

CE Image Analysis

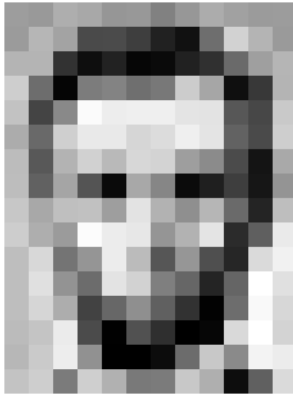
An Overview

Anthony Santella, Zhirong Bao

Conceptualizing Image Analysis

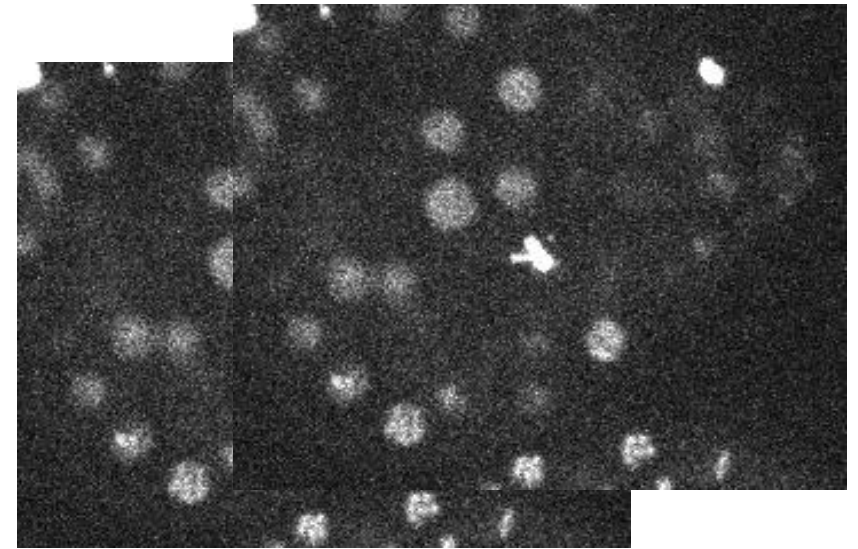
- Images
- Mapping from scientific problems
 - Validation
 - Computational tools
- Some Major Classes of Task
 - Object Segmentation and Measurement
 - Alignment
 - Tracking
- AI and Image Analysis
 - Introduction to deep learning
 - Architectures, Uses, Limitations
 - Advanced Examples Spatial Patterns

Images

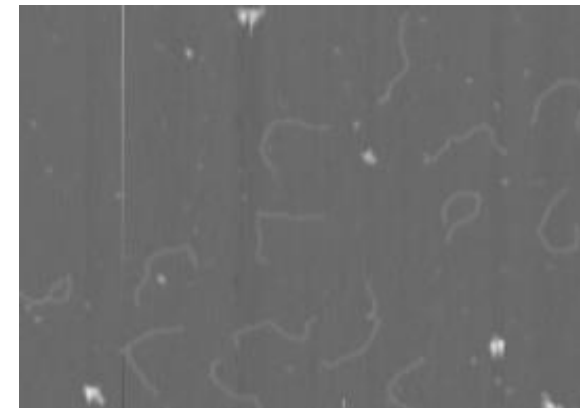
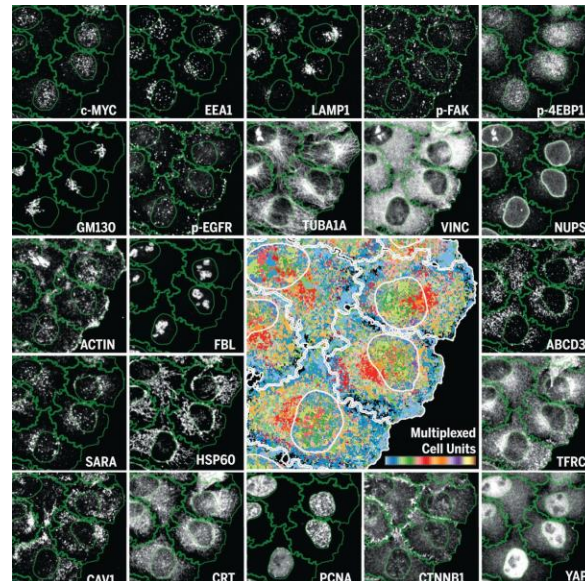
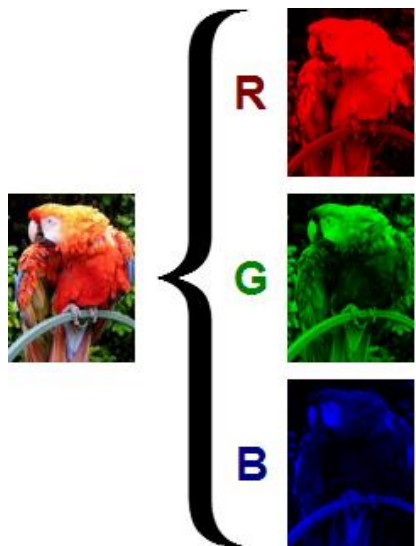


157	153	174	168	150	152	129	151	172	161	155	156
155	182	163	74	75	62	33	17	110	210	180	154
180	180	50	14	34	6	10	33	48	106	159	181
206	109	5	124	131	111	120	204	166	15	56	180
194	68	137	251	237	239	228	227	87	71	201	
172	105	207	233	233	214	220	239	228	98	74	206
188	88	179	209	185	215	211	158	139	75	20	169
189	97	165	84	10	168	134	11	31	62	22	148
199	168	191	193	168	227	178	143	182	106	36	190
205	174	155	252	236	231	149	178	228	43	95	234
190	216	116	149	236	187	86	150	79	38	218	241
190	224	147	108	227	210	127	102	36	101	255	224
190	214	173	66	103	143	96	50	2	109	249	215
187	196	235	75	1	81	47	0	6	217	255	211
183	202	237	145	0	0	12	108	200	138	243	236
195	206	123	207	177	121	123	200	175	13	96	218

157	153	174	168	150	152	129	151	172	161	155	156
155	182	163	74	75	62	33	17	110	210	180	154
180	180	50	14	34	6	10	33	48	106	159	181
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194	68	137	251	237	239	228	227	87	71	201	
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199	168	191	193	168	227	178	143	182	106	36	190
205	174	155	252	236	231	149	178	228	43	95	234
190	216	116	149	236	187	86	150	79	38	218	241
190	224	147	108	227	210	127	102	36	101	255	224
190	214	173	66	103	143	96	50	2	109	249	215
187	196	235	75	1	81	47	0	6	217	255	211
183	202	237	145	0	0	12	108	200	138	243	236
195	206	123	207	177	121	123	200	175	13	96	218



.25um x.25um x 1um pixel



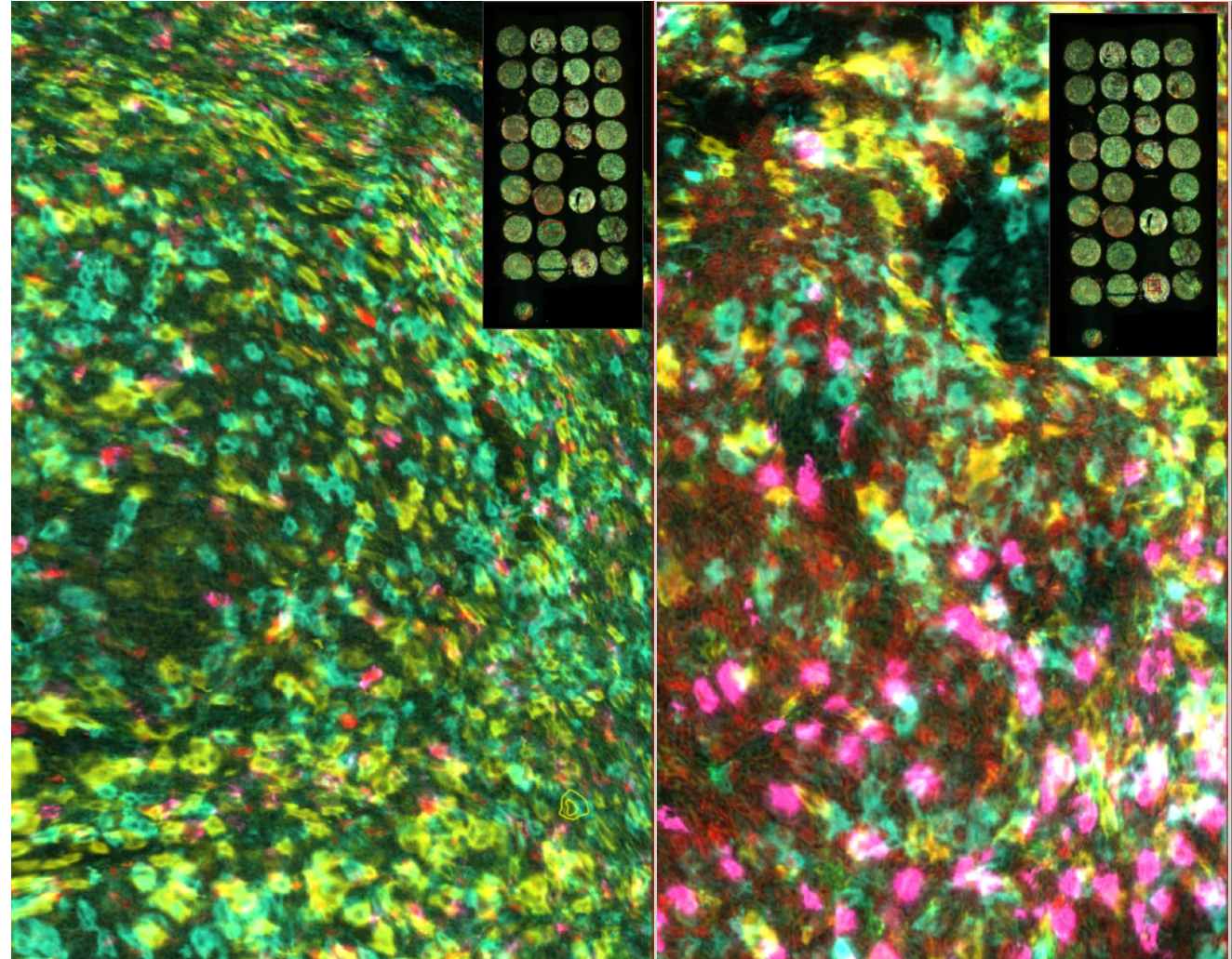
Why might pixels not be square like these?

What is there? Objects, properties, relationships

- Mapping from a scientific problems
- Validation
- Computational tools

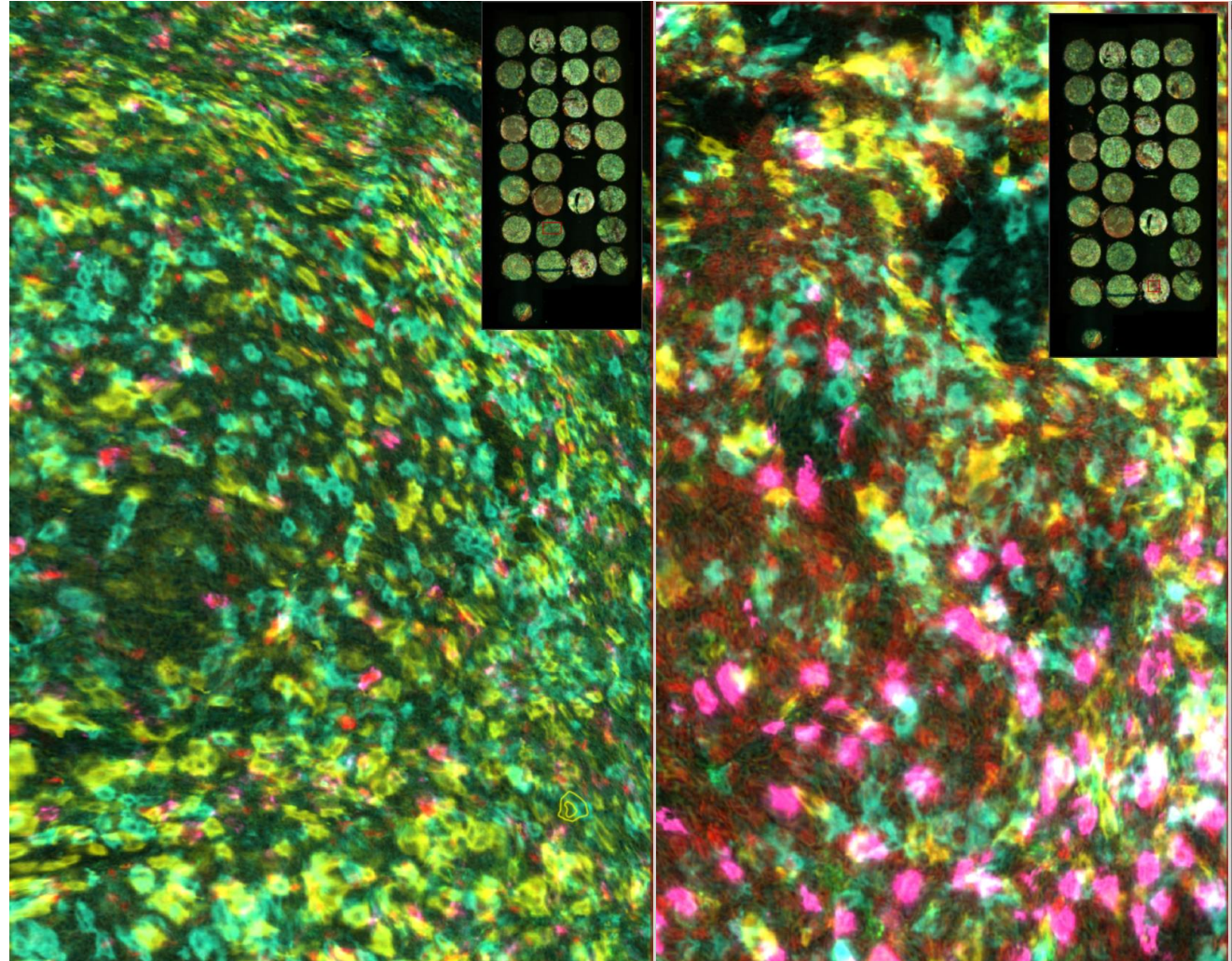
Mapping from a Scientific Problem

- What is the question?
 - Often (not always) prove condition x is different than condition y
- **What do I see?**
- What can I measure?
- What are some differences between these two cores?



Validation

- Spot checking vs formal evaluation
- **Chance of artificial correlation of measurement with experimental condition?**
- **How might this happen here?**



Computational Tools

- How easy is the workflow to
 - Create
 - Use
 - Replicate
- Layering on Flexible Products
 - Free: Cellprofiler, Ilastik, QuPath...
 - commercial: Imaris...
 - Libraries: napari, MATLAB...

Programs versus Software Products

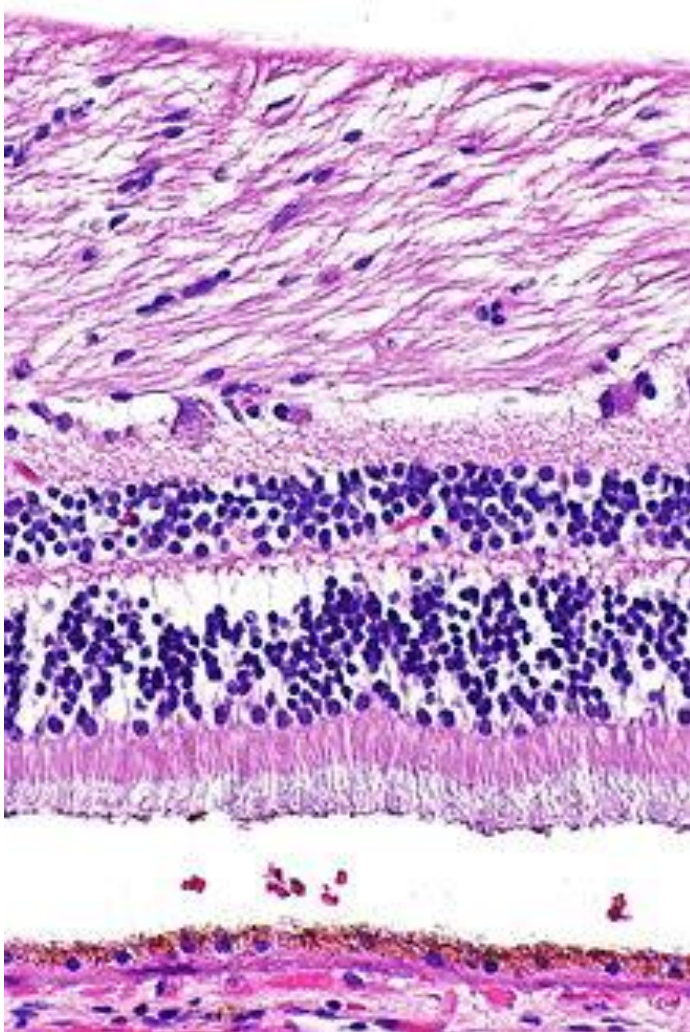
• Usually small in size	• Large
• Author himself is sole user	• Large number of users
• Single developer	• Team of developers
• Lacks proper user interface	• Well-designed interface
• Lacks proper documentation	• Well documented & user-manual prepared
• Ad hoc development.	• Systematic development

Software engineering 101...

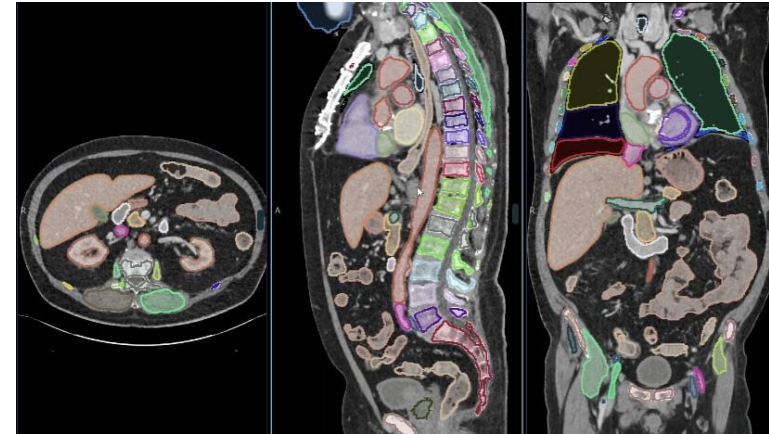
Major Classes of Task and Examples

- Object segmentation and measurement
- Alignment
- Tracking

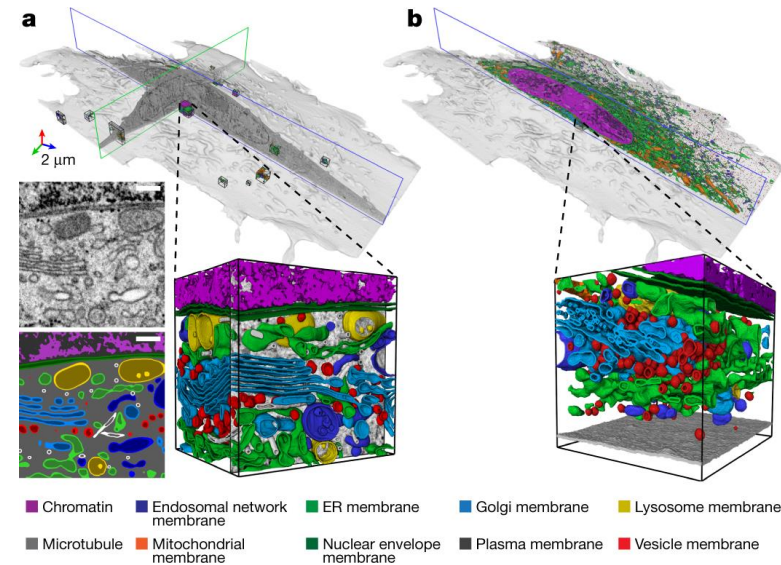
Object Segmentation and Measurement



wikipedia

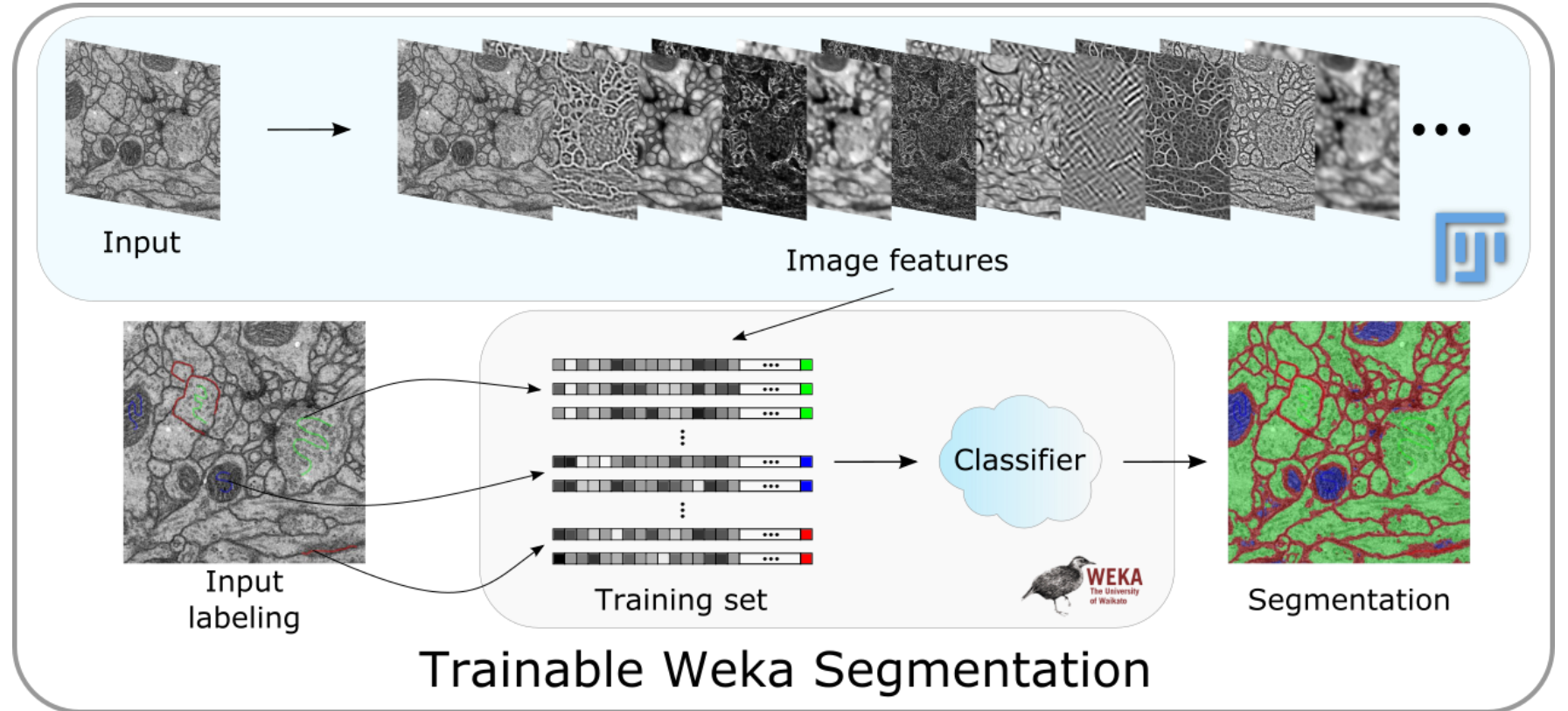
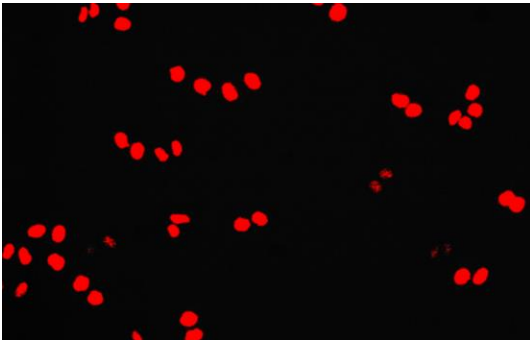
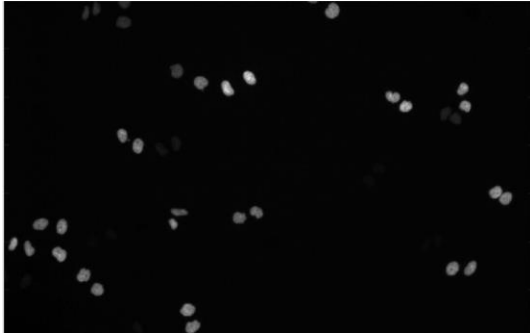


NVIDIA



Heinrich et al Nature, 2021

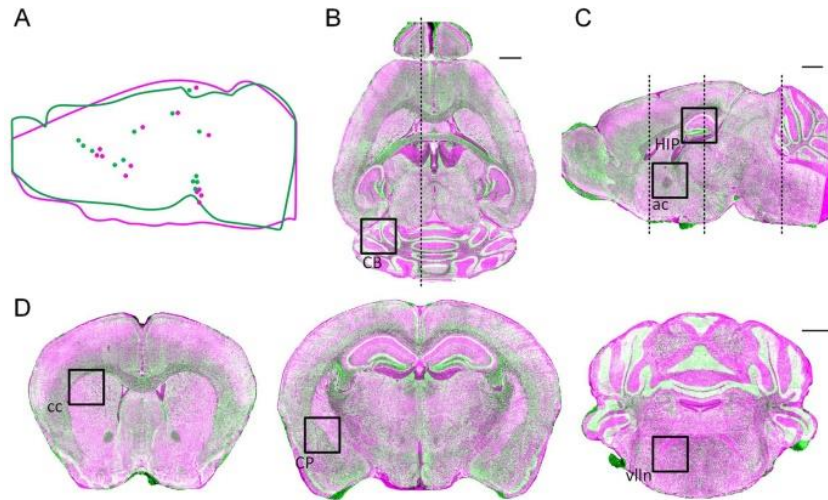
Object Segmentation



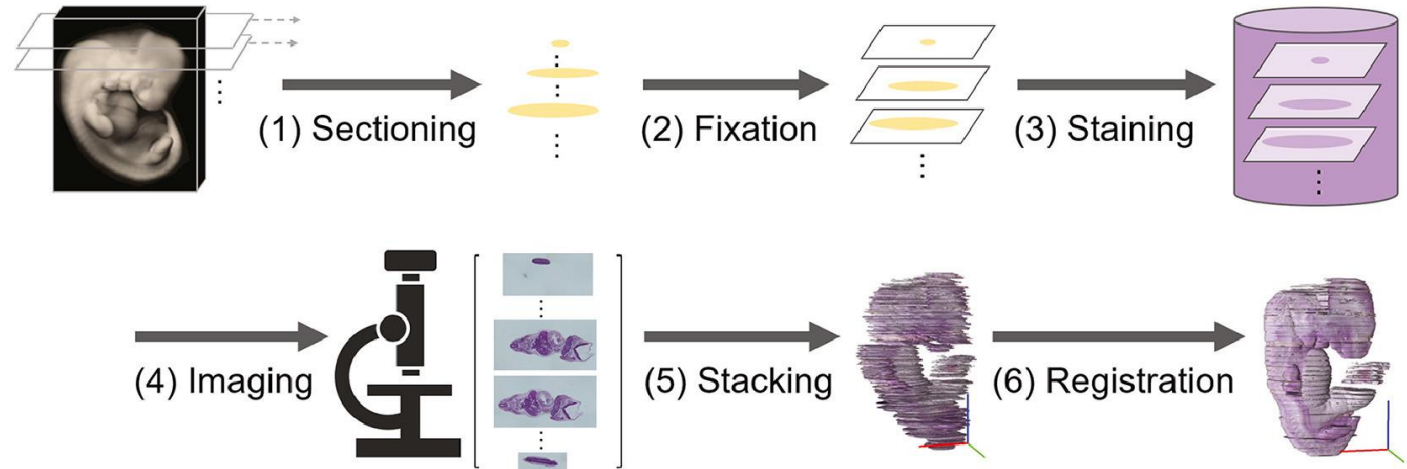
imagej.net/plugins/tws/

What are some likely advantages and disadvantages of these compared to deep learning approaches?

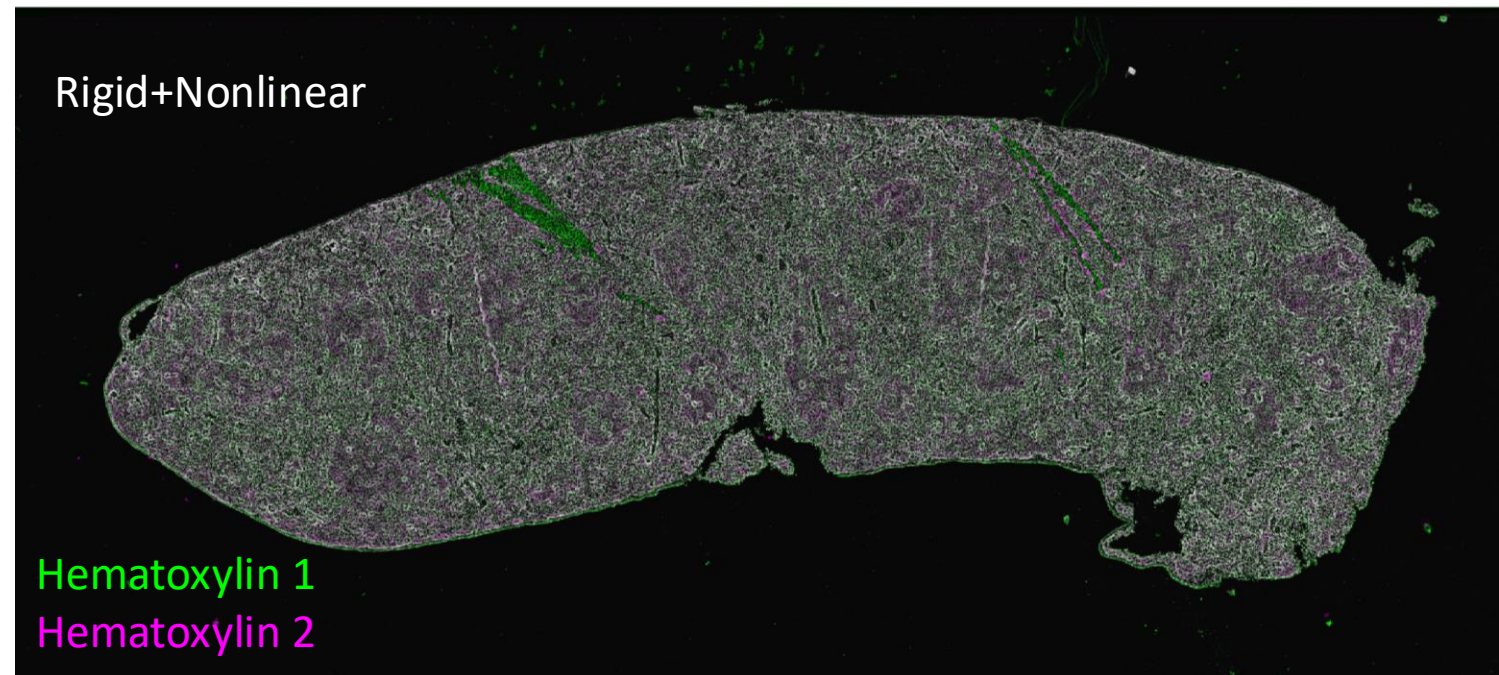
Alignment



Ni et al, Nature Scientific Reports, 2020



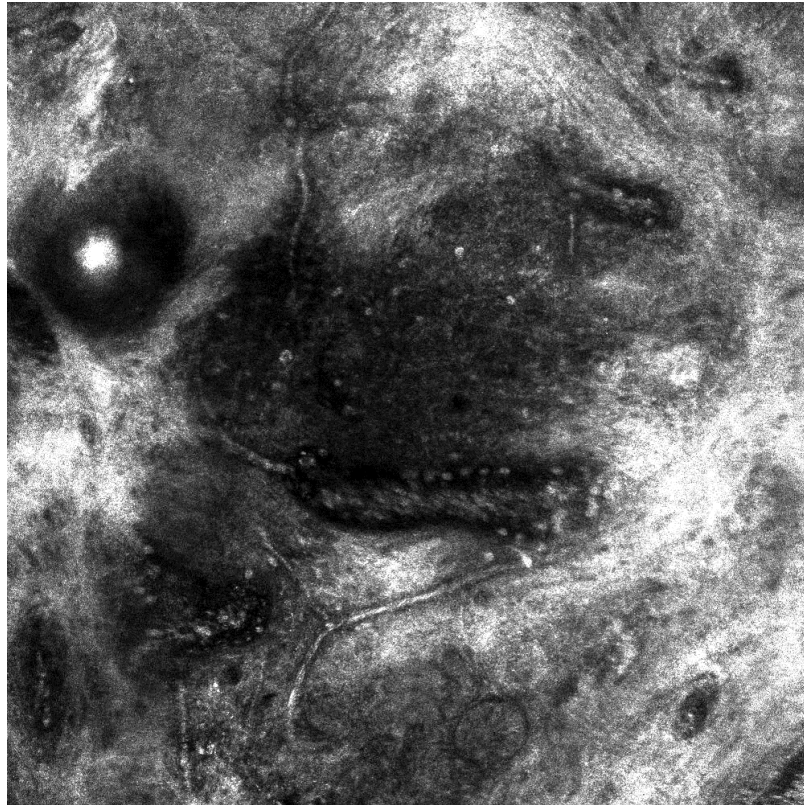
Kajihara et al, Pattern Recognition, 2019



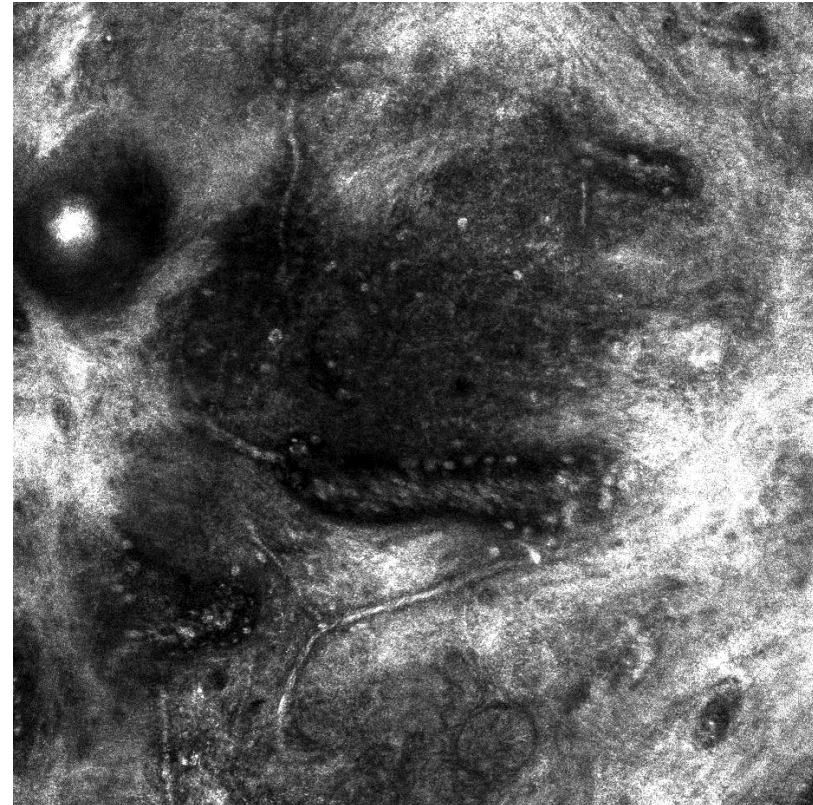
In vivo tumor immune microenvironment phenotypes correlate with inflammation and vasculature to predict immunotherapy response

Sahu et al Nature Communications, 2022

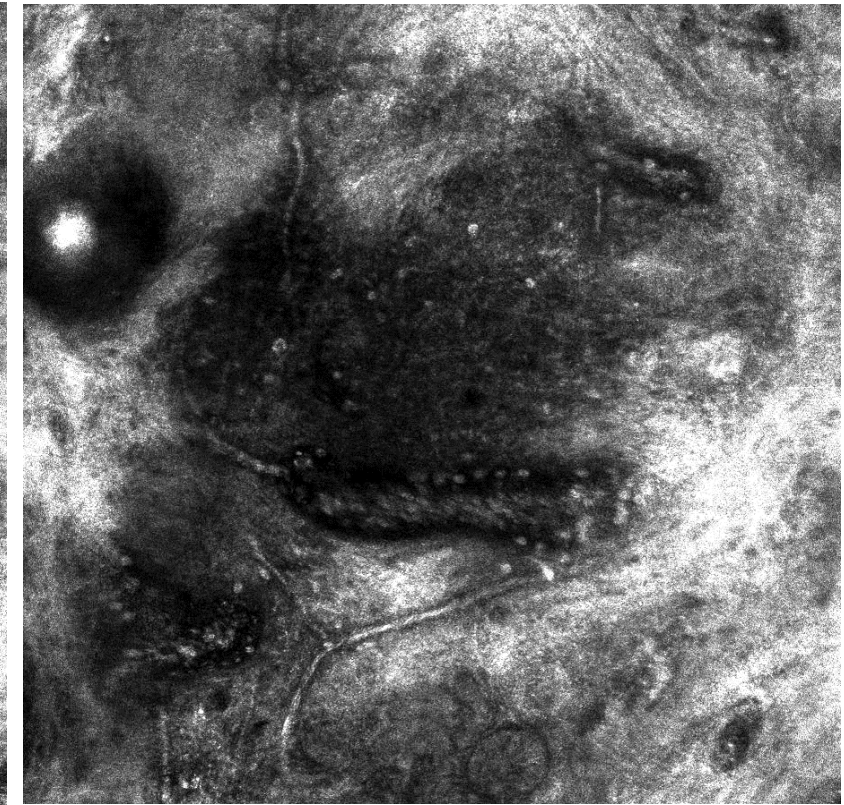
Alignment: Example



Raw data



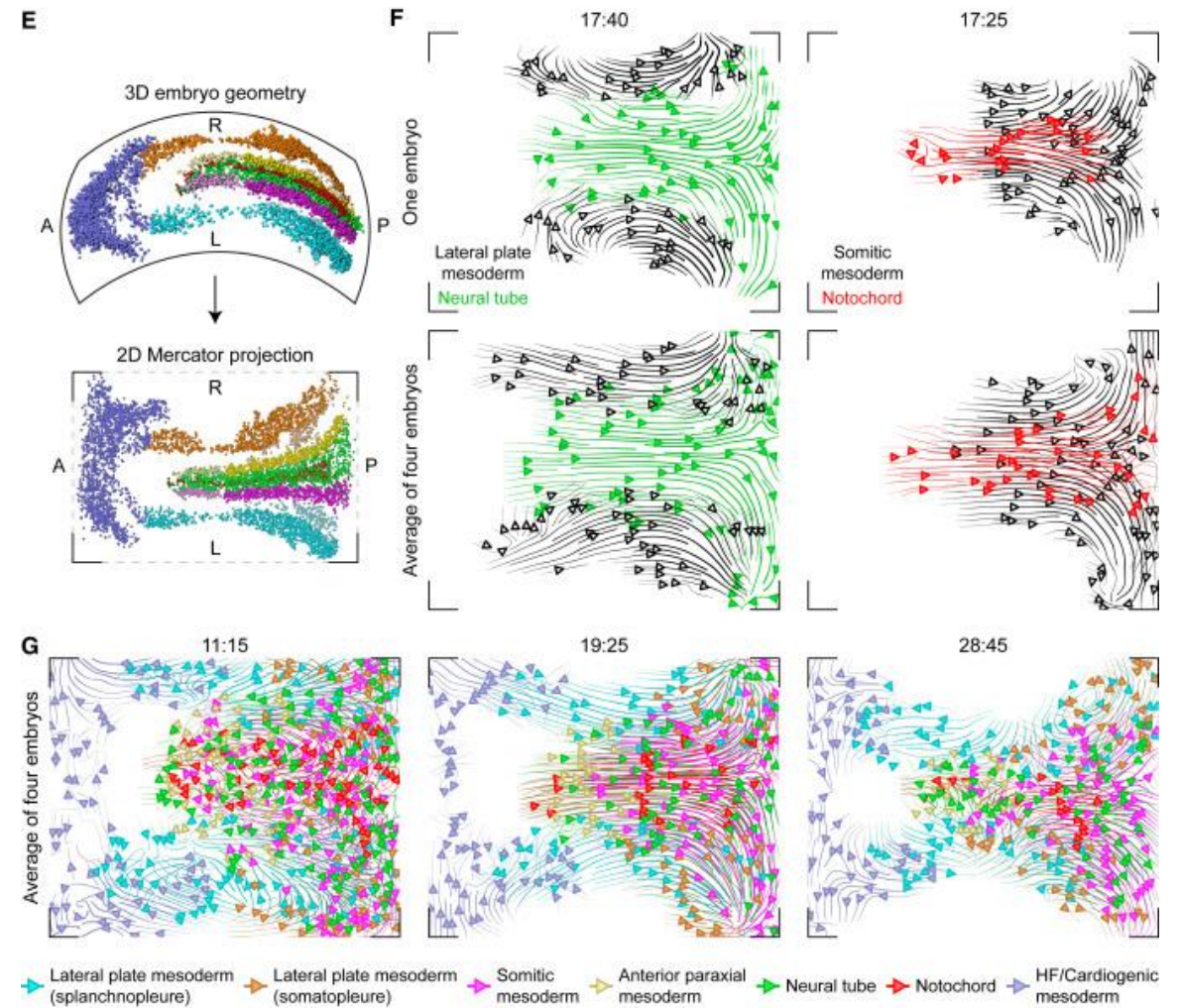
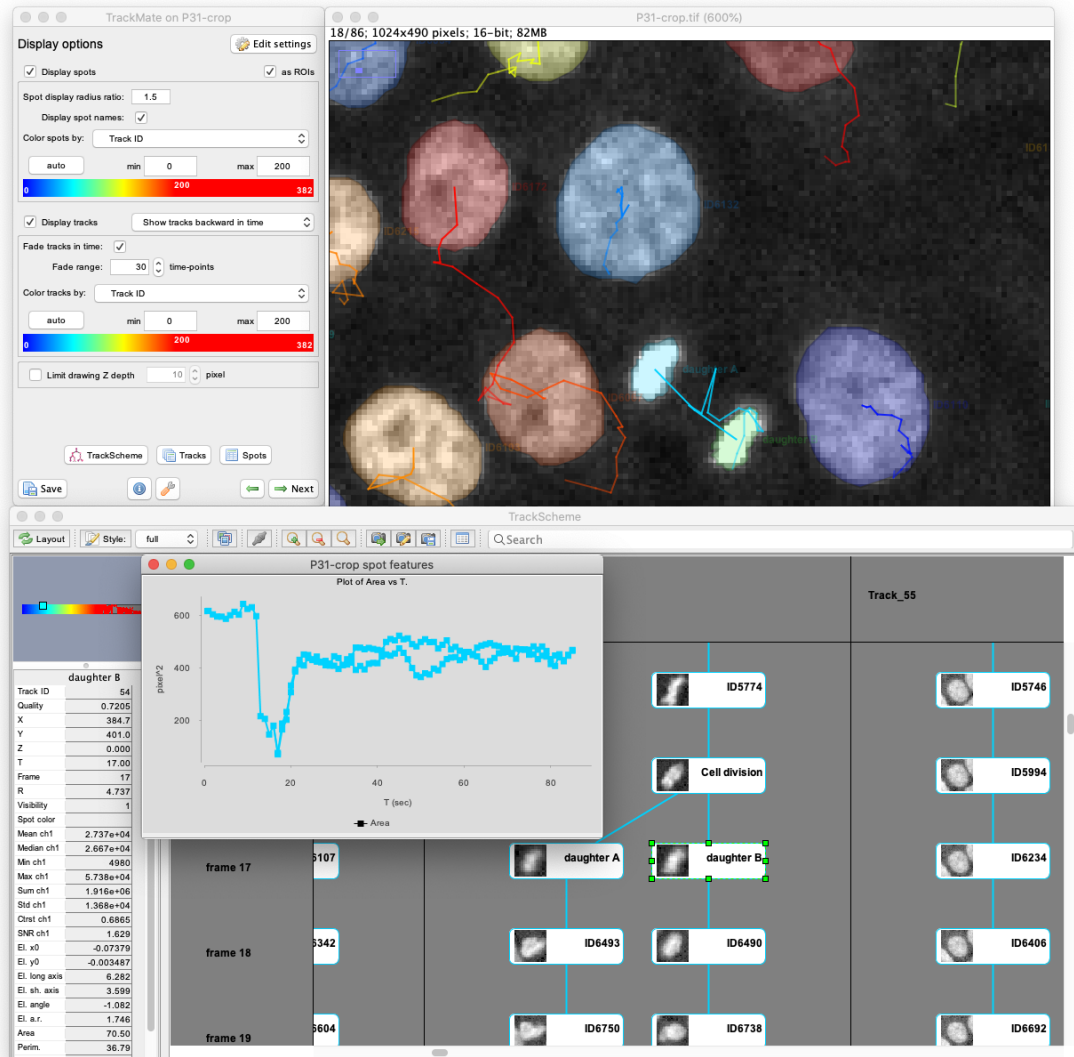
Fiji SIFT feature affine alignment

Plus MATLAB imregdemon
nonlinear alignment

What might be some possible risks of this kind of nonlinear warping?

Tracking

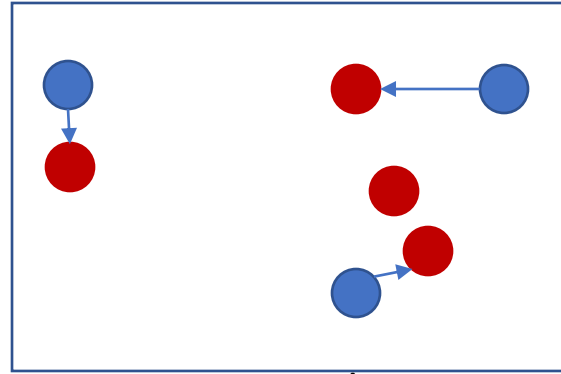
Trackmate



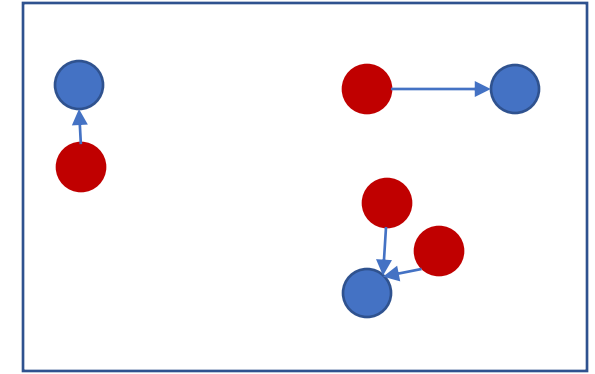
Tracking

- Detect and Link
 - Nearest neighbor
 - Linear Assignment

t1 ● t2 ●

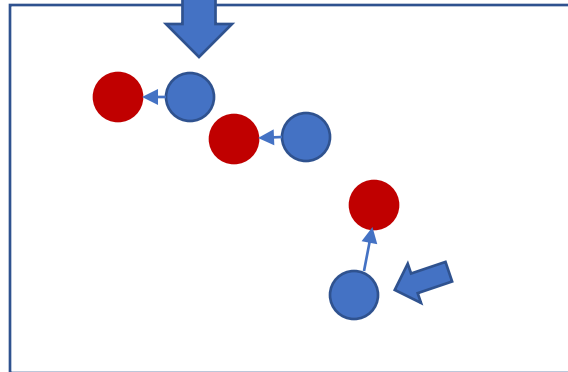


Forward

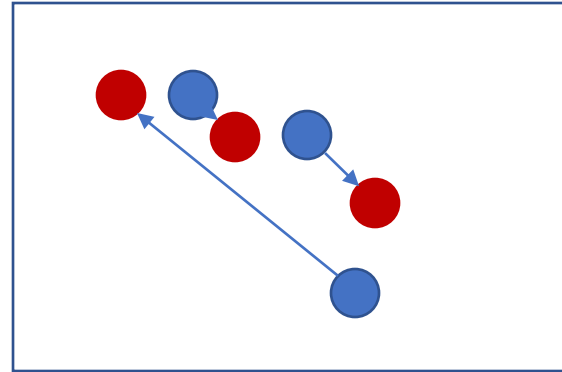


Backward

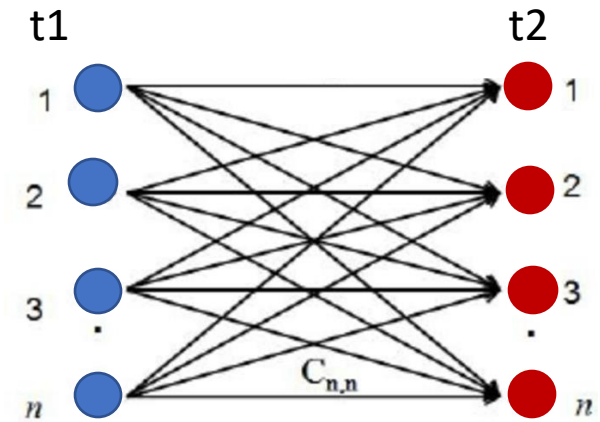
What result would we get if we started here?



Sensible match

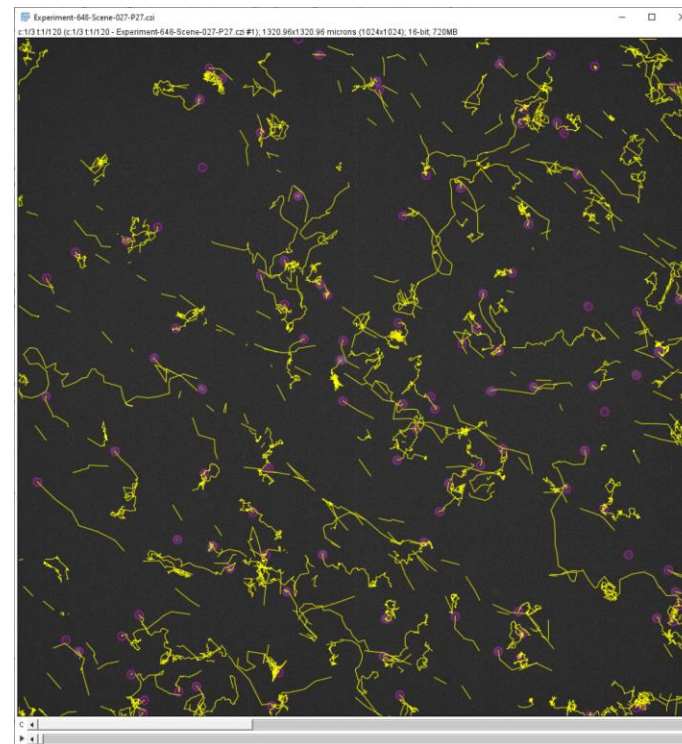
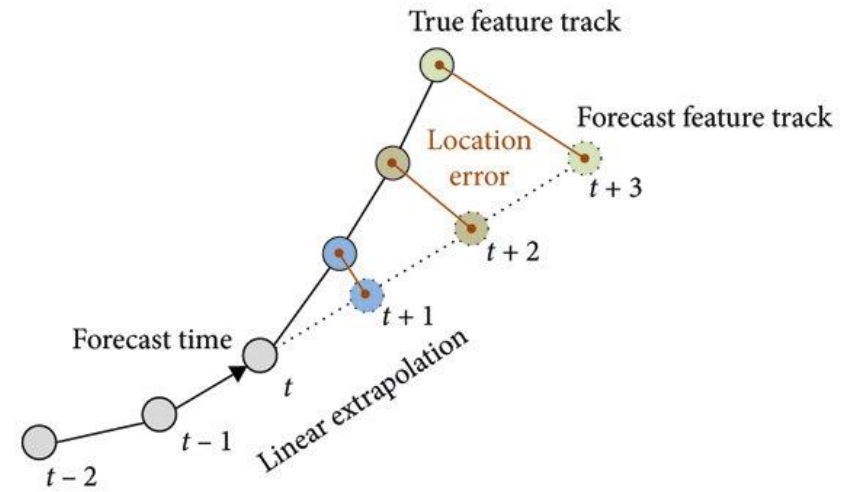
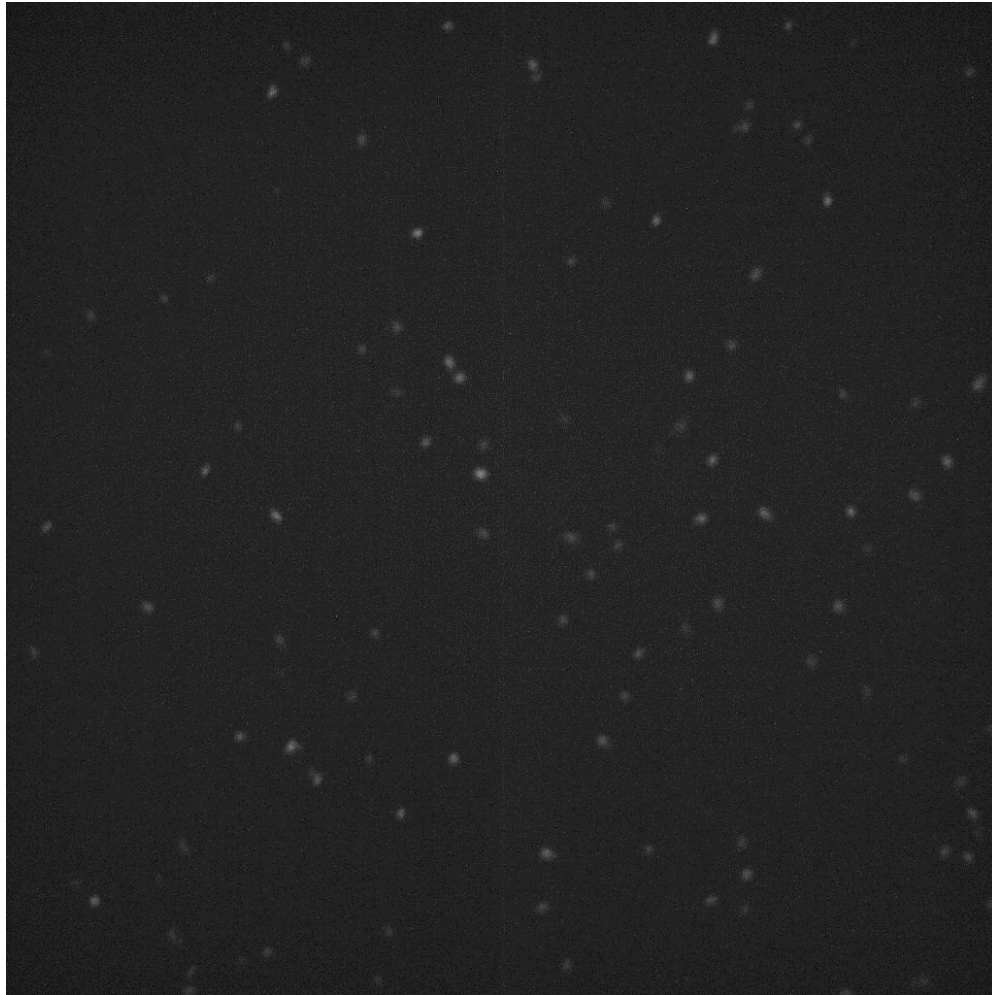


Not so sensible, but 'correct'

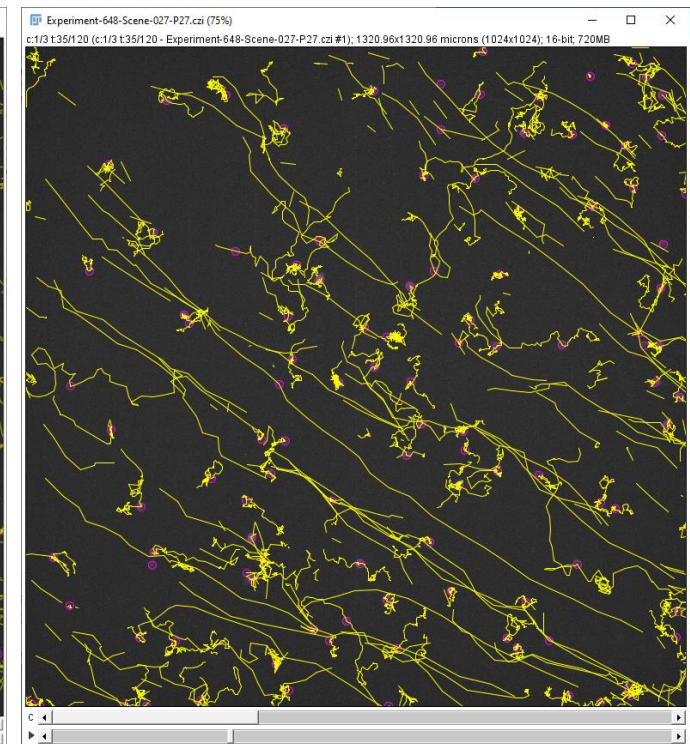


LAP represents the local cost of all possible matches

Motion Models



LAP threshold 40



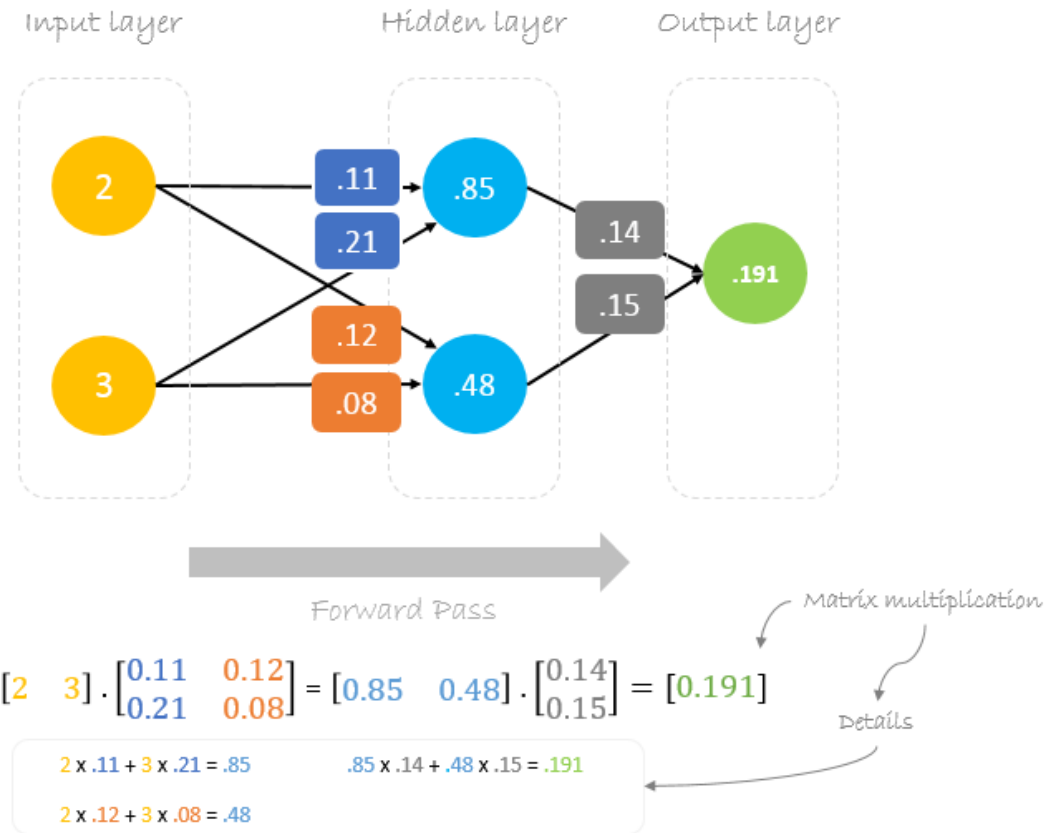
Linear Motion LAP threshold 40

AI and Image Analysis

- Introduction to deep learning
 - Overview of Uses
 - The pieces: Architecture, data, loss, postprocessing
 - Architectures
 - CNN
 - GAN
 - Vision Transformers
 - Limitations
 - Generalization
 - Opacity

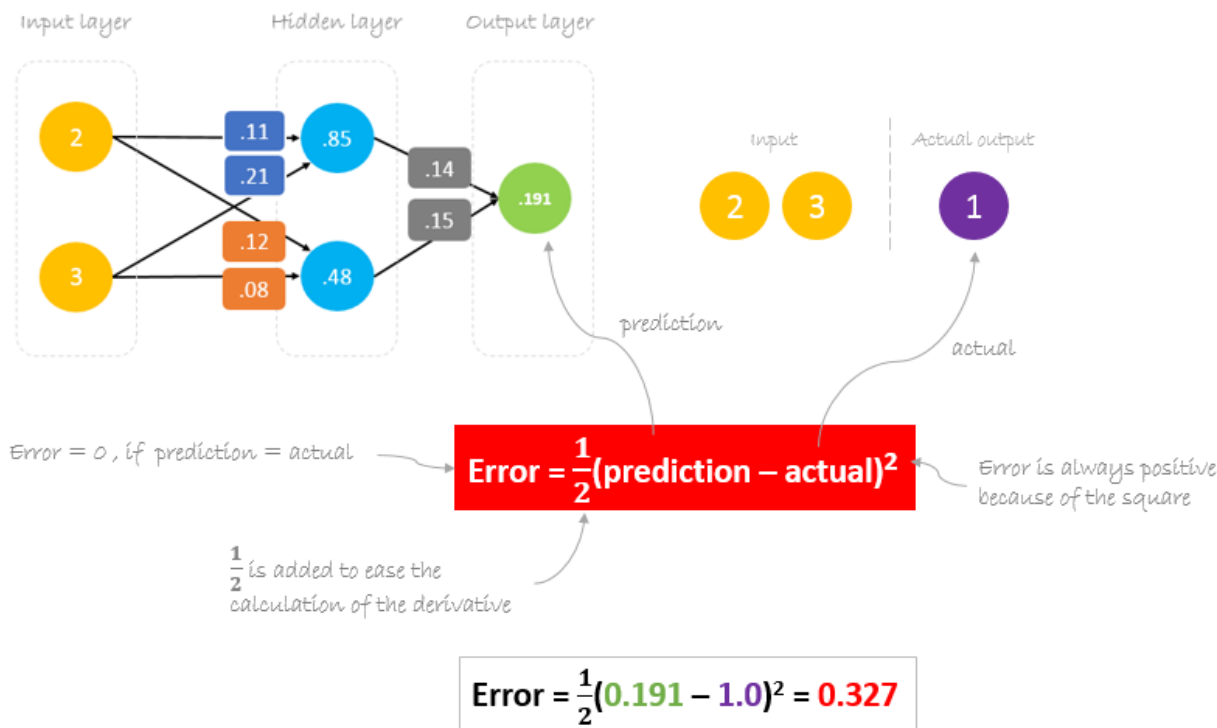
AI is easy!

Forward Propagation



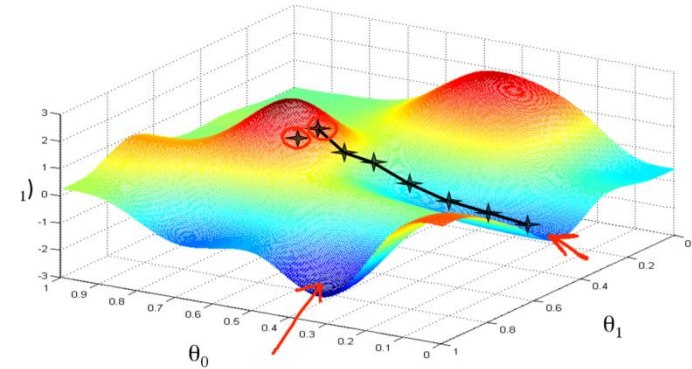
<https://hmkcode.com/ai/backpropagation-step-by-step/>

AI is hard! Backpropagation



What is a potential failure mode here?

Gradient descent



$$*W_x = W_x - a \left(\frac{\partial \text{Error}}{\partial W_x} \right)$$

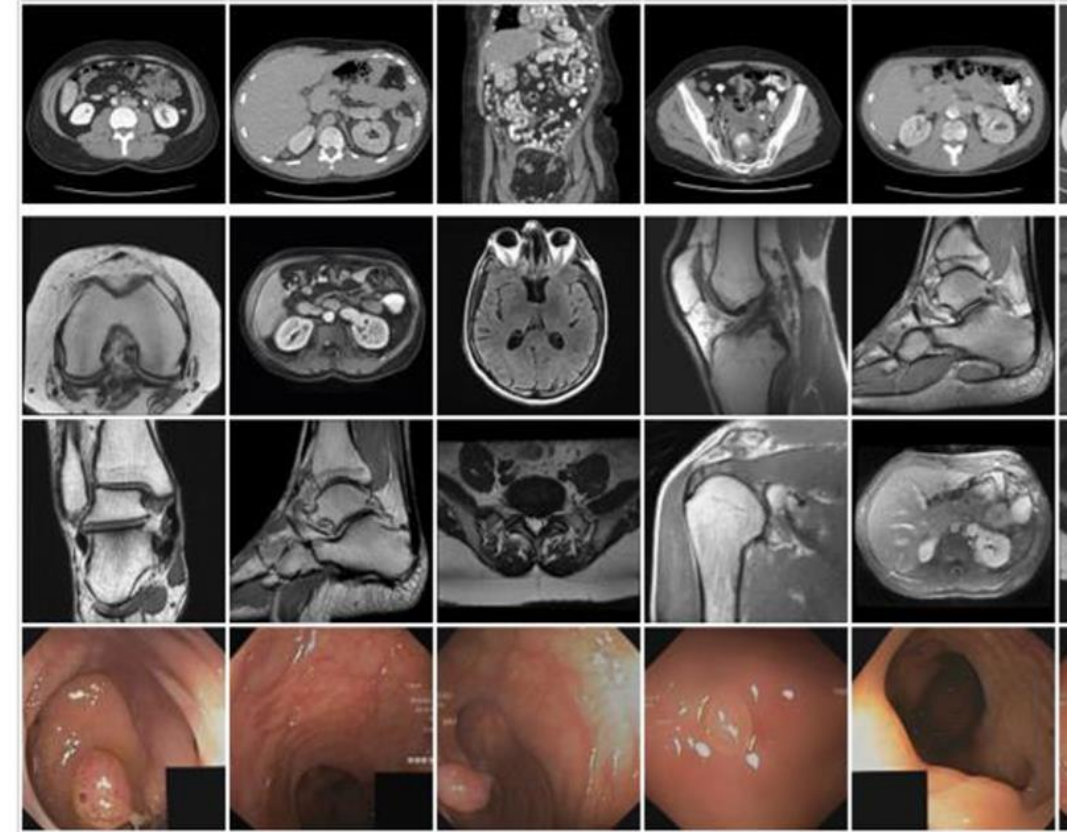
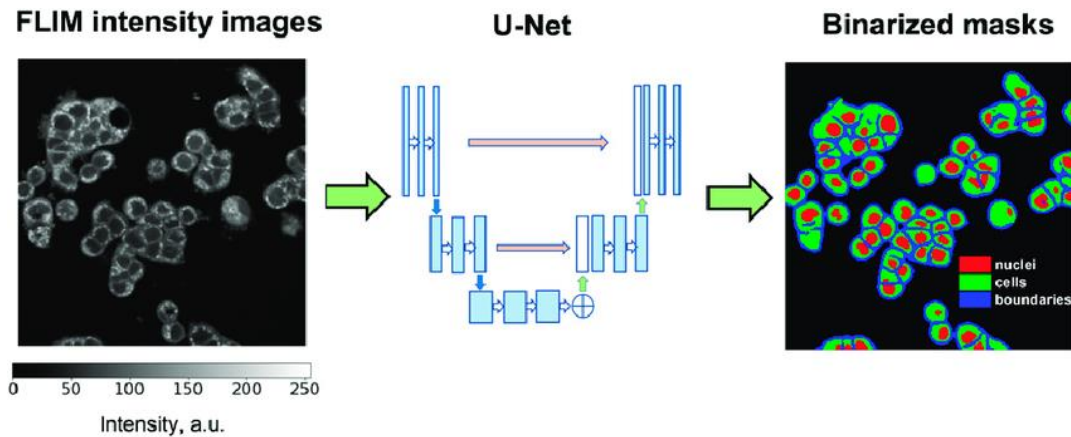
Annotations for the equation:

- $*W_x$: New weight
- W_x : Old weight
- a : Learning rate
- $\left(\frac{\partial \text{Error}}{\partial W_x} \right)$: Derivative of Error with respect to weight

Uses

Regression to some Desired Output

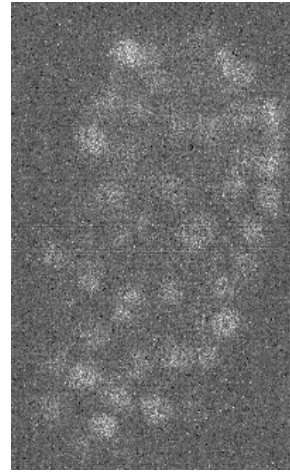
E.G. Segmentation



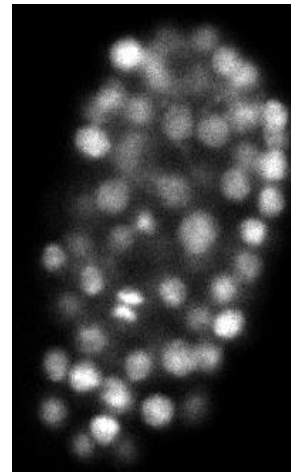
RadImageGAN Liu et al arXiv 2023

Generation of 'Data'

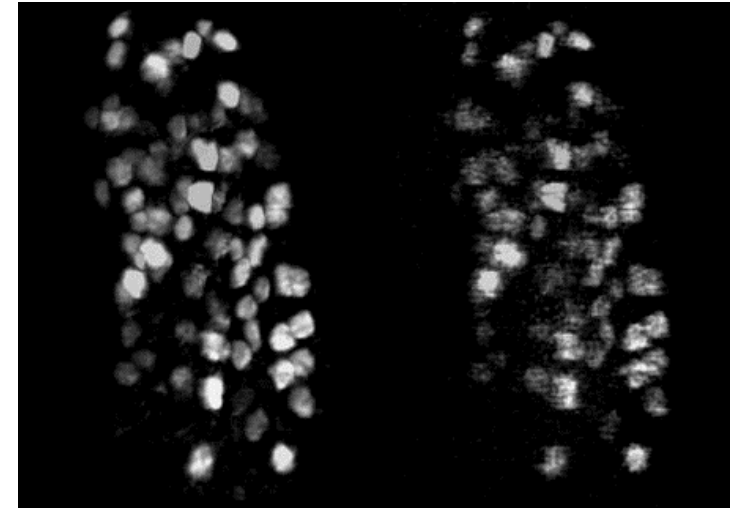
Components



Low contrast

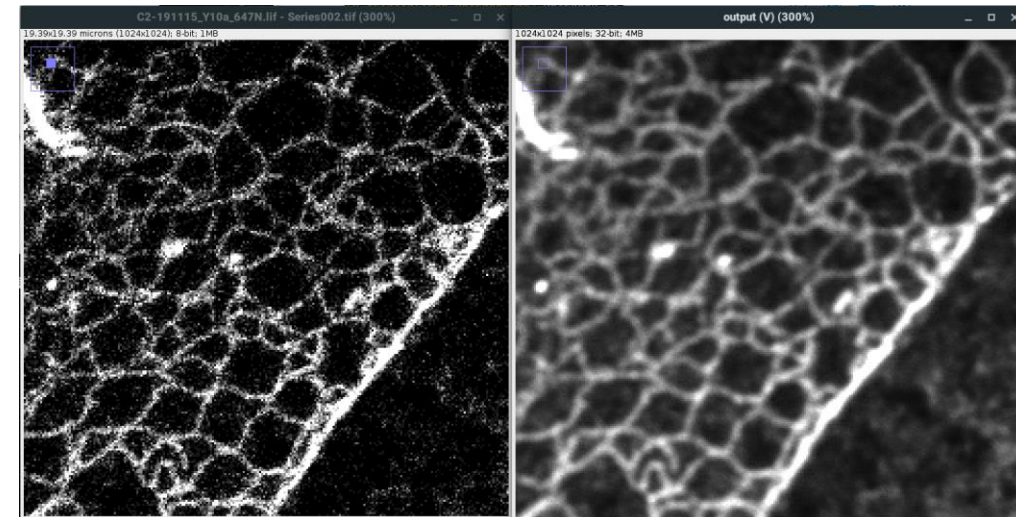


High contrast



CARE (CSBDEEP Unet) Weigert et al Nature Methods 2018

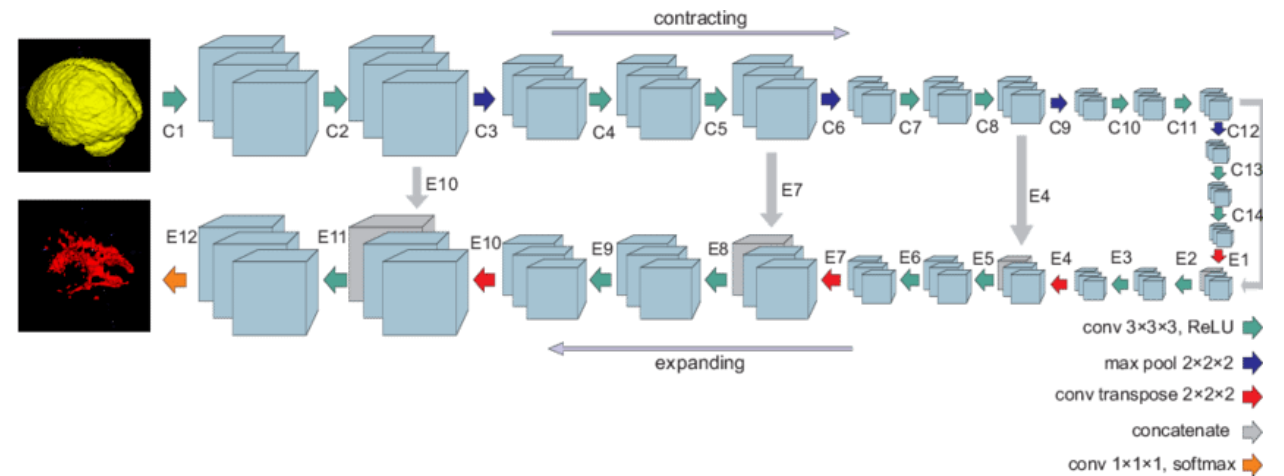
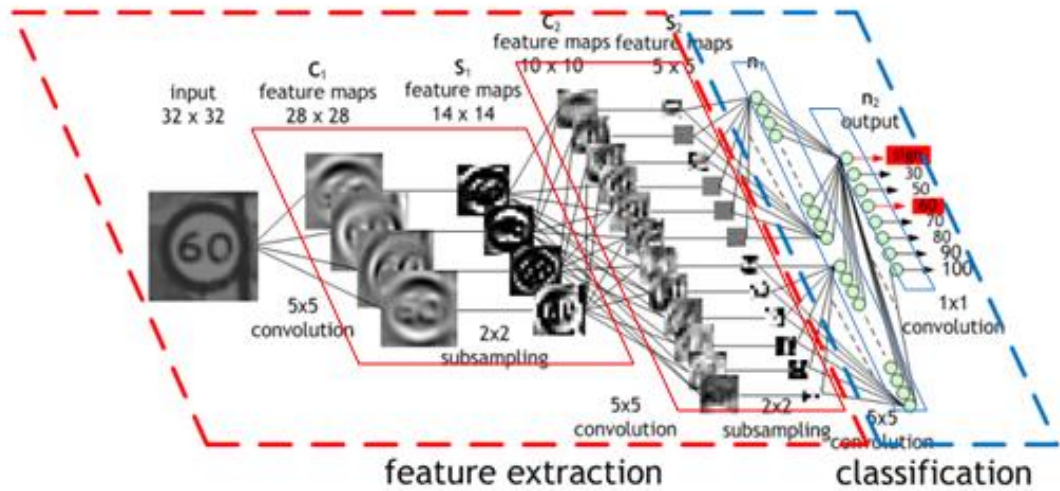
- Architecture –what can it do?
- Format, postprocessing –How to define task?
- Loss –will it learn to do task from this signal?
 - Error on?
 - The task you want to do?
 - An easier to supervise task?
 - Mask and predict
 - Discrimination
 - Relative position
- What might be some pros and cons of each approach?



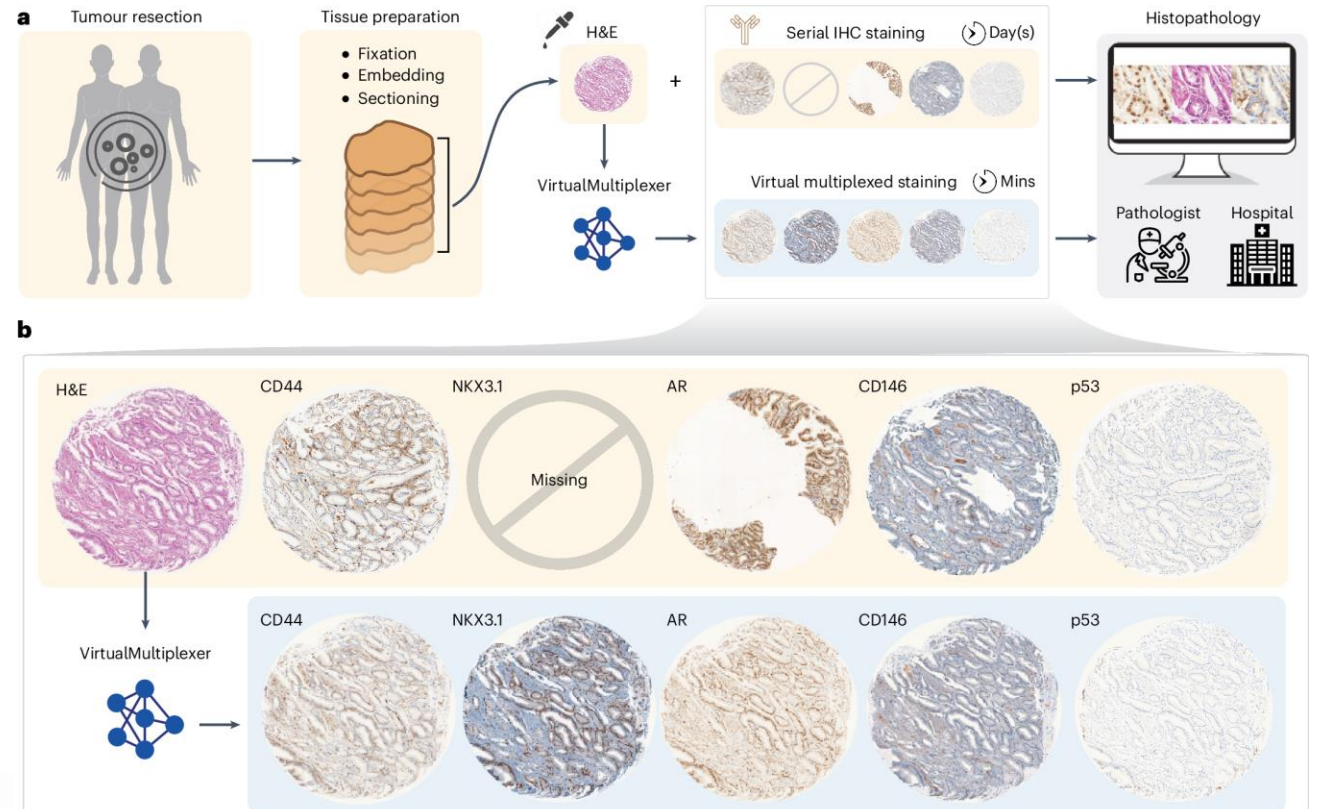
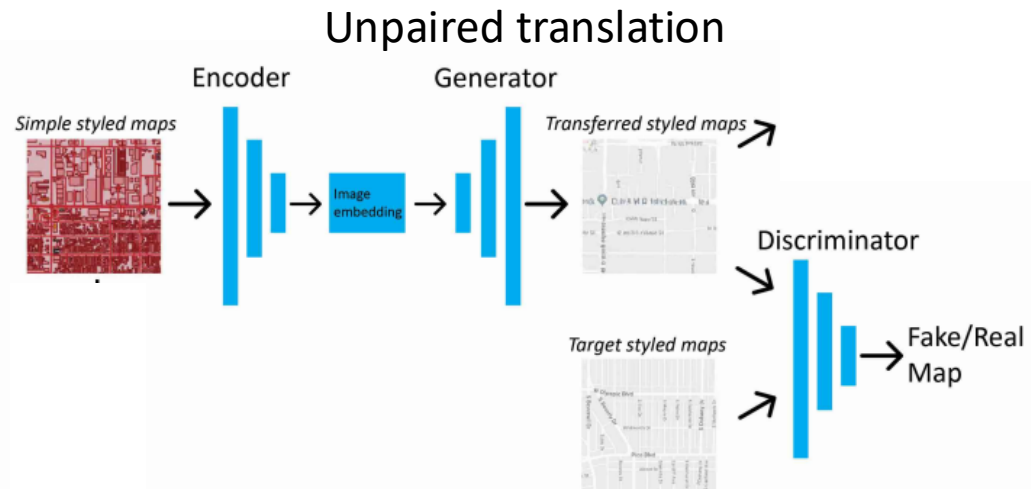
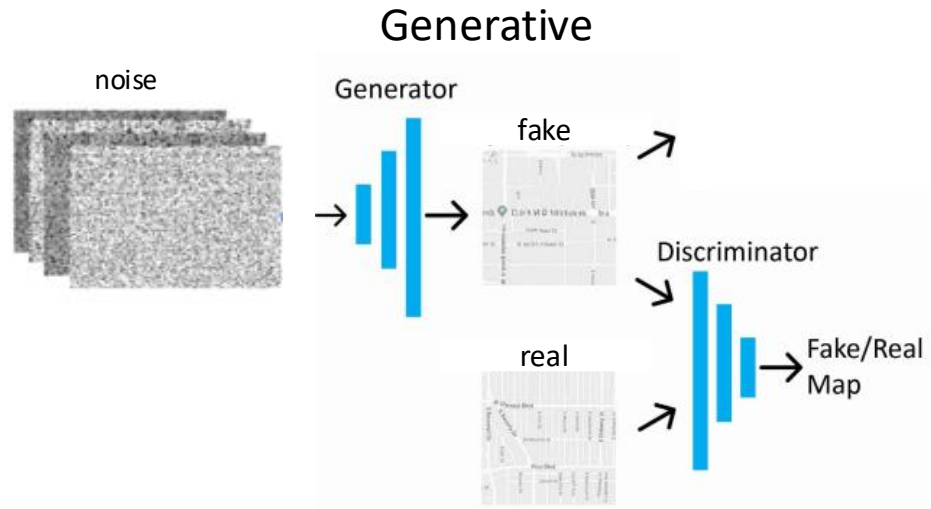
Noise2Void (CSBDEEP Unet) Krull et al CVPR 2019

<https://forum.image.sc/t/noise2void-for-fiji/34552/7>

Architectures -CNN



Architectures GAN

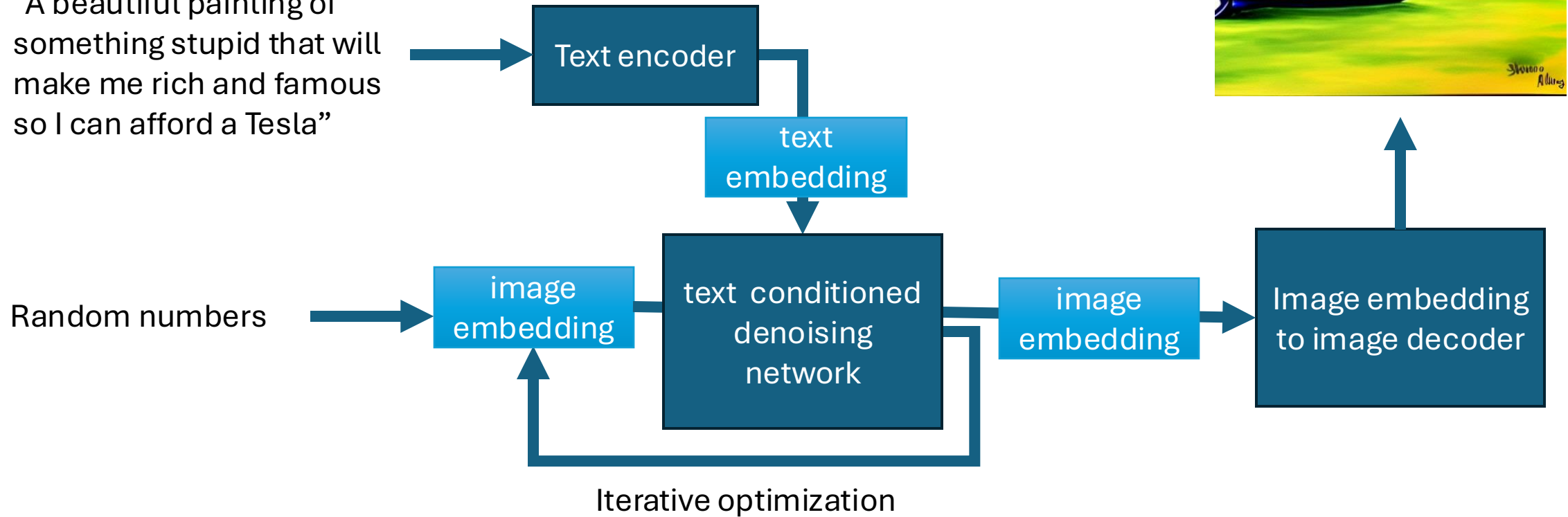


Pati et al Nature Machine Intelligence 2024

Image generation networks

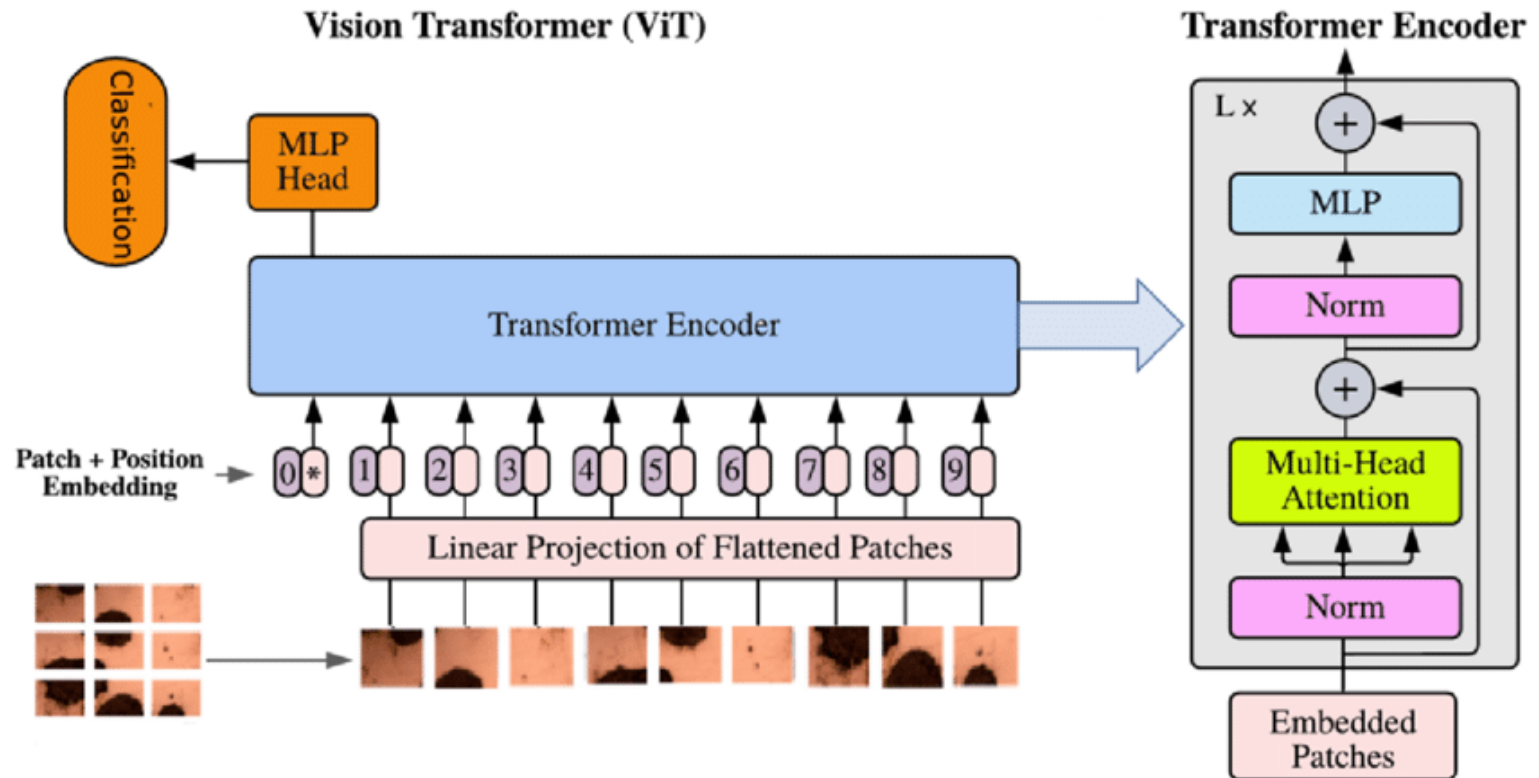
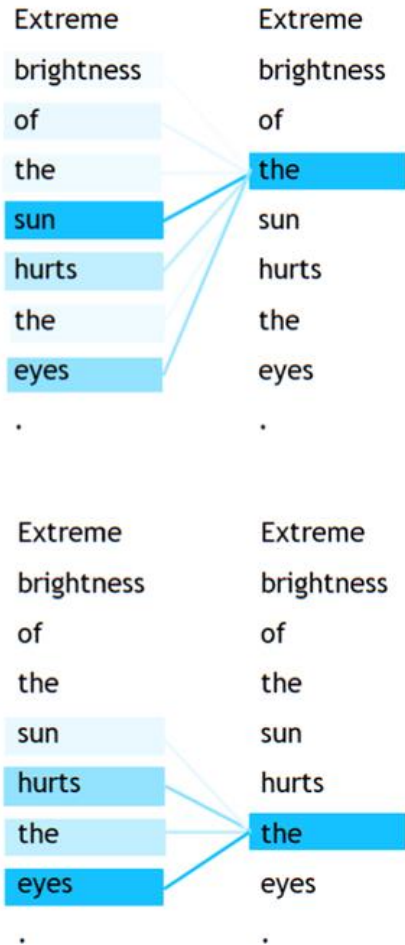
Prompt:

“A beautiful painting of something stupid that will make me rich and famous so I can afford a Tesla”



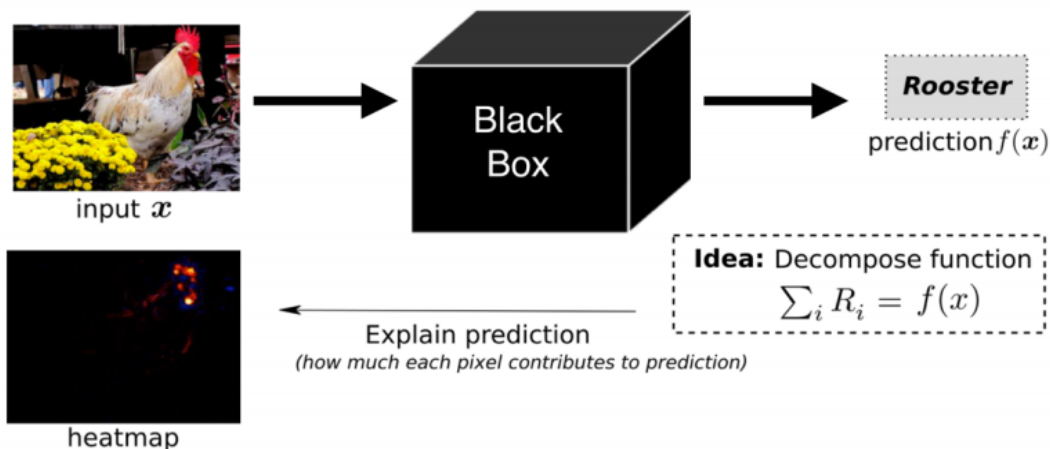
Architectures Vision Transformer

What are some potential tradeoffs for this flexibility?



Limitations

- Wrong and Confident
 - sensitivity to out of sample testing whether hostile or chance
- Largely Opaque



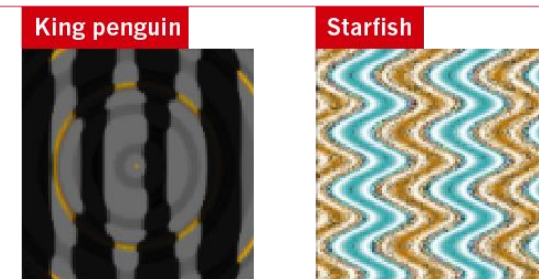
FOOLING THE AI

Deep neural networks (DNNs) are brilliant at image recognition — but they can be easily hacked.

These stickers made an artificial-intelligence system read this stop sign as 'speed limit 45'.



Scientists have evolved images that look like abstract patterns — but which DNNs see as familiar objects.



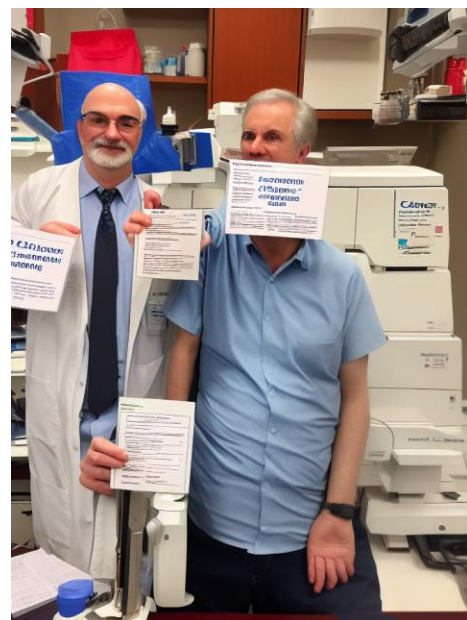
©nature

Even natural images can fool a DNN, because it might focus on the picture's colour, texture or background rather than picking out the salient features a human would recognize.



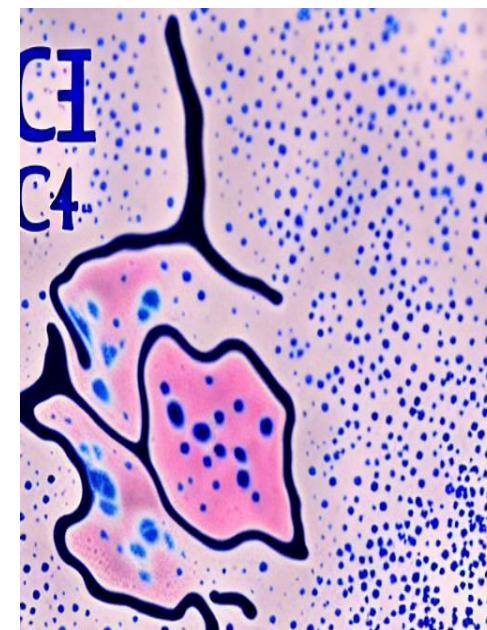
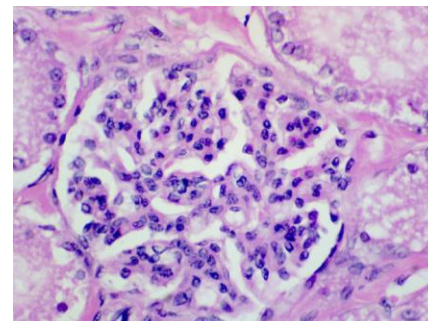
©nature

Why deep-learning AIs are so easy to fool,
Nature Oct 2019

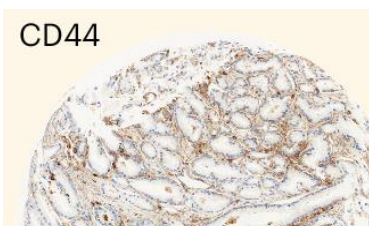


Out of Sample:
You will always get a result, you may not like it
img2img

An oil painting of a man waving by Vincent Van Gough



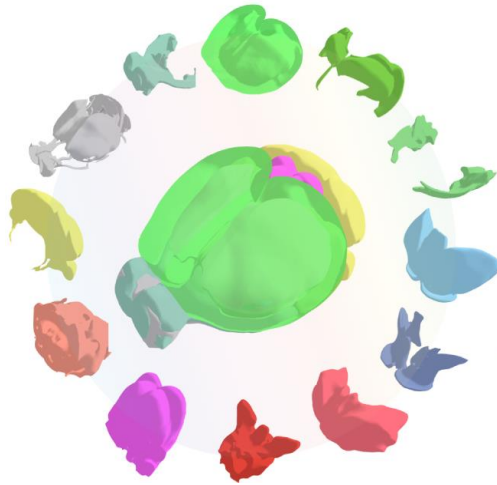
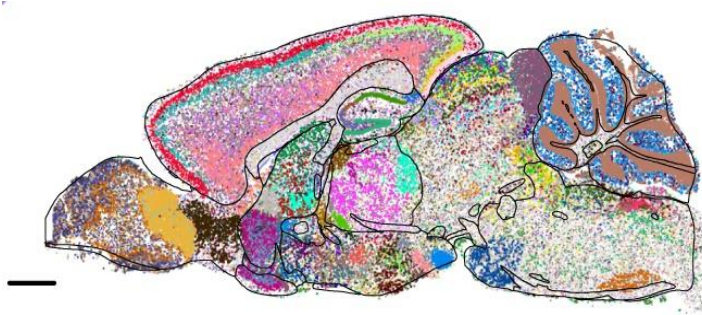
CD44 IHC special stain cancer diagnosis science



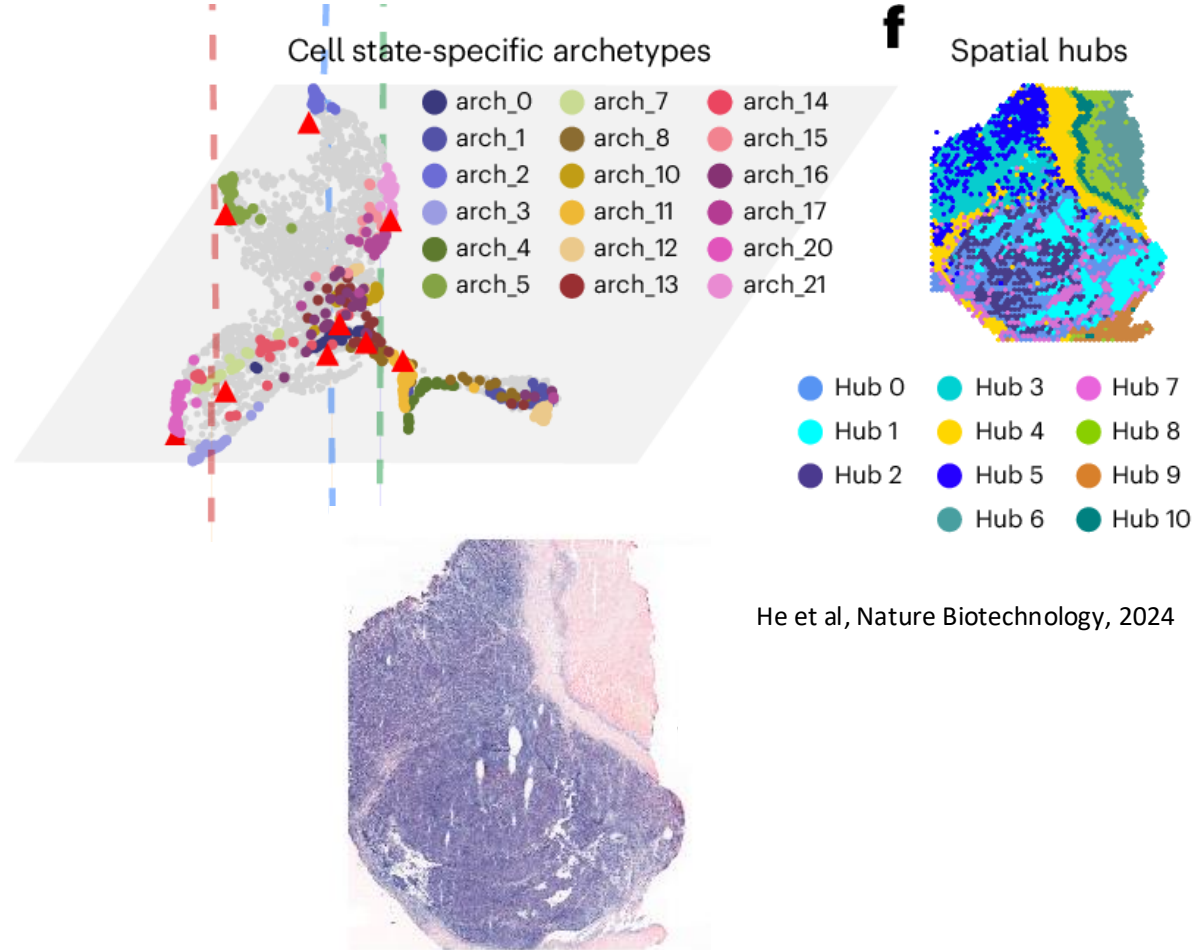
AI in Depth Examples

- Spatial Temporal Pattern Quantification
 - Development
 - Behavior

Spatial Molecular Patterns

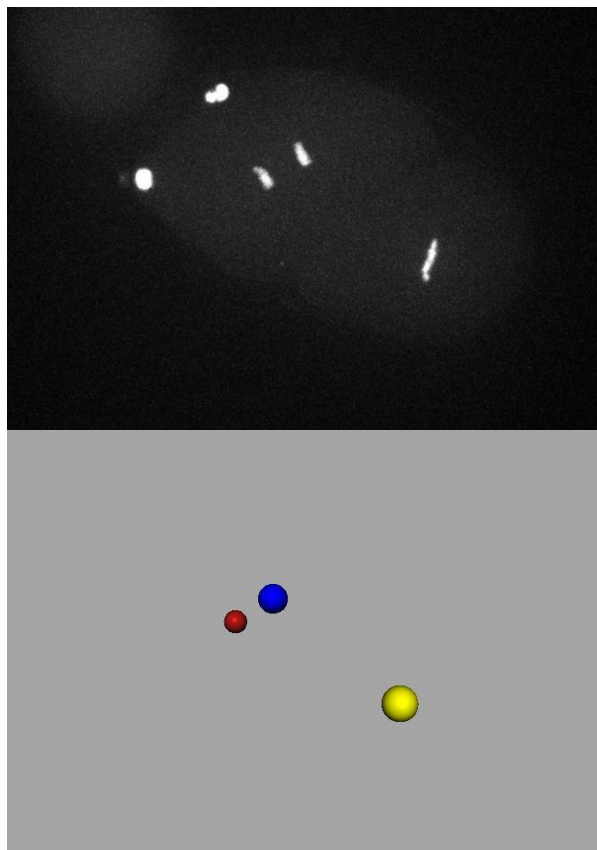
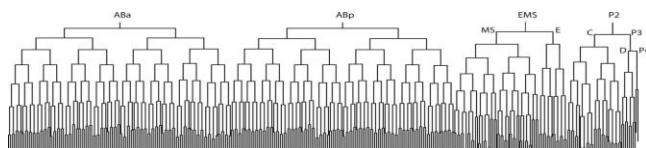


Zhang et al, Nature, 2023

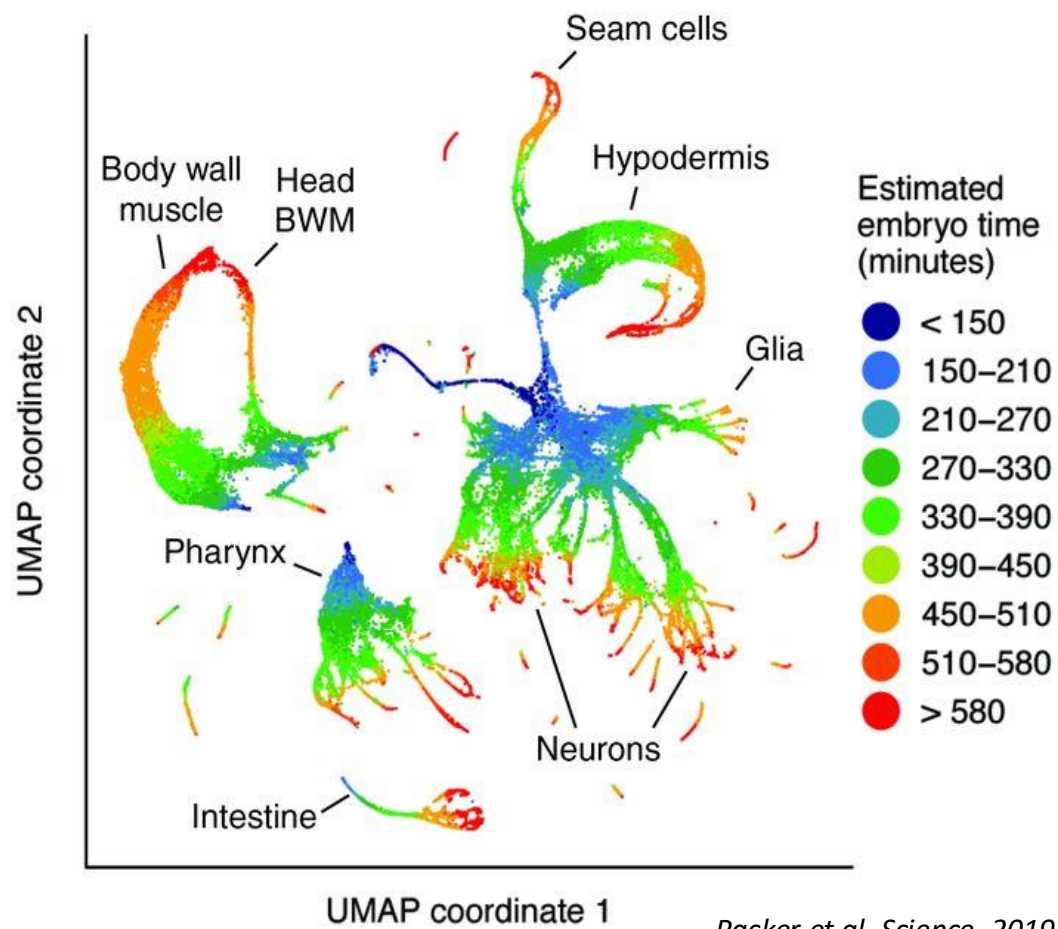


Space, Time and Gene Expression in *C. elegans* embryogenesis

Combining morphogenesis and molecular manifolds



Cell Positions Over Development

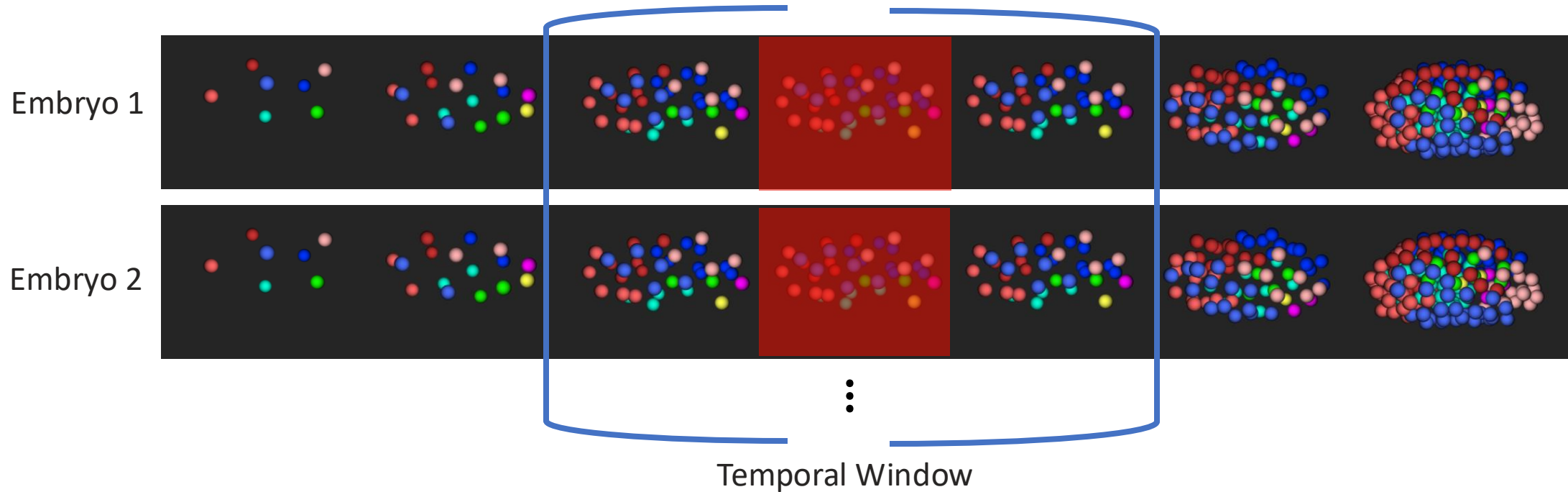


Packer et al. Science, 2019

Molecular manifold

Efficient Joint Spatial Temporal Learning

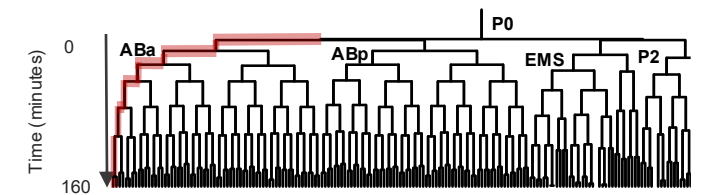
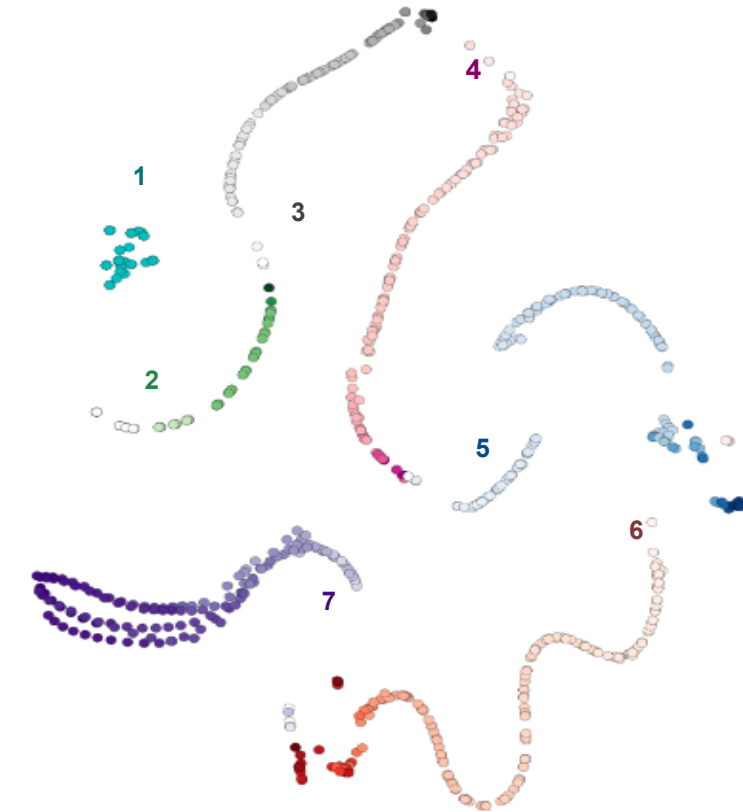
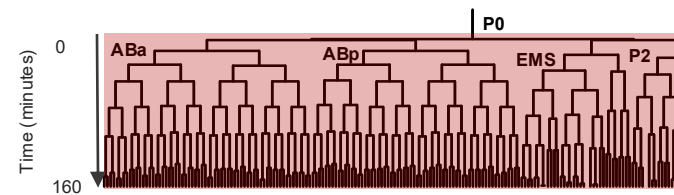
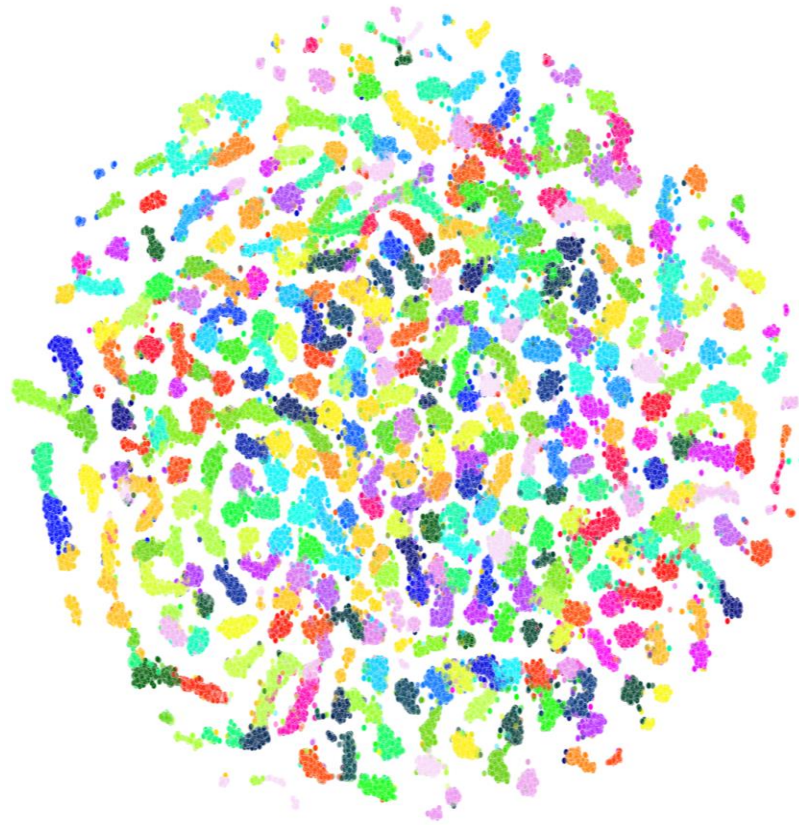
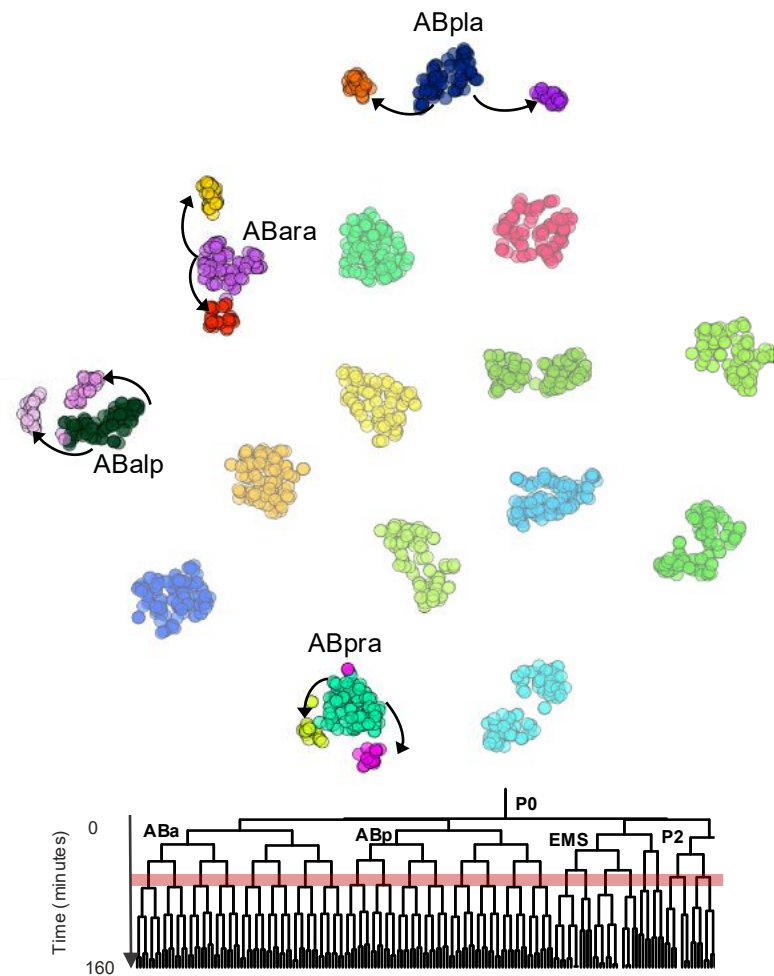
Transformer Based Spatial Representation



1. Transformer Encoder
2. Twin Attention: Training to match cells in similar pairs
3. Sliding Window

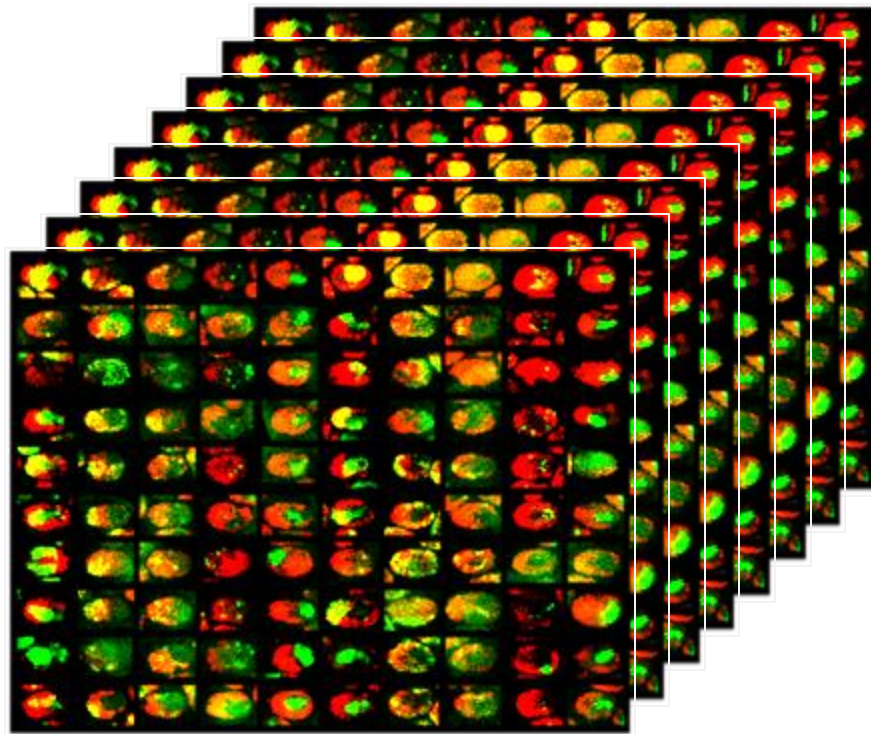
Properties of Learned Embeddings

Learned Manifold of Development

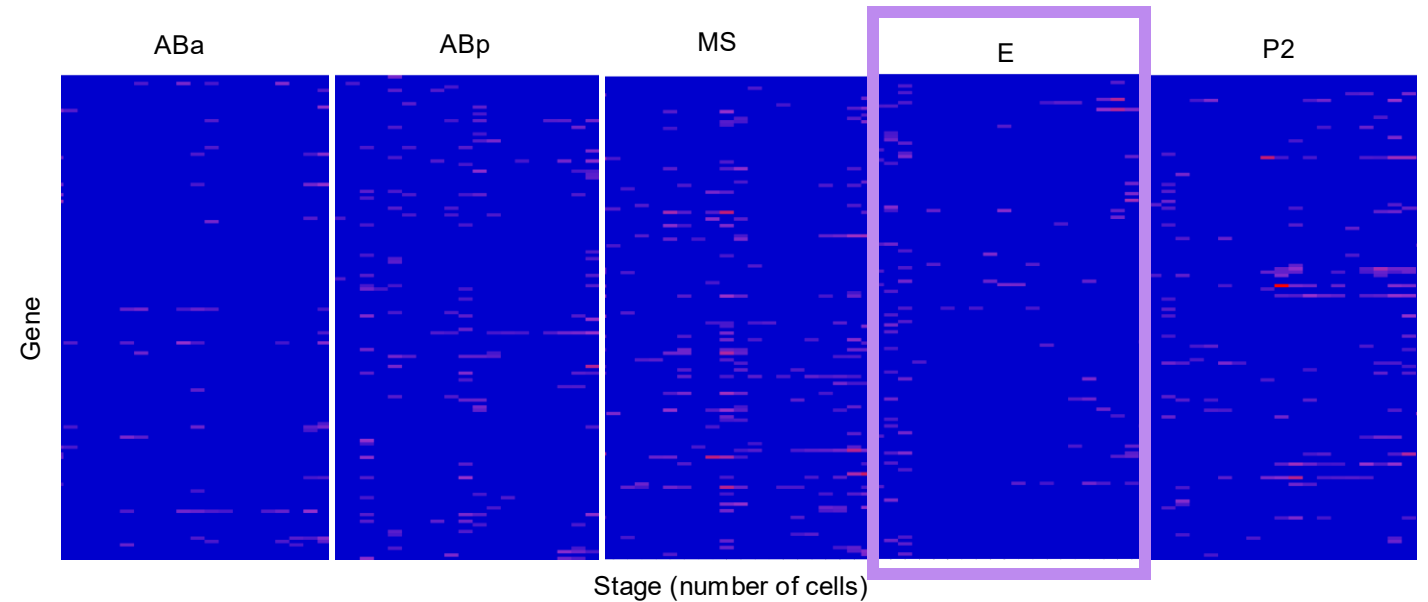
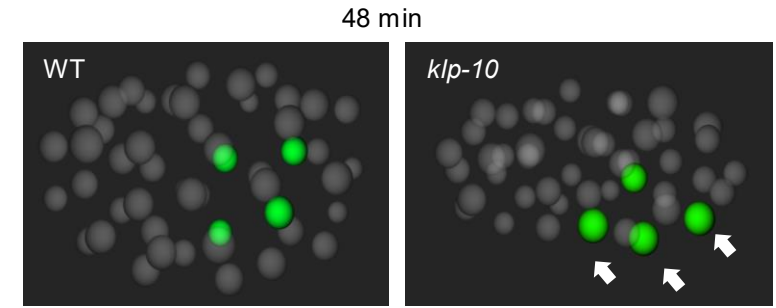
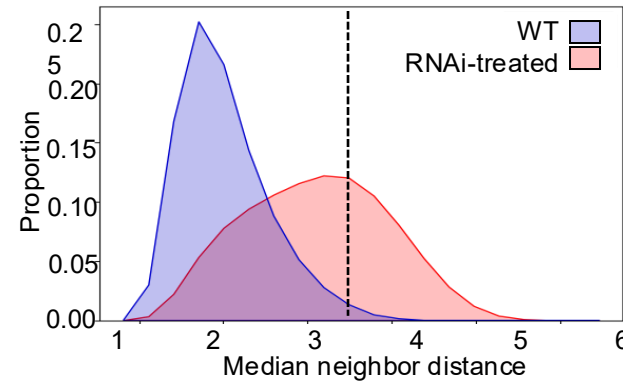


Applications

Transient, Correctable Defects



Du et al Cell, 2014



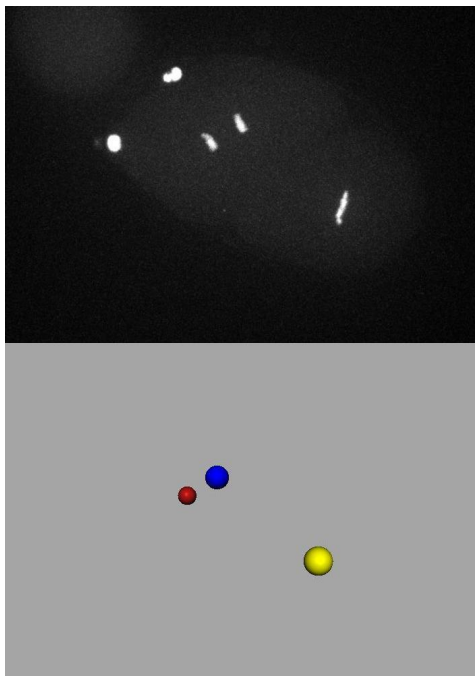
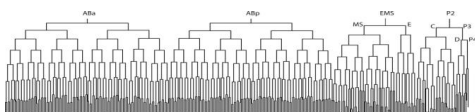
Dataset:

embryos that survived RNAi treatment for 144 embryonic lethal genes

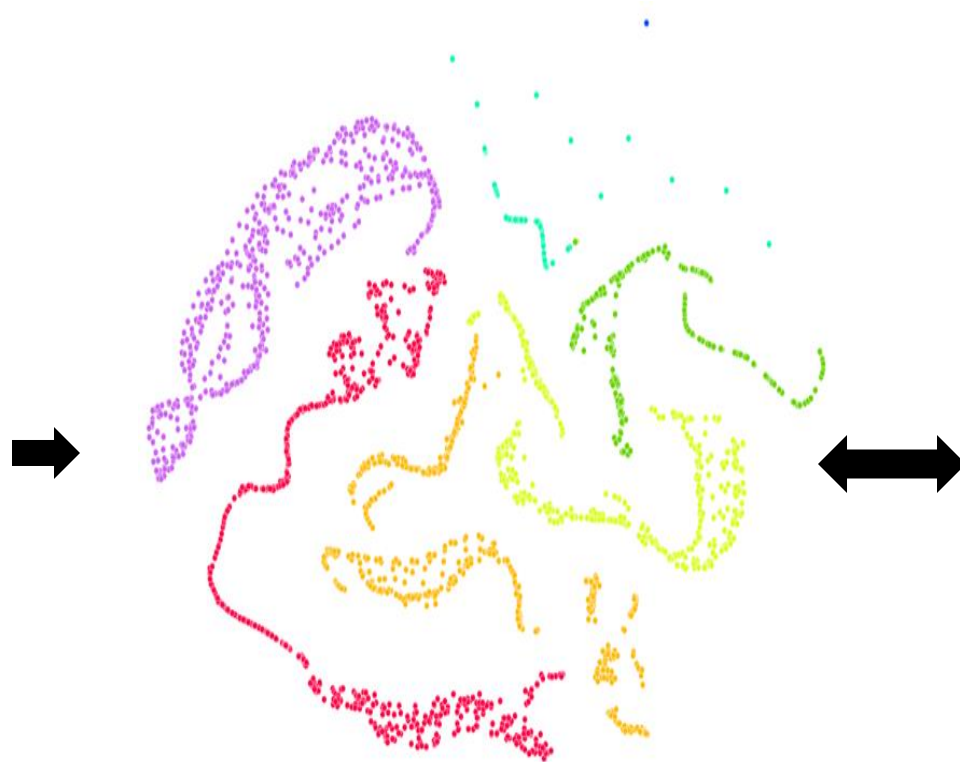
Spatial Molecular Correlation

Combining morphogenesis and molecular manifolds

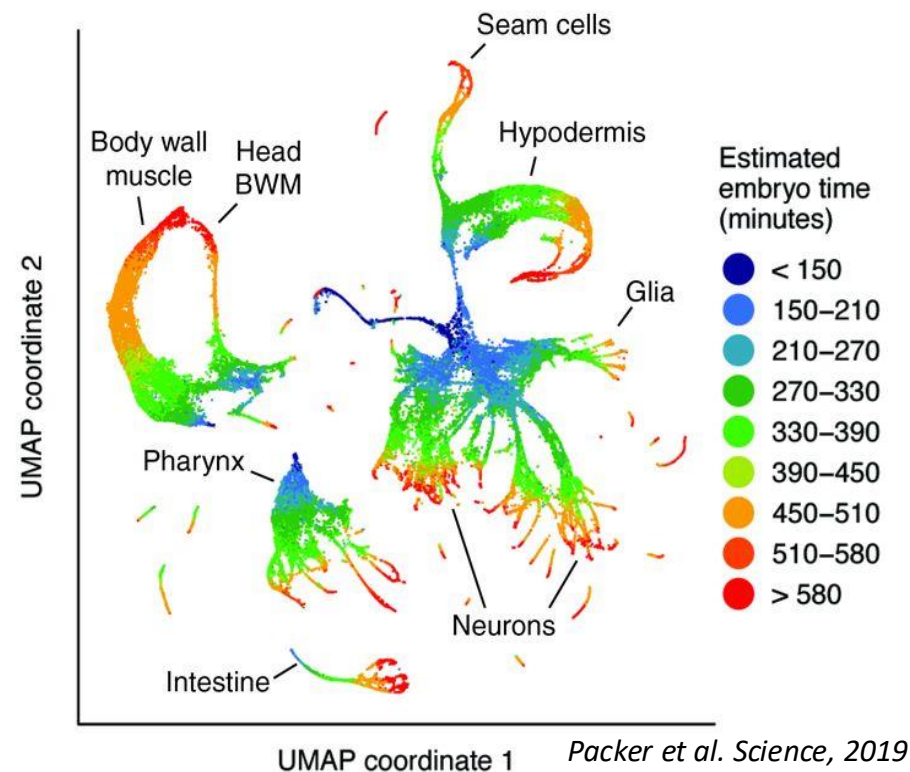
Lineage



How could we use these two manifolds now that we have them and can map between them?

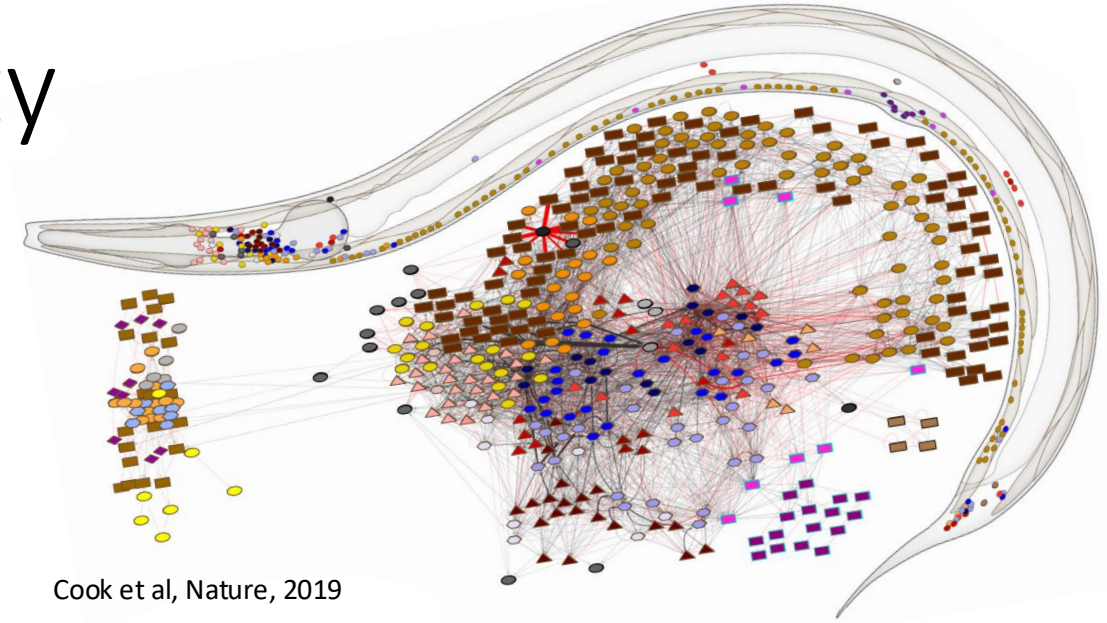
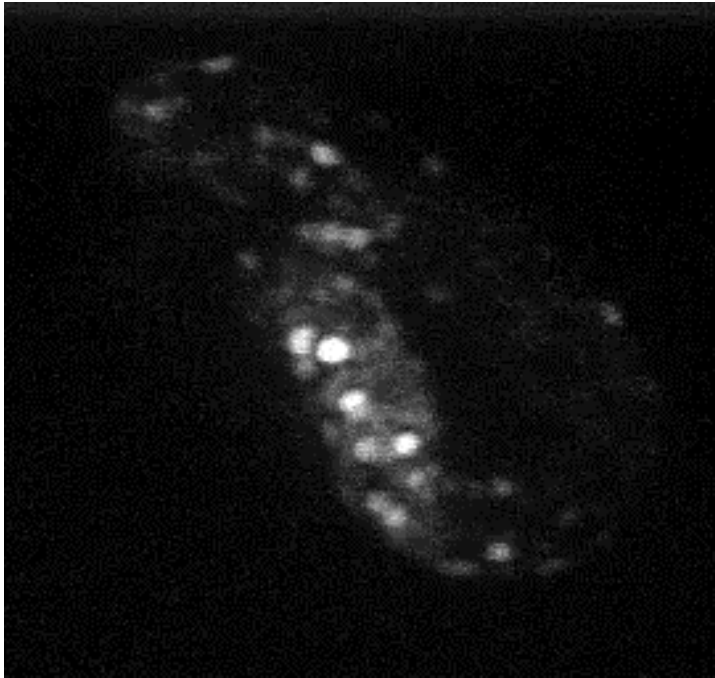


Tissue development manifold

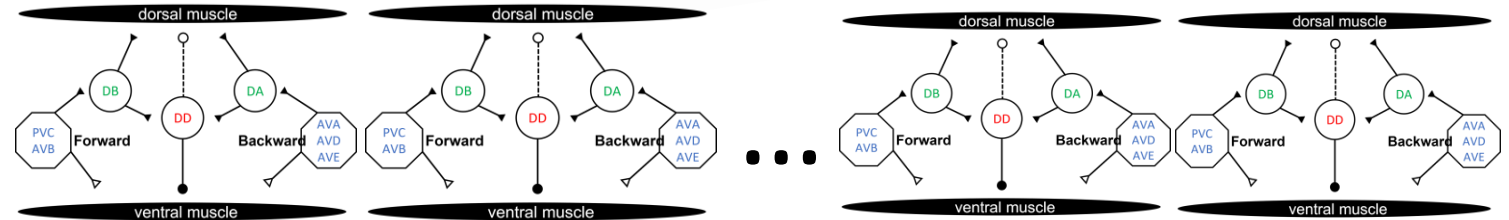


Molecular manifold

Motion and Neural Activity

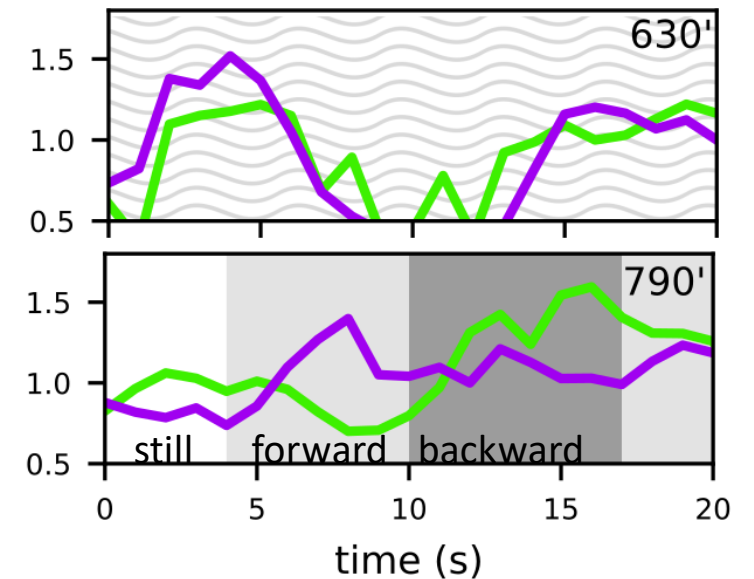
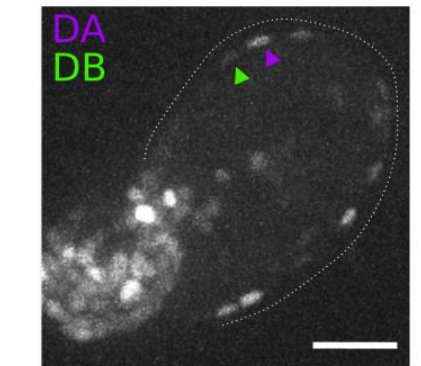
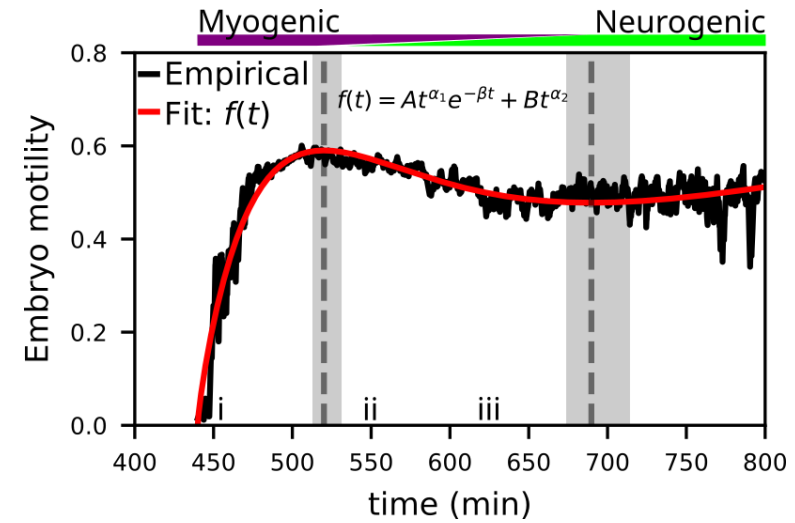
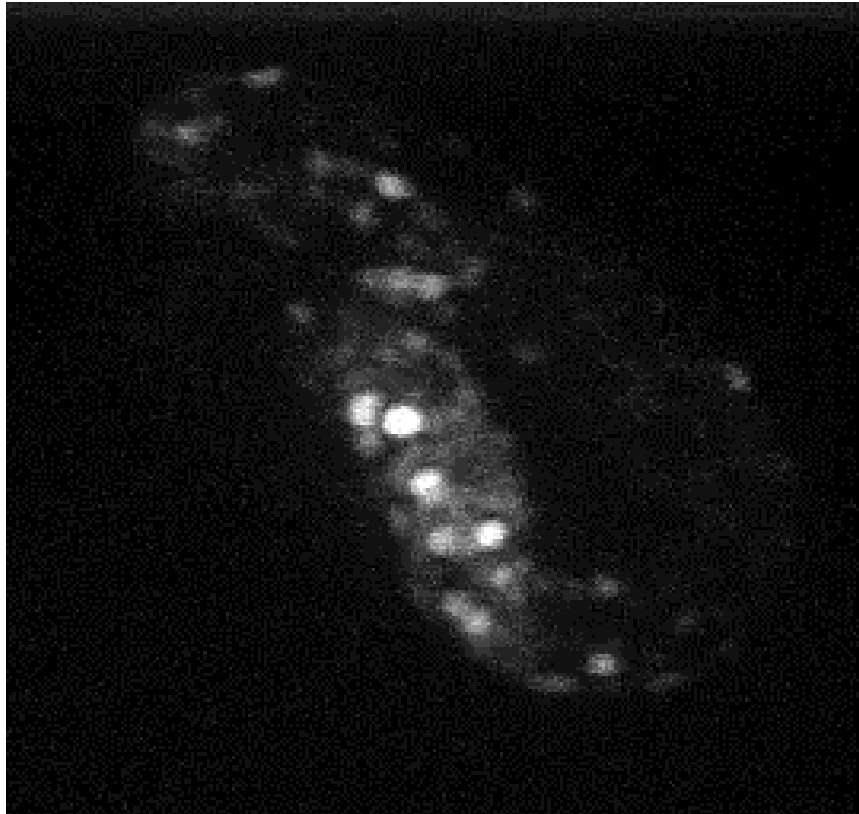


Cook et al, Nature, 2019



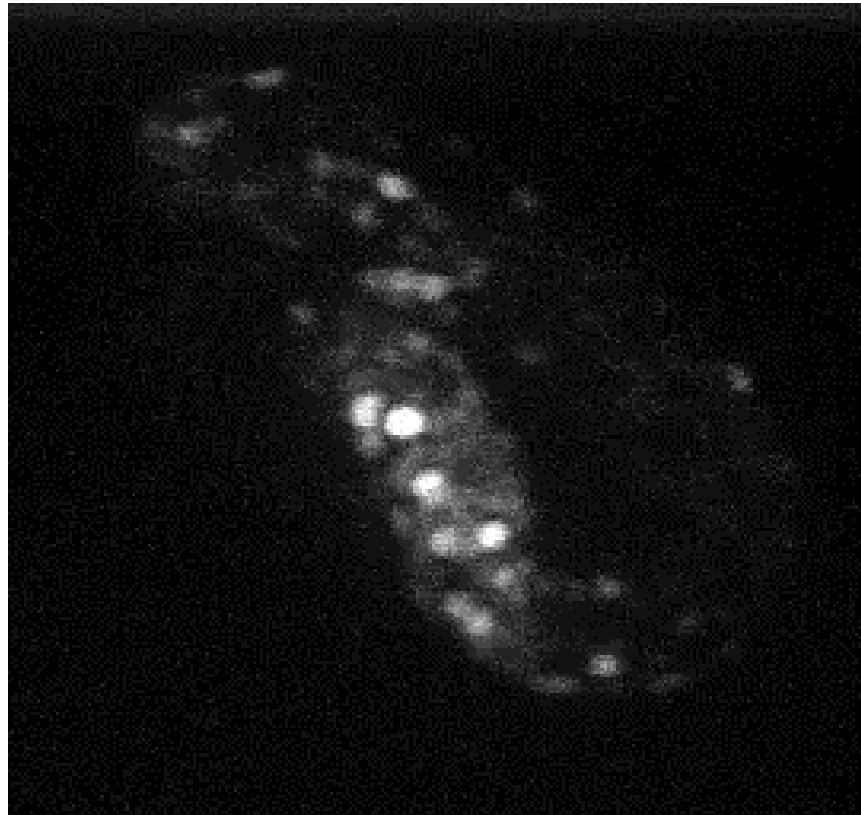
Question: How does circuit activity pattern emerge over development?

Motion and Neural Activity

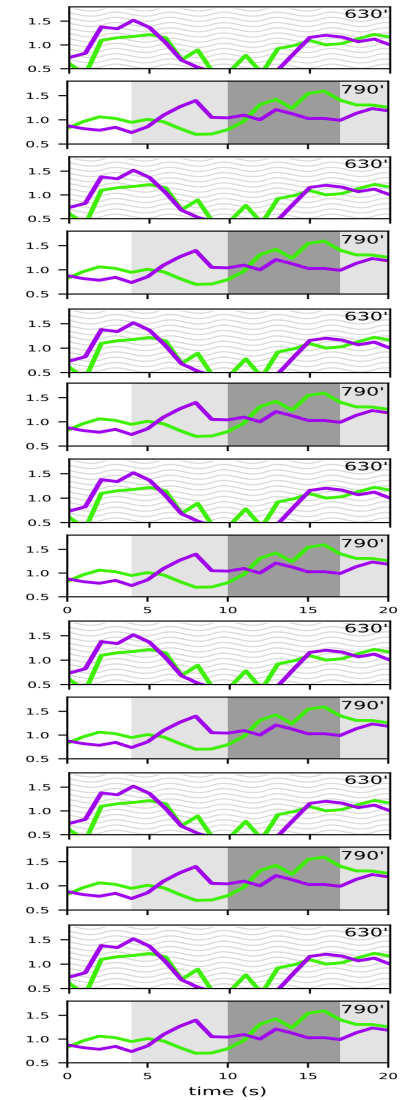
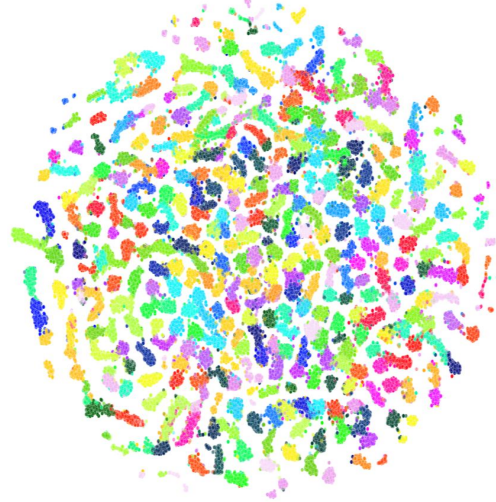


Engineering Challenge

Tracking and Identification

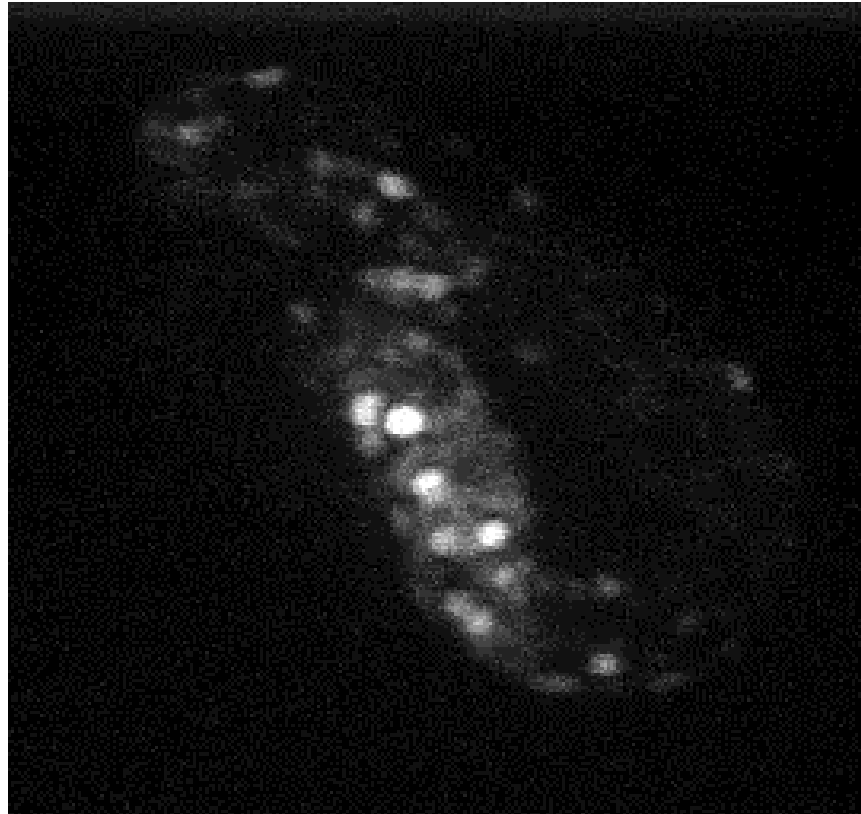


Twin Attention Encoder

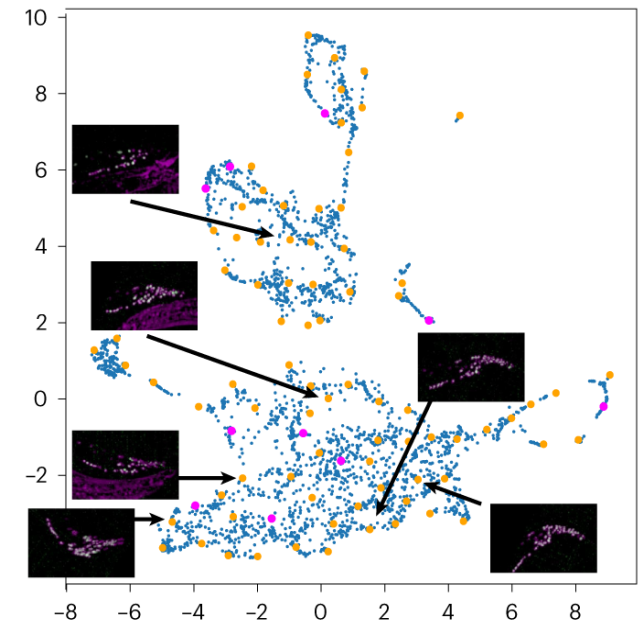


Conceptual Challenge

Posture-Activity Correlation



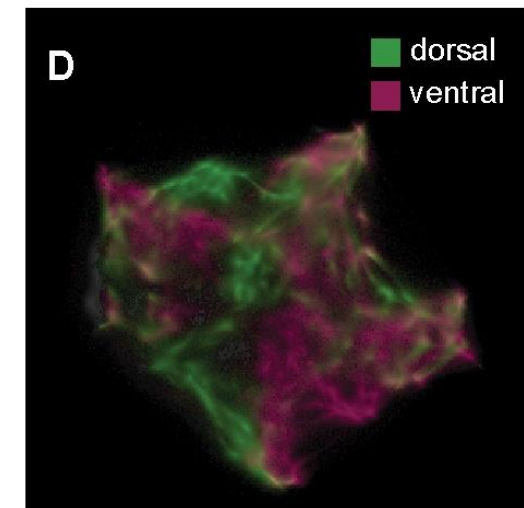
Posture Space



Park et al, Nature Methods 2024

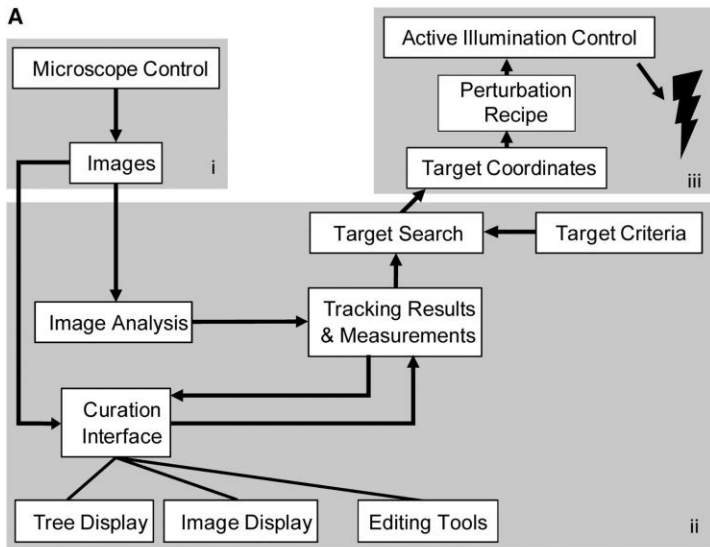
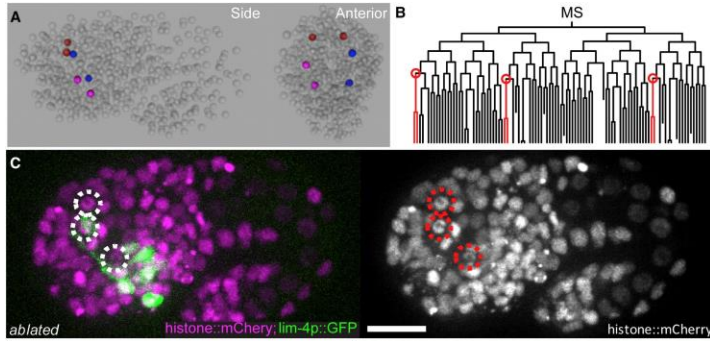


Activity Space



Atanas et al, Cell 2023

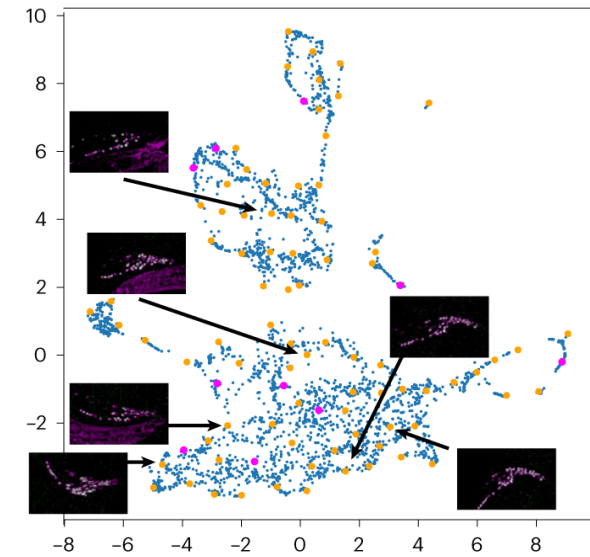
Perturbation



Shah et al, Developmental Cell, 2017

ShootingStar: Identity guided ablation, photoconversion, photoactivation

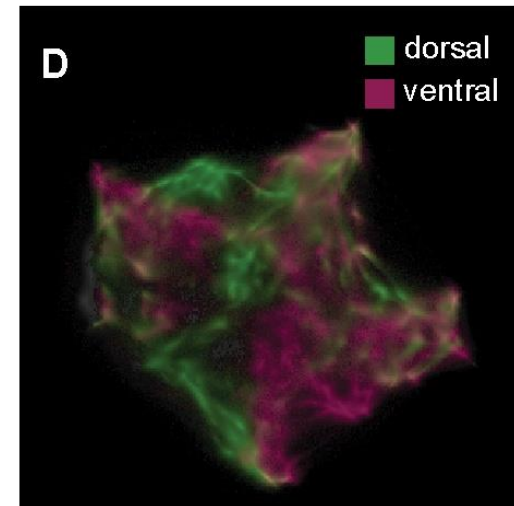
Posture Space



Park et al, Nature Methods 2024



Activity Space



Atanas et al, Cell 2023

All together now

- The fundamental problem underlying image analysis:
 - turn a biological question into a measurement question
- If the question is answerable depends on the interplay of
 - imaging
 - analysis methods
 - tool ergonomics
- This often requires a team, but knowing what is possible informs experimental choices

Questions?