



Single-cell Navigator: visualizing scRNA-seq data

Konstantin Zaitsev, ITMO University August 27th, 2021. Tomsk / My Hotel Room.



Visualizing scRNA-seq data

Main goals:

- To make hypothesis generations easier
- Remove "man-in-the-middle"

Extra goals:

- Fast
- Responsive



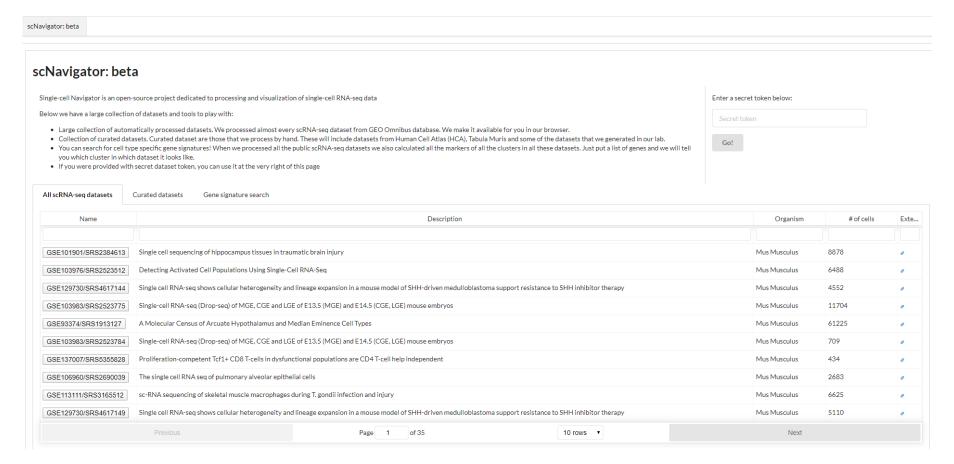
Visualizing scRNA-seq data

https://artyomovlab.wustl.edu/scn/

(still in production, so feedback is very welcome)



Let's open the dataset





Let's open the dataset

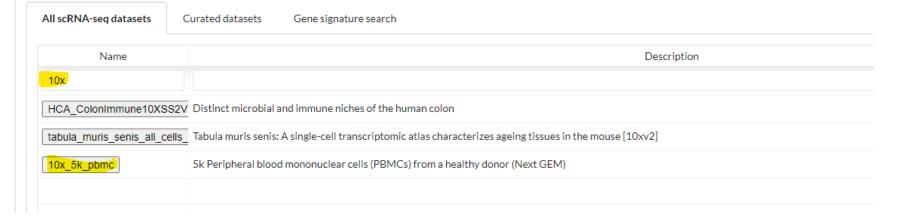
- ✓ Go to https://artyomovlab.wustl.edu/scn/
- Search for 10x
- And click on the dataset

scNavigator: beta

Single-cell Navigator is an open-source project dedicated to processing and visualization of single-cell RNA-seq data

Below we have a large collection of datasets and tools to play with:

- Large collection of automatically processed datasets. We processed almost every scRNA-seq dataset from GEO Omnibus database. We make it available for you in our browser.
- Collection of curated datasets. Curated dataset are those that we process by hand. These will include datasets from Human Cell Atlas (HCA), Tabula Muris and some of the dataset
- You can search for cell type specific gene signatures! When we processed all the public scRNA-seq datasets we also calculated all the markers of all the clusters in all these datase
 you which cluster in which dataset it looks like.
- If you were provided with secret dataset token, you can use it at the very right of this page



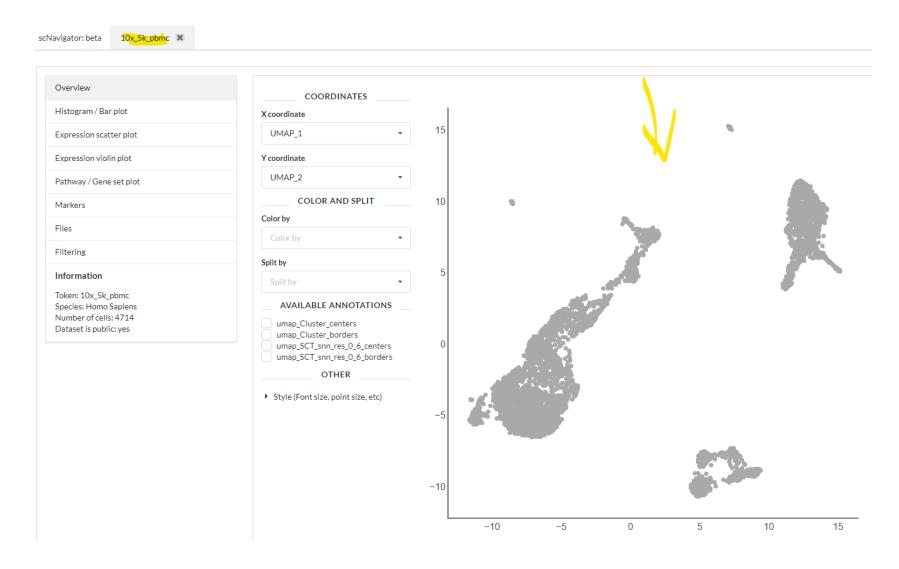


If you have any problem finding dataset

✓ Just go to https://artyomovlab.wustl.edu/scn/?token=10x 5k pbmc



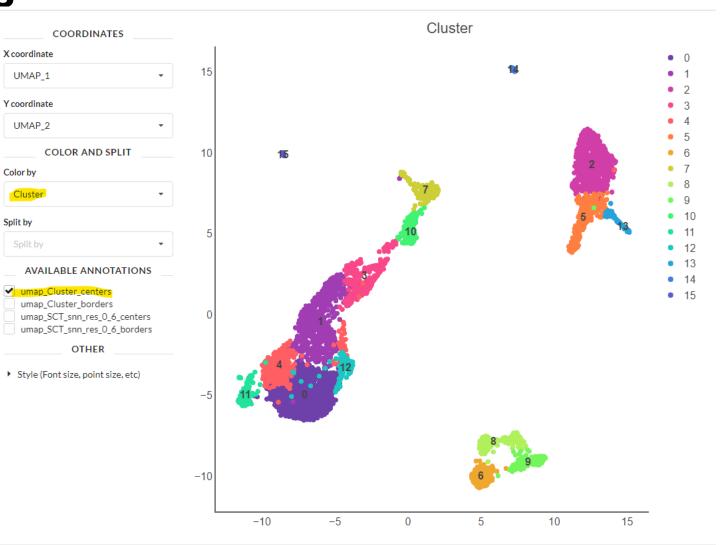
Result should look like that





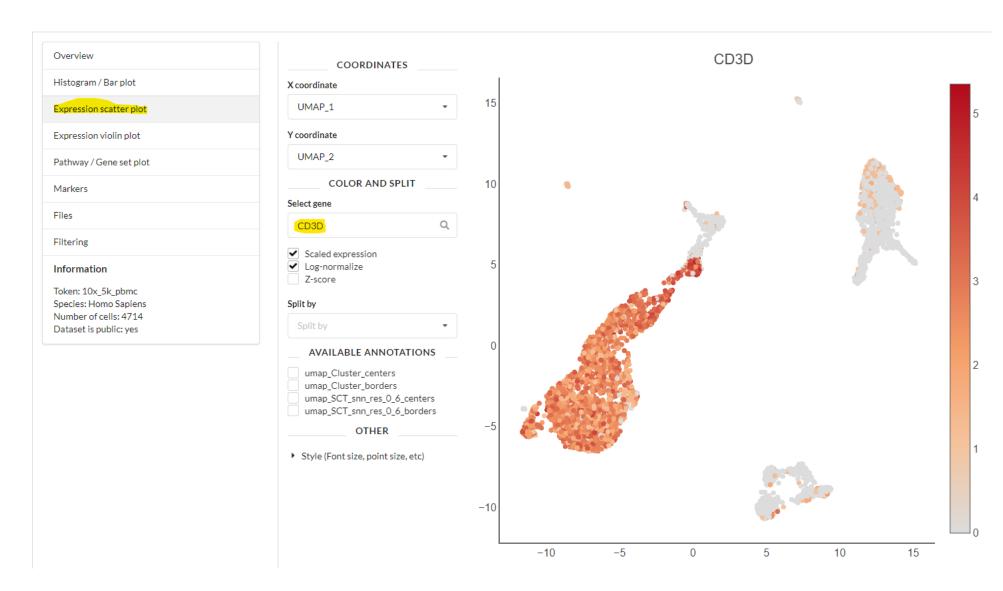
We can color the cells

- Cluster
- Number of UMIs
- Number of genes detected
- umap_Cluster_centers



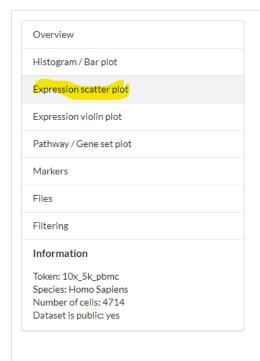


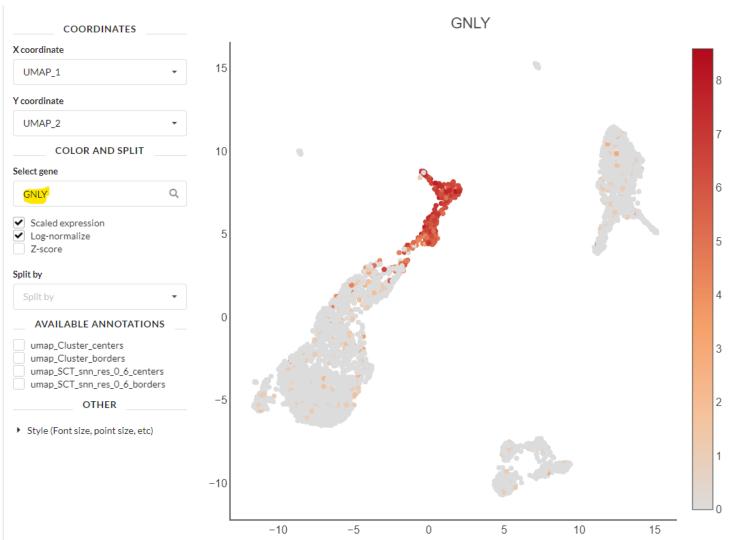
Expression of CD3d





Or you can go for any of your favorite genes







Expression scatter plot

- Expression scatter plot shows gene expression in each cell
- We can see that expression of some genes is localized with clusters



Violin plot

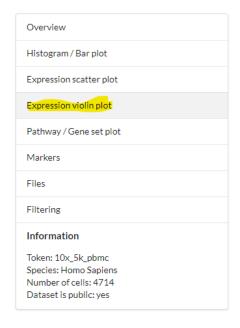
X coordinate

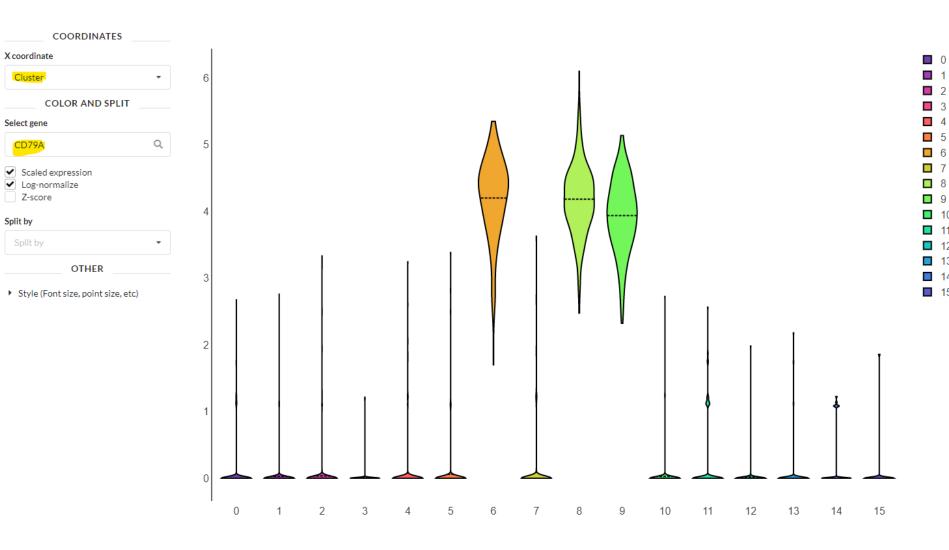
Cluster

Select gene

Z-score

Split by





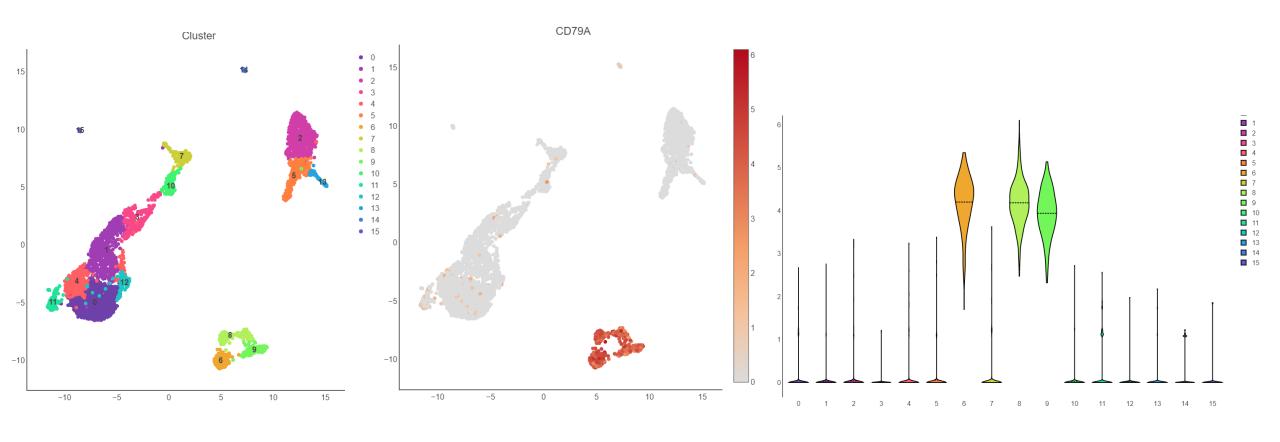


Violin plot

- Violin plot shows distribution of gene expression within several groups of cells (in our case groups are clusters)
- Higher the violin higher the expression in the group



Cd79a: expression scatter and expression violin





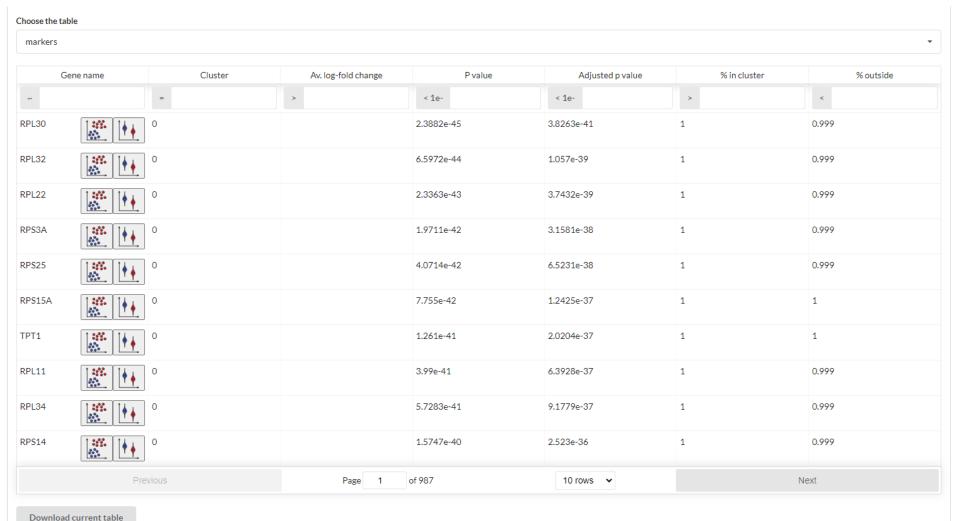
Markers

- Usually we run differential expression to identify cluster markers
- You can compare a cluster against all the other clusters and identify genes that have higher expression than in the other clusters



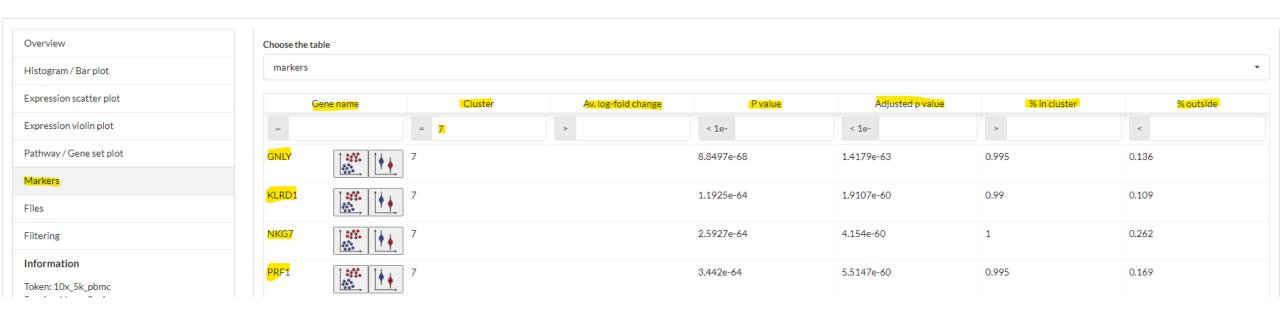
Markers tab







Markers tab: what's the cluster 7?



- **♥** GNLY gene name
- ♥ Cluster 7 we are checking results for cluster 7 vs other clusters
- Average log-fold change: average difference between expression of GNLY in cluster 7 and in other clusters
- ▼ P value (we test difference between average expression of this gene inside and outside cluster 7).
- P adjusted adjusted p value for multiple hypothesis



Markers tab: what's the cluster 7?

- You have two buttons next to the gene name
- 1) First will open gene expression on scatter plot
- 2) Second will open gene expression on violin plot

Choose the table										
markers										
G	ene name	Cluster	Av. log-fold change	P value	Adjusted p value	% in cluster	% outside			
~		= 7	>	< 1e-	< 1e-	>	<			
GNLY	7	,		8.8497e-68	1.4179e-63	0.995	0.136			
KLRD1	7	,		1.1925e-64	1.9107e-60	0.99	0.109			



Now let's play with it

I want you to check out any other genes



Public datasets

- We try to process many other public datasets trying to make them available to scientific community
- You can always go back to the main tab (top left corner)



Public datasets

scNavigator: beta 10x_5k_pbmc X

scNavigator: beta

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- You can search for cell type specific gene signatures! When we processed all the public scRNA-seq datasets we also calculated all the markers of all the clusters in all these datasets. Just put a list of genes and we will tell you which cluster in which dataset it looks like.
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Name	Description		Organism	# of cells	Exte.
SE101901/SRS2384613	Single cell sequencing of hippocampus tissues in traumatic brain injury		Mus Musculus	8878	0
SE103976/SRS2523512	Detecting Activated Cell Populations Using Single-Cell RNA-Seq	Mus Musculus	6488	ø	
SE129730/SRS4617144	Single cell RNA-seq shows cellular heterogeneity and lineage expansion in a mouse model of SHH-driven medulloblastoma support resistance	Mus Musculus	4552	0	
SE103983/SRS2523775	Single-cell RNA-seq (Drop-seq) of MGE, CGE and LGE of E13.5 (MGE) and E14.5 (CGE, LGE) mouse embryos	Mus Musculus	11704	0	
SE93374/SRS1913127	A Molecular Census of Arcuate Hypothalamus and Median Eminence Cell Types	Mus Musculus	61225	0	
SE103983/SRS2523784	Single-cell RNA-seq (Drop-seq) of MGE, CGE and LGE of E13.5 (MGE) and E14.5 (CGE, LGE) mouse embryos			709	8
SE137007/SRS5355828	Proliferation-competent Tcf1+ CD8 T-cells in dysfunctional populations are CD4 T-cell help independent				0
SE106960/SRS2690039	The single cell RNA seq of pulmonary alveolar epithelial cells			2683	0
SE113111/SRS3165512	sc-RNA sequencing of skeletal muscle macrophages during T. gondii infection and injury	Mus Musculus	6625	8	
SE129730/SRS4617149	Single cell RNA-seq shows cellular heterogeneity and lineage expansion in a mouse model of SHH-driven medulloblastoma support resistance	ce to SHH inhibitor therapy	Mus Musculus	5110	0



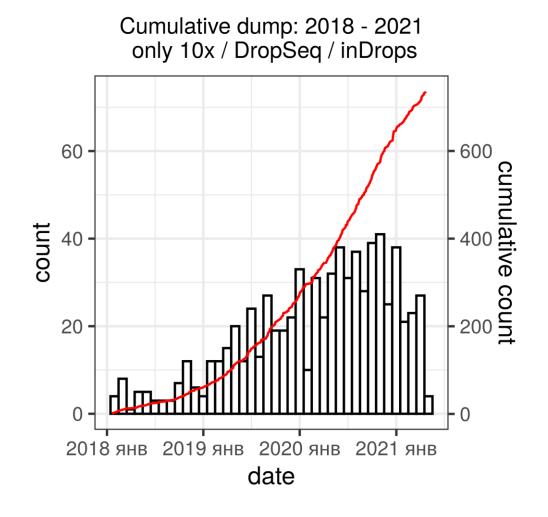
Public datasets

- Two main sources
 - NCBI GEO (Gene Expression Omnibus, https://www.ncbi.nlm.nih.gov/geo/)
 - EMBL EBI (European Bioinformatics Institute, part of EMBL, https://www.ebi.ac.uk/)



Increasing number of public datasets

- Number of public GSE dataset which were qualified as scRNA-seq
- Histogram and cumulative count





Public scRNA-seq datasets

Most of the scRNA-seq datasets are available at NCBI GEO (or SRA) Problems are:

- Different technologies used to perform experiment (10x, DropSeq, SmartSeq2, C1 Fluidigm etc)
- Different pipelines were used to analyze
- Different formats in which data is kept

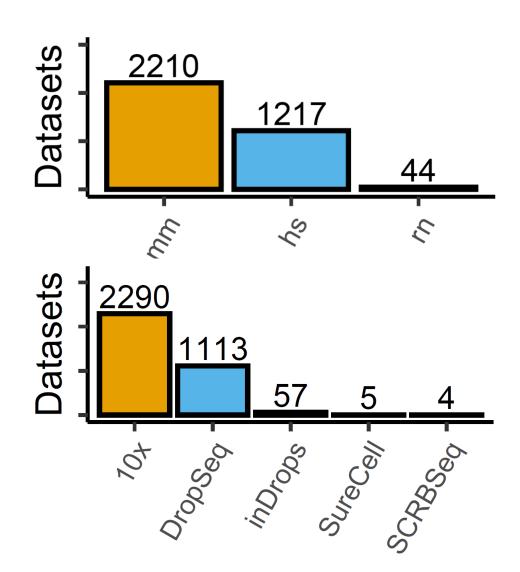
Most of the dataset processing was done by Maria Firuleva





Current database snapshot

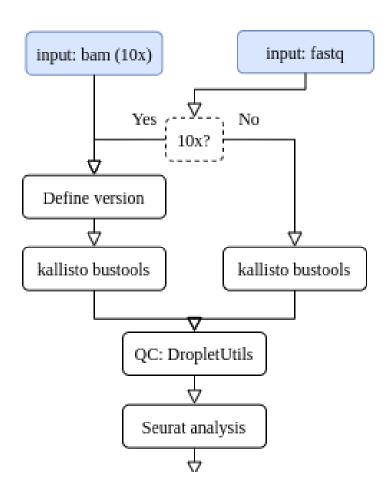
- Total ~3500 single-cell samples (as of two weeks ago)
- Some merged GSEs are available
- We will troubleshoot unprocessed datasets
- We tested the pipeline for bulk-like scRNA-seq dataset, will start those soon





How do we process single-cell RNA-seq

- Determine chemistry version
- Kallisto Bustools (from both reads and bams)
- EmptyDrops to remove noise
- Seurat analysis
- SCNPrep





Seurat analysis

Seurat is an R package for analysis of single-cell RNA-seq data

- Some more QC: removing cells with high mito-content
- Normalization
- (if dataset consists of multiple samples) merging samples together
- PCA
- Dimensionality reductions: both tSNE and UMAP
- Clustering
- Markers identification

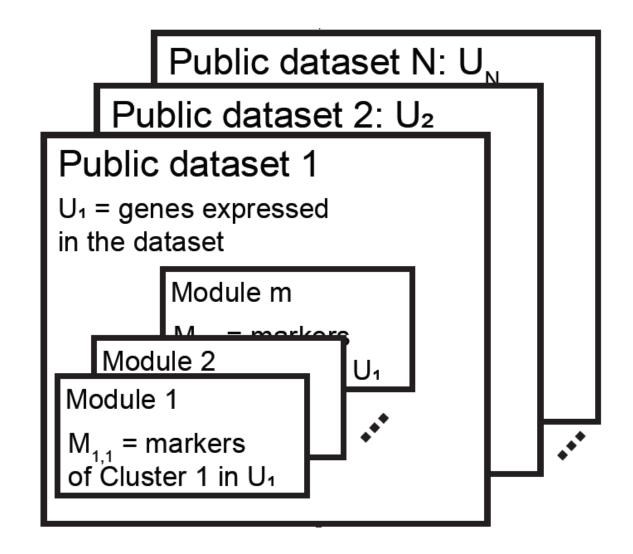


GeneQuery inspired gene signature search

- Having this large database of public scRNA-seq datasets we wanted to implement gene signature search
- Given with a list of gene, we can match it against markers of all the clusters present in our database

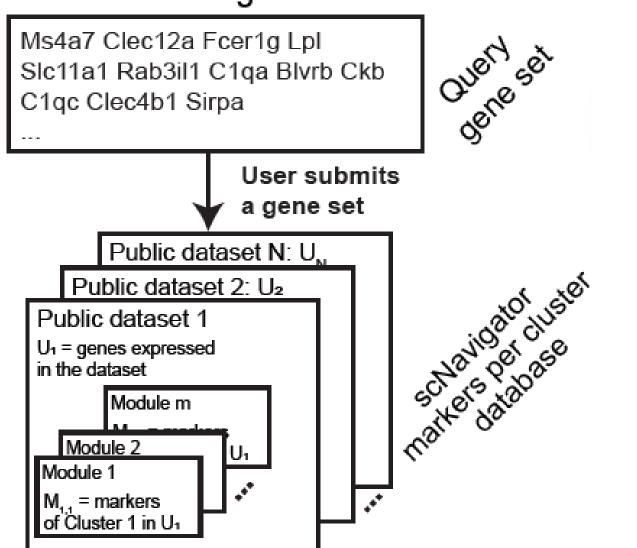


Database of all the markers in all the single-cell clusters





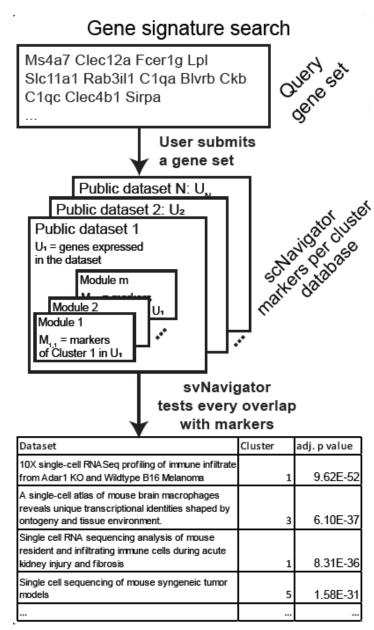






Gene Signature Search

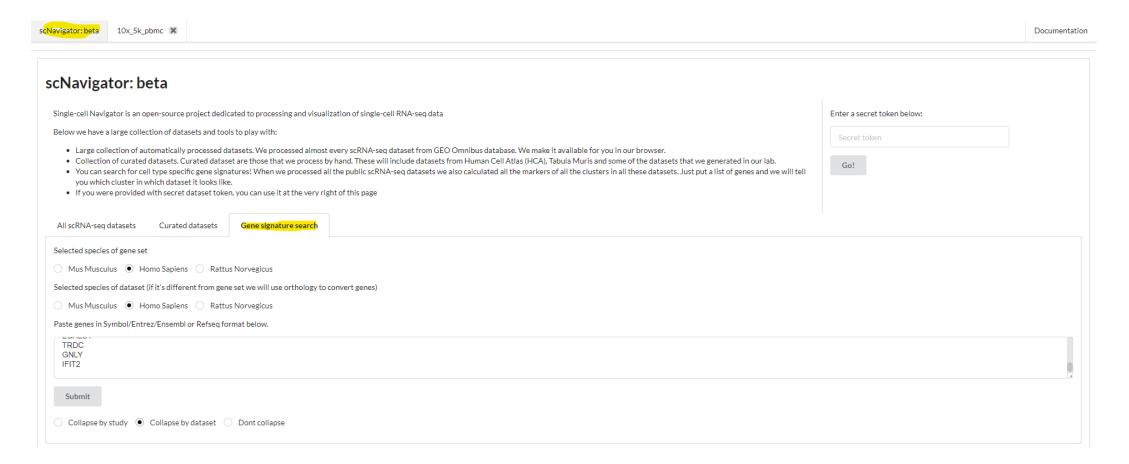
- User submits a gene set
- We compare gene set against markers of all the populations
- Find datasets and clusters where signature is expressed





Example: top 50 genes

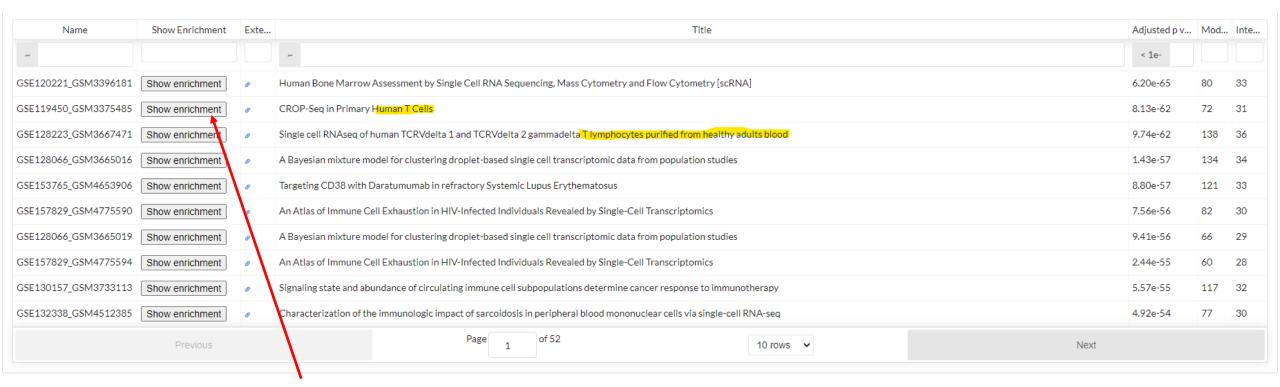
You can select top 50 genes





Example: top 50 genes

We get a lot of results



Press the button



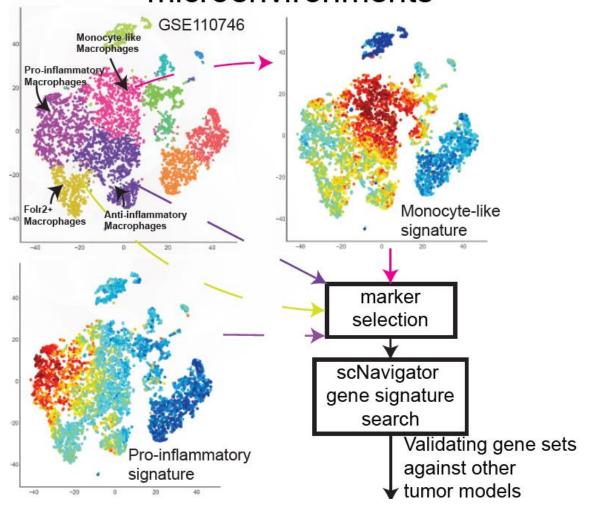
We can see enrichment of these genes in other datasets





Study case: tumor microenvironments

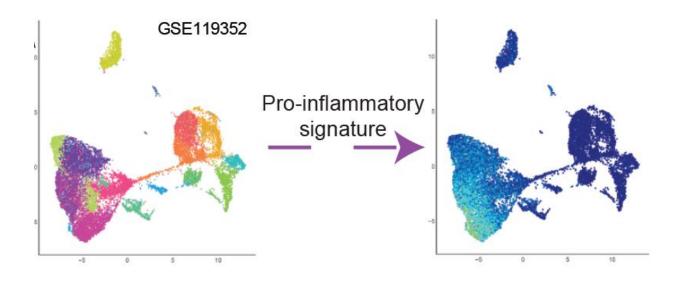
Crossmatching immune tumor microenvironments





Signature of macrophage populations were well-conserved in different mouse tumor models

		Activition	Pro-inflat	nratory Monocyt	EJIKE FOLZEX
GSE110746	B16 tumor model (this dataset)	7.68E-52	4.36E-48	8.01E-46	2.33E-46
GSE119352	D42m1 tumor model	7.41E-29	3.22E-33	2.41E-45	4.48E-16
GSE112865	MC38 tumor model	2.90E-25	2.33E-18	2.54E-26	1.68E-07
GSE121861	LL2/SA1 tumor models	1.59E-31	1.82E-15	8.56E-32	1.01E-10





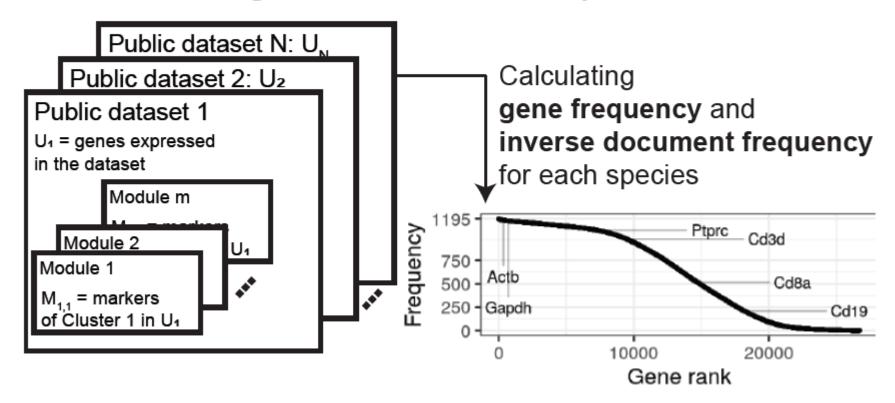
What else can we do with it

- We obtained the markers for all the populations and datasets
- We can figure out the similarity between these populations



Not all the genes are interesting

Universe of the published single-cell RNA-seq data





Building similarity networks

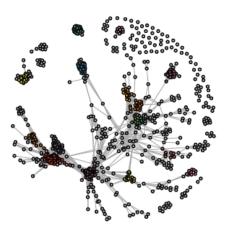
Each dataset and each cluster is an **IDF vector** now



Cosine similarity of dataset IDF vectors



Cosine similarity of cluster IDF vectors



revaling samples that were sorted/taken from same/similar tissue

revaling cell-type specific gene signature across all the datasets



Conclusion

- We hope that single-cell navigator will make interpretation of scRNA-seq data easier
- https://artyomovlab.wustl.edu/scn/
- We try to get there as much datasets as we can
- If you want to use SCN for your private data:
 - You can just e-mail me <u>kzaitsev@itmo.ru</u>
 and I will give you a private link to your data
 - Wait until it gets published (ETA?), you will be able to host SCN locally, or for your department



Extra gene signatures from the study case

- Anti-inflammatory: Ms4a7 Clec12a Fcer1g Lpl Slc11a1 Rab3il1 C1qa Blvrb Ckb C1qc Clec4b1 Sirpa Fcgr4 Grn Pycard C1qb Adgre1 Ctsc Cd72 Clec4a1 Hexa Aif1 Clec4a2 Lst1 Slamf9 Lgmn AF251705 Nr1h3 Cd300e Ctsb
- Pro-inflammatory: Arg1 Adam8 Ninj1 Mmp12 Basp1 Slc2a1 Hilpda Cstb Il1rn Clec4d Il7r Ndrg1 Hmox1 Ftl1 Cd36 Lgals3 Fabp5 Cxcl2 Plin2 Emp1 Rgcc Bnip3 Egln3 Thbs1 Fth1 Ctsl Spp1 Card19 Ero1l Fabp4
- Monocyte-like: Il1b Gm9733 Btg2 Ccr2 Zbp1 Plbd1 Ifitm3 H2-DMa Ly6i Plac8 Spi1 Osm Ms4a6c Samhd1 Cybb Lyz2 Naaa Fos Ms4a4c H2-DMb1 Hp Prdx5 Junb Cd74 Tgfbi Ly6c2 Slamf8 Klra2 Zfp36 Scimp
- Folr2+: Sepp1 Trf Cd163 Apoe Mrc1 Fxyd2 Fcgrt Igf1 Ccl24 Folr2 Itm2b Igfbp4 F13a1 Ednrb Tmem37 Gas6 Ltc4s Glul Cbr2 C4b Wfdc17 Pltp Lyve1 Cd209f Clec10a Npl Pf4 Timp2 Rnase4 C1qc