

# Introduction to Machine Learning

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IAFIG-RMS - Bioimage Analysis With Python  
Cambridge Bioinformatics Training Centre



# Is it just statistics?

- different aim: explanation vs prediction
- effective engineering to allow for mathematical description of environment.
- If it works – it works.

# What is machine “learning”?

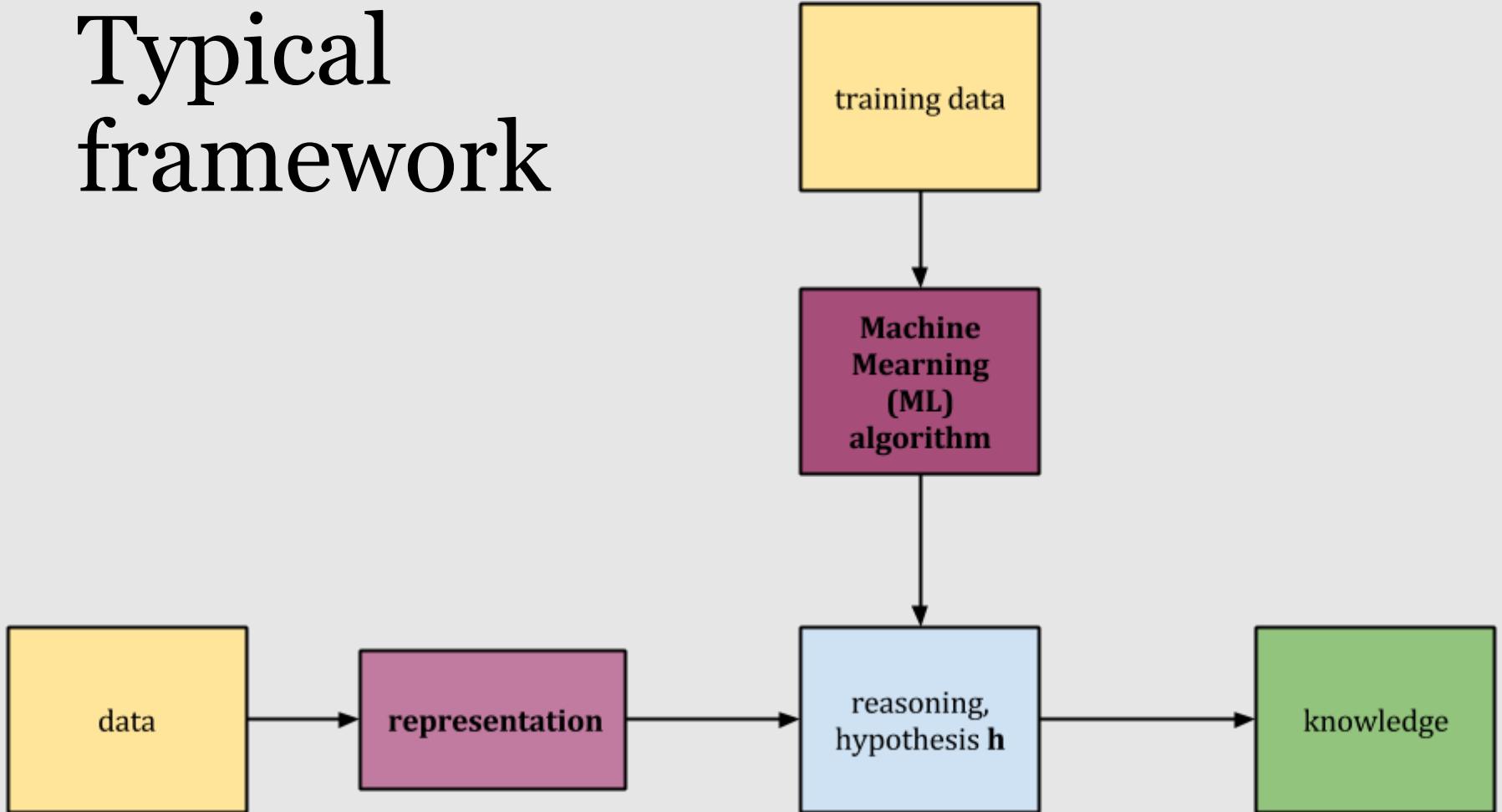
*“A computer program is said to **learn from experience E** with respect to some **task T** and some **performance measure P**, if its performance on T, as measured by P, improves with experience E.”*

Tom Mitchel, 1998

# Types of ML

- Supervised – **task T** with ground-truth labels
  - goal: predict label from data
- Unsupervised – **task T** without labels
  - goal: find groupings and patterns in data
- Reinforcement Learning – **task T** is exploratory
  - goal: optimise processes
  - often RL elements are parts of Supervised and Unsupervised algorithms
  - often agent based (e.g. “game of life”)

# Typical framework



# Machine Learning areas

- **Big Data** – allowing us, humans to learn from data otherwise too sparse or huge to analyse with ordinary statistics .
- **Patter Recognition:** Exploring possible correlations in data which are elusive to ordinary analysis.
- **Computer Vision:** Understanding of visual data (video understanding, object detection, tracking, etc.)
- **Robotics:** physical presence of computers in the real world

# Types of Computer Vision problems:

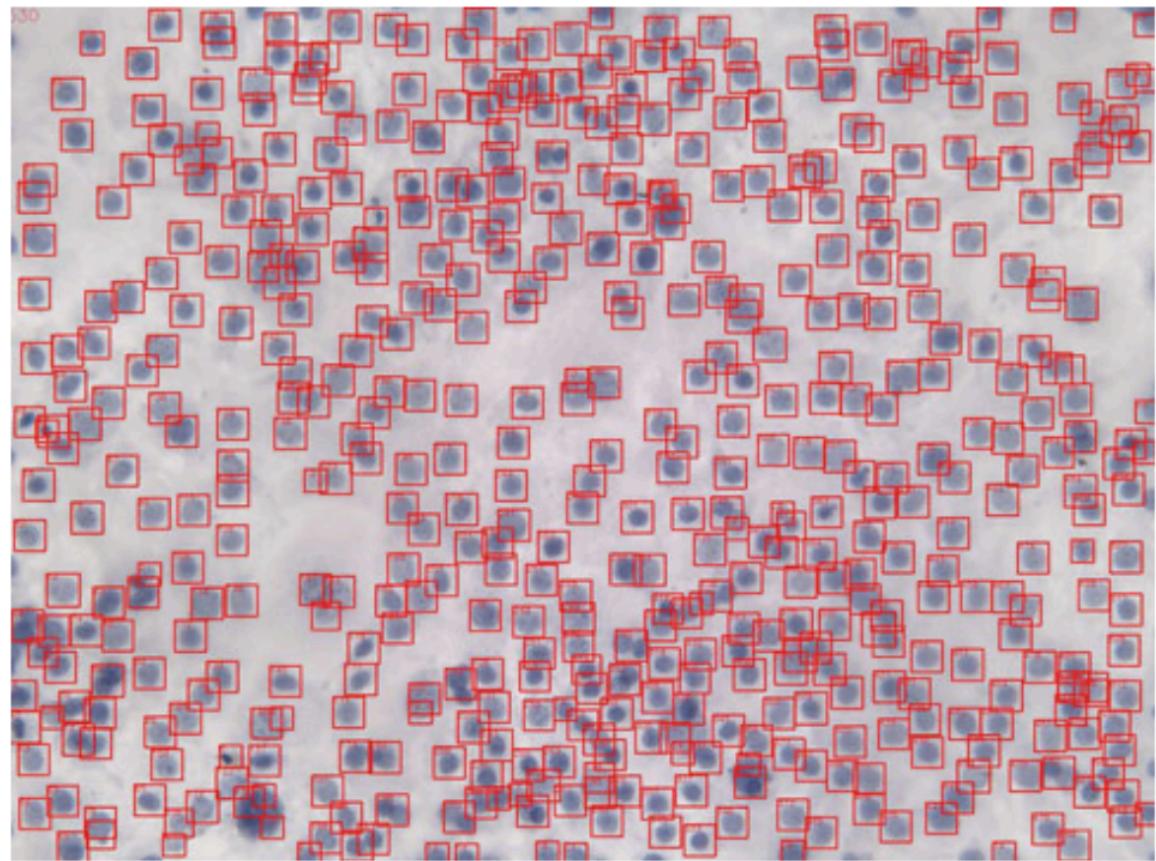
1. classification (binary and non-binary):
  - image recognition
2. regression
  - object localisation
  - object counting
3. association / clustering
  - texture clustering (object segmentation)
  - data mining
4. combination:
  - object detection (localisation + recognition)
  - object tracking (localisation + recognition + association)

# Examples

# Counting via detection

**The application of support vector machine classification to detect cell nuclei for automated microscopy**

Ji Wan Han · Toby P. Breckon · David A. Randell ·  
Gabriel Landini



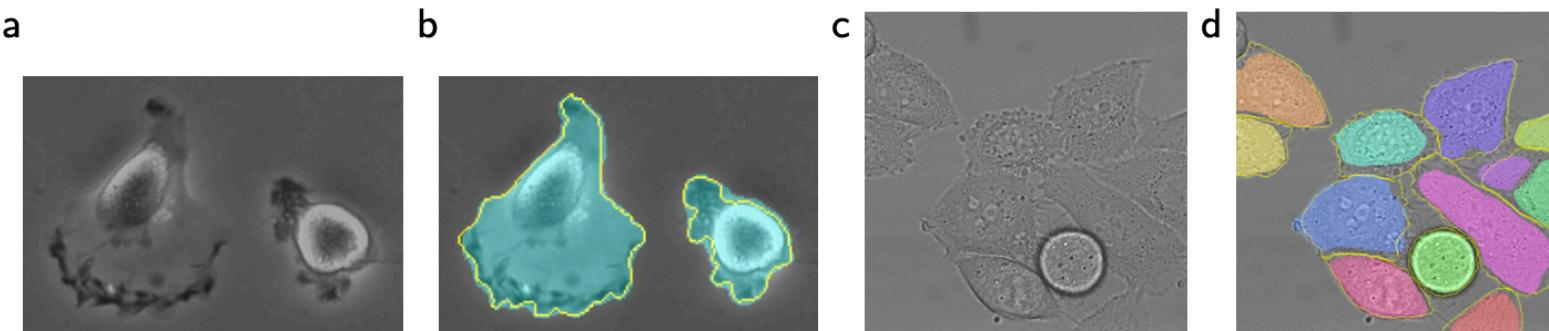
**Fig. 9** Detection of 3t3-H-20× cell nuclei

# Segmentation

## U-Net: Convolutional Networks for Biomedical Image Segmentation

Olaf Ronneberger, Philipp Fischer, and Thomas Brox

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WWW home page: <http://lmb.informatik.uni-freiburg.de/>



**Fig. 4.** Result on the ISBI cell tracking challenge. (a) part of an input image of the “PhC-U373” data set. (b) Segmentation result (cyan mask) with manual ground truth (yellow border) (c) input image of the “DIC-HeLa” data set. (d) Segmentation result (random colored masks) with manual ground truth (yellow border).

# Object (phenotype) recognition



ARTICLE SERIES: Imaging

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## Machine learning in cell biology – teaching computers to recognize phenotypes

Christoph Sommer and Daniel W. Gerlich\*

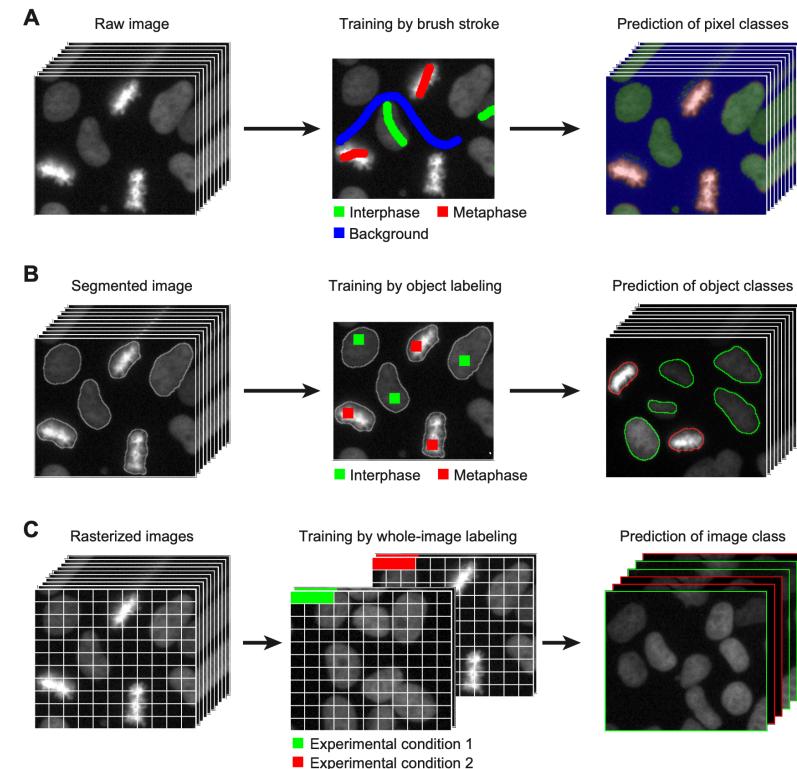
Institute of Molecular Biotechnology of the Austrian Academy of Sciences (IMBA), 1030 Vienna, Austria

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# Object (protein) detection



Featured Prediction Competition

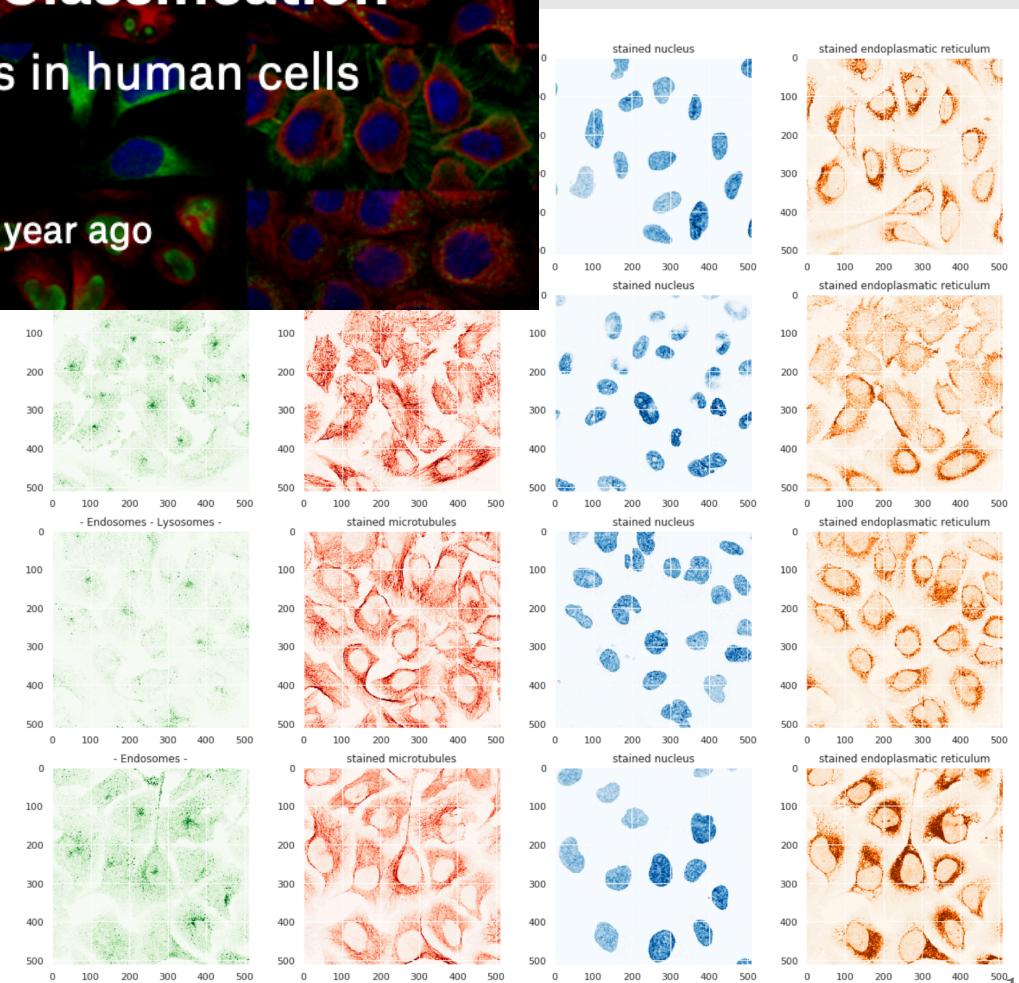
## Human Protein Atlas Image Classification

Classify subcellular protein patterns in human cells



Human Protein Atlas · 2,169 teams · a year ago

Predicting protein  
organelle localization  
labels for each sample.



# Deep Learning on Microscopy Imaging

## Detecting Good, Bad and Ugly Cells with Deep Learning

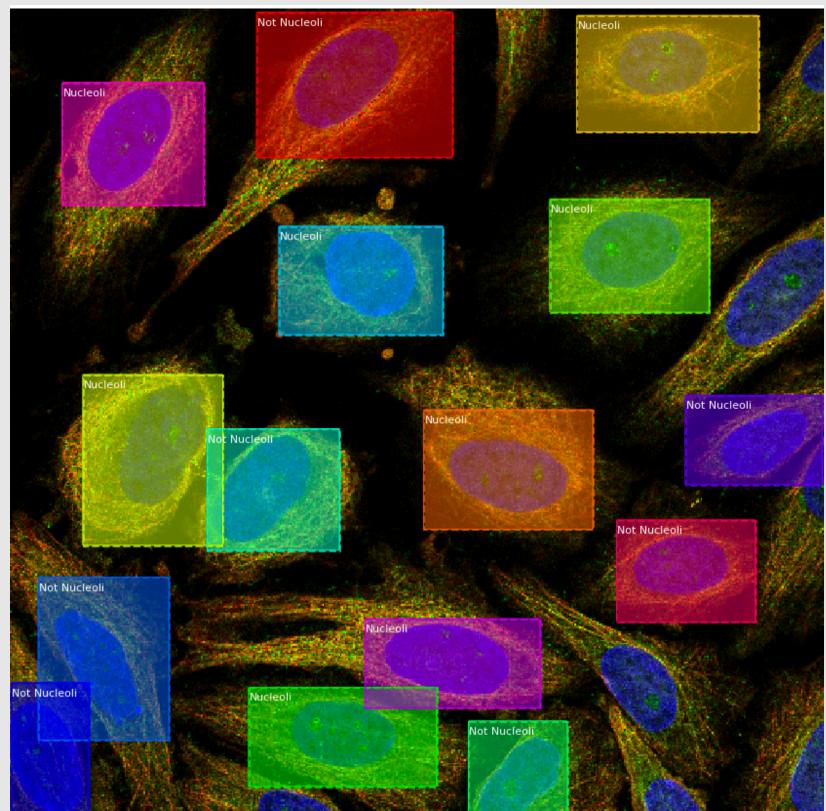
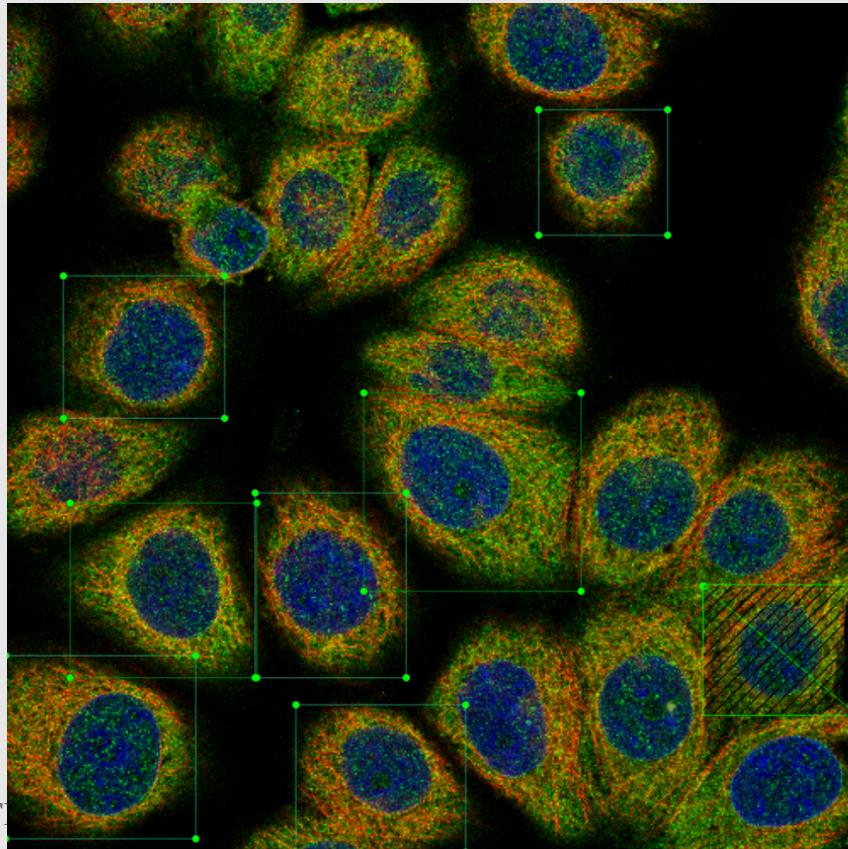


Nikolay Oskolkov

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<https://towardsdatascience.com/deep-learning-on-microscopy-imaging-865b521ec47c>

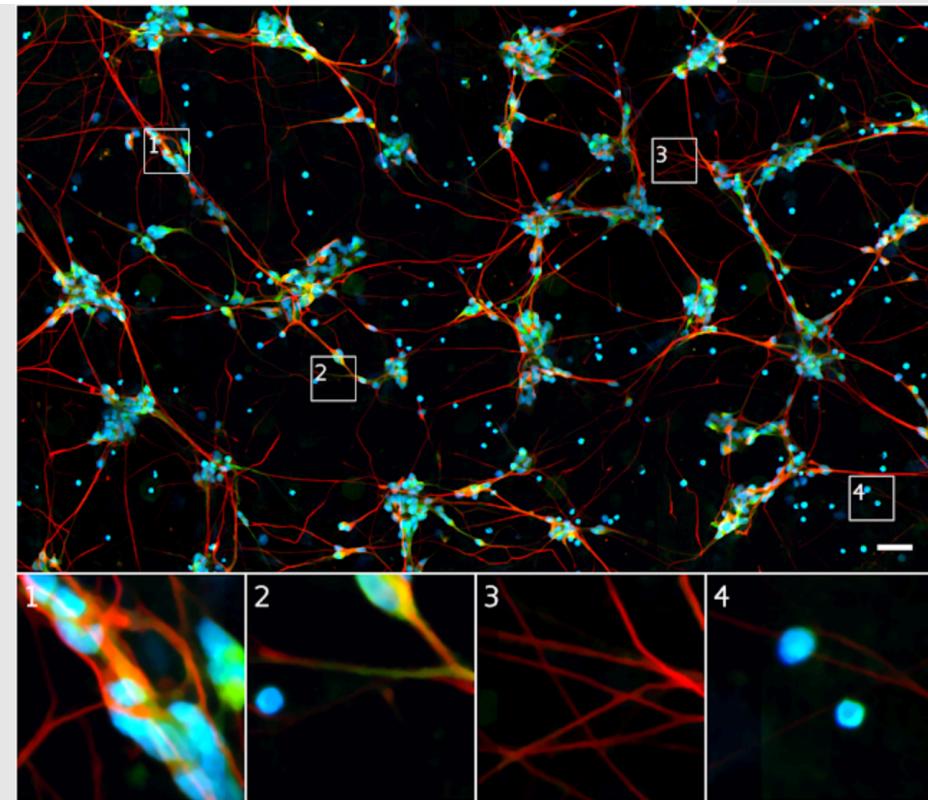
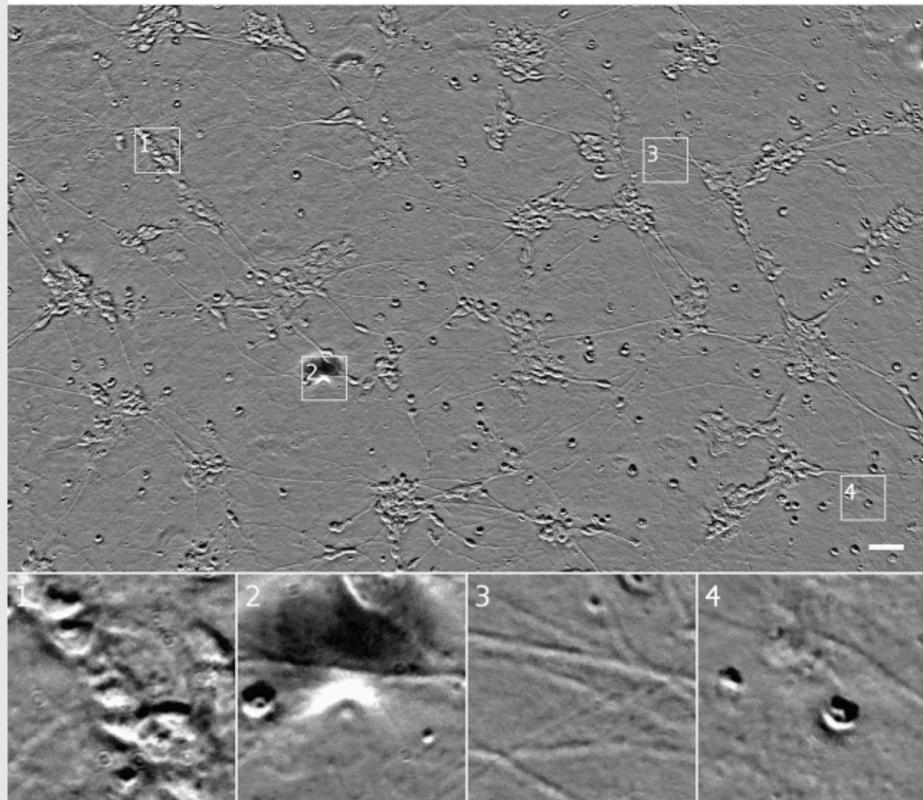


# Deep Learning prediction

## Seeing More with In Silico Labeling of Microscopy Images

Thursday, April 12, 2018

Eric Christiansen, Senior Software Engineer, Google Research



# History of ML

- Early 20<sup>th</sup> century – Modern Statistics
- 1950's – Artificial Intelligence
- 1990's – Traditional Machine Learning
- 2000's – Deep Learning

# Plan for today

- Session 1: **Basic Predictive Models** (statistics differently)
- Session 2: **Features** (representing images in computer world)
- Session 3: **Unsupervised learning**
- Session 4: **Supervised learning**
- Session 5: **Deep Learning**