

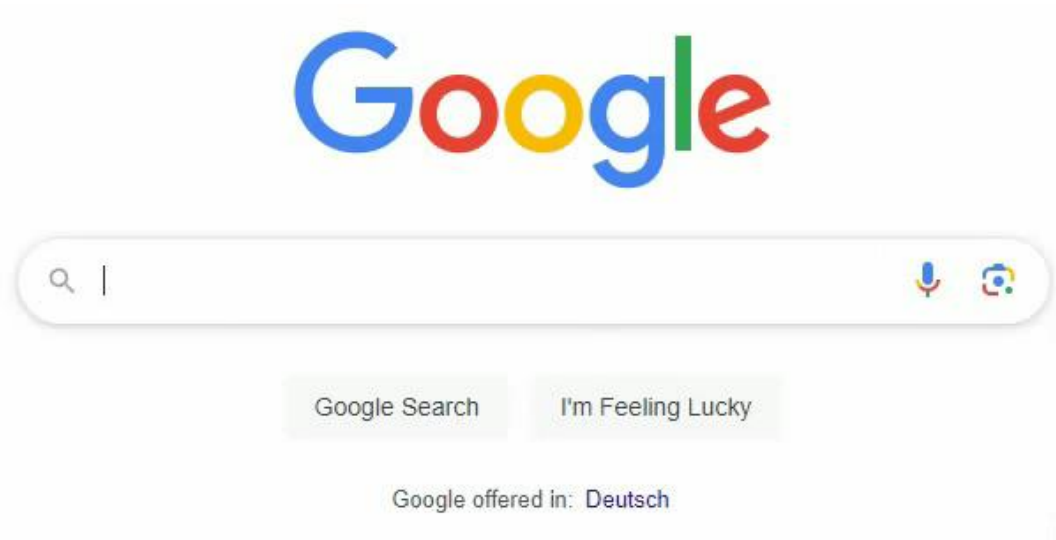
Deep Learning + Large Language Models for Bio-image Analysis

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)

- How it started

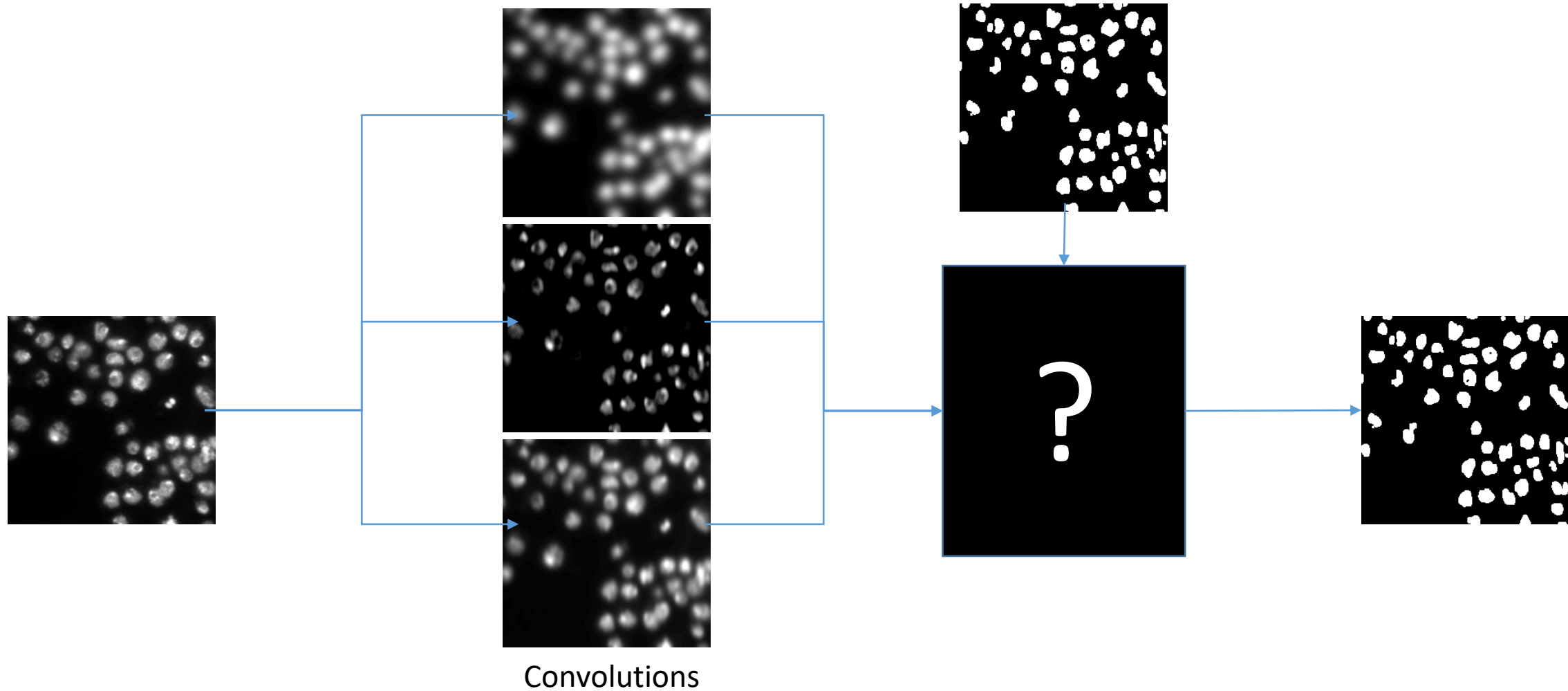


- How it's going

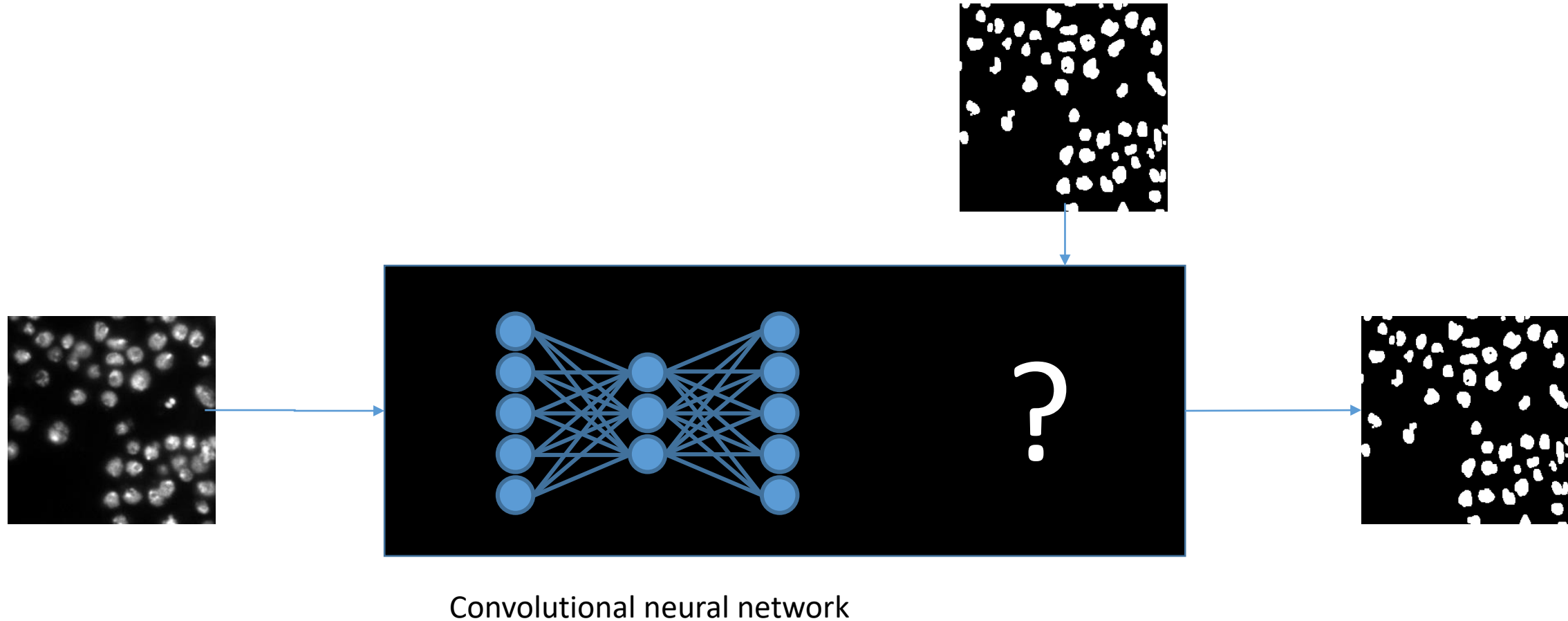
(GitHub copilot)

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

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

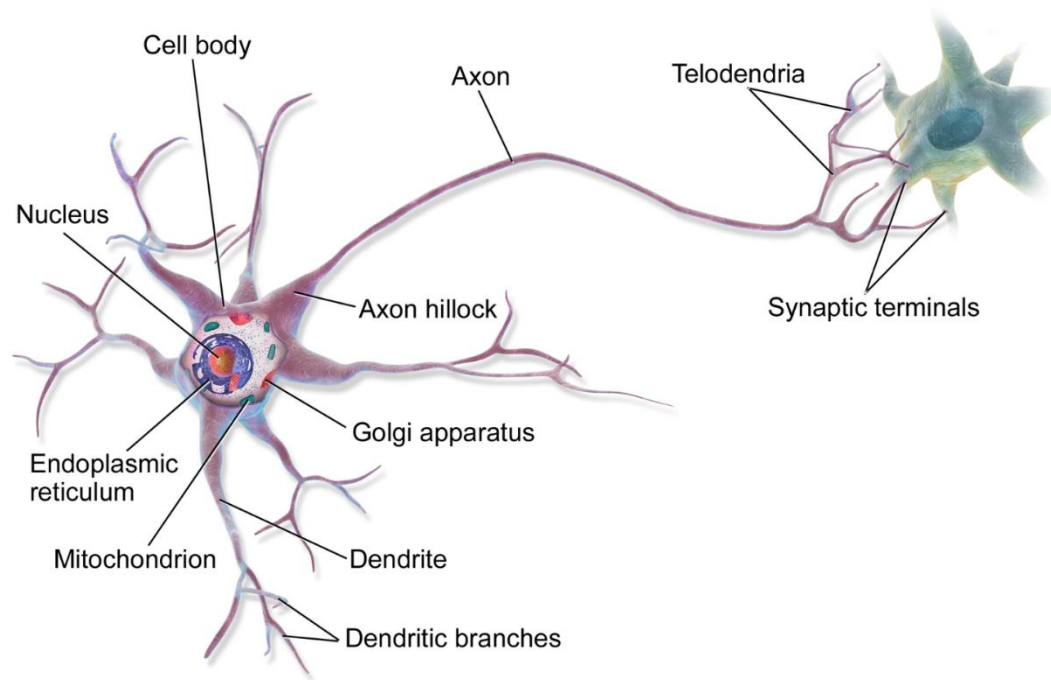


- In deep learning, this selection becomes part of the black box

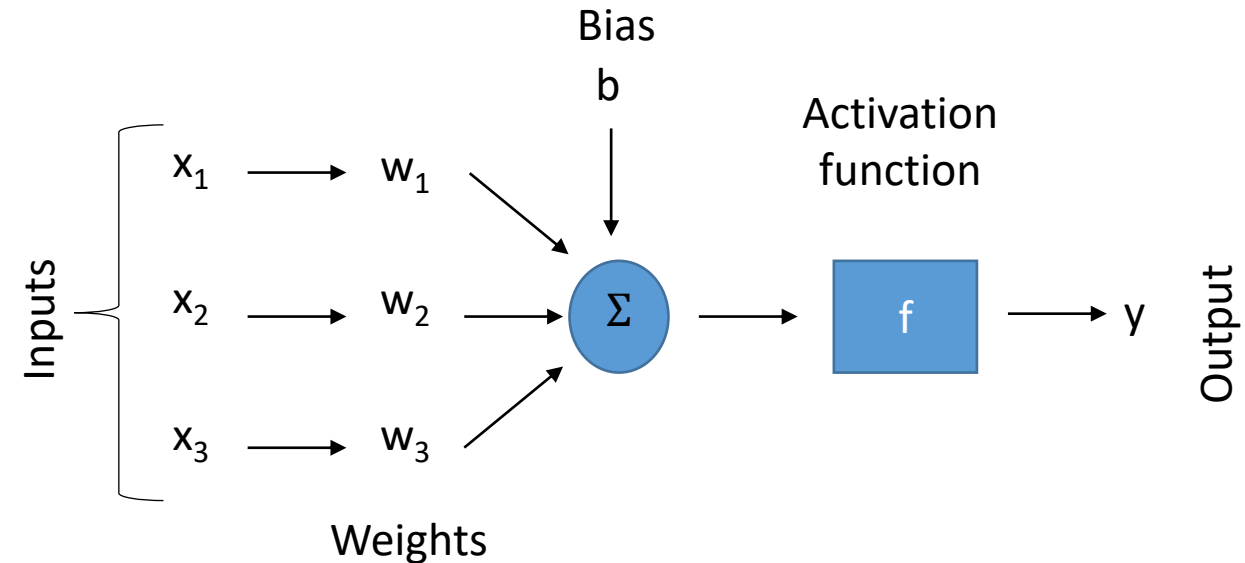


Neural networks

- How biologists see neurons



- How computer scientists see neurons
“perceptron”



$$y = f(w_1x_1 + w_2x_2 + w_3x_3 + b)$$

Neuron image source:

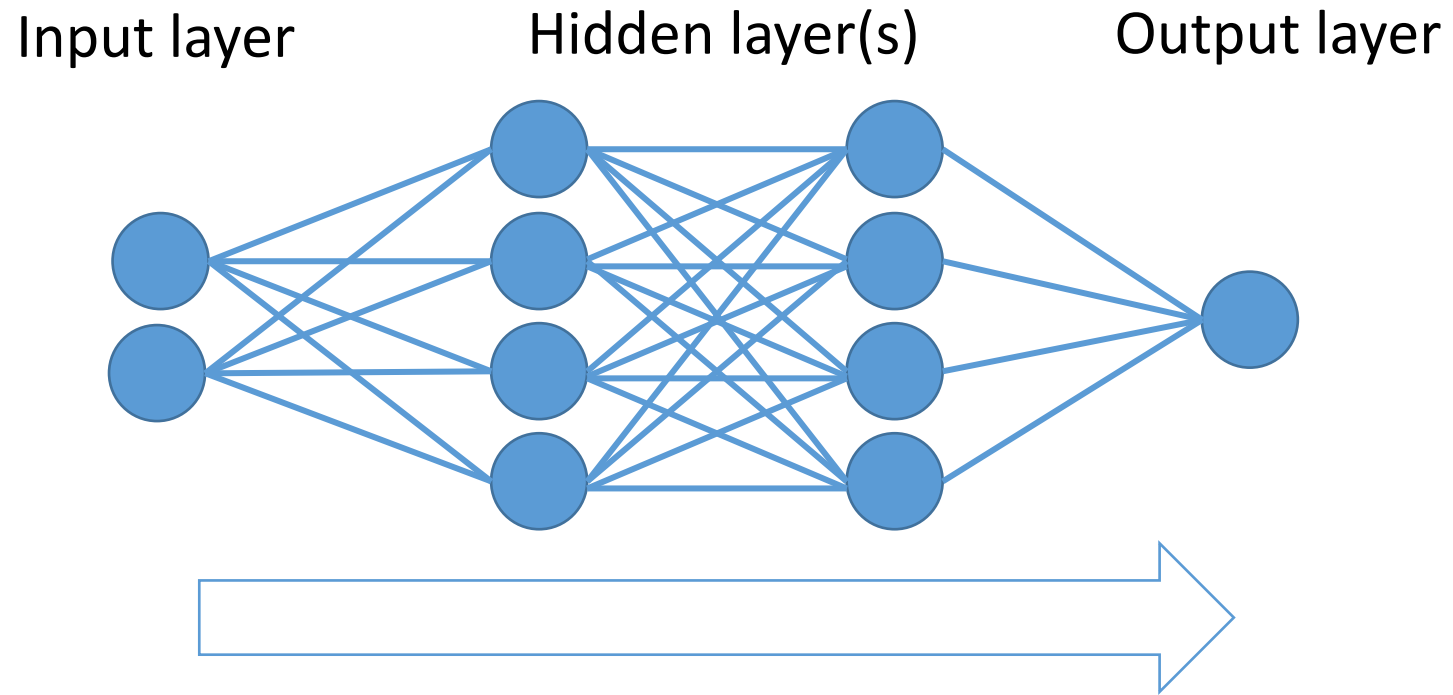
https://commons.wikimedia.org/wiki/File:Blausen_0657_MultipolarNeuron.png

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Neural Networks

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

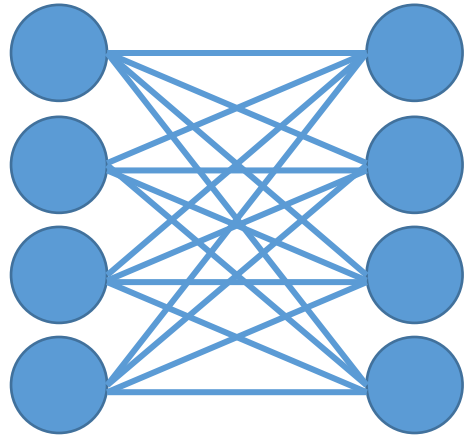
If there are *many* hidden layers, we speak of a *deep* neural network



Convolutional neural networks

- Layer types

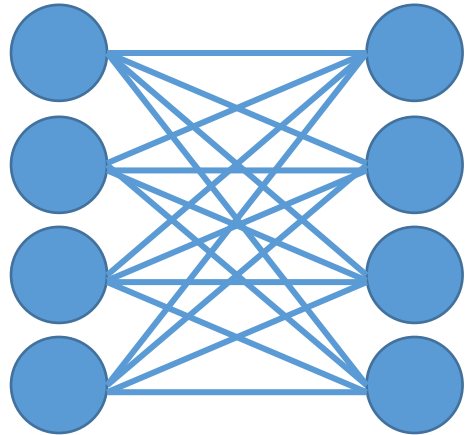
Fully connected layer



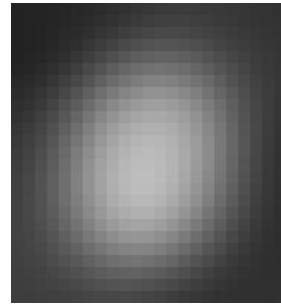
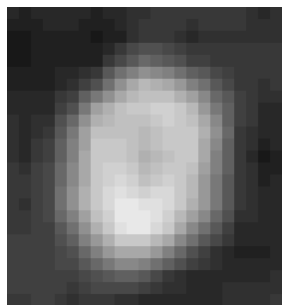
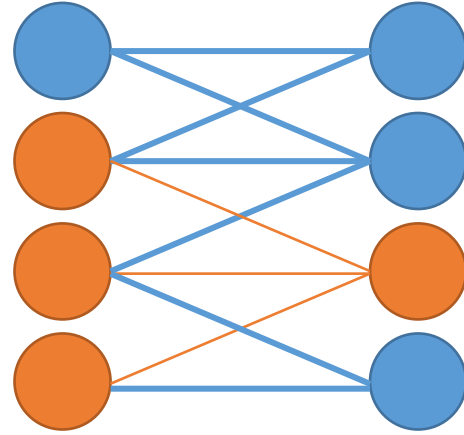
Convolutional neural networks

- Layer types

Fully connected layer



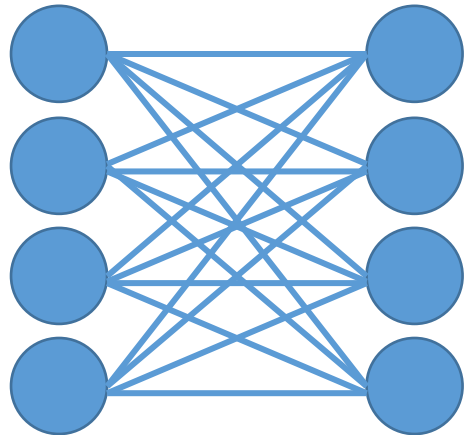
Convolutional layer



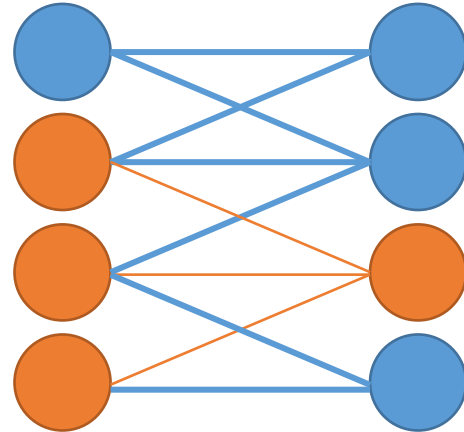
Convolutional neural networks

- Layer types

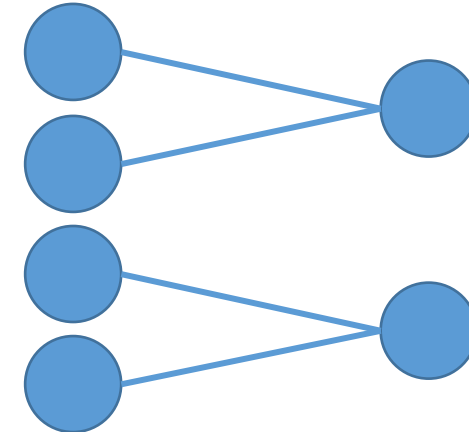
Fully connected layer



Convolutional layer



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



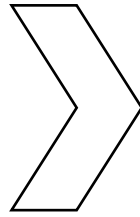
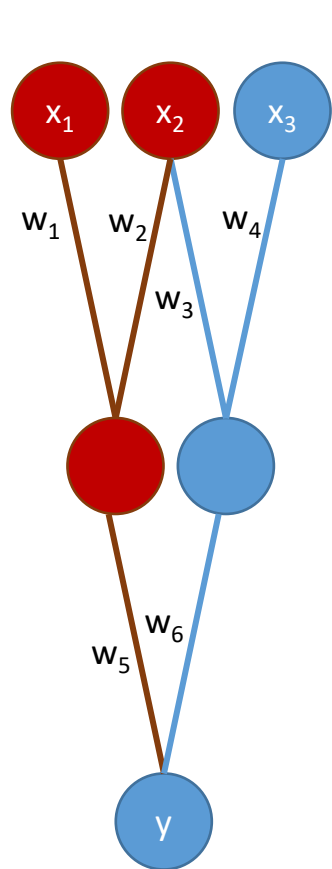
3	15	1	13
9	7	0	10
11	5	5	3
1	8	9	6

Max pooling

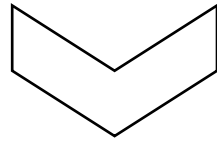
15	13
11	9

Convolutional neural networks

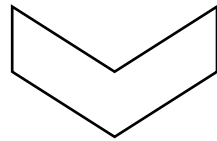
- Assuming we had no activation functions in the network layers can be reduced by eliminating brackets!



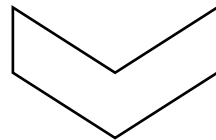
$$y = w_5(w_1x_1 + w_2x_2) + w_6(w_3x_2 + w_4x_3)$$



$$y = w_5w_1x_1 + w_5w_2x_2 + w_6w_3x_2 + w_6w_4x_3$$



$$y = w_5w_1x_1 + (w_5w_2 + w_6w_3)x_2 + w_6w_4x_3$$

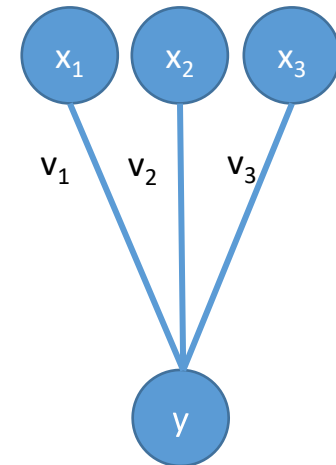


$$v_1 = w_5w_1$$

$$v_2 = w_5w_2 + w_6w_3$$

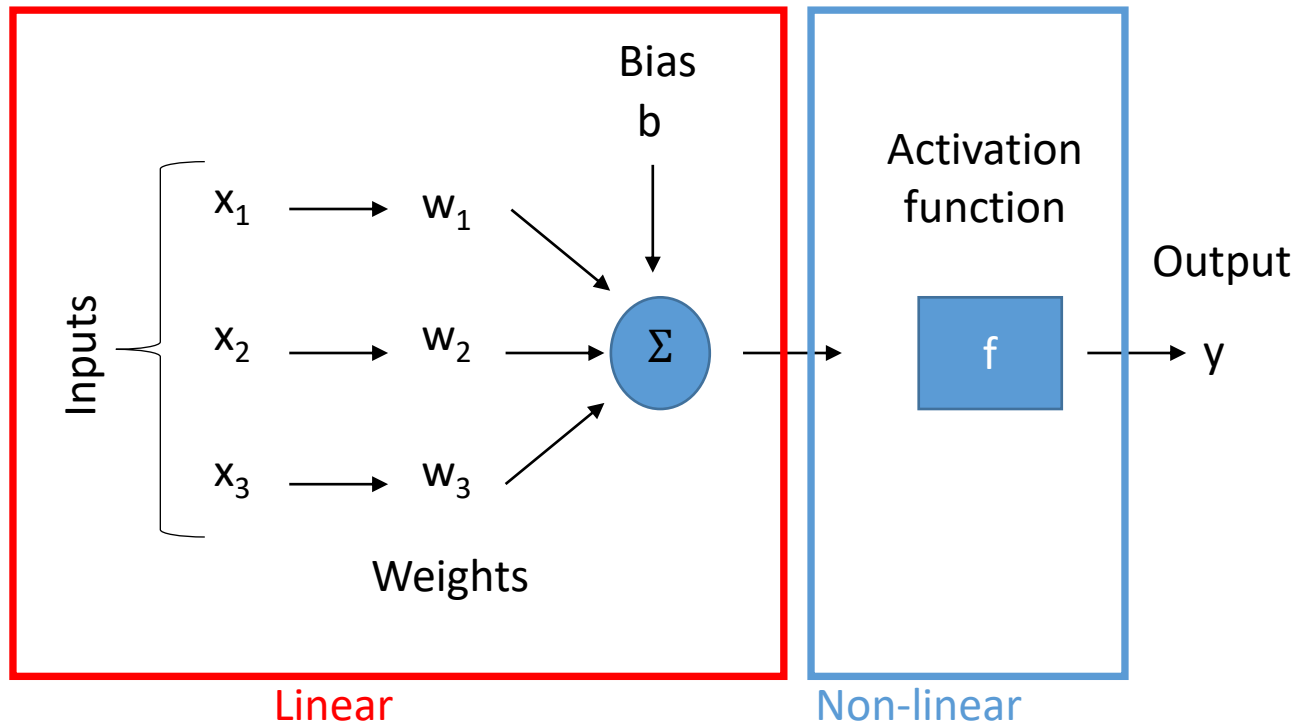
$$v_3 = w_6w_4$$

$$y = v_1x_1 + v_2x_2 + v_3x_3$$



Activation functions

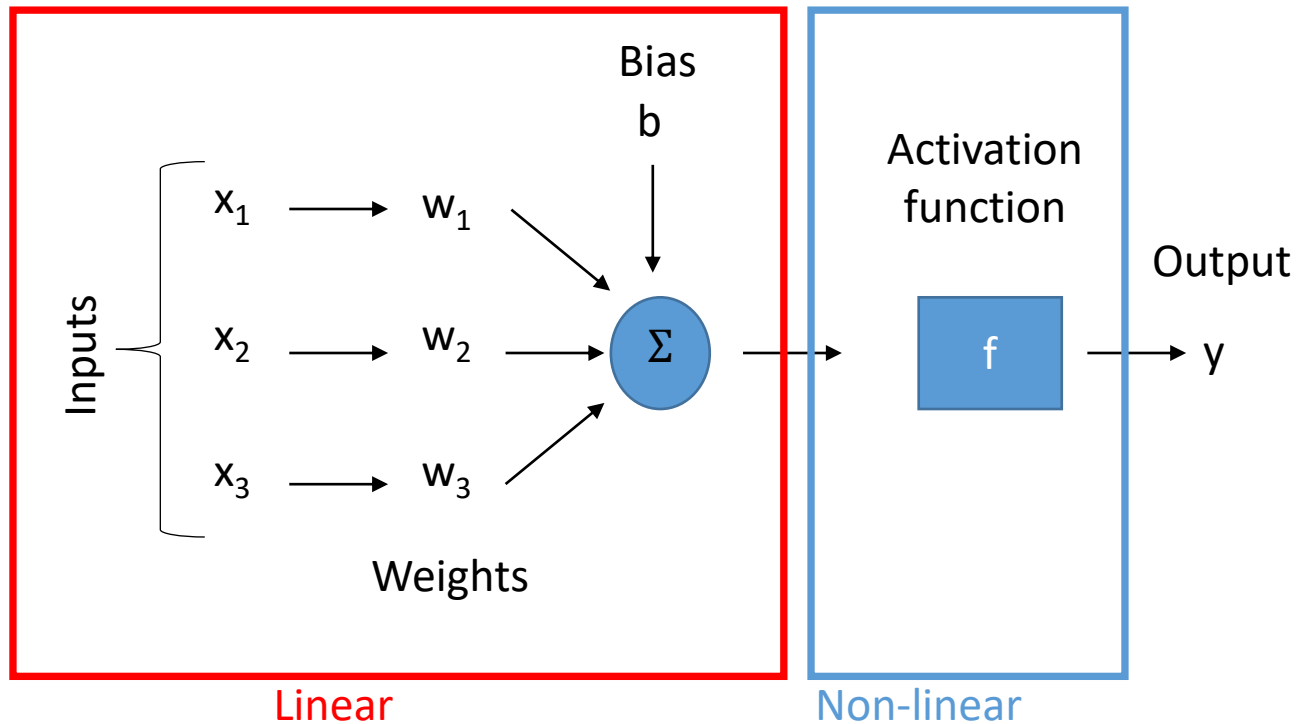
- Introduction of *non-linearity* and *activation functions* enabled what we call *deep-learning* today.

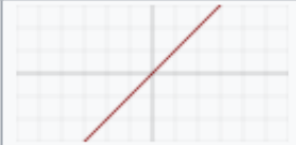





$$y = f(w_1x_1 + w_2x_2 + w_3x_3 + b)$$

Activation functions

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

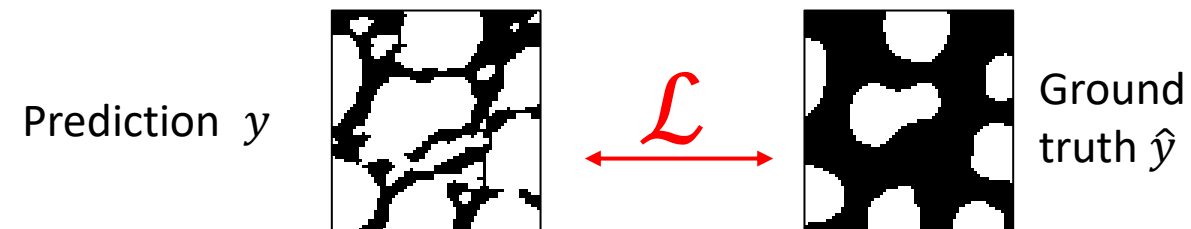
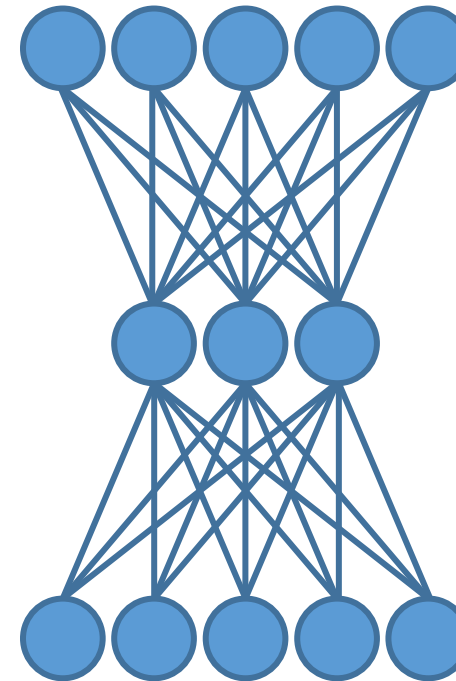
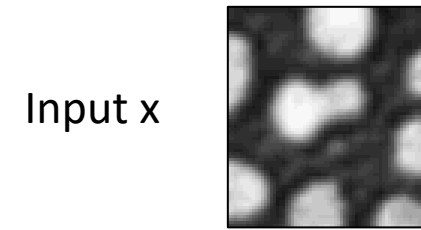


Identity		x
Binary step		$\begin{cases} 0 & \text{if } x < 0 \\ 1 & \text{if } x \geq 0 \end{cases}$
Logistic, sigmoid, or soft step		$\sigma(x) \doteq \frac{1}{1 + e^{-x}}$
Rectified linear unit (ReLU) ^[8]		$(x)^+ \doteq \begin{cases} 0 & \text{if } x \leq 0 \\ x & \text{if } x > 0 \end{cases}$ $= \max(0, x) = x \mathbf{1}_{x>0}$

- 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



The loss function can be expanded from

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

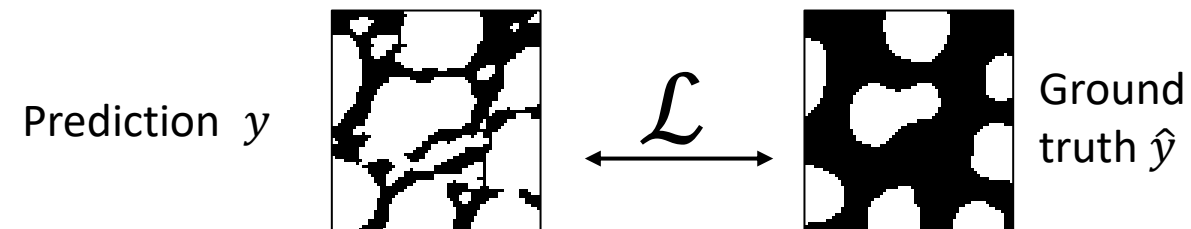
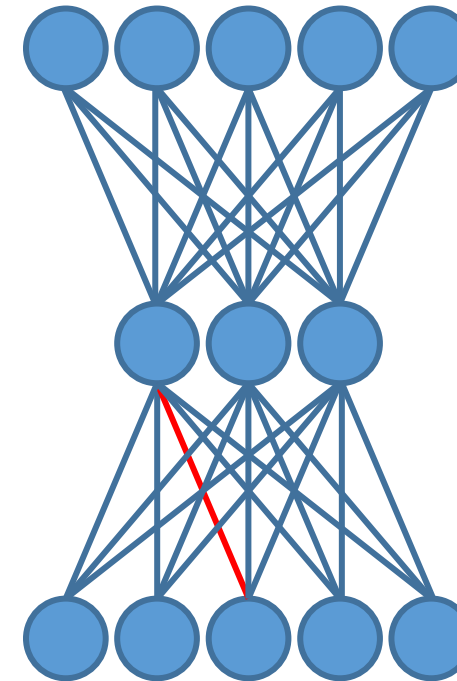
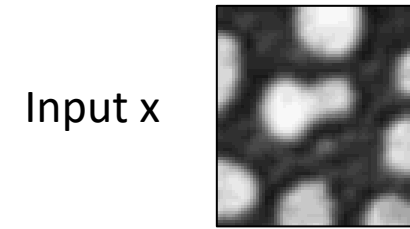
as the prediction depends on inputs x weights w and bias b

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

Derivatives with respect to w and b :

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

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



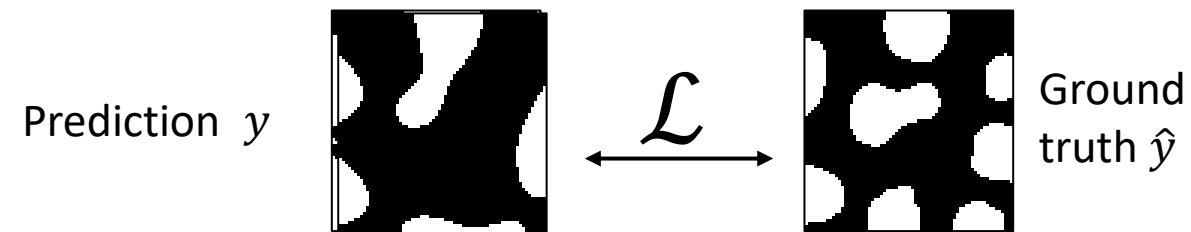
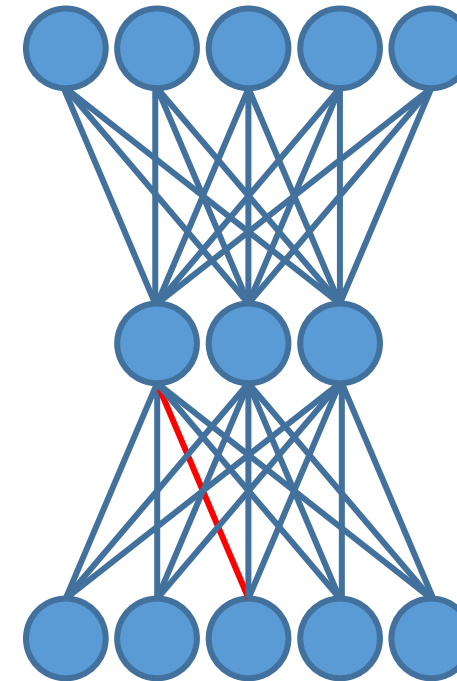
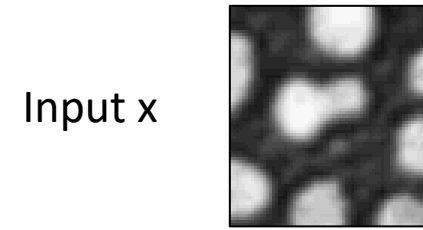
- The principle of the BackProp algorithm is to calculate the gradient of the loss function with respect to each trainable parameters of the network, i.e.

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

- 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 η is the step length, in this context known as the **learning rate**.
 η 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)

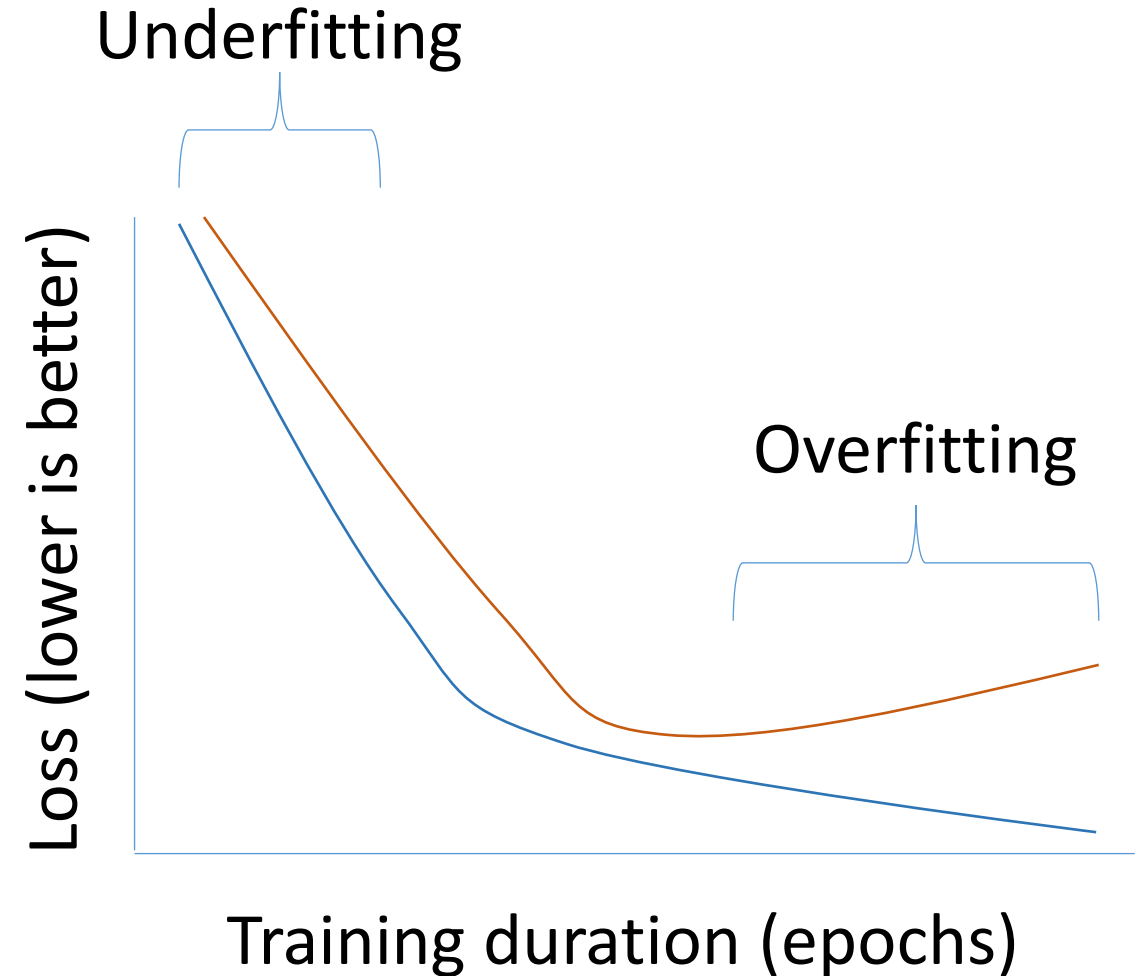
- Final evaluation after training is finished (once)

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”

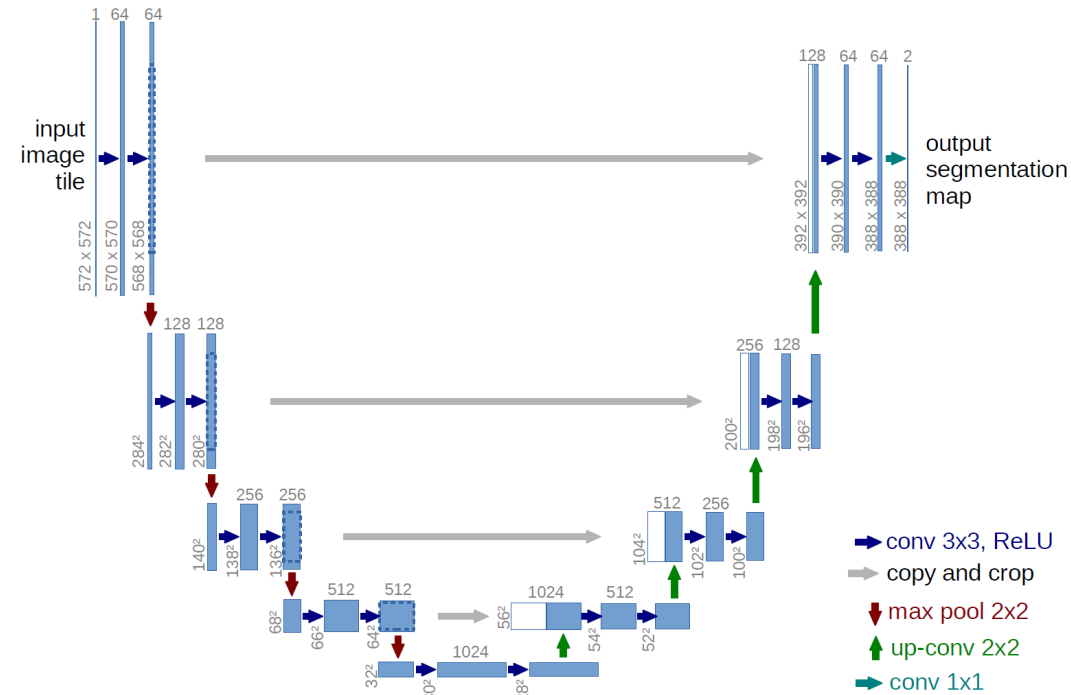


Image denoising: CARE

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

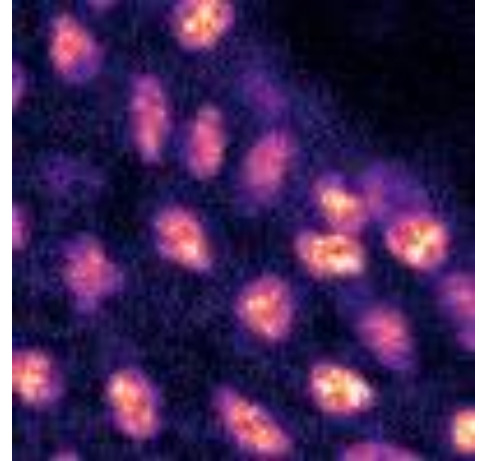
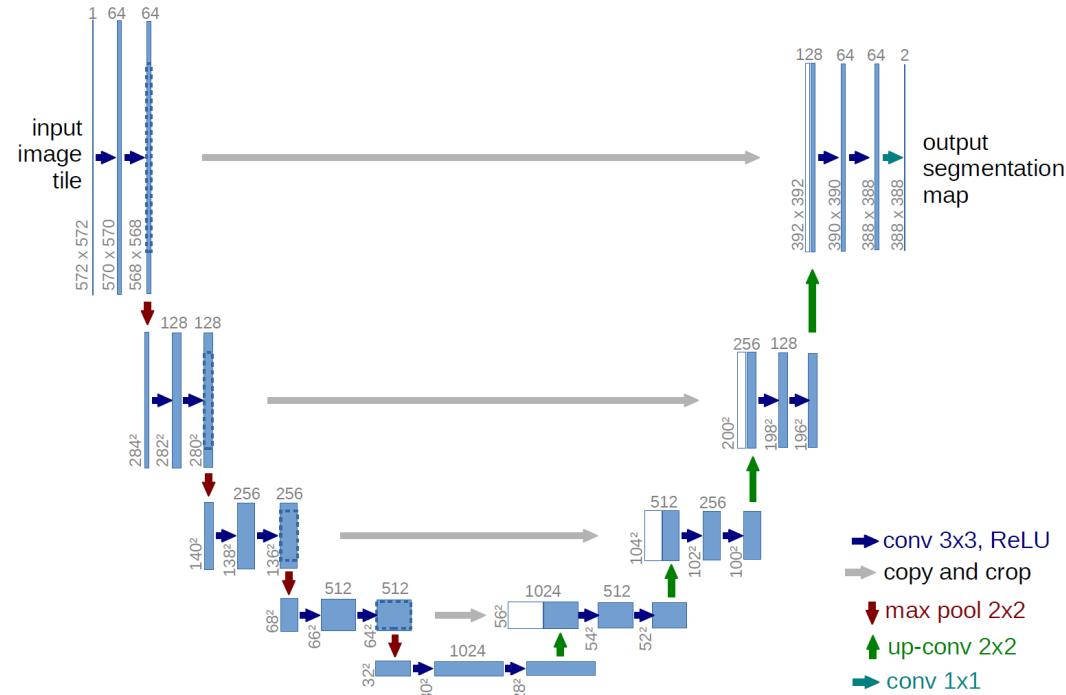
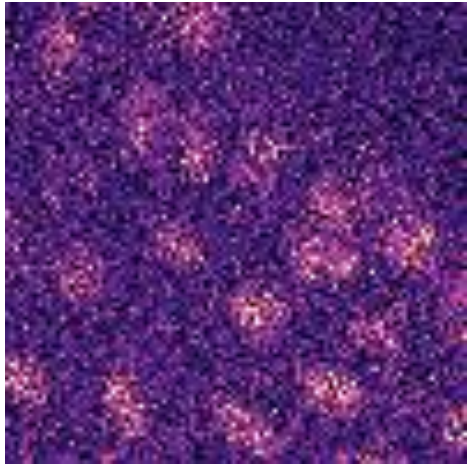


Image denoising: CARE

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

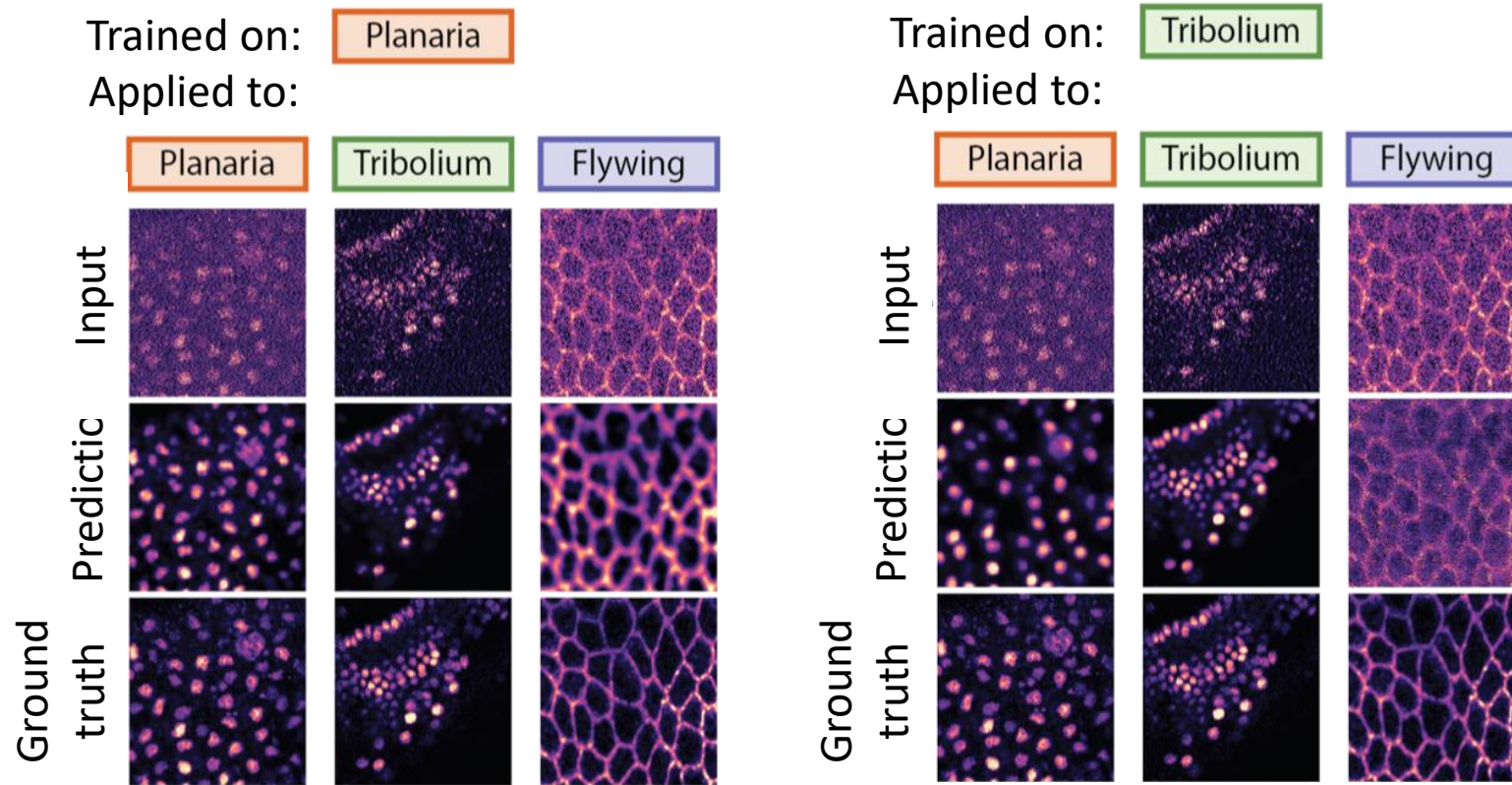
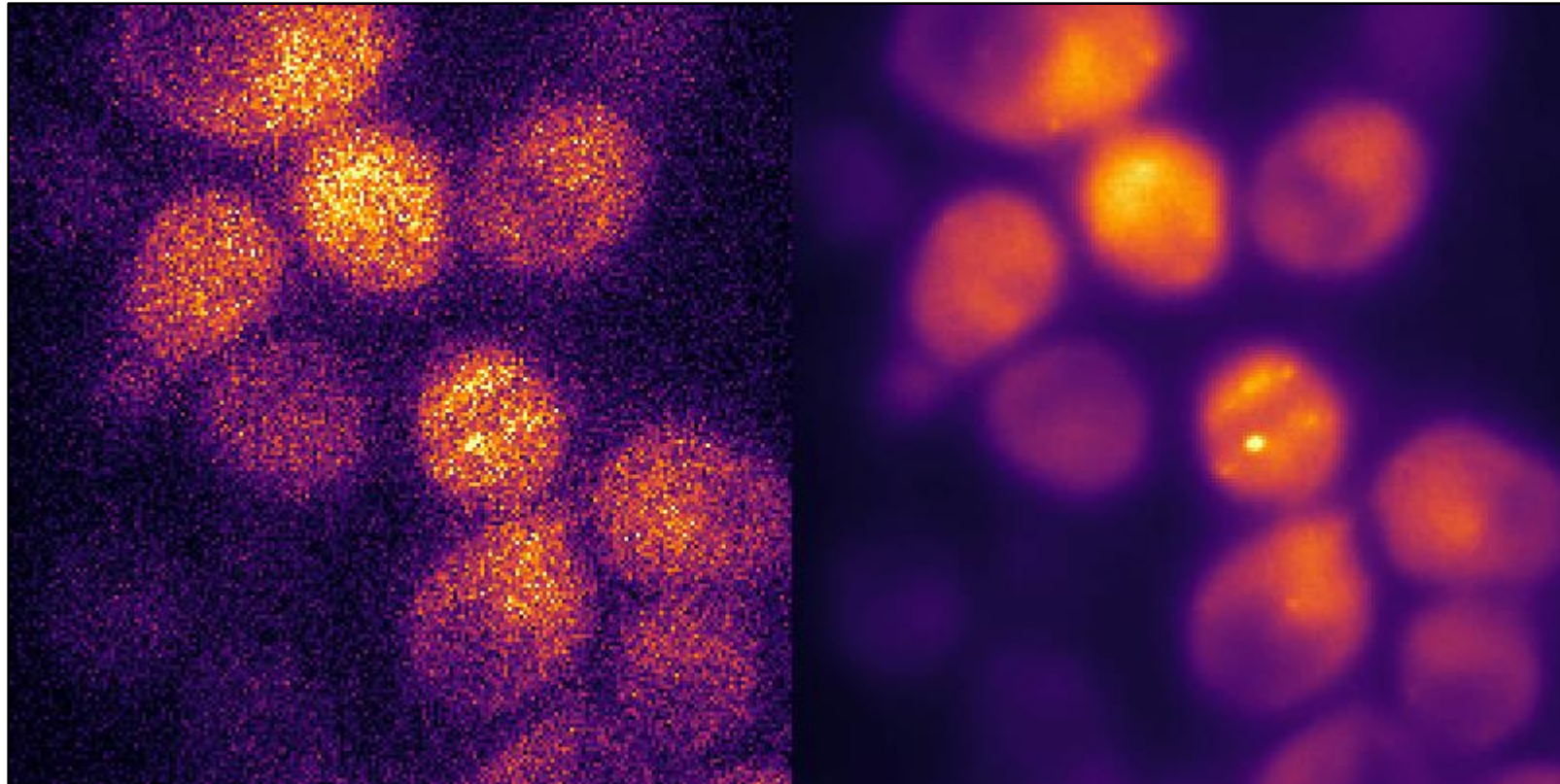


Image denoising

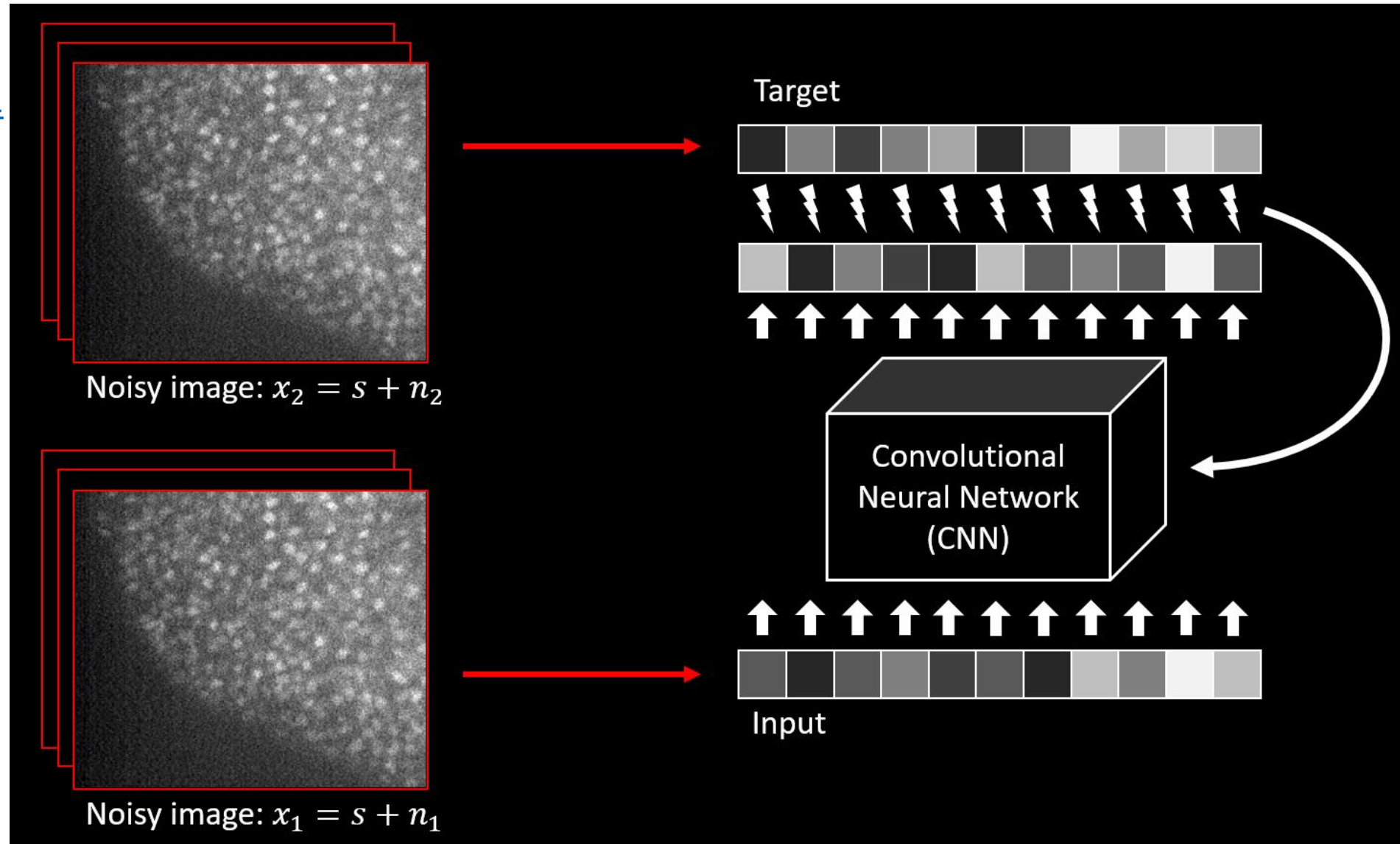
- Noise2Void



Data by Stephanie Heinrich

Slide adapted from: Alexander Krull, MPI CBG

- 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>

Noise2Void - Blind Spot Network

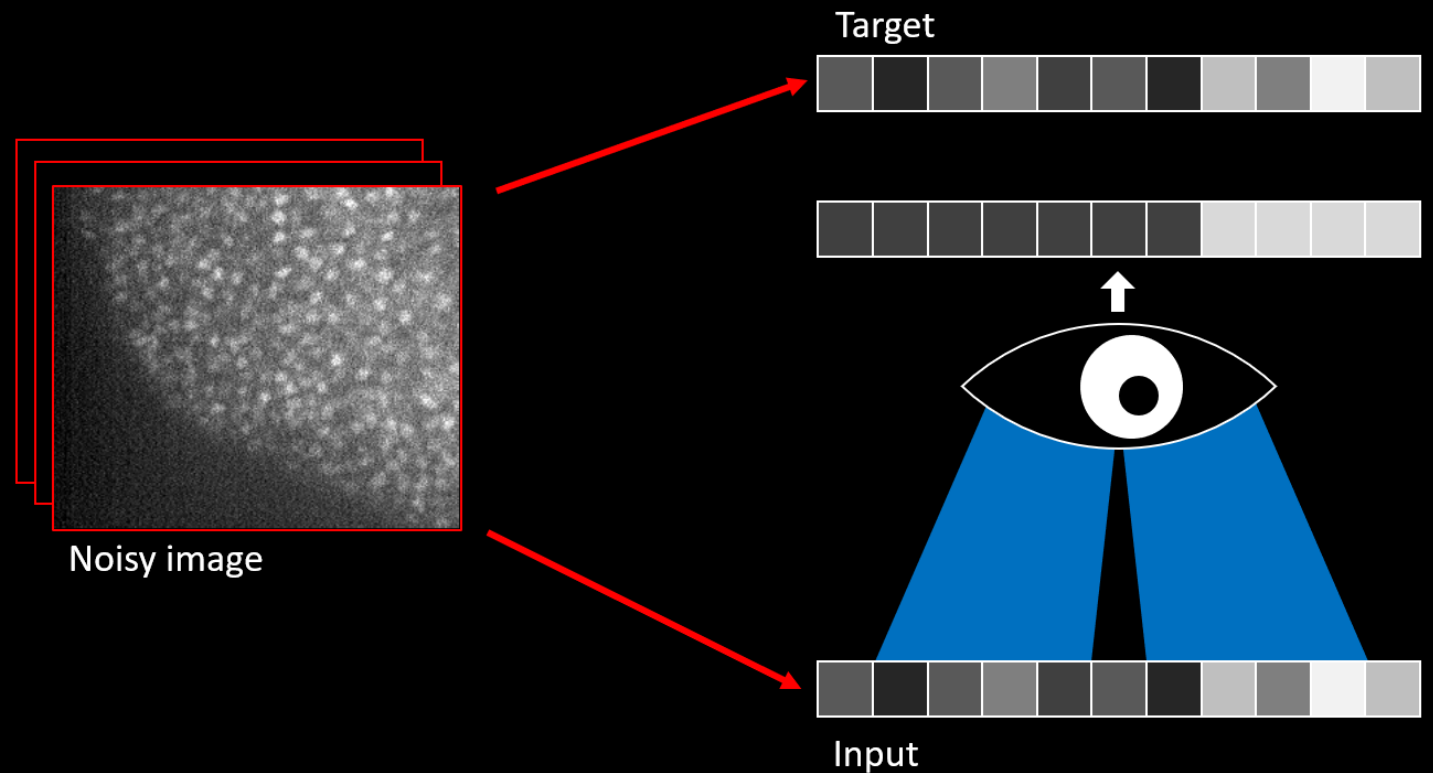
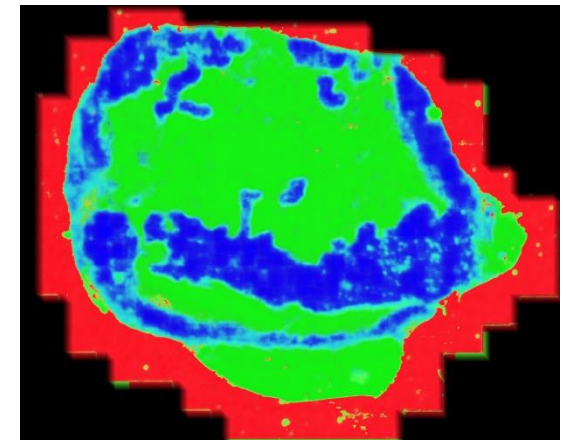
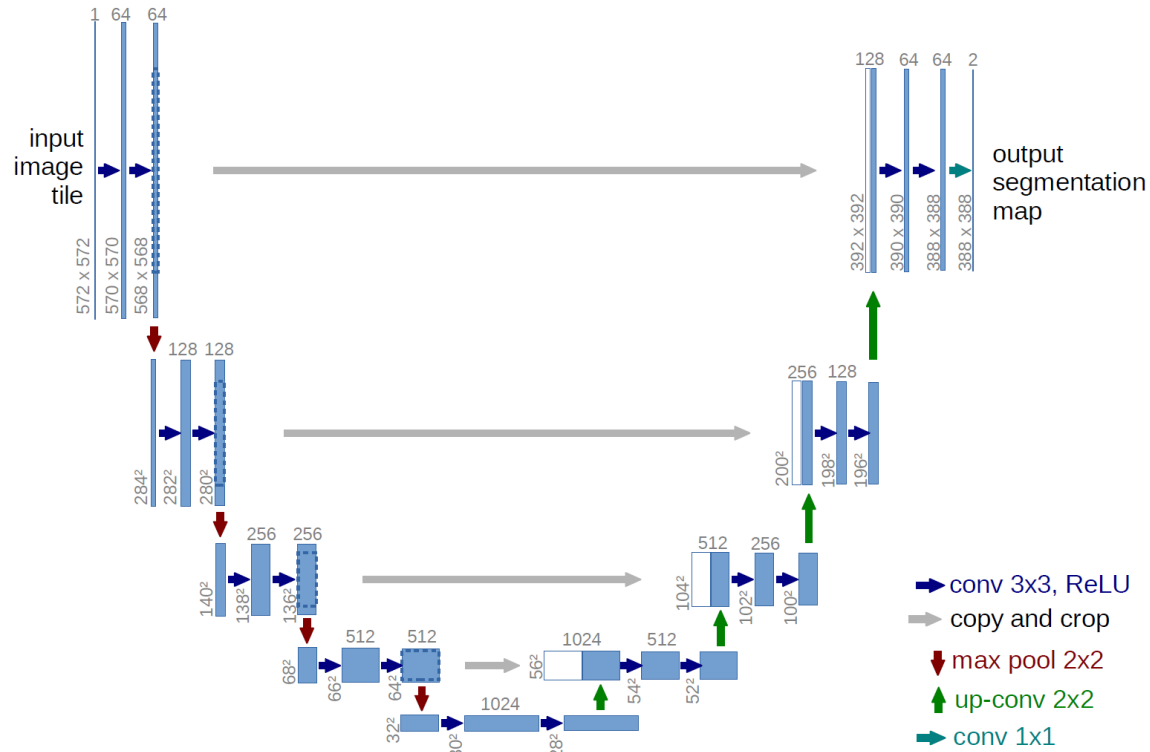
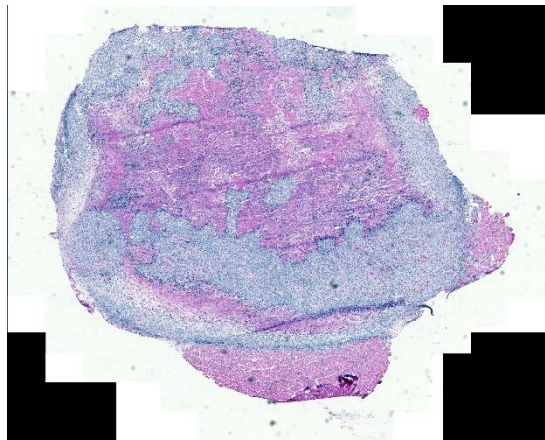
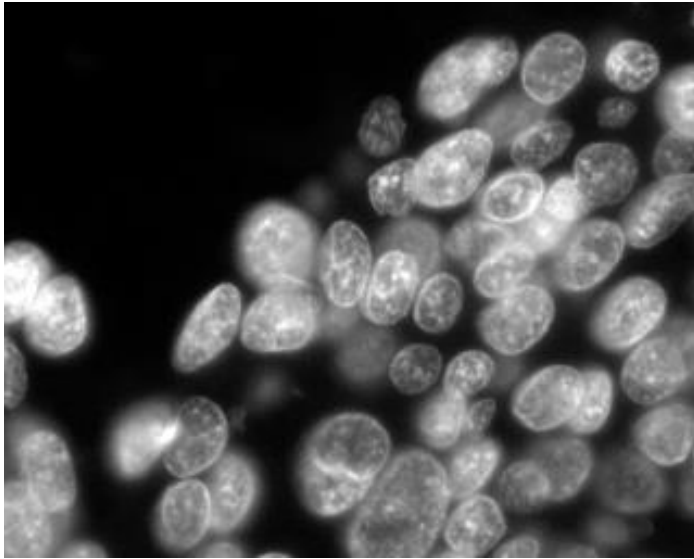


Image segmentation

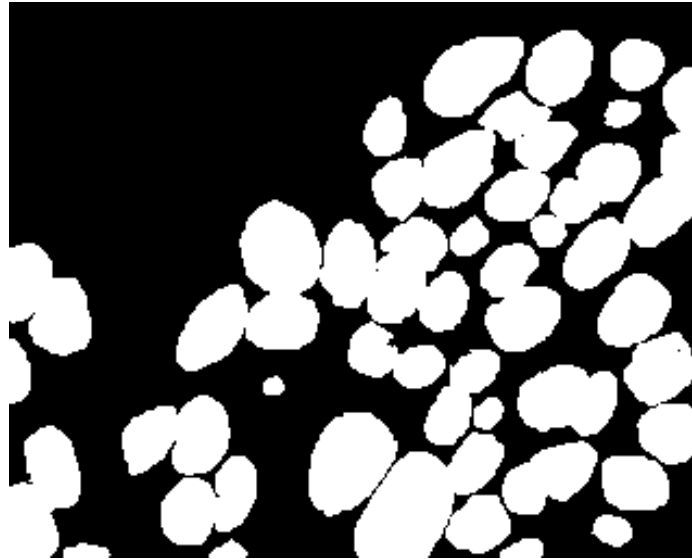
- U-net for pixel classification
 - > Semantic segmentation
 - > Probability maps



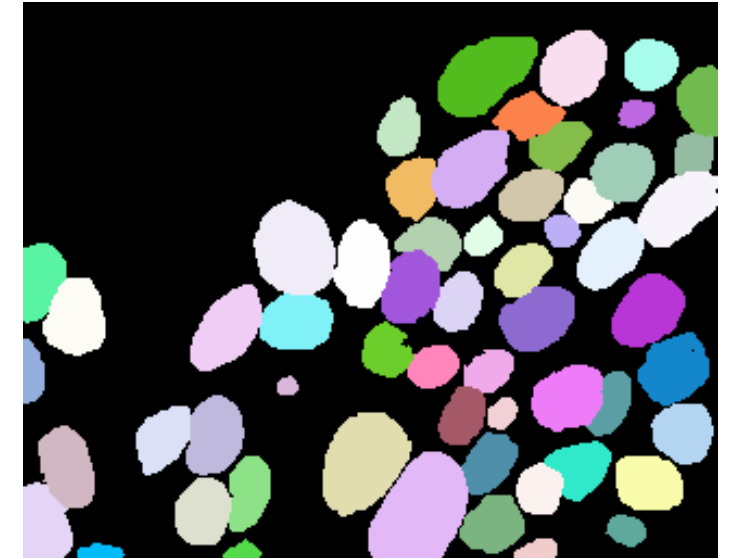
- Recap: Terminology



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

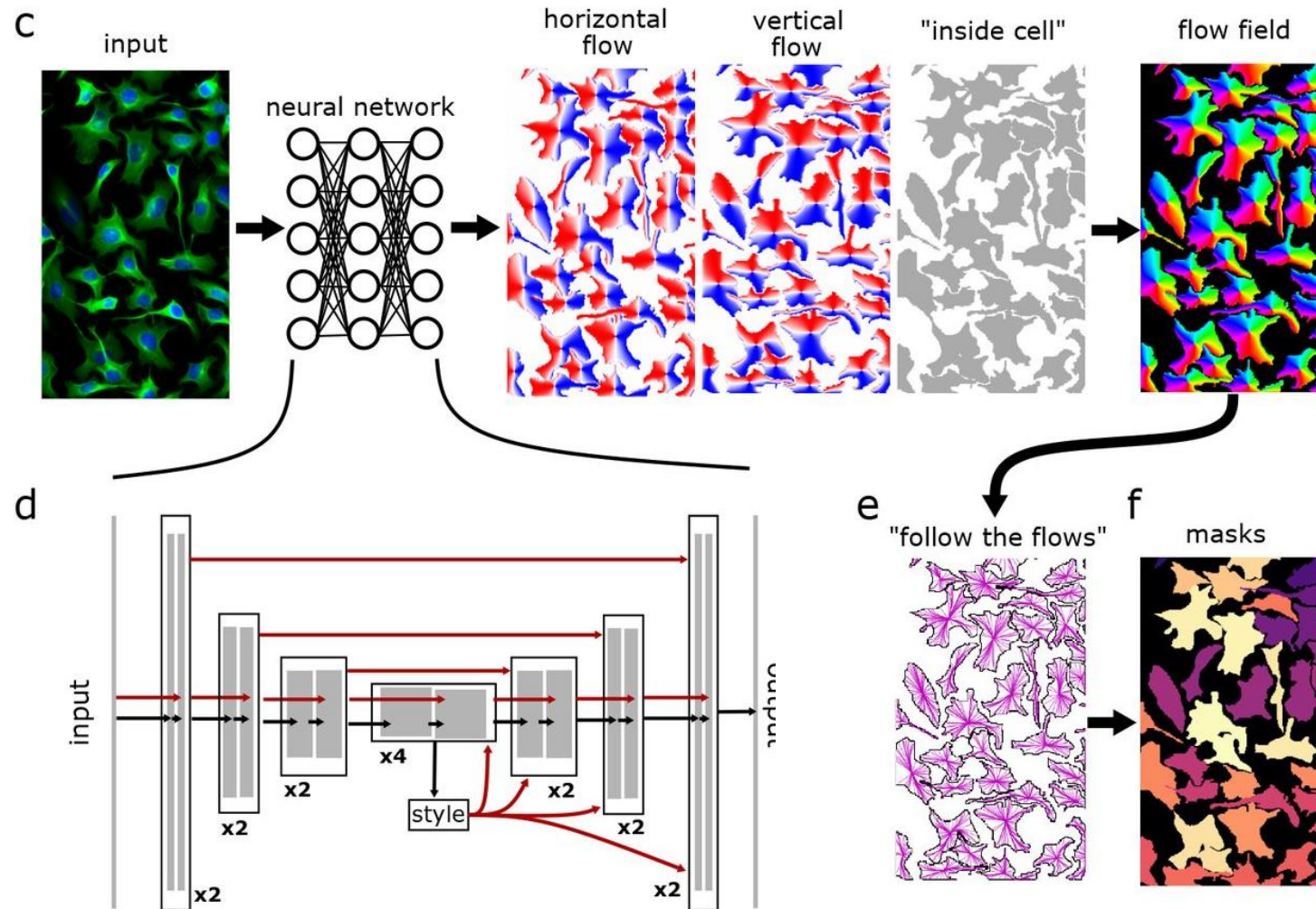
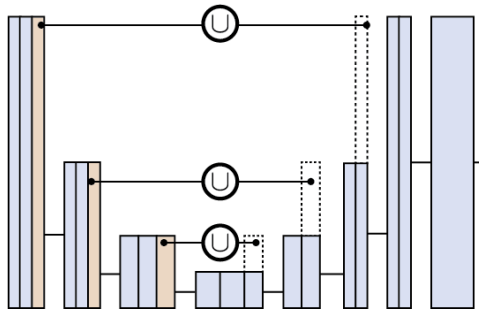
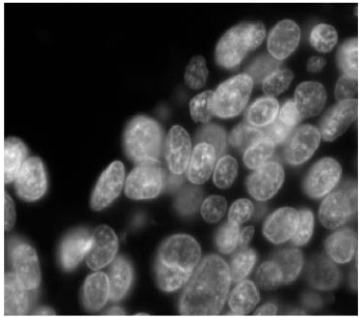


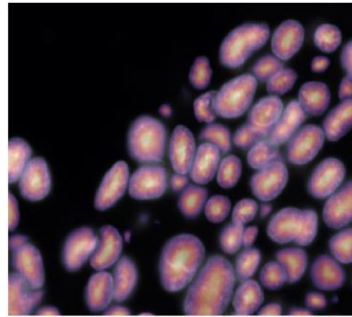
Image source: Modified from Stringer et al (2020) licensed [CC-BY-NC 4.0](https://creativecommons.org/licenses/by-nc/4.0/)
<https://www.biorxiv.org/content/10.1101/2020.02.02.931238v2.full>

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

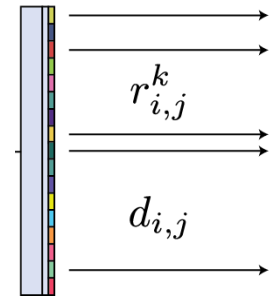
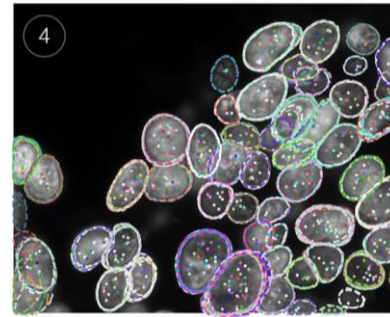
Input



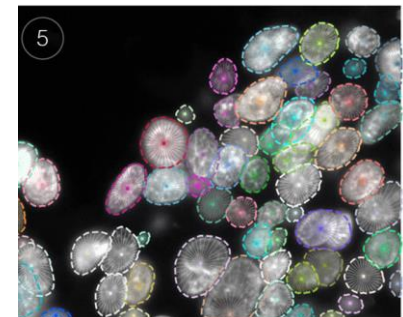
Object probabilities



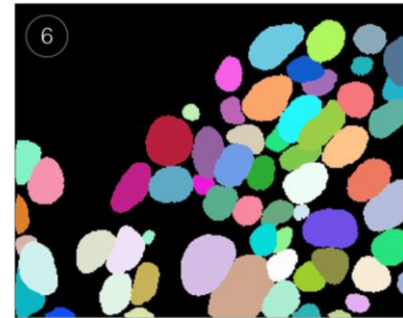
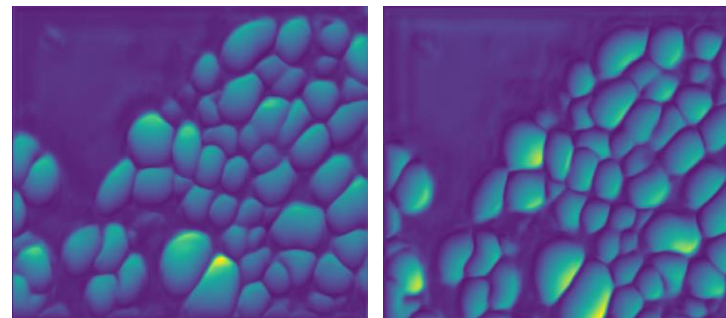
Polygon candidates



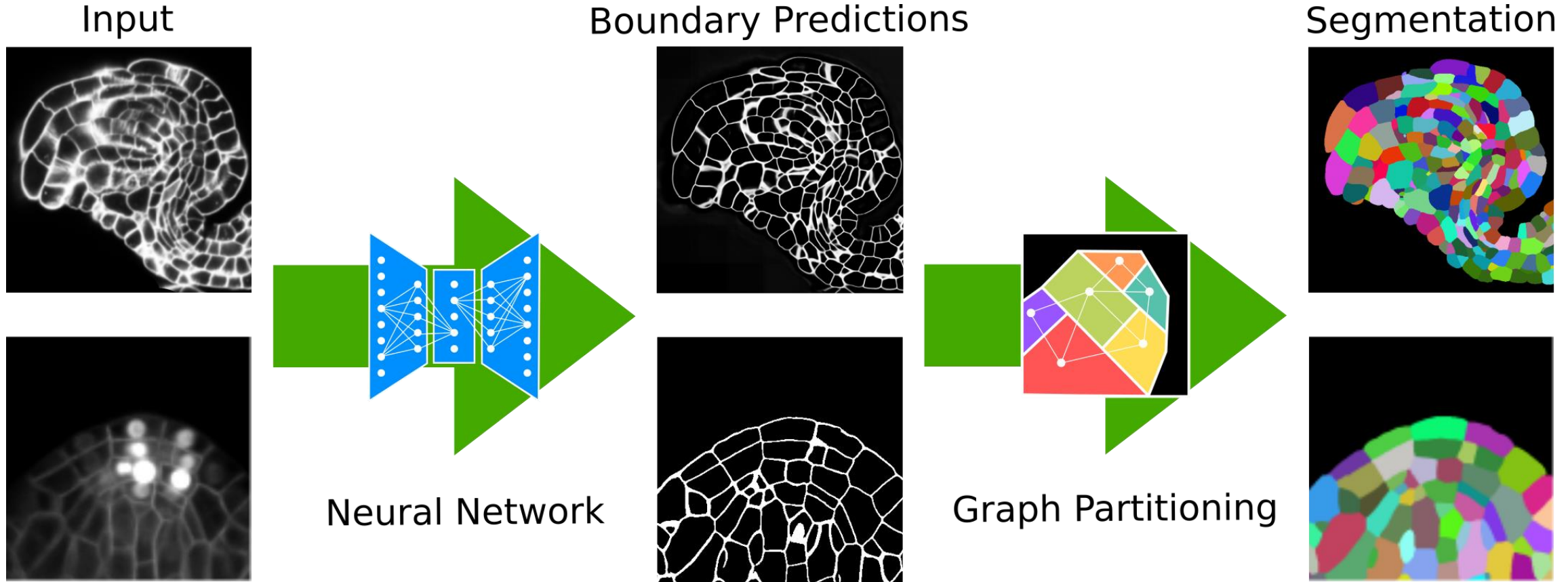
Final segmentation



Directional distance maps



- Combination of neural networks + 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!



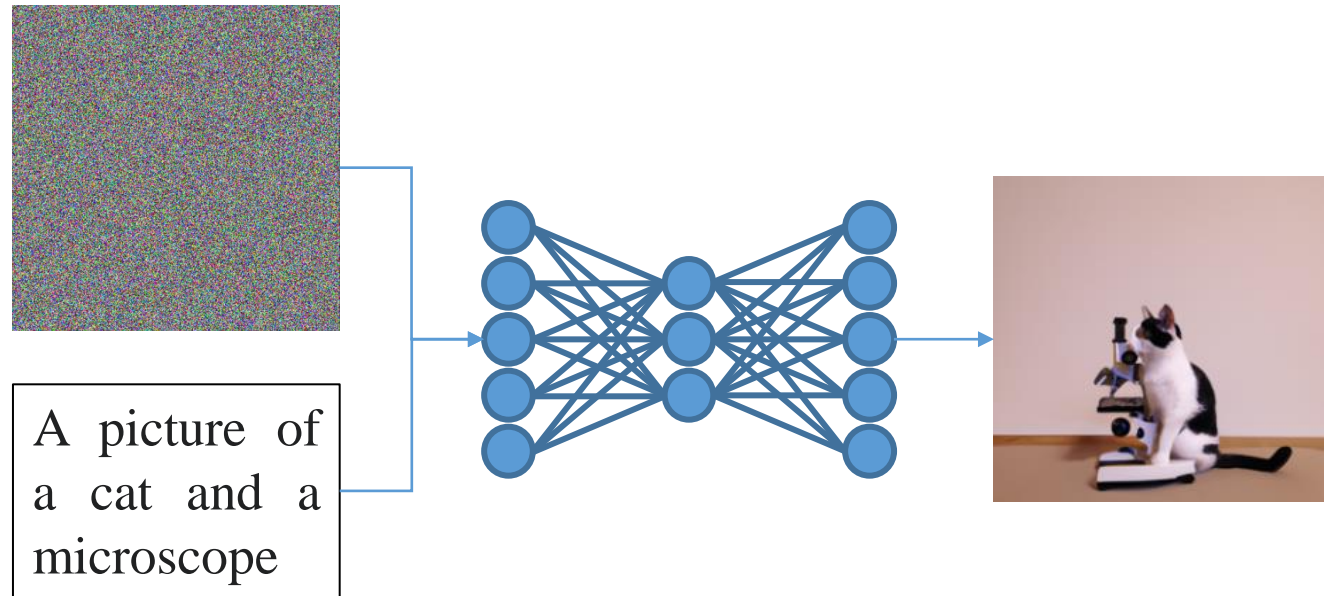
Generative Artificial Intelligence: Foundations, Applications and Implications

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)

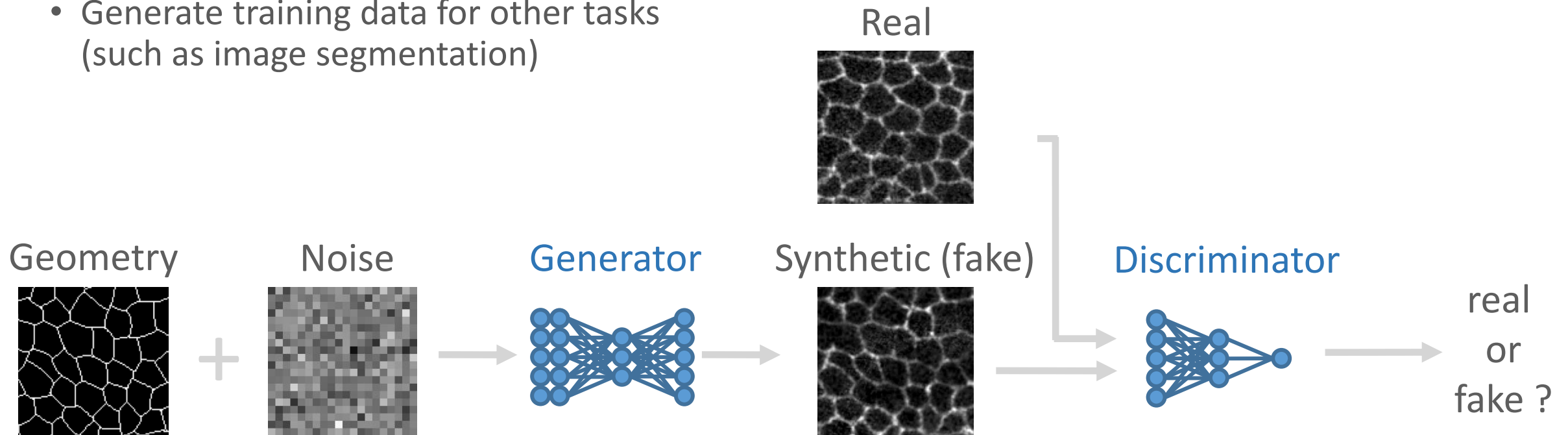
Generative Artificial Intelligence

- Definition: “Generative artificial intelligence [...] is a type of artificial intelligence (AI) system capable of generating text, images, or other media in response to prompts.”¹
- 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



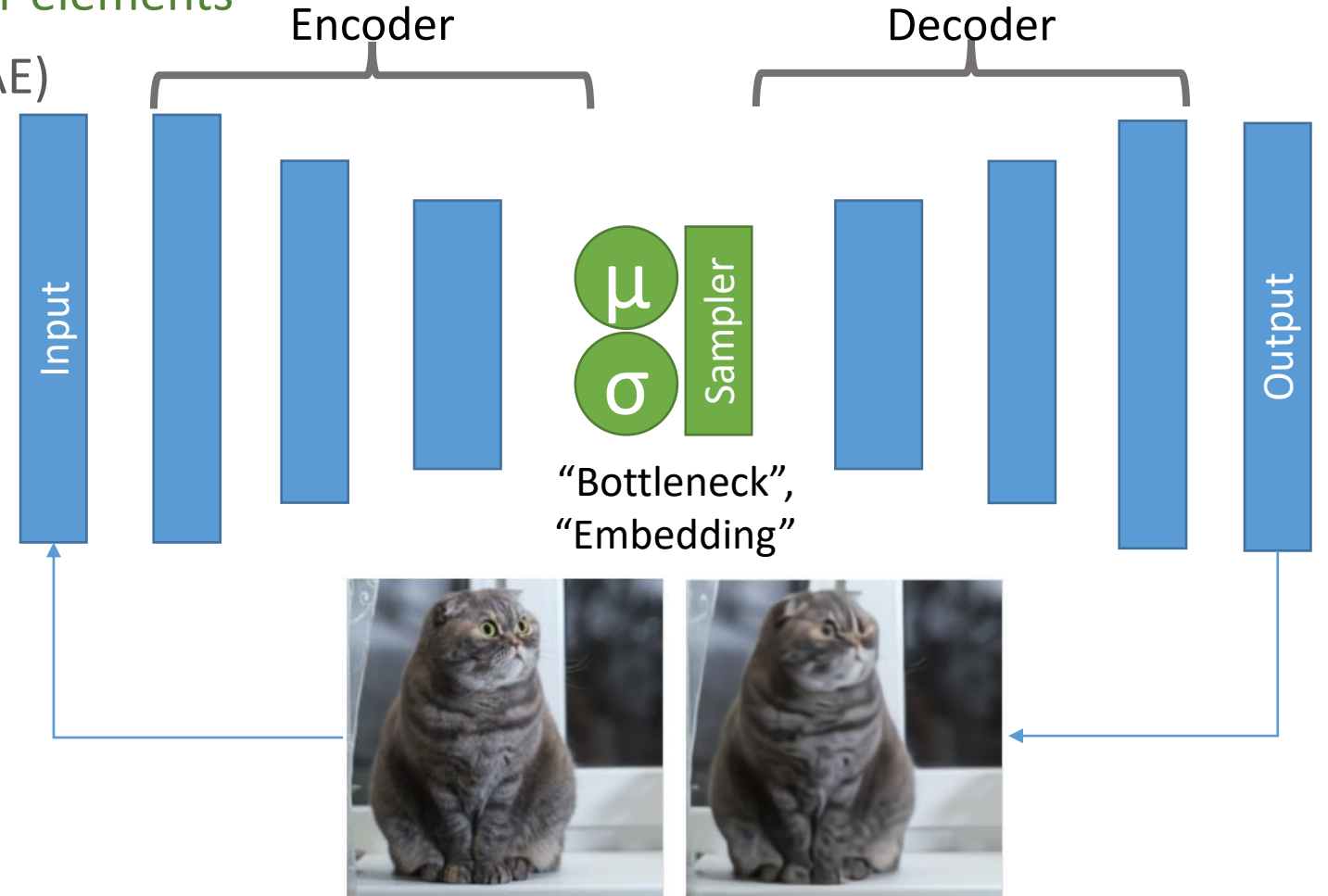
How does it work?

- 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?

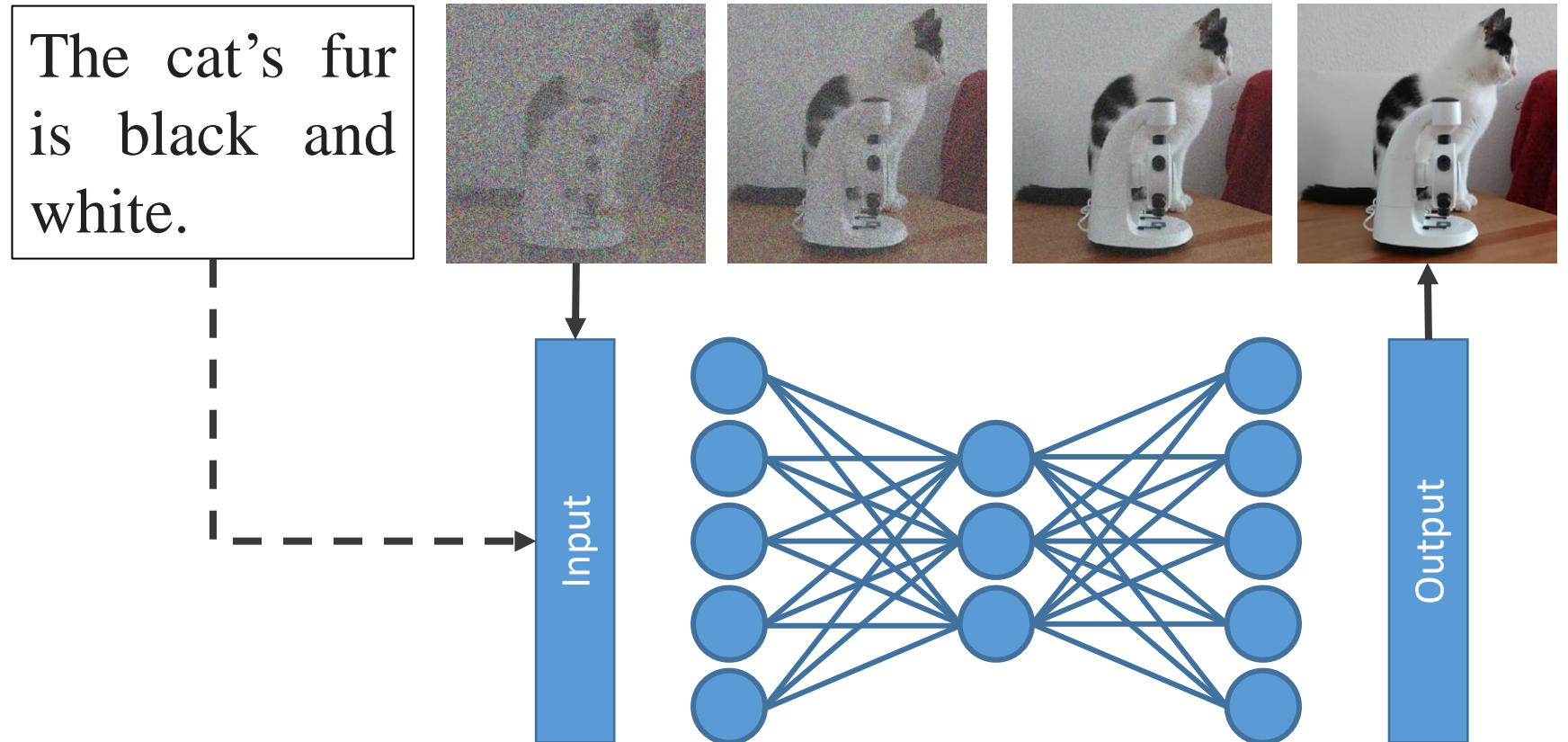
- Combination of **neural networks** + **other elements**
- Example: Variational Auto-Encoders (VAE)
- Use-cases:
 - Combine images (cat, microscope)



Cat picture source: Ramesh et al., licensed [CC-BY 4.0](https://creativecommons.org/licenses/by/4.0/)
<https://arxiv.org/pdf/2102.12092.pdf>

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

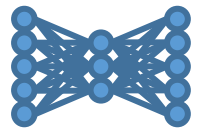


How does it work?

- Combination of neural networks + other elements + various data sources

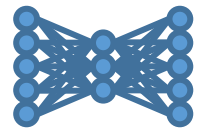
- Examples: GPT / DALL-E, Stable Diffusion

- Use-case:
 - Generate image from noise + text

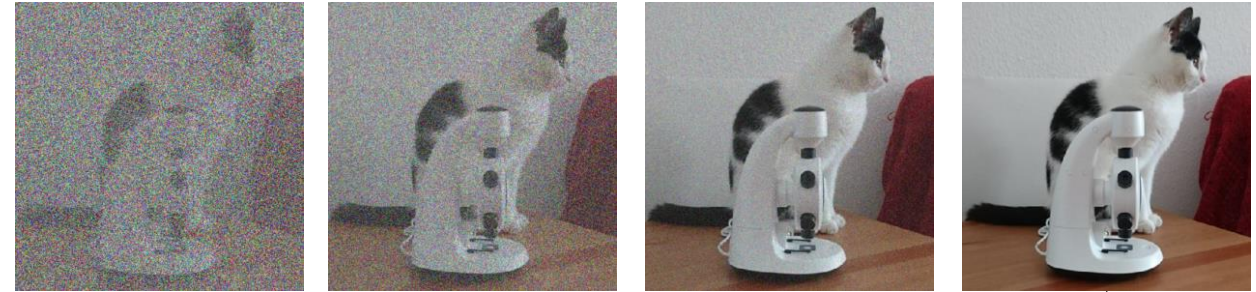
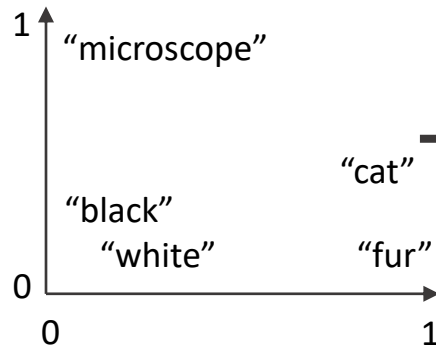


Self-attention

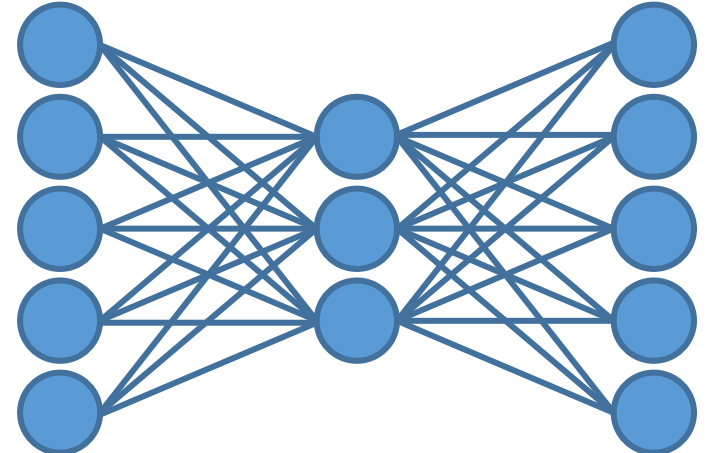
The cat's fur is black and white.



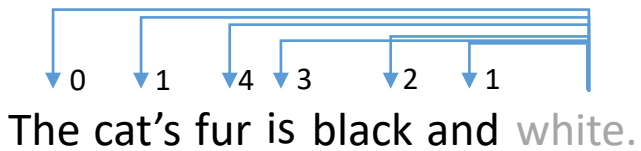
Word Embedding



Input



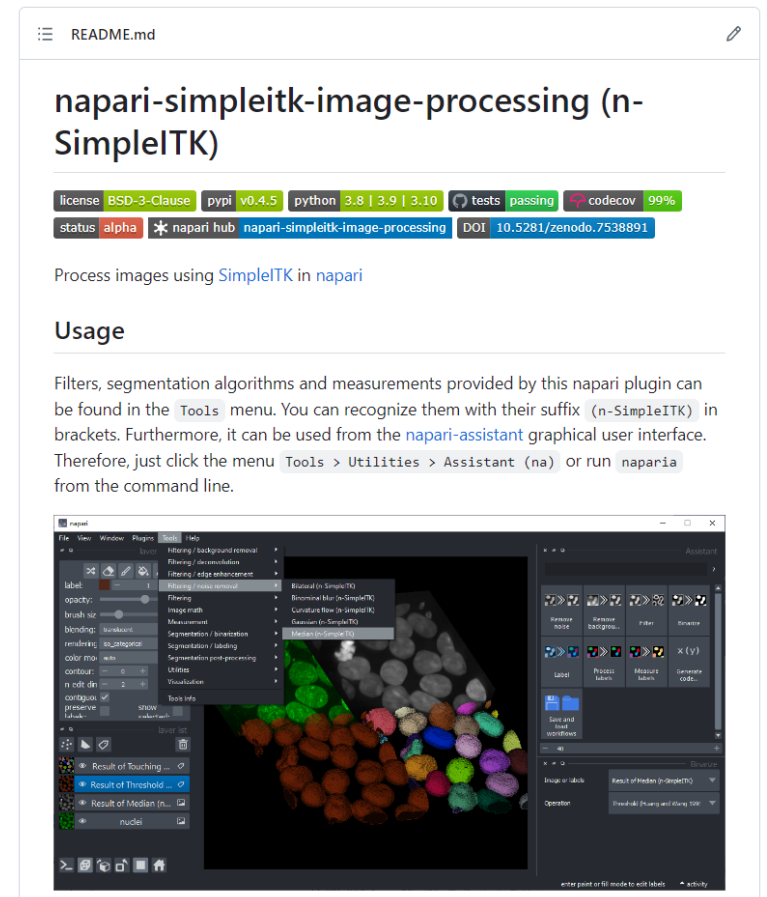
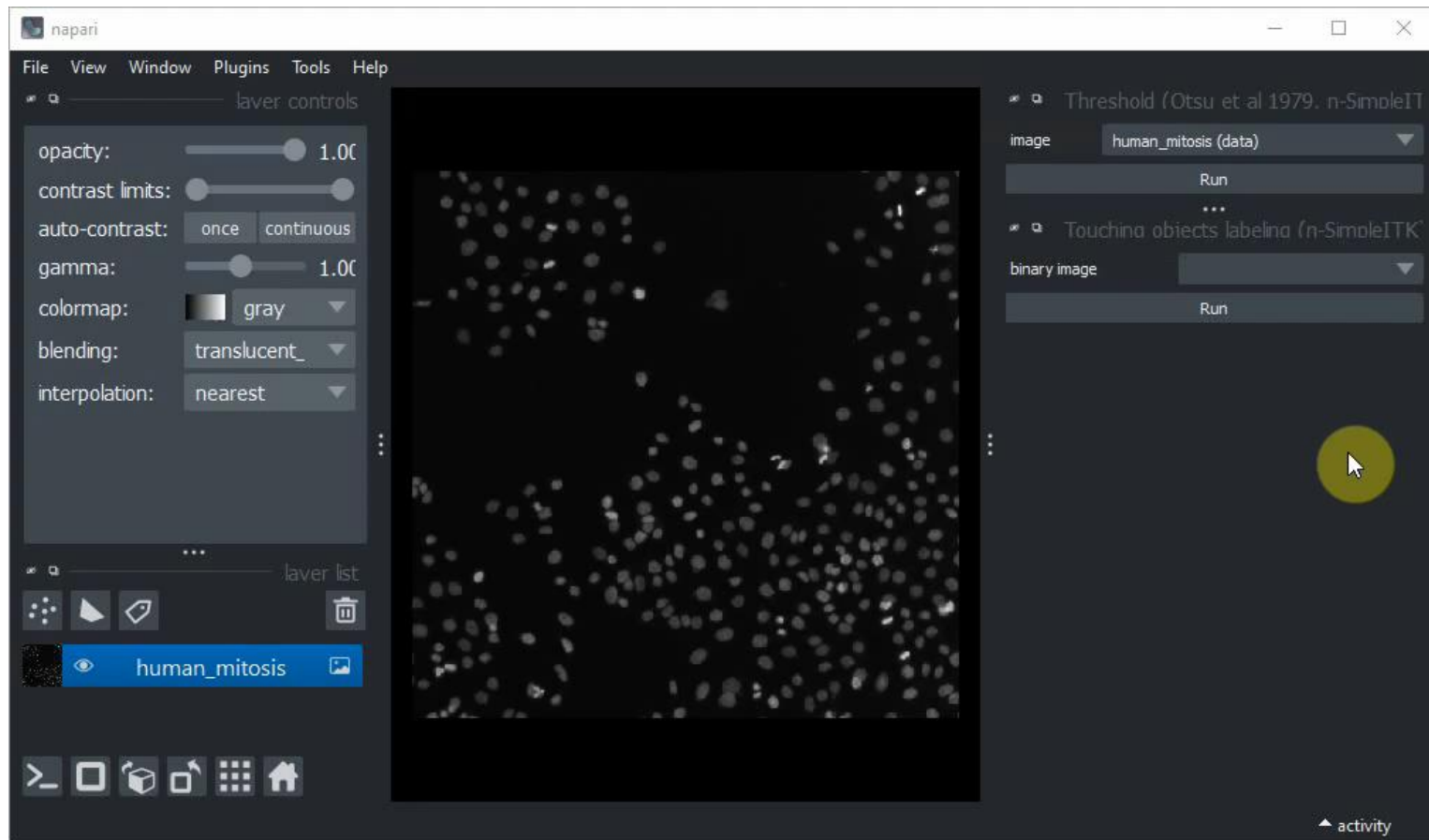
Output



- 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.

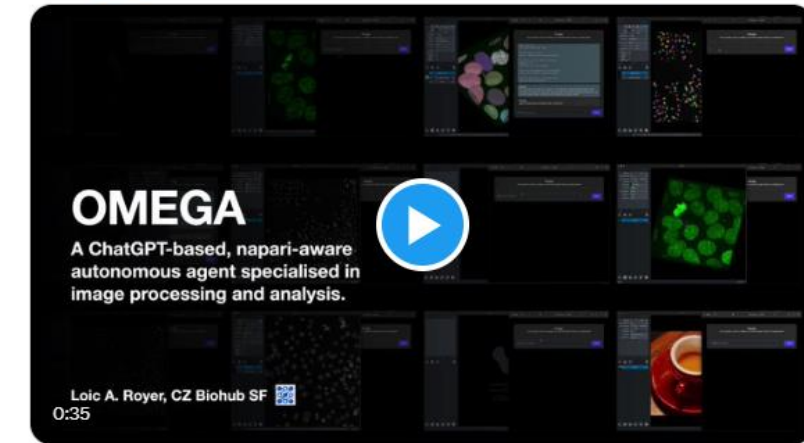
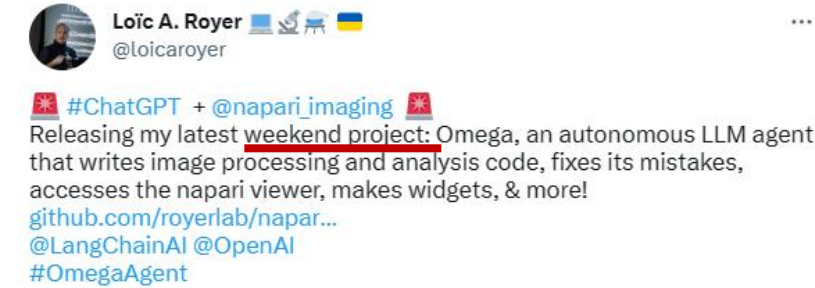
Deconstruction of napari-chatGPT

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



Deconstruction of napari-chatGPT

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

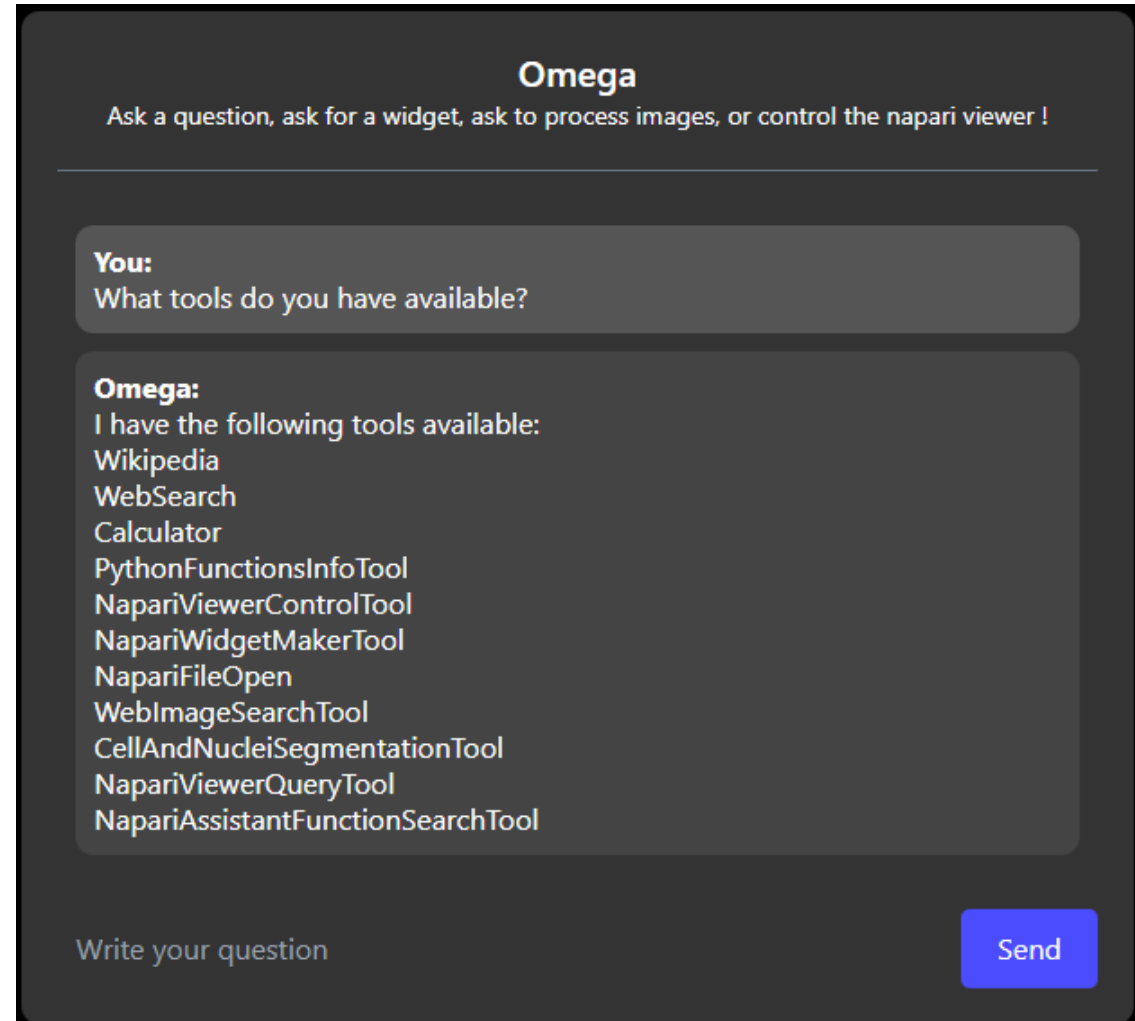
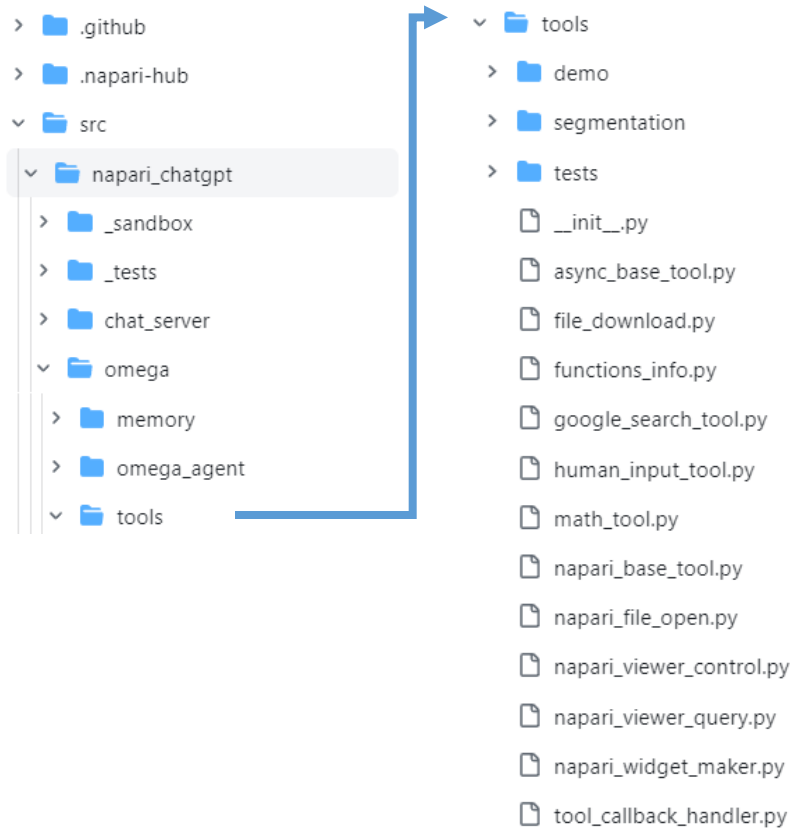


5:19 AM · May 3, 2023 · 183.6K Views

208 Retweets 44 Quotes 770 Likes 424 Bookmarks

Deconstruction of napari-chatGPT

- Napari-chatGPT defines a list of “tools”



Deconstruction of napari-chatGPT

- Napari-chatGPT defines a list of “tools”



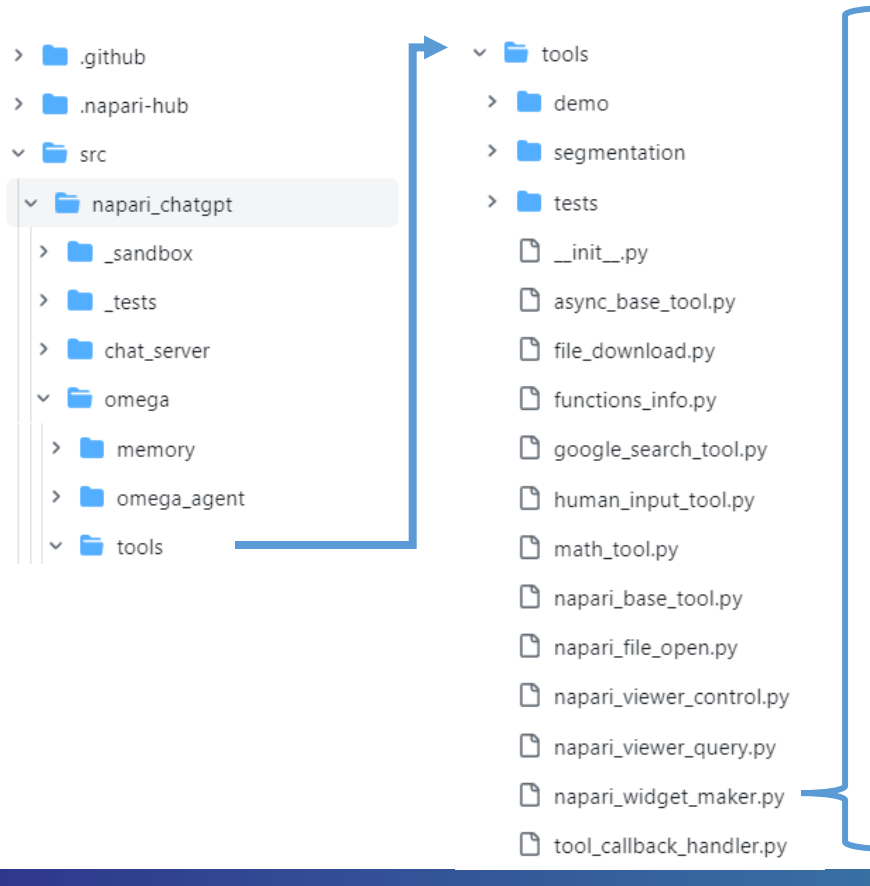
```
1 from napari_chatgpt.omega.tools.async_base_tool import AsyncBaseTool
2 from napari_chatgpt.utils.google import search_overview
3
4
5 class GoogleSearchTool(AsyncBaseTool):
6     name = "GoogleSearch"
7     description = "Useful for when you need to answer questions by querying the web."
8
9     def _run(self, query: str) -> str:
10         """Use the tool."""
11         result = search_overview(query=query)
12         return result
```



```
58 def search_overview(query: str,
59                     num_results: int = 3,
60                     lang: str = "en",
61                     max_text_snippets: int = 5,
62                     max_query_freq_hz: float = 1.0) -> str:
63     url = f"https://www.google.com/search?q={query}&num={num_results}&hl={lang}"
64     text = text_from_url(url,
65                         max_text_snippets=max_text_snippets,
66                         max_query_freq_hz=max_query_freq_hz)
67     return text
```

Deconstruction of napari-chatGPT

- Napari-chatGPT defines a list of “tools”



```
10  _napari_widget_maker_prompt = """
11  Task:
12  You competently write image processing and image analysis functions in python given a plain text request.
13  The function should be pure, self-contained, effective, well-written, syntactically correct.
14  The function should work on 2D and 3D images, and images of any number of dimensions (nD),
15  unless the request is explicit about the number of dimensions.
16  The widget should do all and everything that is asked, but nothing else or superfluous.
27  Instructions for Function Signature:
28  - Integers, floats, boolean, or any other type accepted by the magicgui library.
29  - Decorate the function with the magicgui decorator: '@magicgui(call_button='Run')' where <function_name>
30  - DO NOT CREATE A NEW INSTANCE OF A NAPARI VIEWER, use the one provided in the variable: 'viewer'.
31  - DO NOT write code to add the widget to the napari window by calling viewer.window.add_dock_widget().
46  The function signature should have a type hint for the return, e.g. -> ImageData or -> Image
47
48  {generic_codegen_instructions}
49
50  Request:
51  {input}
```

← Our prompt goes here

https://github.com/royerlab/napari-chatgpt/blob/aa00b48d0523da588ece5dc8cdb006c895c4b473/src/napari_chatgpt/omega/tools/napari_widget_maker.py#L10

- 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

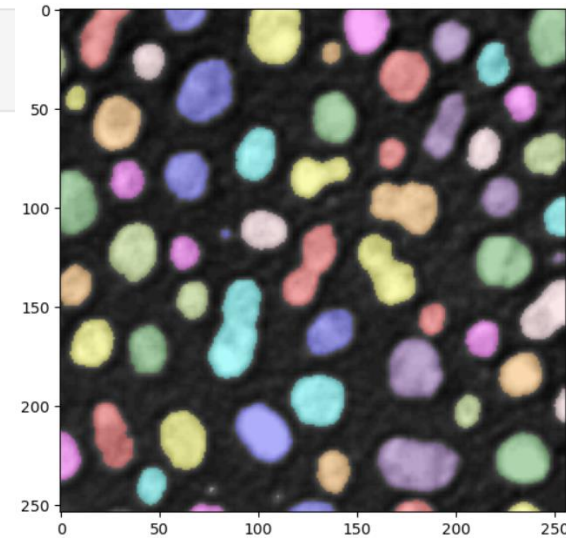
```
code = prompt(simple_question)  
print(code)  
  
```python  
from skimage.io import imread
from skimage.filters import threshold_otsu
from skimage.measure import label
from skimage.color import label2rgb
import matplotlib.pyplot as plt

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



- 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.

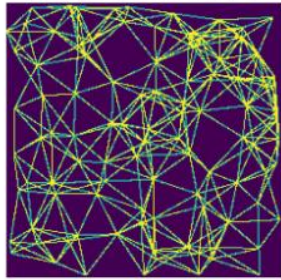
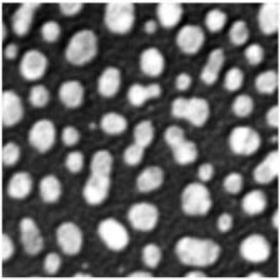
Attempt 1

```
<string>:3: UserWarning: viewer requires Qt
```

Attempt 2

Original Image

Mesh between objects



Attempt 3

```
NameError                                Traceback (most recent call last)
File ~\mambaforge\envs\my_first_env\lib\site-packages\ipywidgets\widgets\interact.py:240, in interact.interact(*args)
    238     value = widget.get_interact_value()
    239     self.kwarg[widget._kwarg] = value
--> 240 self.result = self.f(**self.kwarg)
    241 show_inline_matplotlib_plots()
    242 if self.auto_display and self.result is not None:
```

```
File <string>:21, in show_image(image)
```

```
NameError: name 'io' is not defined
```

```
NameError                                Traceback (most recent call last)
File ~\mambaforge\envs\my_first_env\lib\site-packages\ipywidgets\widgets\interact.py:240, in interact.interact(*args)
    238     value = widget.get_interact_value()
    239     self.kwarg[widget._kwarg] = value
--> 240 self.result = self.f(**self.kwarg)
    241 show_inline_matplotlib_plots()
    242 if self.auto_display and self.result is not None:
```

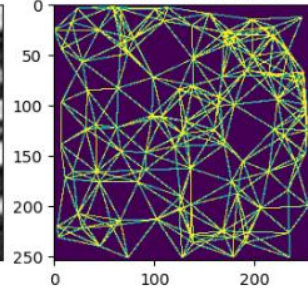
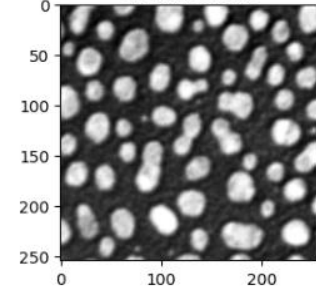
```
File <string>:25, in show_label_image(label_image)
```

```
NameError: name 'np' is not defined
```

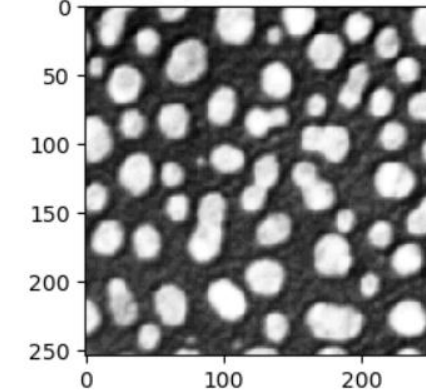
Attempt 6

Original image

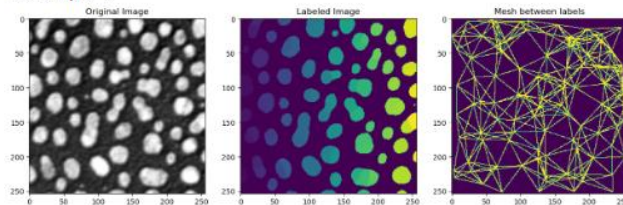
Meshed labeled objects



Attempt 8



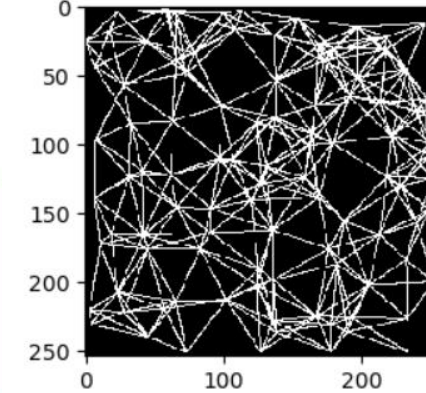
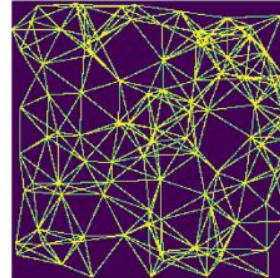
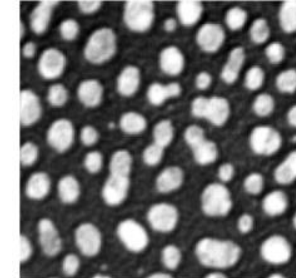
Attempt 9



Attempt 10

Original Image

Mesh between Labels



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

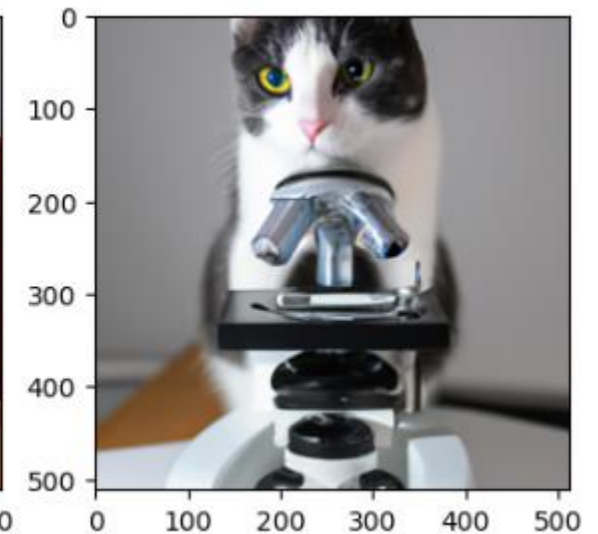
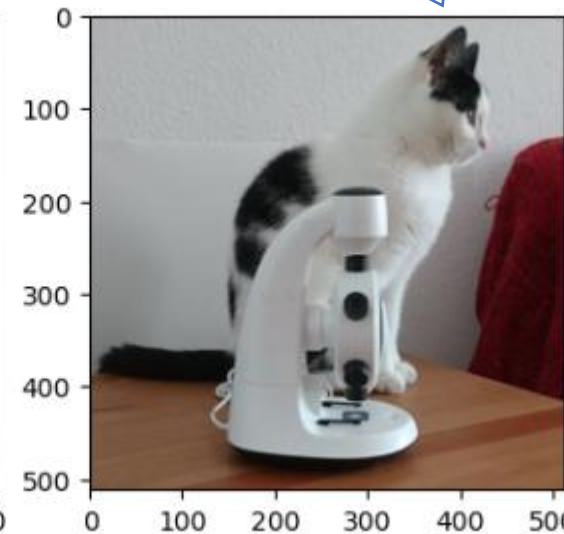
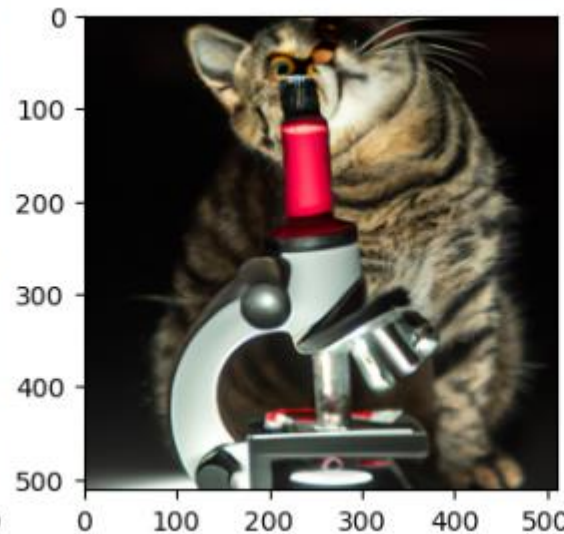
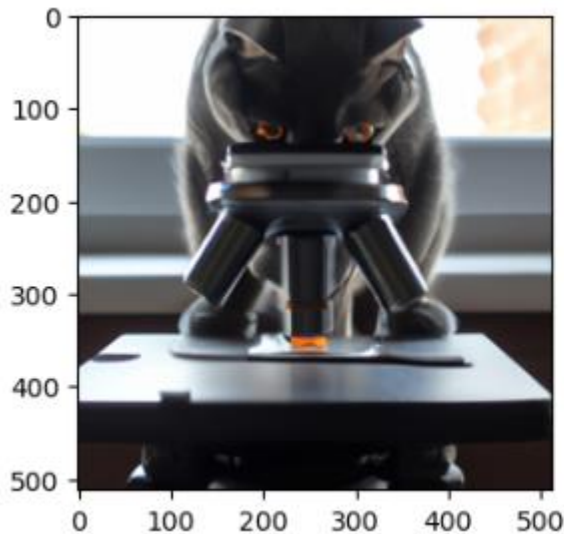
The *more sophisticated* prompt had errors in 4 out of 10 runs.

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:

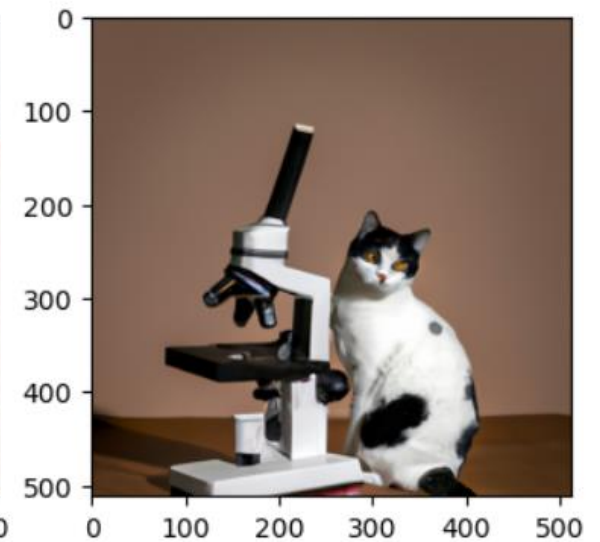
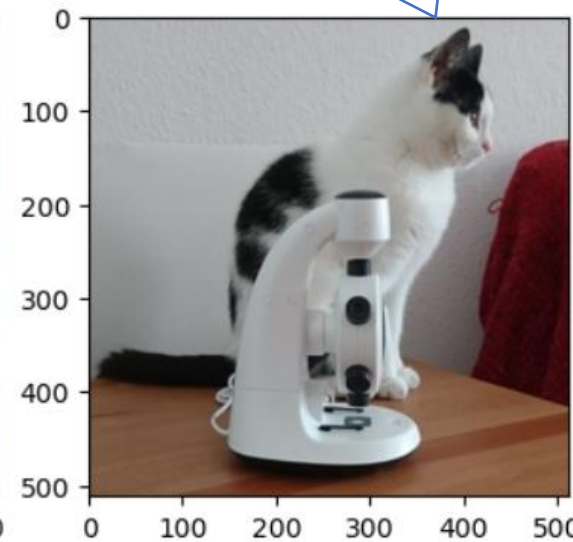
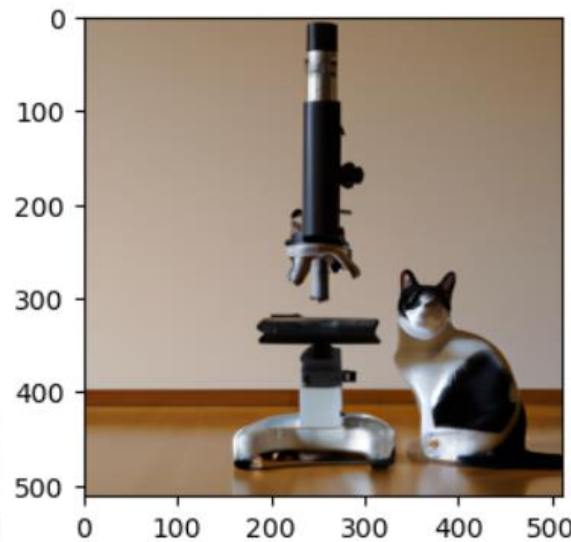
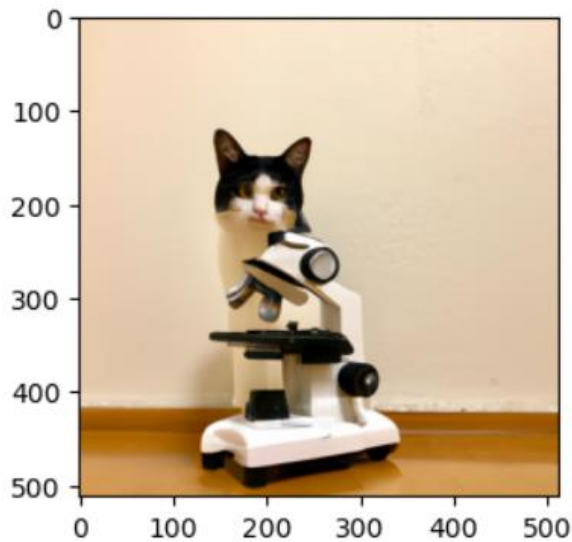
https://haesleinhuepf.github.io/BioImageAnalysisNotebooks/07_prompt_engineering/02_generating_images.html

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.  
"""
```

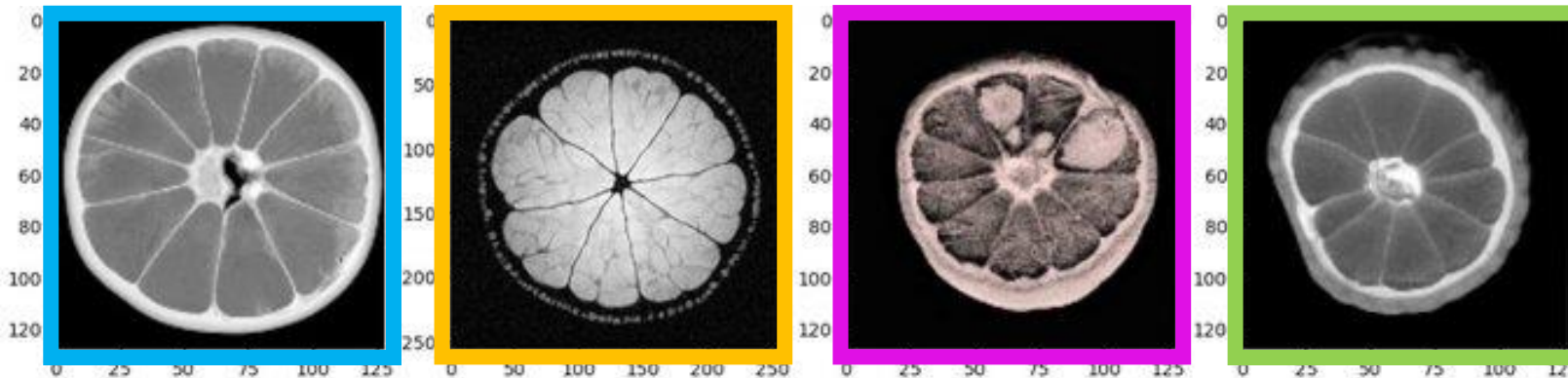
One cat
is real.



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.

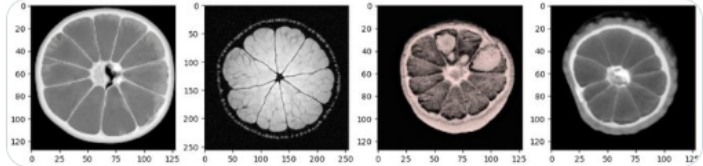
```
mri_prompt = """  
A single, high resolution, black-white image of  
a realistically looking orange fruit slice  
imaged with T2-weighted magnetic resonance imaging (MRI).  
"""
```



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! 😊



8 6 15 4,880

Robert Haase @haesleinhuepf · 20h ...

| | |
|---|-------|
| 1 | 16.8% |
| 2 | 67.6% |
| 3 | 3.4% |
| 4 | 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_DESCR  
    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?")
```

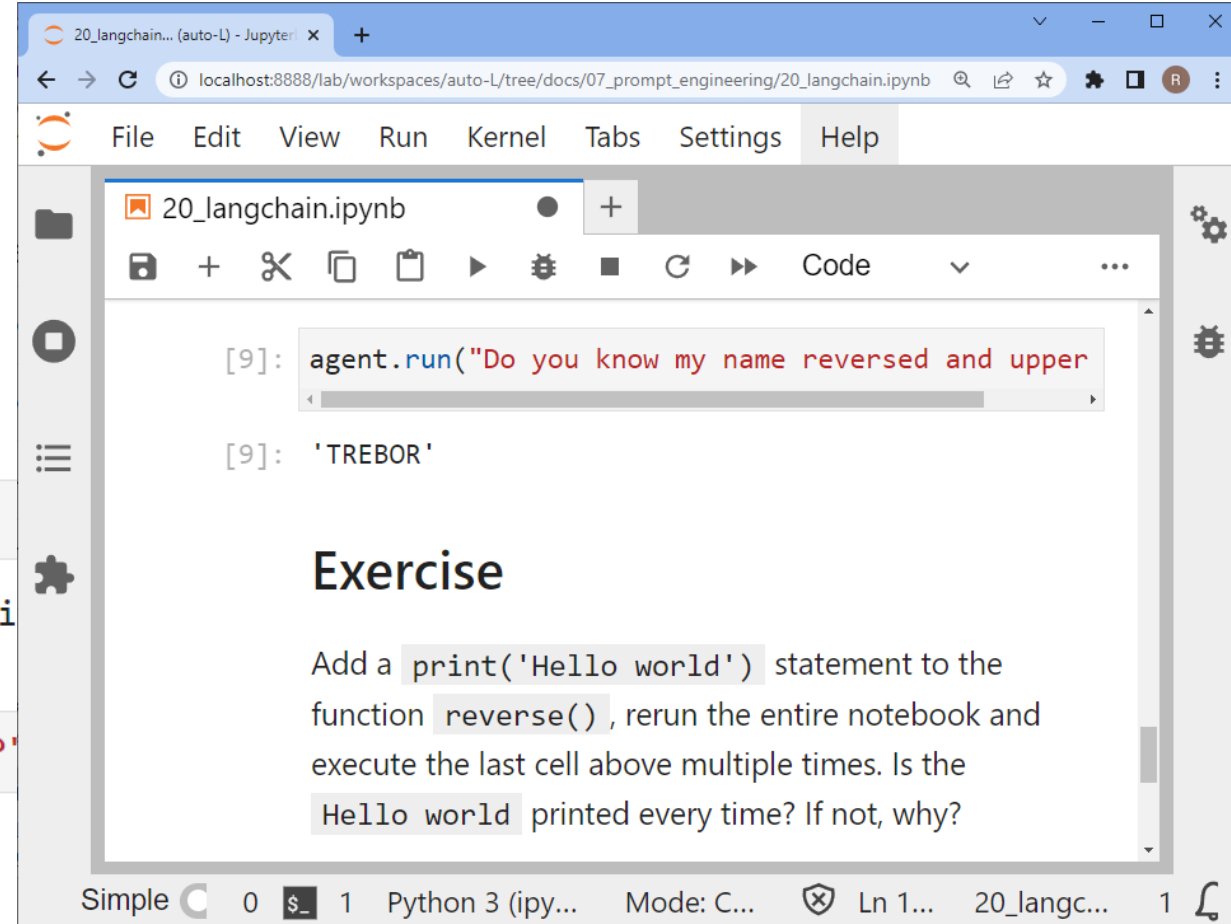
```
[7]: 'Your name is Robert'
```

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

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

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

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



The screenshot shows a JupyterLab notebook window titled "20_langchain... (auto-L) - Jupyter". The browser address bar shows "localhost:8888/lab/workspaces/auto-L/tree/docs/07_prompt_engineering/20_langchain.ipynb". The notebook interface includes a menu bar (File, Edit, View, Run, Kernel, Tabs, Settings, Help) and a toolbar with icons for file operations and execution. The notebook content shows a code cell with the following code and output:

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

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

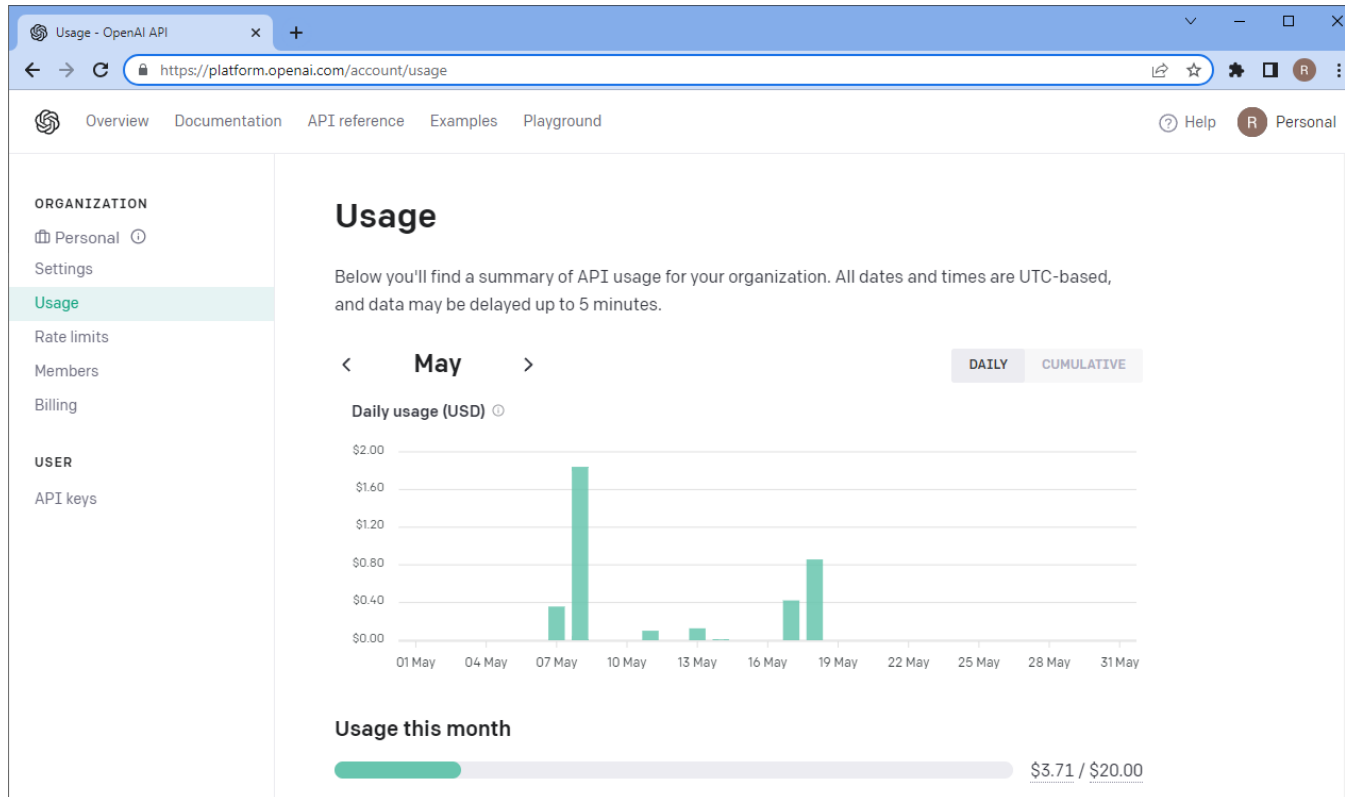
Below the code cell, there is a section titled "Exercise" with the following text:

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 bottom status bar of the notebook shows "Simple C 0 \$ 1 Python 3 (ipy... Mode: C... Ln 1... 20_langc... 1".

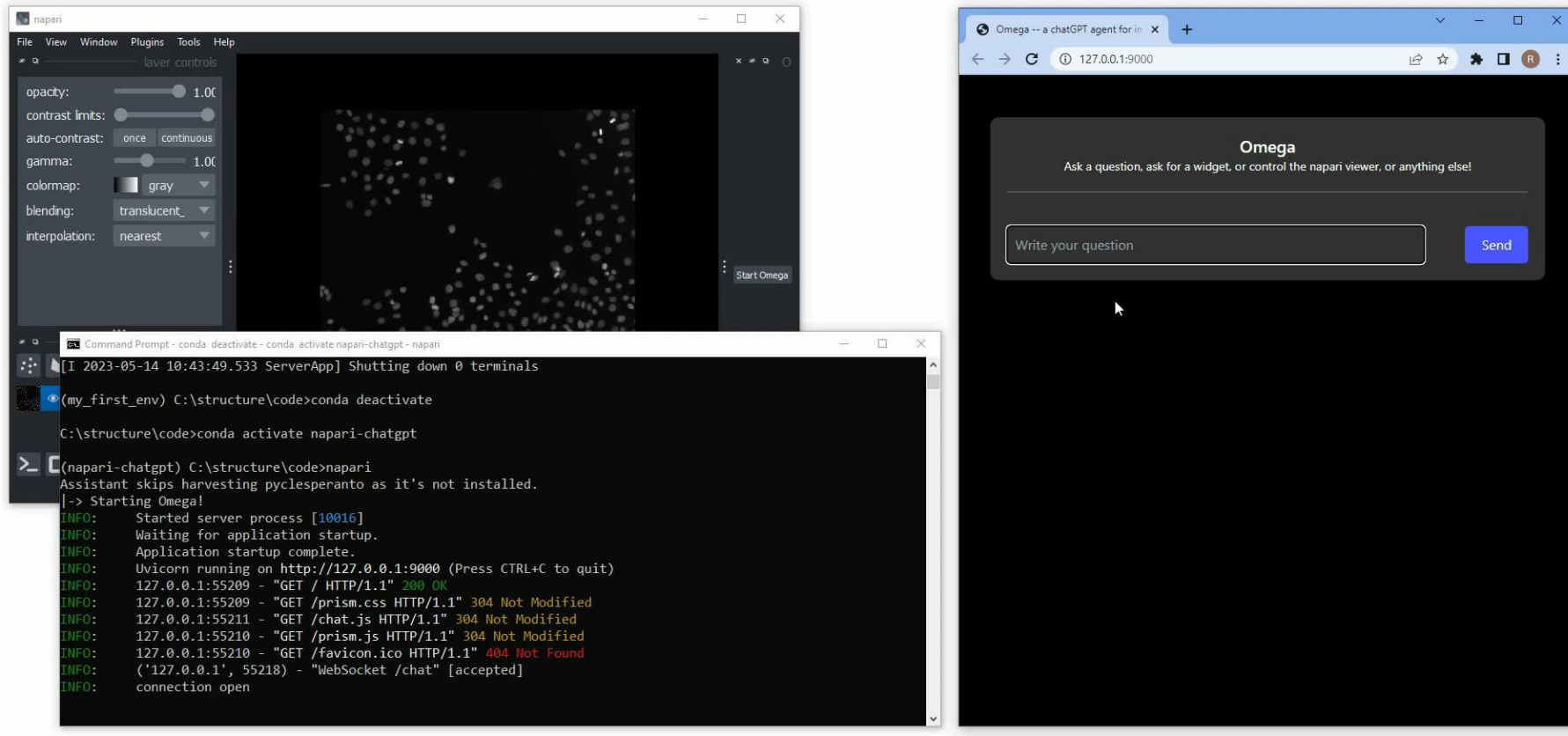
How much is the fish?

- 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



The image shows two windows. The left window is the napari viewer, displaying a dark image with many small white dots. The right window is a browser showing the Omega chat interface. Below the browser is a terminal window showing the command prompt output for starting the Omega server.

```
Command Prompt - conda deactivate - conda activate napari-chatgpt - napari
[I 2023-05-14 10:43:49.533 ServerApp] Shutting down 0 terminals
(my_first_env) C:\structure\code>conda deactivate
C:\structure\code>conda activate napari-chatgpt
(napari-chatgpt) C:\structure\code>napari
Assistant skips harvesting pyclesperanto as it's not installed.
|-> Starting Omega!
INFO: Started server process [10016]
INFO: Waiting for application startup.
INFO: Application startup complete.
INFO: Uvicorn running on http://127.0.0.1:9000 (Press CTRL+C to quit)
INFO: 127.0.0.1:55209 - "GET / HTTP/1.1" 200 OK
INFO: 127.0.0.1:55209 - "GET /prism.css HTTP/1.1" 304 Not Modified
INFO: 127.0.0.1:55211 - "GET /chat.js HTTP/1.1" 304 Not Modified
INFO: 127.0.0.1:55210 - "GET /prism.js HTTP/1.1" 304 Not Modified
INFO: 127.0.0.1:55210 - "GET /favicon.ico HTTP/1.1" 404 Not Found
INFO: ('127.0.0.1', 55218) - "WebSocket /chat" [accepted]
INFO: connection open
```

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

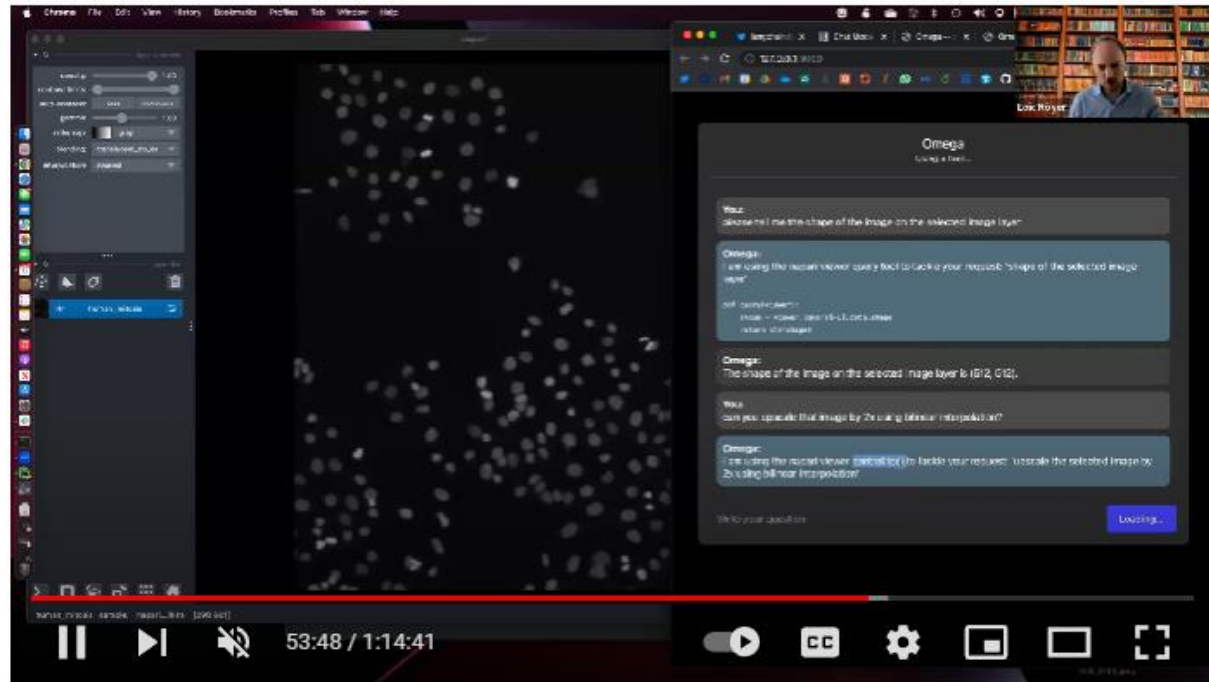
Who developed Gradoop?



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)



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>



Exercises

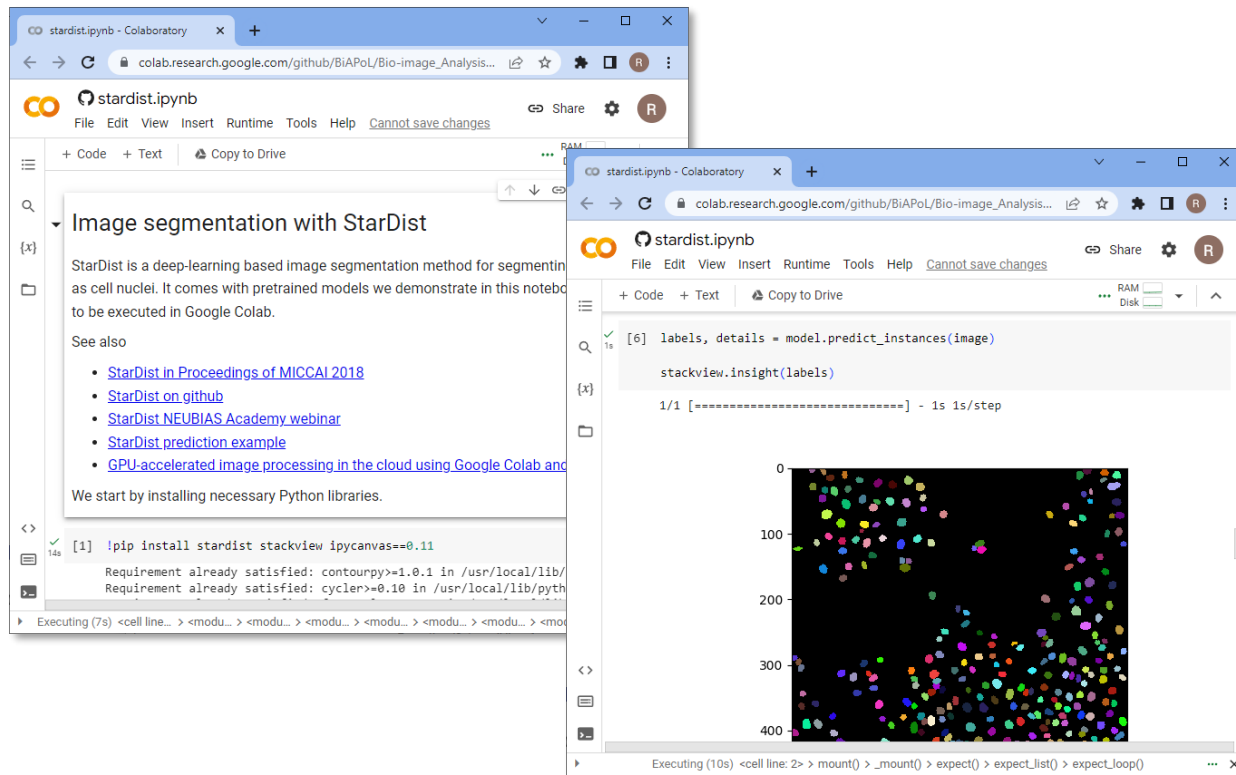
Robert Haase



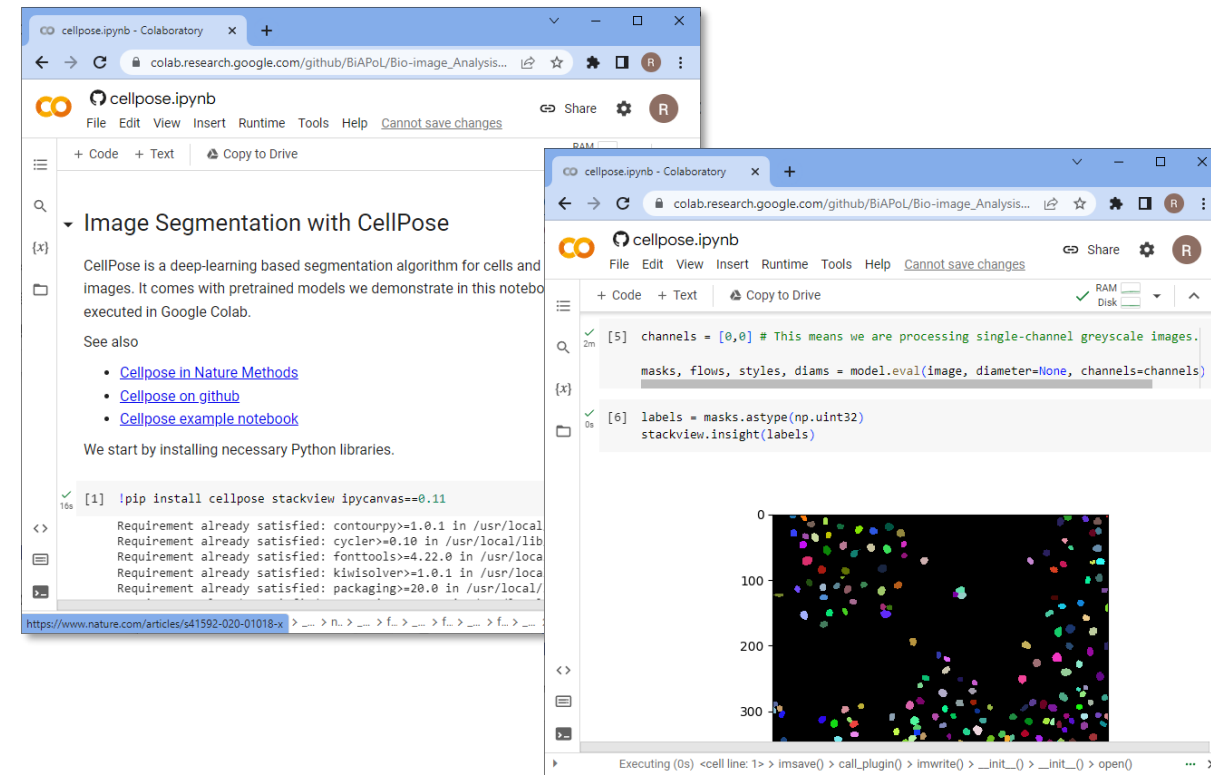
Deep-learning based nuclei segmentation

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/cellpose.ipynb



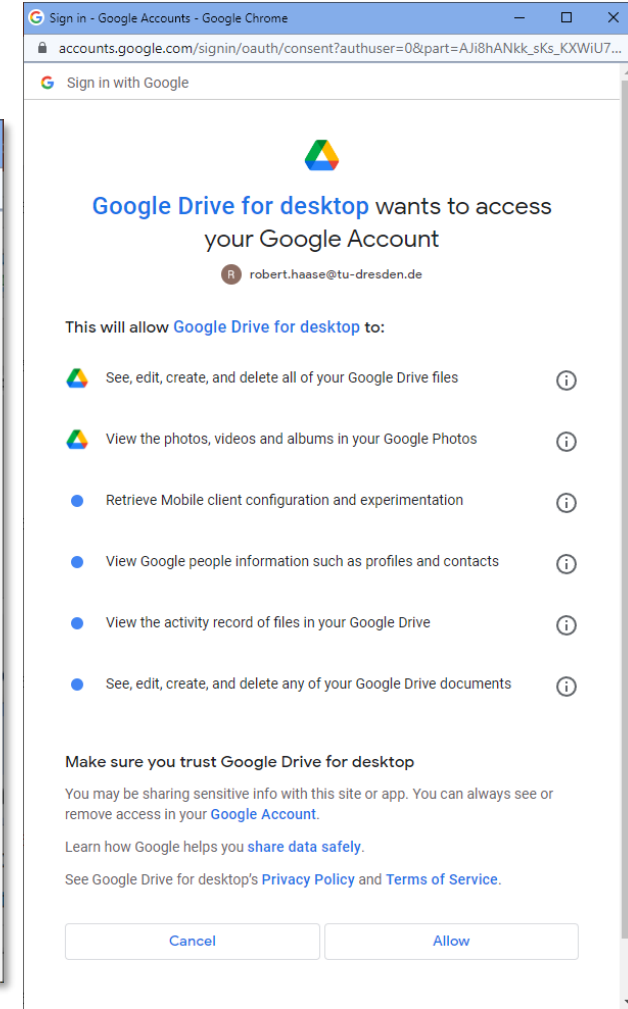
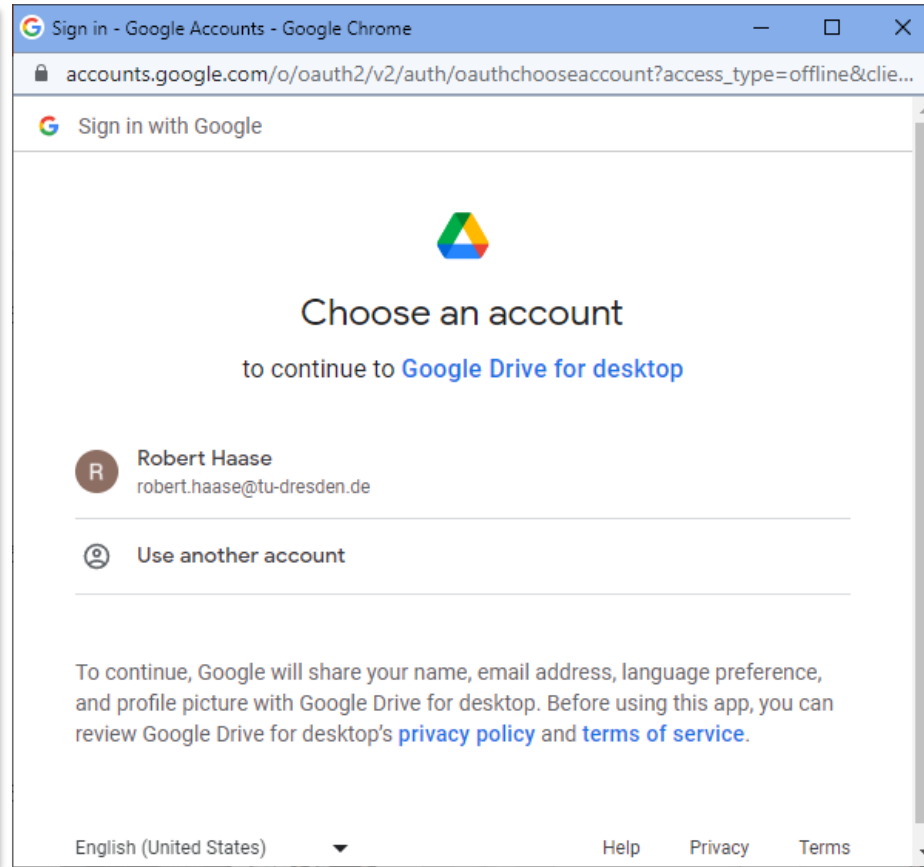
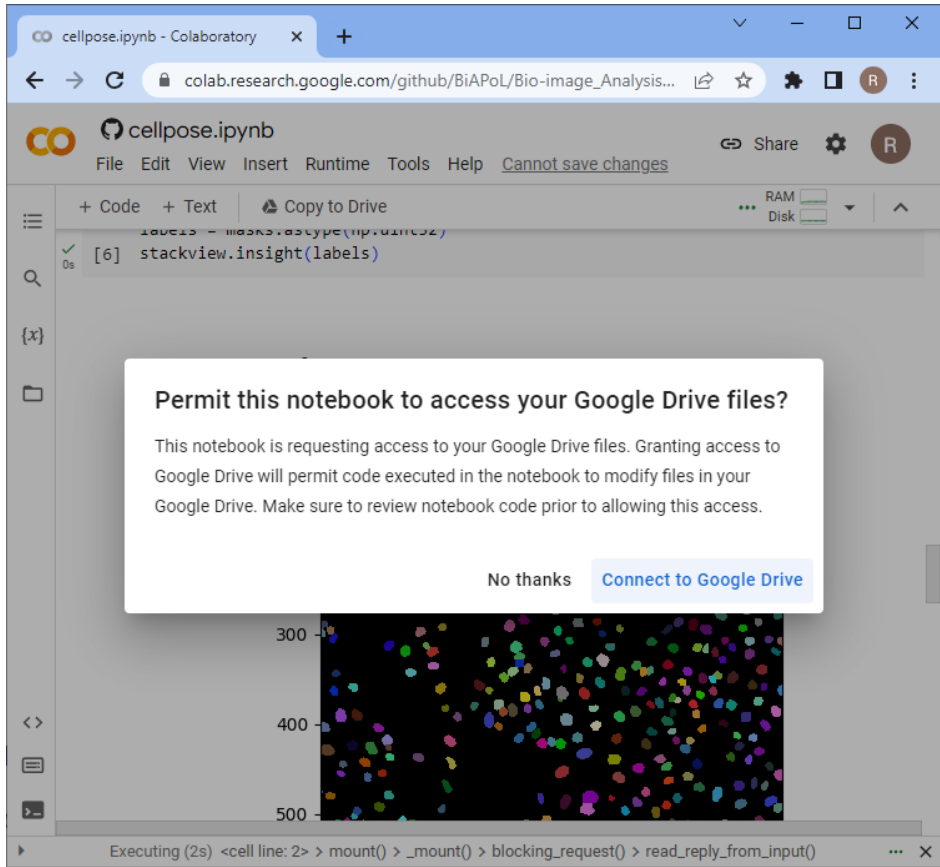
The screenshot shows a Google Colab notebook titled "stardist.ipynb". The main heading is "Image segmentation with StarDist". The text explains that StarDist is a deep-learning based method for segmenting cell nuclei. It includes a list of links for further reading: "StarDist in Proceedings of MICCAI 2018", "StarDist on github", "StarDist NEUBIAS Academy webinar", "StarDist prediction example", and "GPU-accelerated image processing in the cloud using Google Colab and". The code cell shows the installation of libraries and the execution of the segmentation model. The output is a stackview insight of the labels, which is a 400x400 pixel image showing a dense field of multi-colored nuclei on a black background.



The screenshot shows a Google Colab notebook titled "cellpose.ipynb". The main heading is "Image Segmentation with CellPose". The text explains that CellPose is a deep-learning based segmentation algorithm for cells and images. It includes a list of links for further reading: "Cellpose in Nature Methods", "Cellpose on github", and "Cellpose example notebook". The code cell shows the installation of CellPose and the execution of the segmentation model. The output is a stackview insight of the labels, which is a 400x400 pixel image showing a dense field of multi-colored nuclei on a black background.

Connecting Google Drive

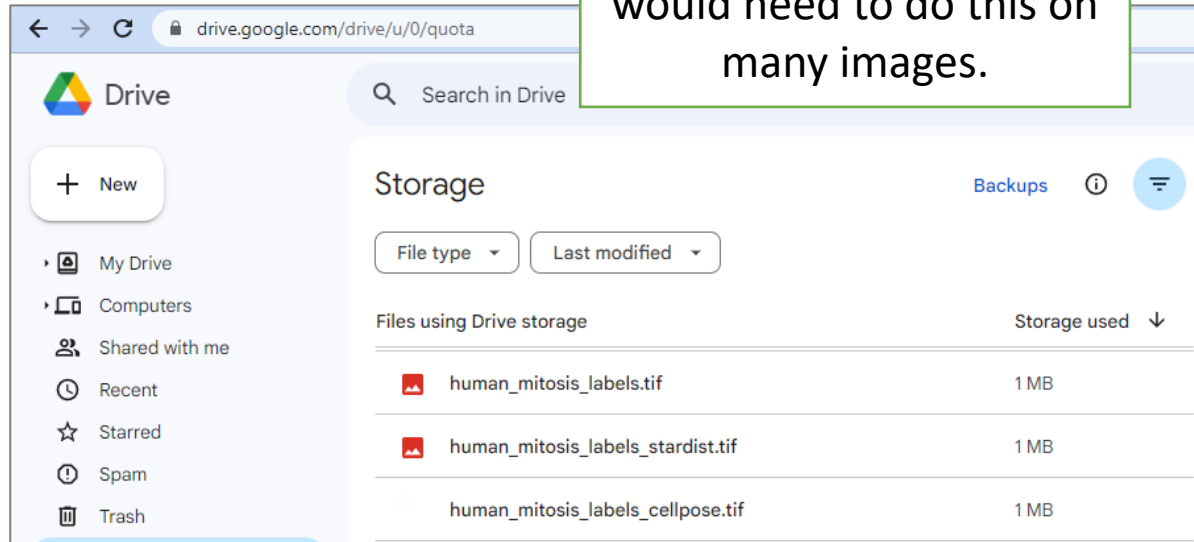
- Store the resulting label images to your Google Drive



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?

Note: In a real setting, we would need to do this on many images.



Pixel-wise versus Object-wise evaluation

- Average Overlap for all ground-truth objects
- <https://github.com/haesleinhuepf/the-segmentation-game>



@haesleinhuepf

April 2023

<https://drive.google.com/drive/u/0/quota>

<https://github.com/haesleinhuepf/the-segmentation-game>

