tests: Fusion transcripts
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- **NGS** Bulk NGS: Mutations
 - Similarly limited, less sensitive



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- **MFC** Single cell methods Multiparameter flow cytometry (MFC):
 - More broadly applicable
 - Limited in sensitivity
 - Challenging to standardize and difficult to interpret

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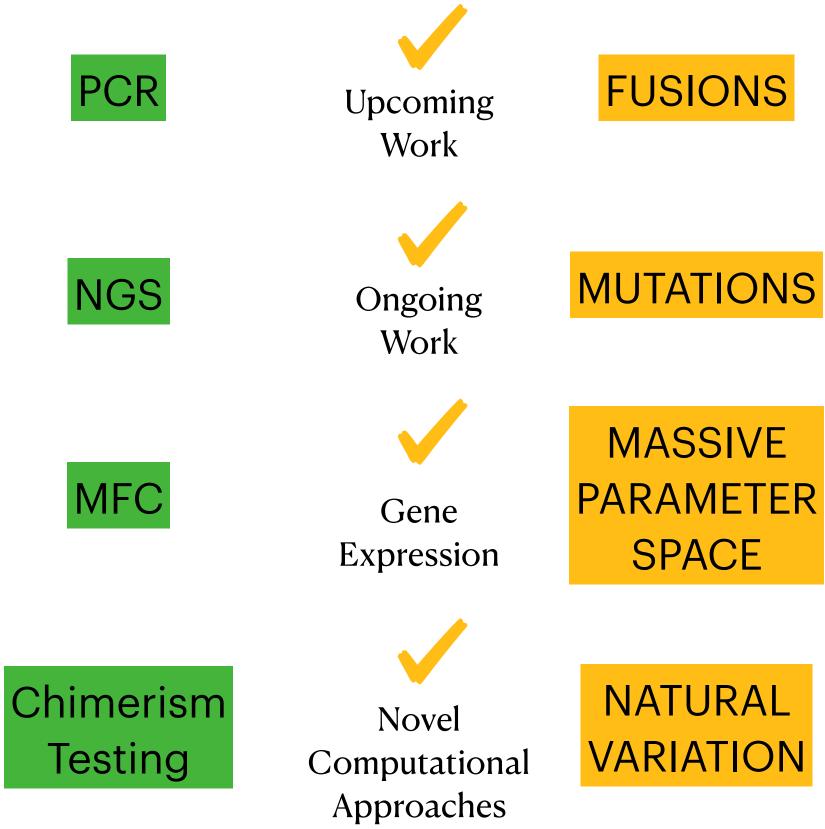


Chimerism

Testing

- **MFC** Single cell methods Multiparameter flow cytometry (MFC):
 - More broadly applicable
 - Limited in sensitivity
 - Challenging to standardize and difficult to interpret
 - Not currently sensitive enough (+/- 5%)

How can we improve?



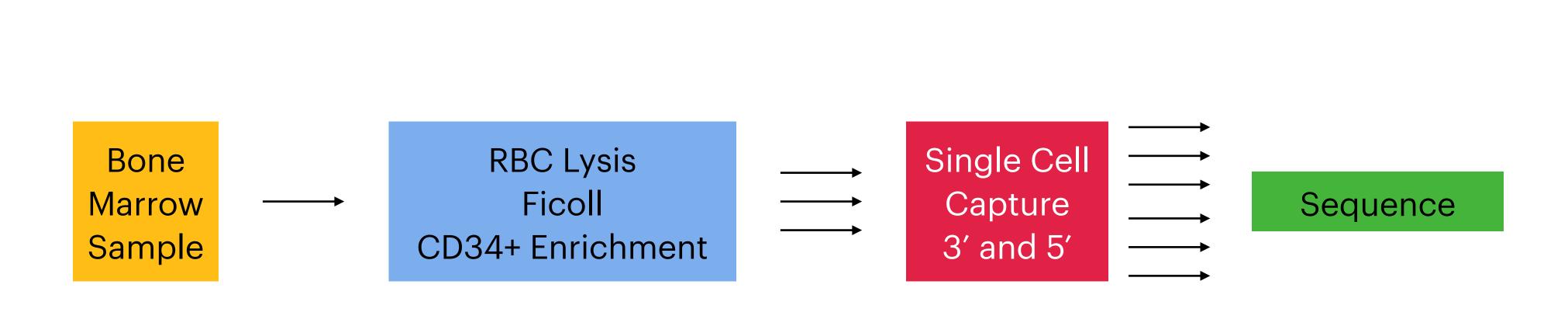




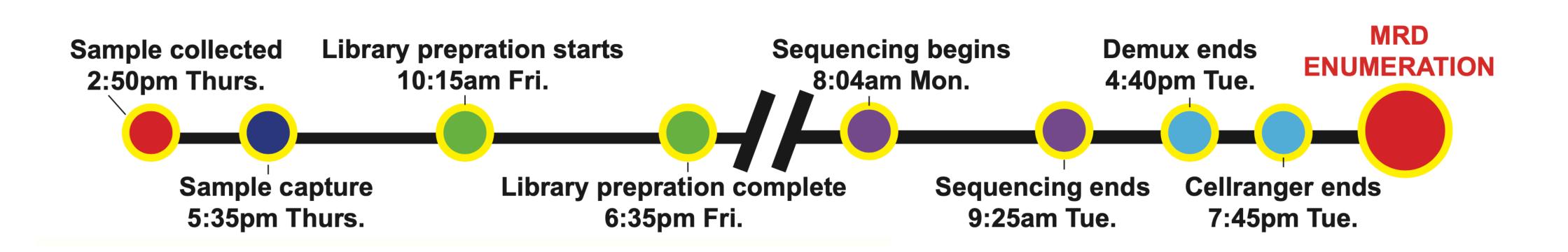
SINGLE CELL RNA SEQUENCING MAY BE ABLE TO MEET THIS NEED!

> **Deliverable:** More CONFIDENT assessment of MRD

Overview of our first experiment

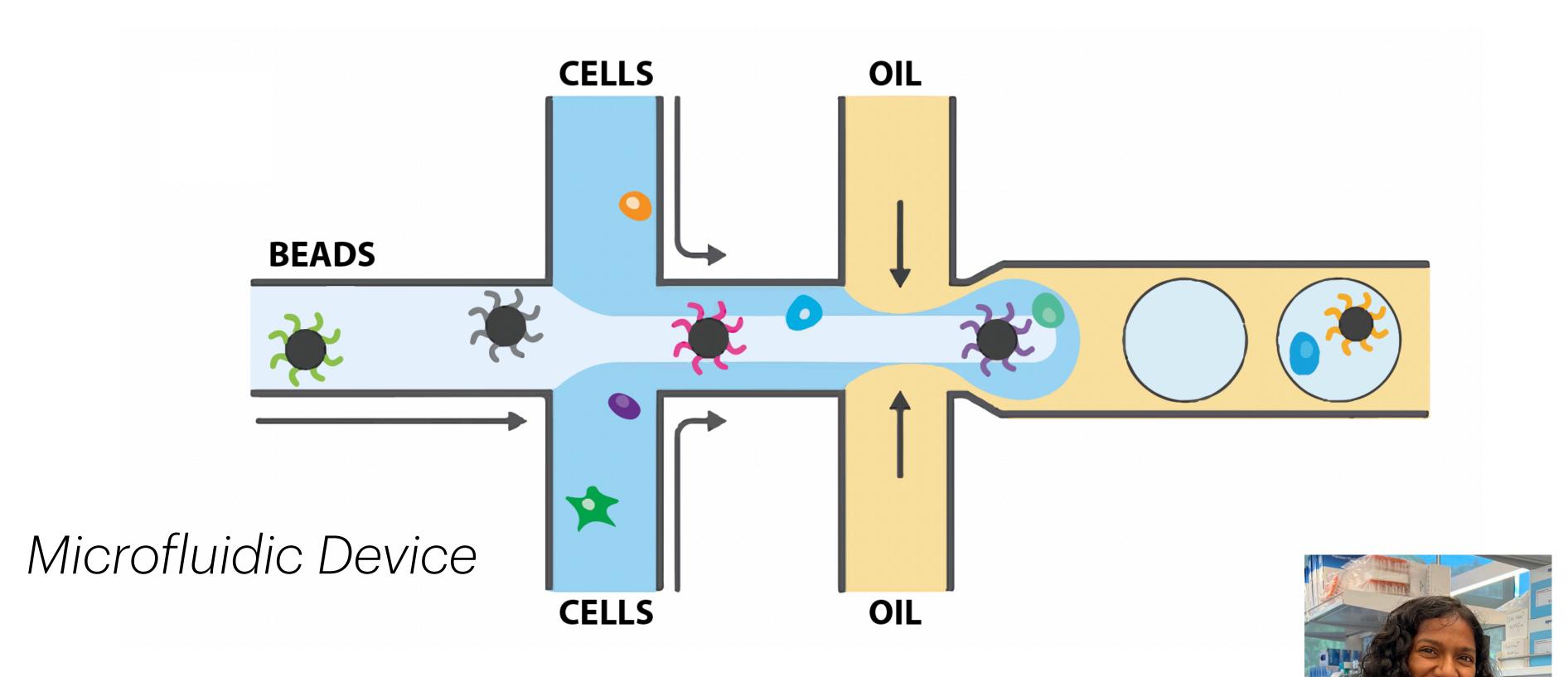


A note on feasibility

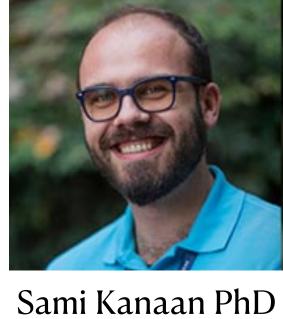




droplet partitioning single-cell RNA sequencing



cell capture



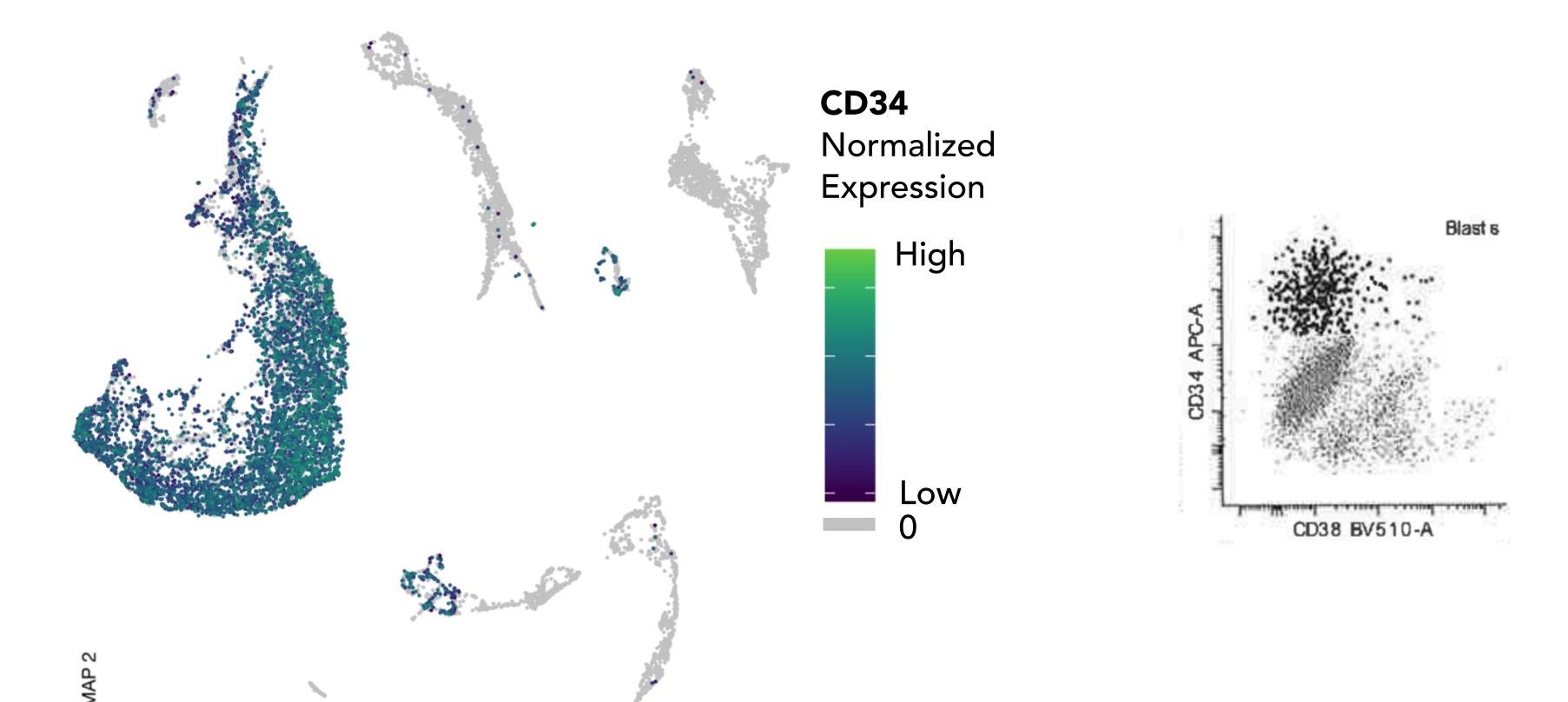
Shruti Bhise MS

Higher than expected numbers of cells expressing CD34

RBC Lysis Ficoll CD34+ Enrichment

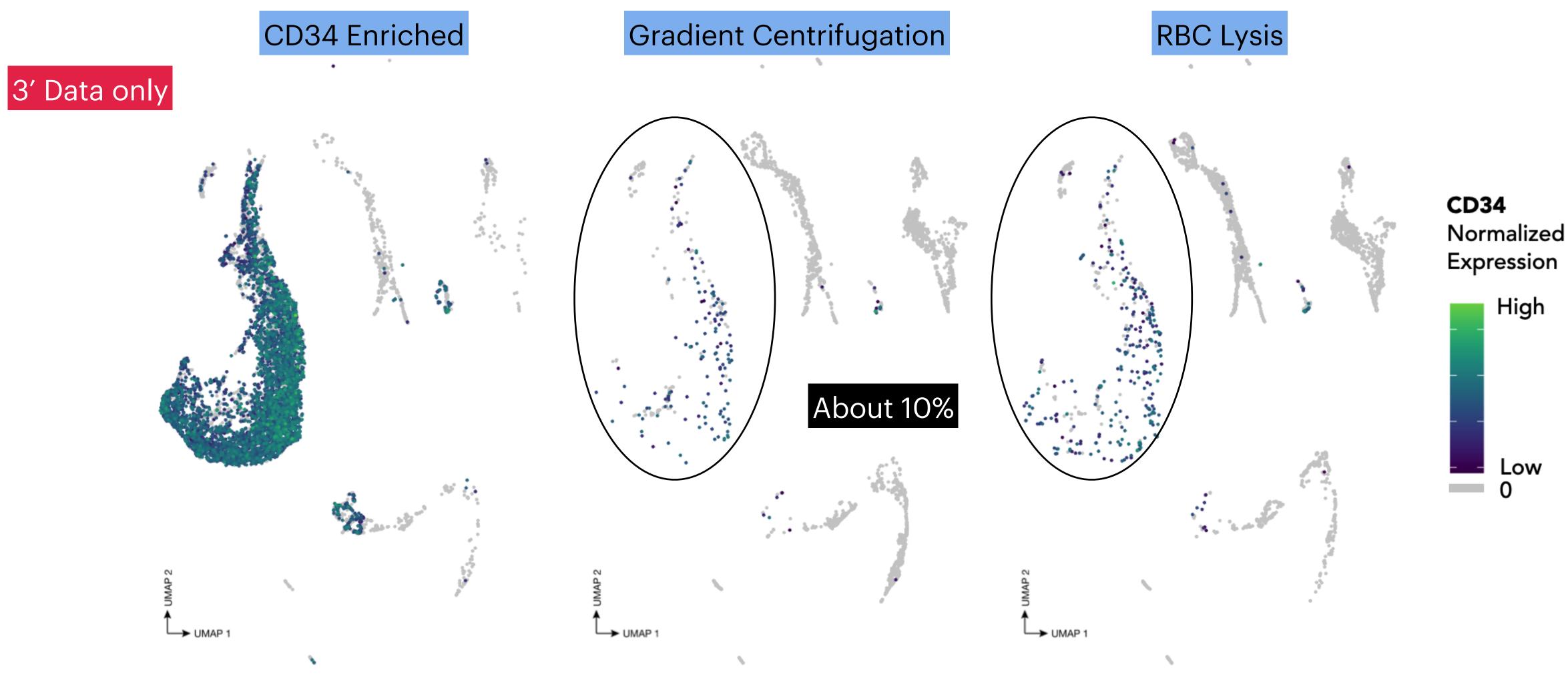
3' Data only

UMAP 1



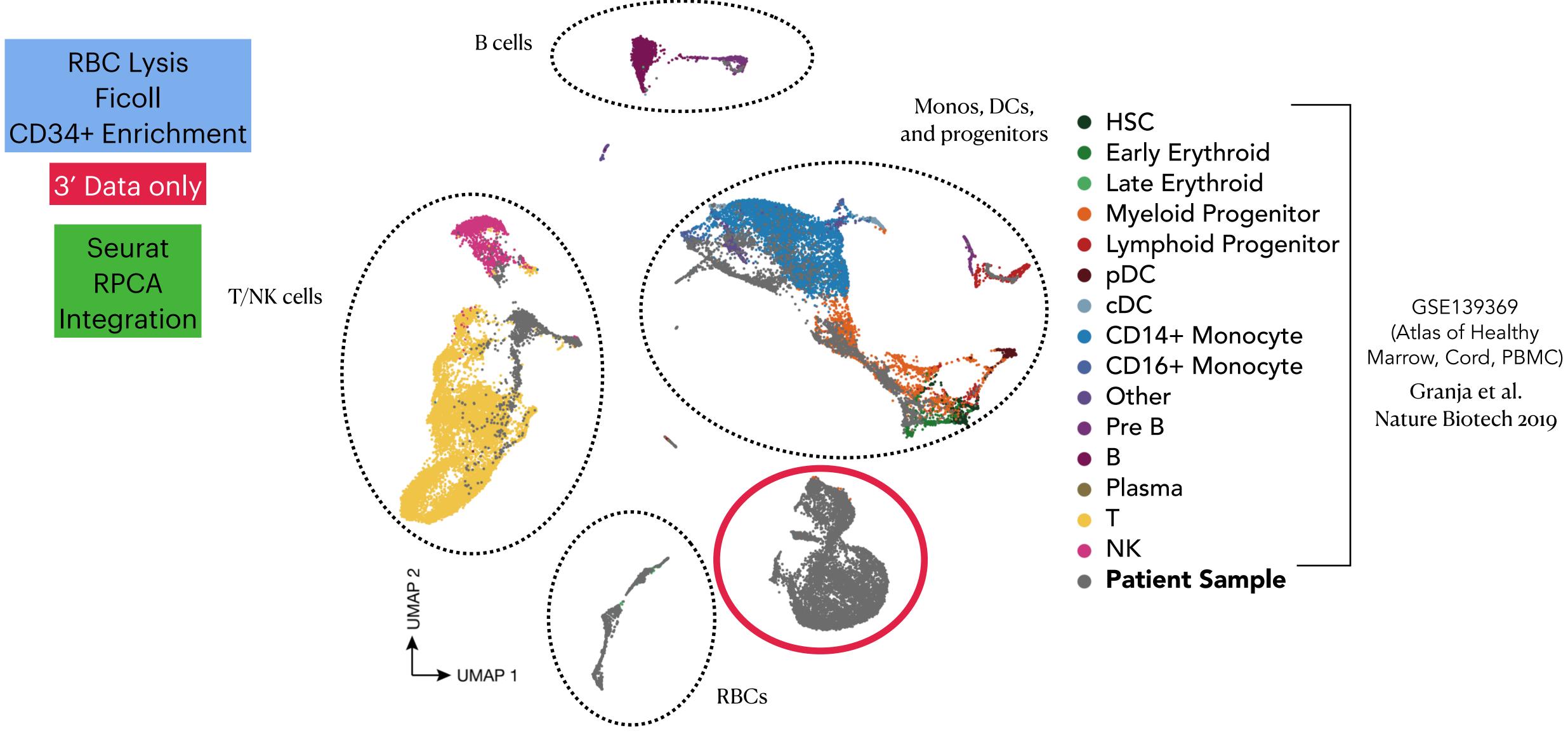


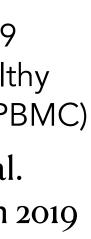
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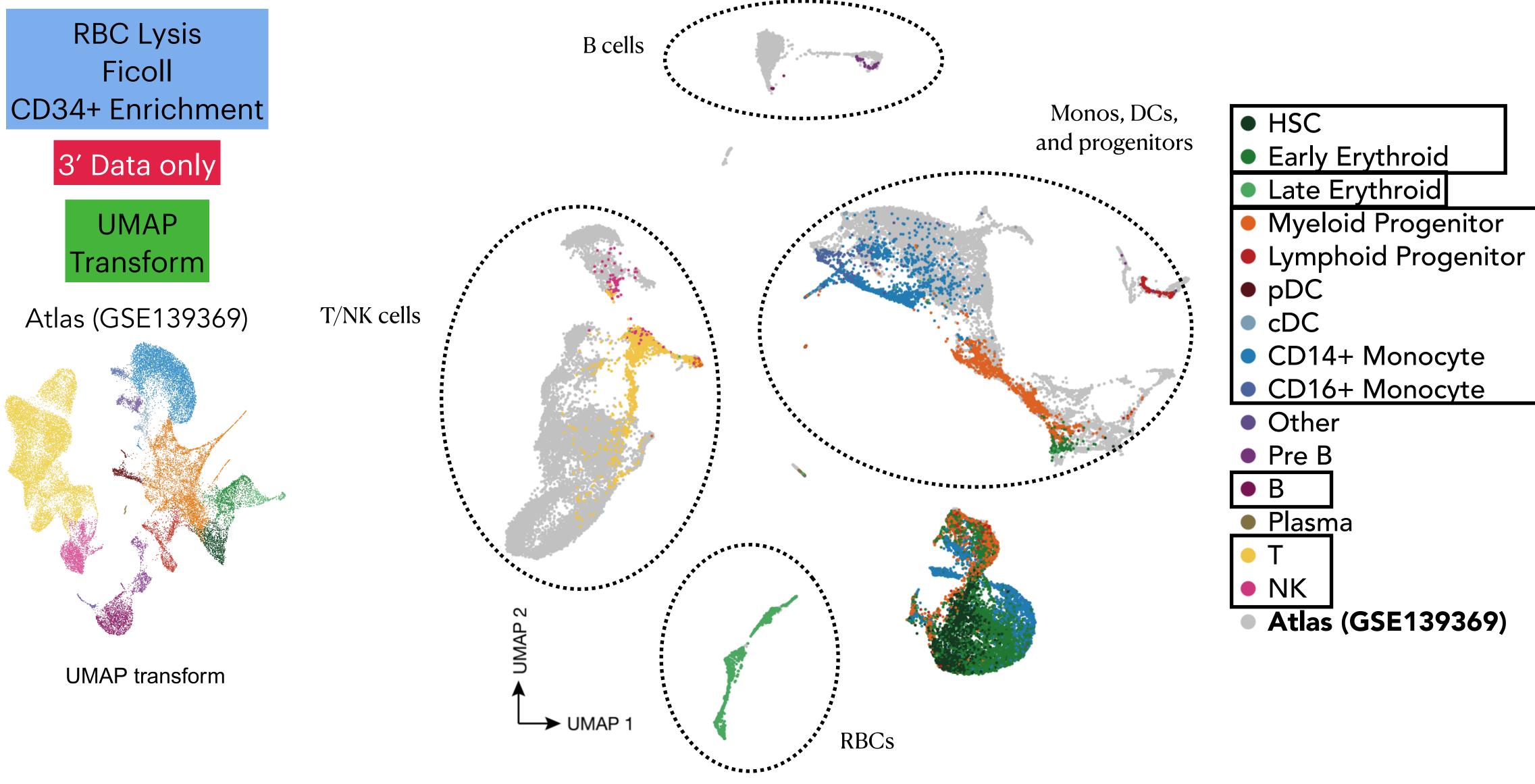


Co-embedding patient cells with healthy atlas



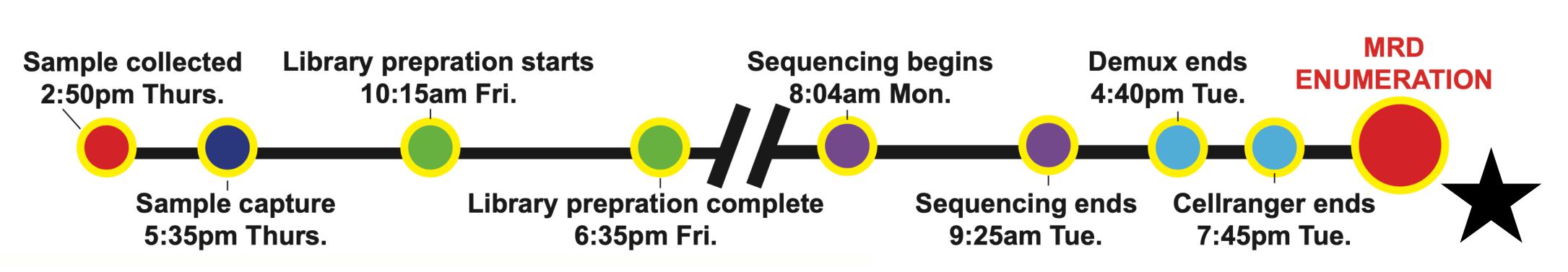


Annotation of cell types





Celltype classification needs to be quick (and accurate)



Motivation for a new cell classifier

- Wish list:
 - R interactive session
 - One line of code
 - Fast
 - Modular
 - UMI count-based (not have to embed)







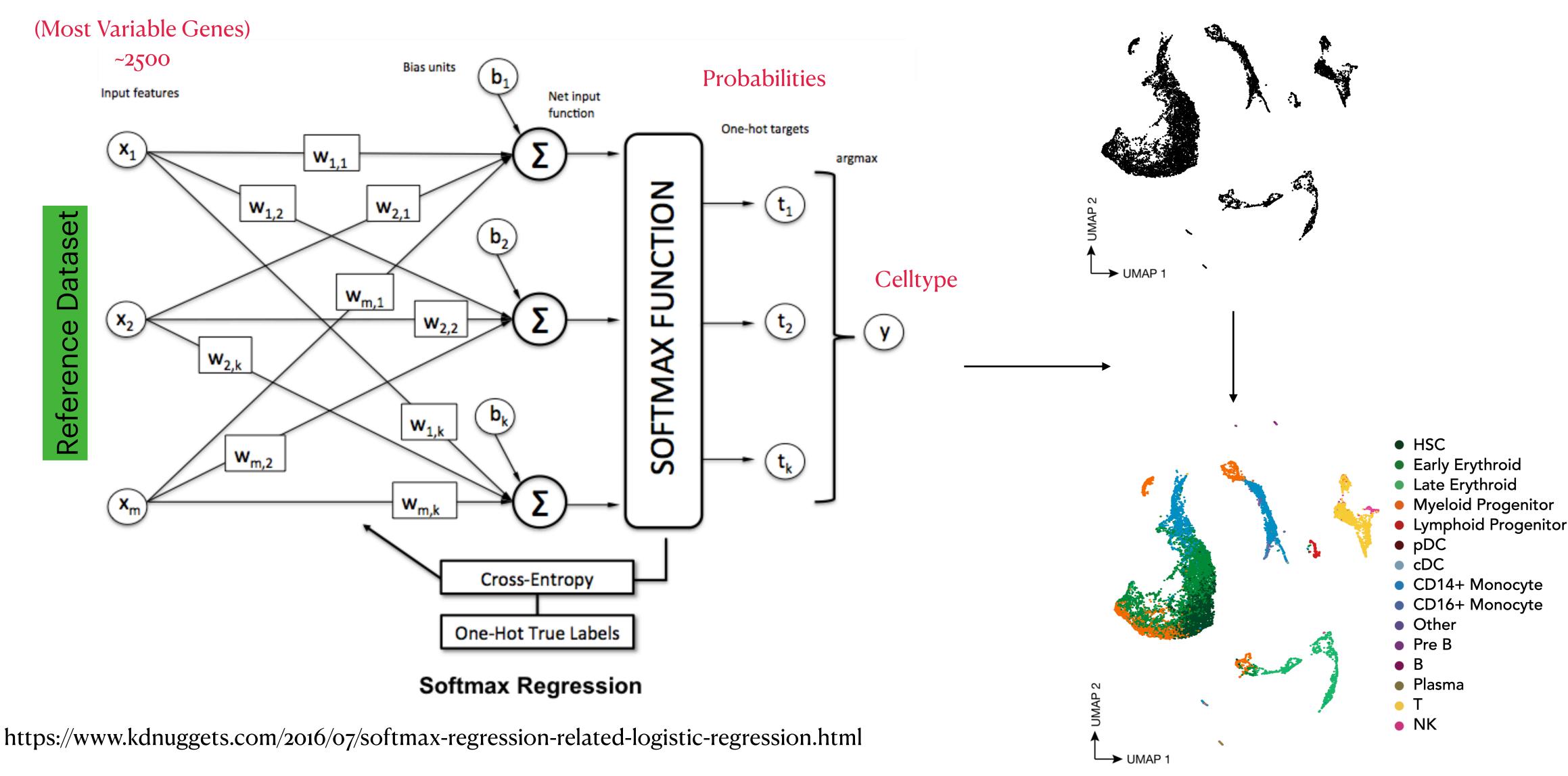
ARRAYFIRE

RcppArrayFire

azuki Fukui and Ralf Stubne

Viewmaster - Softmax Regression

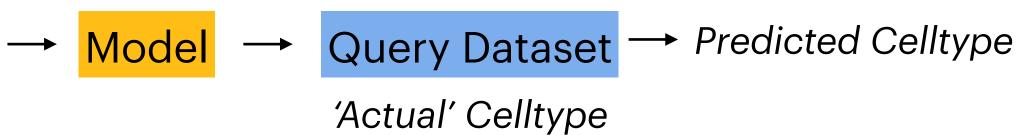
(Most Variable Genes)





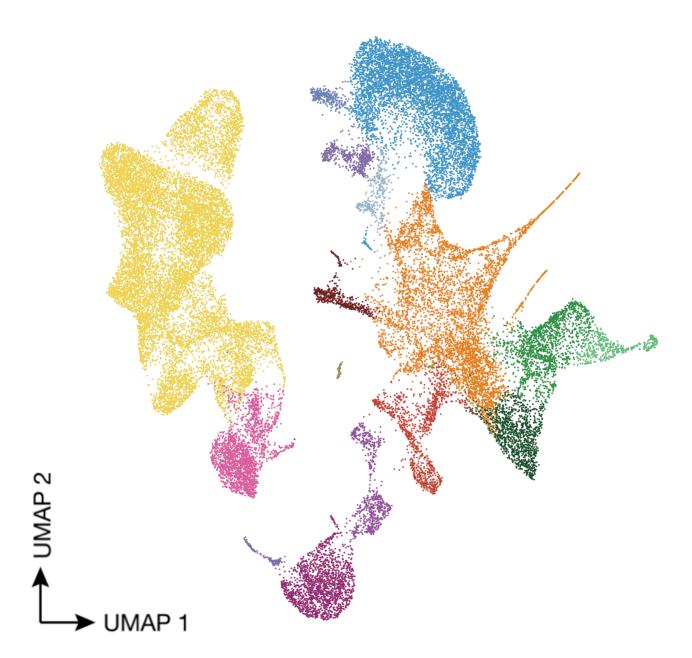






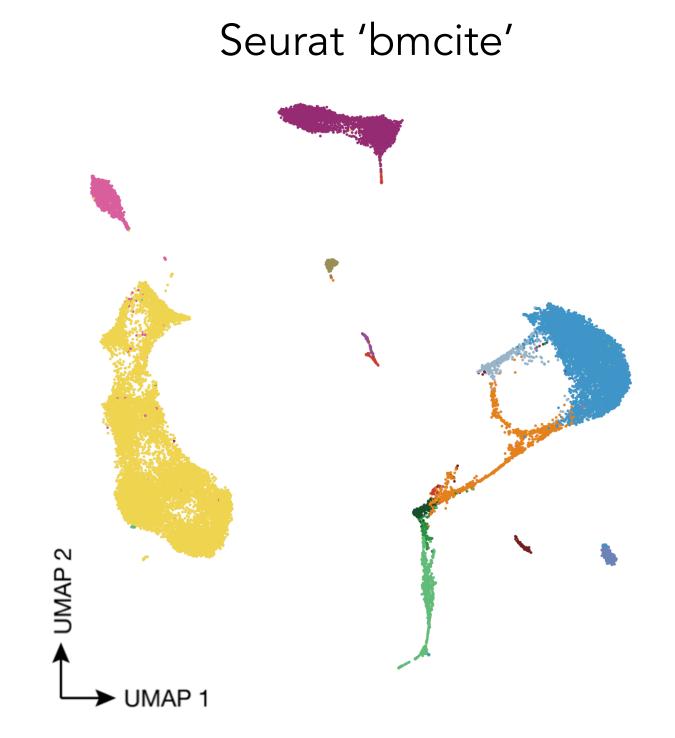
Benchmarking Viewmaster

GSE139369



Granja et al. Nature Biotech 2019

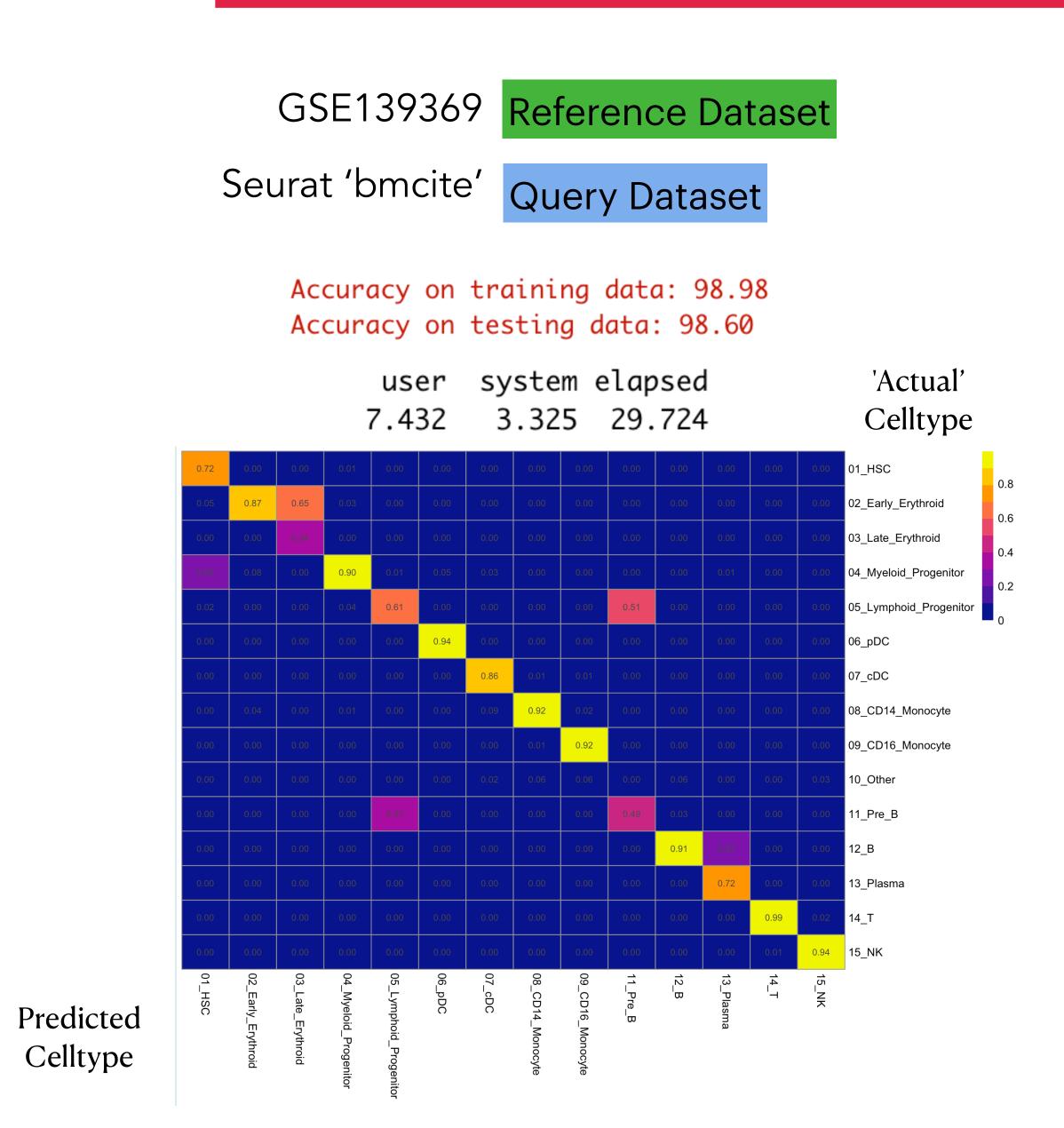




Harmonized Labels

- HSC
- Early Erythroid
- Late Erythroid
- Myeloid Progenitor
- Lymphoid Progenitor
- pDC
- cDC
- CD14+ Monocyte
- CD16+ Monocyte
- Other
- Pre B
- B
- Plasma
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- NK





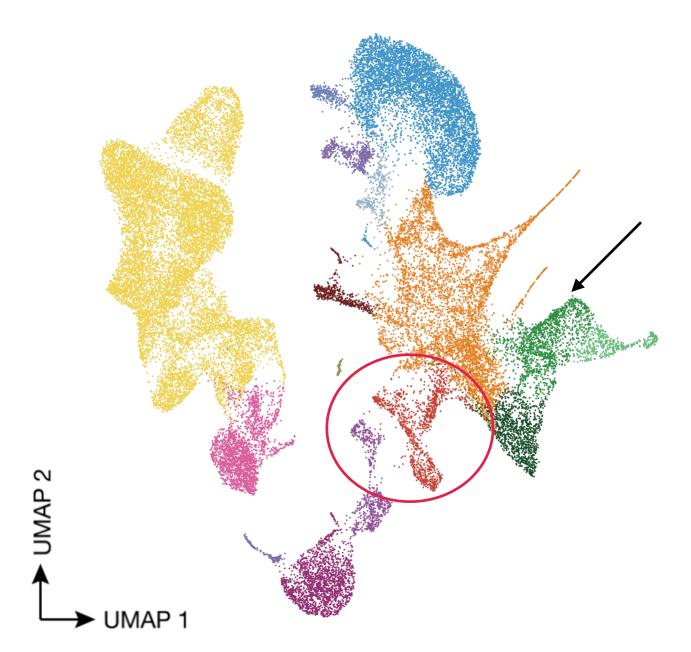
Viewmaster





Rare cell types may confound softmax regression

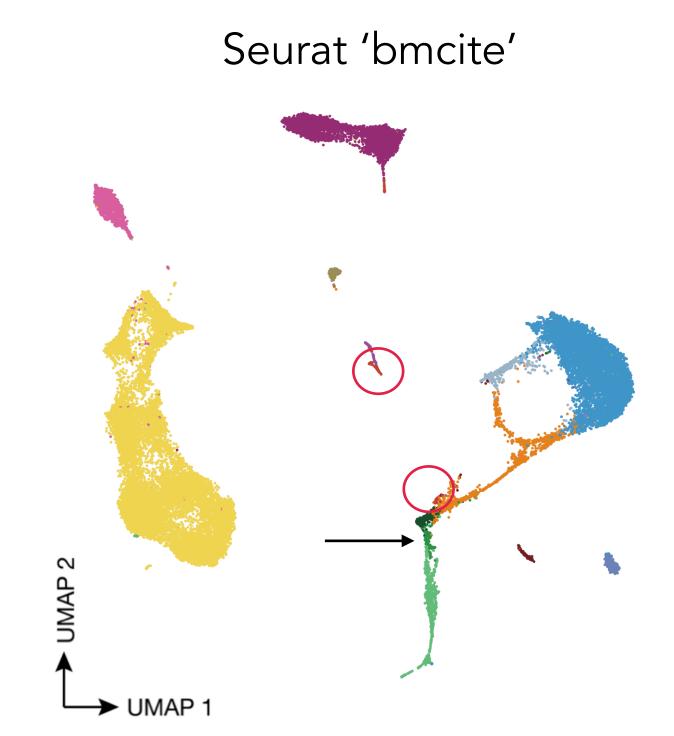




Granja et al. Nature Biotech 2019



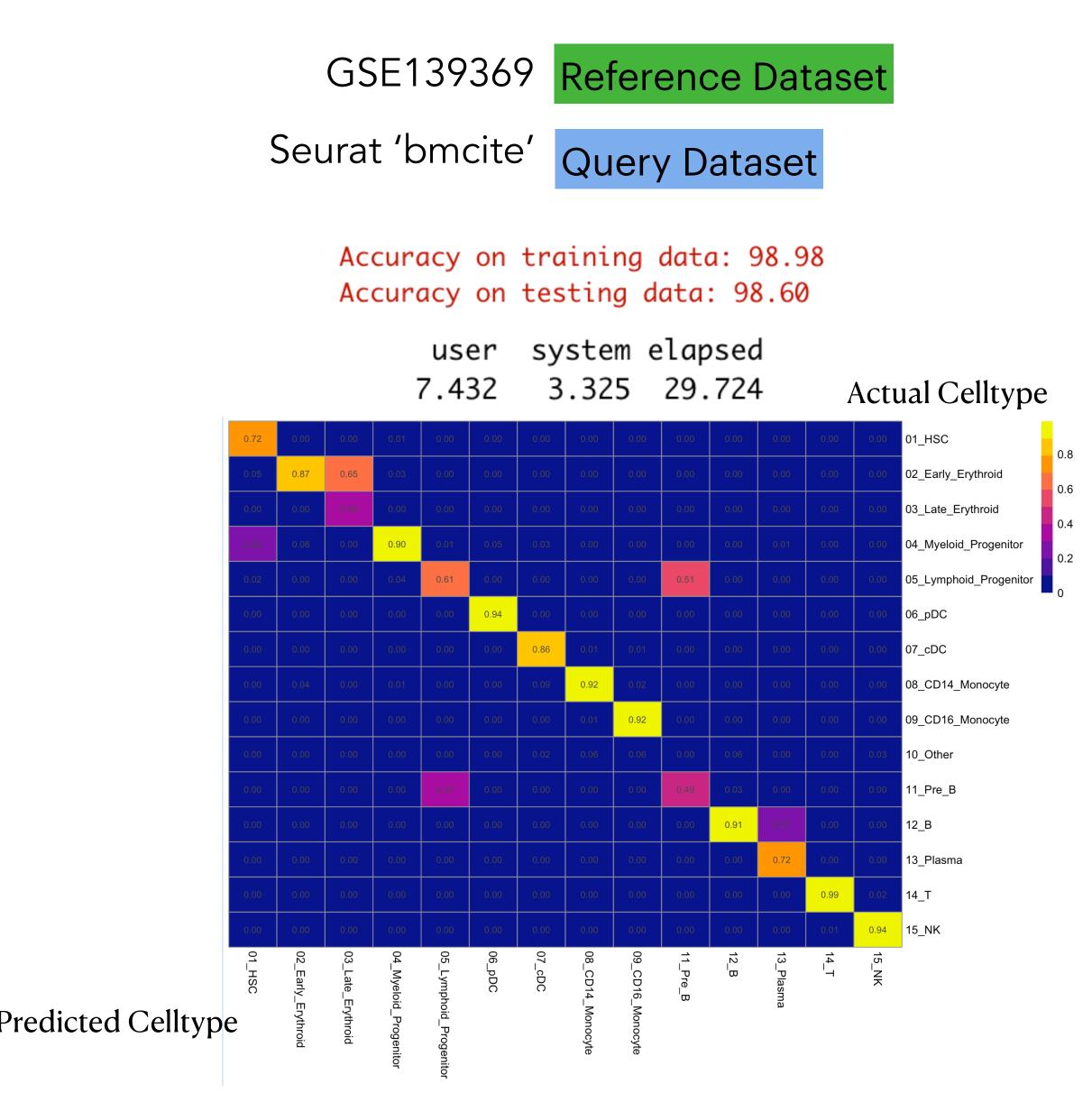




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Viewmaster



GSE139369 Query Dataset



Accuracy on training data: 98.99 Accuracy on testing data: 97.49

> system elapsed user 7.795 35.641 8.405

0.94	0.03	0.00	0.03	0.02	0.00	0.00	0.00	0.00	0.01	0.00	0.00	0.00	0.00	0.00	01_HSC	1
0.01	0.34	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	02_Early_Erythroid	0.8
0.00	0.60	0.98	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	03_Late_Erythroid	0.6
0.01	0.03	0.00	0.91	0.09	0.01	0.03	0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	04_Myeloid_Progenitor	0.4
0.03	0.00	0.00	0.00	0.57	0.00	0.00	0.00	0.00	0.00	0.30	0.00	0.00	0.00	0.00	05_Lymphoid_Progenitor	0.2
0.00	0.00	0.00	0.01	0.00	0.98	0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	06_pDC	0
0.00	0.00	0.00	0.01	0.00	0.00	0.90	0.02	0.00	0.01	0.00	0.00	0.00	0.00	0.00	07_cDC	
0.00	0.00	0.01	0.01	0.00	0.00	0.05	0.97	0.06	0.30	0.00	0.00	0.00	0.00	0.00	08_CD14_Monocyte	
		0.00	0.00													
0.00	0.00			0.00	0.00	0.00	0.00	0.94	0.00	0.00	0.00	0.00	0.00	0.00	09_CD16_Monocyte	
0.00	0.00	0.00	0.00	0.32	0.00	0.00	0.00	0.00	0.00	0.61	0.00	0.00	0.00	0.00	11_Pre_B	
0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.37	0.09	0.99	0.00	0.00	0.00	12_B	
0.00	0.00	0.00	0.00	0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	1.00	0.00	0.00	13_Plasma	
0.00	0.01	0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.16	0.00	0.00	0.00	0.99	0.15	14_T	
0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.14	0.00	0.00	0.00	0.00	0.85	15_NK	
01_HSC)2_Earl)3_Late)4_Mye)5_Lym	06_pDC	07_cDC)8_CD1)9_CD1	10_Other	11_Pre_B	12_B	13_Plasma	14_T	15_NK		
.,	02_Early_Erythroid	03_Late_Erythroid	04_Myeloid_Progenitor	05_Lymphoid_Progenitor			08_CD14_Monocyte	09_CD16_Monocyte	Ъ	'ϖ		ma				
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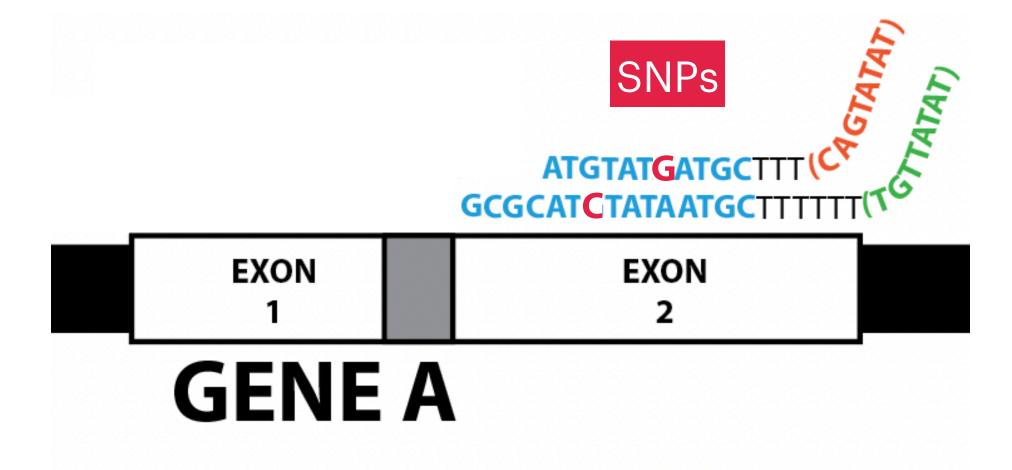
Olivia Waltner BA







Can we leverage natural genetic variation to improve leukemia detection?

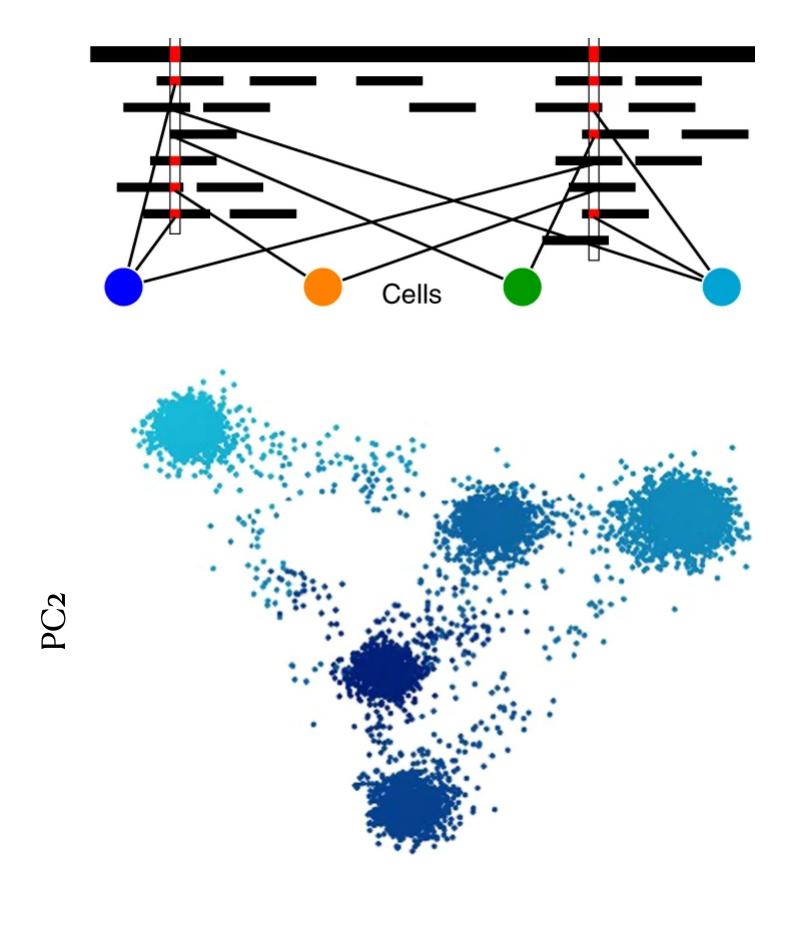


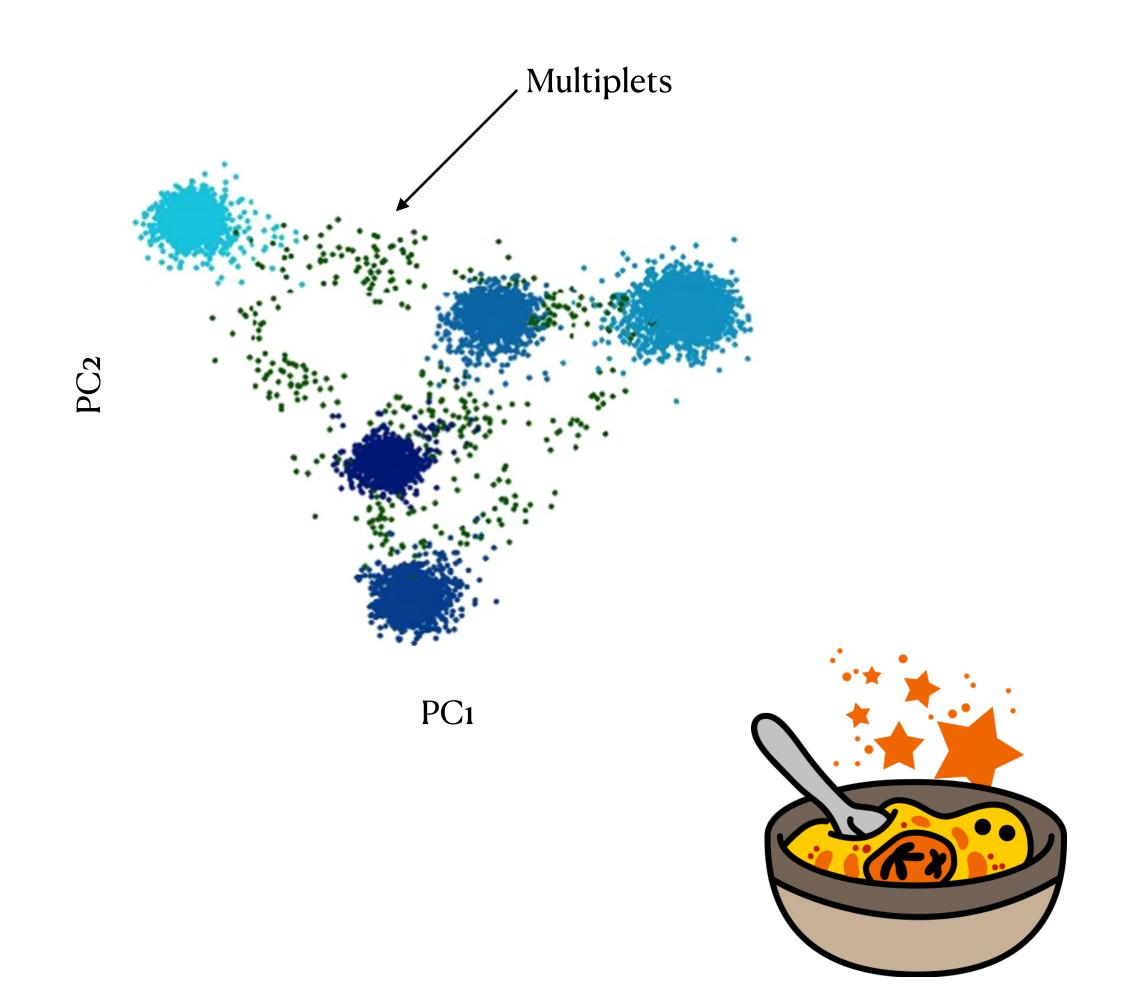
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Imperative to pair this with cell type

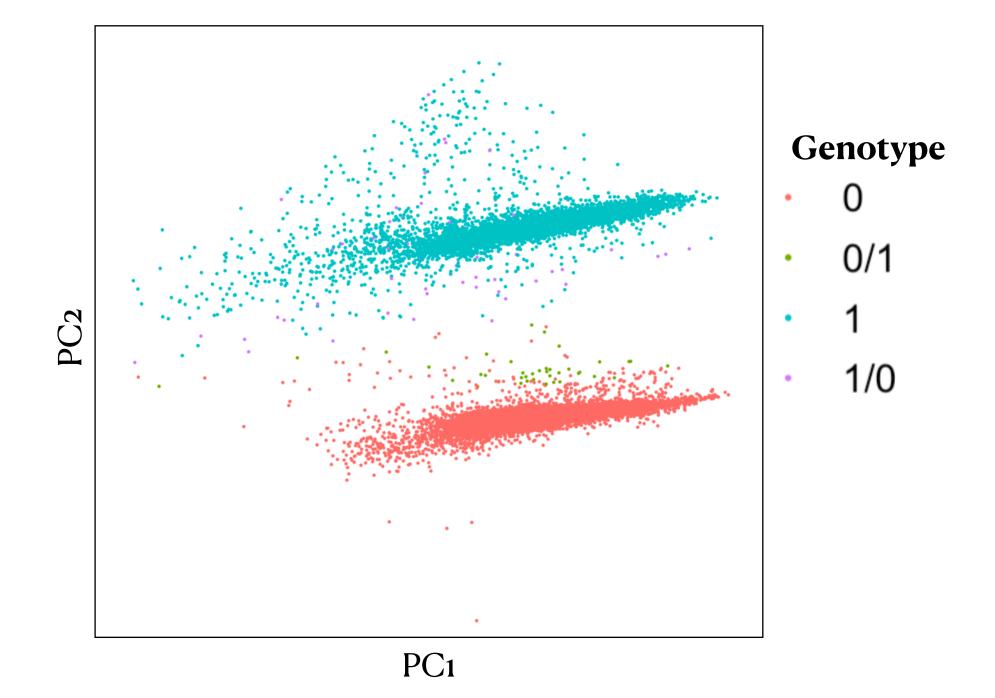
Souporcell genetically demultiplexes scRNAseq data



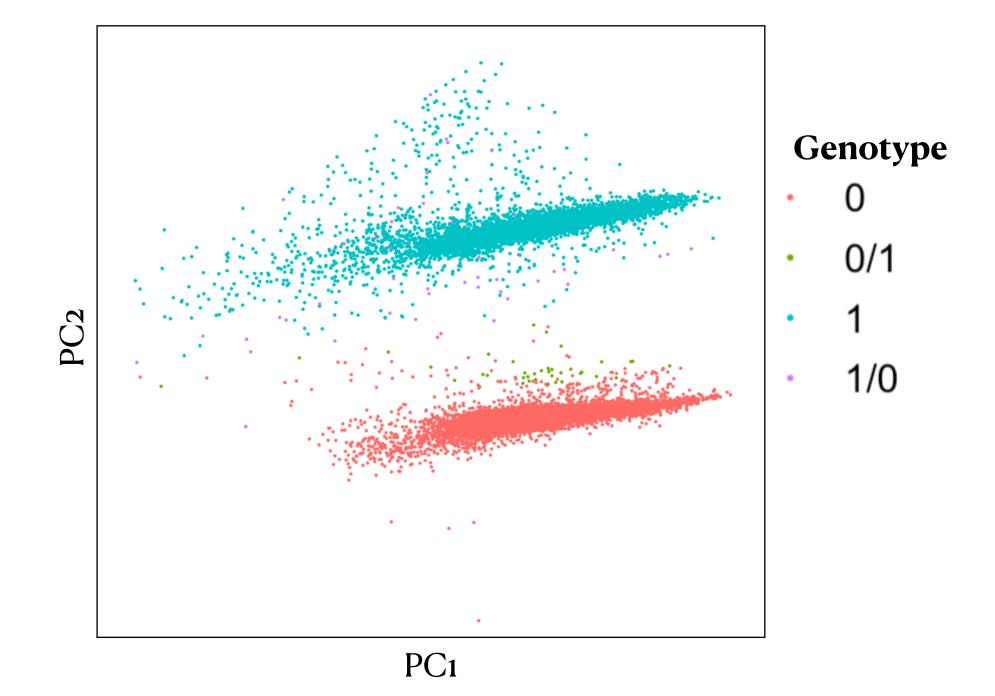


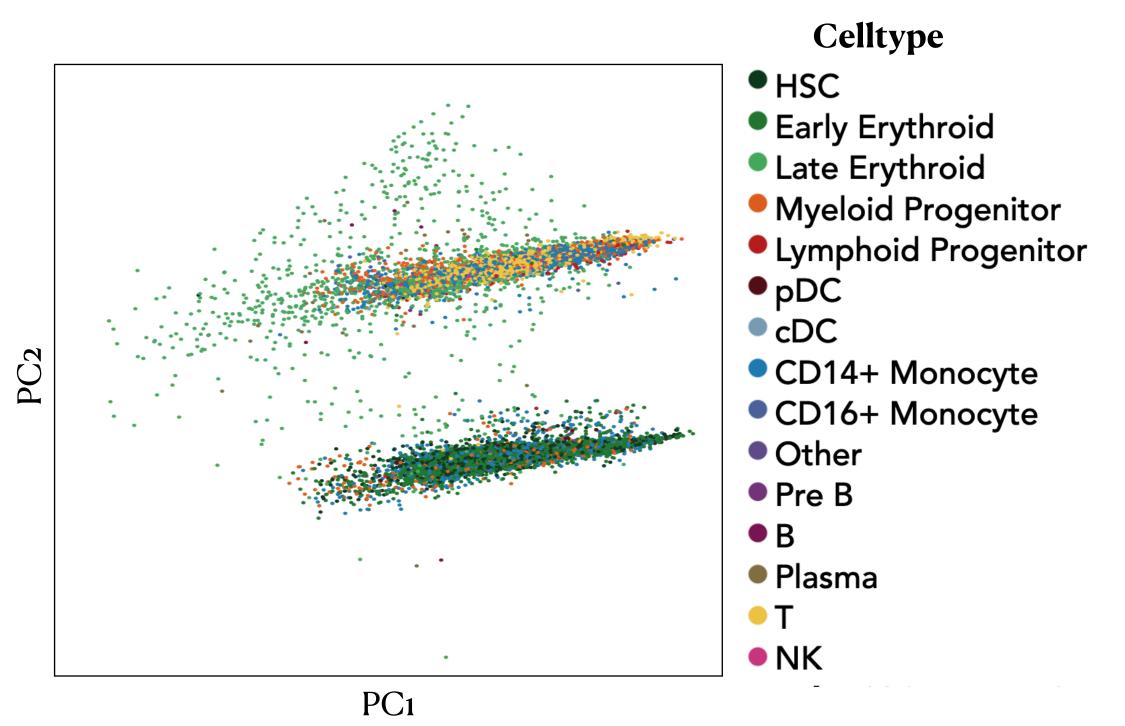
Heaton H, Talman AM, Knights A, et al. Nat Methods. 2020;17(6):615-620.

Souporcell identifies two genotypes in our sample

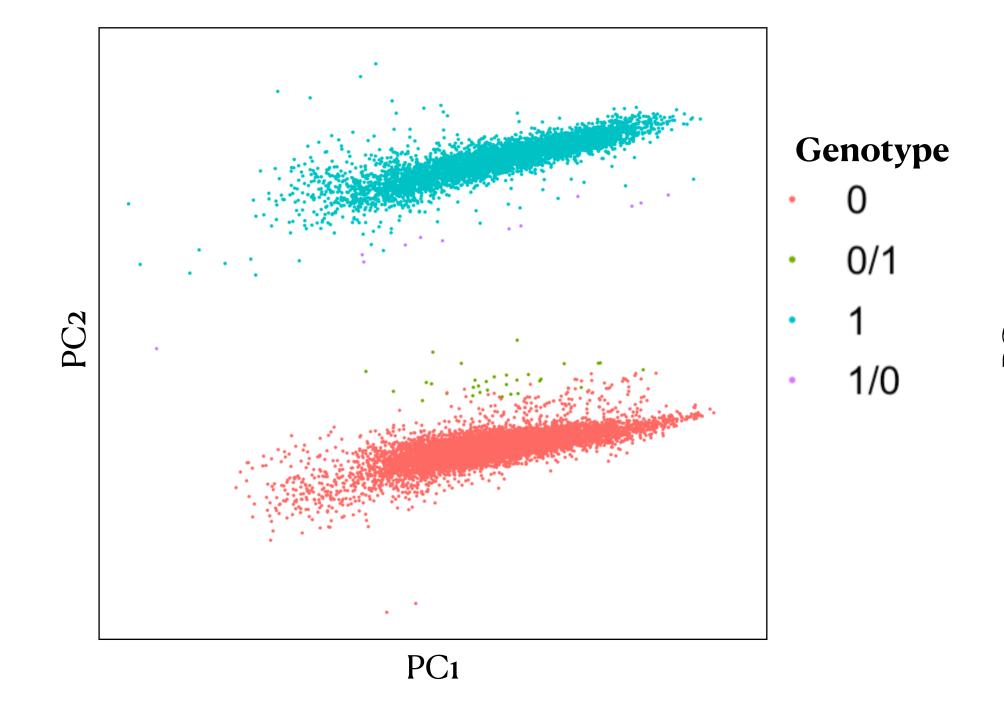


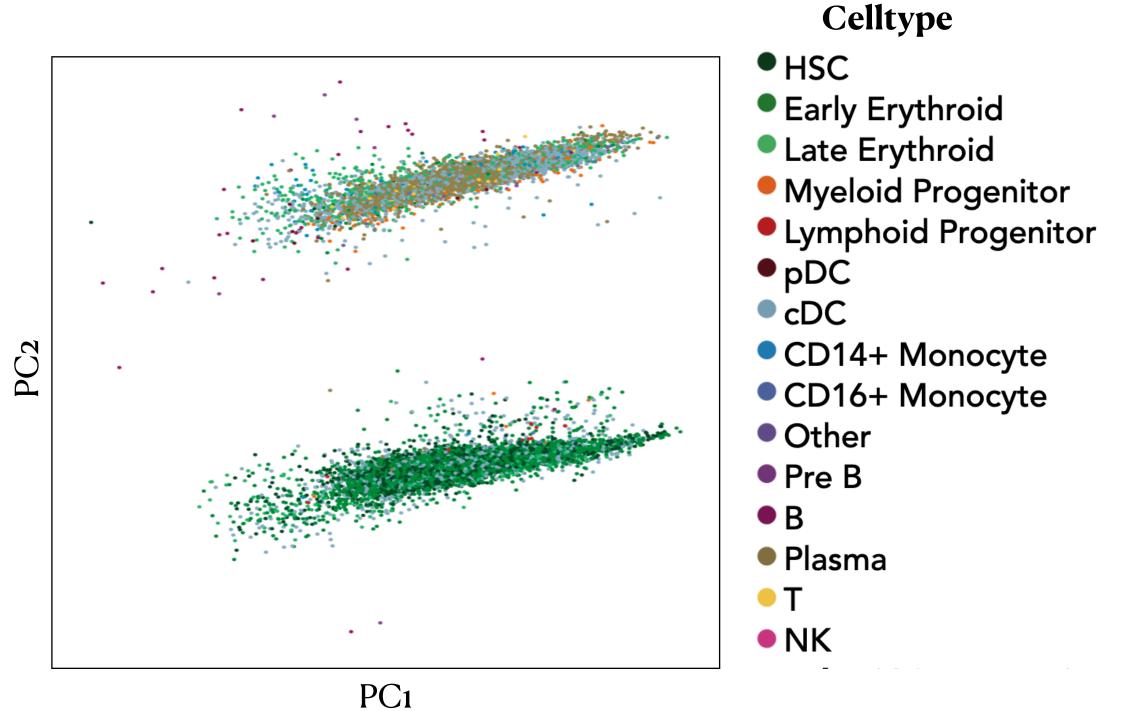
Souporcell struggles with RBCs



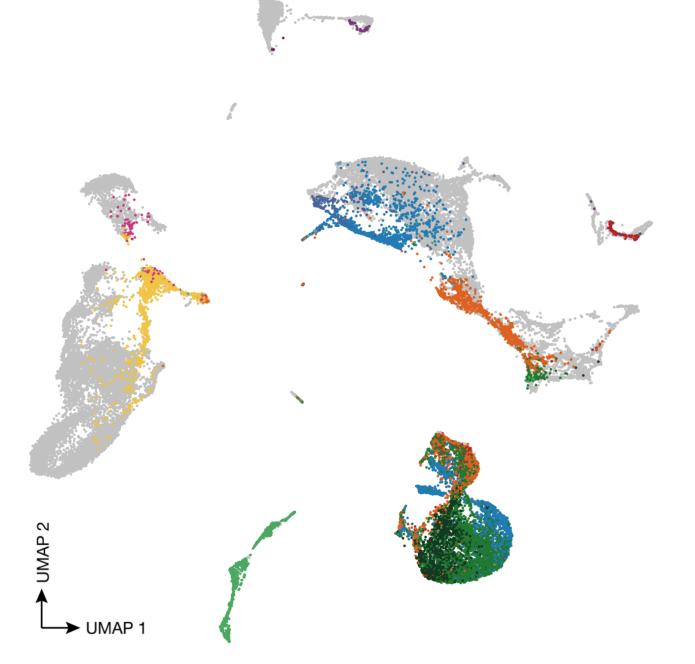


Souporcell struggles with RBCs





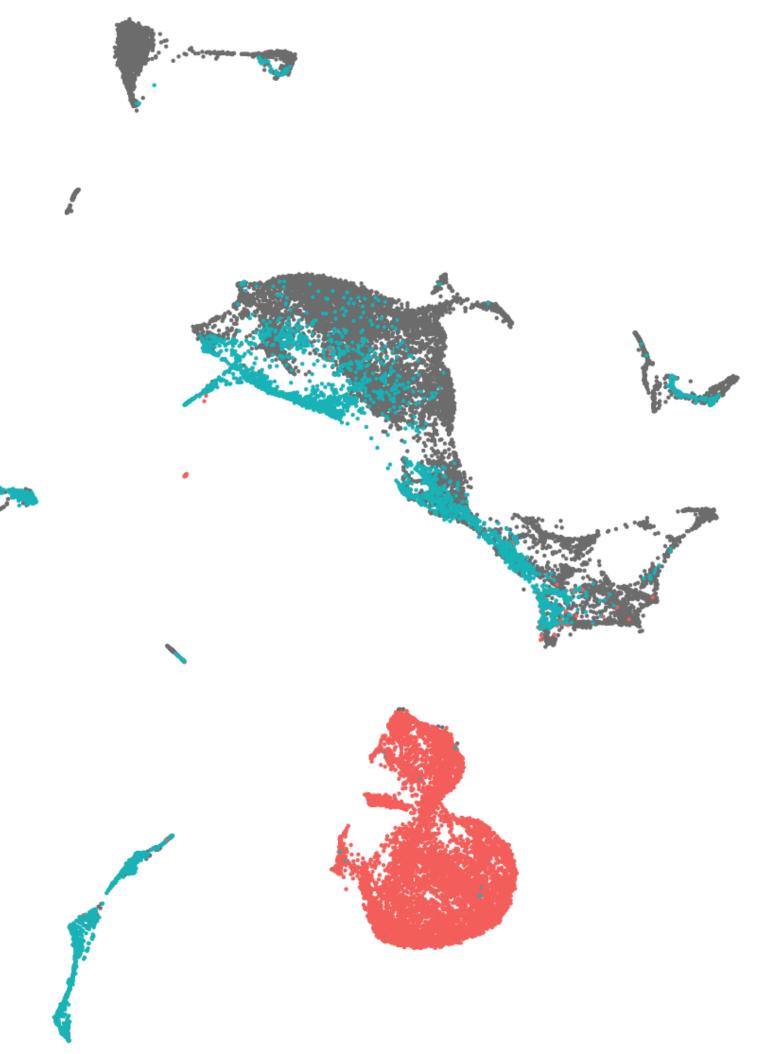
Convincing evidence for relapse using souporcell



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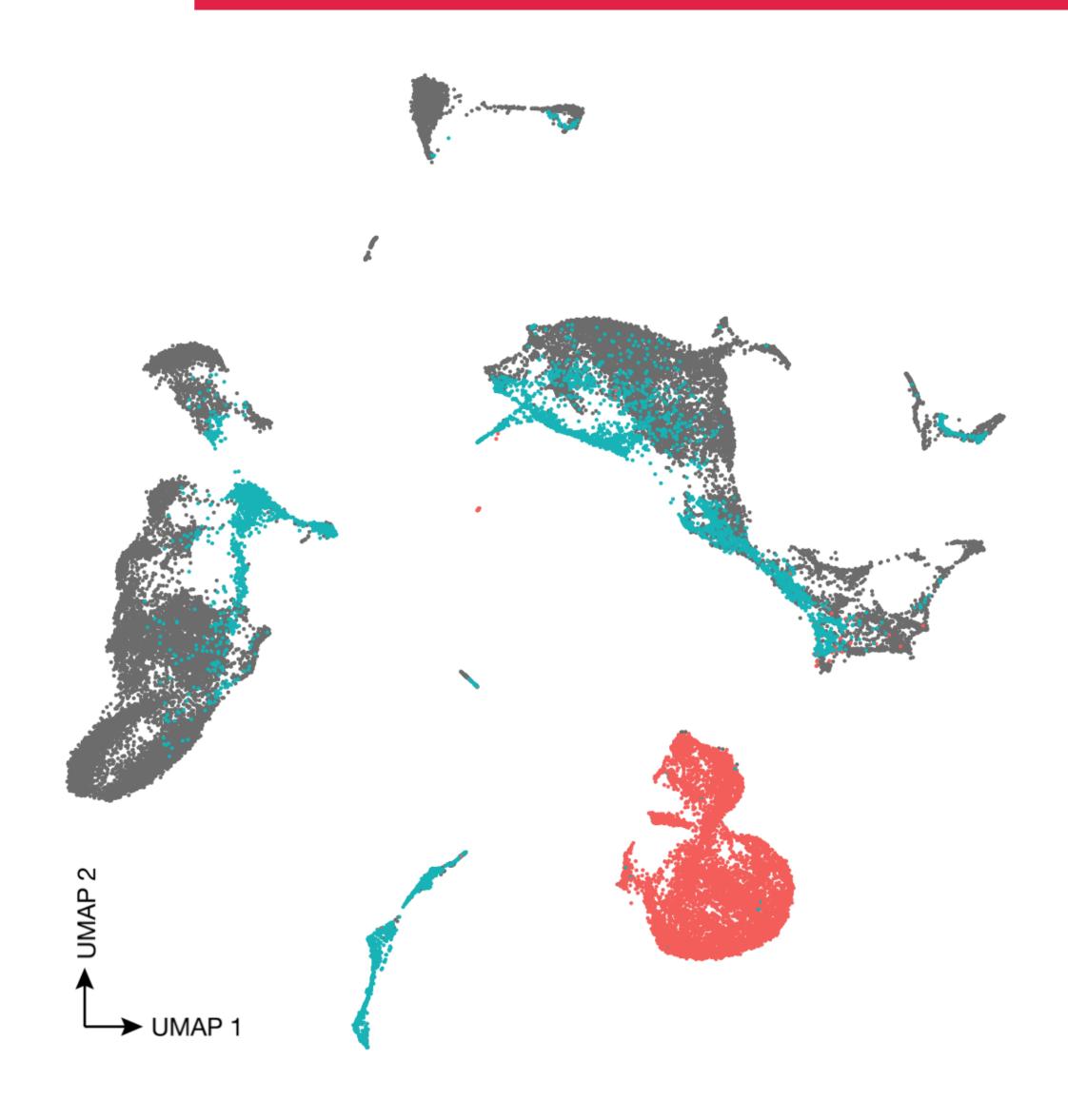
 \sim UMAP 1



Genotype 1 Genotype 2 • Atlas (GSE139369)



Convincing evidence for relapse using souporcell

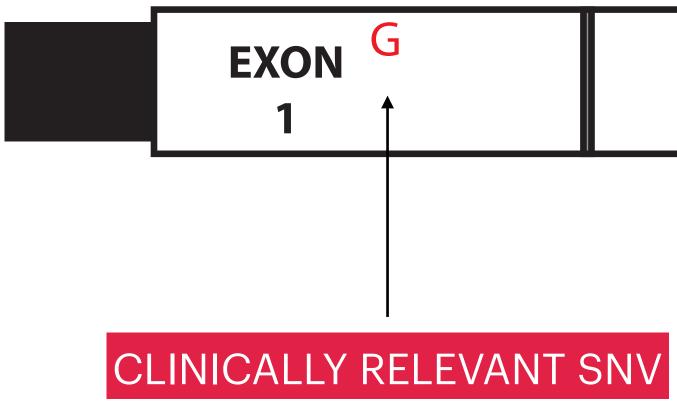


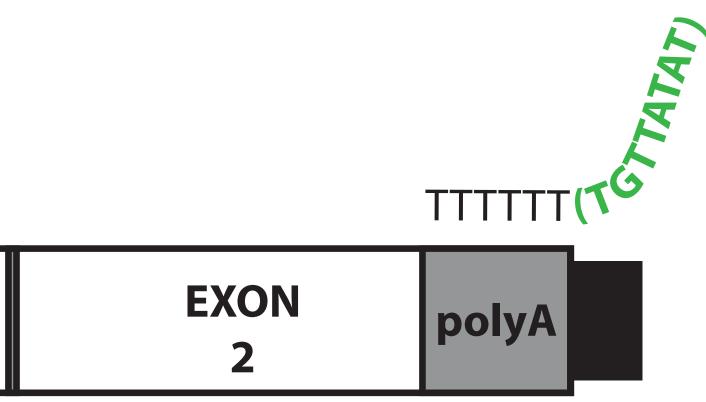
- Genotype 1
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- Atlas (GSE139369)

Genotype 1Genotype 2

Can we detect AML mutations in scRNAseq data?

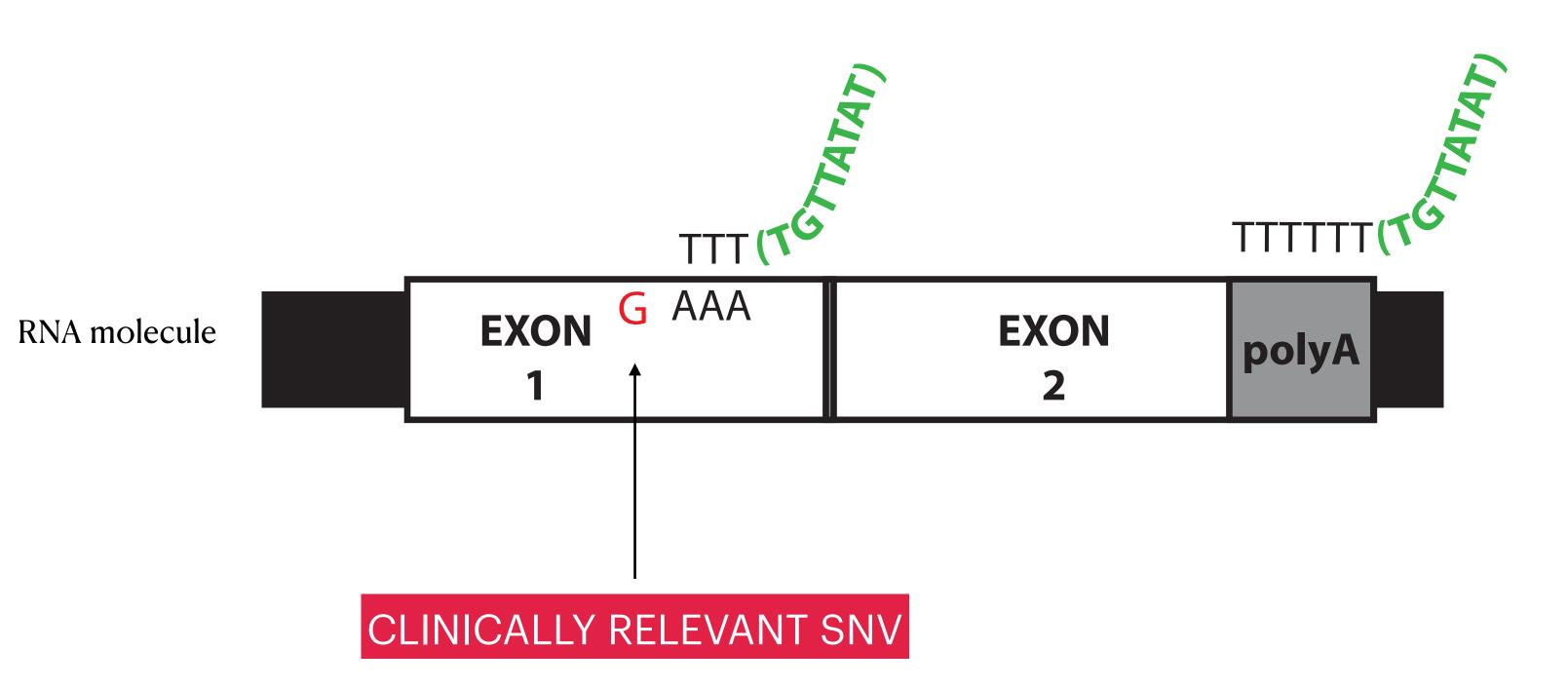








Can we detect AML mutations in scRNAseq data?



cb_sniffer

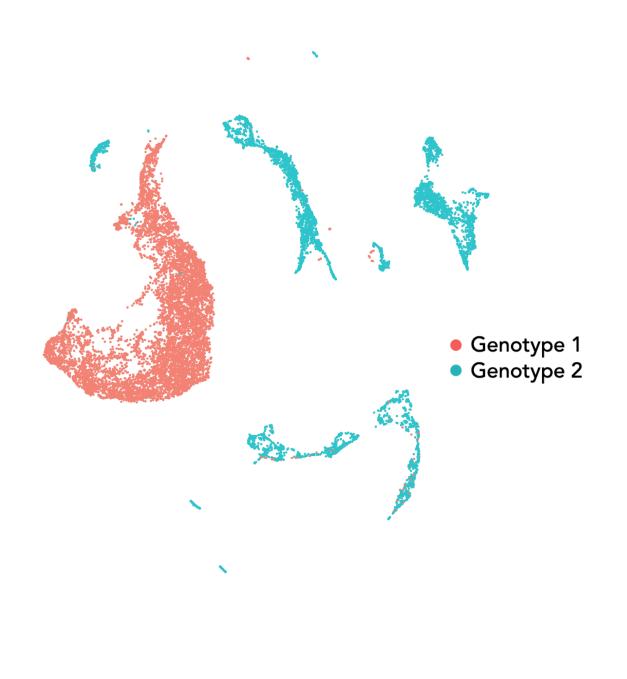
Petti et. al. Nature Communications 2019



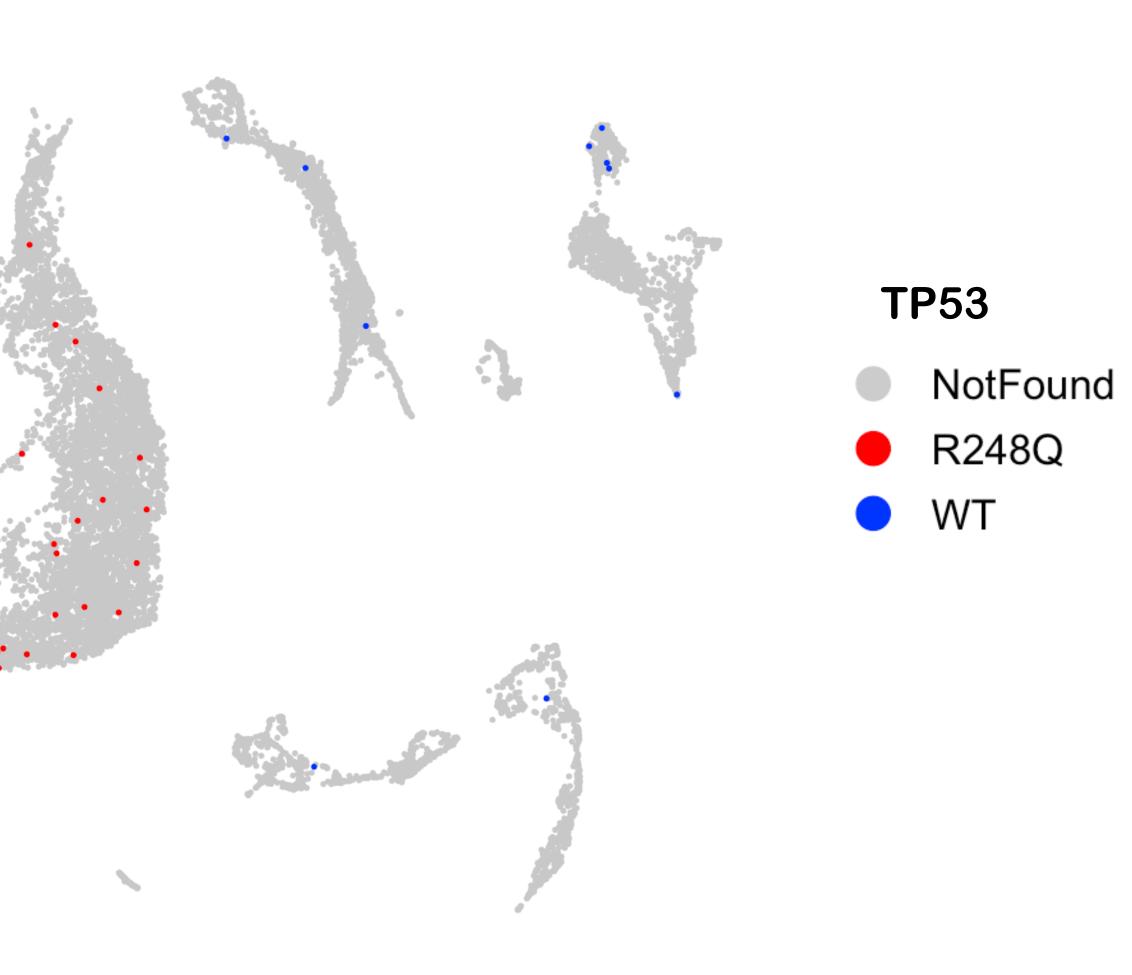
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 \sim UMAP → UMAP 1

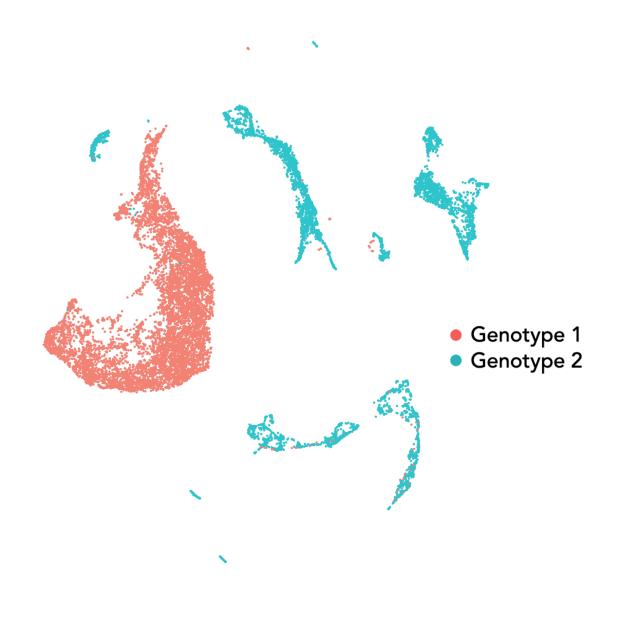




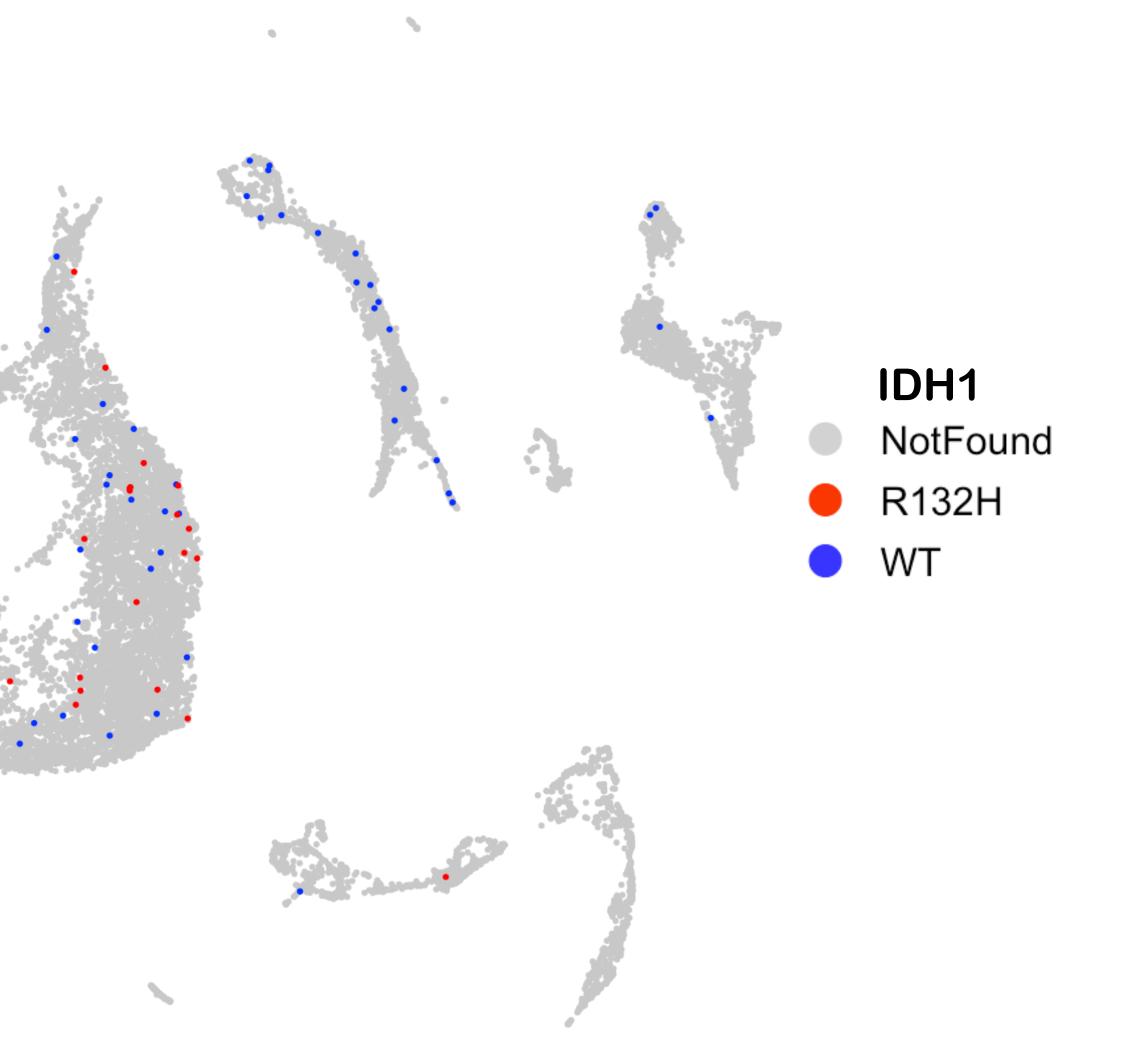
Not sufficient coverage for the study of clonal heterogeneity

cb_sniffer

Petti et. al. Nature Communications 2019



 \sim UMAP → UMAP 1





CD34+ Enrichment Unfragmented

3' Data only

Hybridization capture with 1253 cancer gene panel PCR, concatenation, SMRT sequencing

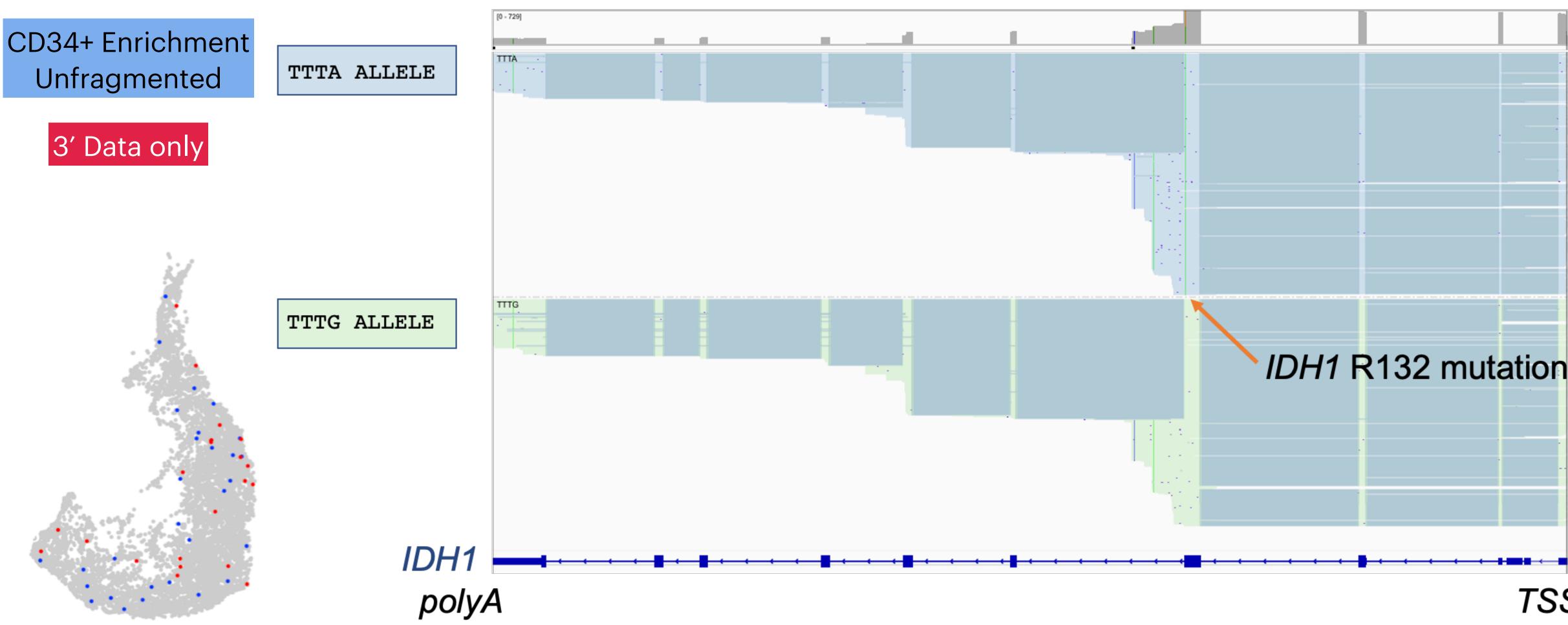






Jason Underwood PhD

We can dramatically increase coverage using IsoSeq







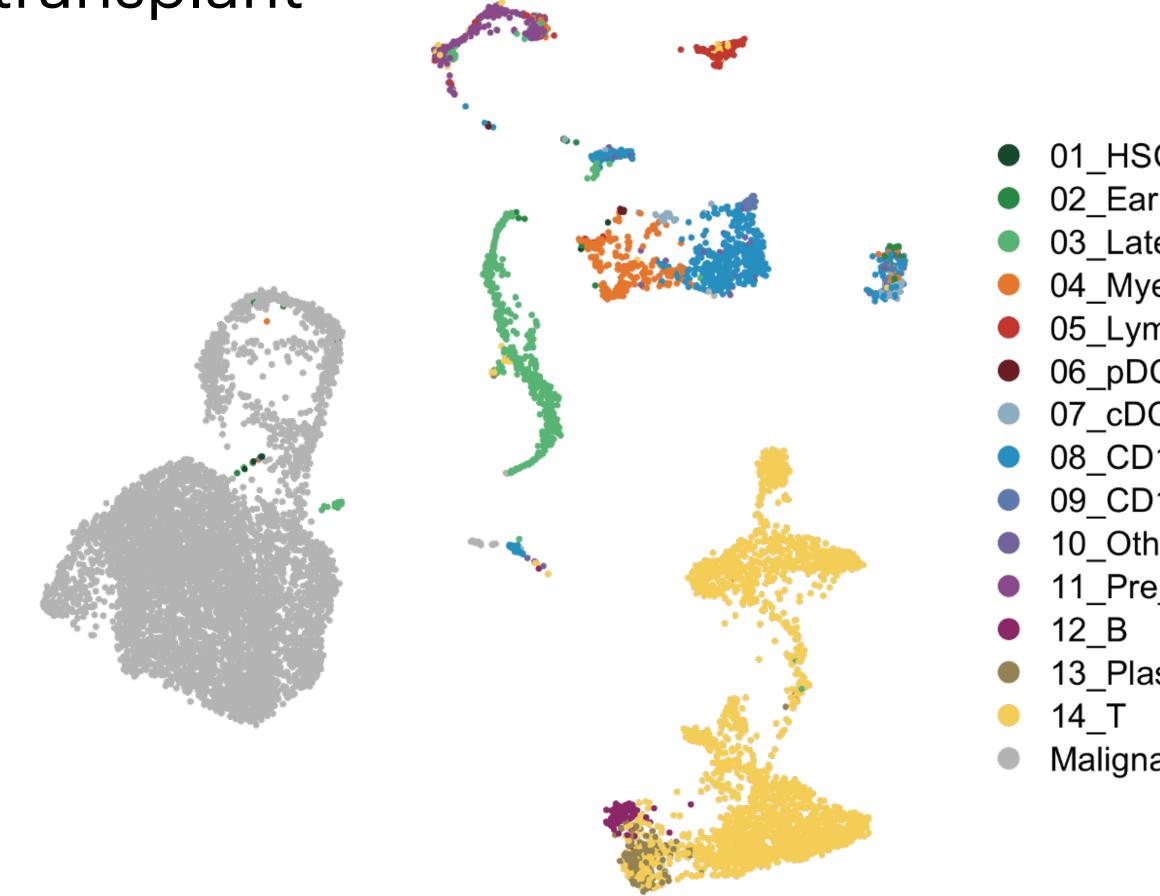
How does this extend to other patients?

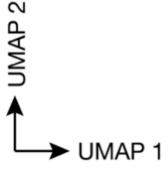
 2 yo with undifferentiated leukemia, with suspected relapse after unrelated cord blood transplant





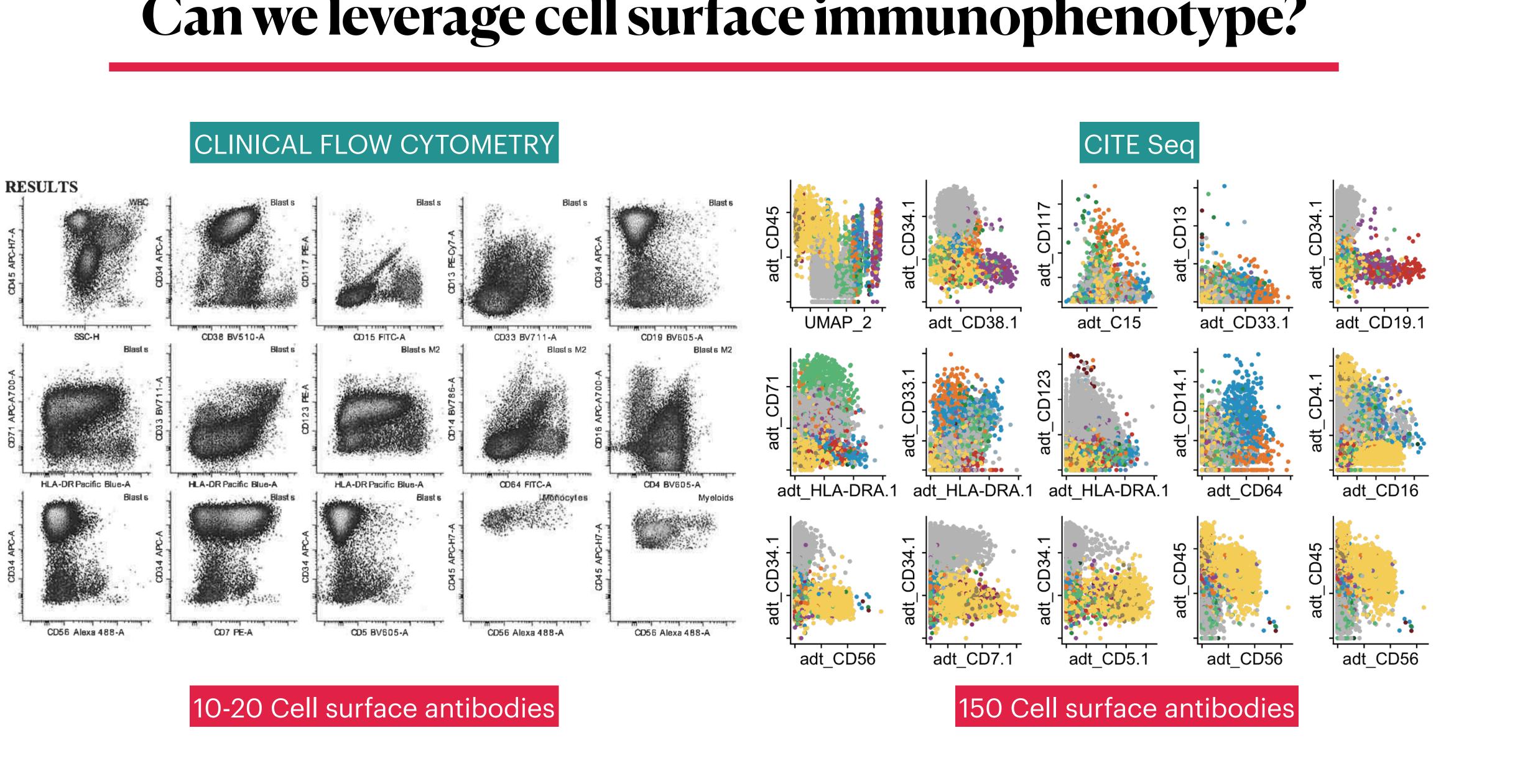




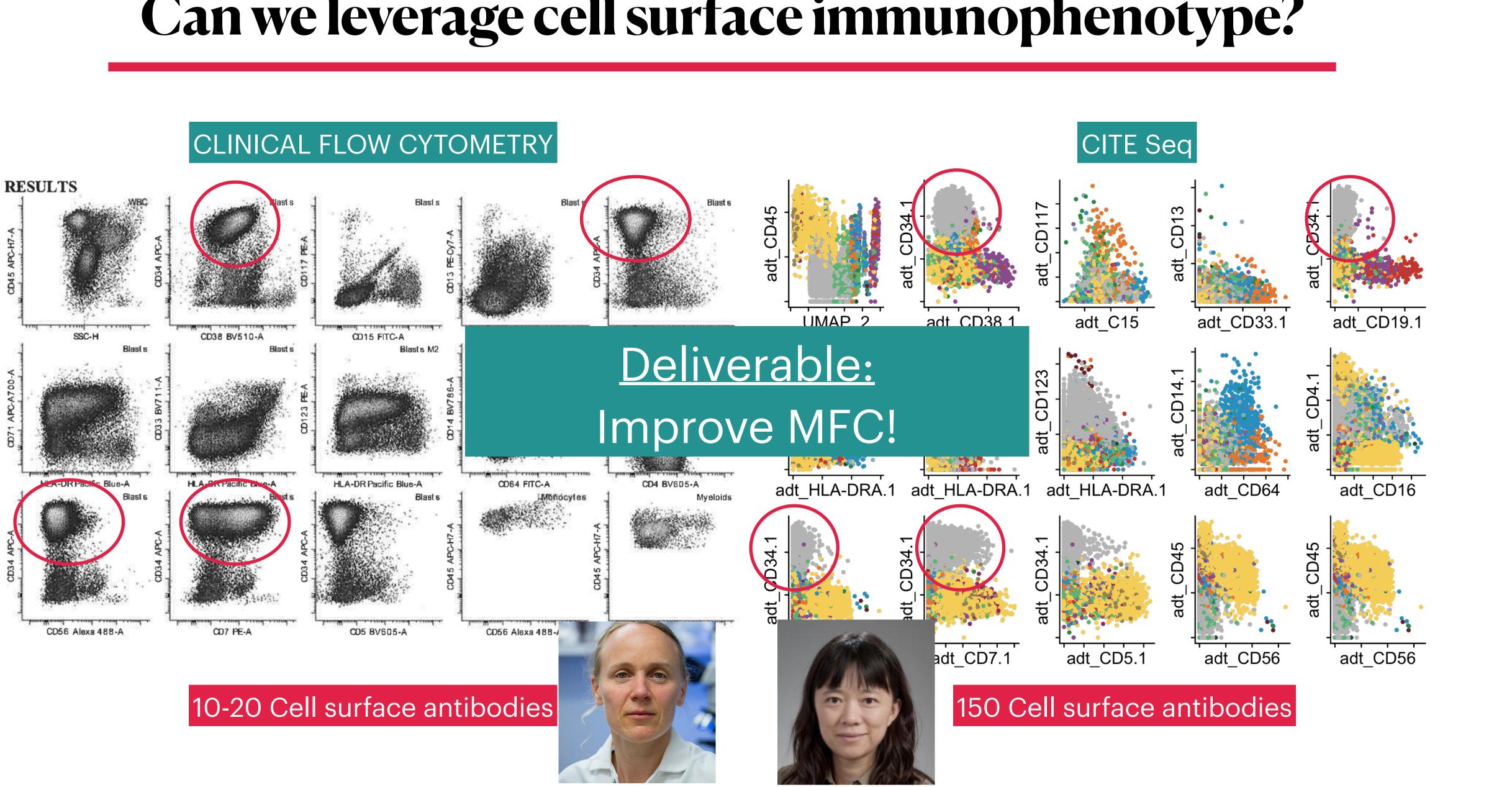


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- 05_Lymphoid_Progenitor
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- 07_cDC
- 08_CD14_Monocyte
- 09_CD16_Monocyte
- 10_Other
- 11_Pre_B
- 13_Plasma
- Malignant

Can we leverage cell surface immunophenotype?



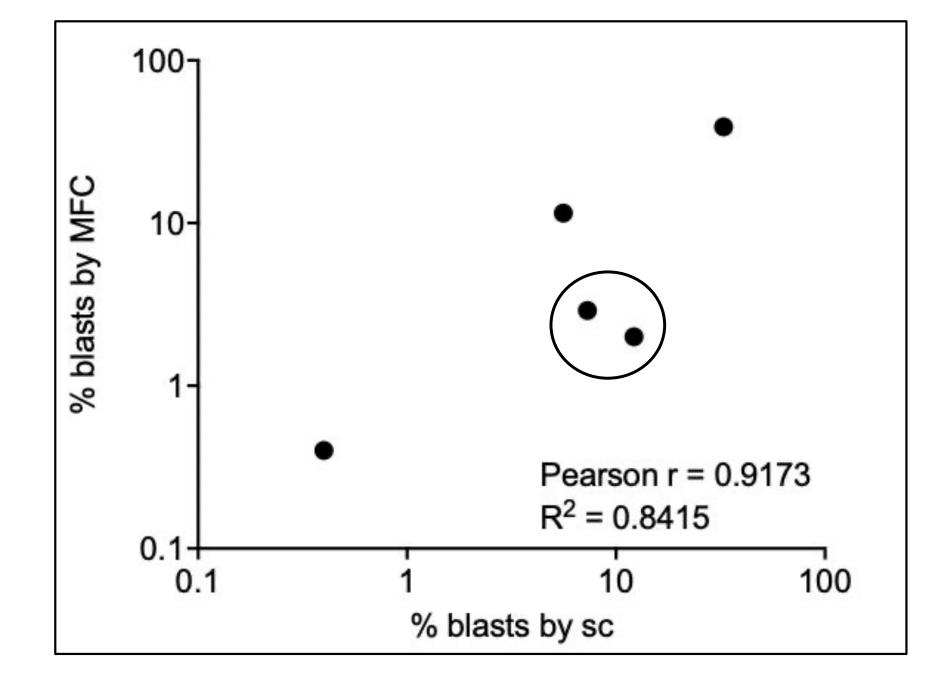
Can we leverage cell surface immunophenotype?



Melinda Biernacki MD

Xueyan Chen MD PhD

Where are we now?



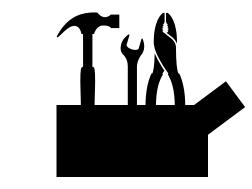
- Two patients with PCR+/MFC negative
- One cell of early myeloid lineage and recipient origin

- Inconsistencies in clinical assays are important motivators to improve diagnostics
- Integration of single expression data with genetic demultiplexing can provide a confident assessment of burden of relapsed leukemia
- Promising preliminary data suggesting that we can augment coverage of specific loci
- Immunoproteomic data show promise in recapitulating clinical flow cytometric data.



Future directions

- Increase sample numbers
- Working to detect fusion transcripts using PacBio sequencing
- Mechanisms of relapse
 - HLA expression / Antigen expression
 - T cell exhaustion
 - Myeloid suppressor cells
- Resources...



- Computational Biologists/Data Scientists
- Positions are available for Lead Data Scientist and Data Scientist I positions
- and algorithms is required
- cancer!

https://careers-seattlechildrens.icims.com/

Email: Sean.Taylor@seattlechildrens.org or jay.sarthy@seattlechildrens.org for more information

• The Ben Towne Center for Childhood Cancer Research (BTCCCR) at SCRI is looking for

• Experience with omics or other high dimensional data analysis and pipelines, e.g. nextgeneration sequencing, proteomics, metabolomics, epigenomics, phenotypic readouts, imaging, and clinical data, including experience developing and using statistical models

• Work together with pediatric cancer researchers to improve the lives of children with





Acknowledgements

Sami Kanaan **Shruti Bhise Olivia Waltner Rula Green-Gladden**

Jeffrey Stevens Todd Cooper Melinda Biernacki Marie Bleakley **Monica Thakar**

Cole Trapnell

Jason Underwood Vijay Ramani

UNIVERSITY of WASHINGTON

Soheil Meshinchi Rhonda Reis Jenny Smith

OUR PATIENTS AND THEIR FAMILIES



FRED HUTCH **CURES START HERE®**







Chan Zuckerberg Initiative 🛞





Questions?