

# Deep Learning + Large Language Models for Bio-image Analysis

Robert Haase

Reusing Material from

Alex Krull (MPI CBG), Martin Weigert (EPFL Lausanne), Uwe Schmidt (MPI CBG), Johannes Soltwedel, Till Korten (PoL TU Dresden), Loic A. Royer (CZ Biohub), Alexandr Dibrov (CSBD/MPI-CBG Dresden), Aditya Ramesh et al (OpenAI) and Alexandr Khrapichev (University of Oxford), Carsen Stringer et al (HHMI Janelia)

# Deep Learning and Large language models

- How it started



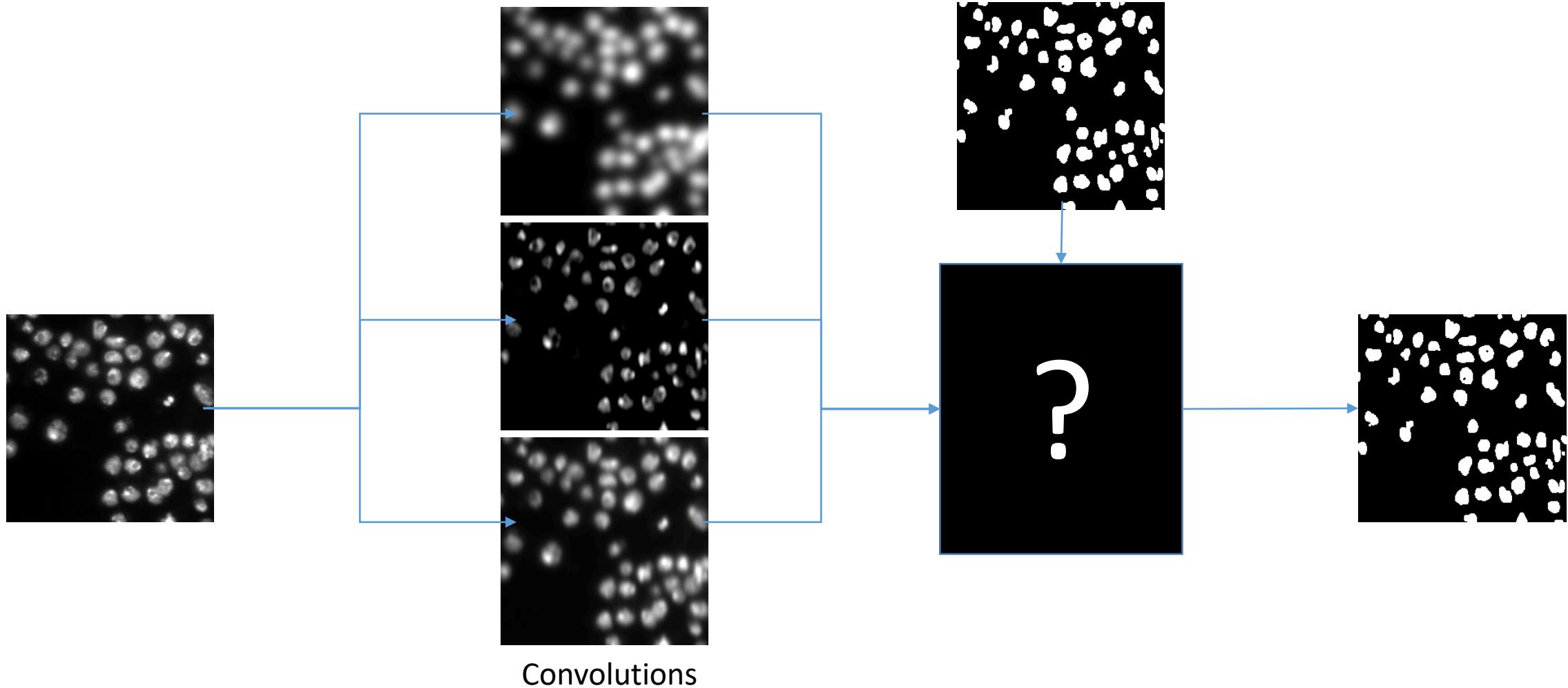
- How it's going

(GitHub copilot)

```
👤 Robert Haase +4 *
@catch_NaN
def agglomerative_clustering(
    reg_props: pd.DataFrame, cluster_number: int, n_neighbors: int
) -> Tuple[str, np.ndarray]:
    """
    """
```

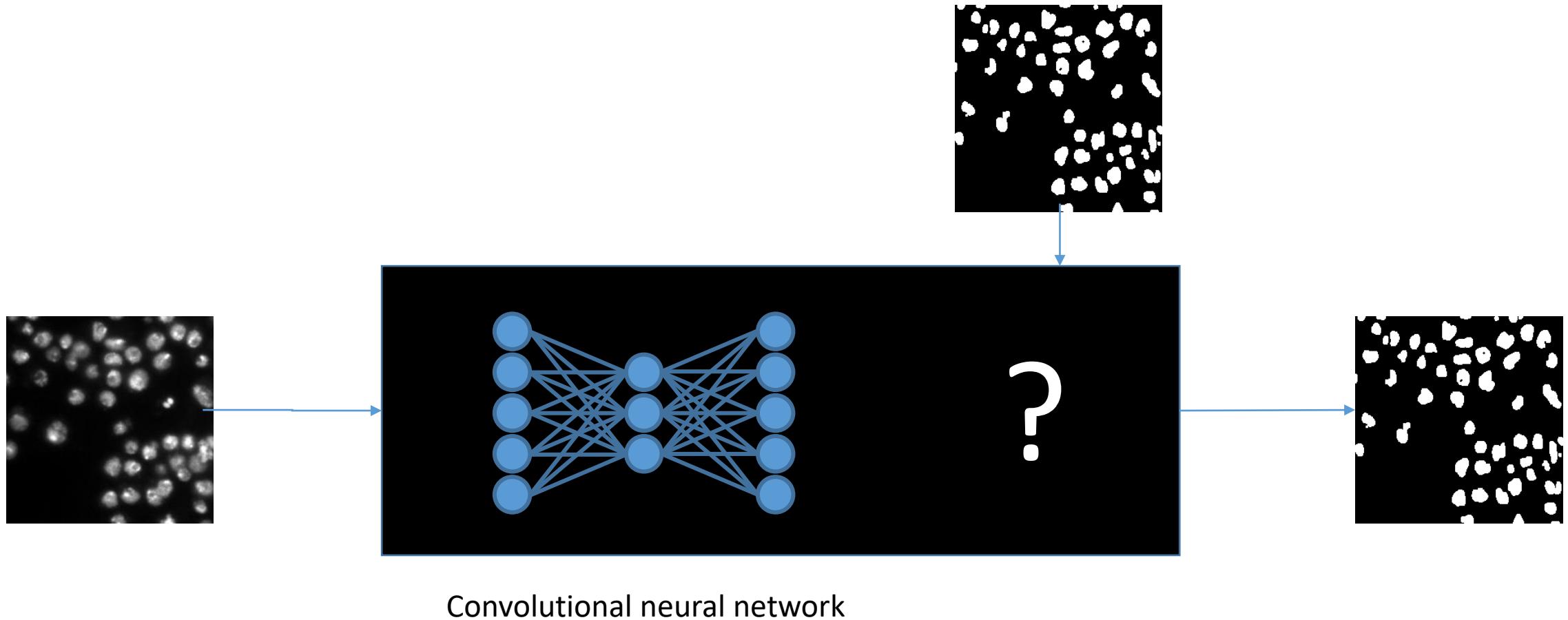
# Machine learning for image analysis

- Recap: In classical supervised machine learning, we typically select features for training our classifier



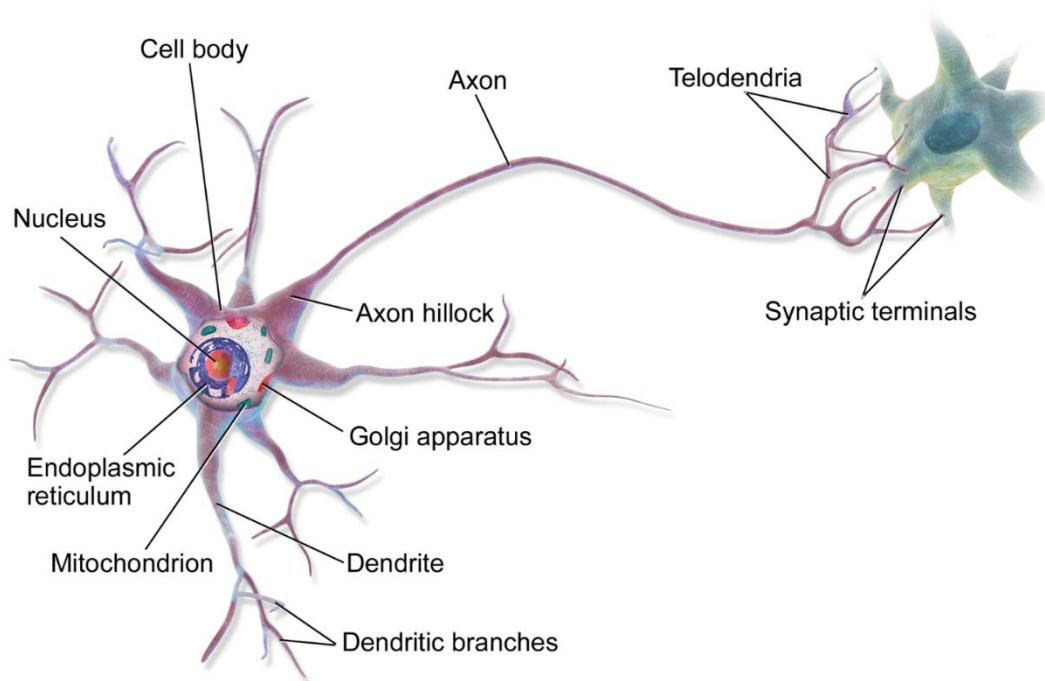
# Deep learning for image analysis

- In deep learning, this selection becomes part of the black box

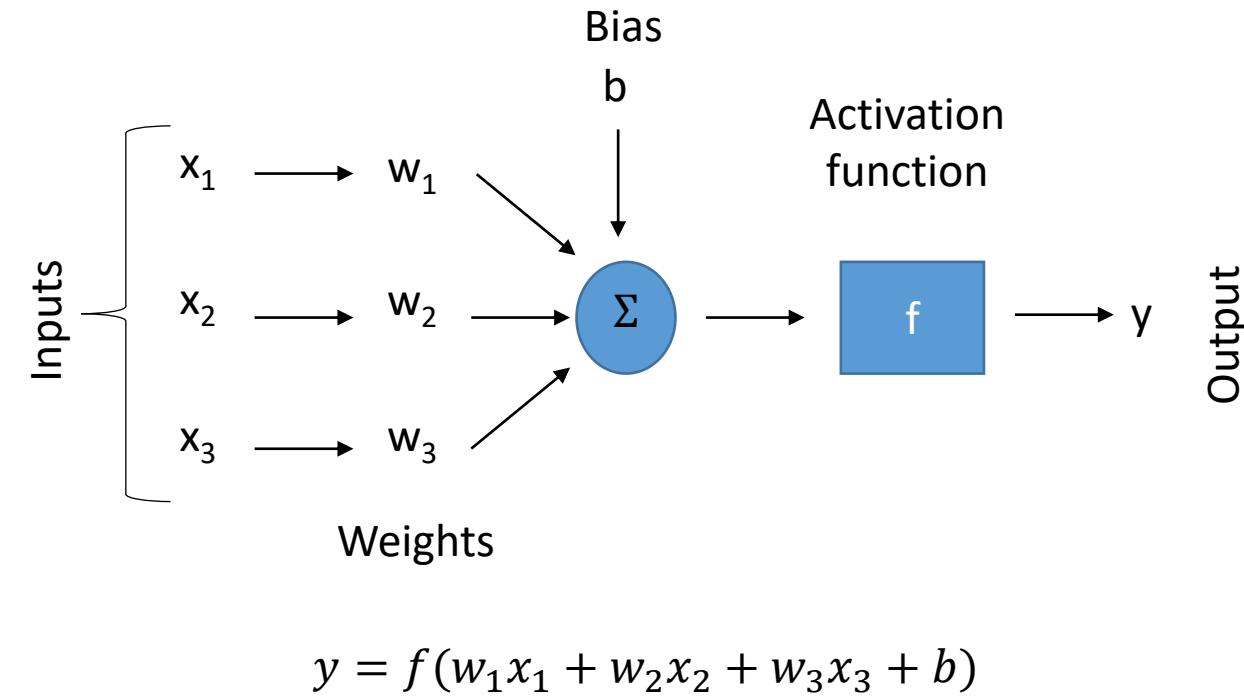


# Neural networks

- How biologists see neurons



- How computer scientists see neurons  
“perceptron”



Neuron image source:

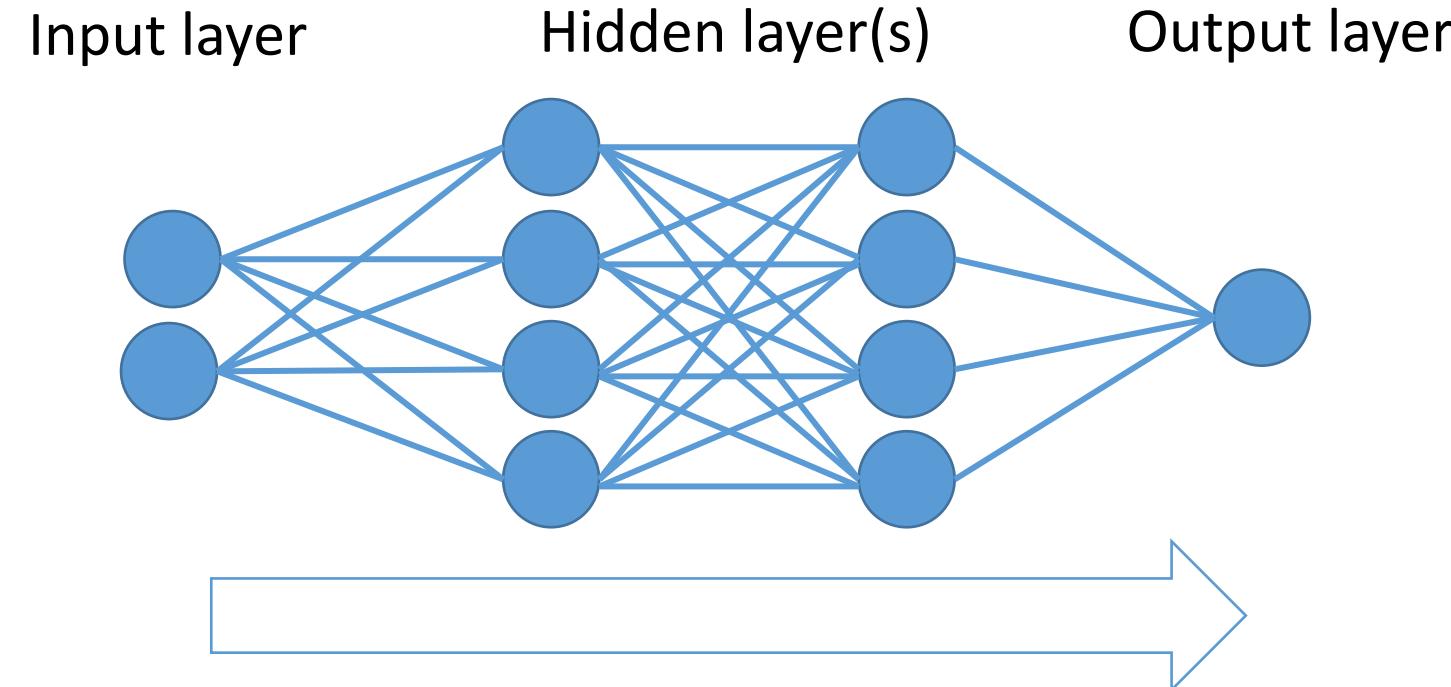
[https://commons.wikimedia.org/wiki/File:Blausen\\_0657\\_MultipolarNeuron.png](https://commons.wikimedia.org/wiki/File:Blausen_0657_MultipolarNeuron.png)

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# Neural Networks

- Early form: “Multilayer Perceptron”
- fully connected class of feedforward artificial neural network

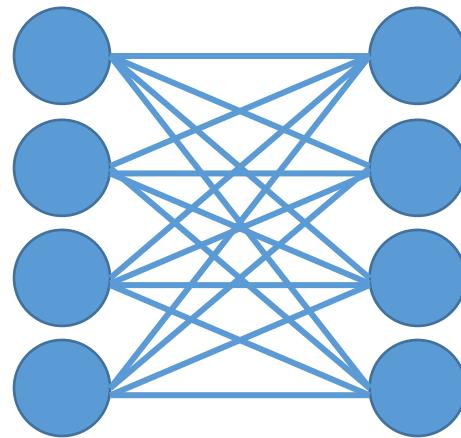
If there are *many* hidden layers, we speak of a *deep* neural network



# Convolutional neural networks

- Layer types

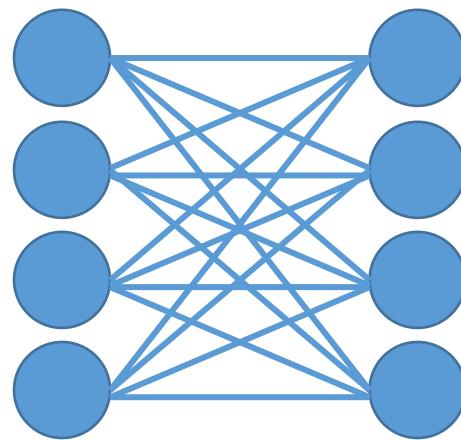
Fully connected layer



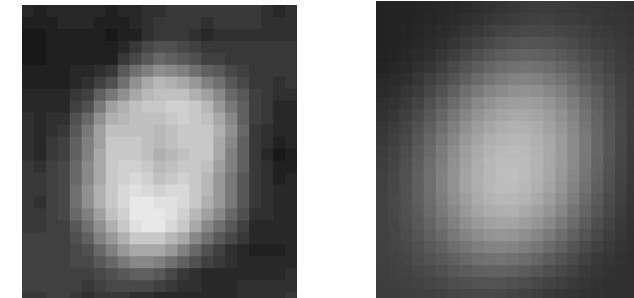
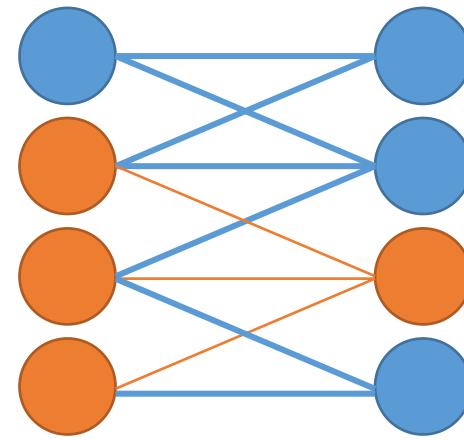
# Convolutional neural networks

- Layer types

Fully connected layer



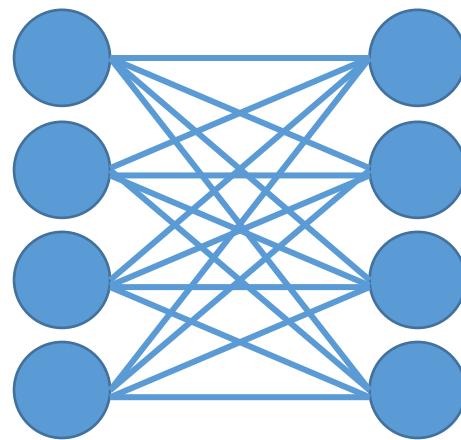
Convolutional layer



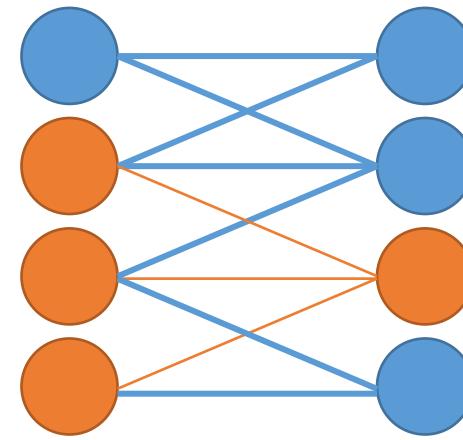
# Convolutional neural networks

- Layer types

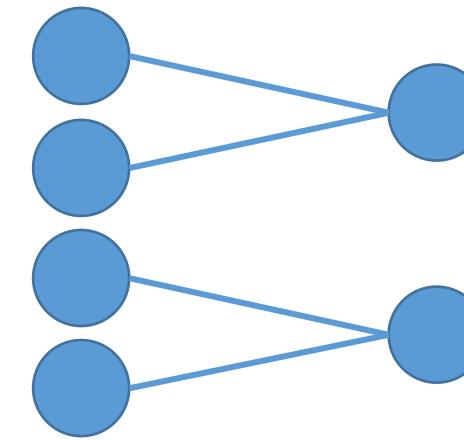
Fully connected layer



Convolutional layer



Pooling layer  
("Max pool", "Average pool")



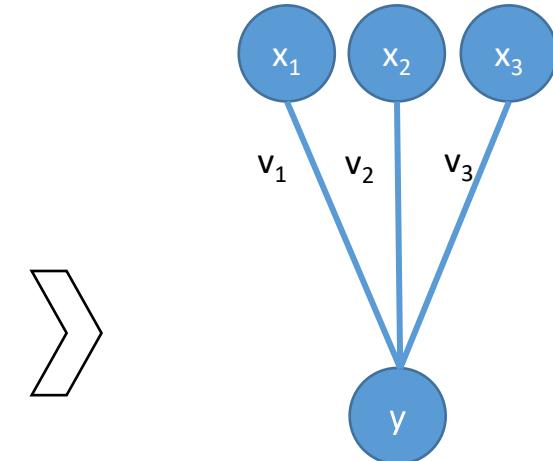
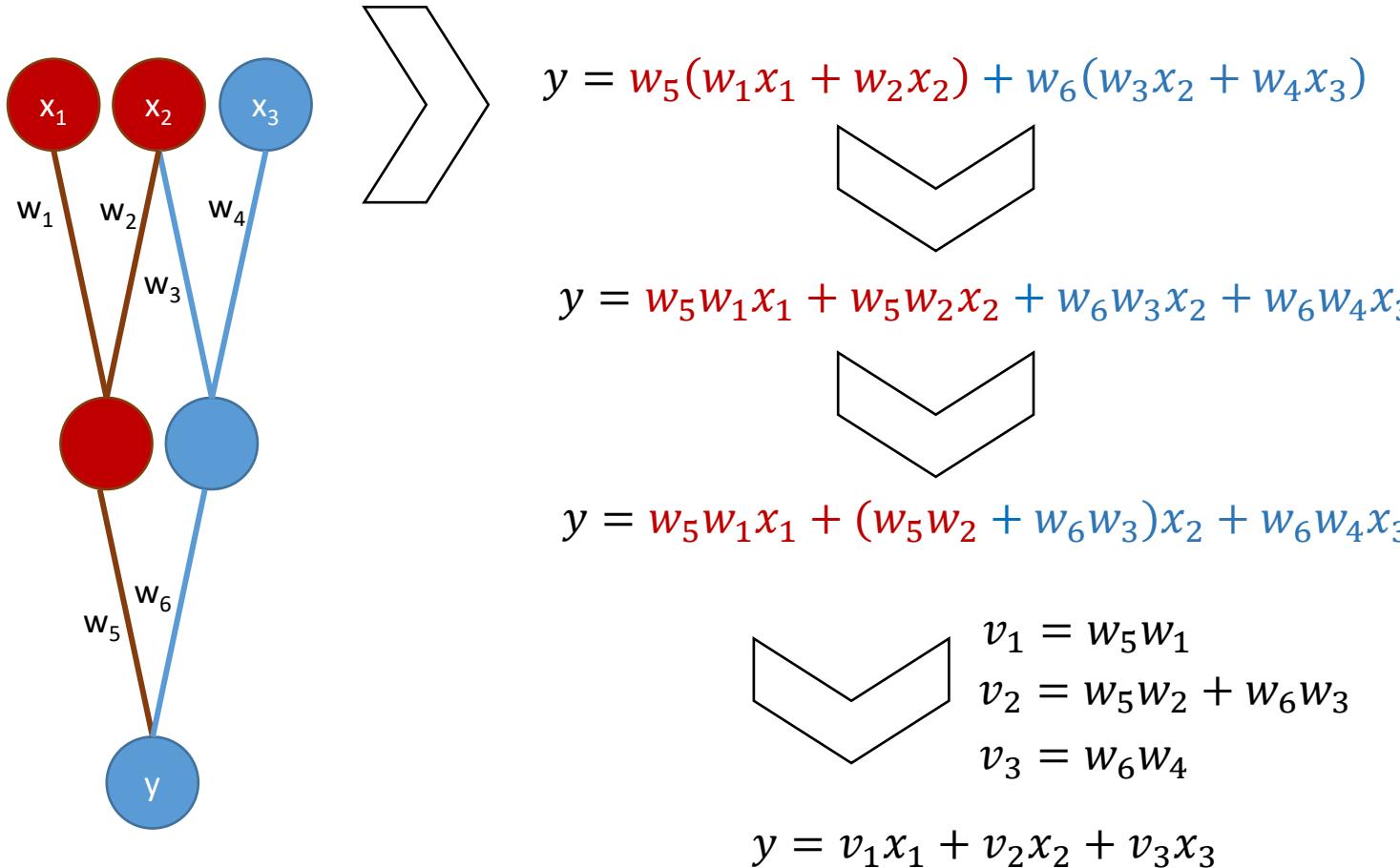
3	15	1	13
9	7	0	10
11	5	5	3
1	8	9	6

Max pooling

15	13
11	9

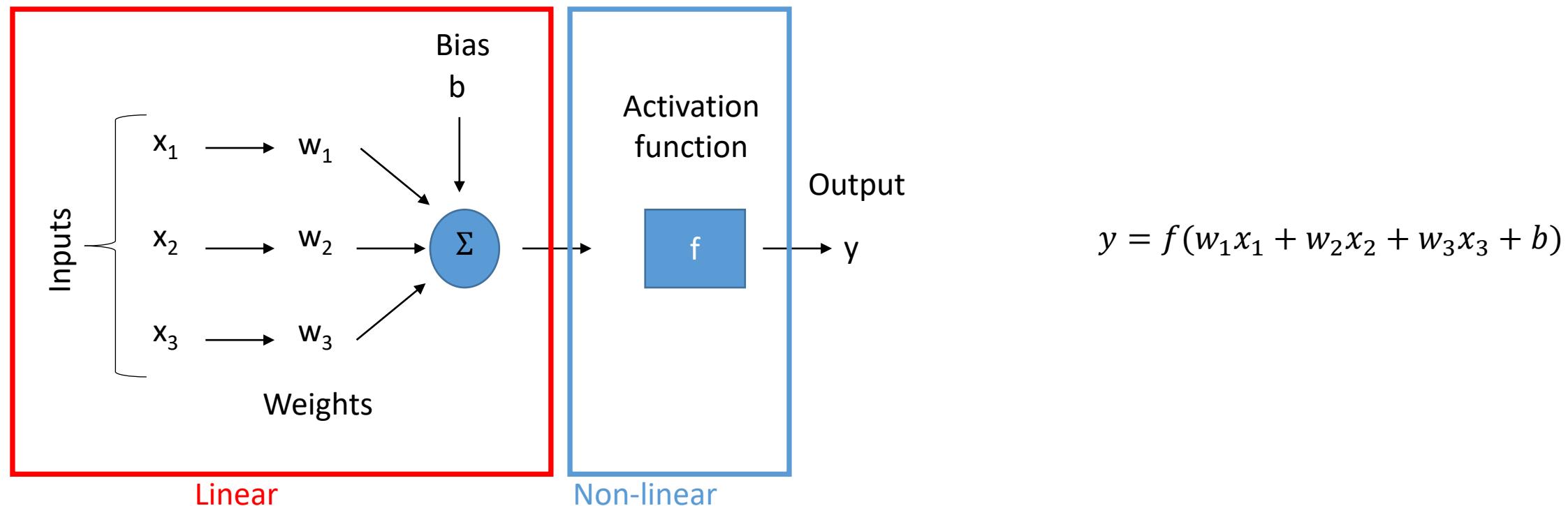
# Convolutional neural networks

- Assuming we had no activation functions in the network layers can be reduced by eliminating brackets!



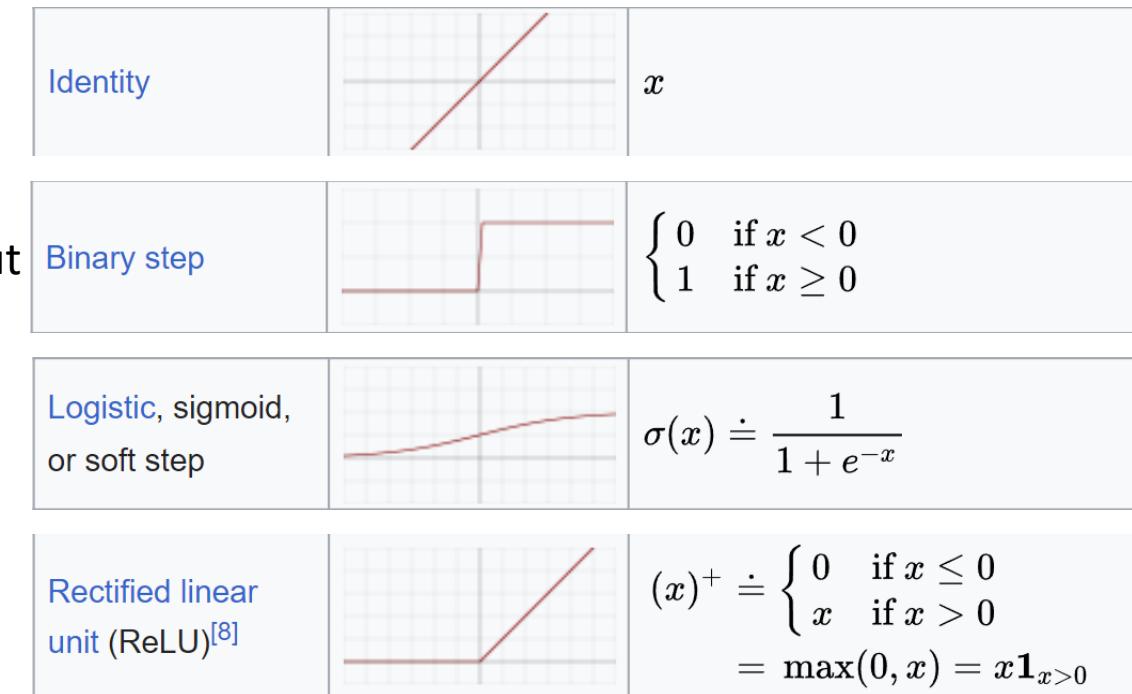
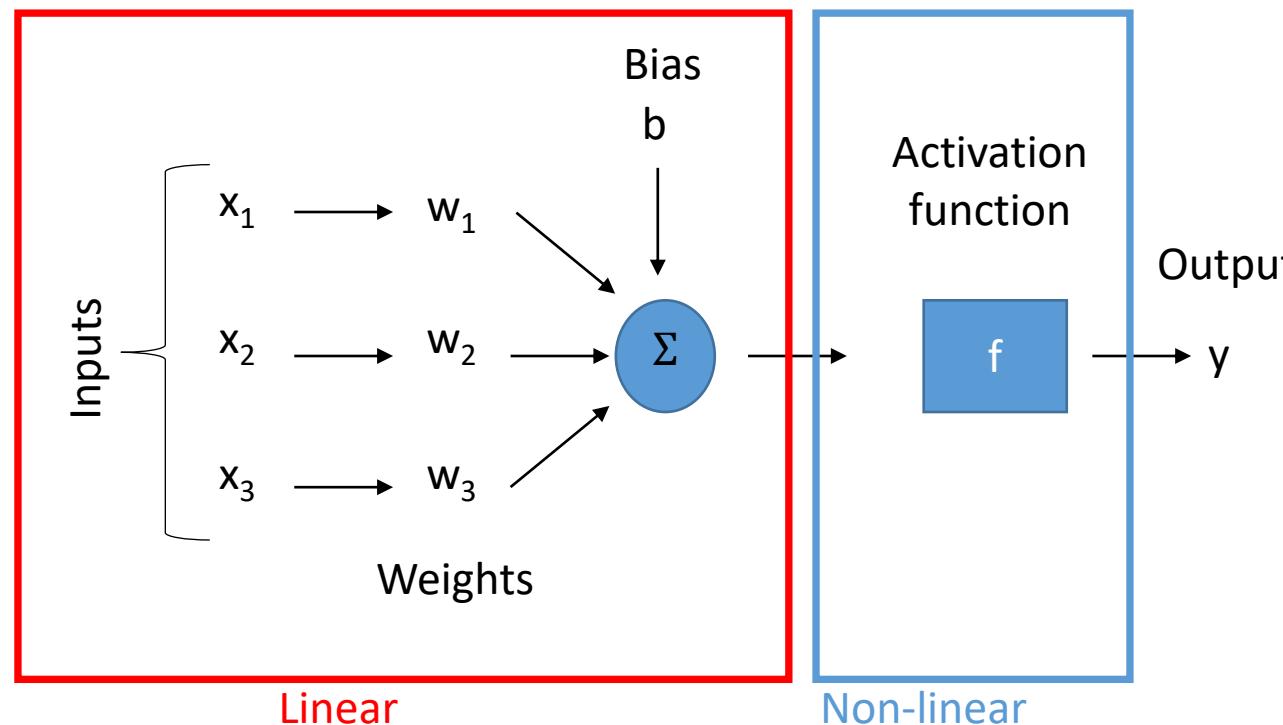
# Activation functions

- Introduction of *non-linearity* and *activation functions* enabled what we call *deep-learning* today.



# Activation functions

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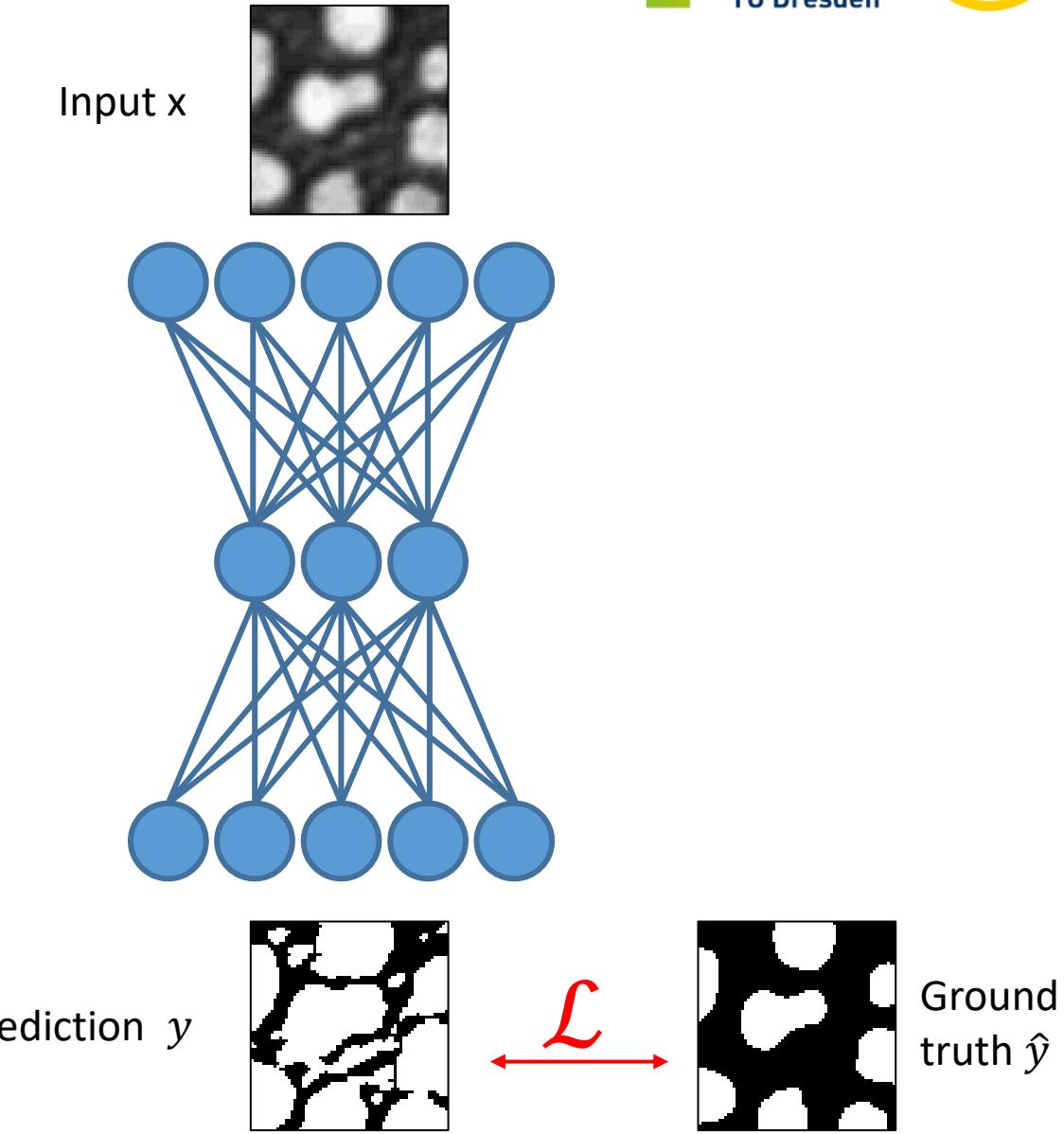


# Learning: Back propagation

- Learning is an optimization problem
- Step 0: Initialize the network randomly
  - Weights
  - Bias
- Step 1: Forward pass the input through the network, get an initial prediction
- Step 2: Compare the output with the ground truth, compute the error (loss function)
  - The loss function can be freely defined.
  - Example: mean squared error

$$\mathcal{L}(y, \hat{y}) = \frac{1}{M} \sum_{i=1}^M (\hat{y}_i - y_i)^2$$

- Step 3: Update weights



# Back-Propagation Algorithm

The loss function can be expanded from

$$\mathcal{L}(y, \hat{y}) = \frac{1}{M} \sum_{i=1}^M (\hat{y}_i - y_i)^2$$

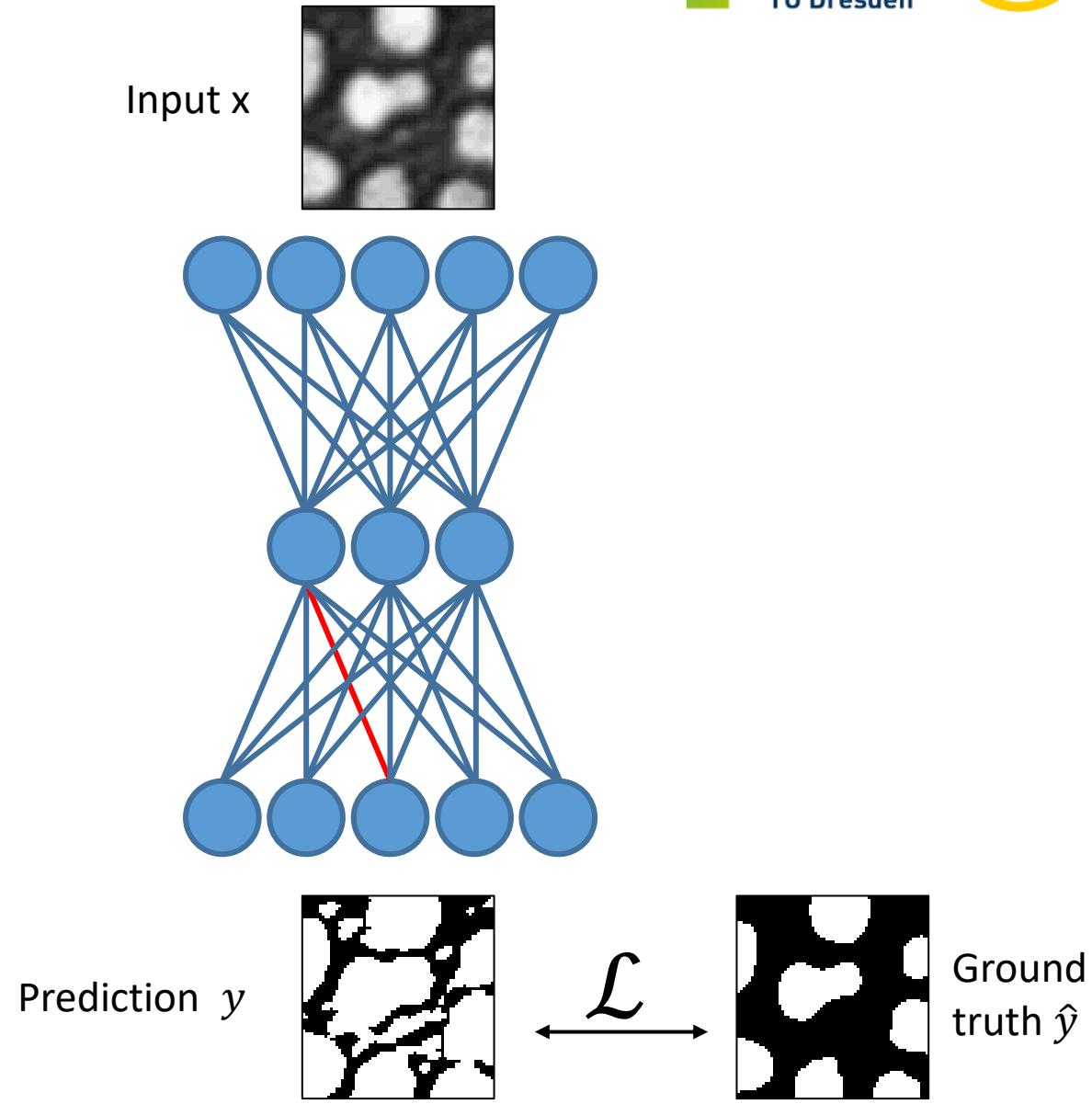
as the prediction depends on inputs  $x$  weights  $w$  and bias  $b$

$$\mathcal{L}(\hat{y}, x, w) = \frac{1}{M} \sum_{i=1}^M (\hat{y}_i - (w^T x_i + b))^2$$

Derivatives with respect to  $w$  and  $b$ :

$$\frac{\partial \mathcal{L}(\hat{y}, x, w)}{\partial w} = -\frac{1}{N} \sum_{i=1}^N 2x_i(\hat{y}_i - (w^T x_i + b))$$

$$\frac{\partial \mathcal{L}(\hat{y}, x, w)}{\partial b} = -\frac{1}{N} \sum_{i=1}^N 2(\hat{y}_i - (w^T x_i + b))$$



# Back-Propagation Algorithm

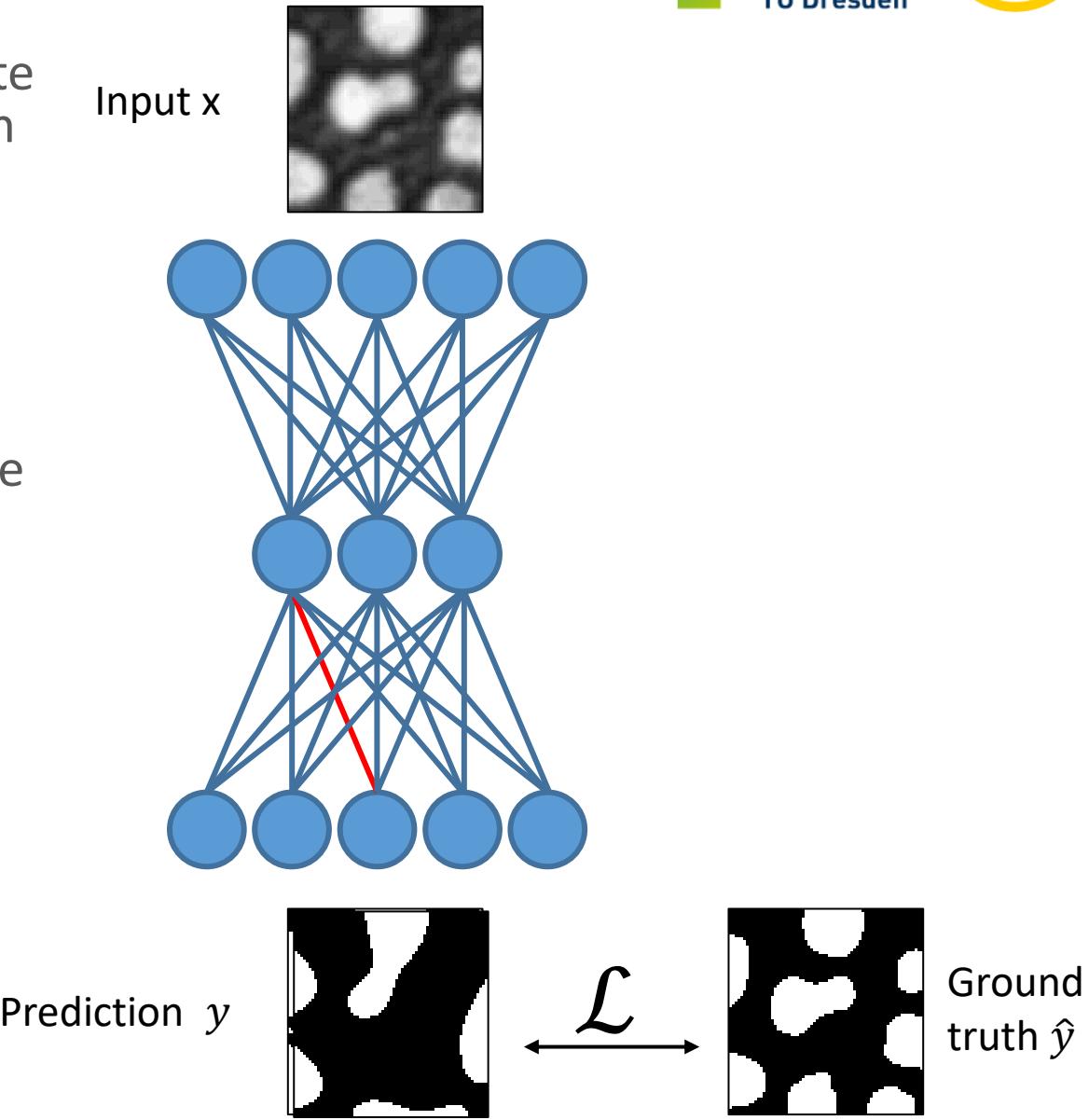
- The principle of the BackProp algorithm is to calculate the gradient of the loss function with respect to each trainable parameters of the network, i.e.

$$\frac{\partial \mathcal{L}}{\partial w_{ij}^k}$$

- where  $w_{ij}^k$  is the i:th weight of node j in layer k, which will allow the optimization algorithm to update the weights step by step using stochastic gradient descent

$$w_{ij}^k = w_{ij}^k + \eta \frac{\partial \mathcal{L}}{\partial w_{ij}^k}$$

- where  $\eta$  is the step length, in this context known as the **learning rate**.  
 $\eta$  can be varied during training (e.g. from epoch to epoch).
- Epoch*: Updating all parameters considering all input/ground-truth pairs



## Train dataset (e.g. 80% of the data)

- Used for training directly

## Validation dataset (10% of the data)

- After every iteration see if the model overfits

## Test dataset (10% of the data)

- Final evaluation after training is finished (once)

## Underfitting

- A trained model that is not even able to properly process the data it was trained on

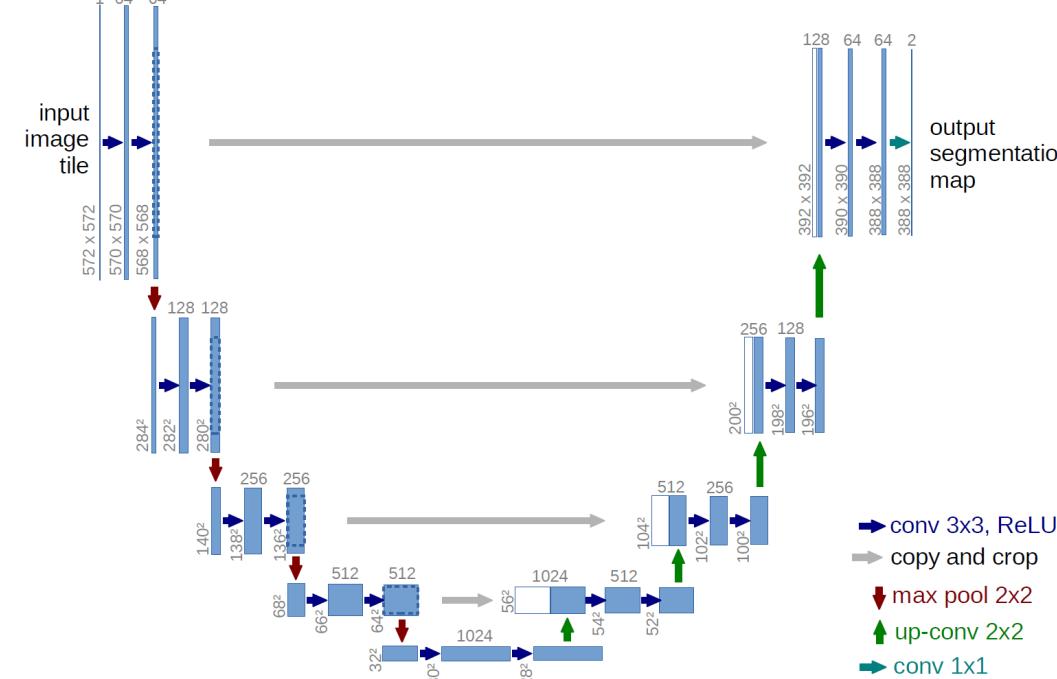
## Overfitting

- A model that is able to process data it was trained on well
- It processes other data poorly

<https://towardsdatascience.com/how-to-split-data-into-three-sets-train-validation-and-test-and-why-e50d22d3e54c>

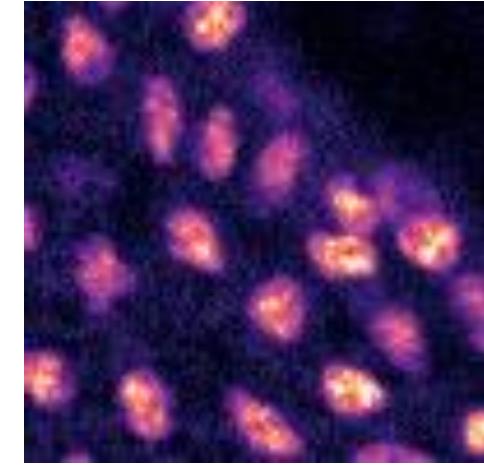
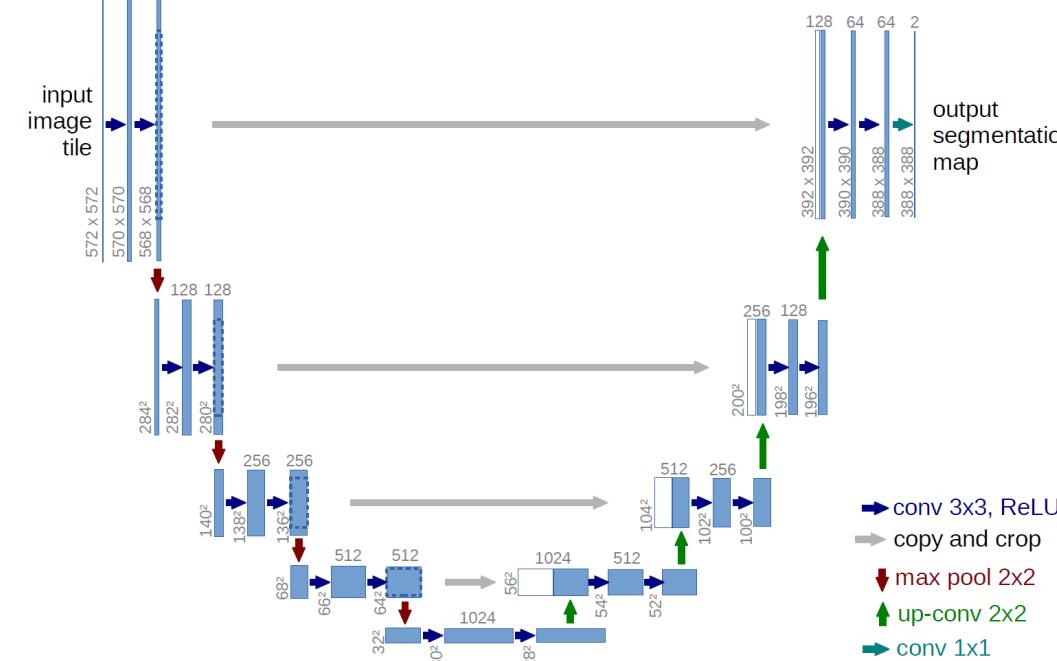
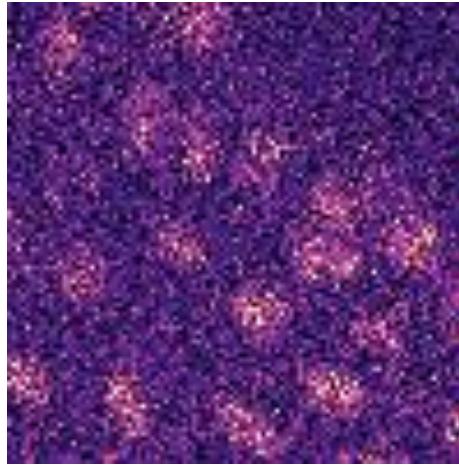
# The U-net

- The U-net is a very common DNN architecture in biological image processing.
  - Contraction: Increase the “What”, decrease the “Where”
  - Expansion: Increase the “Where”, decrease the “What”



# Image denoising: CARE

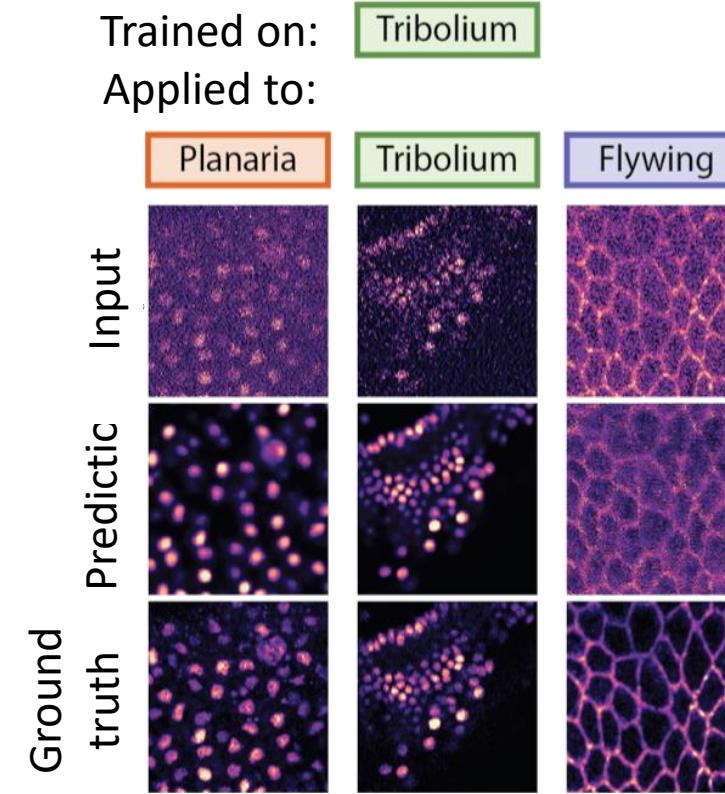
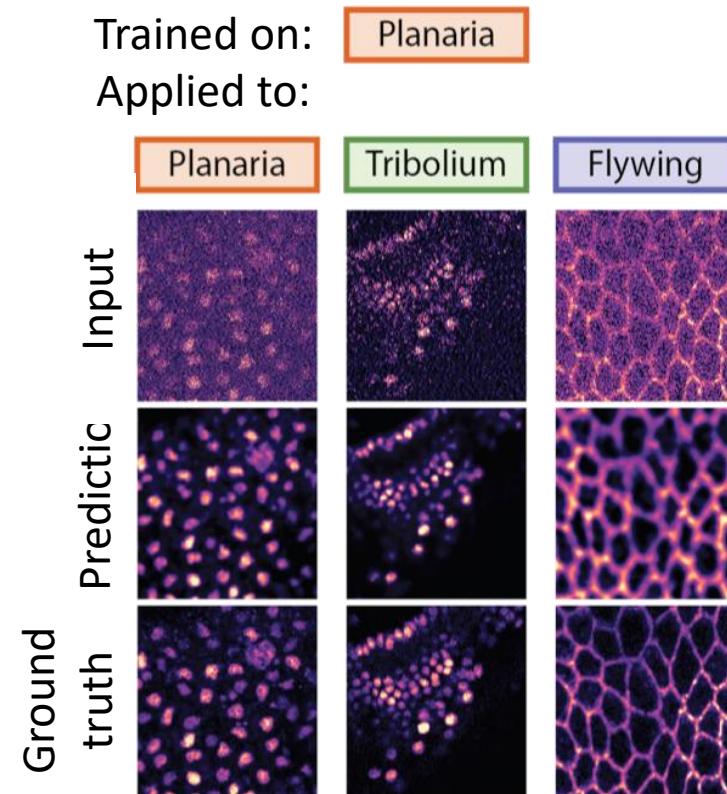
- Content aware image restoration (CARE)
- Image acquisition of pairs of images: A high-quality and a low-quality image.
- Problem: shot noise, biology moves!



Source: Ronneberger (2015) <https://arxiv.org/pdf/1505.04597.pdf>  
Slide adapted from: Martin Weigert, EPFL Lausanne

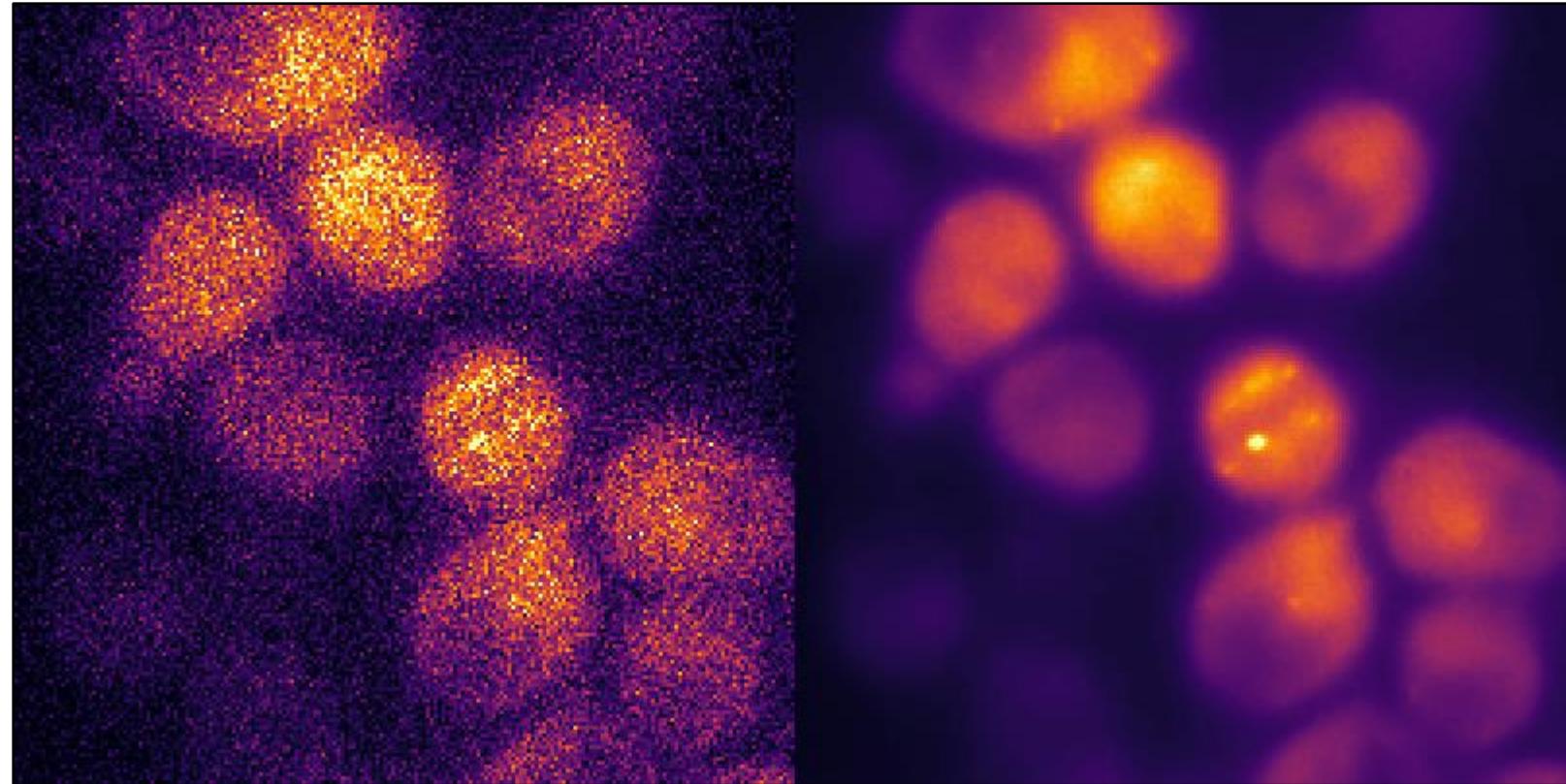
# Image denoising: CARE

- Important to use on the same conditions/structures/staining that the networks were trained on!



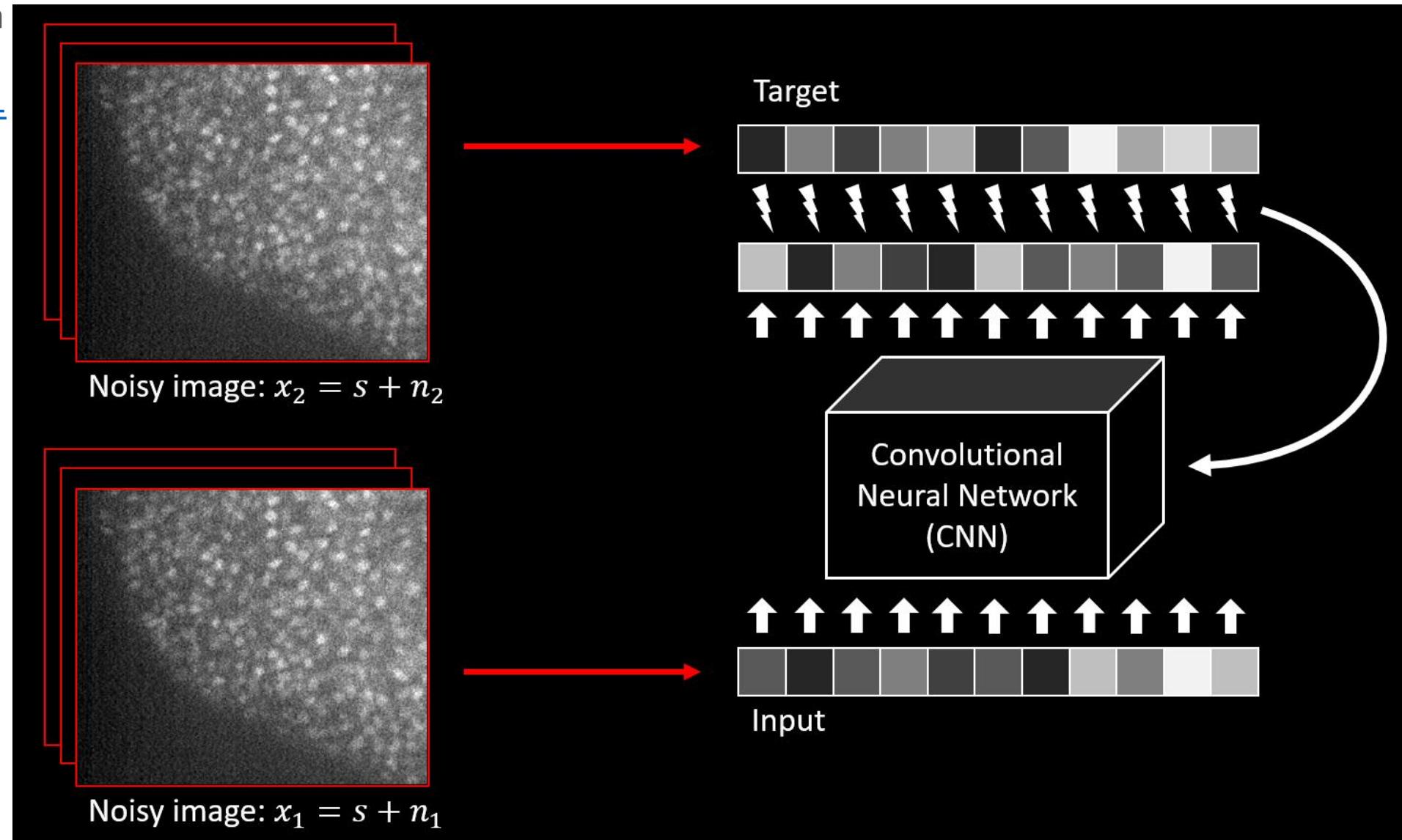
# Image denoising

- Noise2Void

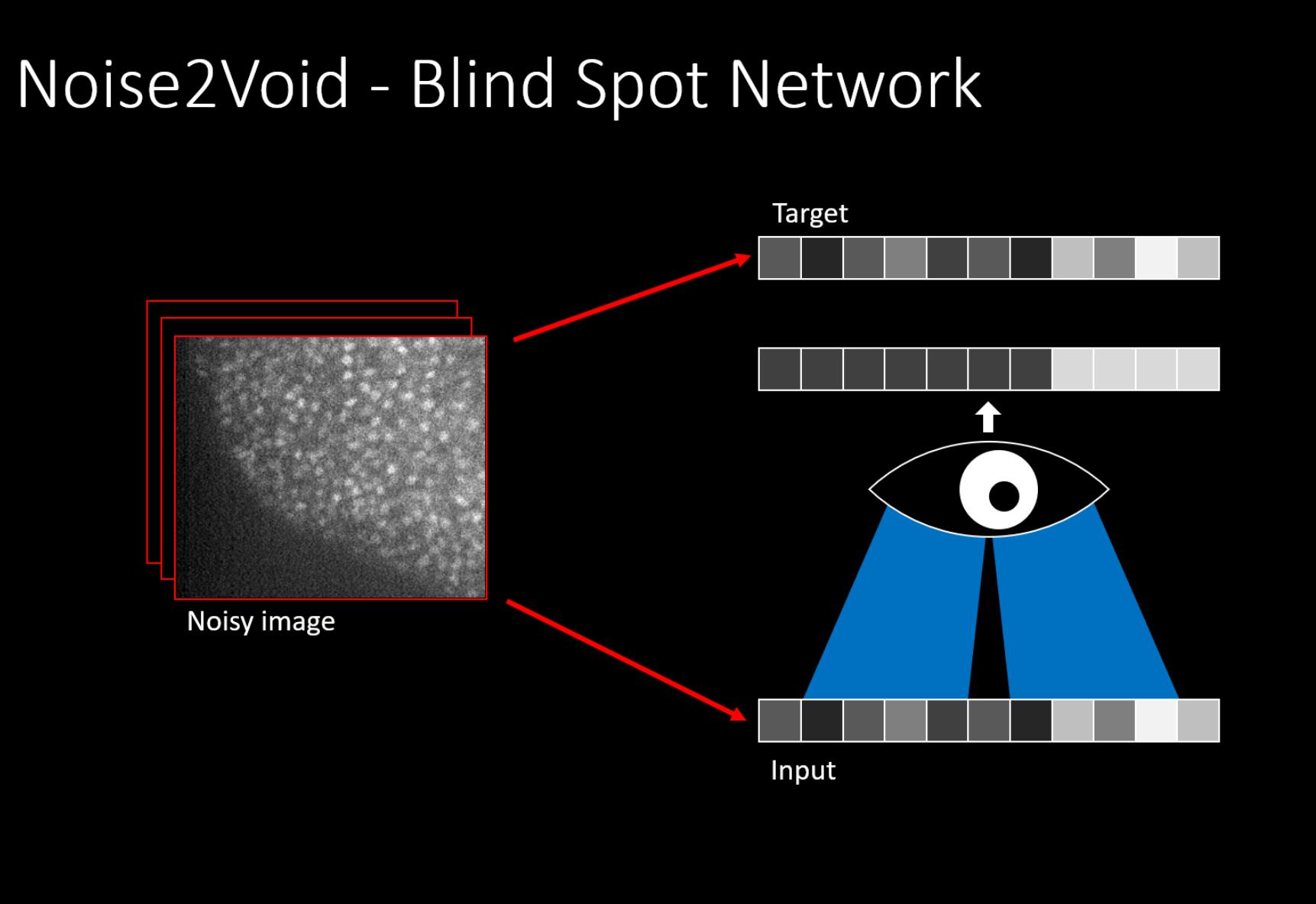


# Image denoising

- Noise2Noise by NVidia  
(Lehtinen 2018)  
<https://arxiv.org/pdf/1803.04189.pdf>

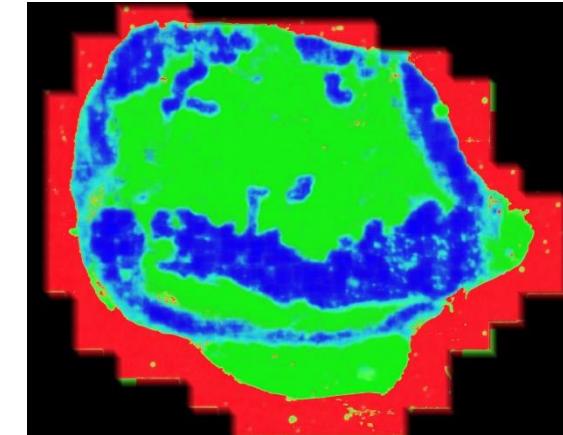
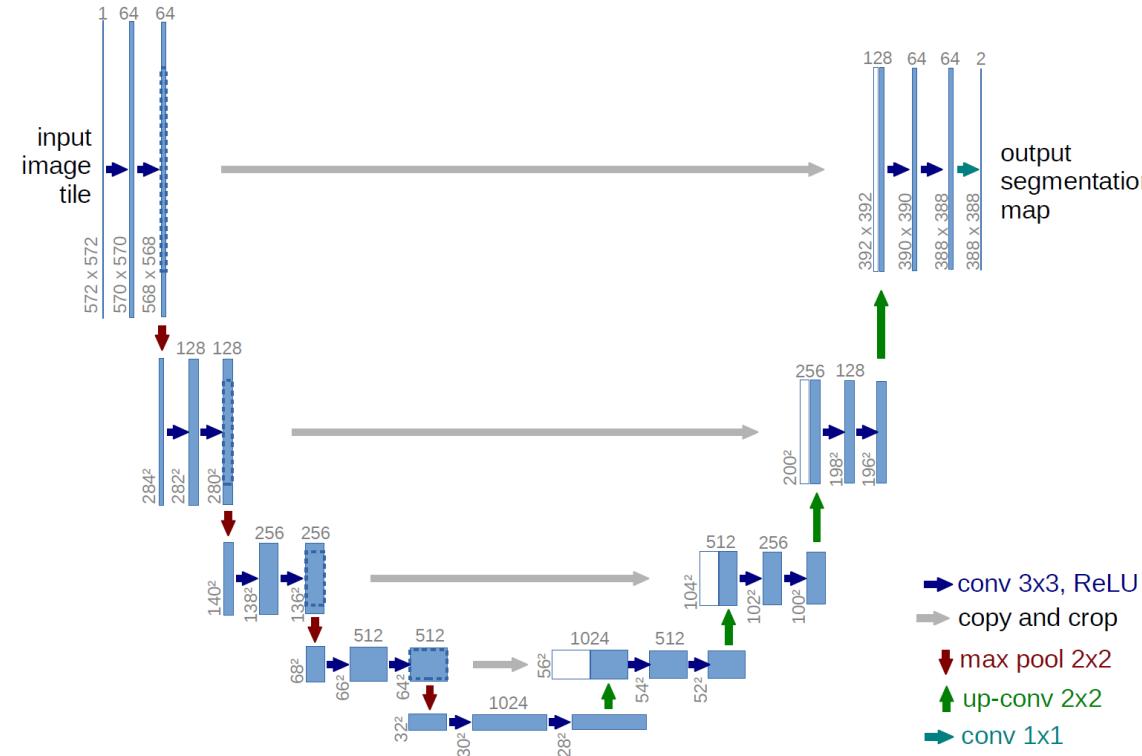
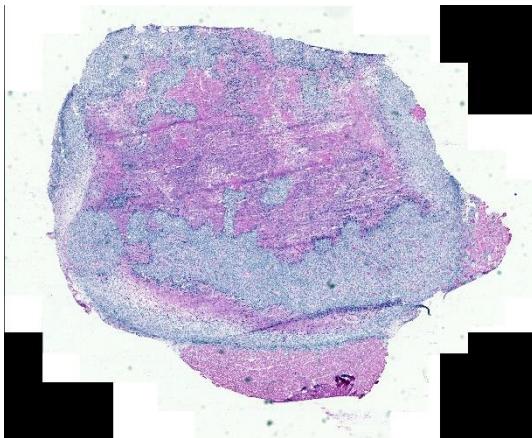


- Noise2void: Krull et al (2019)  
<https://arxiv.org/abs/1811.10980>
- Noise2self: Batson and Royer (2019)  
<https://arxiv.org/abs/1901.11365>



# Image segmentation

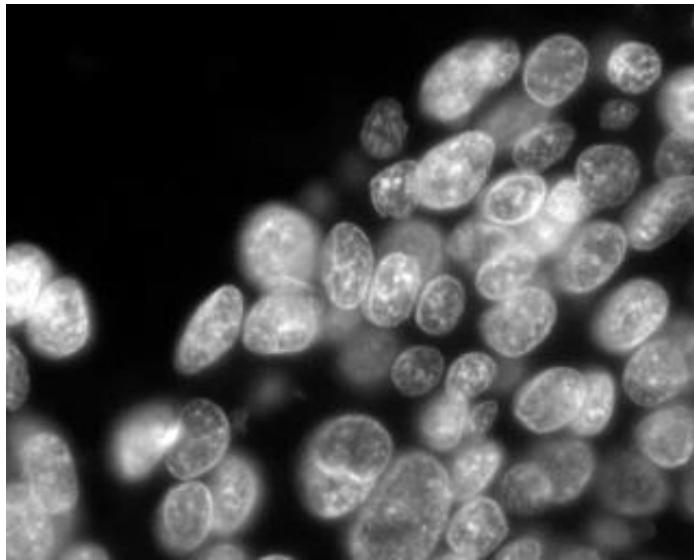
- U-net for pixel classification
  - > Semantic segmentation
  - > Probability maps



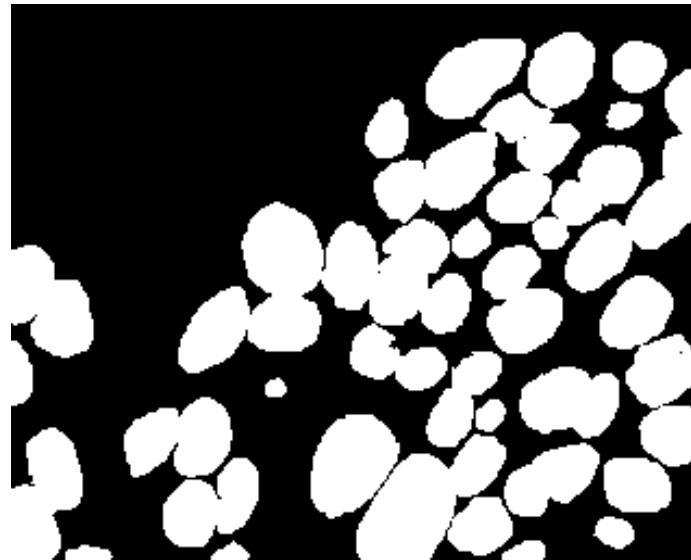
Legend:  
→ conv 3x3, ReLU  
→ copy and crop  
↓ max pool 2x2  
↑ up-conv 2x2  
→ conv 1x1

# Nuclei/cell segmentation

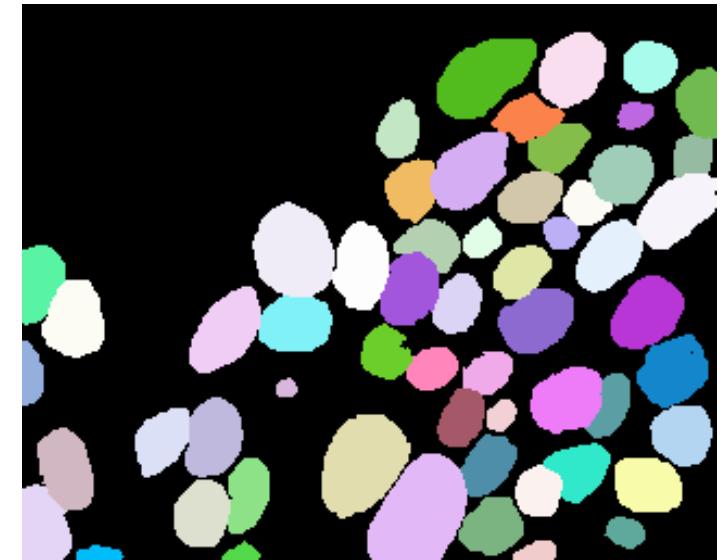
- Recap: Terminology



Input Image



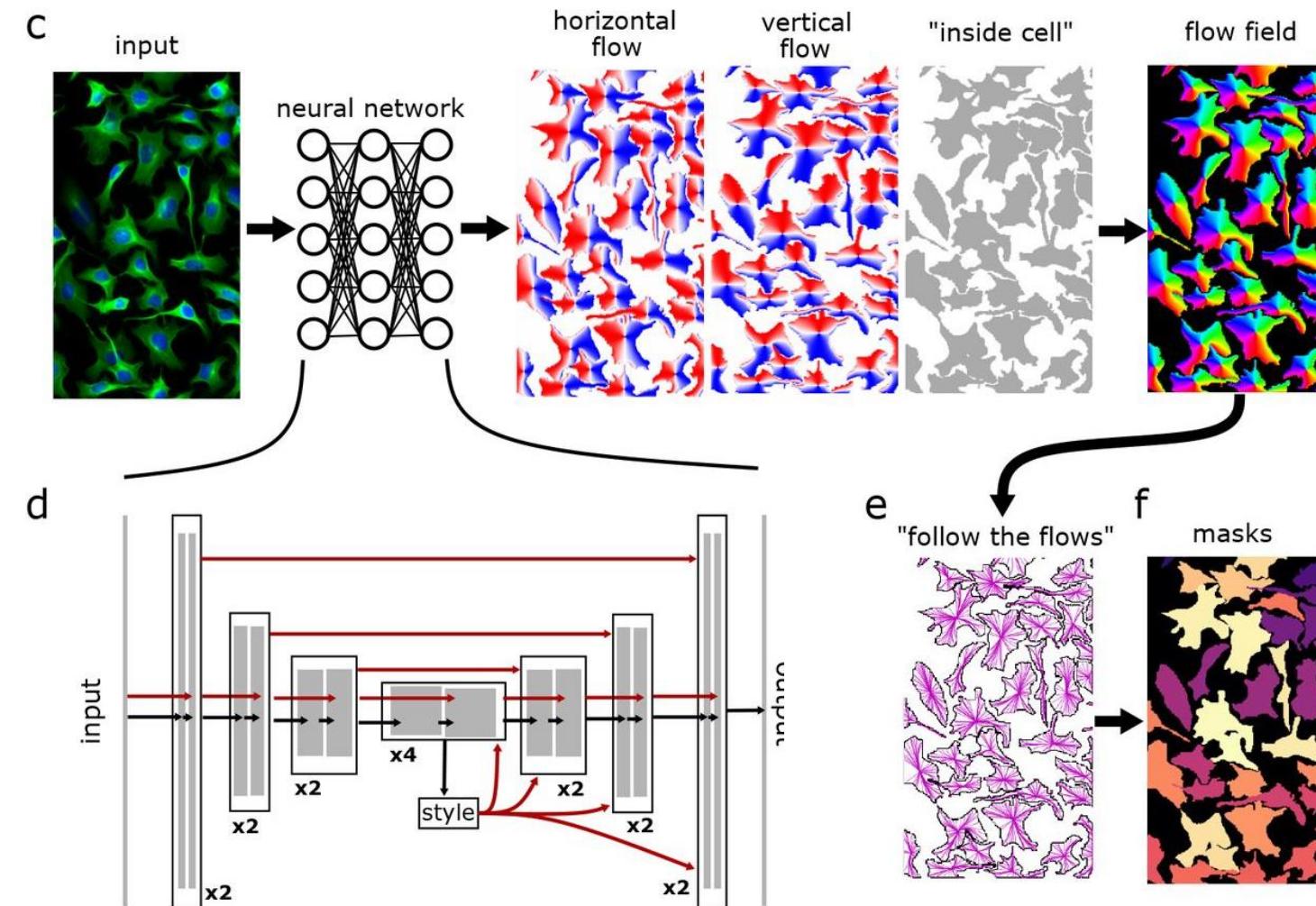
Semantic Segmentation  
(foreground/background)



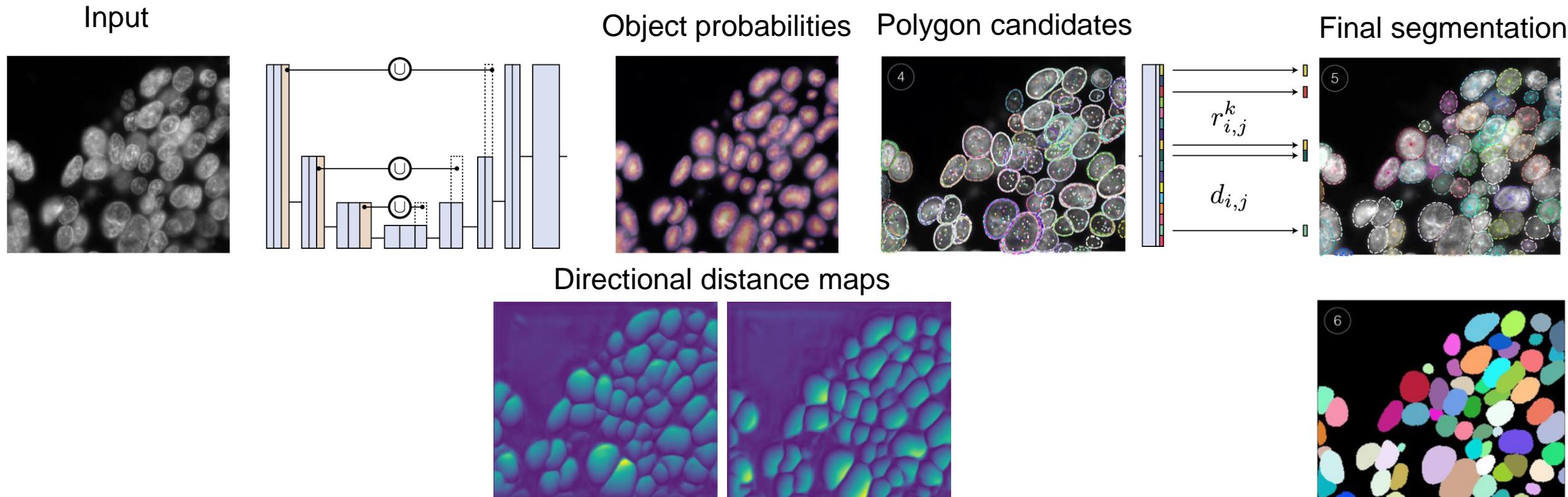
Instance Segmentation  
(individual cells)

# CellPose

- Cell/Nuclei – segmentation based on flow-fields
- Technically similar to Watershed, but with a deep-learning based altitude-image



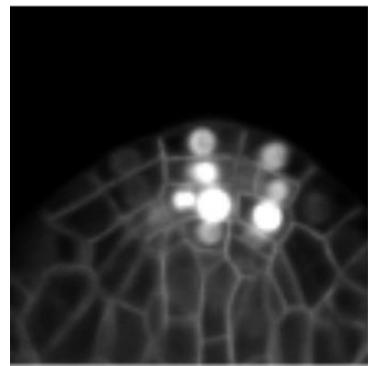
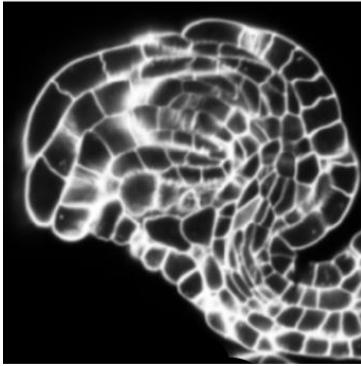
- Prediction of probable object centers + polygon outlines
- Non-maximum-suppression of less likely polygons



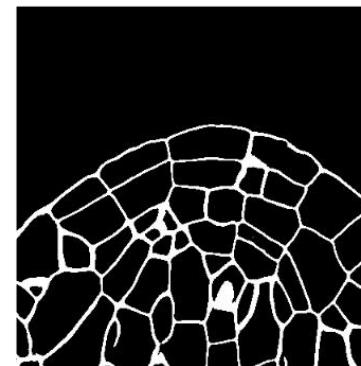
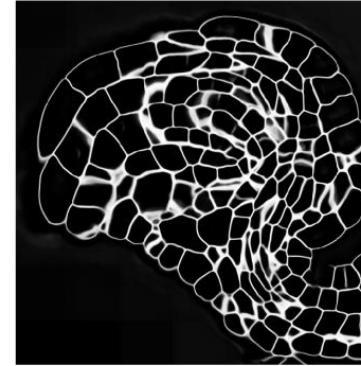
Slide adapted from: Martin Weigert, EPFL Lausanne and  
from <https://github.com/stardist/stardist> (license: BSD3)

- Combination of neural networks + graph partitioning

Input

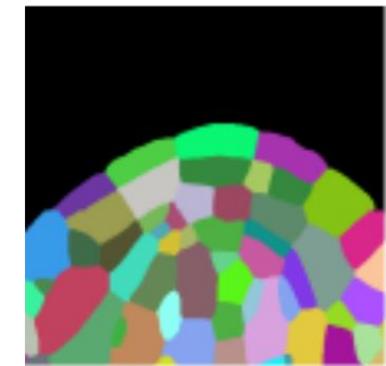
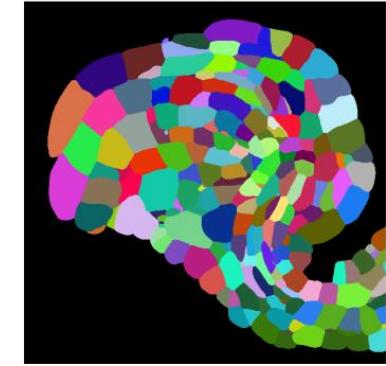


Boundary Predictions



Neural Network

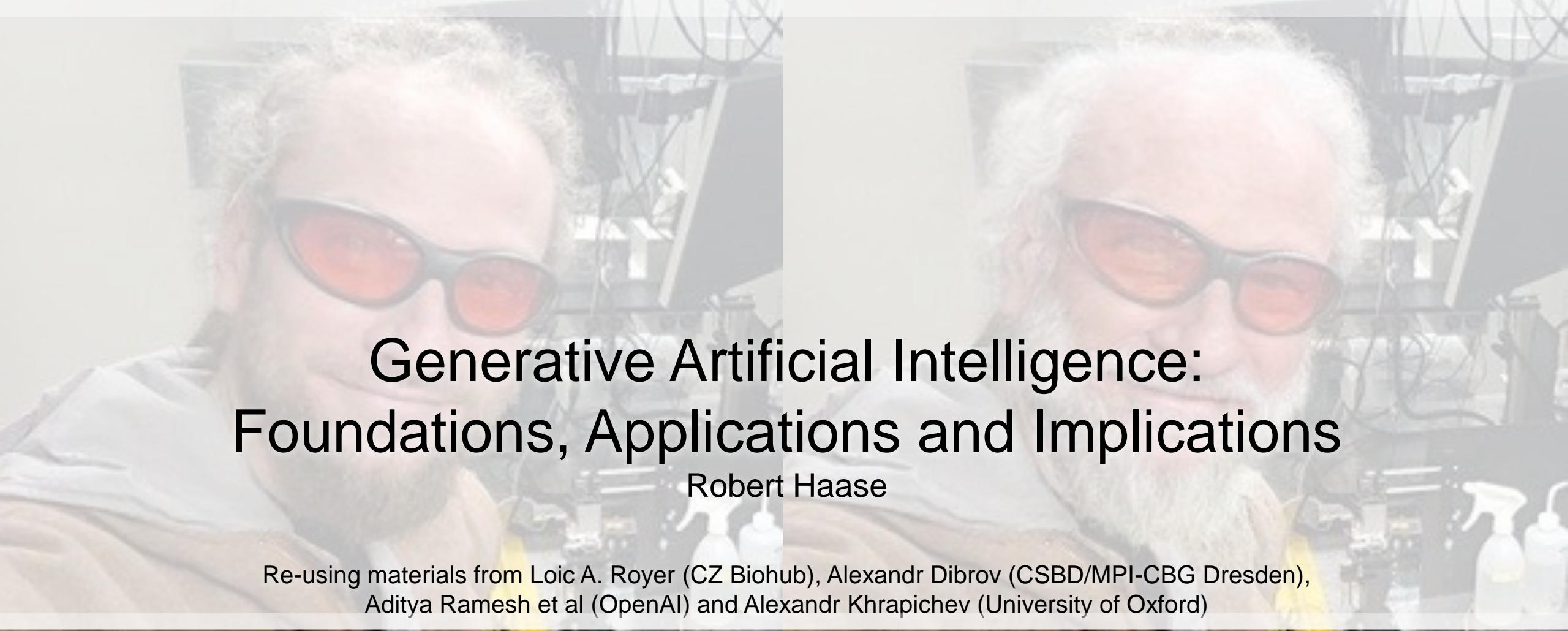
Segmentation



Graph Partitioning

# Summary: Deep Learning for Bio-image Analysis

- [Convolutional] Neural Networks is a decade old technology that enabled breakthroughs in various fields during the last decade. Examples in Microscopy:
  - Image Denoising
  - Image Segmentation
- Common scheme: Smart algorithms for processing input/output of neural networks
  - ~~Image in, segmentation out~~
- Deep-Learning based often represent state-of-the-art techniques with respect to result quality
- Training these models is
  - computationally expensive,
  - needs large amounts of training data (~~single images~~),
  - requires a certain level of expertise
- If Otsu-Thresholding + Connected Component Labeling does the job, don't dive into deep learning!



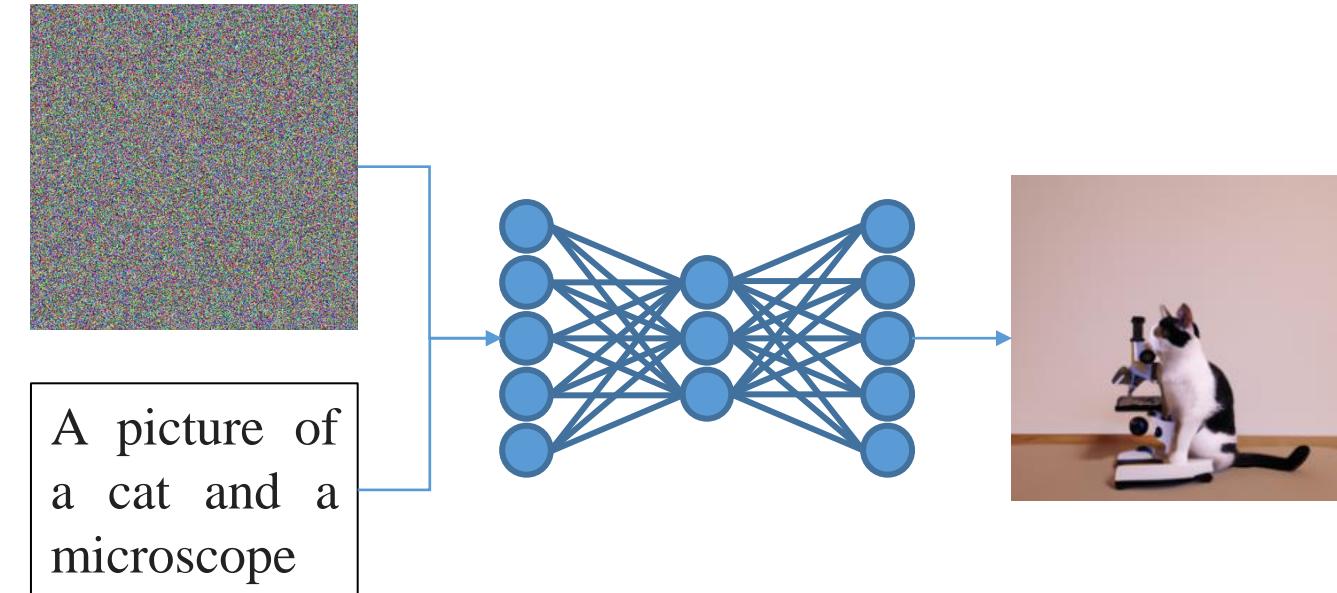
# Generative Artificial Intelligence: Foundations, Applications and Implications

Robert Haase

Re-using materials from Loic A. Royer (CZ Biohub), Alexandr Dibrov (CSBD/MPI-CBG Dresden),  
Aditya Ramesh et al (OpenAI) and Alexandr Khrapichev (University of Oxford)

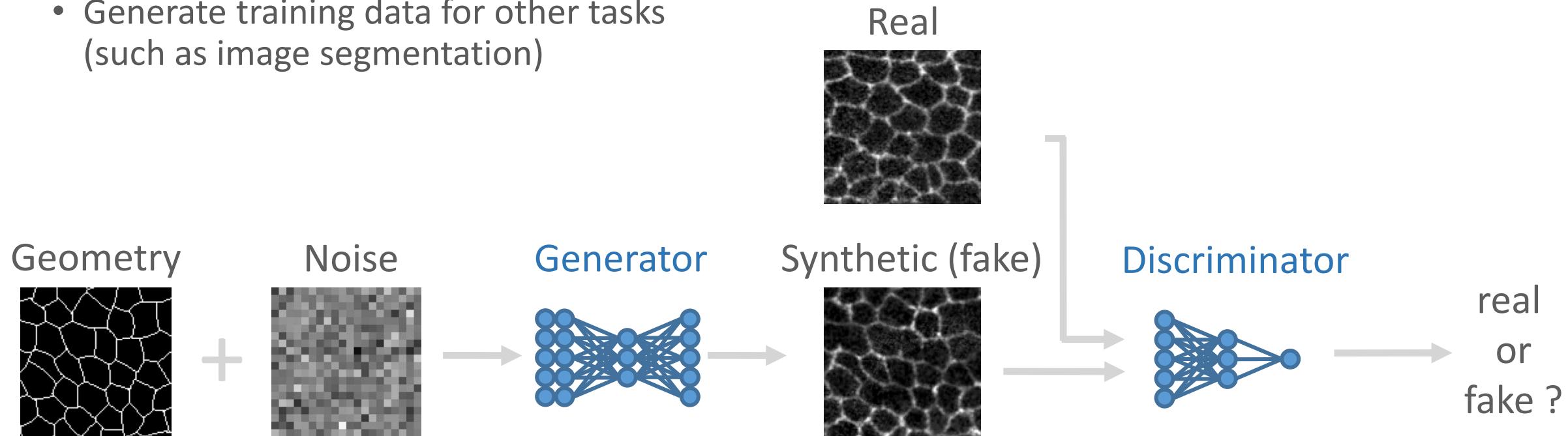
# Generative Artificial Intelligence

- Definition: “Generative artificial intelligence [...] is a type of artificial intelligence (AI) system capable of generating text, images, or other media in response to prompts.”<sup>1</sup>
- Commonly based on Neural Networks
- Bridges fields:
  - Natural Language Processing (NLP)
  - Computer Vision (CV)
- Use-cases
  - Translating text
  - Writing emails, text, grant proposals
  - Summarizing articles
  - Writing code
  - General question answering
  - Image generation



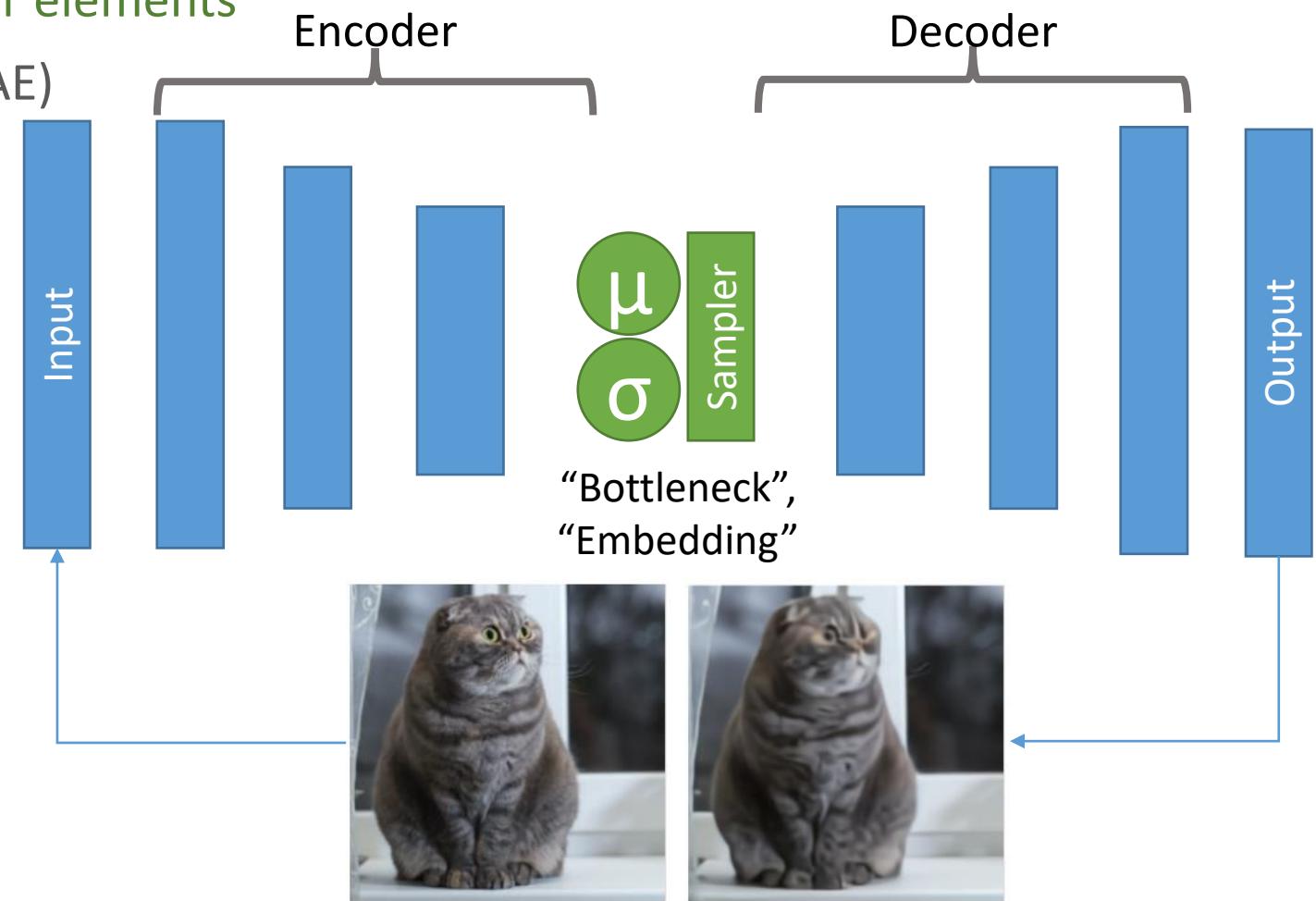
# How does it work?

- Combination of **neural networks**
- Example: Generative Adversarial Networks (GANs)
- Use-cases:
  - Generate training data for other tasks (such as image segmentation)



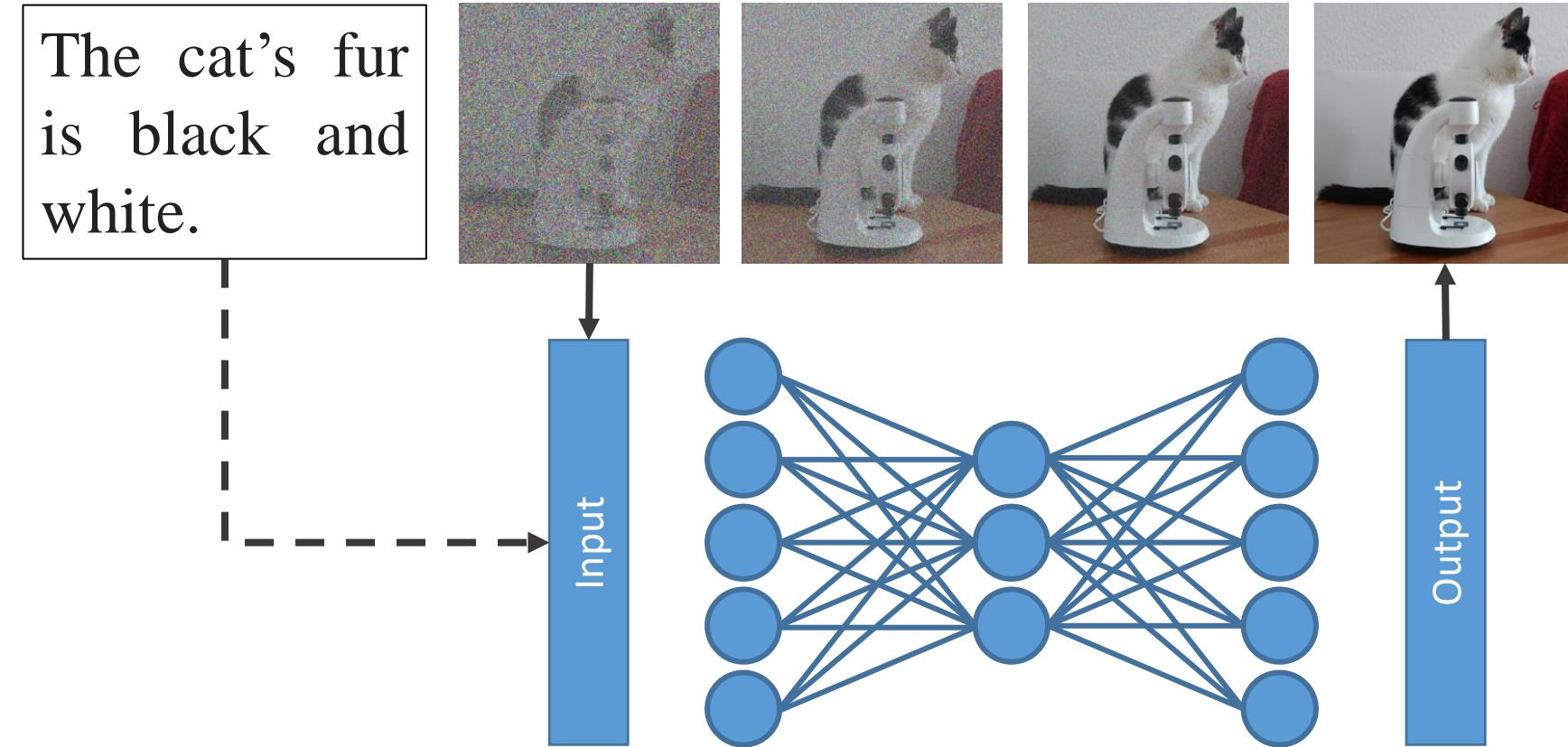
# How does it work?

- Combination of **neural networks** + **other elements**
- Example: Variational Auto-Encoders (VAE)
- Use-cases:
  - Combine images (cat, microscope)



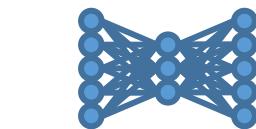
# How does it work?

- Combination of neural networks + other elements + various data sources
- Examples: GPT / DALL-E, Stable Diffusion
- Use-case:
  - Generate image from noise + text



# How does it work?

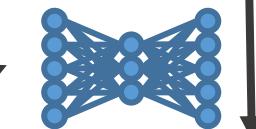
- Combination of neural networks + other elements + various data sources
- Examples: GPT / DALL-E, Stable Diffusion
- Use-case:
  - Generate image from noise + text



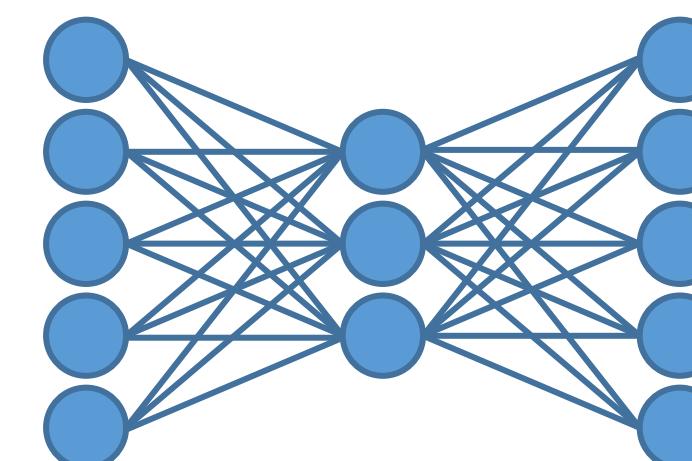
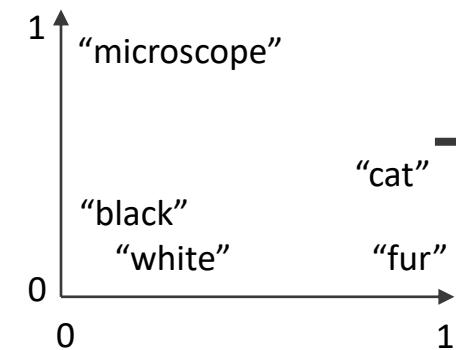
Self-attention

The cat's fur is black and white.

The cat's fur  
is black and  
white.



Word Embedding

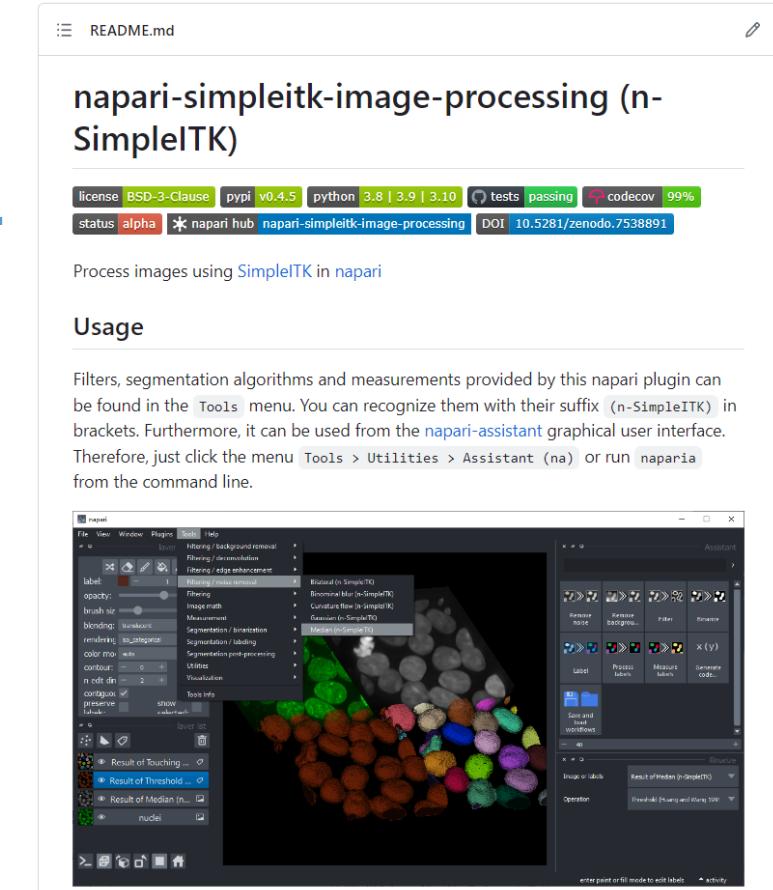
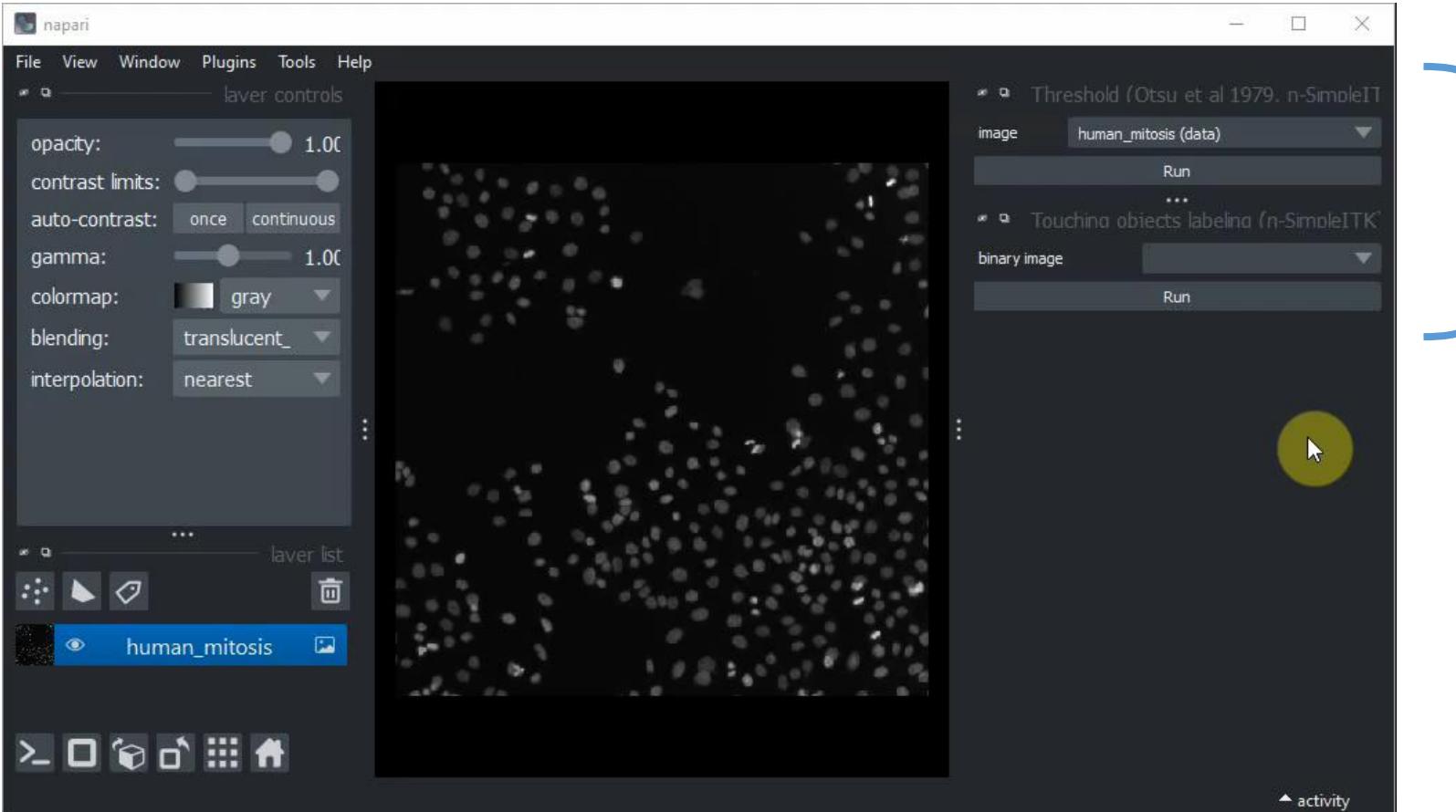


# Deconstruction

- Deconstruction is a method in software engineering to
  - understand how existing software works
  - prevent reinventing the wheel
  - identify limitations
  - identify bottlenecks
- Related methods
  - Reverse engineering
  - Code review
  - Pair programming
- In the age of computers writing code, reading code is a key skill.

# Deconstruction of napari-chatGPT

- Napari is a Python-based image viewer, extensible via plugins / “widgets”.



The top part shows the GitHub README page for the 'napari-simpleitk-image-processing' plugin. It includes the plugin's name, license (BSD-3-Clause), PyPI version (v0.4.5), supported Python versions (3.8 | 3.9 | 3.10), test status (tests passing), code coverage (codecov 99%), and GitHub status (alpha, napari hub). Below is a brief description: 'Process images using SimpleITK in napari'. The bottom part shows a screenshot of the graphical user interface for this plugin, which is a tabbed extension to the Napari interface. It includes tabs for 'Label', 'Filtration / denoising', 'Segmentation / binarization', and 'Segmentation / labeling'. The 'Label' tab shows a 3D volume with colored nuclei. The 'Segmentation / binarization' tab shows a 2D thresholded image labeled 'Result of Threshold ...'. The 'Segmentation / labeling' tab shows a 3D volume with colored nuclei labeled 'Result of Median (n-SimpleITK) ...'. The 'Label' tab has a 'Label' dropdown menu with options like 'Binary (n-SimpleITK)', 'Binarize (n-SimpleITK)', 'Curvature Flow (n-SimpleITK)', 'Gaussian (n-SimpleITK)', and 'Median (n-SimpleITK)'.

# Deconstruction of napari-chatGPT

- Napari-chatGPT can automate programming plugins / “widgets”.



Loïc A. Royer      
@loicaroyer

#ChatGPT + @napari\_imaging

Releasing my latest ~~weekend~~ project: Omega, an autonomous LLM agent that writes image processing and analysis code, fixes its mistakes, accesses the napari viewer, makes widgets, & more!  
[github.com/royerlab/napari...](https://github.com/royerlab/napari-chatgpt)  
@LangChainAI @OpenAI  
#OmegaAgent

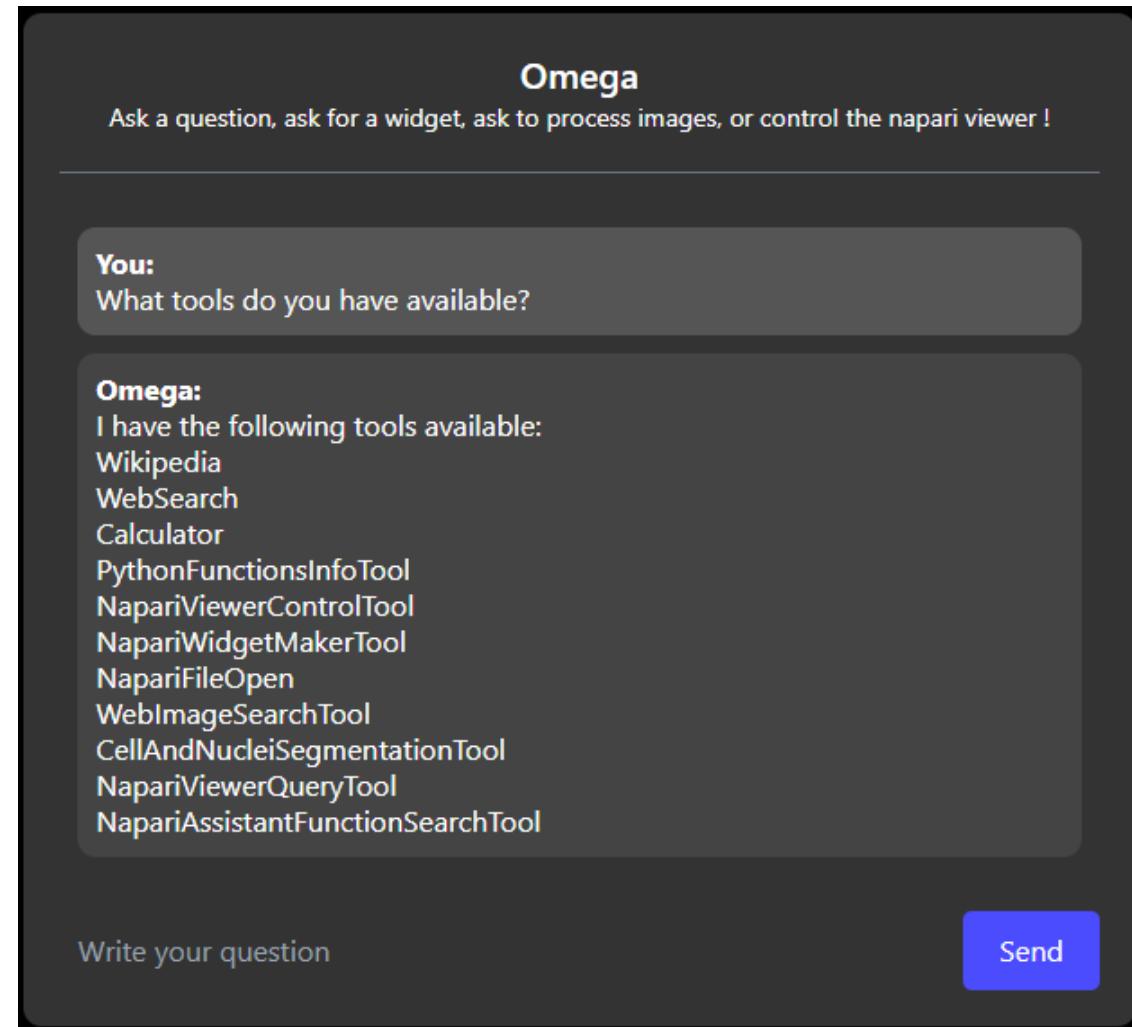
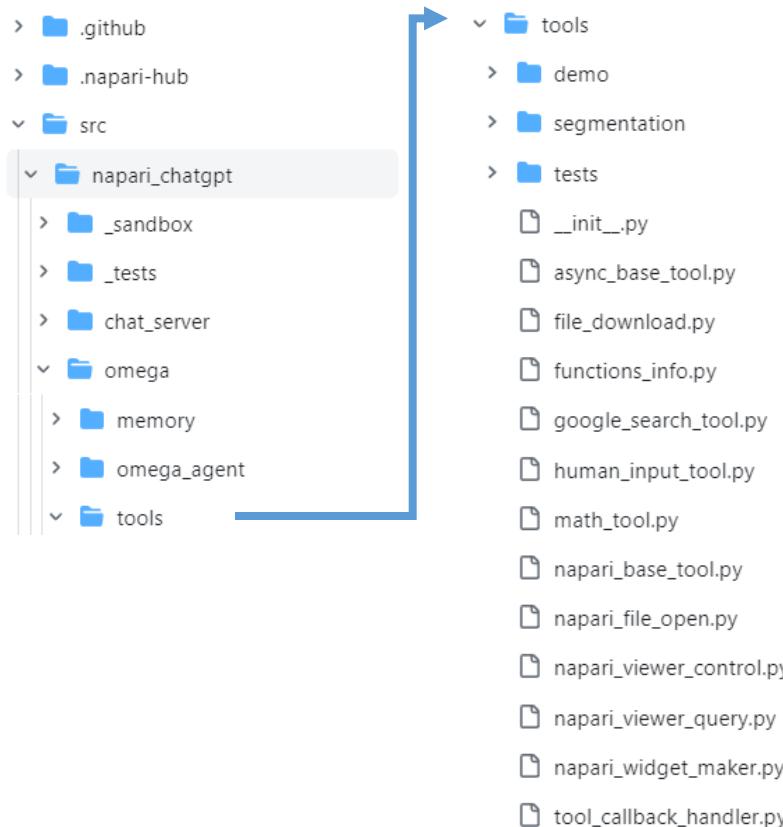


5:19 AM · May 3, 2023 · 183.6K Views

208 Retweets 44 Quotes 770 Likes 424 Bookmarks

# Deconstruction of napari-chatGPT

- Napari-chatGPT defines a list of “tools”



The screenshot shows a dark-themed AI interface titled "Omega". At the top, it says "Ask a question, ask for a widget, ask to process images, or control the napari viewer !". Below this, there's a "You:" input field containing "What tools do you have available?". The AI response, under "Omega:", lists various tools: Wikipedia, WebSearch, Calculator, PythonFunctionsInfoTool, NapariViewerControlTool, NapariWidgetMakerTool, NapariFileOpen, WebImageSearchTool, CellAndNucleiSegmentationTool, NapariViewerQueryTool, and NapariAssistantFunctionSearchTool. At the bottom, there's a "Write your question" input field and a blue "Send" button.

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- Napari-chatGPT defines a list of “tools”

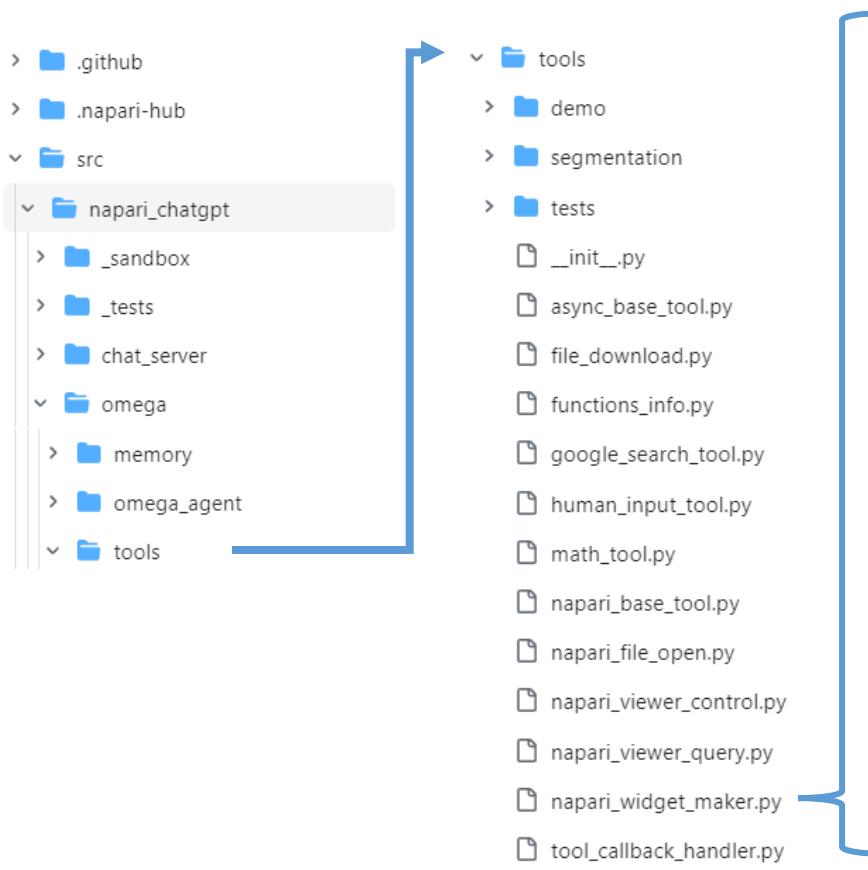
The image shows a file browser on the left and a code editor on the right. The file browser displays the directory structure of the napari-chatgpt project, with a focus on the tools folder. The code editor shows the implementation of the GoogleSearchTool and its search\_overview method.

```
1  from napari_chatgpt.omega.tools.async_base_tool import AsyncBaseTool
2  from napari_chatgpt.utils.google import search_overview
3
4
5  class GoogleSearchTool(AsyncBaseTool):
6      name = "GoogleSearch"
7      description = "Useful for when you need to answer questions by querying the web."
8
9      def _run(self, query: str) -> str:
10          """Use the tool."""
11          result = search_overview(query=query)
12          return result
13
14
15  def search_overview(query: str,
16                      num_results: int = 3,
17                      lang: str = "en",
18                      max_text_snippets: int = 5,
19                      max_query_freq_hz: float = 1.0) -> str:
20      url = f"https://www.google.com/search?q={query}&num={num_results}&hl={lang}"
21      text = text_from_url(url,
22                           max_text_snippets=max_text_snippets,
23                           max_query_freq_hz=max_query_freq_hz)
24
25      return text
```

[https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari\\_chatgpt/omega/tools/google\\_search\\_tool.py](https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari_chatgpt/omega/tools/google_search_tool.py)  
[https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari\\_chatgpt/utils/google.py#L58](https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari_chatgpt/utils/google.py#L58)  
Code origin: <https://github.com/Nv7-GitHub/googlesearch>

# Deconstruction of napari-chatGPT

- Napari-chatGPT defines a list of “tools”



```
10    _napari_widget_maker_prompt = """
11
12 Task:
13 You competently write image processing and image analysis functions in python given a plain text request.
14 The function should be pure, self-contained, effective, well-written, syntactically correct.
15 The function should work on 2D and 3D images, and images of any number of dimensions (nD),
16 unless the request is explicit about the number of dimensions.
17 The widget should do all and everything that is asked, but nothing else or superfluous.
18
19 Instructions for Function Signature:
20 - Integers, floats, boolean, or any other type accepted by the magicgui library.
21 - Decorate the function with the magicgui decorator: '@magicgui(call_button='Run')' where <function_name>
22 - DO NOT CREATE A NEW INSTANCE OF A NAPARI VIEWER, use the one provided in the variable: 'viewer'.
23 - DO NOT write code to add the widget to the napari window by calling viewer.window.add_dock_widget().
24
25 The function signature should have a type hint for the return, e.g. -> ImageData or -> Image
26
27 {generic_codegen_instructions}
28
29 Request:
30 {input}
```

Our prompt goes here

[https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari\\_chatgpt/omega/tools/napari\\_widget\\_maker.py#L10](https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari_chatgpt/omega/tools/napari_widget_maker.py#L10)

# Prompt engineering

- chatGPT can solve simple image analysis tasks

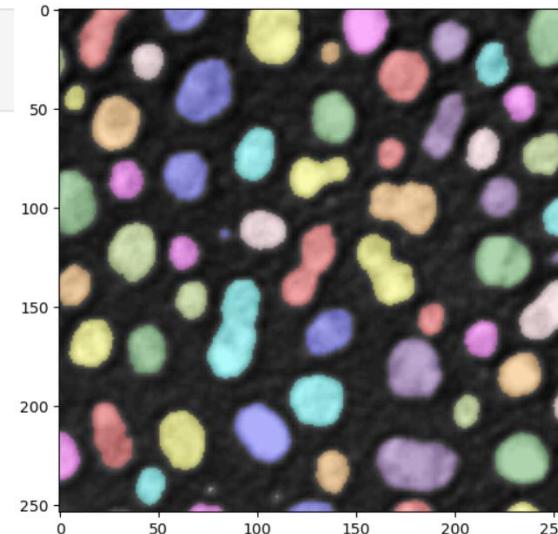
Prompt

```
simple_question = """  
Write Python code only and no additional explanatory text.  
  
Write a python program, that  
* loads the file `../../data/blobs.tif`,  
* labels objects in this image, and  
* visualize results.  
  
Assume this program would be executed in a Jupyter notebook.  
It is not necessary to save the results. Show the results in Jupyter.  
"""
```

Response

```
code = prompt(simple_question)  
print(code)  
  
```python  
from skimage.io import imread  
from skimage.filters import threshold_otsu  
from skimage.measure import label  
from skimage.color import label2rgb  
import matplotlib.pyplot as plt  
  
image = imread(' ../../data/blobs.tif')  
thresh = threshold_otsu(image)  
binary = image > thresh  
labels = label(binary)  
image_label_overlay = label2rgb(labels, image=image)  
  
fig, ax = plt.subplots(figsize=(10, 6))  
ax.imshow(image_label_overlay)  
  
plt.show()  
```
```

Result



# Prompt engineering

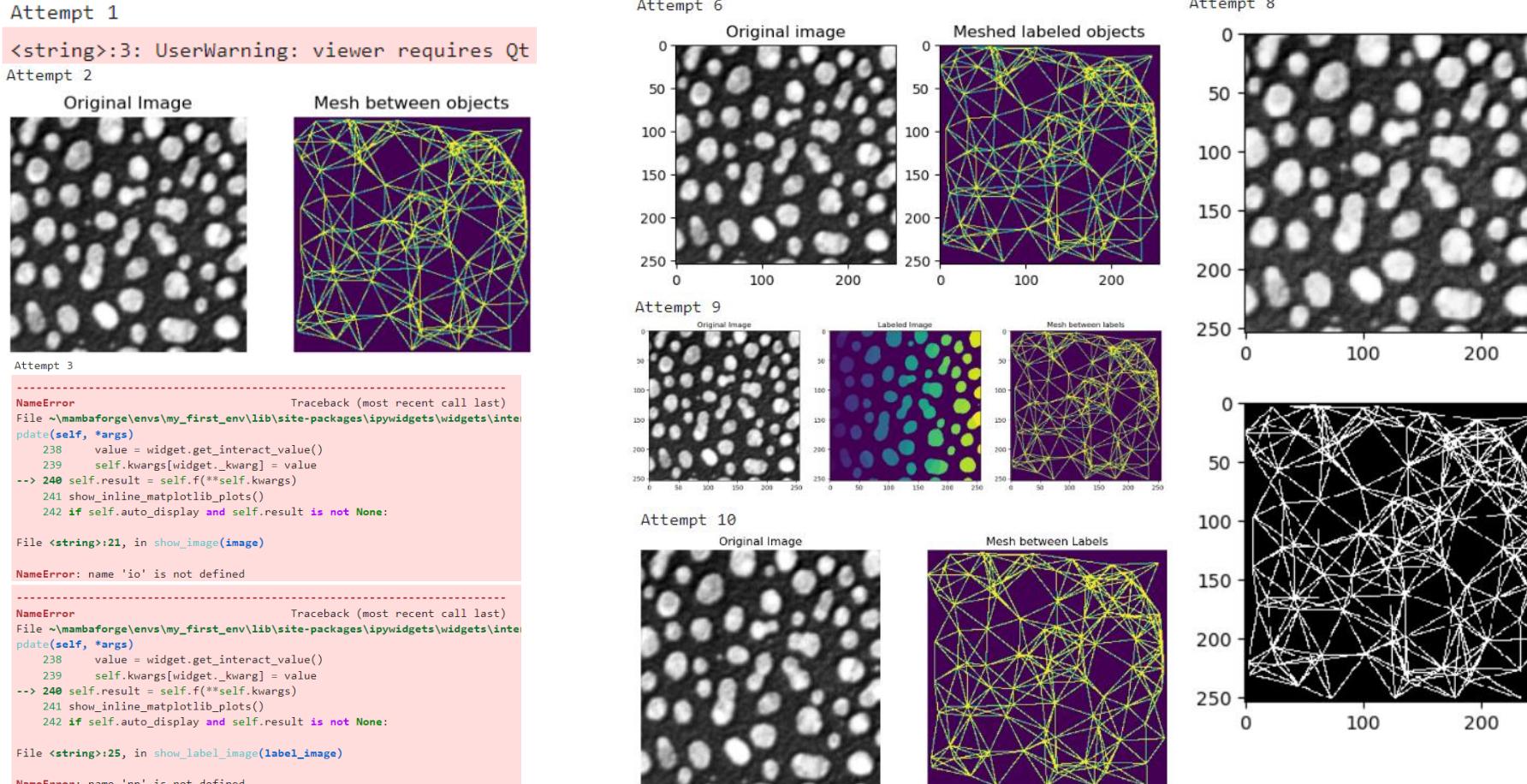
- With more advanced tasks, it might need additional help.

```
simple_question = """  
Write Python code only and no additional explanatory text.  
  
Write a python program, that  
* loads the file `../../data/blobs.tif`,  
* labels objects in this image,  
* and draws a mesh between labels with a maximum distance of 50 pixels.  
  
Assume this program would be executed in a Jupyter notebook.  
It is not necessary to save the results. Show the results in Jupyter.  
"""
```

```
more_sophisticated_question = """  
Please program some python code like a professional would.  
Write Python code only and no additional explanatory text.  
  
Write a python program, that  
* loads the file `../../data/blobs.tif`,  
* labels objects using voronoi-otsu-labeling,  
* and draws a mesh between labels with a maximum distance of 50 pixels.  
  
I have this code snippet for segmenting an image:  
import pyclesperanto_prototype as cle  
label_image = cle.voronoi_otsu_labeling(image)  
  
And this is the code snippet for drawing a mesh between objects in a label image:  
mesh = cle.draw_mesh_between_proximal_labels(labels, maximum_distance:int)  
  
Assume this program would be executed in a Jupyter notebook.  
It is not necessary to save the results. Show the results in Jupyter.  
"""
```

# Prompt engineering

- With more involved tasks, it might need additional help.



The *more sophisticated* prompt produced useful results in 5 out of 10 runs.

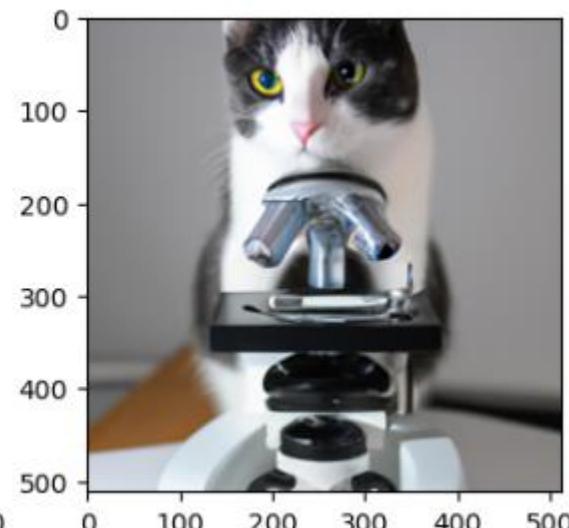
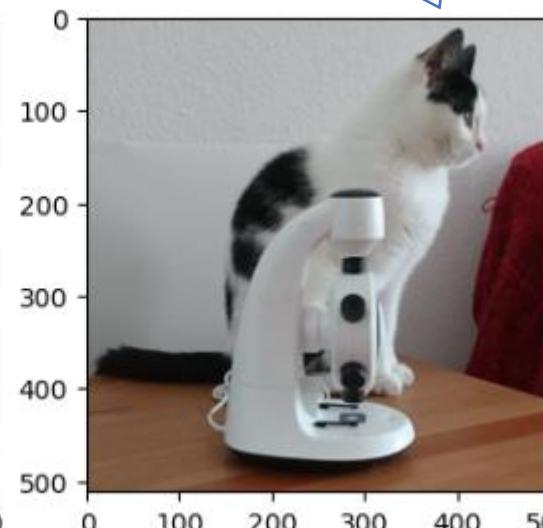
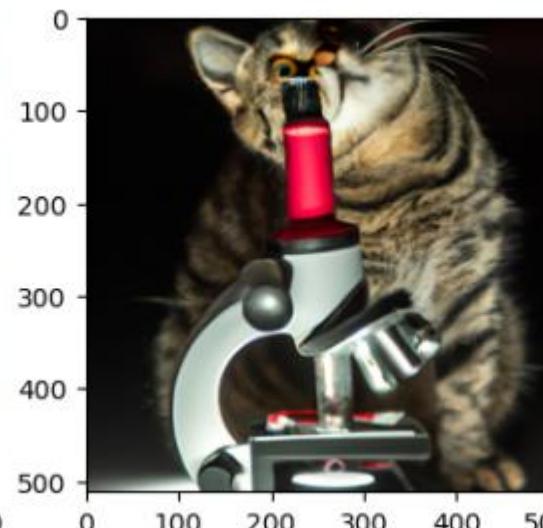
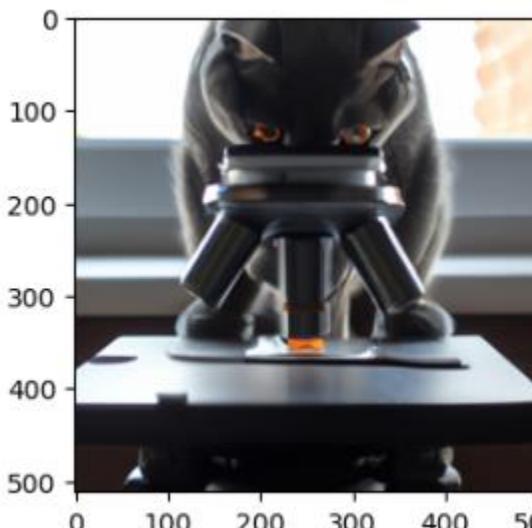
The *more sophisticated* prompt had errors in 4 out of 10 runs.

# Prompt engineering

- Prompts can be used to write code, but also to generate images, e.g. with DALL-E.  
One can generate quite realistic images given a detailed prompt.

```
cat_microscope_prompt = """  
Image of a cat sitting behind a microscope.  
"""
```

One cat  
is real.



Adapted from:

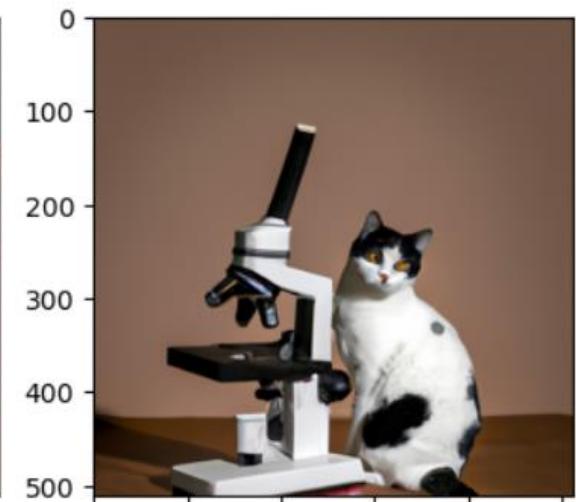
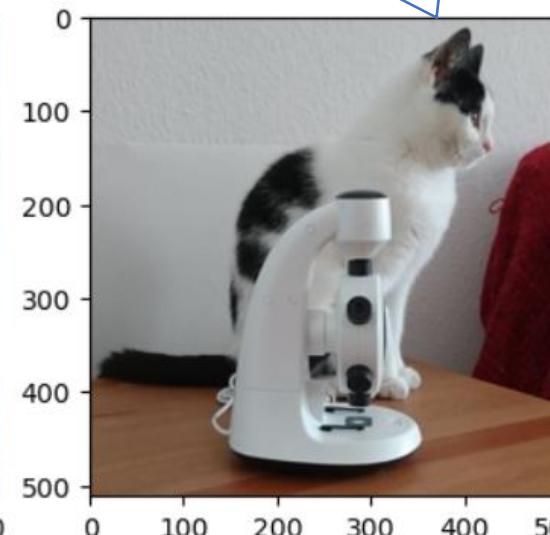
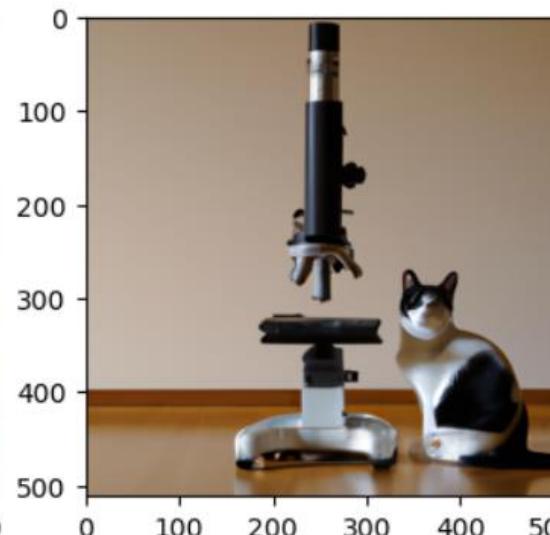
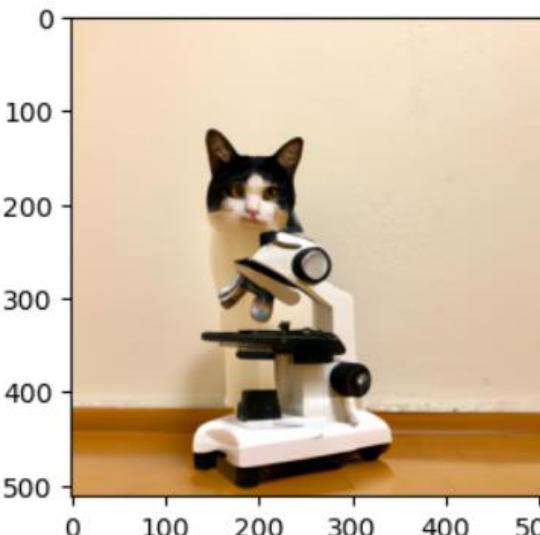
[https://haesleinhuepf.github.io/BioImageAnalysisNotebooks/07\\_prompt\\_engineering/02\\_generating\\_images.html](https://haesleinhuepf.github.io/BioImageAnalysisNotebooks/07_prompt_engineering/02_generating_images.html)

# Prompt engineering

- Prompts can be used to write code, but also to generate images, e.g. with DALL-E.  
One can generate quite realistic images given a detailed prompt.

```
[5]: cat_microscope_prompt = """  
Image of a cat sitting behind a microscope.  
Both are on a brown floor in front of a white wall.  
The cat is mostly white and has some black dots.  
The cat sits straight.  
The cat is a bit larger than the microscope.  
"""
```

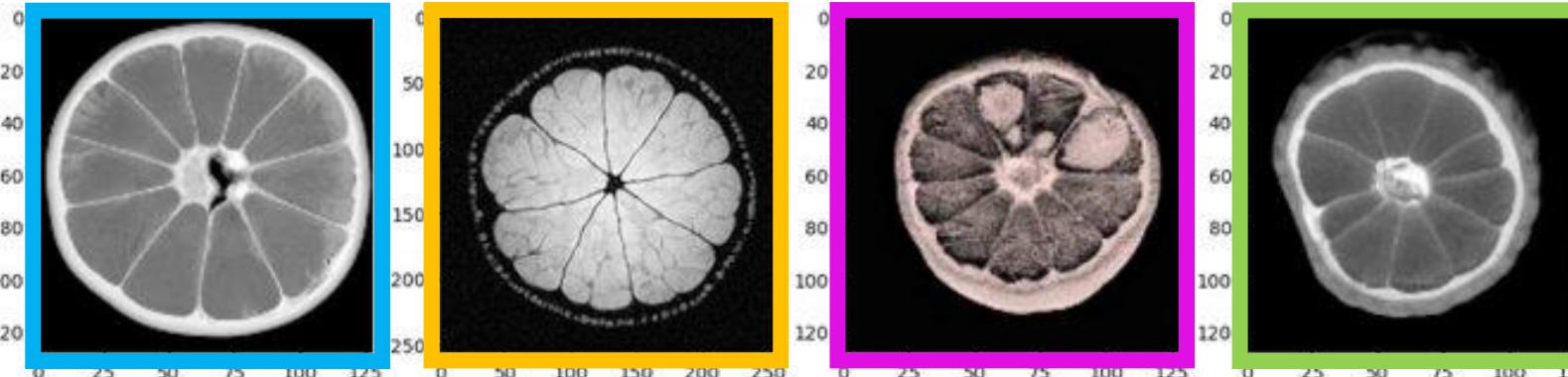
One cat  
is real.



# Prompt engineering

- Prompts can be used to write code, but also to generate images, e.g. with DALL-E.  
One can generate quite realistic images given a detailed prompt.

```
mri_prompt = """  
A single, high resolution, black-white image of  
a realistically looking orange fruit slice  
imaged with T2-weighted magnetic resonance imaging (MRI).  
"""
```



Robert Haase @haesleinhuepf · 20h

Fun poll time! Which of these images shows a real MRI image of an orange? (Credits: licensed CC-BY 4.0 by Alexandr Khrapichev, University of Oxford; the other images were generated by @openai's DALL-E)

Please vote below, RT and if you can explain why, please comment! 😊

Four MRI slices are shown in a grid. Below them is a poll with four options:

| Option | Votes | Percentage |
|--------|-------|------------|
| 1      | 1     | 16.8%      |
| 2      | 6     | 67.6%      |
| 3      | 1     | 3.4%       |
| 4      | 1     | 12.3%      |

179 votes · Final results

- Quality assurance results depend not only on the language model, but also on the knowledge of the observers.

# Deconstruction of napari-chatGPT

- LangChain is used to combine tools.
- It uses chatGPT under the hood.

```
def upper_case(text:str):
    return text.upper()

def reverse(text:str):
    return text[::-1]
```

```
tools = [
    Tool(
        name="Upper case",
        func=upper_case,
        description="Useful for making a text uppercase or capital letters."
    ),
    Tool(
        name="Reverse",
        func=reverse,
        description="Useful for making reversing order of a text."
    ),
]
```



⚡ Building applications with LLMs through composability ⚡

 lint passing  test passing  linkcheck passing  downloads/month 1M  License MIT

```
[4]: memory = ConversationBufferMemory(memory_key="c")
      llm=ChatOpenAI(temperature=0)
```

```
[5]: agent = initialize_agent(
    tools,
    llm,
    agent=AgentType.CHAT_CONVERSATIONAL_REACT_DESCRIPTION,
    memory=memory
)
```

# Deconstruction of napari-chatGPT

- After combining tools, large language model and memory in an *agent*, you can interact with it.

```
[6]: agent.run(input="Hi, I am Robert")
```

```
[6]: 'Nice to meet you, Robert!'
```

```
[7]: agent.run(input="What's my name?")
```

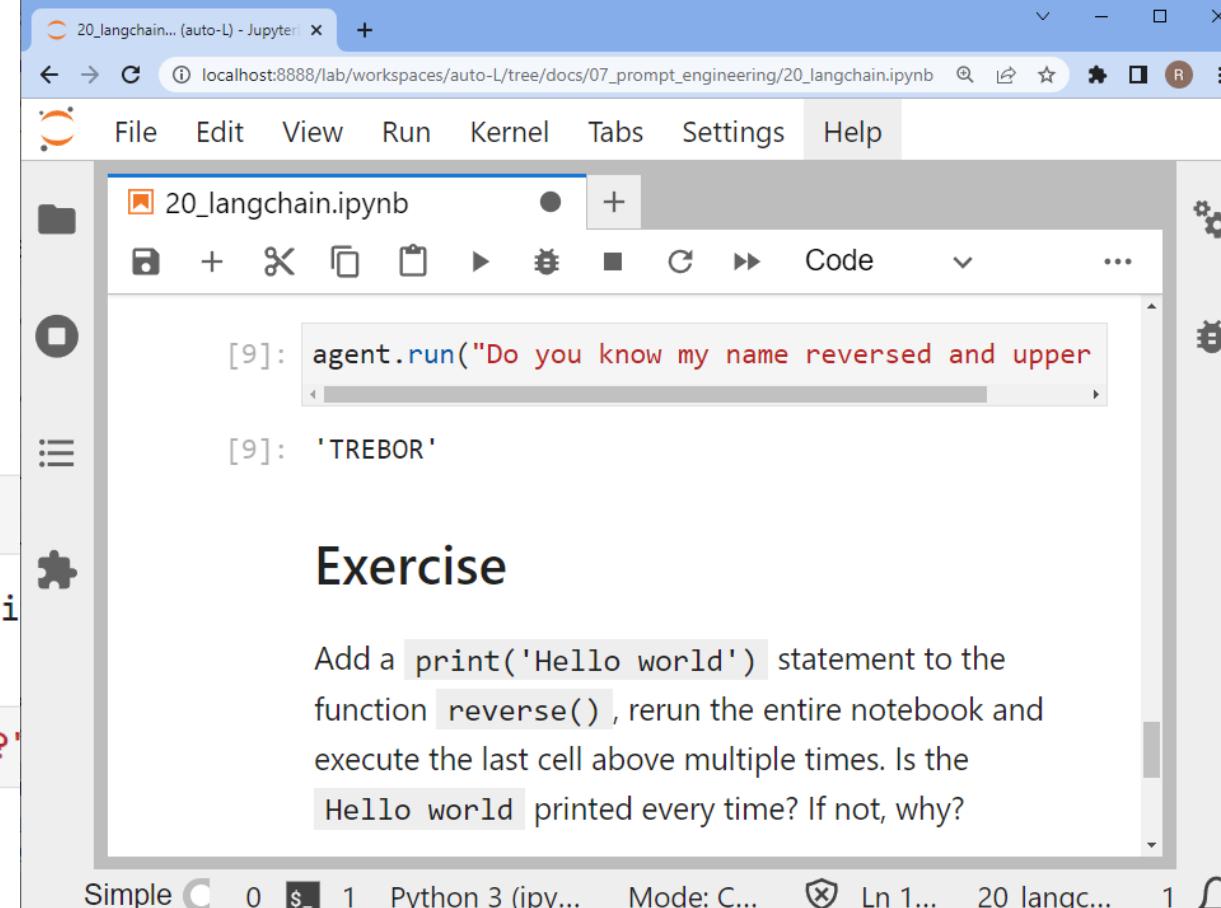
```
[7]: 'Your name is Robert'
```

```
[8]: agent.run("Can you reverse my name?")
```

```
[8]: "The response to your last comment was 'treboR', which is"
```

```
[9]: agent.run("Do you know my name reversed and upper case?")
```

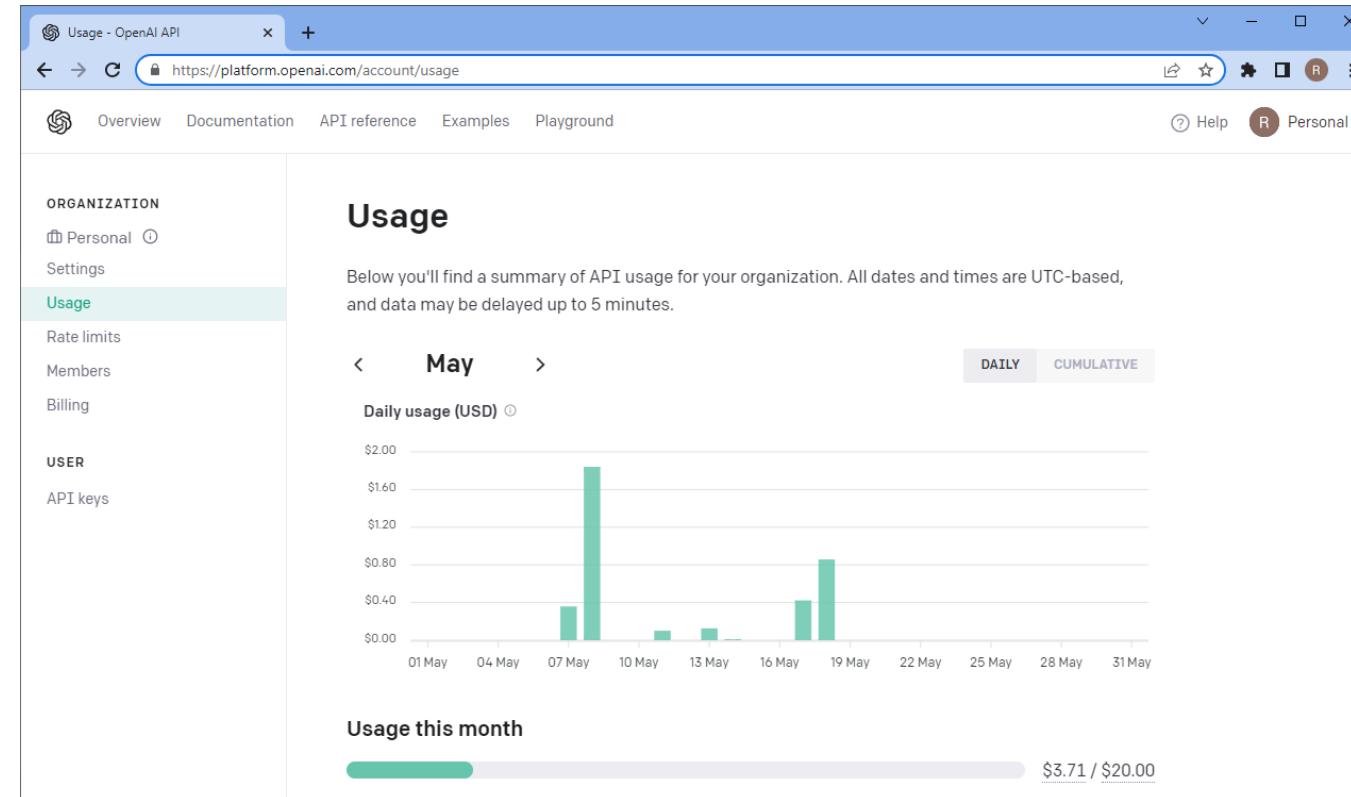
```
[9]: 'TREBOR'
```



The screenshot shows a Jupyter Notebook window titled '20\_langchain.ipynb' running on 'localhost:8888'. The notebook contains several code cells and their outputs. Cell [6] outputs 'Nice to meet you, Robert!'. Cell [7] outputs 'Your name is Robert'. Cell [8] outputs the message 'The response to your last comment was 'treboR', which is'. Cell [9] outputs 'TREBOR'. A new cell [9] is currently being edited with the command 'agent.run("Do you know my name reversed and upper case?")'. The output of this cell is 'TREBOR'. Below the notebook, there is a section titled 'Exercise' with instructions: 'Add a `print('Hello world')` statement to the function `reverse()`, rerun the entire notebook and execute the last cell above multiple times. Is the `Hello world` printed every time? If not, why?' The status bar at the bottom shows 'Simple C 0 \$ 1 Python 3 (ipy... Mode: C... 20\_langc... 1 L'.

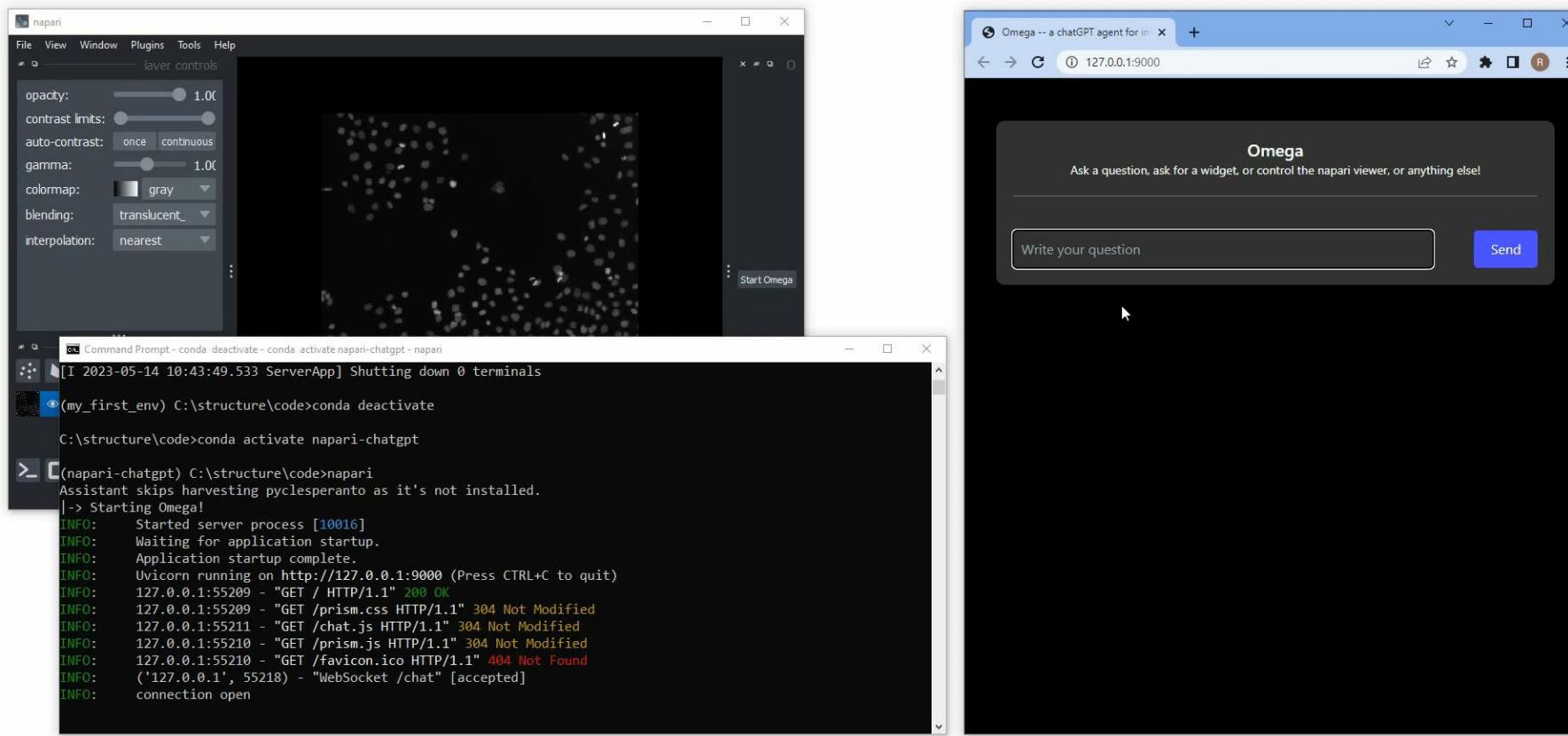
# How much is the fish?

- Executing DALL-E via Python, napari-ChatGPT and LangChain requires an OpenAI account.
- Using OpenAI infrastructure costs real money.



# A little warning by the end

- napari-chatGPT executes code and installs software on your machine.
- Use it with care! E.g. in a virtual machine / sandbox



# Generative Artificial Intelligence

- Challenges
  - Data safety / security
  - Computational cost of training neural networks
    - CO2-footprint/climate change
    - Accessibility
  - Bias: “a nice photo of a human”
  - Hallucinations
  - Glitch tokens
  - False information / fake news
  - Debugging is hard



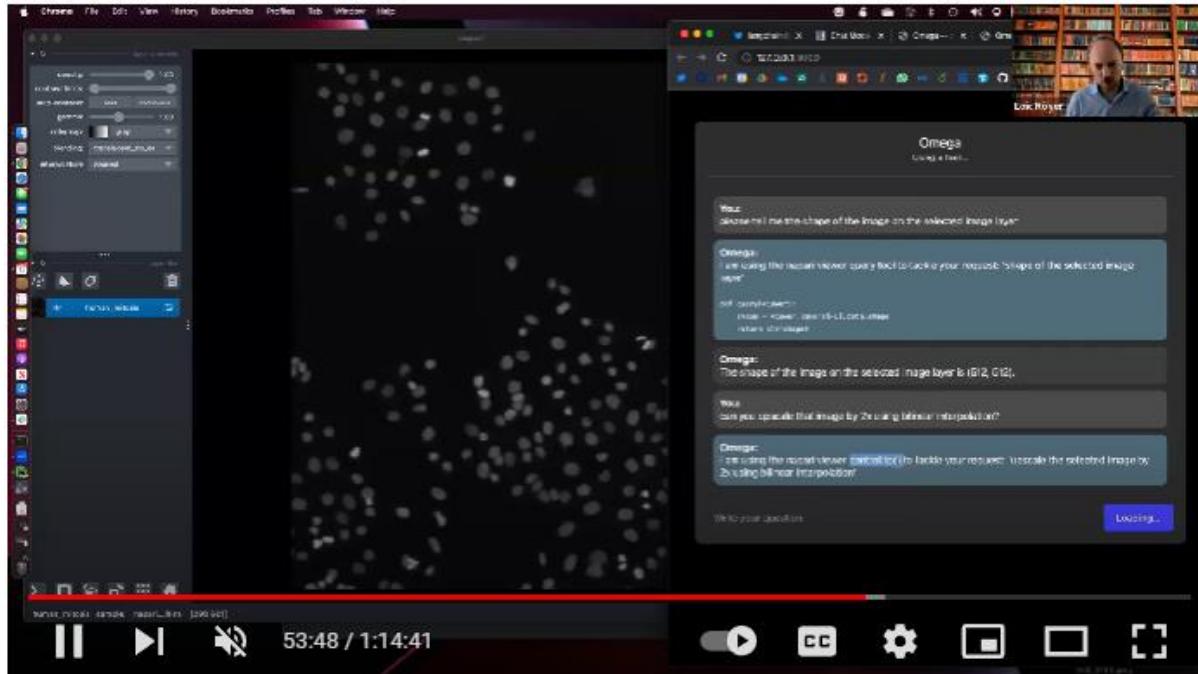
Model: Default (GPT-3.5)

R Who developed Gradoop?



Gradoop is an open-source framework for scalable graph analytics. It was developed by the [Hasso Plattner Institute \(HPI\) in Potsdam, Germany](#). The main contributors to Gradoop are researchers and developers from the HPI's Database Systems and Information Management (DIMA) group, led by [Prof. Dr. Felix Naumann](#). The project started in 2015 as part of the HPI's research on large-scale graph data management and analysis. Since then, it has gained popularity among researchers and industry professionals working with graph data.

# Further reading / watching



Code review of Napari-ChatGPT by Loic Royer (CZI Biohub)



Scientific Computing ...  
122 subscribers

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Like 6

Dislike

Share

...

[https://www.youtube.com/watch?v=JMo6Sn-L\\_j4](https://www.youtube.com/watch?v=JMo6Sn-L_j4)



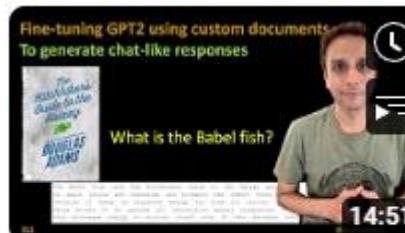
DigitalSreeni

@DigitalSreeni 71.3K subscribers 380 videos

This channel walks you through the entire process of learning to code in P ... >



308 - An introduction to language models with focu...



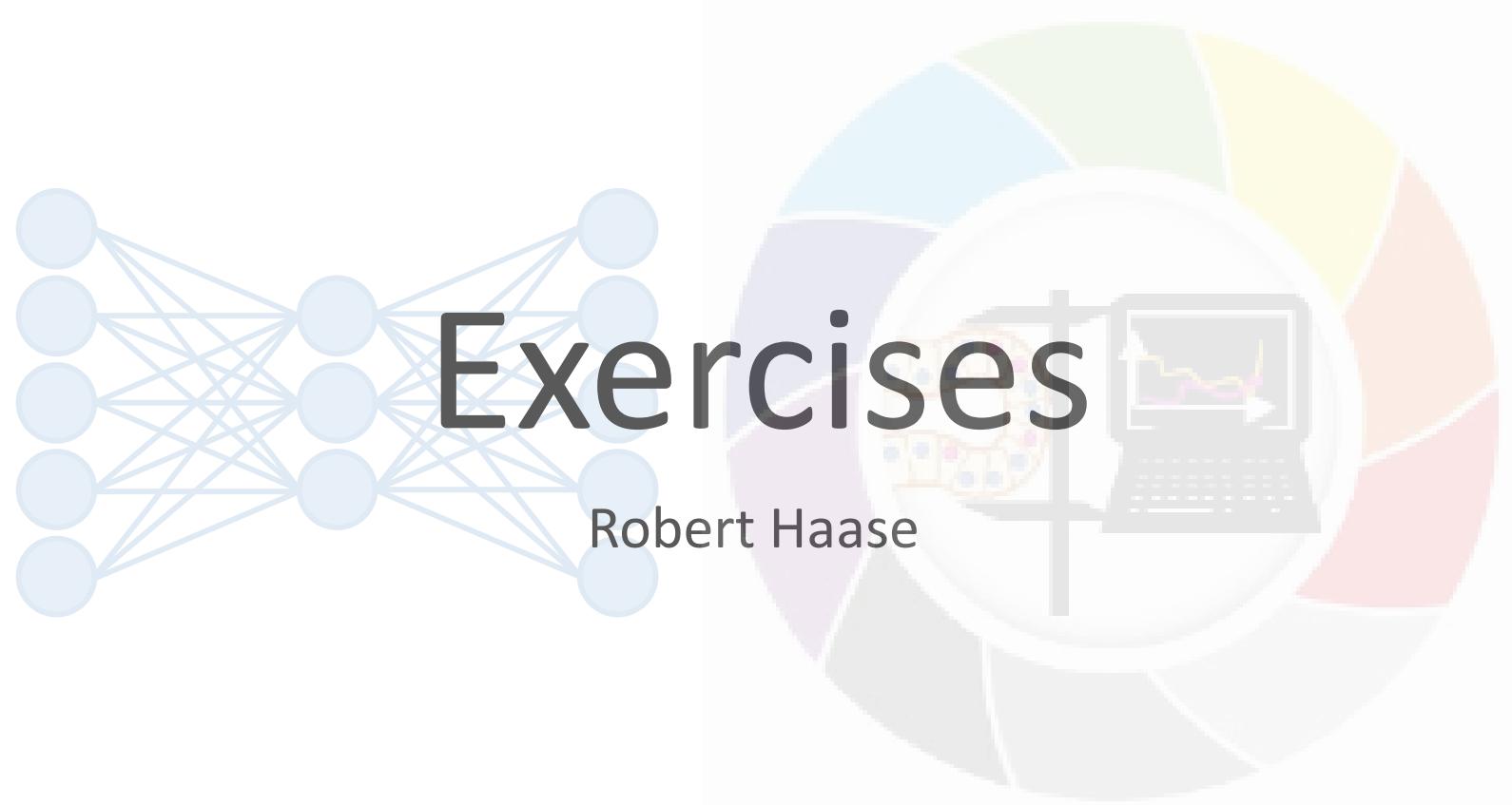
311 - Fine tuning GPT2 using custom documents



309 - Training your own Chatbot using GPT



<https://www.youtube.com/c/digitalsreeni>



# Deep-learning based nuclei segmentation

Use CellPose and StarDist to segment the nuclei in the human-mitosis example dataset of scikit-image.

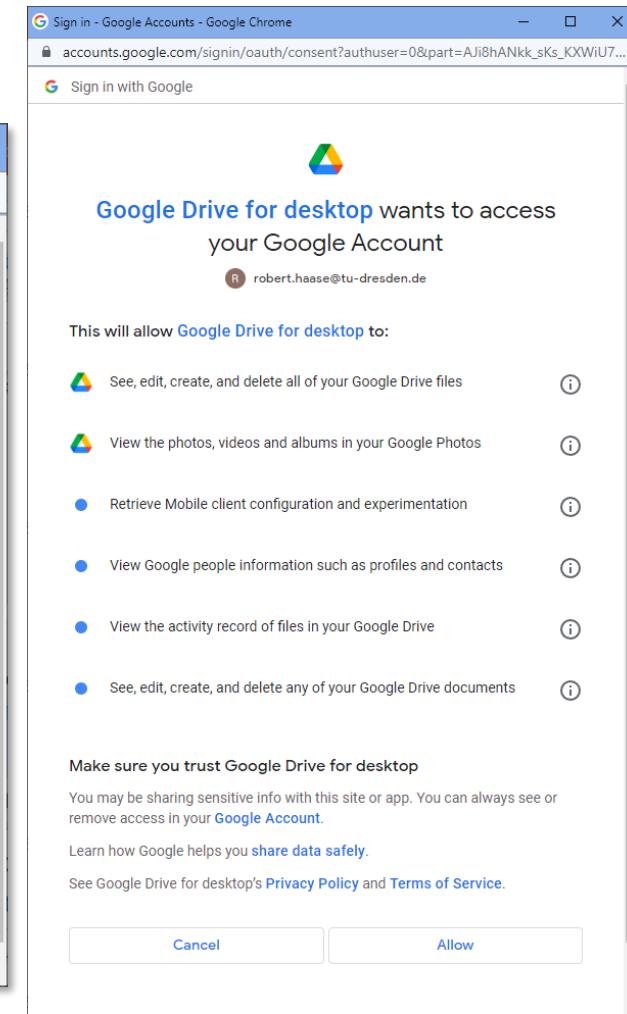
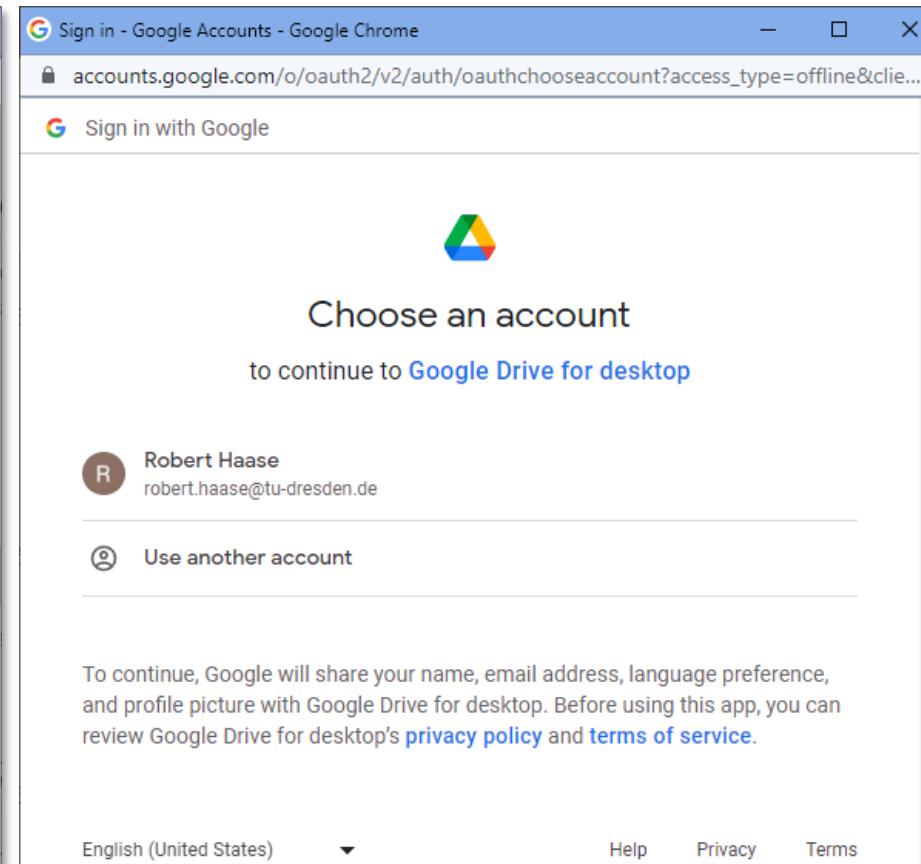
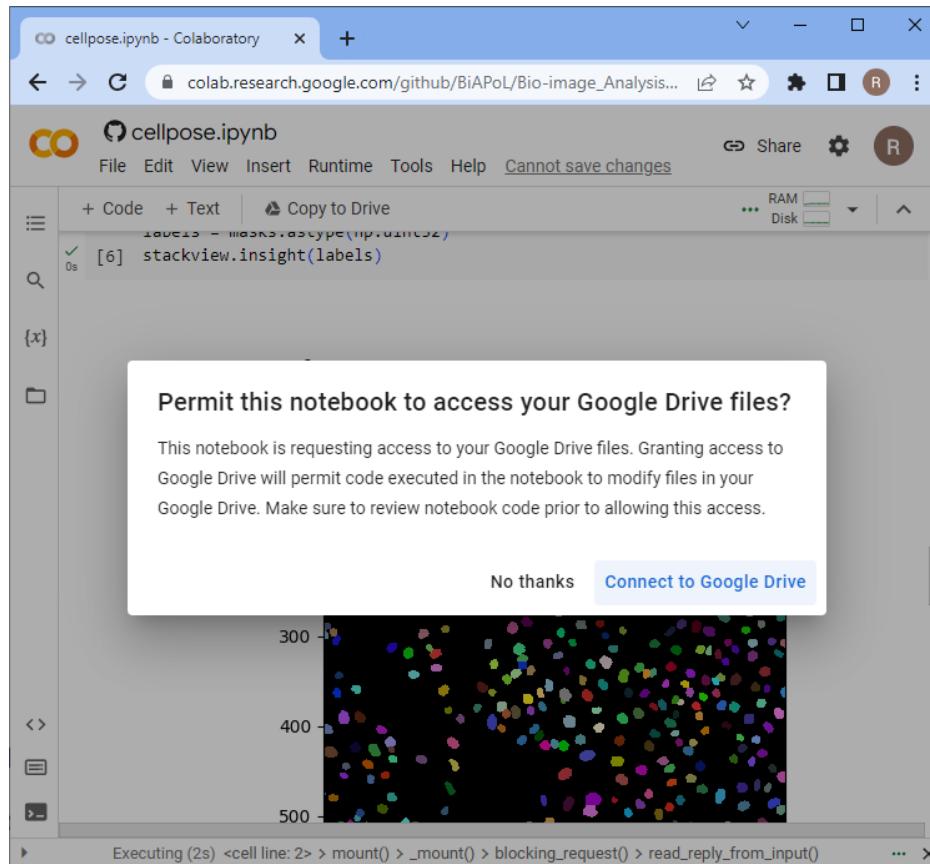
- [https://colab.research.google.com/github/BiAPoL/Bio-image Analysis with Python/blob/main/11 deep learning/stardist.ipynb](https://colab.research.google.com/github/BiAPoL/Bio-image_Analysis_with_Python/blob/main/11_deep_learning/stardist.ipynb)
- [https://colab.research.google.com/github/BiAPoL/Bio-image Analysis with Python/blob/main/11 deep learning/cellpose.ipynb](https://colab.research.google.com/github/BiAPoL/Bio-image Analysis with Python/blob/main/11_deep_learning/cellpose.ipynb)

This screenshot shows two Google Colab notebooks side-by-side. The left notebook, titled 'stardist.ipynb', contains code for 'Image segmentation with StarDist'. It includes a brief introduction to the method, a list of references, and a section on installing necessary Python libraries. The right notebook, also titled 'stardist.ipynb', shows the execution of a cell containing code to predict instances from an image and display the results using stackview.insight. A scatter plot of colored dots representing segmented nuclei is visible.

This screenshot shows two Google Colab notebooks side-by-side. The left notebook, titled 'cellpose.ipynb', contains code for 'Image Segmentation with CellPose'. It includes a brief introduction to the algorithm, a list of references, and a section on installing necessary Python libraries. The right notebook, also titled 'cellpose.ipynb', shows the execution of a cell containing code to evaluate a model on an image and display the results using stackview.insight. A scatter plot of colored dots representing segmented nuclei is visible.

# Connecting Google Drive

- Store the resulting label images to your Google Drive



# Segmentation quality comparison

- Download the segmentation results of CellPose and StarDist from your Google Drive.
- Measure the quality of both compared to a sparse annotation.
- Which algorithm is better on this one image?

