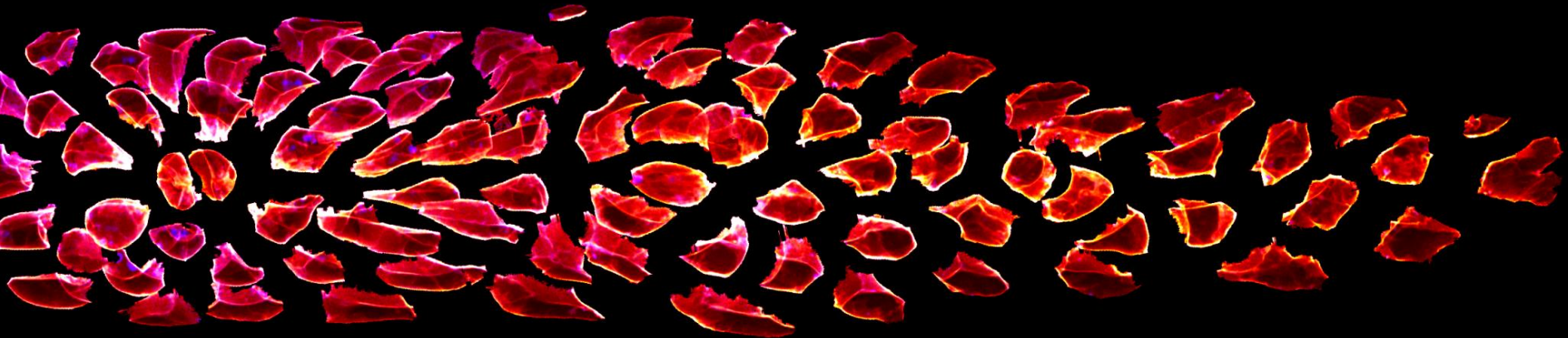


Bonus Lecture! Yay!

EMBL Bio-IT/ALMF Course

Image Analysis with Python 2018

Sessions 3 – 5



Jonas Hartmann

Gilmour group, EMBL Heidelberg

Agenda

- ▶ **Parameter optimization**
- ▶ **Smart microscopy**
- ▶ **Object tracking**
- ▶ **Machine learning for segmentation**
- ▶ **More machine learning**

Parameter Optimization

Parameter Optimization

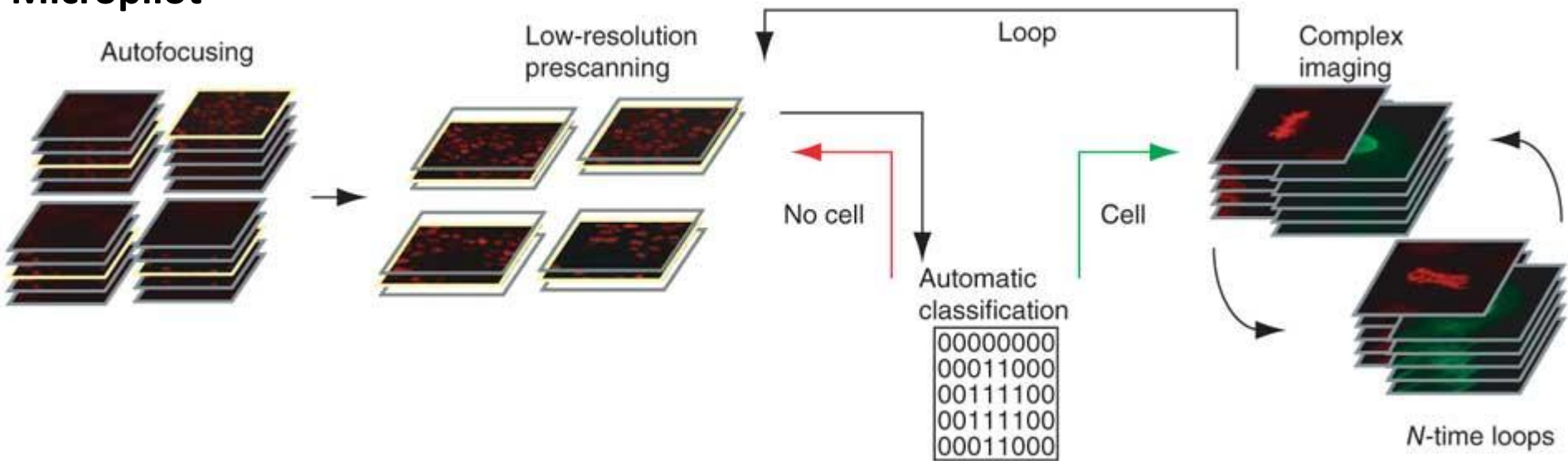
- ▶ Selecting parameters is important but difficult!
 - Examples: *sigma*, *SE size*, *morphology params*, *DT sigma*, *min_distance*, ...
 - Option 1: Use reasoning based on 'limits'
 - Option 2: Trial and error
 - This can be automated!
 1. Create some manual gold standard segmentations
 2. Automatically screen through 100s of parameter combinations
 3. This becomes feasible by using the compute cluster
 4. Tobias Rasse (ALMF) is developing a framework for this at EMBL!

Smart Microscopy

Smart Microscopy

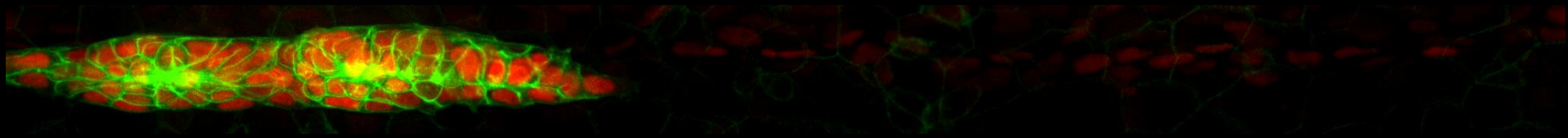
- ▶ Direct coupling of microscopy and image analysis
- ▶ Example 1: Region Of Interest (ROI) detection
 - Identify ROI in low-res high-FOV image
 - Instruct microscope to acquire ROI in high-res

Micropilot

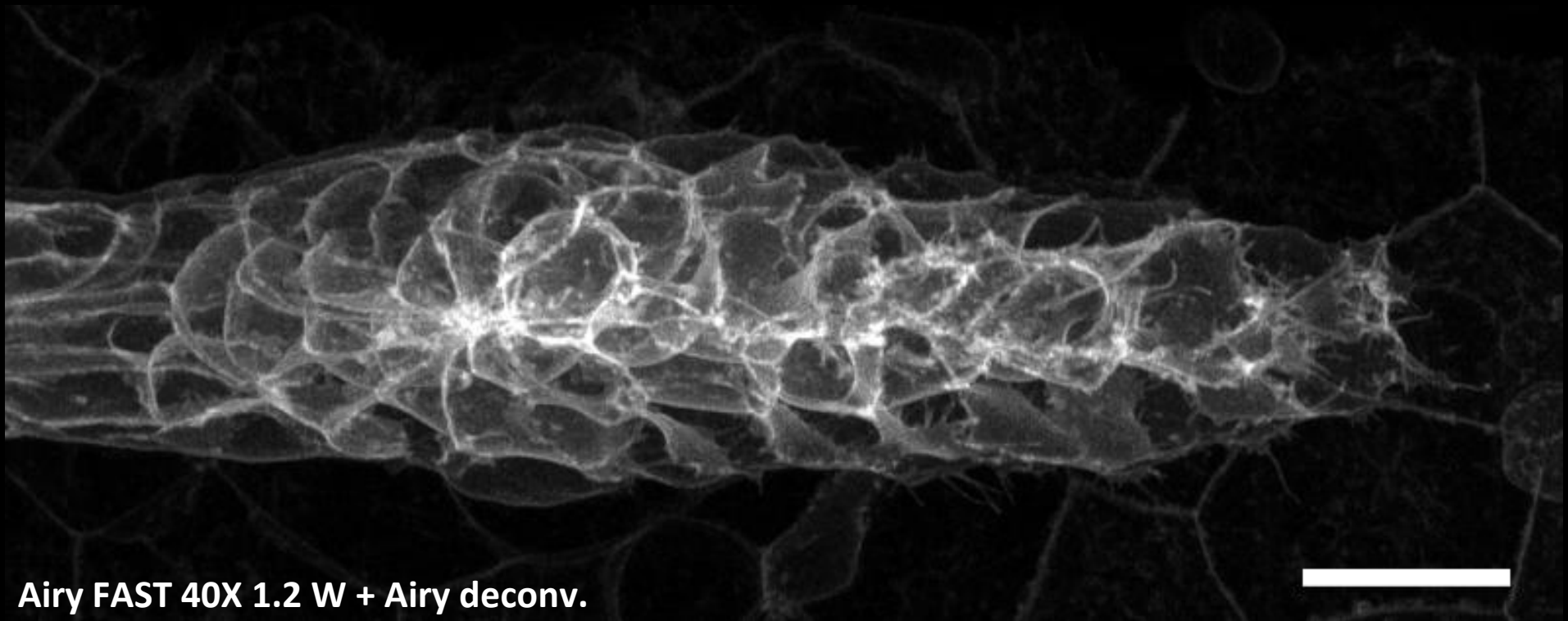


Published 2011 in Nature Methods by Ellenberg & Pepperkok groups

Smart Microscopy



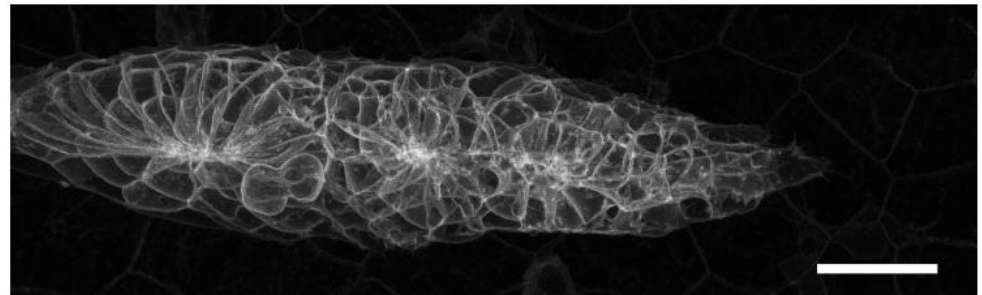
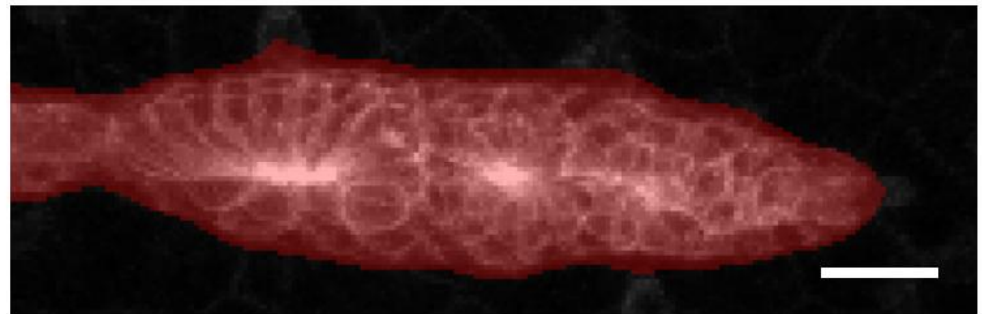
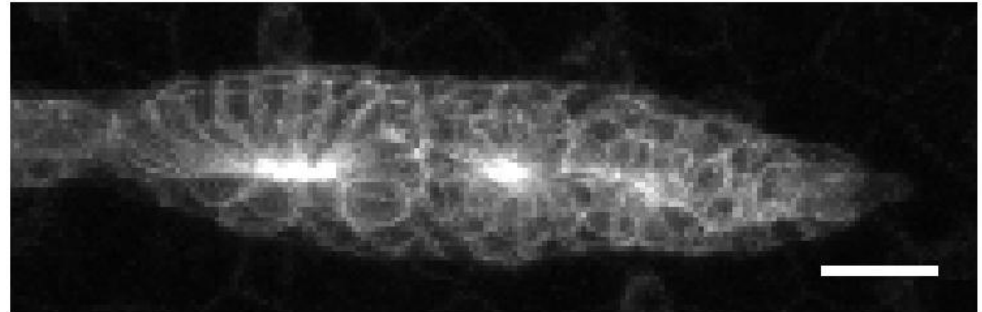
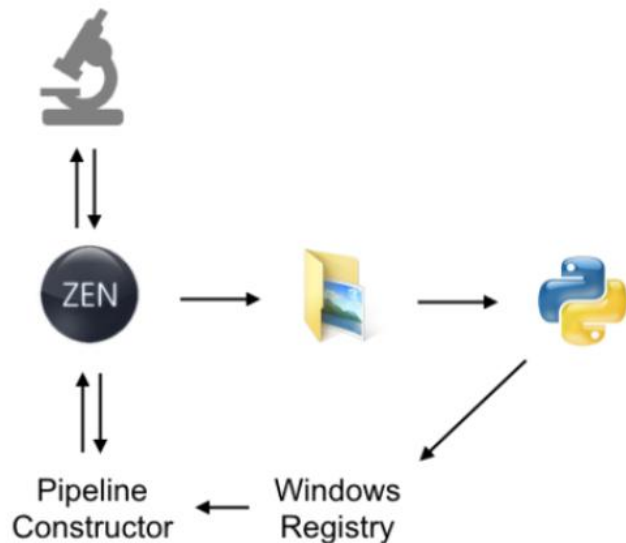
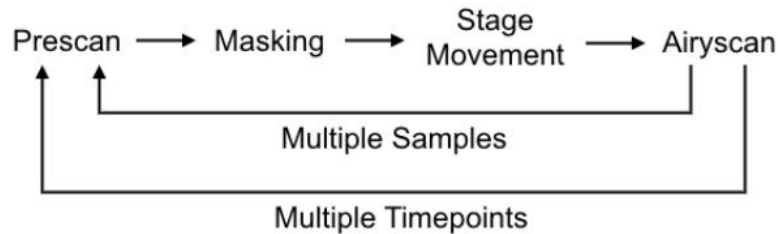
VoX Spinning Disk 40X 1.2 W



Airy FAST 40X 1.2 W + Airy deconv.

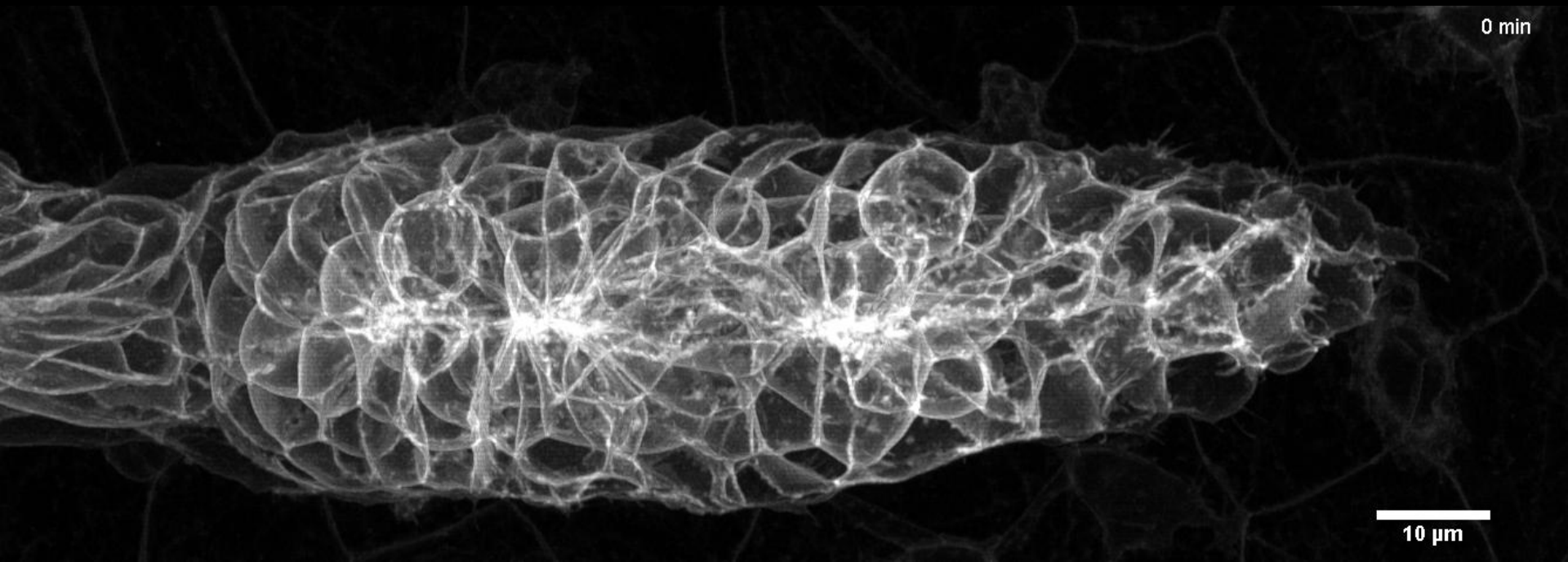
Smart Microscopy

- ▶ Direct coupling of microscopy and image analysis
- ▶ Example 2: Live object tracking



Scale bars: 15μm

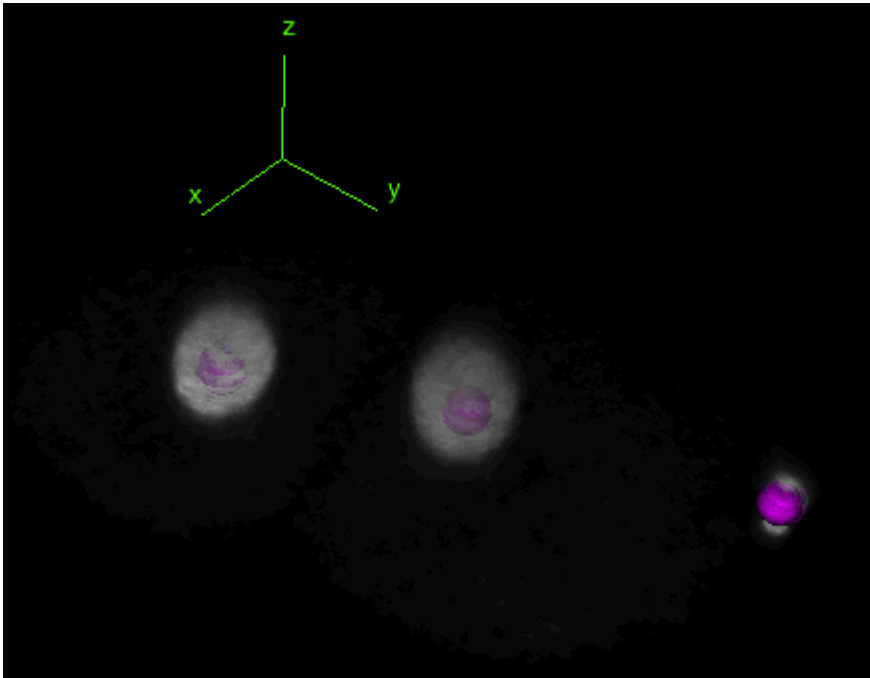
Smart Microscopy



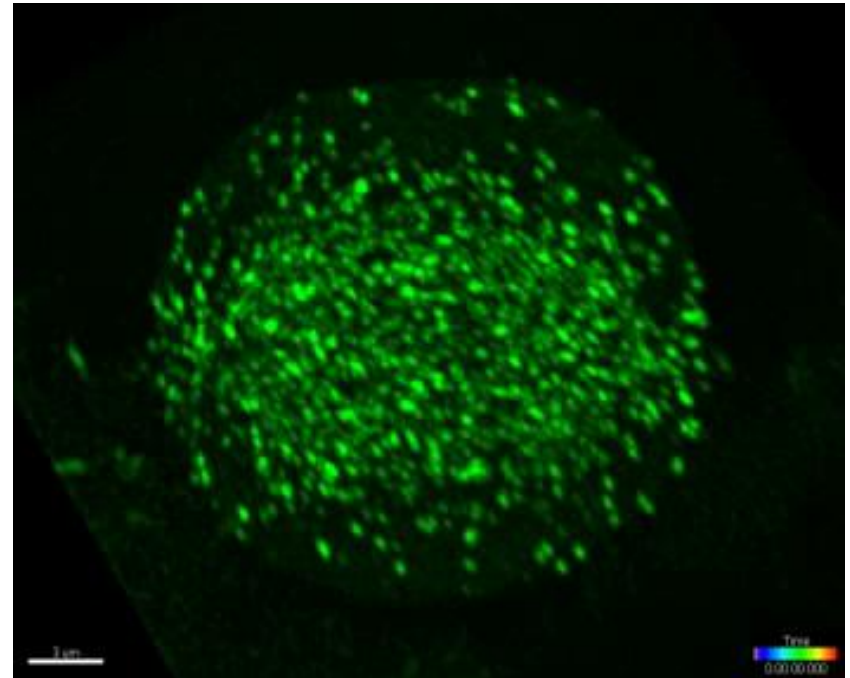
Object Tracking

Object Tracking

- Assign each labeled object at time t_0 a corresponding object at t_1
 - Approach: segment independently, then link objects (by optimization)
 - Linking based on: space, movement, object properties
 - Easier for small δt
 - Challenges: poor segmentation, overlapping objects, dividing/merging objects



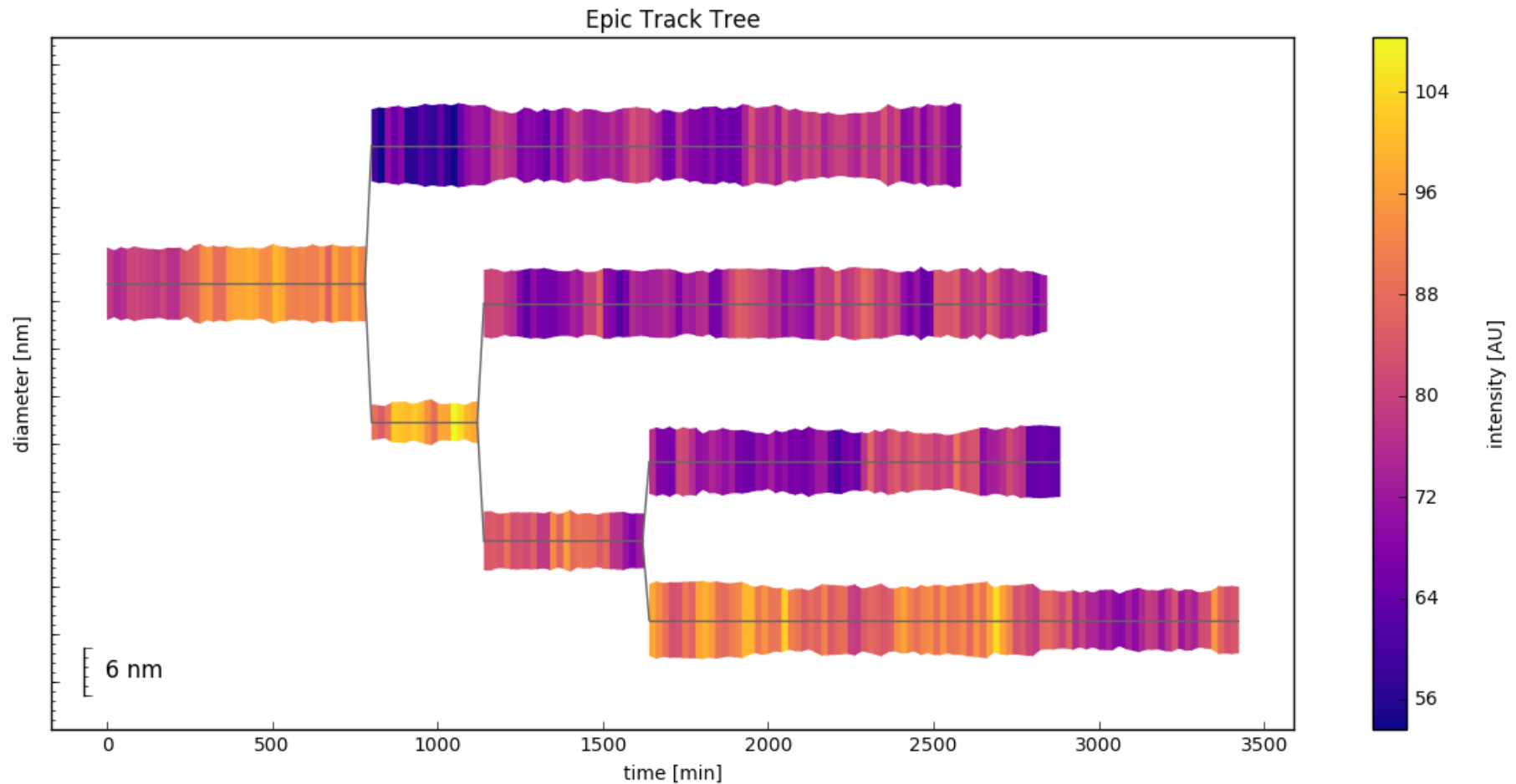
From Fiji's TrackMate plugin



From Eric Betzig's group

Object Tracking

- Assign each labeled object at time t_0 a corresponding object at t_1



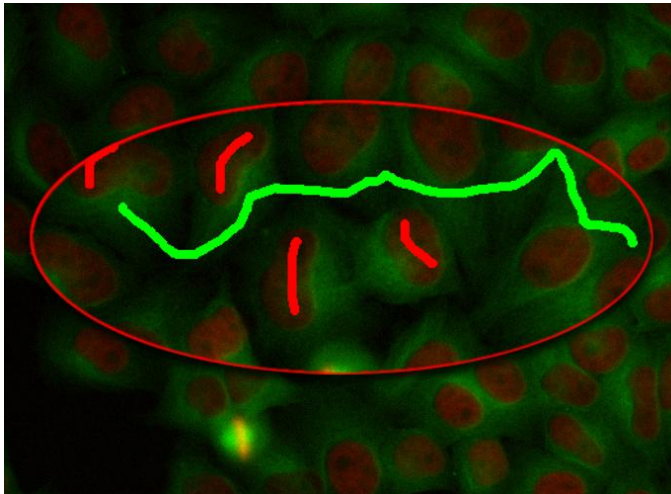
Machine Learning for Segmentation

Machine Learning for Segmentation

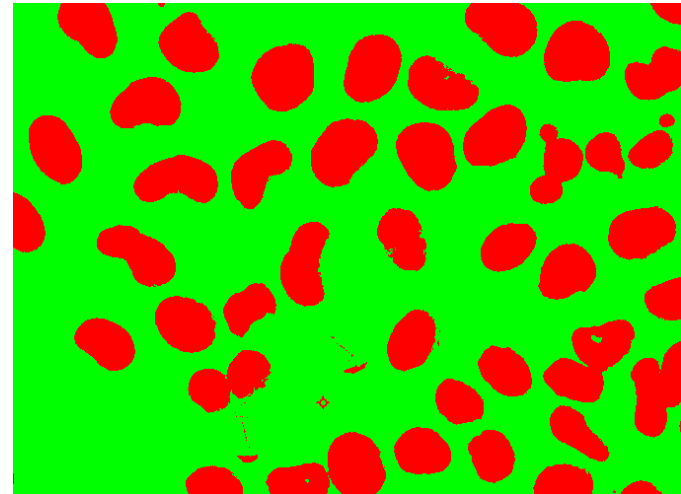
- ▶ Goal: foreground-background detection by machine learning
 - This is a classification task: classify pixels into groups
 - Approach: supervised learning
 - Manually label example pixels
 - Extract features (intensity, neighborhood, ...)
 - Train classifier, e.g. random forest
 - Make prediction for all other pixels



Ilastik



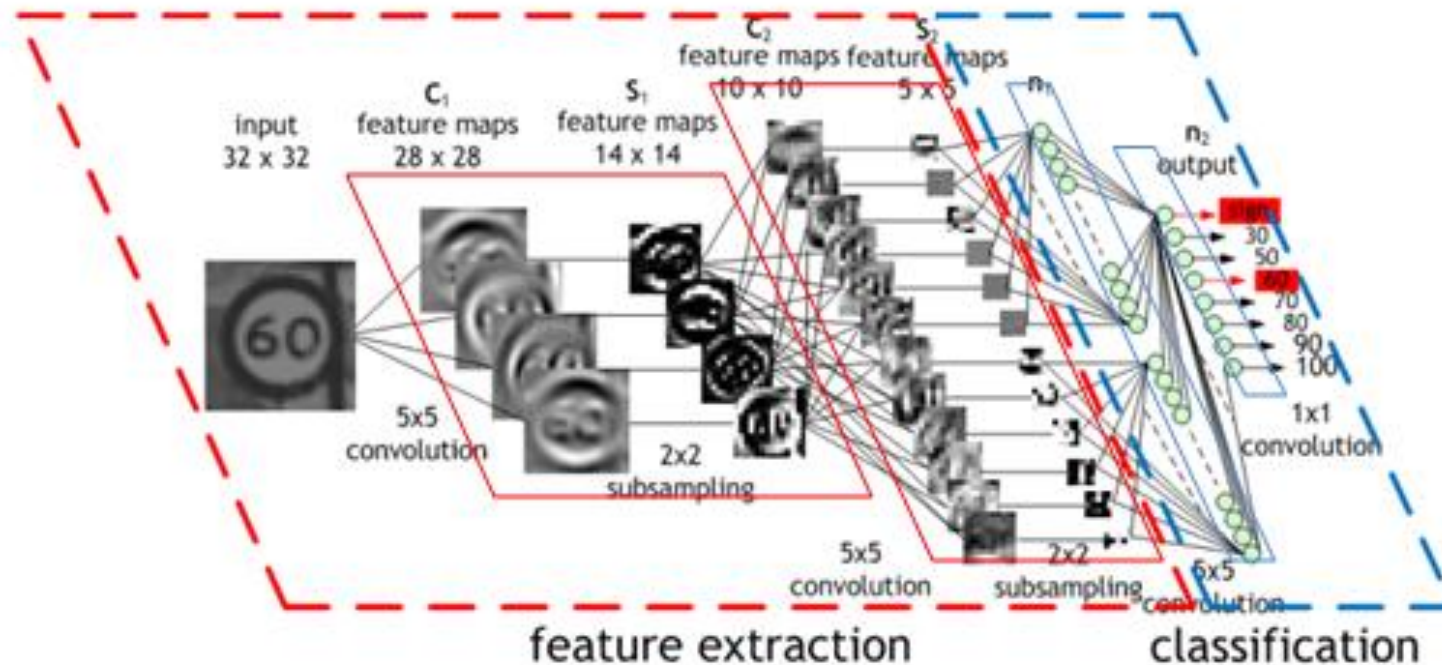
From the Ilastik website



Also useful as `preprocessing` (use probabilities)!

Machine Learning for Segmentation

- ▶ Machine learning in python
 - scikit-learn (sklearn)
- ▶ Deep Learning
 - Keras, Tensorflow, PyTorch



Similar approach (but without neural networks): speak to [Christian Tischer, ALMF](#)

Machine Learning for Segmentation

Machine Learning for Image Analysis

Date: Monday 29 - Wednesday 31 October 2018

Venue: EMBL - EMBL- Heidelberg, Meyerhofstraße, 69117, Heidelberg, Germany

Application opens: Monday May 07 2018

Application deadline: Friday June 15 2018

Participation: Open application with selection

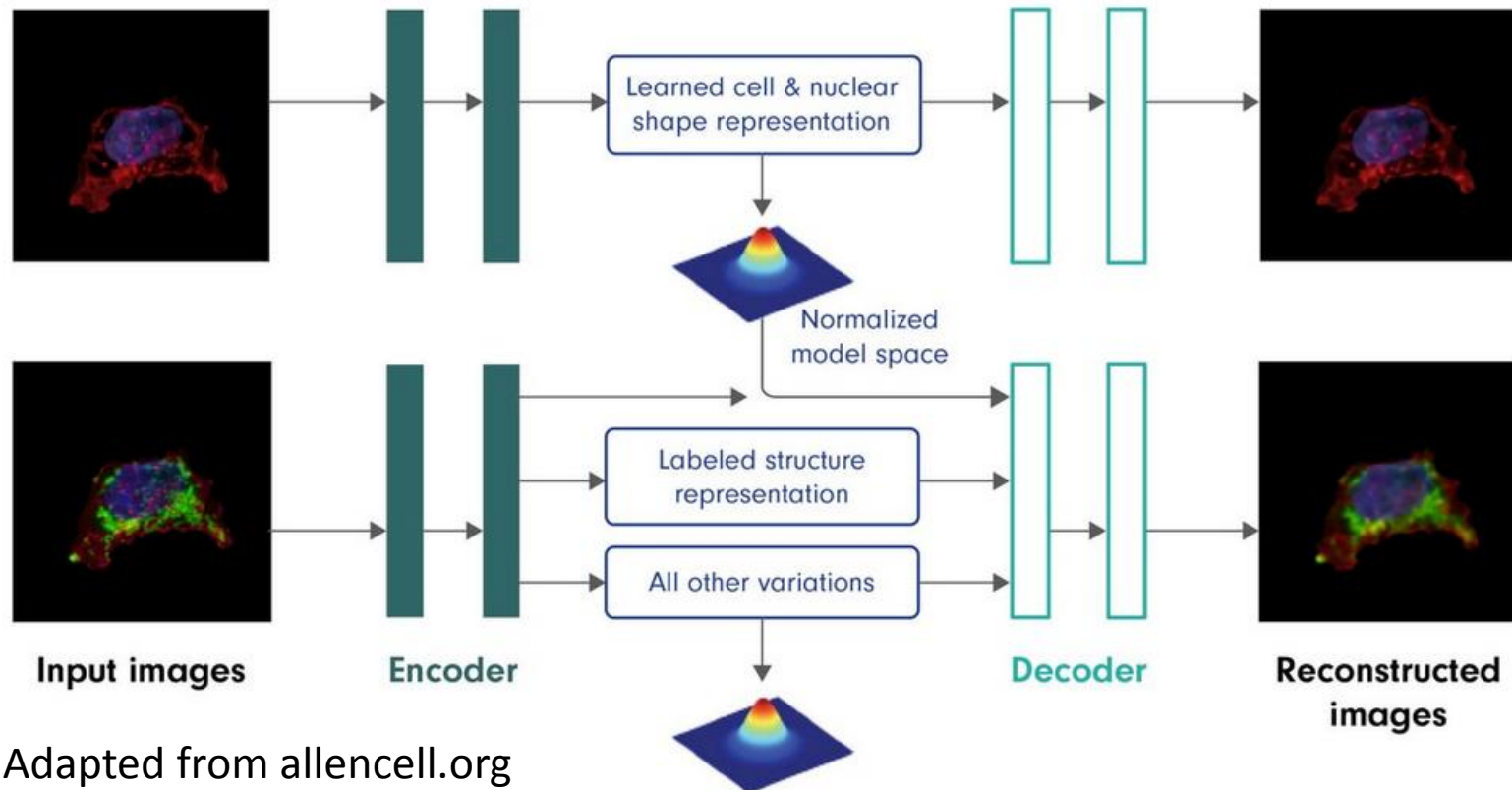
This is a blended learning course on Machine Learning for Image Analysis, consisting of three online sessions with associated hands-on exercises prior to the workshop, a three day face-to-face workshop at EMBL Heidelberg and two optional online sessions with associated hands-on exercises after the workshop.

More Machine Learning

More Machine Learning

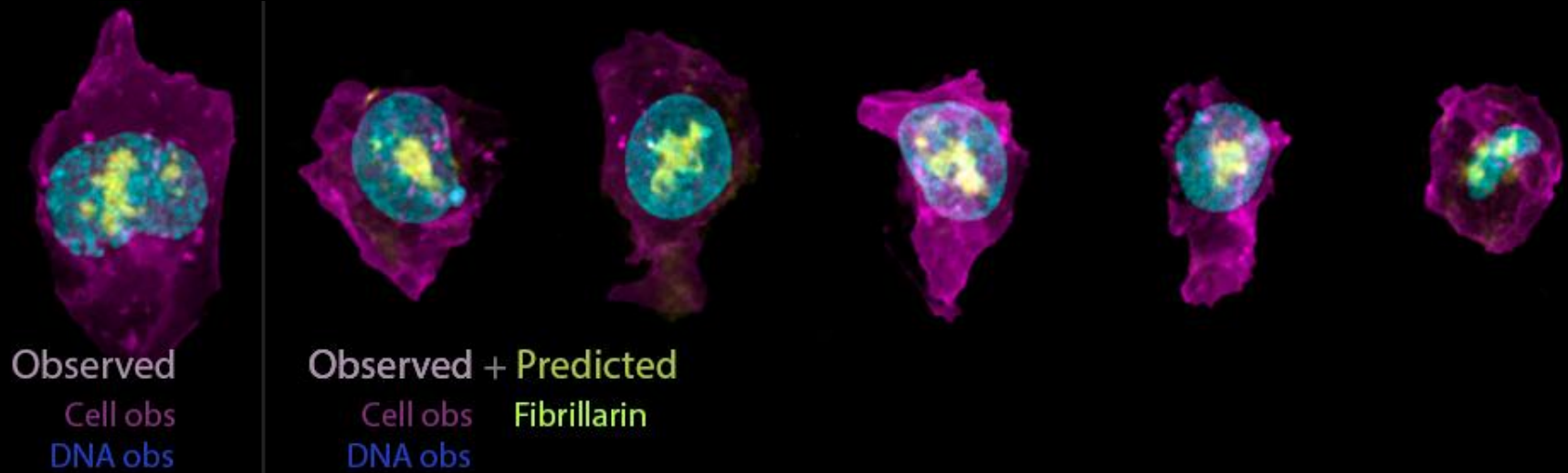
► Deep neural network autoencoder

- First Learn to reconstruct mem+nuc image from more sparse representation
- Next learn to reconstruct other channels based on mem+nuc



More Machine Learning

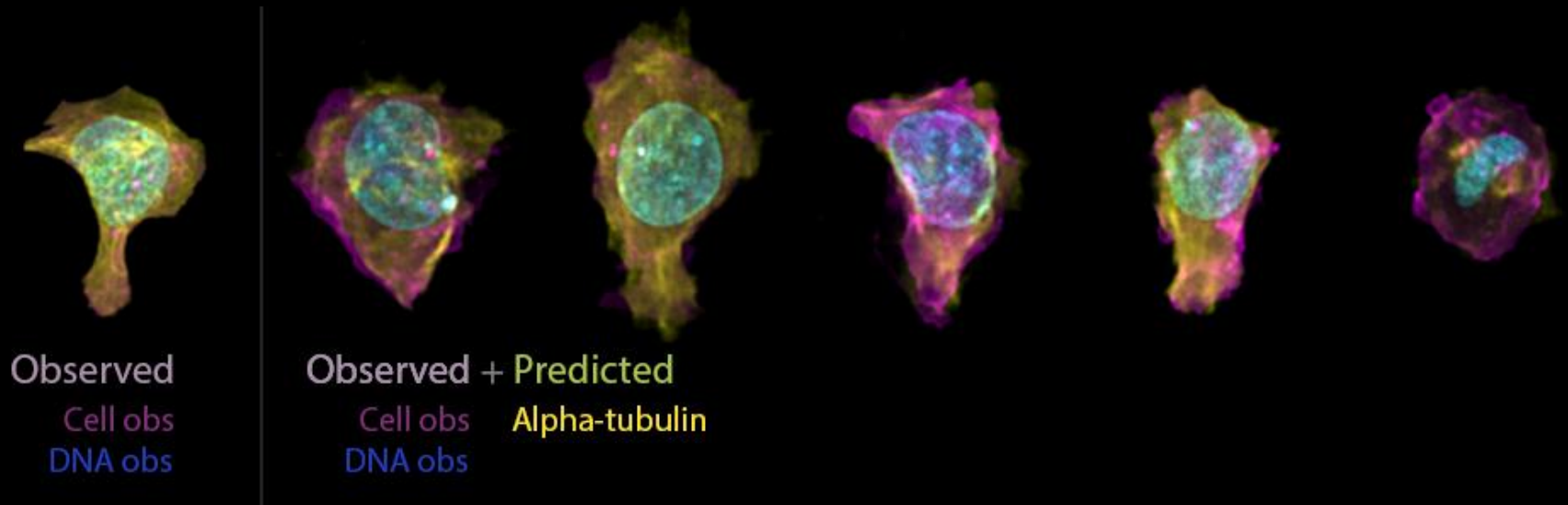
► Deep neural network autoencoder



Adapted from allencell.org

More Machine Learning

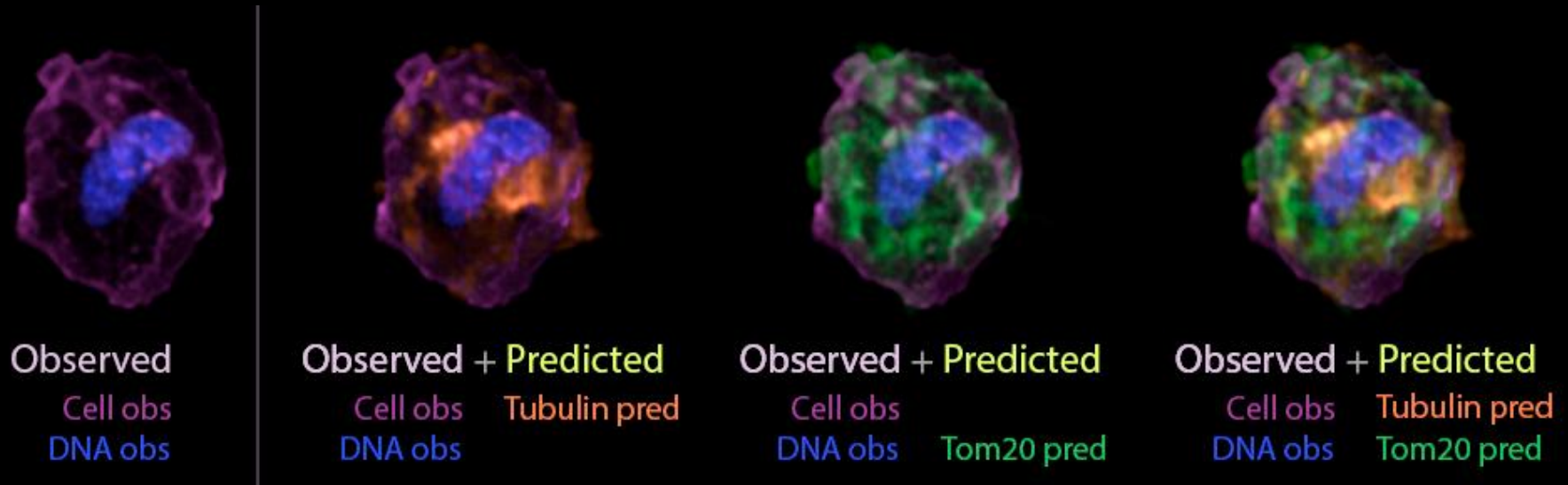
► Deep neural network autoencoder



Adapted from allencell.org

More Machine Learning

► Deep neural network autoencoder



Adapted from allencell.org

- **Advantages:** construction of 'atlases' without feature engineering!
- **Disadvantages:** encoded model is a 'black box'

That's all!

Thanks!

:)