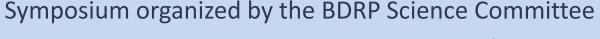
# Single-Cell Revolution: Embryogenesis at High-Resolution



July 2, 2020

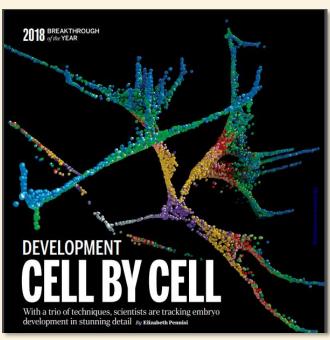
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(virtual)



# Profiling development at the single-cell level





#### Three technical pillars:

- scRNAseq analysis (cell-state landscape)



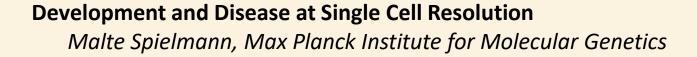
- light sheet microscopy (cell-imaging in situ)
- CRISPR/Cas9 (functional analysis)

https://science.sciencemag.org/content/sci/362/6421/1344.full.pdf



# Symposium Overview What does scRNAseq bring to 21<sup>st</sup> Century embryology and birth defects research?

Teratogenesis at the Single-Cell Level: Opportunities and Challenges Thomas Knudsen, USEPA/ORD-CCTE



What scRNAseq Is Now Telling Us about Embryology
Sean Megason, Harvard Medical School

How Single-Cell Profiling Data Can Be Applied to Improve Children's Health Elaine Faustman, University of Washington



#### Single-Cell Revolution: Embryogenesis at High-Resolution

# TERATOGENESIS AT THE SINGLE-CELL LEVEL: OPPORTUNITIES AND CHALLENGES

Thomas B. Knudsen, PhD

Developmental Systems Biologist

US EPA, Center for Computational Toxicology and Exposure

Research Triangle Park, NC 27711 <u>knudsen.thomas@epa.gov</u> ORCHID 0000-0002-5036-596x



#### **Disclosure Statement**

#### **DISCLAIMER:**

The views expressed in this presentation are my own and do not reflect Agency policy.

#### **DISCLOSURE:**

Editor-in-Chief of 'Current Research in Toxicology' (Elsevier).

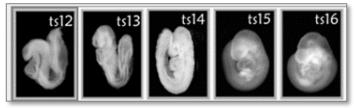
#### **CONFLICTS of INTEREST:**

None.

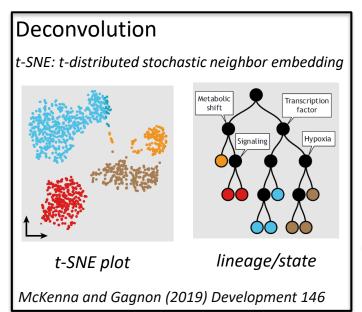
# Single-cell RNA-seq (scRNAseq)

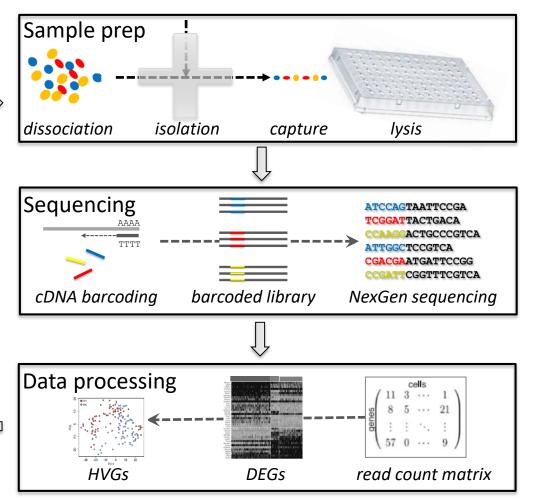
- The fundamental unit of form and function in embryology is 'the cell', each with its own unique history (lineage) and environment (microphysiology).
- NexGen sequencing (RNA-seq) averages the transcriptome across a composite system but cannot decode individual lineages or state dynamics.
- With methods to separate and tag individual cells, scRNAseq enables annotation of tens of thousands of cells in parallel from a composite sample.
- Computational methods are then used to unravel cellular complexity and reconstruct gene expression dynamics for quantitative lineage tracing.

# scRNAseq Workflow



https://www.emouseatlas.org/emap/ema/home.php





# **Challenges**

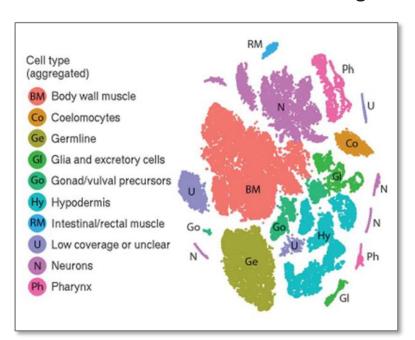
- <u>Cell separation tools</u>: how well does the scRNA-ome survive preparation protocols required to generate isolated cells from a composite system?
- <u>Bioinformatics</u>: how can individual cell provenance be retained and annotations decoded while deep-sequencing thousands of cells in parallel?
- <u>Analysis</u>: how to distinguish highly-variable genes (HVGs) from technical variation (eg, PCR artifacts and over-sequencing)?
- <u>Computational</u>: how to <del>validate</del> transitional cell types, map them to developmental trajectories, and spatially reconstruct a composite system?

#### **Opportunites**

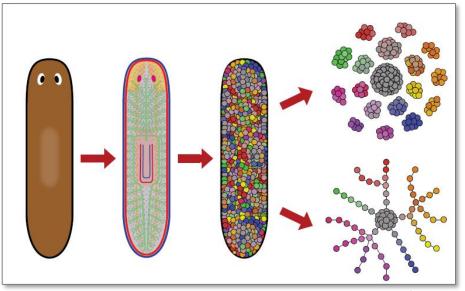
- Comprehensive 'cell atlases' for modeling gene expression dynamics as cells acquire specified fates during morphogenesis and differentiation; examples:
  - simple model organisms (planaria, C elegans, Drosophila)
  - experimental embryology (ascidian, zebrafish, frog, chick)
  - mammalian embryos (mouse, human)
  - differentiating embryonic stem cell lines (mouse, human)
  - and more …
- Despite many studies to date that have used genomic approaches to characterize dysmorphogenesis, very few examples using scRNAseq.

#### **Roundworm & Flatworm**

t-SNE plot decomposing *C elegans* into 29 cell clusters at L2 larval stage.



Planaria (regenerative) decomposed to multiple cell states → reconstitution of lineages by divergence of transcriptomes.

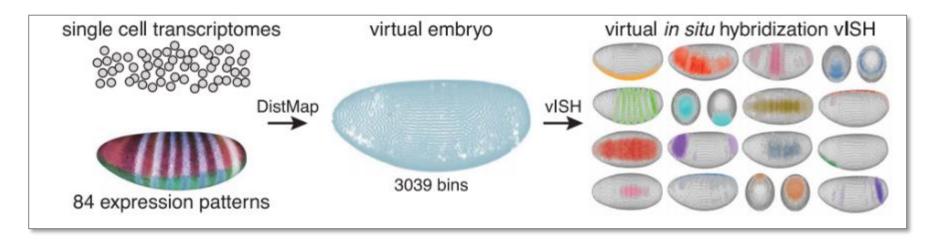


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Plass et al. (2018) Science 360

# Drosophila

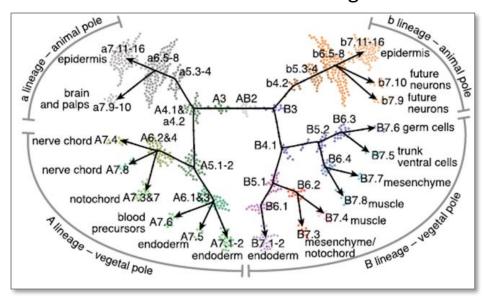
- scRNAseq profiles decomposed into 84 cell clusters that could be annotated by wellcorrelated marker genes (coverage = 87% of cells, >8k genes per cell);
- spatial reconstruction mapped expression patterns in a 'virtual embryo' that predicts developmental trajectories linked to key transcription factors and signal gradients.



#### **Ascidian**

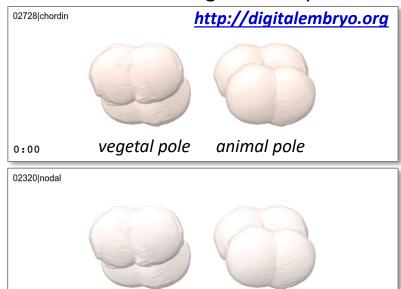
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Lineage tree from 1,042 cells (>8K genes per cell) → gene expression history (scRNAseq) and physical position (DLSM) of every cell from the 4- to 64-cell stage.



Sladitschek et al. (2020) Cell 181

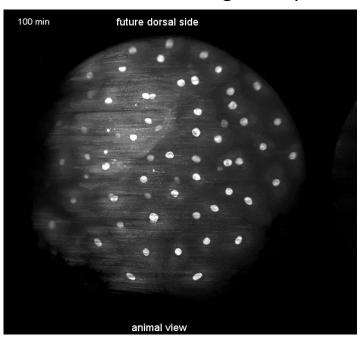
vISH: reconstruction of expression dynamics up to gastrulation; two genes (chordin, nodal) selected from the online digital library.



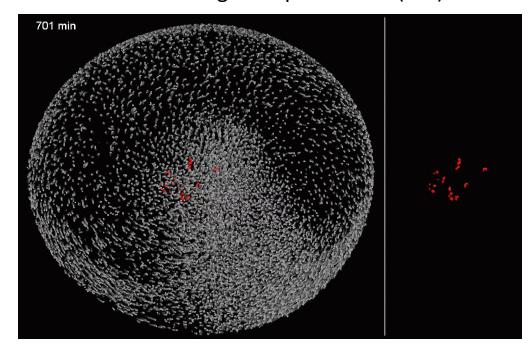
12

#### Zebrafish - at single-cell resolution

Digital LSM reconstruction at 90 sec intervals through 18 hpf.



Reverse engineering the physical position of cells forming the optic vesicle (red).



Keller et al. (2008) Science 322

#### Zebrafish

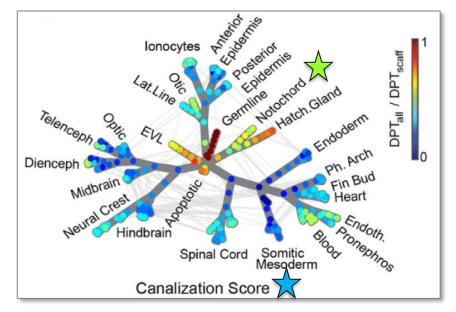
Lineage tree: 25 cell types built on 38,731 cells captured 3.3 hpf (pluripotent blastula) to 12 hpf (6-somite pharyngula).

Primordial

Farrell et al. (2018) Science 360

-rom Science cover 2018

Canalization tree built on 92,000 cells 4 hpf through organogenesis (24 hpf); note buffering notochord > somitic lineages.



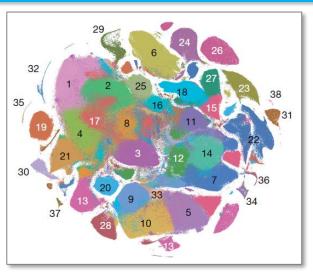
Wagner et al. (2018) Science 360

#### Mouse - organogenesis

1-Connective tissue progenitors 2-Chondrocytes and osteoblasts 3-Intermediate mesoderm 4-Jaw and tooth progenitors 

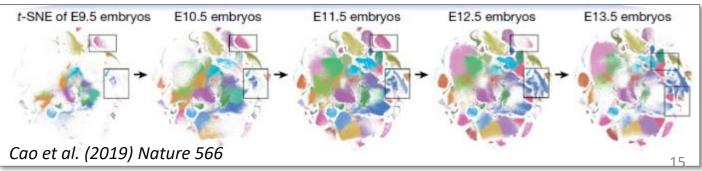
5-Excitatory neurons 6-Epithelial cells 7-Radial glia 8-Early mesenchyme 9-Neural progenitor cells 

10-Postmitotic premature neurons 11-Oligodendrocyte progenitors 12-Isthmic organizer cells 13-Myocytes 14-Dorsal neural tube cells 15-Inhibitory neurons 16-Stromal cells 17-Osteoblasts 18-Inhibitory neuron progenitors 19-Premature oligodendrocytes 20-Endothelial cells 21-Chondroctye progenitors 22-Definitive erythrocyte lineage 23-Schwann cell precursors 24-Sensory neurons 25-Limb mesenchyme 26-Primitive erythroid lineage 27-Inhibitory interneurons 28-Granule neurons 29-Hepatocytes 30-Notochord and floor plate cells 31-White blood cells 32-Ependymal cells 33-Cholinergic neurons 34-Cardiac muscle lineage 35-Megakaryocytes 36-Melanocytes . 38-Neutrophils (



t-SNE annotated 38 cell clusters during organogenesis, the period when most organ systems form:

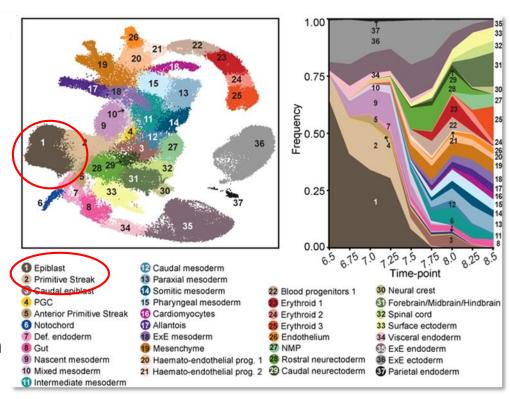
- GD 9.5 (151K cells)
- GD 10.5 (370K cells)
- GD 11.5 (603K cells)
- GD 12.5 (468K cells)
- GD 13.5 (435K cells)



#### Mouse - gastrulation



- The vast majority of cell lineages derive from the epiblast during gastrulation;
- t-SNE lineage staged from 116,312 cells epiblast (GD 6.5) to headfold (GD 8.5);
- body's fundamental genomic blueprint is also decoded during this period;
- while autopoiesis rules the epiblast each cell's destiny determined by its position.



Pijuan-Sala et al. (2019) Nature 566

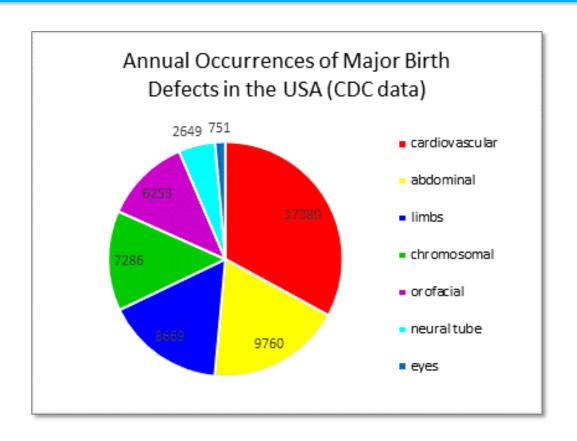
In toto live cell imaging atlas reveals individual cellular dynamics by position, fate, movement, and division (GD 6.5 – GD 8.5).

Mouse - at single-cell resolution

Joint visualization of dynamic cell fate reconstruction and cell divisions across a developing embryo

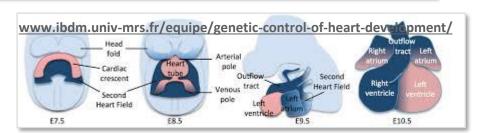
McDole et al. (2018) Cell 175

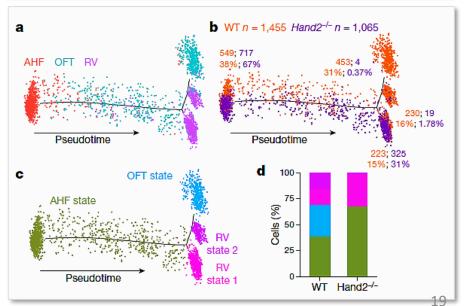
#### Birth Defects - 1 in 33 babies in the USA



#### **Congenital Heart Defects**

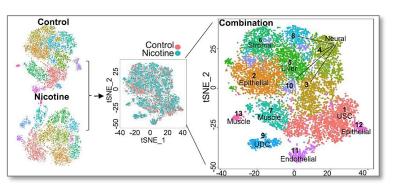
- scRNAseq on 36,000 cells from the cardiogenic region (GD 7.75 to 9.25);
- analysis pointed to Hand2 specification of myocardial cells in the OFT;
- however, Hand2-null mice display RV hypoplasia in addition to OFT defects;
- scRNAseq of Hand2(-/-) rudiments pointed to retinoic acid (ATRA) signaling;
- altered ATRA distribution 'posteriorized' RV to an atrial-like morphology.



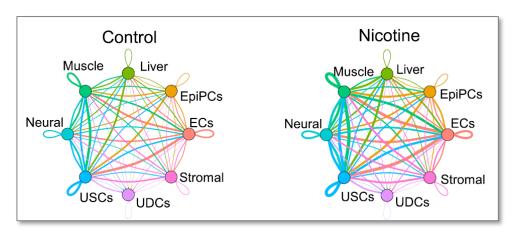


de Soysa et al. (2019) Nature 572

#### Nicotine and hESC differentiation



- t-SNE plots from hESC-derived embryoid bodies on day 21 of culture;
- control condition (n= 6,847 individual cells) and 10  $\mu$ M nicotine (n= 5,646 cells) exposure.



- Nicotine increased specific cell-tocell communication pathways;
- network connects ligands from to receptors for target cell populations.

#### Pluripotency – embryonic stem cells and the epiblast

#### hESCs (0-8 days in culture)

tSNE map of 4822 barcoded progenitors at naïve (n=1491), primed (n=695), and embryoid body (EB, n=2636) stages.

#### **Epiblast (GD 5.5 – GD 6.5)**

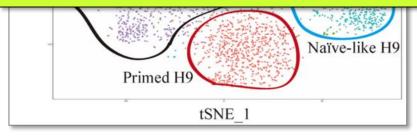
t-SNE map of 1,724 cells from 28 mouse embryos as they acquire the propensity for 'Primitive Streak' formation.

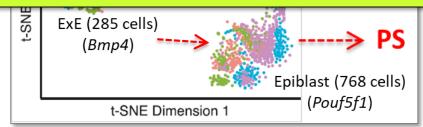


VE (671 cells) (*Amn*)

"It is not birth, marriage, or death, but **gastrulation** which is truly the most important time in your life."

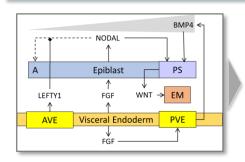
- Lewis Wolpert

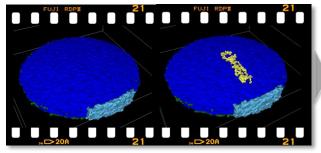




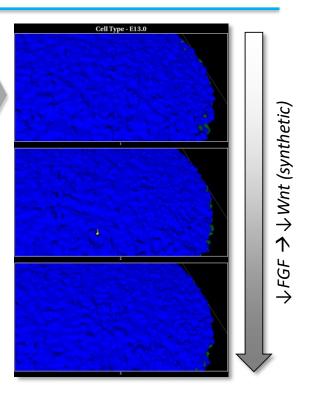
Cheng et al. (2019) Cell Reports 26

# Quasi-gastrulation – epiblast in silico





- When and where an epiblast cell migrates through the PS ultimately determines it's regional destiny;
- morphological programming logic of the epiblast can be tapped to virtually reconstruct gastrulation in silico;
- simulation manifold for developmental computation with data from chemical effects on pluripotent hESCs.



# Single-cell profiling for DevTox

- Developmental toxicology has yet to experience scRNAseq profiling as a nextgeneration blueprint for chemical testing.
- 'Cell atlases' (e.g., HCA, MCA) for modeling gene expression dynamics as cells acquire specified fates during morphogenesis and differentiation.
- Precision toxicology due to unprecedented resolution to examine biological systems and their perturbation by chemicals.
- Computational approaches needed for data integration/clustering, HVGs for state landscape, models for cell-swarming and positional information, ...).

#### **References for Images**

- Slide 07 McKenna and Gagnon (2019) Recording development with single cell dynamic lineage tracing. Development 146, dev169730. doi:10.1242/dev.169730 (10 pages).
- Slide 10 Cao et al. (2017) Comprehensive single-cell transcriptional profiling of a multicellular organism. Science 357: 661–667.
- Slide 10 Plass et al. (2018) Cell type atlas and lineage tree of a whole complex animal by single-cell transcriptomics. Science 360, eaaq1723 (10 pages).
- Slide 11 Karaiskos et al. (2017) The Drosophila embryo at single-cell transcriptome resolution. Science 358: 194-199.
- Slide 12 Sladitschek et al. (2020) MorphoSeq: full single-cell transcriptome dynamics up to gastrulation in a Chordate. Cell 181: 922–935.
- Slide 13 Keller et al. (2008) Reconstruction of zebrafish early embryonic development by scanned light sheet microscopy. Science 322: 1065-1069.
- Slide 14 Farrell et al. (2018) Single-cell reconstruction of developmental trajectories during zebrafish embryogenesis. Science 360: doi:10.1126/science.aar3131.
- Slide 14 Wagner et al. (2018) Single-cell mapping of gene expression landscapes and lineage in the zebrafish embryo. Science 360: 981-987.
- Slide 15 Cao et al. (2019) The single-cell transcriptional landscape of mammalian organogenesis. Nature 566: doi.org/10.1038/s41586-019-0969-x.
- Slide 16 Pijuan-Sala et al. (2019) A single-cell molecular map of mouse gastrulation and early organogenesis. Nature 566: doi.org/10.1038/s41586-019-0933-9.
- Slide 17 McDole et al. (2018) *In toto* imaging and reconstruction of postimplantation mouse development at the single cell level. Cell 175: 859–876.
- Slide 19 de Soysa et al. (2019) Single-cell analysis of cardiogenesis reveals basis for organ-level developmental defects. Nature 572: 120-124.
- Slide 20 Guo et al. (2019) Single-cell RNA sequencing of human embryonic stem cell differentiation delineates adverse effects of nicotine on embryonic development. Stem Cell Reports 12: 772–786.
- Slide 21 Han et al. (2018) Mapping human pluripotent stem cell differentiation pathways using high throughput single-cell RNA-sequencing. Genome Biol. 19: 47. doi.org/10.1186/s13059-018-1426-0.
- Slide 21 Cheng et al. (2019) Single-cell RNA-seq reveals cellular heterogeneity of pluripotency transition and X chromosome dynamics during early mouse development. Cell Reports 26:2593–2607.
- Slide 22 Knudsen et al. (2020) Developmental computation with embryonic stem cells (SOT 2020 ePoster #2043).