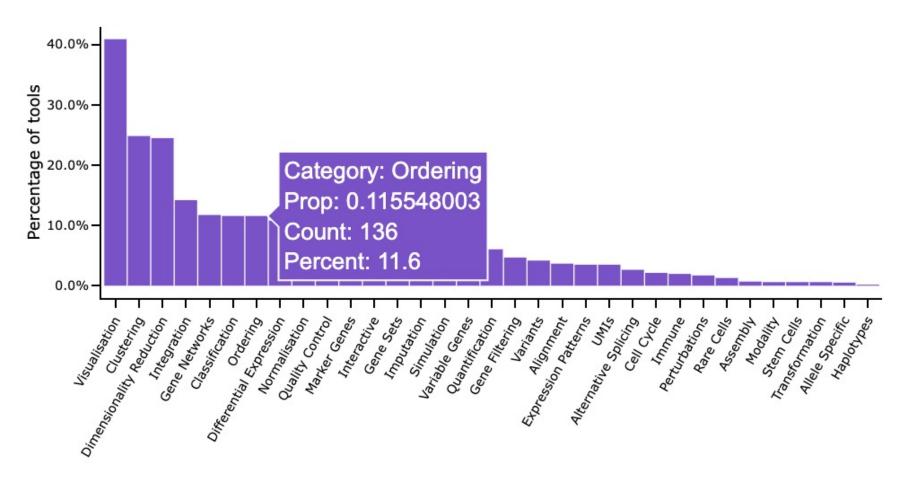
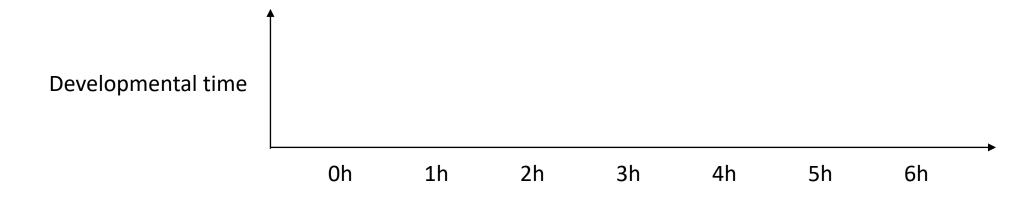


### Again long list of possible tools



#### Example of Trajectories



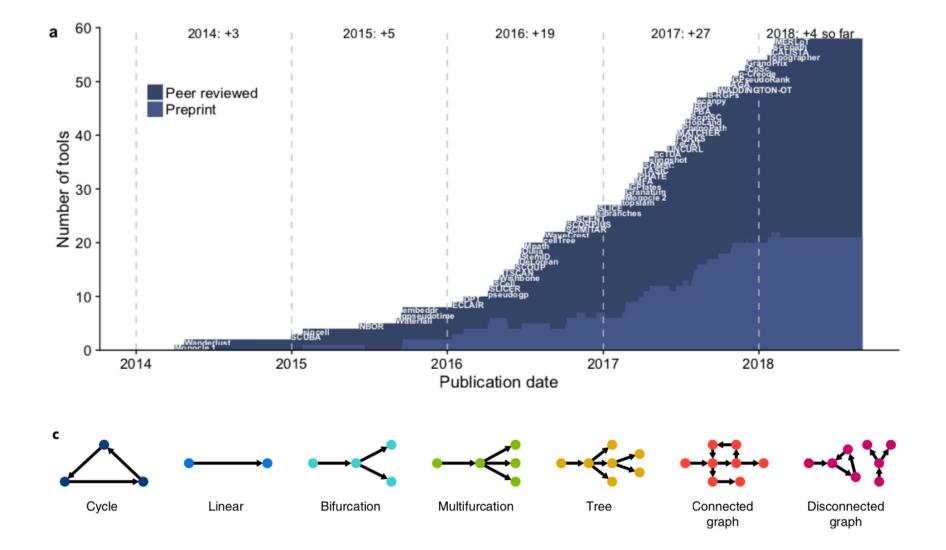
- •In the analysed data set one might encounter:
- •1. Cells that differentiate display a **continuous spectrum** of states Transcriptional program for activation and differentiation
- •2. Individual cells will differentiate in an unsynchronized manner Each cell is a snapshot of **differentiation** time
- •OR 3. Pseudotime abstract **unit of progress**Distance between a cell and the start of the trajectory

## Should you run trajectory inference

- Are you sure that you have a developmental trajectory?
- Do you have intermediate states?
- Do you believe that you have branching in your trajectory?

Be aware, any dataset can be forced into a trajectory without any biological meaning!

First make sure that gene set and dimensionality reduction captures what you expect.



Saelens et al (2019) Nat Biotechnology

#### Trajectory analysis

- Differences in gene expression between cells, might be attributed to dynamic processes:
  - Cell cycle
  - Cell differentiation
  - Response to an external stimuli
- Trajectory inference can order a set of individual cells along a path / trajectory / lineage
- Some methods project cells onto a pseudotime axis others project each cell along a path.
- This can be a starting point for further analysis to determine gene expression programs driving interesting cell phenotypes.

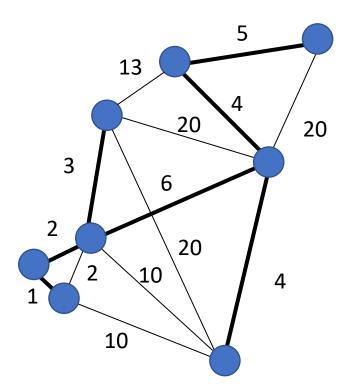
#### Example of application

- From the paper Single-Cell RNA-Seq Reveals Dynamic, Random Monoallelic Gene Expression in Mammalian Cells (Deng et al. 2014)
- « To investigate allele-specific gene expression at single-cell resolution, we isolated 269 individual cells dissociated from in vivo F1 embryos (CAST/EiJ × C57BL/6J, hereafter abbreviated as CAST and C57, respectively) from oocyte to blastocyst stages of mouse preimplantation development (PD)»
- Here finding a trajectory between the cells might be of high interest.

#### Minimum spanning tree

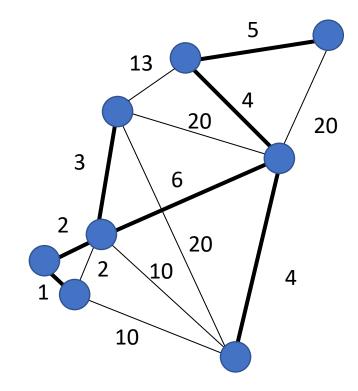
- Take a weighted graph.
- Take a spanning tree
- Take the minimum of all spanning trees.

### Example

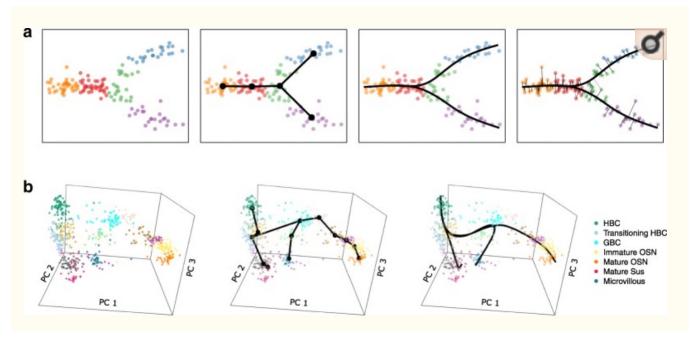


# Minimum spanning tree (MST)

- Sum of all distances in the tree (graph) is at its minimum
- Having more transitional cells improves the definition of the tree
- The weights can be a distance in the dimensionality reduction space (ICA, T-SNE, UMAP, diffusion maps) or a correlation between cells, etc.
- MST has no cycles, cell cycles will not work in here



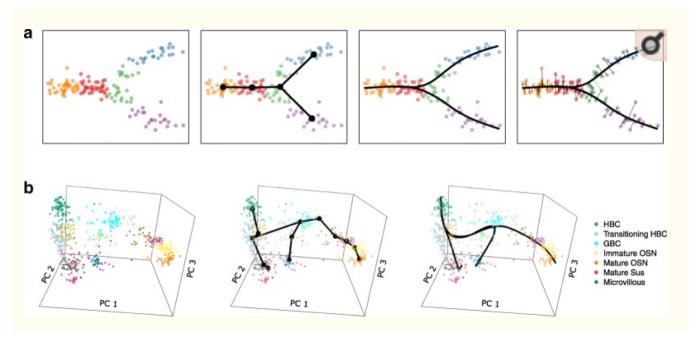
#### Slingshot (Street et al 2018)



1. Distance between clusters

$$d^2\Big(\mathscr{C}_i,\mathscr{C}_j\Big) \equiv \left(\overline{X}_i - \overline{X}_j\right)^T \left(S_i + S_j\right)^{-1} \left(\overline{X}_i - \overline{X}_j\right),$$

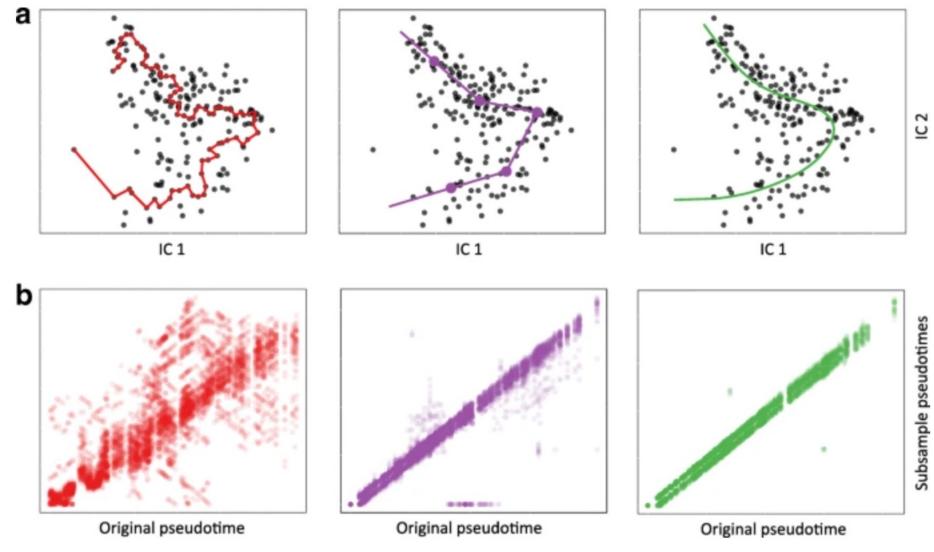
### Slingshot (Street et al 2018)



- 1. Distance between clusters
- 2. Infer lineages by ordering cell clusters and construct MST
- 3. Construct principal curves\*

<sup>\*</sup>Principal curves are smooth one-dimensional curves that pass through the middle of a p-dimensional data set, providing a nonlinear summary of the data. They are nonparametric, and their shape is suggested by the data

### Slingshot vs others



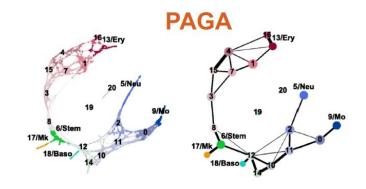
### Monocle3 uses an algorithm based on PAGA (python)

- PAGA constructs a k-nearest neighbour graph on cells and then identifies 'communities' of cells via the Louvain method.
- Two vertices (**Louvain communities**) are linked with an edge, when the cells in the respective communities are neighbours in the *k*-nearest neighbour graph.
- Monocle 3 constructs a k-nearest neighbour graph (k = 20) on cells in the UMAP space, then grouping them into Louvain communities, and testing each pair of communities for a **significant number** of links between their respective cells.
- Those communities that have more links than expected under the null hypothesis of spurious linkage (FDR <1%) remain connected in the PAGA graph, and those links that fail this test are severed. (correction of spurious linkage)

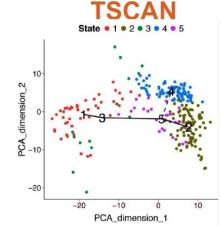
### RNA Velocity a quite different algorithm

- « RNA velocity is a high-dimensional vector that predicts the future state of individual cells on a timescale of hours»
- «aid the analysis of developmental lineages and cellular dynamics»
- Method: calculate the relative abundance of nascent (unspliced) and mature (spliced) mRNA to estimate the rates of gene splicing and degradation
- During a dynamic process:
  - increase in the transcription rate=> rapid increase in unspliced mRNA=> increase in spliced mRNA until a new steady state is reached.
  - a drop in the rate of transcription => drop in unspliced mRNA => reduction in spliced mRNAs.
- During induction of gene expression: => unspliced mRNAs are present in excess,
- During repression: => unspliced mRNAs are present in lower amounts.
- Hence: The balance of unspliced and spliced mRNA abundance is, therefore, an indicator of the future state of mature mRNA abundance, and thus the future state of the cell.

#### Some additional tools

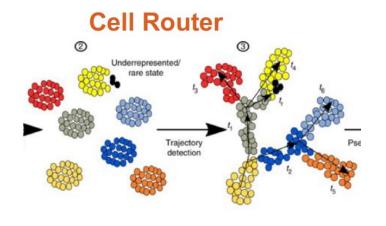


Street et al (2019) Genome Biology



Zhicheng et al (2016) Nuc Acid Res

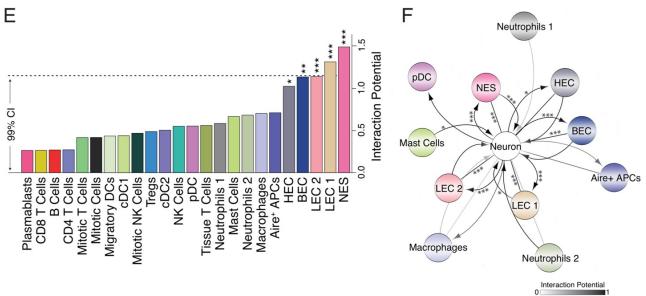
Spade, StemID 2, Eclair, TSCAN and Mpath use different clustering algorithms such as k-means, k-medoids, hierarchical clustering or DBSCAN in a dimensionality-reduced space.



Da Rocha et al (2018) Nat Commun

#### Other Post-hoc analysis

- LRIP Ramilowski et al, Nat comm, 2015- bioarxiv, Huang et al,xmol, 2020
- CellphoneDB <a href="https://www.cellphonedb.org/">https://www.cellphonedb.org/</a> online « clickable » Mirjana Efremova, Nat protocols, 2020.
- NicheNet needs apriori knowledge, Robin Browaeys, Nat met, 2020.
- CellChat- http://www.cellchat.org/



LRIP - Bioarxiv, <a href="https://doi.org/10.1101/833509">https://doi.org/10.1101/833509</a>

#### NicheNet- Ligand receptor analysis

 Question: In your analysis you have a certain list of DGE genes. Can one associate a pair of Ligand and Receptor responsible for the change in expression of those genes?

 This is extremely useful as it will point biologist to possible pathways to target.

#### Prior model of ligand-target regulatory potential

	Target_1	Target_2	Target_3	•••	Target_n
Ligand_1	P11	P12	P13		P1n
Ligand_2	P21				
Ligand_m	Pm1	Pm2	Pm3		Pmn

#### Prior model of ligand-target regulatory potential

Calculate correlation scores

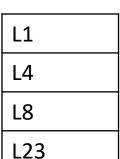
Select highest scores (scores are not meant to be so high due to high number of 0s)

Select only expressed ligands

	Target_1	Target_2	Target_3	•••	Target_n
Ligand_1	P11	P12	P13		P1n
Ligand_2	P21				
Ligand_m	Pm1	Pm2	Pm3		Pmn

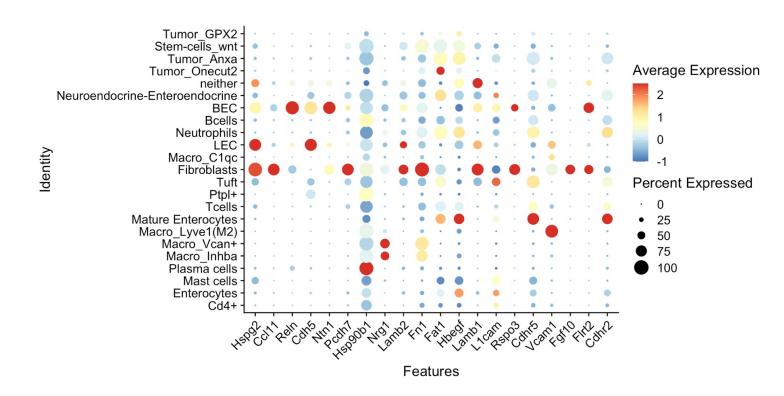
	Target_1	Target_2	Target_3	•••	Target_n
DGE	0	1	1		0

List of Ligands Where are they expressed + Are they actually Differentially expressed as well in that cell type



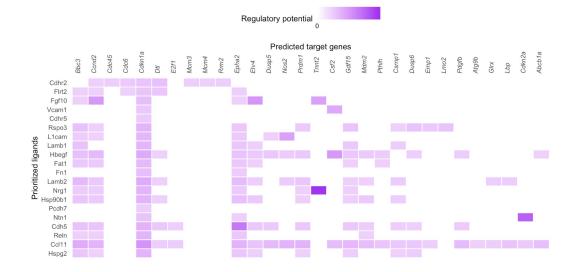
L50



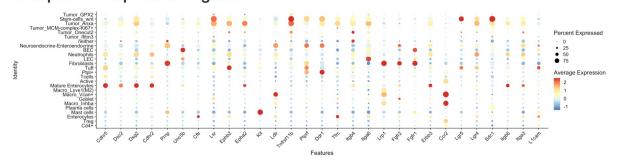


List of Receptors associated to the potential ligands

Where are they expressed + Are they actually Differentially expressed as well in that cell type

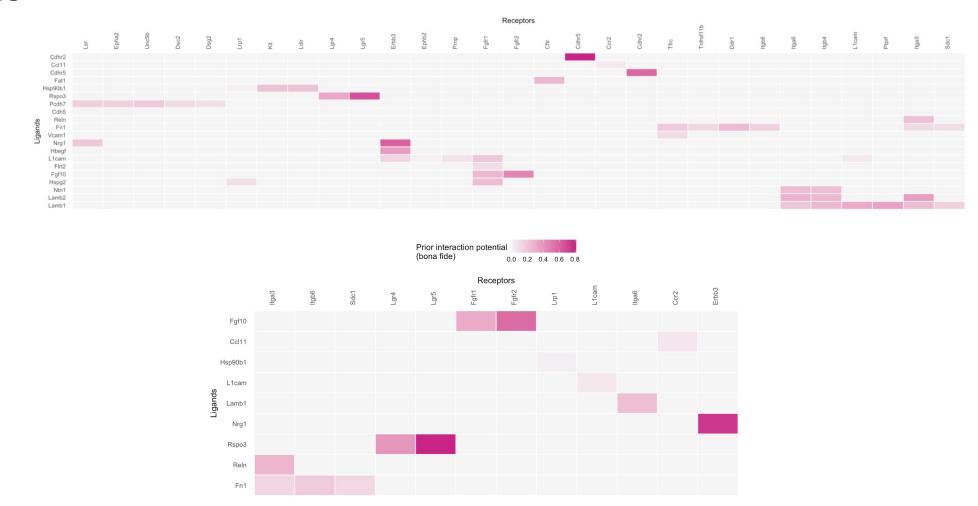


#### Receptors of top-ranked ligands



List of Ligand Receptors with interaction potential

Bona fide



# In the cluster of cells where the receptor is expressed

