Trajectory analysis

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CQS Summer Academy (8/15/2018)

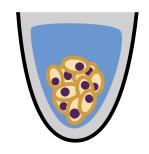
http://www.mc.vanderbilt.edu/vumcdept/cellbio/laulab/index.html
Twitter: @KenLauLab



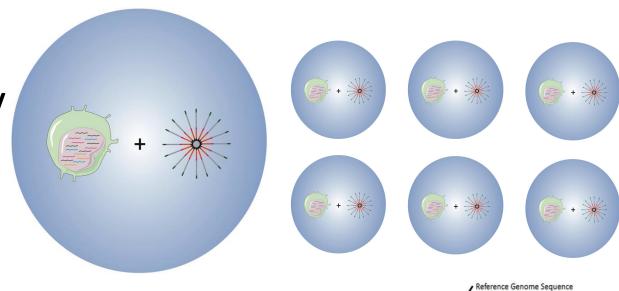




Single cell suspension***

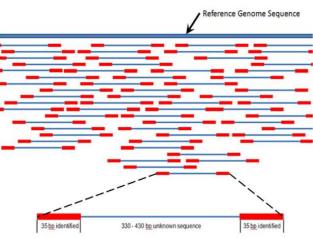


Single-cell encapsulation/ Library preparation

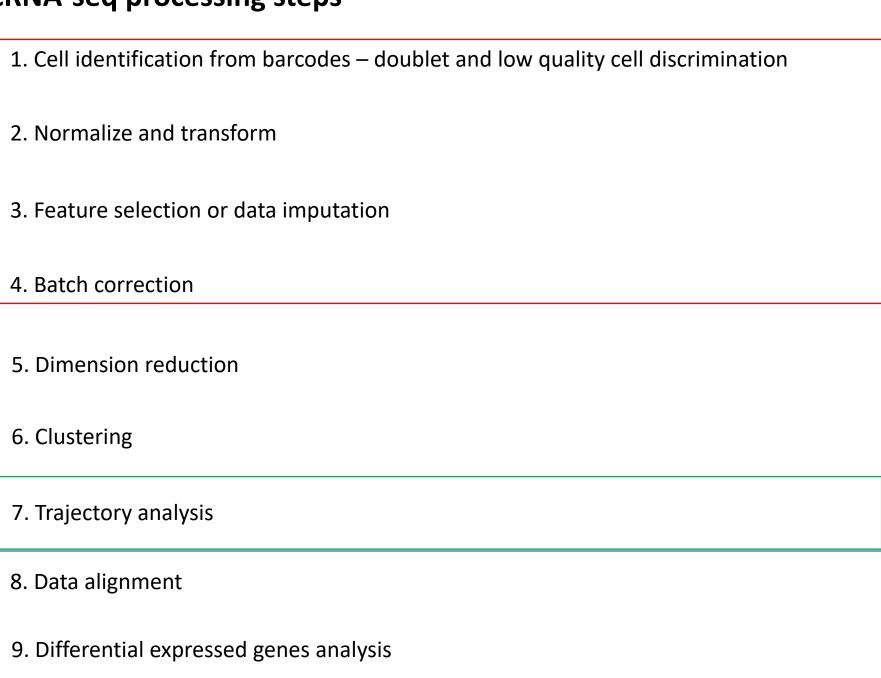


Sequencing and alignment (Bioinformatics I)



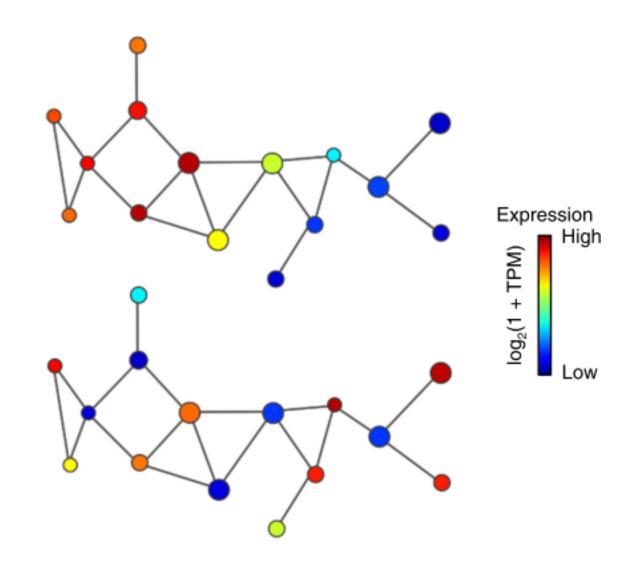


scRNA-seq processing steps

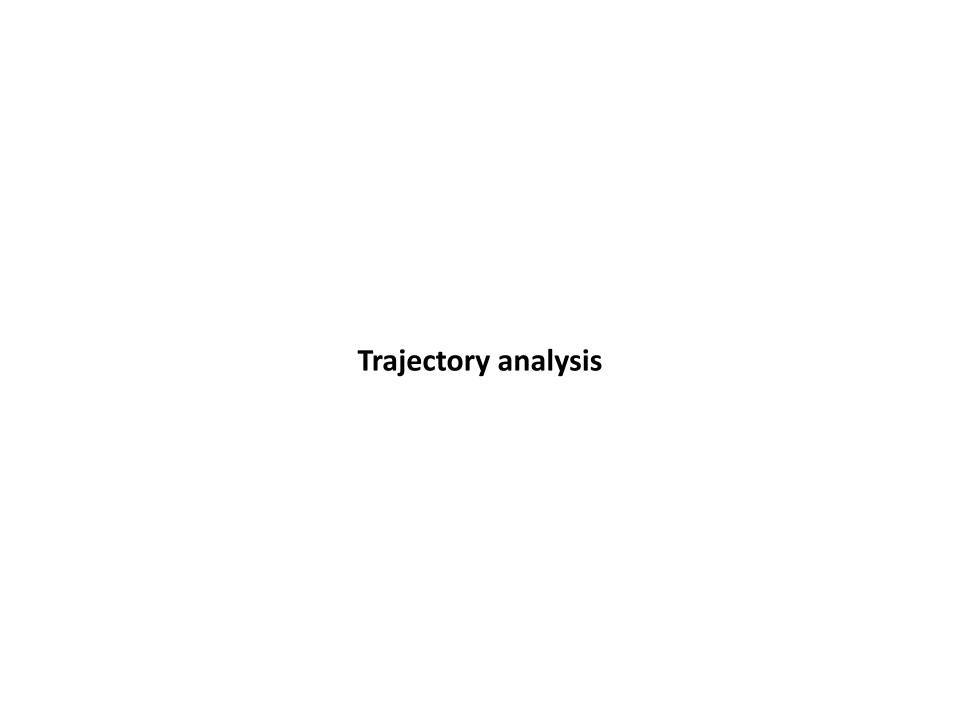




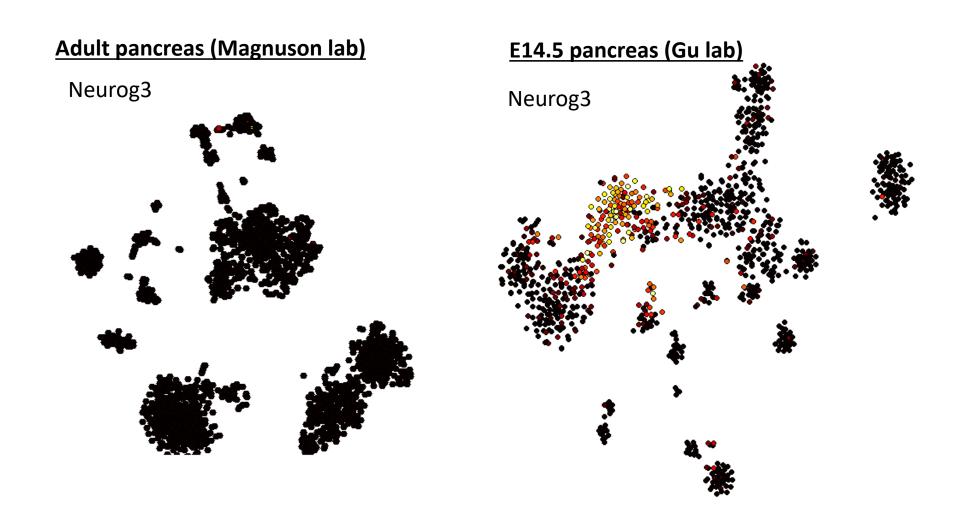
Neighborhood variance feature selection



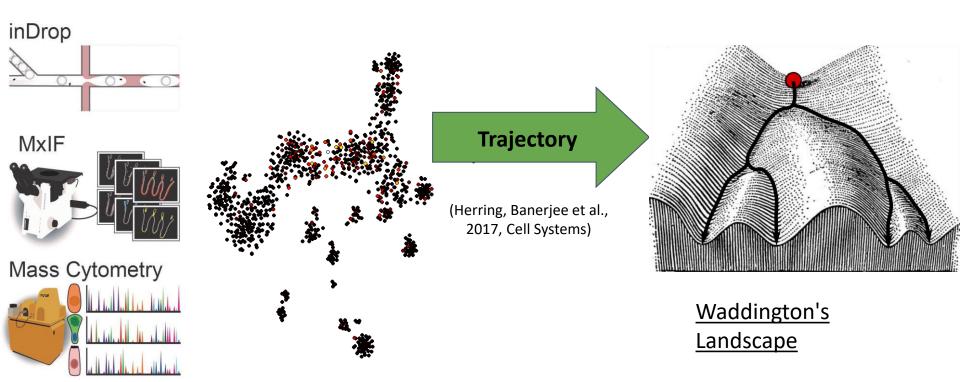
Allows for selection of genes that change monotonically over all cells



The pancreatic landscape is different between the adult and embryonic stages

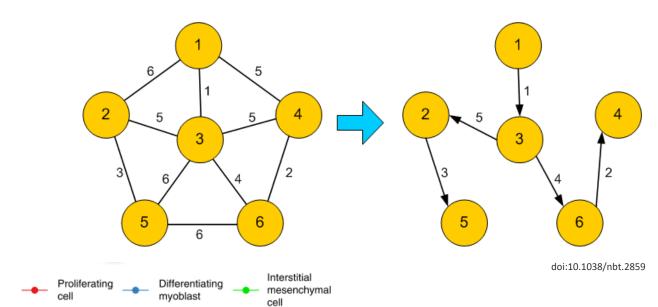


Single-cell data continnum represented as transitional trajectories



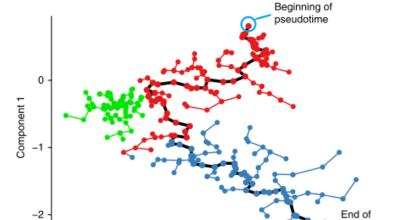
Minimum spanning tree algorithms

$$1 + 5 + 3 + 4 + 2 = 15 = Tree Score$$



pseudotime

-2



Component 2

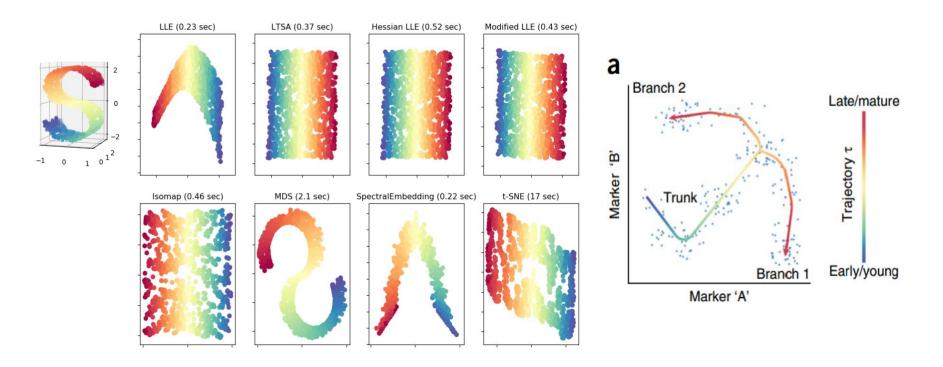
-3

- Monocle (Trapnell et al., 2014)
- TSCAN (Ji and Ji, 2016)
- Waterfall (Shin et al., 2015)
- SPADE (Qiu *et al.*, 2011)
- Slingshot (Street et al. BioRxiv)
- ... and many others

Problem: Not robust. Different results at different runs with the same data.

Non-linear embedding algorithms

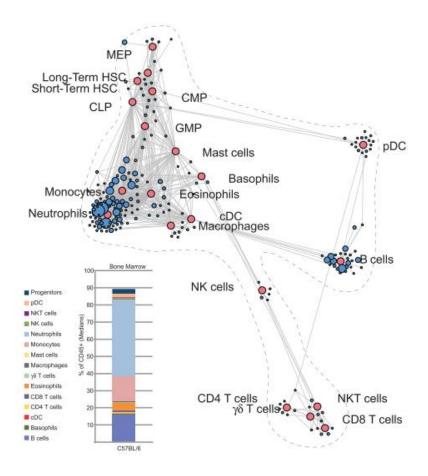
Non-linear embedding



- SLICER (Welch et al., 2016) LLE
- Diffusion Map Pseudotime(Haghverdi et al., 2015) Diffusion Map
- Wishbone (Setty et al., 2016) Diffusion Map plus network
- scTDA (Rizvi et al., 2017) MDS, Mapper

Problem: Stretching and compressing of data a tradeoff between increased resolution and noise. Thus, limited to simple structures.

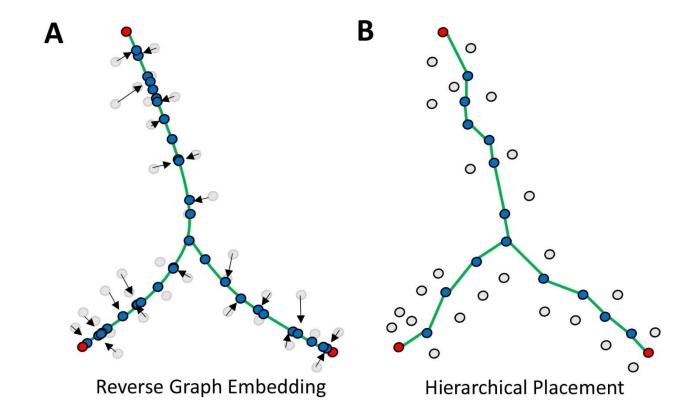
Force directed algorithms



- FLOW-Map (Eli Zunder)
- Scaffold (Garry Nolan)
- SPRING (Allon Klein)

Problem: Complex structures difficult to interpret; visualization only

Graph methods



- Monocle2 (Trapnell, 2017)
- p-Creode (Lau, 2017)
- CellRouter (Collins and Daley, 2018)
- PAGA (Wolf and Theis, 2018)
- URD (Regev and Schier, 2018)

cmgh

CELLULAR AND MOLECULAR
GASTROFATEROLOGY AND HEPATOLOGY

REVIEW

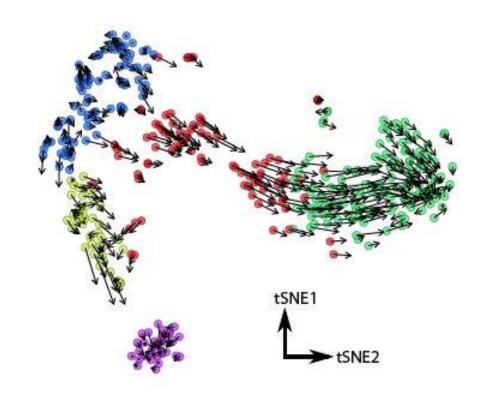
Single-Cell Computational Strategies for Lineage Reconstruction in Tissue Systems



Charles A. Herring, 1,2 Bob Chen, 1,3 Eliot T. McKinley, 1,4 and Ken S. Lau 1,2,3

¹Epithelial Biology Center, ⁴Department of Medicine, Vanderbilt University Medical Center, Nashville, Tennessee; ²Program in Chemical and Physical Biology, ³Department of Cell and Developmental Biology, Vanderbilt University School of Medicine, Nashville, Tennessee

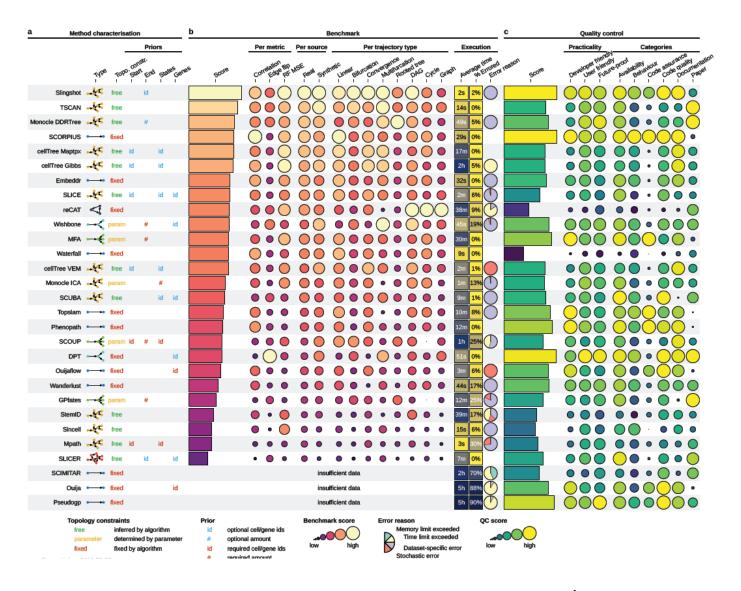
Methods that cheat (include prior/extra knowledge)

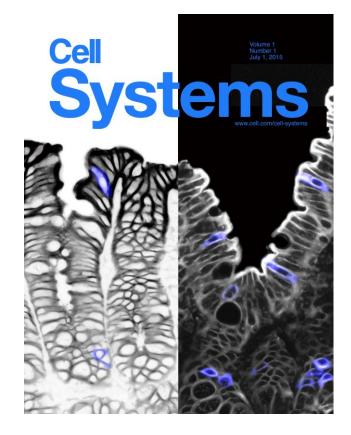


- Velocyto (Kharchenko splicing)
- SLICE (Yan Xu gene ontology grouping entropy)
- scdiff (Ding et al. TF→gene expression relationships)
- SCENT (Teschendorff et al. protein-protein interaction node entropy)

Problem: Depends on how good is the prior knowledge

Single-cell data analysis is a field in its infancy



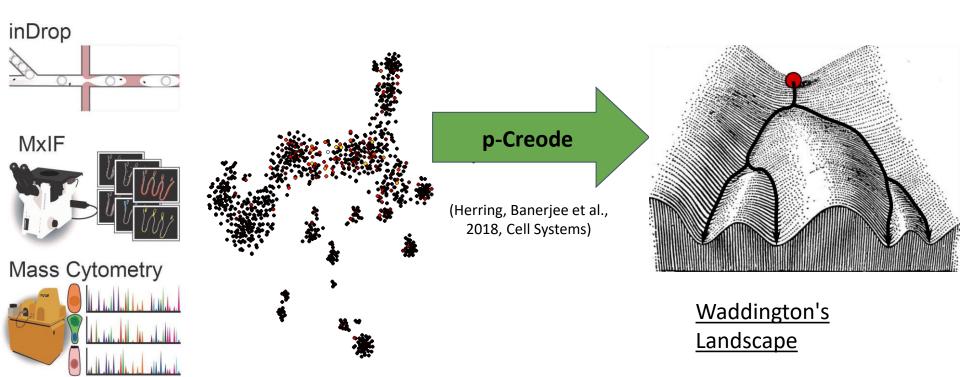


The p-Creode algorithm

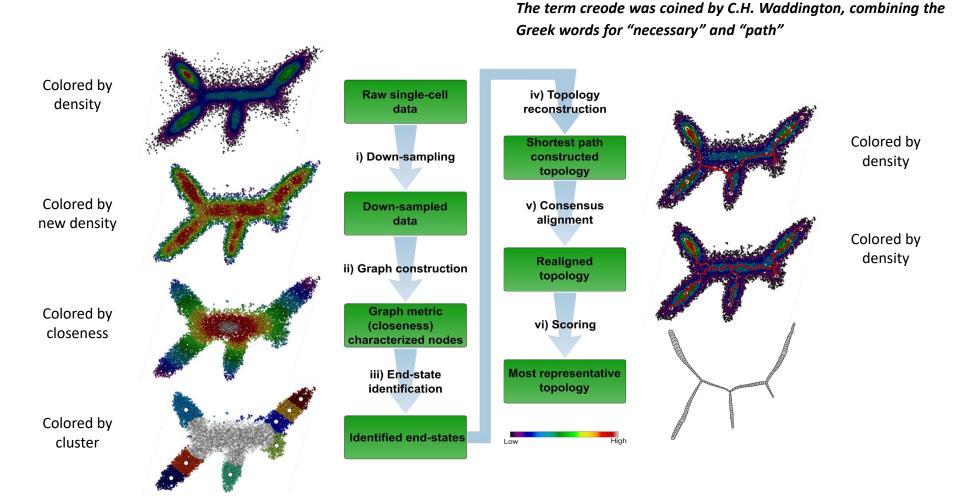
Unsupervised Trajectory Analysis of Single-Cell RNA-Seq and Imaging Data Reveals Alternative Tuft Cell Origins in the Gut

Charles A. Herring, 1,2 Amrita Banerjee, 1,3 Eliot T. McKinley, 1,4 Alan J. Simmons, 1,3 Jie Ping, 5,5 Joseph T. Roland, 1 Jeffrey L. Franklin, 1,3 Qi Liu, 5,5 Michael J. Gerdes, 7 Robert J. Coffey, 1,3,4,8 and Ken S. Lau^{1,2,3,5,9,*}

Single-cell data continnum represented as transitional trajectories



putative-Creode Overview



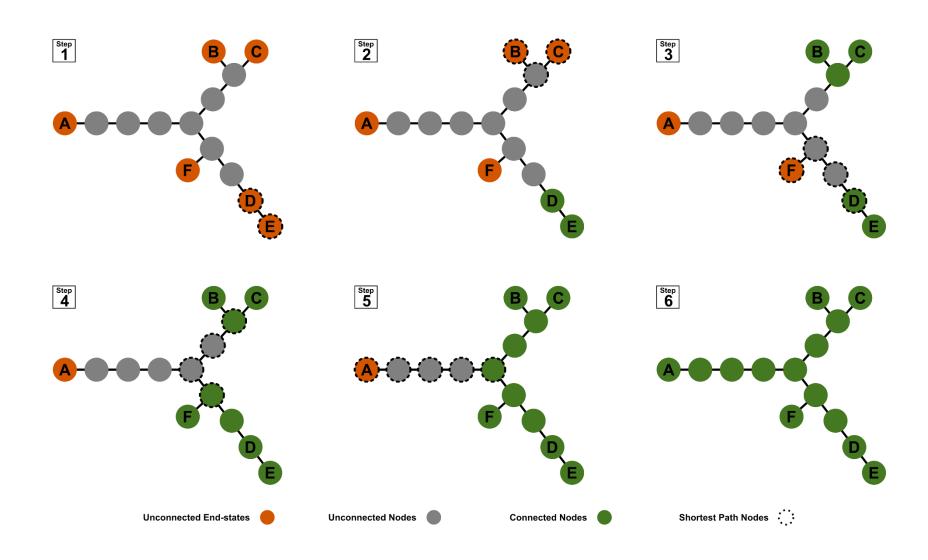
p-Creode characteristics

- Map complex multi-branching topologies
- Results reproducible when run on well-regulated processes
- Ensemble approach for deconvolving dysregulated transition processes

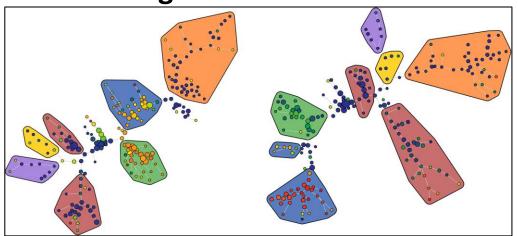
Three major novel developments of p-Creode

- Unsupervised end state identification
- Hierarchical placement algorithm (coupled to consensus alignment)
- Unique scoring method that leverages an ensemble approach to statistically assess the quality of a result

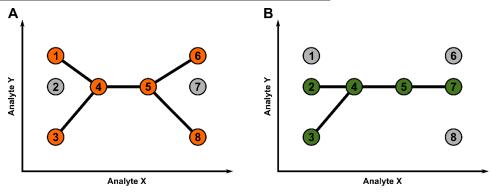
p-Creode's hierarchical placement strategy



p-Creode scoring method



All graphs are different, some more than other



Breaking New Ground in the Landscape of Single-Cell Analysis

Kenji Kamimoto^{1,2,3} and Samantha A. Morris^{1,2,3,*}

¹Department of Developmental Biology

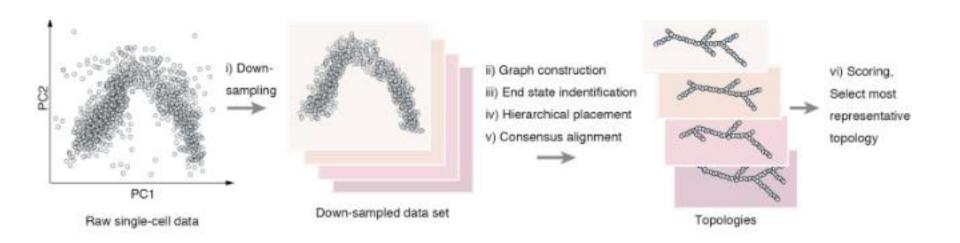
²Department of Genetics

³Center of Regenerative Medicine

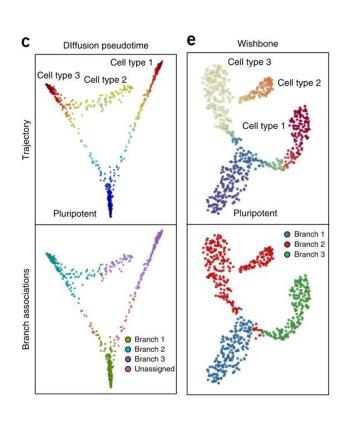
Washington University School of Medicine in St. Louis. 660 S. Euclid Avenue, Campus Box 8103, St. Louis, MO 63110, USA

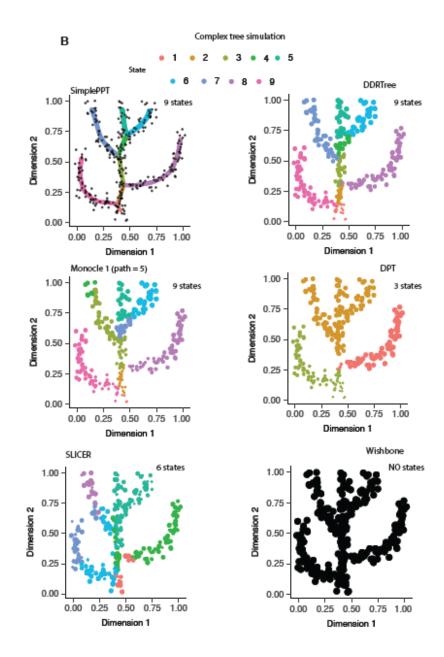
*Correspondence: s.morris@wustl.edu

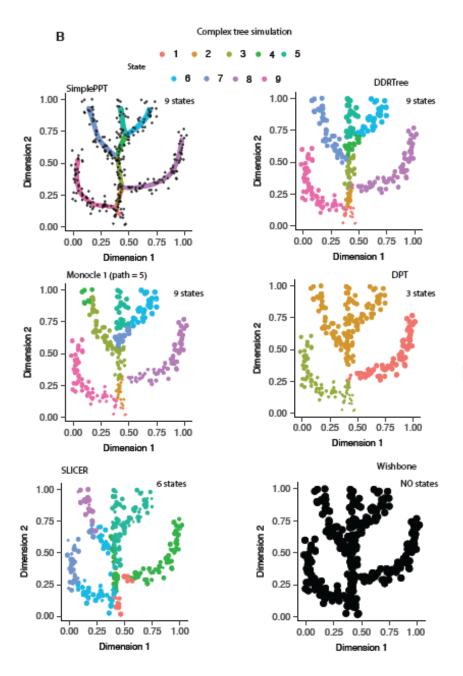
https://doi.org/10.1016/j.cels.2017.12.015



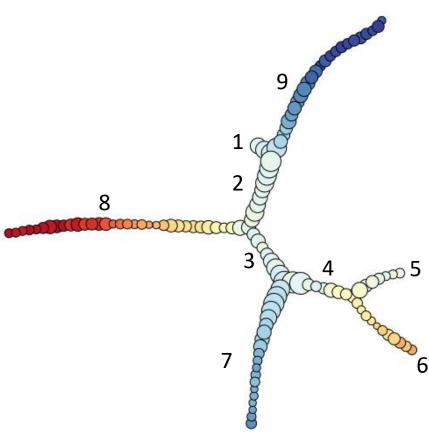
Some examples of algorithm comparisons



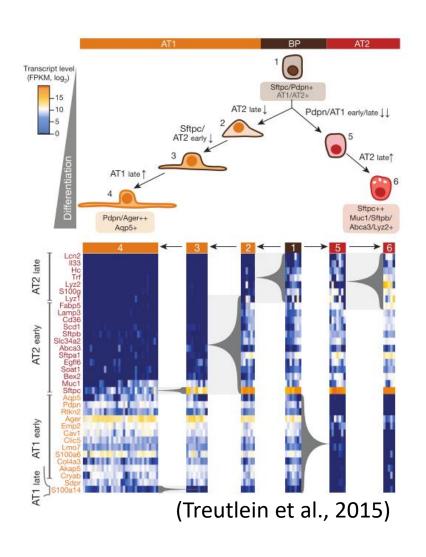


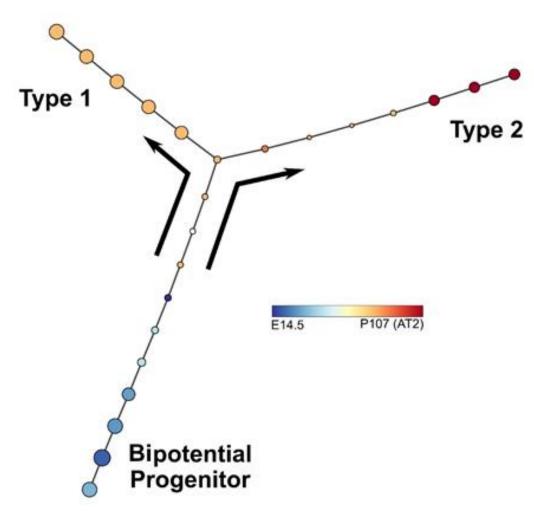


p-Creode application on other people's synthetic data



Developmental time of alveolar differentiation reconstructed by p-Creode applied to Fluidigm C1 sc-RNA-seq data



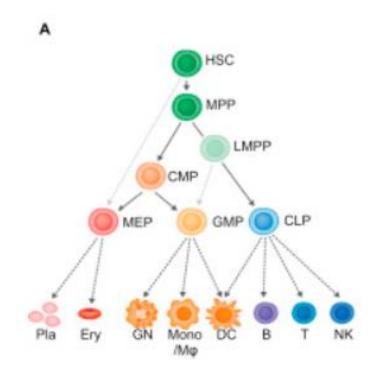


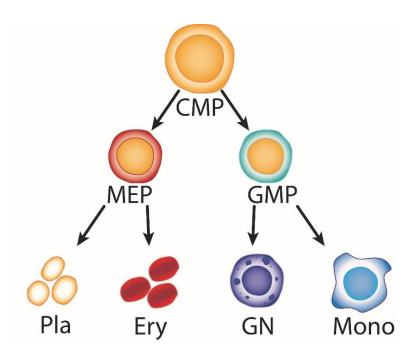
p-Creode applied to multi-branching myeloid differentiation

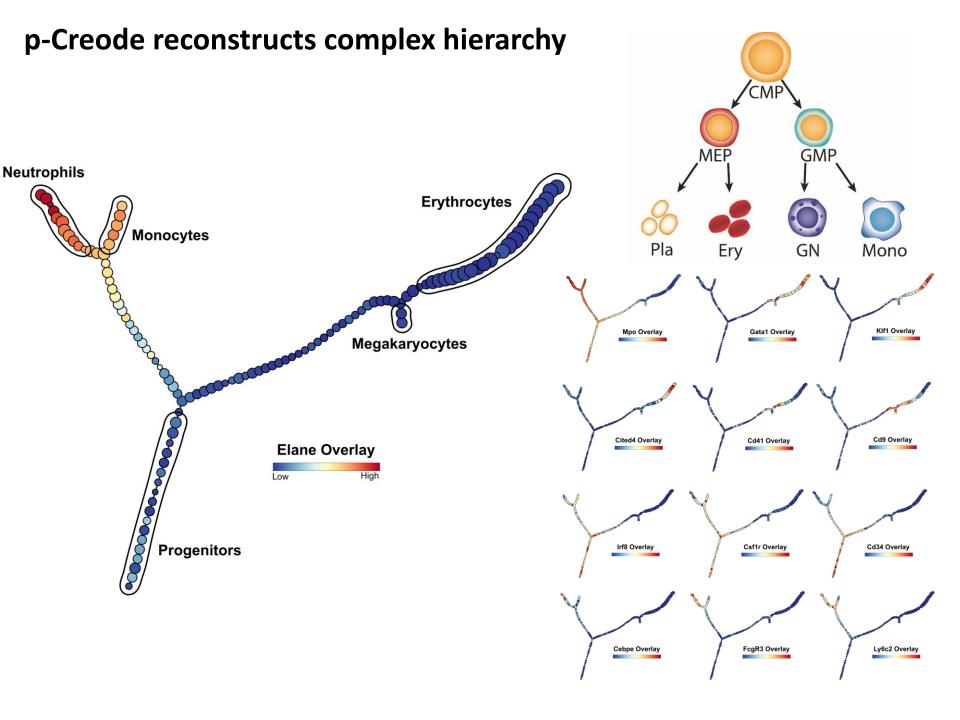
Transcriptional Heterogeneity and Lineage Commitment in Myeloid Progenitors

Franziska Paul,^{1,9} Ya'ara Arkin,^{2,9} Amir Giladi,^{1,9} Diego Adhemar Jaitin,¹ Ephraim Kenigsberg,² Hadas Keren-Shaul,¹ Deborah Winter,¹ David Lara-Astiaso,¹ Meital Gury,¹ Assaf Weiner,¹ Eyal David,¹ Nadav Cohen,² Felicia Kathrine Bratt Lauridsen,^{3,4,5} Simon Haas,⁶ Andreas Schlitzer,^{7,8} Alexander Mildner,¹ Florent Ginhoux,⁷ Steffen Jung,¹ Andreas Trumpp,⁶ Bo Torben Porse,^{3,4,5} Amos Tanay,^{2,10,*} and Ido Amit^{1,10,*}

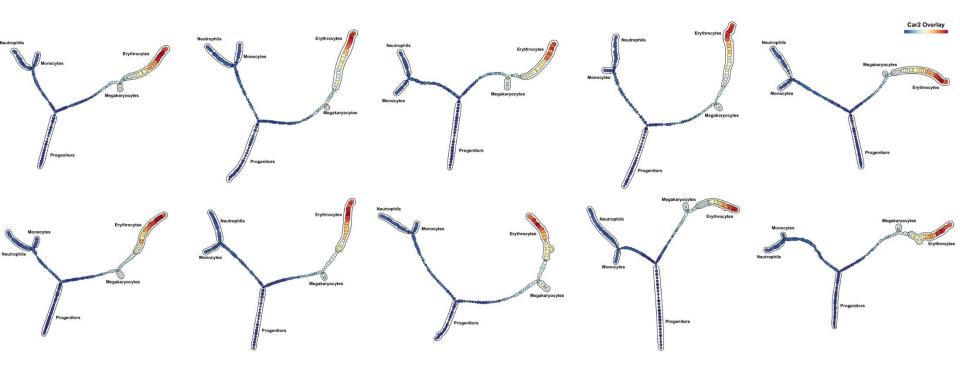
Plate-based MARS-seq - >4000 cells



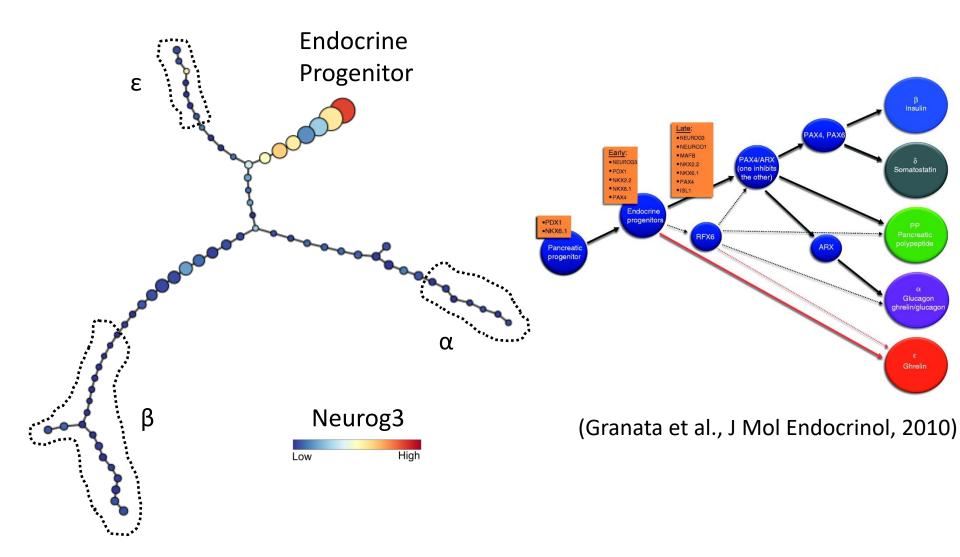




Robustness of p-Creode demonstrated on complex trajectories

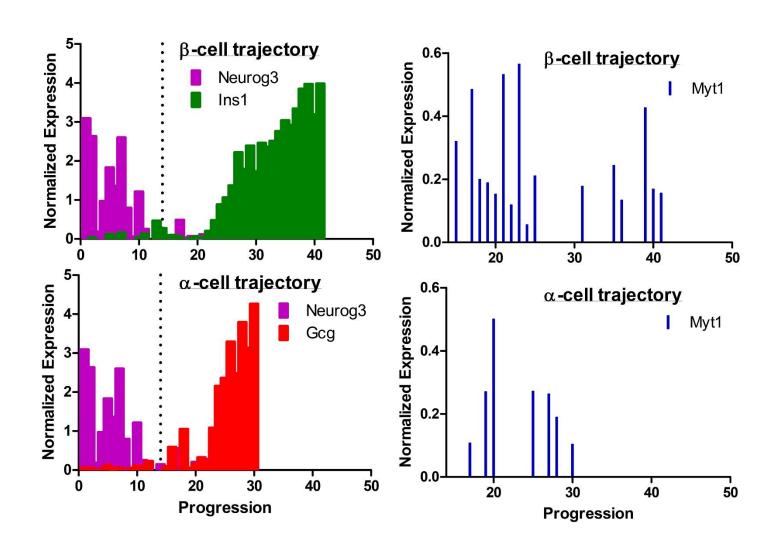


p-Creode generates epsilon, beta, alpha cell trajectories for E14.5 pancreas

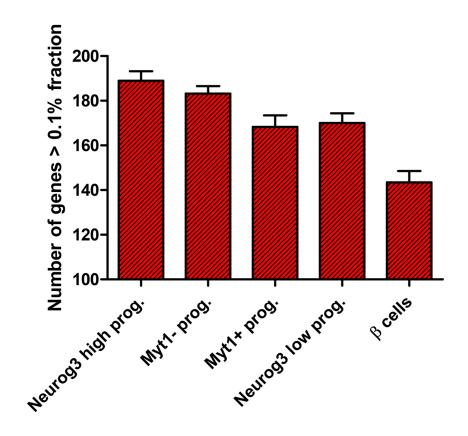


(Liu et al., Developmental Cell, in revision)

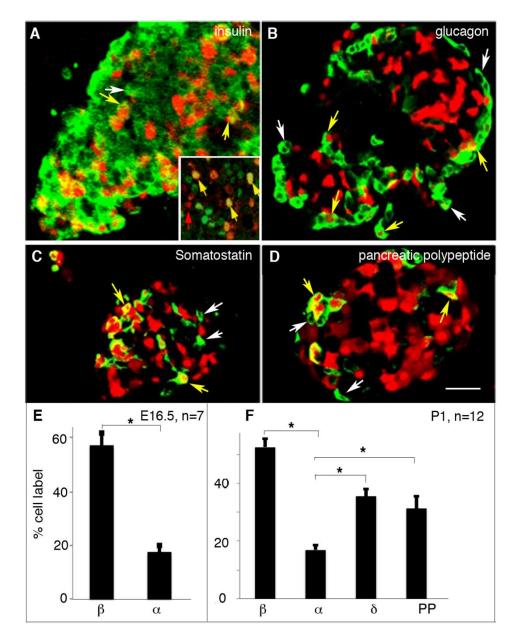
Trajectory analysis demonstrates the beta cell lineage bias of Myt1 expression



Trajectory analysis demonstrates the beta cell lineage bias of Myt1 expression



Ngn3/Myt1 lineage tracing is biased towards beta cells compared to alpha cells



Intestinal tuft cell biology

- ~1.0% epithelial cells in mouse intestine
- Express chemoreceptors (TRPM5,etc.)
- Characteristic "tufted" morphology due to apical acetylated tubulin bundle

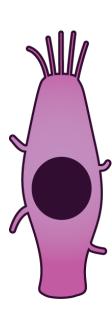
Tuft cells, taste-chemosensory cells, orchestrate parasite type 2 immunity in the gut

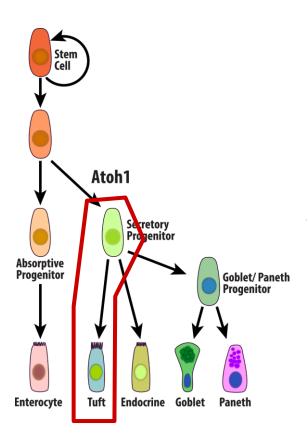
Michael R. Howitt, ¹ Sydney Lavoie, ¹ Monia Michaud, ¹ Arthur M. Blum, ² Sara V. Tran, ³ Joel V. Weinstock, ² Carey Ann Gallini, ¹ Kevin Redding, ³ Robert F. Margolskee, ³ Lisa C. Osborne, ⁴* David Artis, ⁴ Wendy S. Garrett^{1,5,6}†

Intestinal epithelial tuft cells initiate type 2 mucosal immunity to helminth parasites

Tuft-cell-derived IL-25 regulates an intestinal ILC2-epithelial response circuit

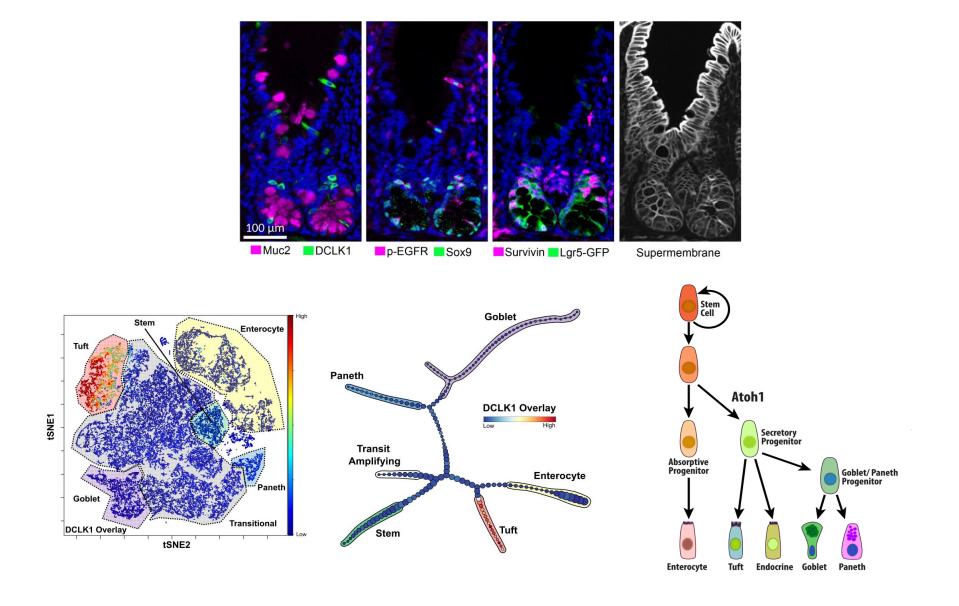
Jakob von Moltke¹, Ming Ji^{1,2}, Hong-Erh Liang¹ & Richard M. Locksley^{1,2,3}



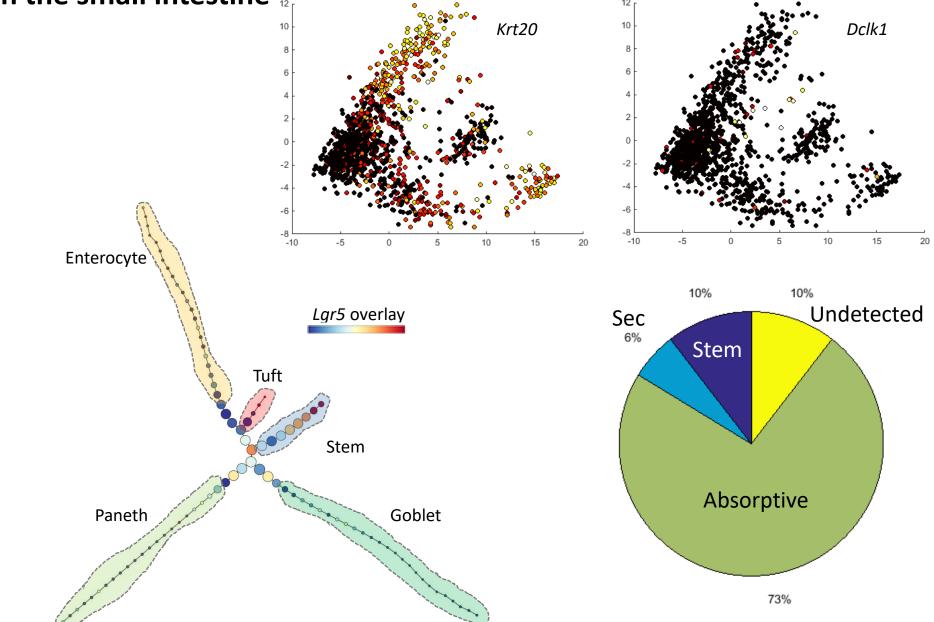


(Gerbe et al., 2011)

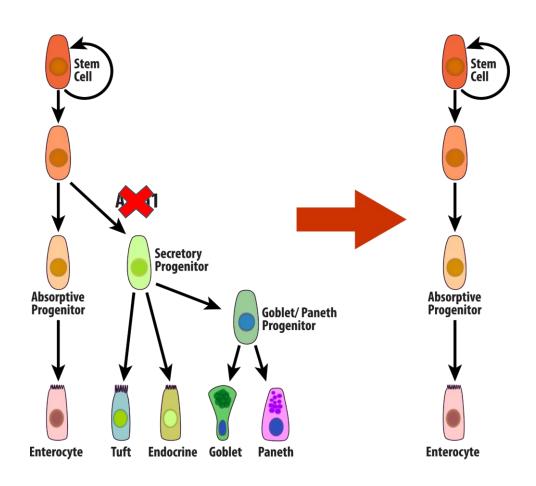
p-Creode suggests tuft cells are specified outside of secretory lineage in the small intestine



p-Creode suggests tuft cells are specified outside of secretory lineage in the small intestine 12 Krt20 Krt20 Dc/k1

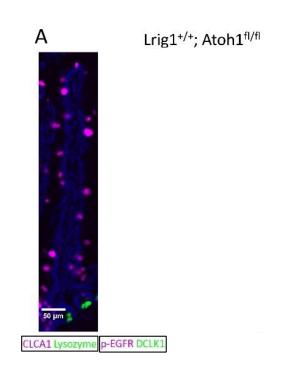


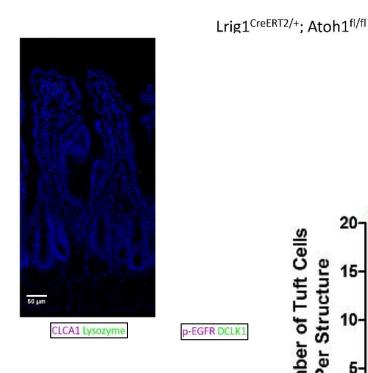
The Atoh1 transcription factor is an early determinant of secretory fate

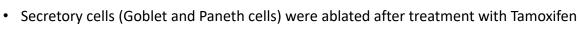


Epithelial specific Atoh1 KO suppresses secretory cells but increases tuft cells in the small intestine

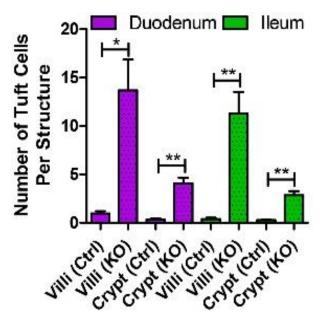
Small Intestine







- Tuft cells (DCLK1) increased in number in the small intestine
- Markers and Morphology point towards this not being a separate cell type



Epithelial specific Atoh1 KO suppresses secretory cells but increases tuft cells in the small intestine

Small Intestine

