# **Bonus Lecture! Yay!**

EMBL Bio-IT/ALMF Course

Image Analysis with Python 2018

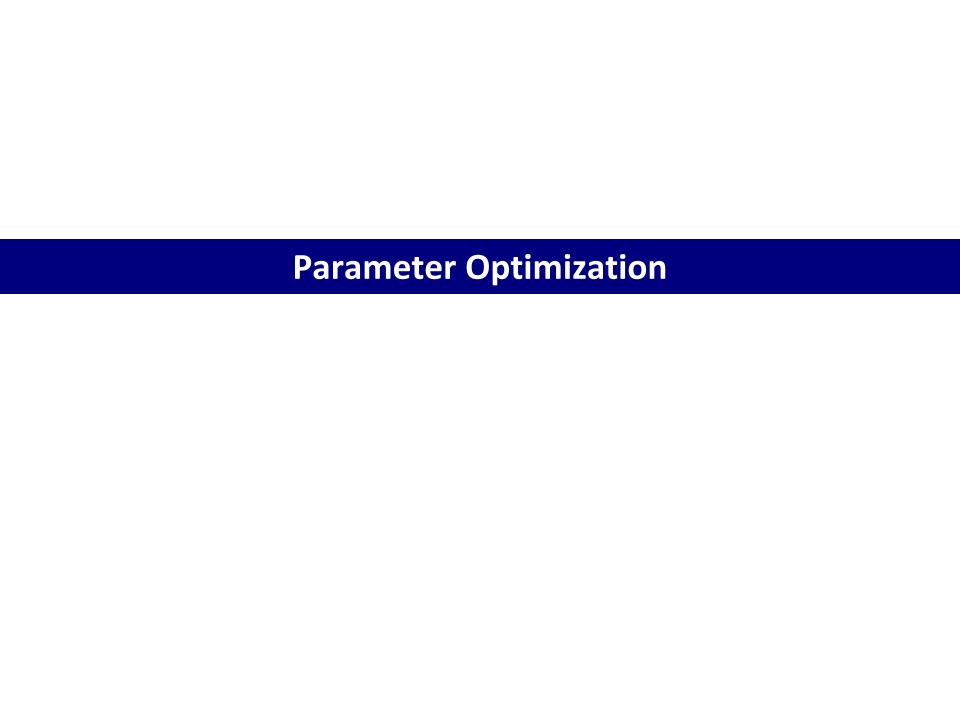
Sessions 3 – 5



Jonas Hartmann
Gilmour group, EMBL Heidelberg

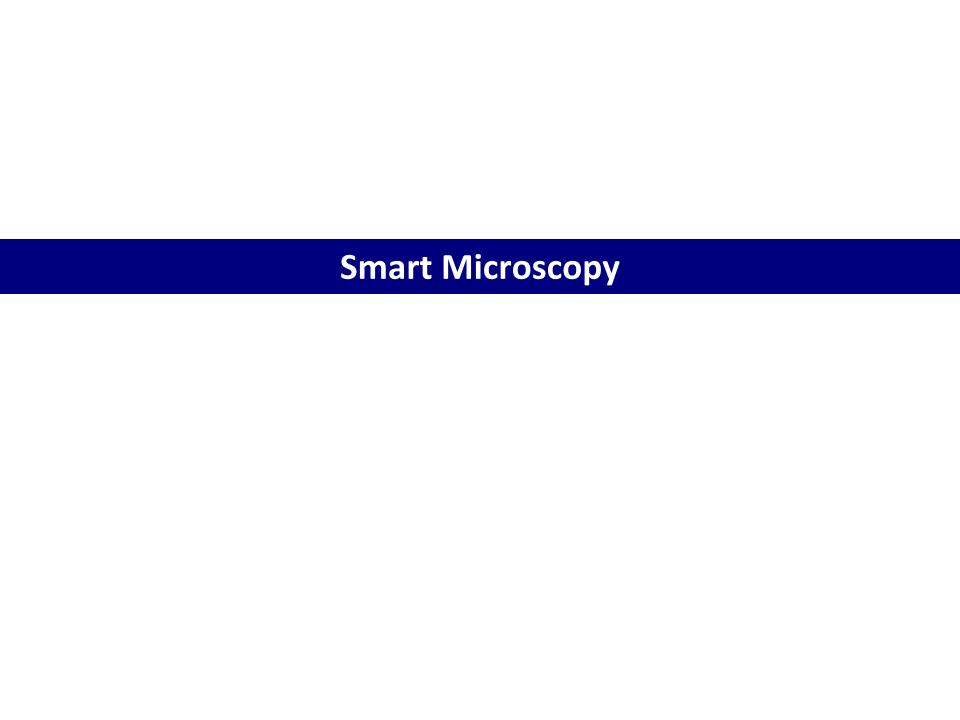
## **Agenda**

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

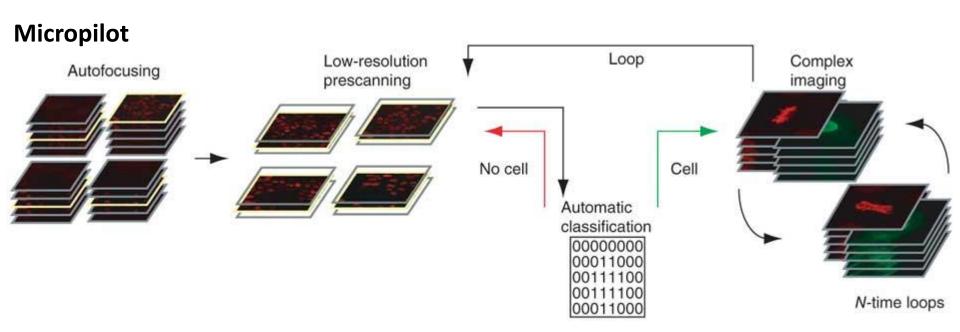


#### **Parameter Optimization**

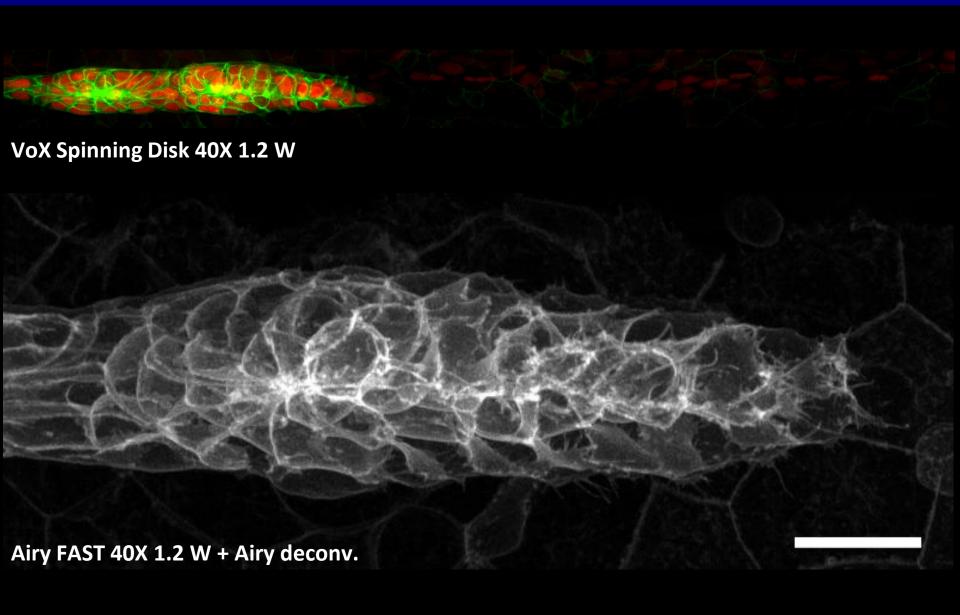
- Selecting parameters is important but difficult!
  - Examples: sigma, SE size, morphology params, DT sigma, min\_distance, ...
  - Option 1: Use reasoning based on 'limits'
  - Option 2: Trial and error
    - This can be automated!
      - 1. Create some manual gold standard segmentations
      - 2. Automatically screen through 100s of parameter combinations
      - 3. This becomes feasible by using the compute cluster!



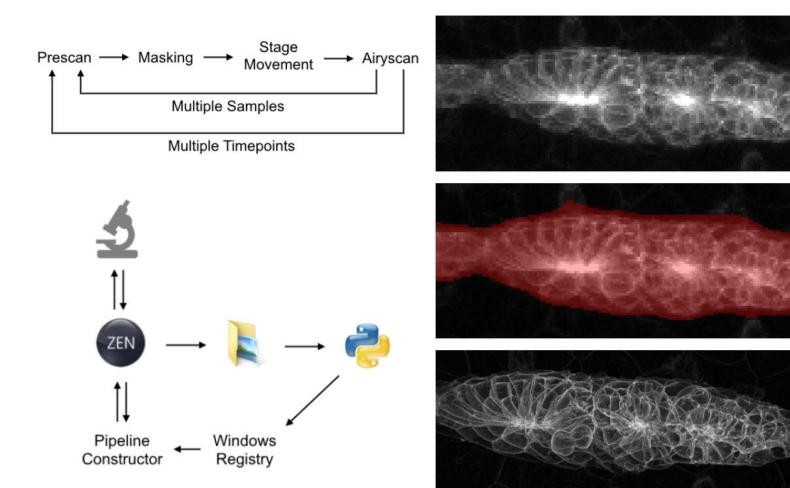
- Direct coupling of microscopy and image analysis
- ► Example 1: Region Of Interest (ROI) detection
  - Identify ROI in low-res high-FOV image
  - Instruct microscope to acquire ROI in high-res



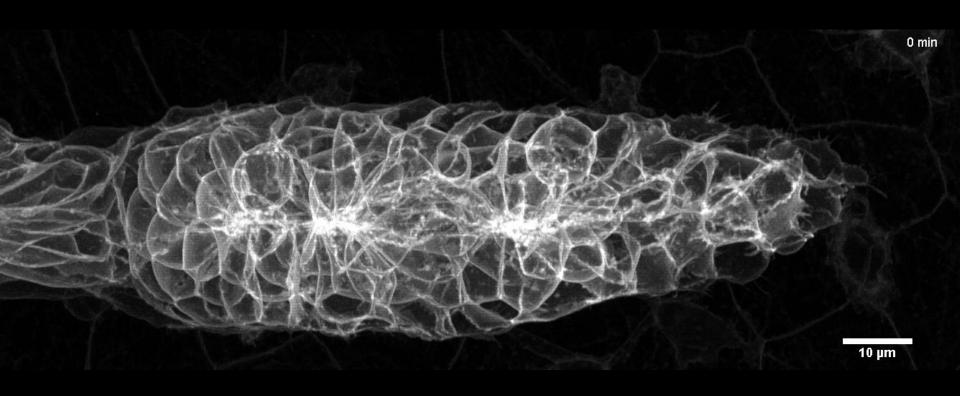
Published 2011 in Nature Methods by Ellenberg & Pepperkok groups



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



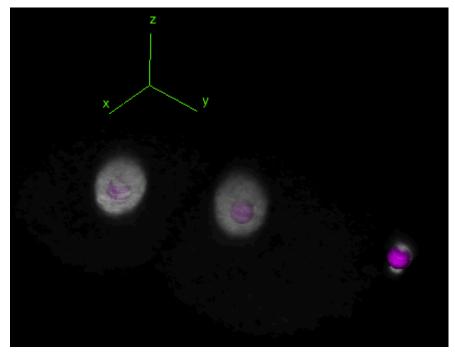
Scale bars: 15µm



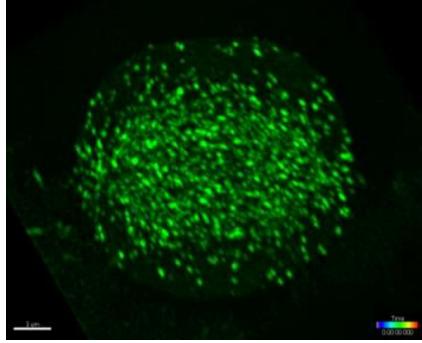


#### **Object Tracking**

- Assign each labeled object at time to a corresponding object at t1
  - 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



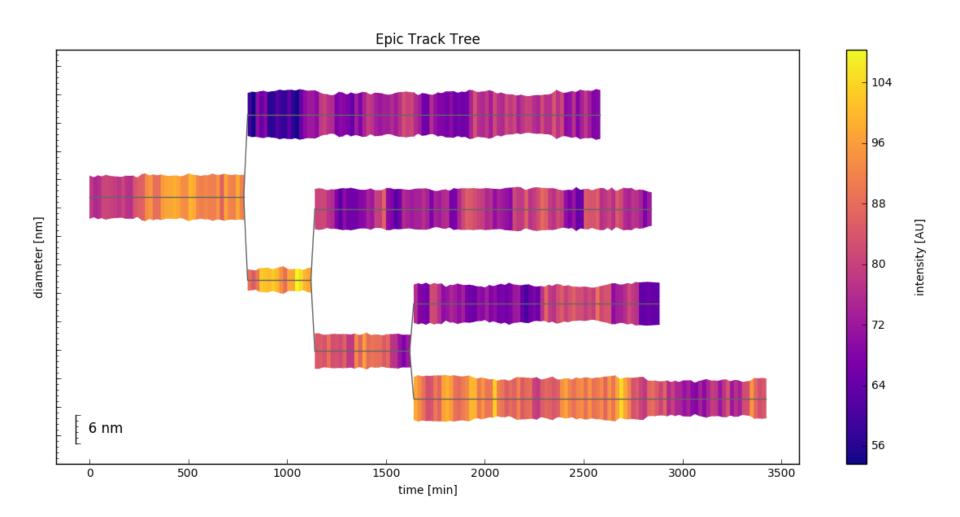
From Fiji's TrackMate plugin

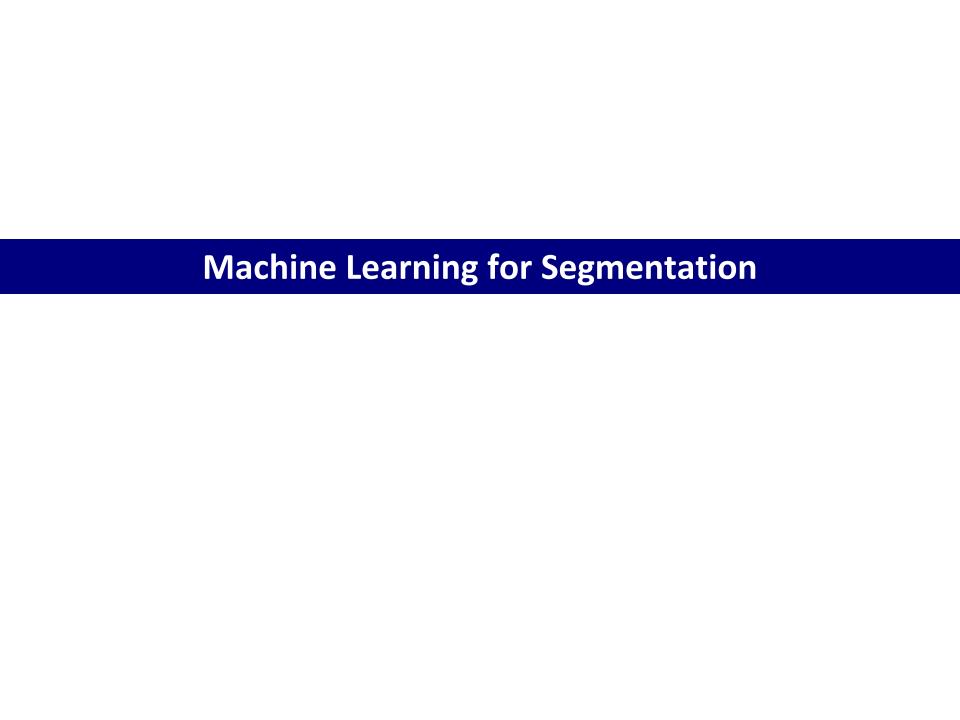


From Eric Betzig's group

## **Object Tracking**

► Assign each labeled object at time t0 a corresponding object at t1

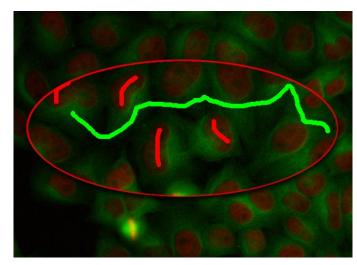


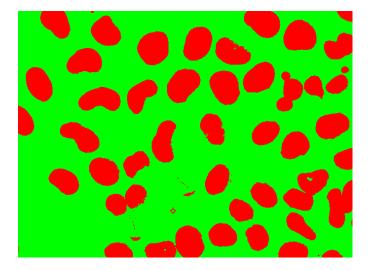


## **Machine Learning for Segmentation**

- 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



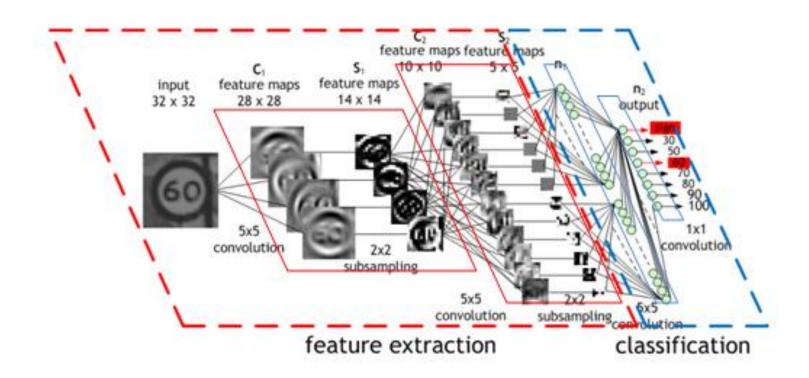




From the Ilastik website

## **Machine Learning for Segmentation**

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



#### **Machine Learning for Segmentation**

# Machine Learning for Image Analysis

Date: Monday 29 - Wednesday 31 October 2018

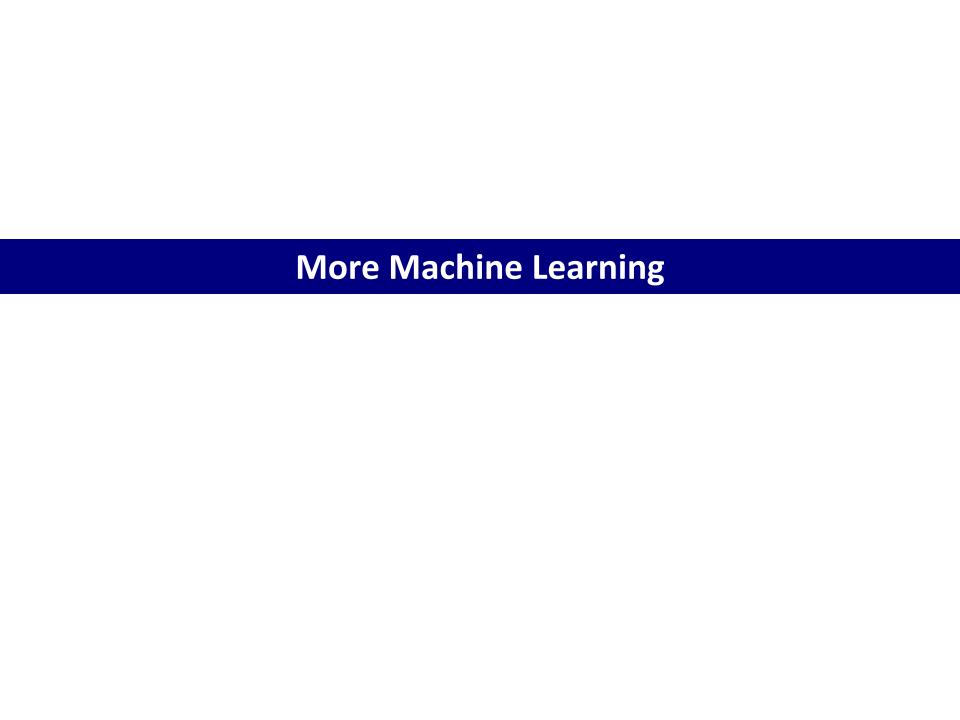
Venue: EMBL - EMBL- Heidelberg, Meyerhofstraße, 69117, Heidelberg, Germany

Application opens: Monday May 07 2018

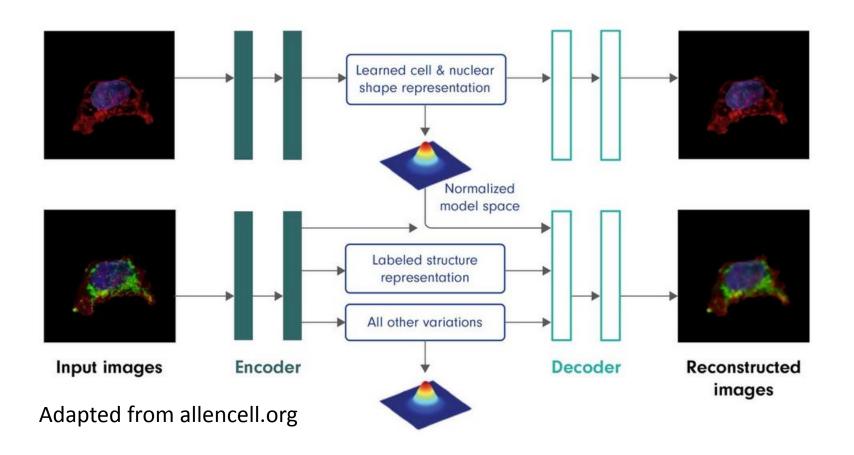
Application deadline: Friday June 15 2018

Participation: Open application with selection

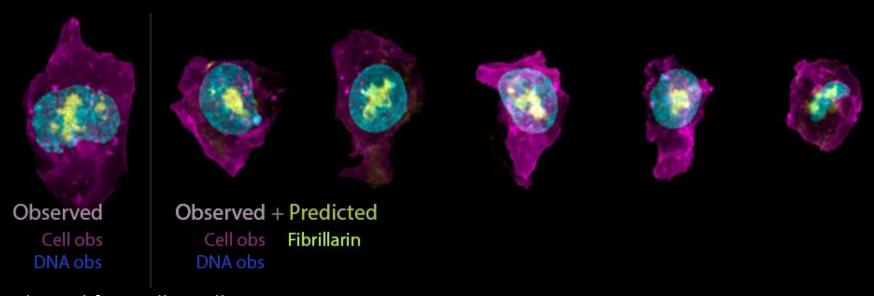
This is a blended learning course on Machine Learning for Image Analysis, consisting of three online sessions with associated hands-on exercises prior to the workshop, a three day face-to-face workshop at EMBL Heidelberg and two optional online sessions with associated hands-on exercises after the workshop.



- 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

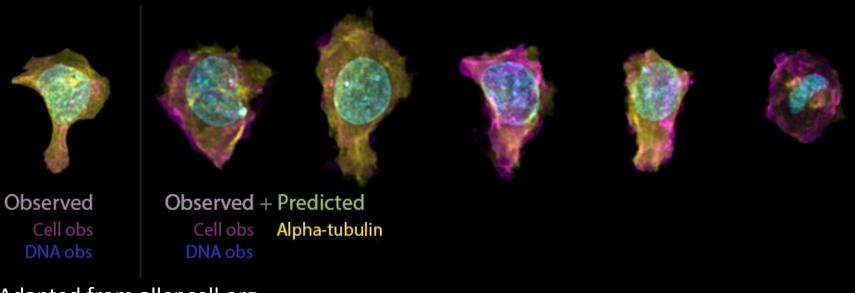


Deep neural network autoencoder



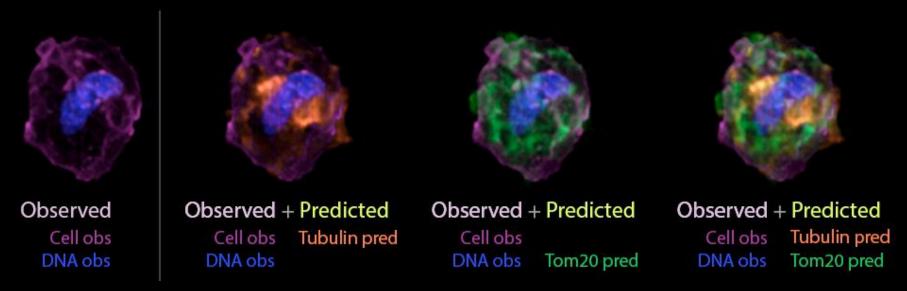
Adapted from allencell.org

Deep neural network autoencoder



Adapted from allencell.org

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!