

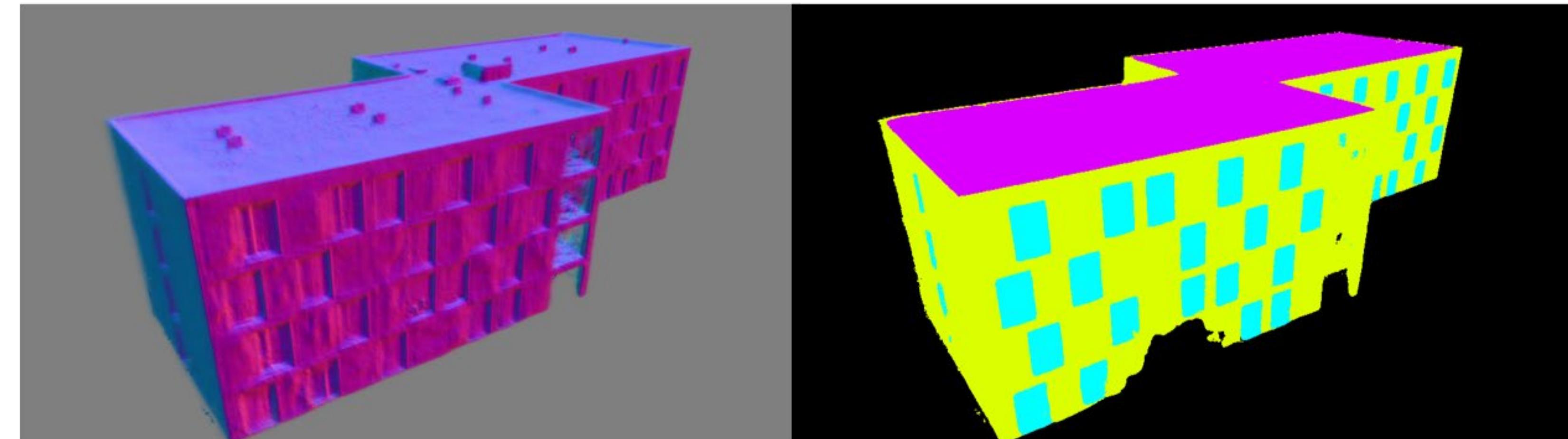
# Introduction to Machine Learning

# Estimating Building Envelope Characteristics



Ground Truth

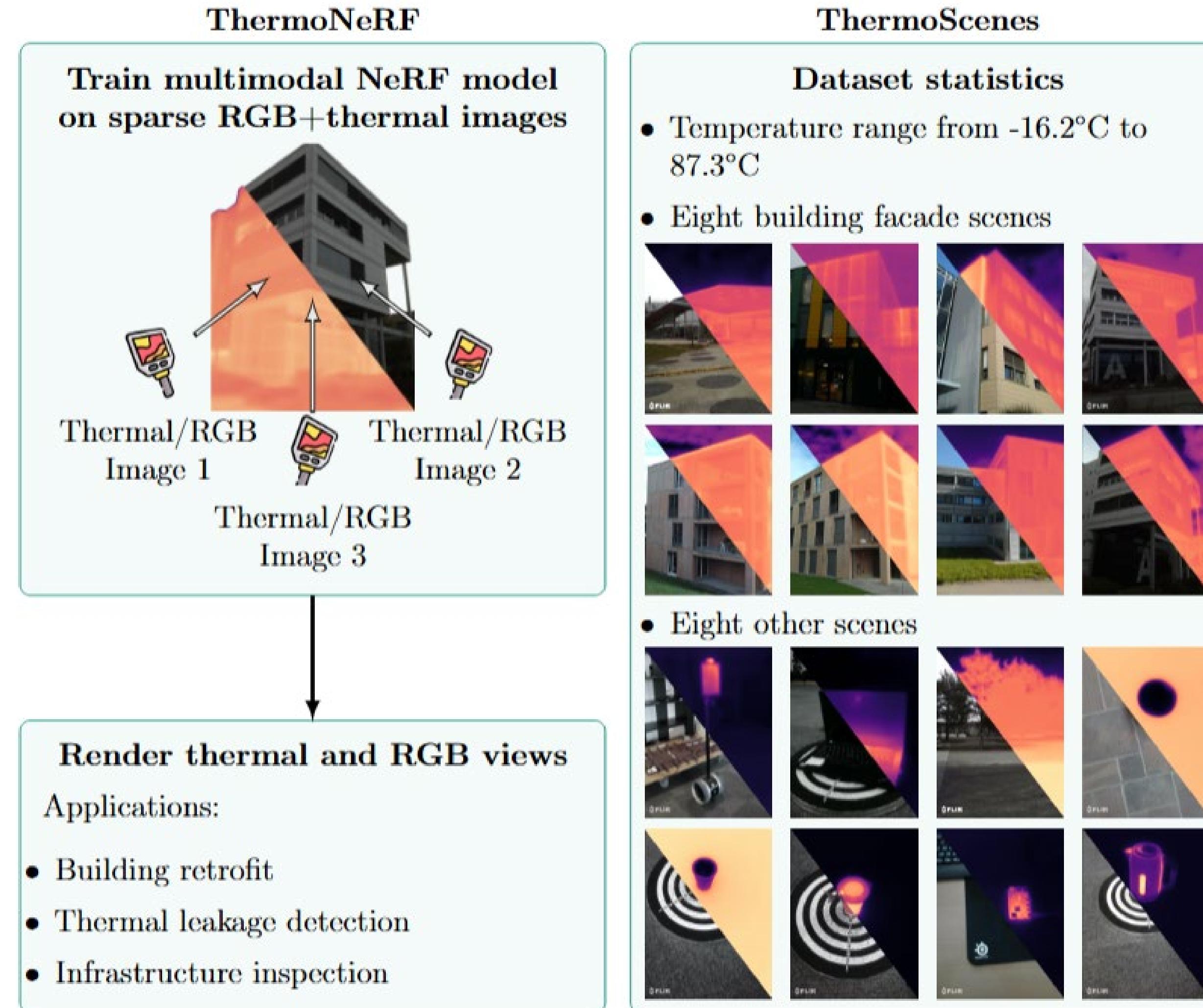
Rendered RGB - BuildNet3D



Rendered Normal - BuildNet3D

Rendered Semantics- BuildNet3D

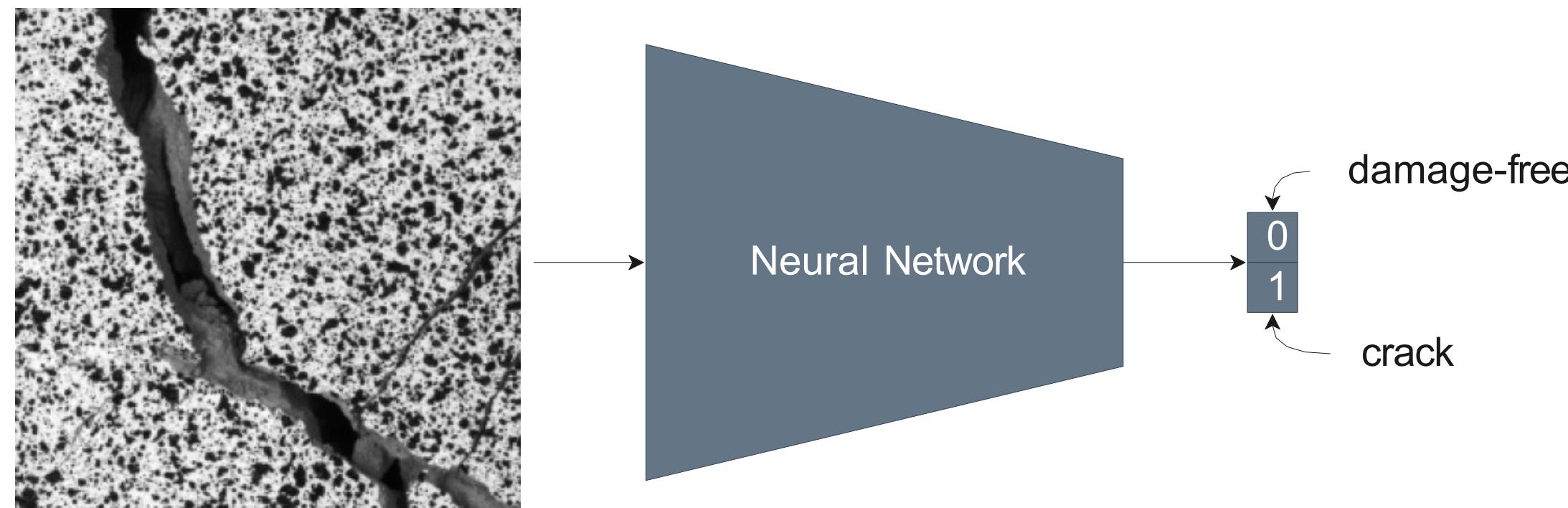
# Thermal 3D reconstructions from sparse images



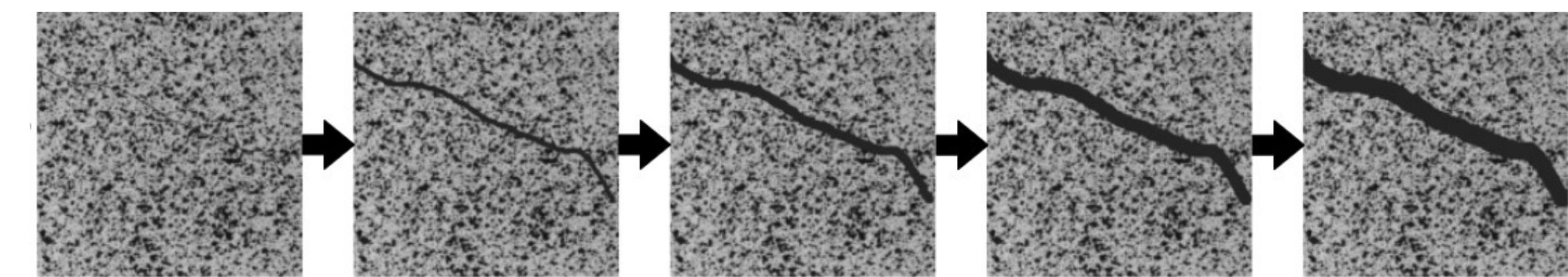
# Machine learning for image-based crack detection

Data-driven approaches based on supervised deep learning have demonstrated excellent performance in detection of cracks in images, but they require **large annotated datasets** for training.

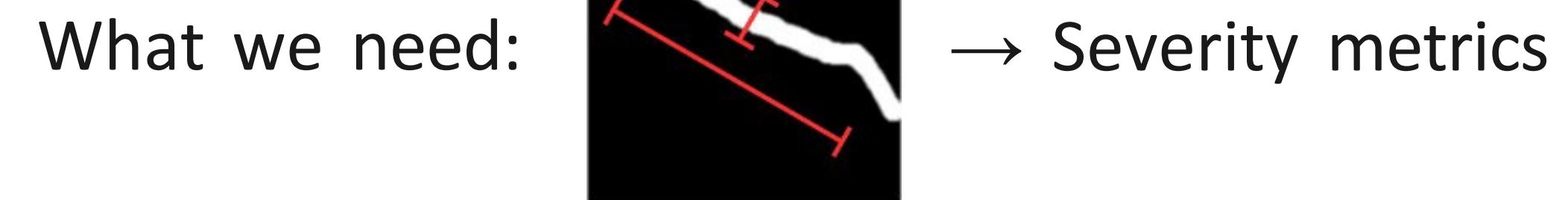
## Classification task:



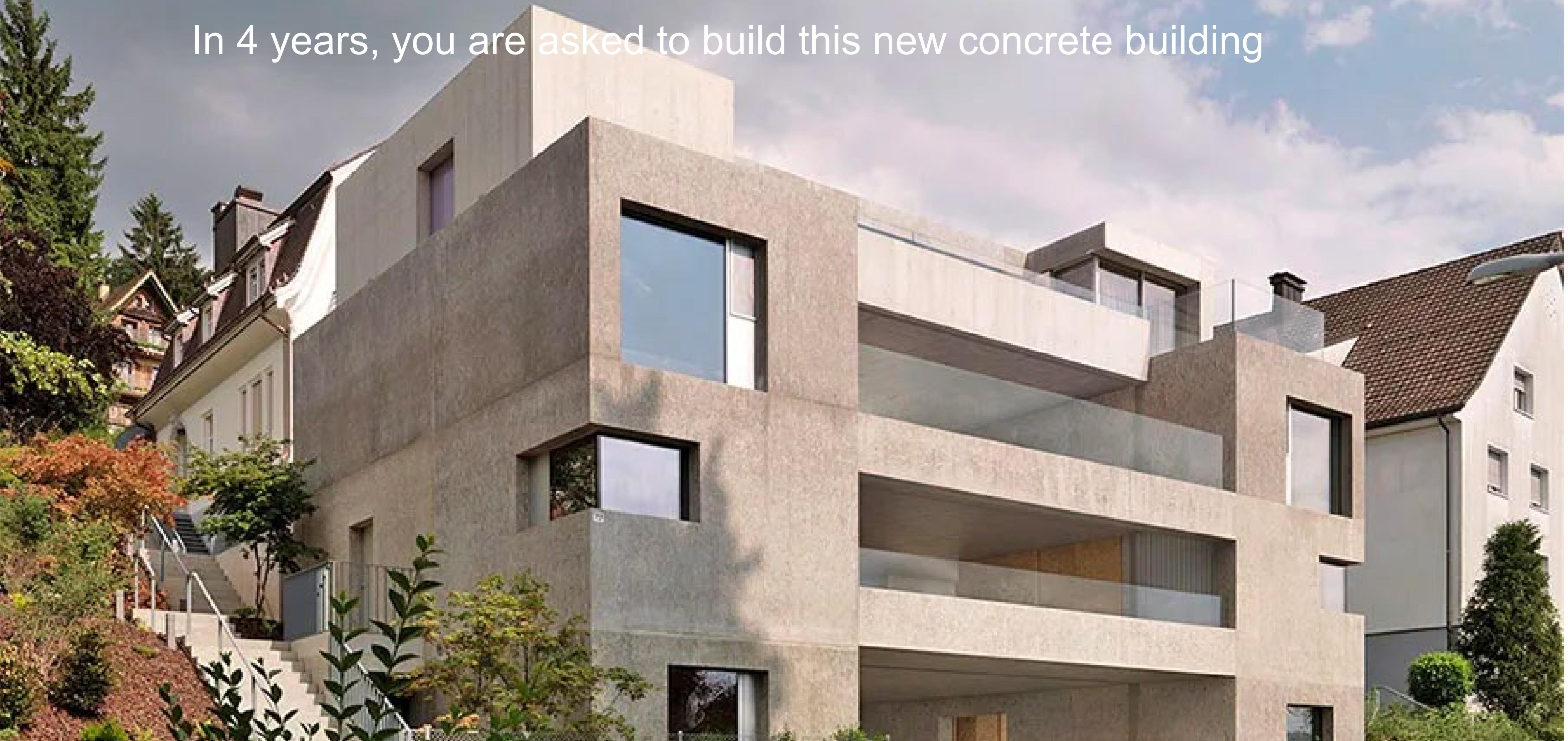
- ▶ Does not allow severity quantification
- ▶ Fast and easy image-level annotation (1 bit)



Severity quantification and monitoring is crucial for timely decision-making.



In 4 years, you are asked to build this new concrete building

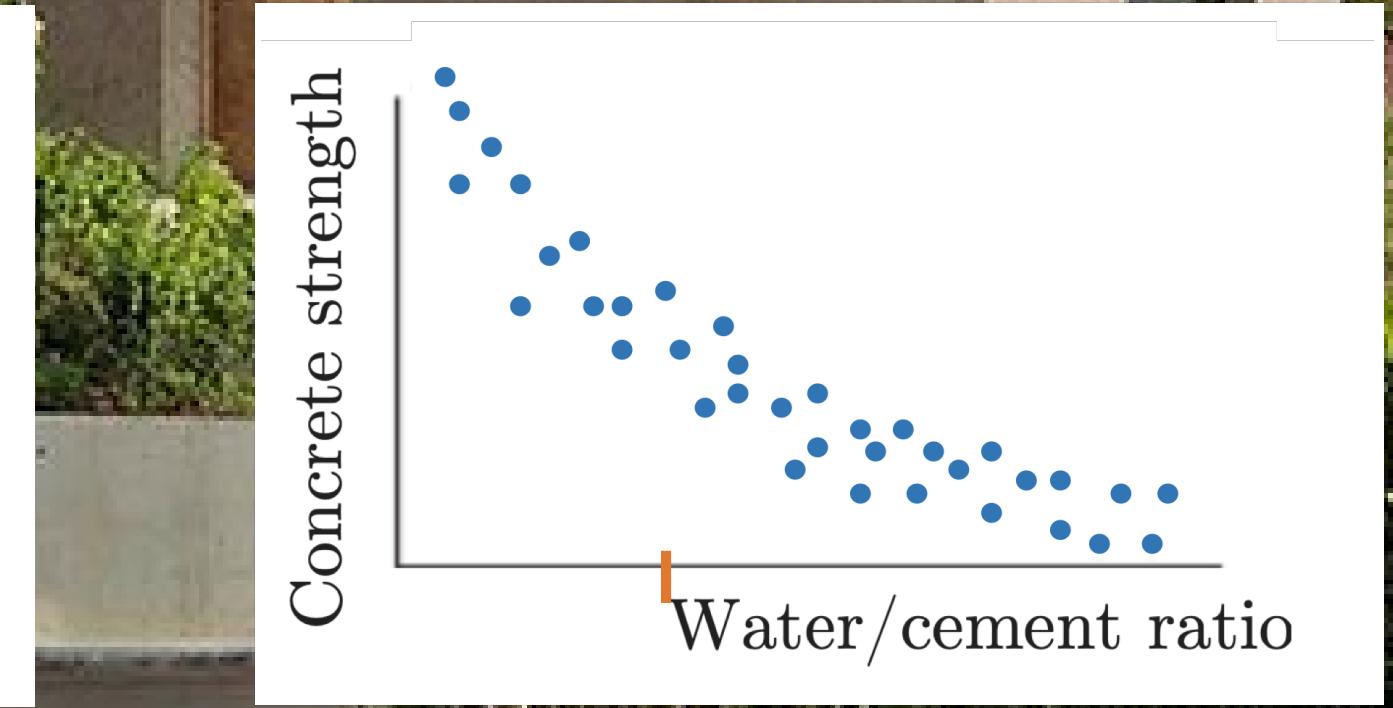


**Situation:**

You received concrete (with a certain water cement ratio) from your supplier and need to check its strength.

**Task:**

How to predict concrete strength based on its water /cement ratio?



# Mathematically,

Let:

$X$  = Water/cement ratio

$Y$  = Concrete strength

Can I find a function  $f$  of parameters  $W$  such that:

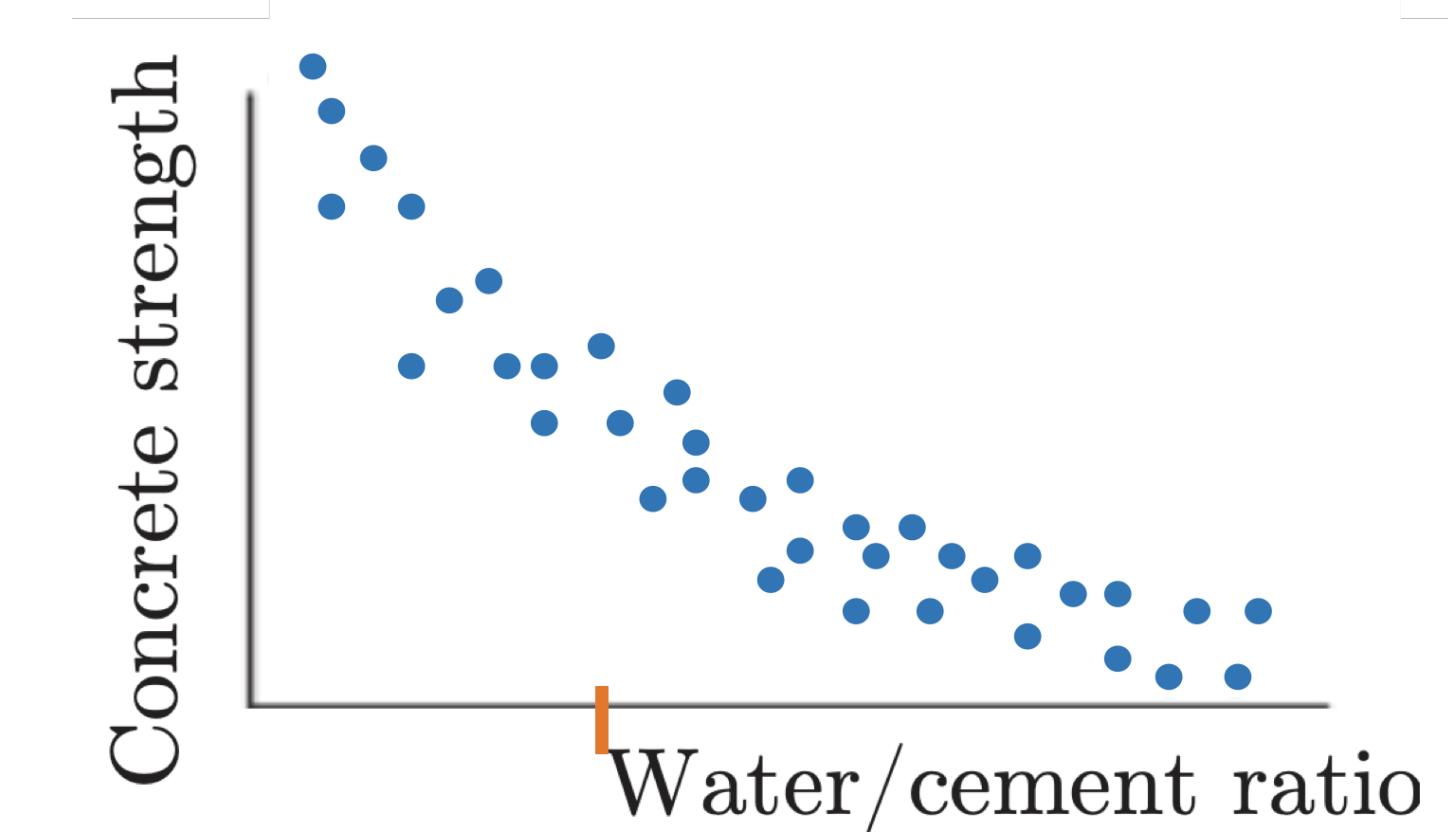
$$Y = f(X, W)$$

$$\text{Attempt1 : } f(X, W) = X^*W_1 + W_2$$

$$\text{Attempt2 : } f(X, W) = X^2*W_1 + X^*W_2 + W_3$$

$W$  is a vector commonly referred to as unknown parameters

The goal of this class is to learn how to find a good function  $f$  and its parameters  $W$ !



**Situation:**

Your goal is to make the production of steel sheet more efficient by identify defects.

**Task:**

How to detect and classify defects from images of steel sheets?

**Mathematically:**

Let,

$X$  = Image (matrix of numbers)

$Y$  = No Defect (0), or Defect (1)

Can I find a function  $f$  of parameters  $W$  such that:  
 $Y = f(X, W)$



No Defect



Defect

14 24 24 45 22 35 54  
17 76 43 54 78 48 72  
65 54 65 34 76 99 32

46 44 34 45 24 35 84  
87 86 63 55 78 48 32  
25 44 65 36 75 97 82



# Welcome in civil-226

# Outline

- Administrative
- AI revolution
- Data revolution
- ML revolution



Artificial Intelligence  
AI



Machine Learning  
ML

# Administrative

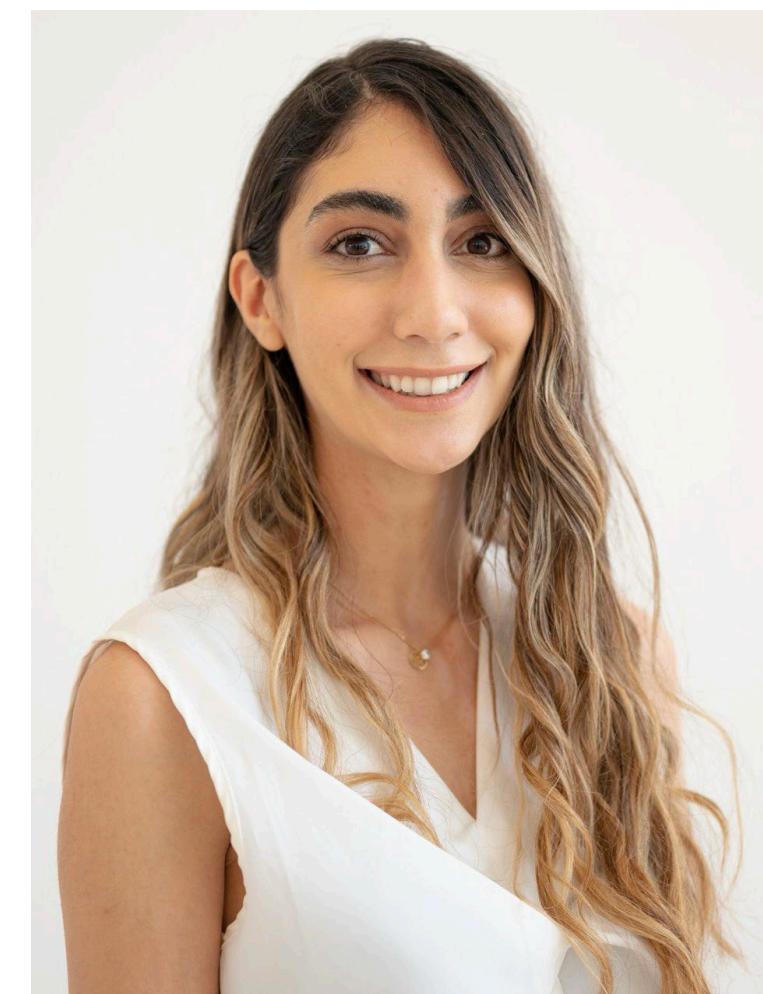
# Administrative Teacher Staff



Olga Fink



Alexandre Alahi



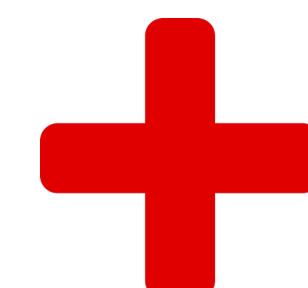
Yasaman  
Haghghi



Leandro Von  
Krannichfeldt



Bastien Van Delft



Amelie Menoud

Pedro Rodrigues Da Costa Savini

# Administrative

## Grading

This course will be graded as follows:

- **Exercises:** Not graded
  - Every week - To practice the techniques learned in class.
- **Graded Homework (individual):** 10%
  - Similar to a weekly exercise, except it will be graded.
- **Projects (in groups):** 40%
  - Coding project to solve a problem using ML.
- **Final Exam:** 50%
  - Standard final exam.

# Administrative Requirements

Domains covered by ML

- Linear algebra
- Calculus
- Probability & statistics
- Programming (in Python)

Course Requirements:

- MATH-111
- CS-119(h)

# Lectures

- Start with a section specific real-world example
- Real-time demo (from any domain)
  - e.g., <https://playground.tensorflow.org>
- Have references to on-line resources/animations
  - e.g., <https://www.deeplearning.ai/ai-notes/optimization/>
- Guest speaker from industry

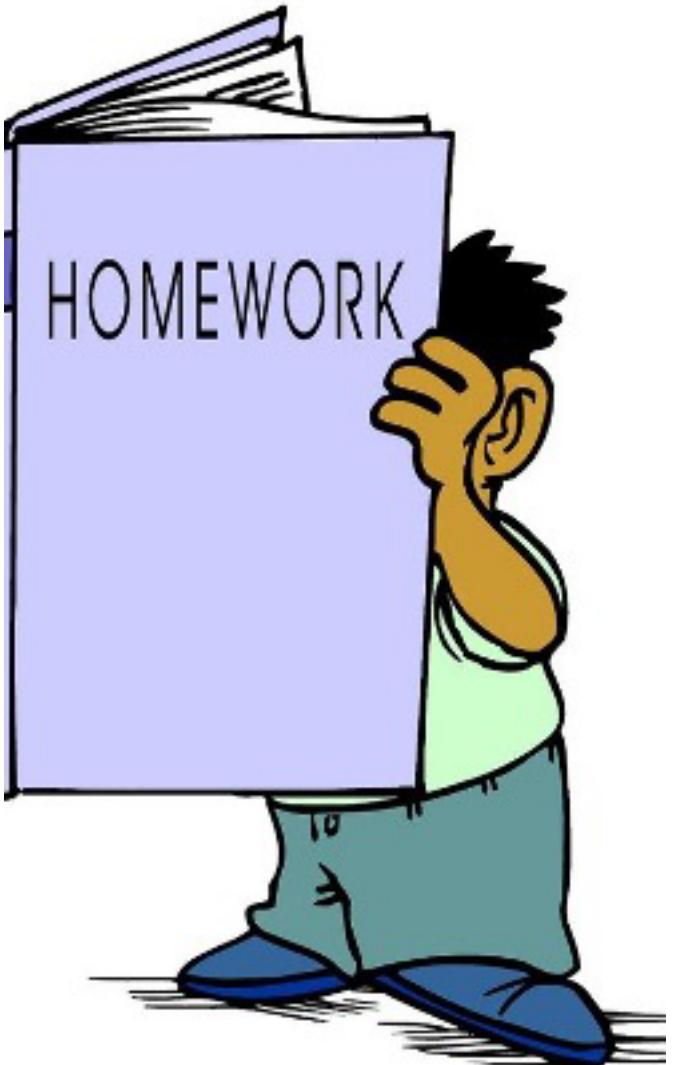
# Exercices

- **Notebooks**

- Visual, direct feedback
- [EPFL NOTO](#)
- 3rd party tools ([colab](#))

- **Generic**

<https://github.com/vita-epfl/IntroML-2025>



# Project

- Alcrowd
  - Leaderboard
  - <https://www.aicrowd.com/challenges/a-i-in-civil-engineering-building-collapse>
- Poster session
- Awards:
  - 1<sup>st</sup> & 2<sup>nd</sup> in the leaderboard
  - Best poster

## A.I. in Civil Engineering

### Building Collapse

By  VITA VITA Lab, EPFL 7837 65 20 1359 6 Follow

[Overview](#) Leaderboard Notebooks Discussion Insights Submissions Rules

#### PROBLEM STATEMENTS



##### Regression - Predicting seismic collapse capacity

Effects of earthquakes and their ground motions on a 4-story steel structure

2084 933



##### Classification - Tsunami induced building collapse detection

Before and after satellite images of the Tohoku tsunami 2011

1537 426

#### Overview

This is the challenge page for the project. It consists in completing the 2 milestones on Building collapse (challenge participation, code submission, and README file with code explanations) as well as a poster to summarize and present your ideas.

#### Submission

Please submit only as a TEAM -> Simply click on create team on the top right of

# Administrative

## Additional resources

### Books

- K.P. Murphy, [Probabilistic Machine Learning: An Introduction](#), 2021
- M.P. Deisenroth, A.A. Faisal, C.S. Ong, [Mathematics for Machine Learning](#), 2020
- J. VanderPlas, [Python Data Science Handbook](#), 2016

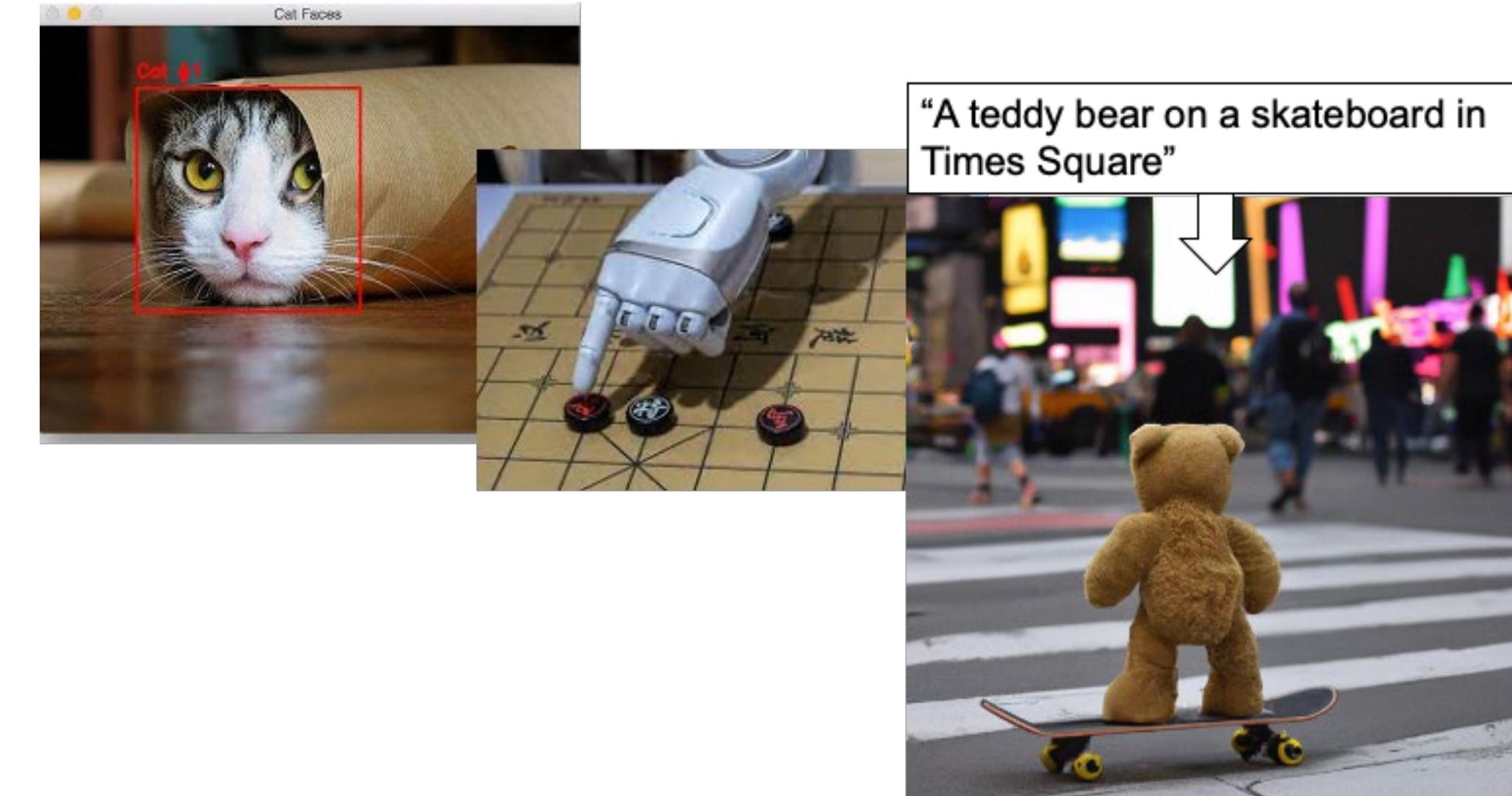
### Online courses

- [Machine Learning Course](#) by Andrew Ng (Stanford)
- [Deep Learning Specialization](#) by Andrew Ng (deep learning.ai)

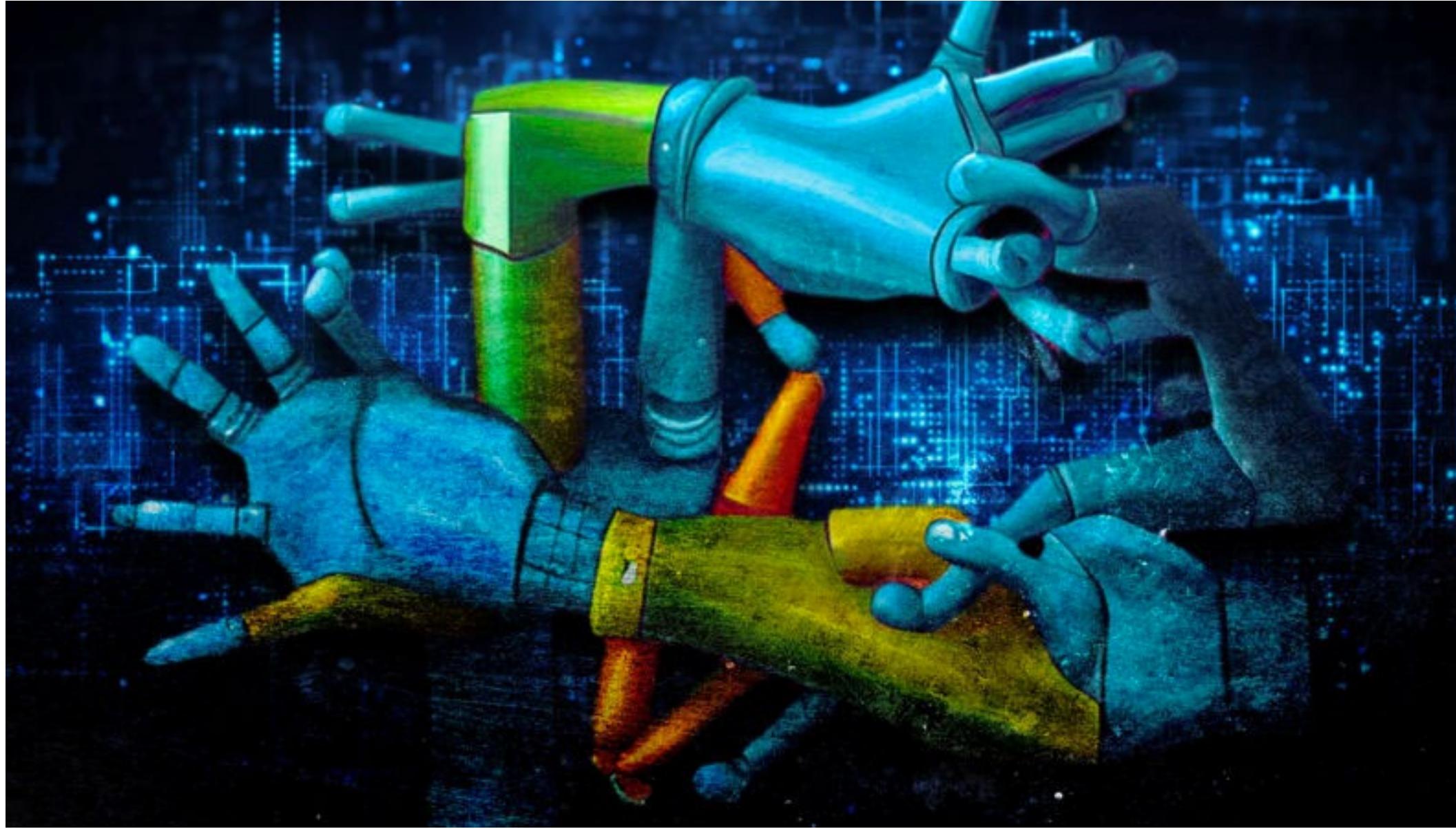
### Other

- [Google Machine Learning Glossary](#)

# AI revolution



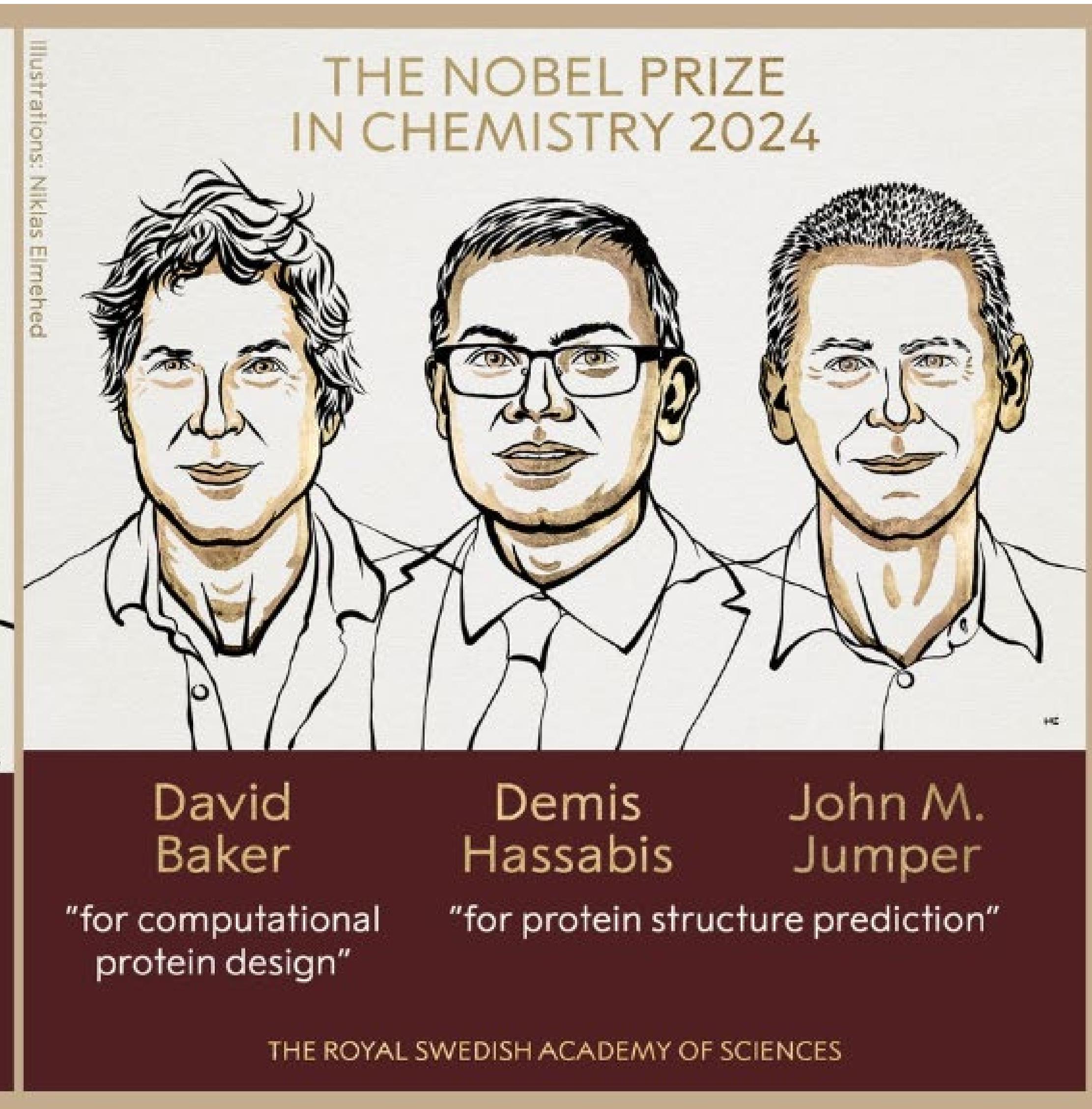
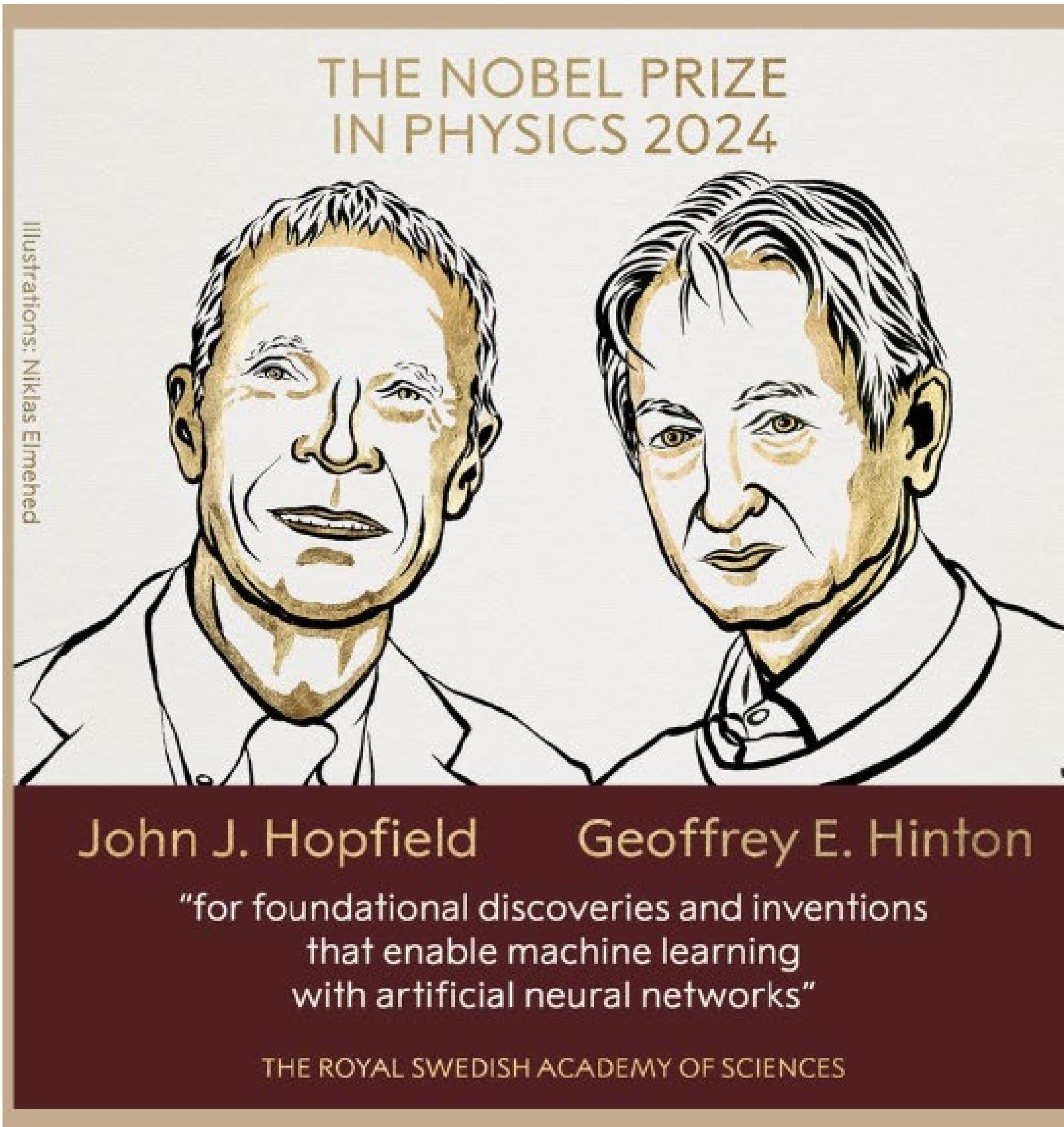
“a pair of robot hands holding pencils drawing a pair of human hands, oil painting, colorful,”



“a shark that looks like a car that looks like a face”



# AI Takes the Nobel Prize



# What is AI?

# What is AI?

AI is intelligence demonstrated by machines...

# What is AI?

## ...acting like a human?

**But:**

- Not all 'intelligent' behaviour is something that humans can do.
- For example, automated 'intelligent' assembly robots can operate with much greater speed and precision than a human can.

**Some say:**

- Building intelligent machines by imitating human behaviour is like building a plane by imitating the way birds fly.
- Instead, why not try and understand the principles behind that behaviour?



## ...thinking like a human?

**But:**

- Human thinking is **not good at everything**.
- For example, human thinking is good at recognising faces...
- ...but bad at multiplying large numbers.

**Also:**

- Human thinking is affected by **biases** which cloud our decisions.
- Should these biases be included in intelligent machines too?



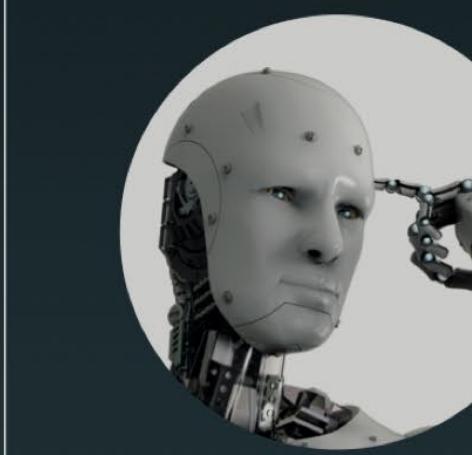
## ...thinking rationally?

**But:**

- If 'thinking rationally' is about using logic to make decisions...
- ...not all decision-making can **easily be expressed in logic**.
- For example, when driving a car and making a quick decision, we do not think in words. We just know what to do instinctively.

**So:**

- There is more to intelligence than just thinking rationally.



## ...acting rationally?

**This means:**

- Acting in the way most likely to achieve an objective.
- Using both logic-based thinking and intuitive thinking depending on what is more appropriate for the situation.

**So acting rationally:**

- Does not depend on acting like something we already see.
- Does not depend on thinking in a particular way.
- Is the **most flexible definition of intelligence**.



**What is Intelligence? Ability to accomplish complex goals**

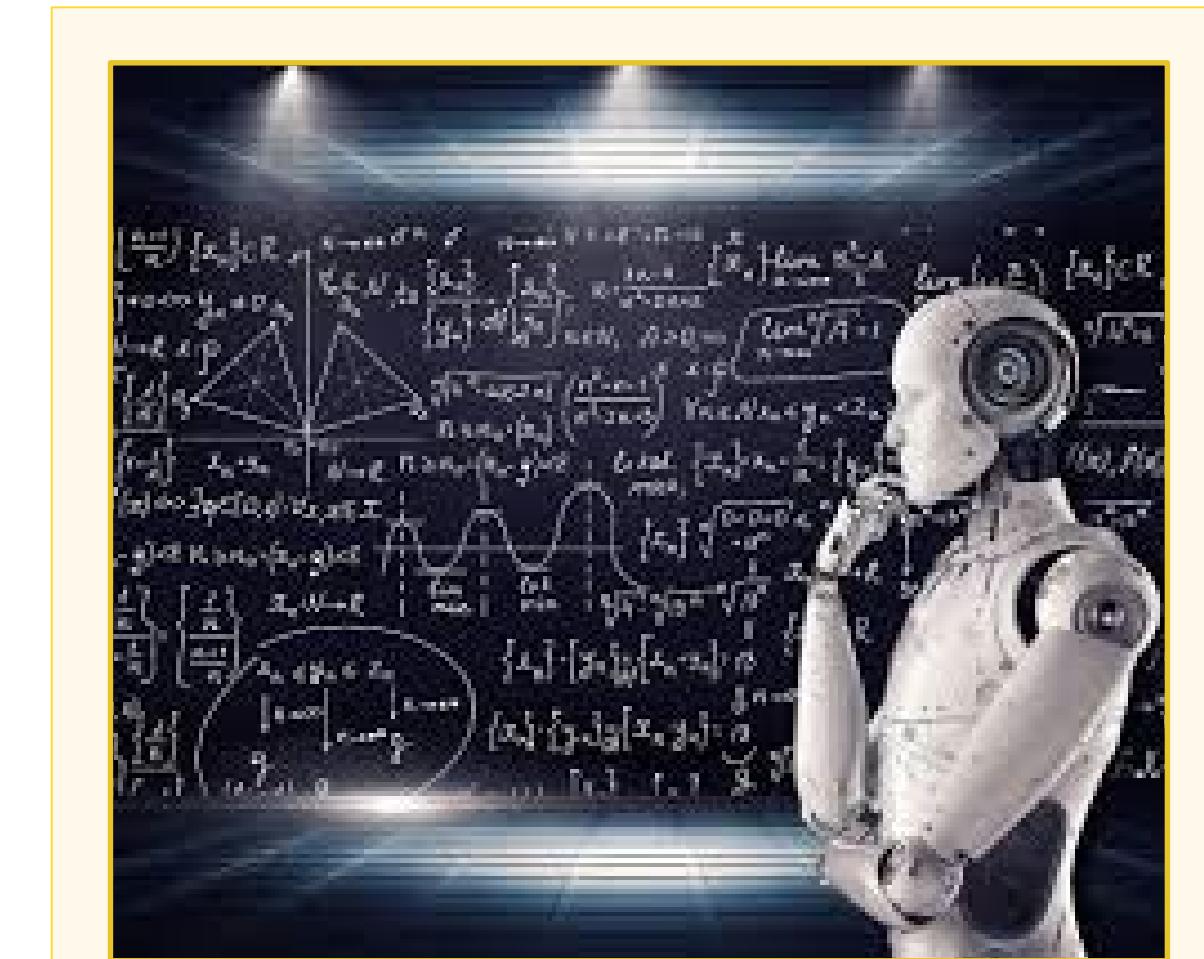
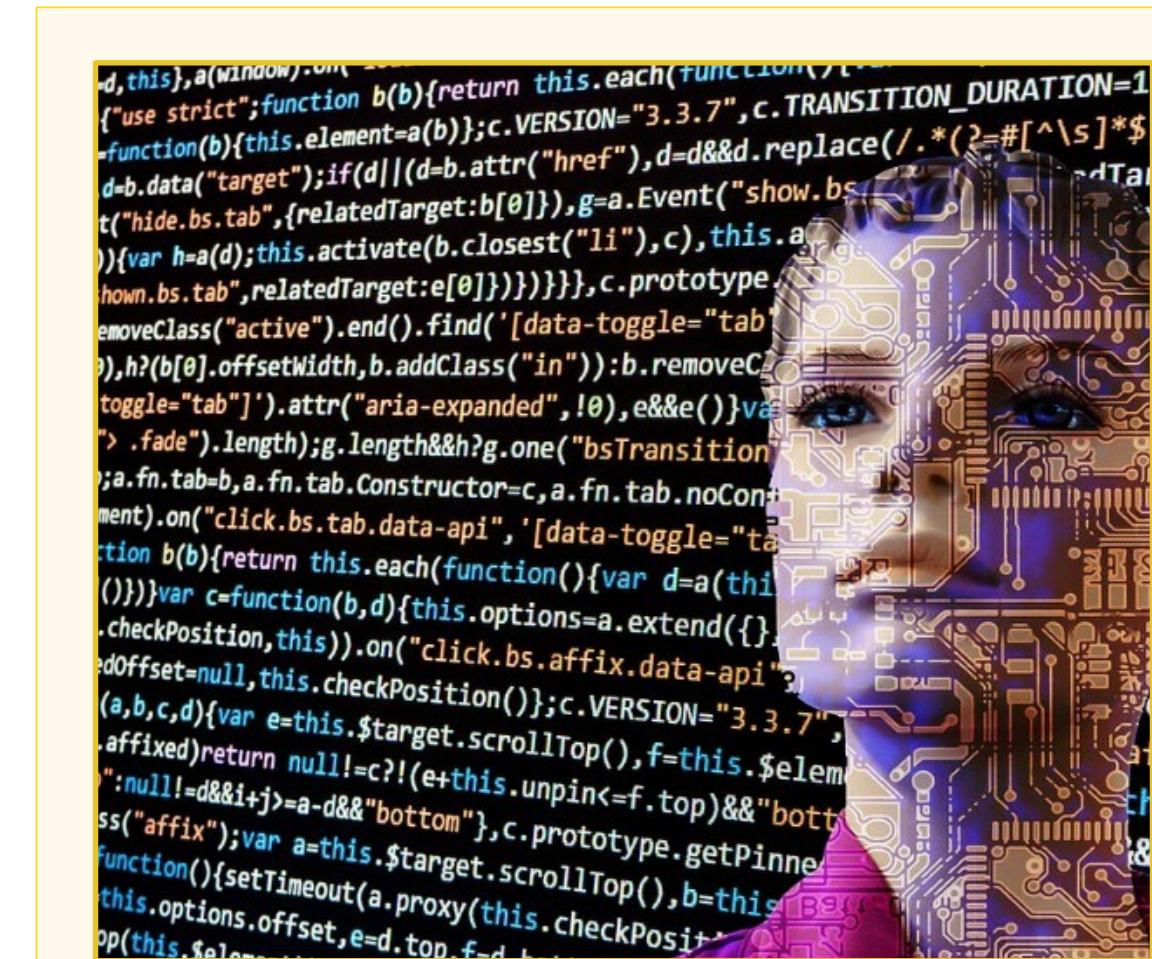
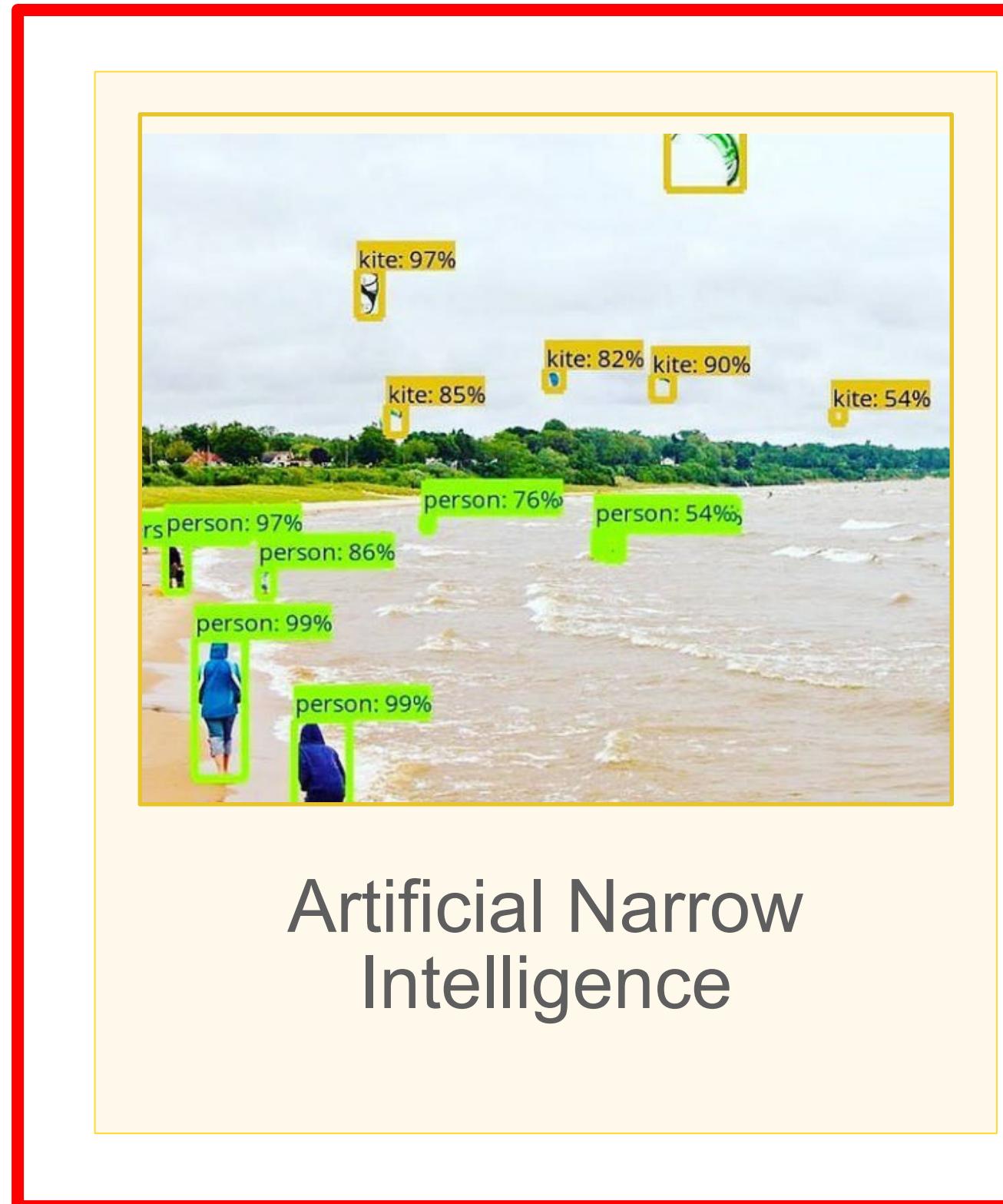
# What is AI?

“any device/system/software that **perceives** its environment  
and **takes actions** that maximise success for its goal”

## What is artificial intelligence?

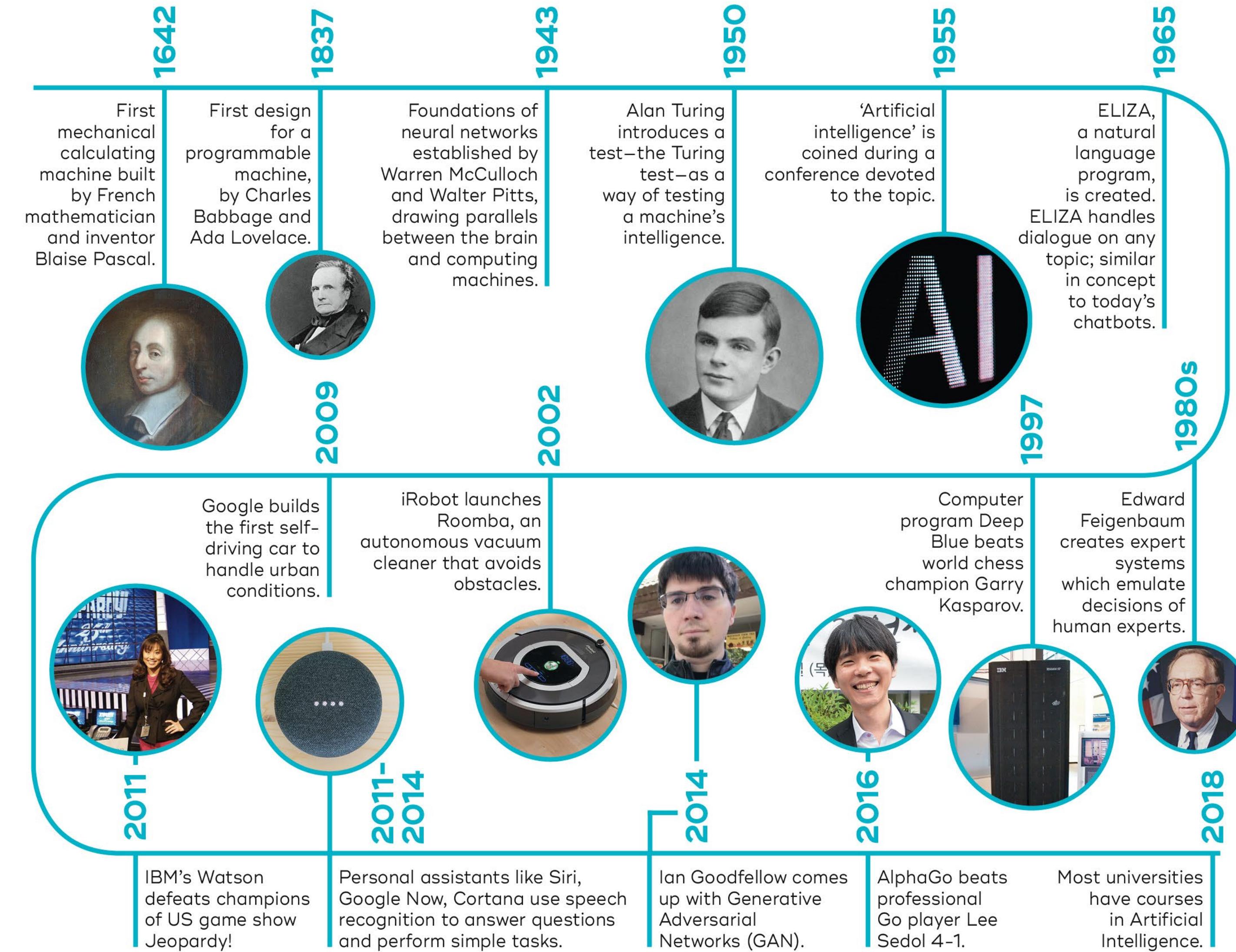
Artificial intelligence (AI) refers to the ability of a computer or machine to perform tasks that would normally require human intelligence, such as learning, problem solving, decision making, and language understanding.

# In which areas does AI excel today?

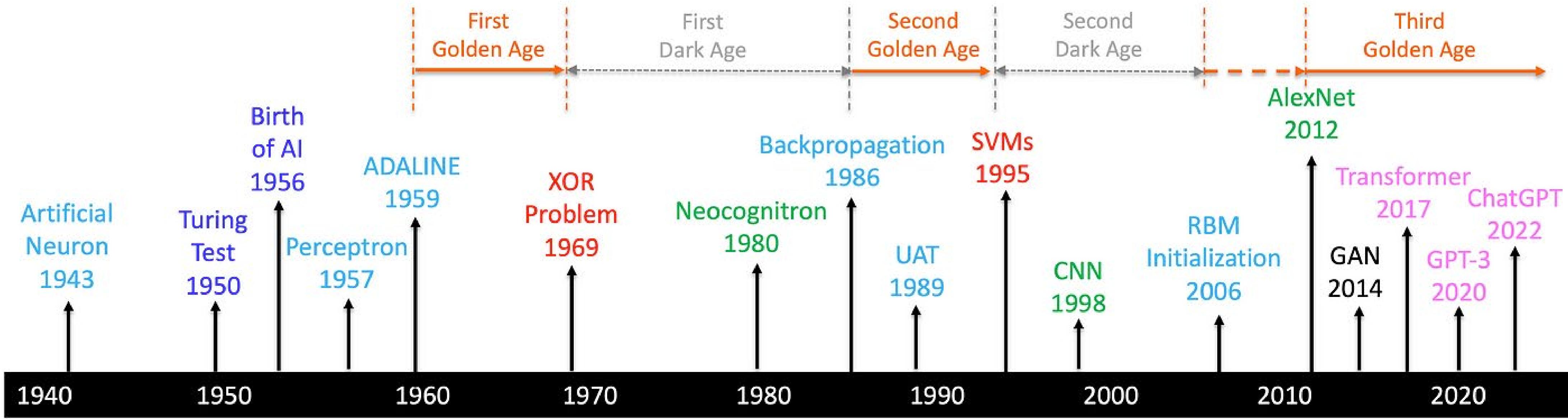


# History of AI

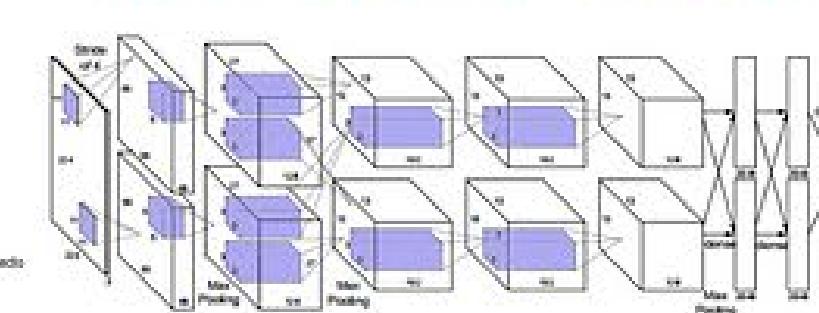
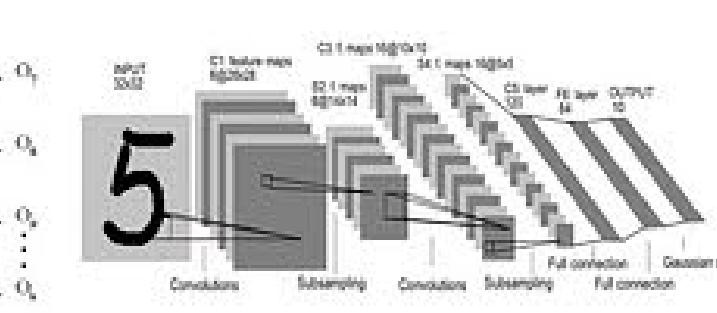
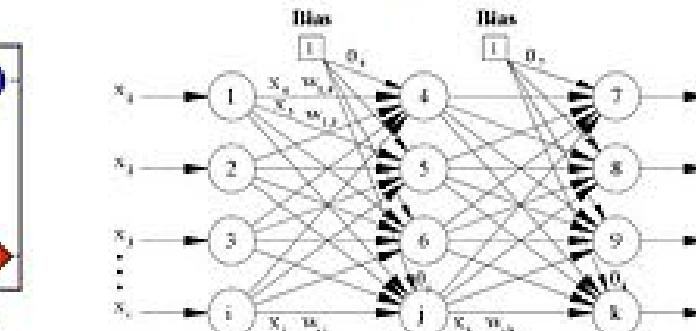
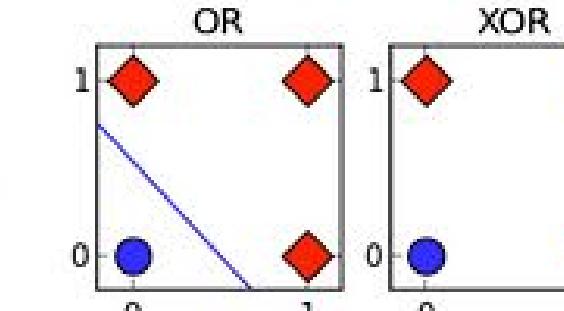
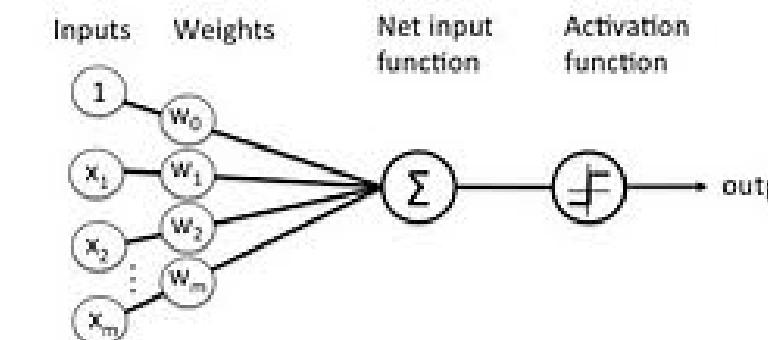
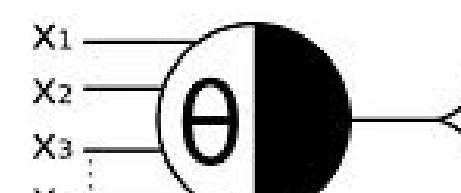
# History of AI



# History of AI: a bit different perspective



## ■ Intro to ML for engineers



# History of AI

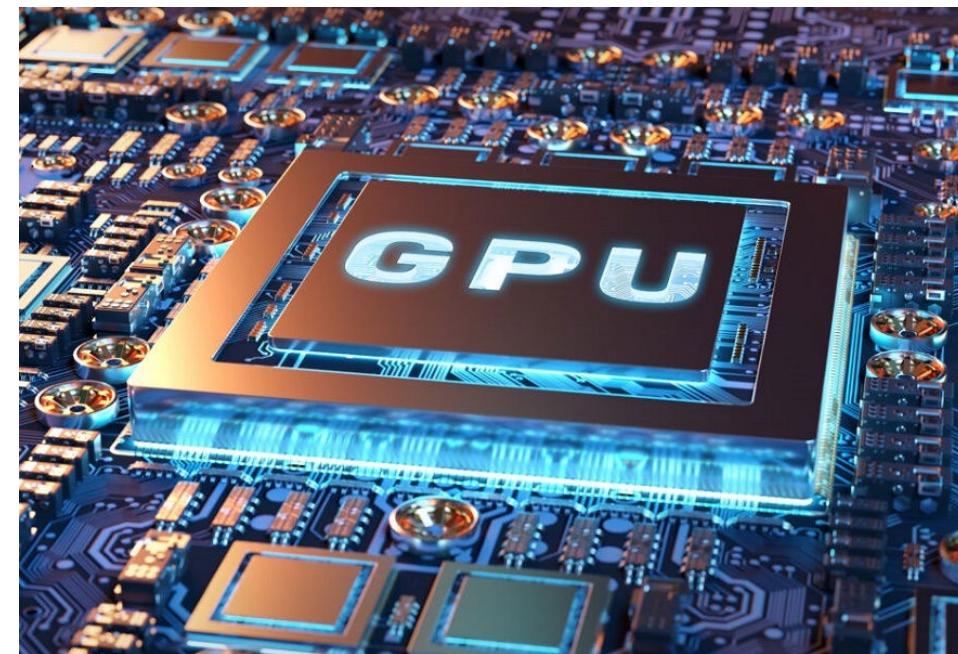
## Interplay between Data and algorithms

Most AI breakthroughs used old algorithms but recent data sets.

Year	Breakthroughs in AI	Datasets (First Available)	Algorithms (First Proposed)
1994	Human-level spontaneous speech recognition	Spoken Wall Street Journal articles and other texts (1991)	Hidden Markov Model (1984)
1997	IBM Deep Blue defeated Garry Kasparov	700,000 Grandmaster chess games, aka “The Extended Book” (1991)	Negascout planning algorithm (1983)
2005	Google’s Arabic- and Chinese-to-English translation	1.8 trillion tokens from Google Web and News pages (collected in 2005)	Statistical machine translation algorithm (1988)
2011	IBM Watson became the world Jeopardy! champion	8.6 million documents from Wikipedia, Wiktionary, Wikiquote, and Project Gutenberg (updated in 2010)	Mixture-of-Experts algorithm (1991)
2014	Google’s GoogLeNet object classification at near-human performance	ImageNet corpus of 1.5 million labeled images and 1,000 object categories (2010)	Convolution neural network algorithm (1989)
2015	Google’s Deepmind achieved human parity in playing 29 Atari games by learning general control from video	Arcade Learning Environment dataset of over 50 Atari games (2013)	Q-learning algorithm (1992)
<b>Average No. of Years to Breakthrough:</b>		<b>3 years</b>	<b>18 years</b>

# Why is everyone talking about AI now?

Some pillars of AI aren't recent. So why is everyone excited about it now?



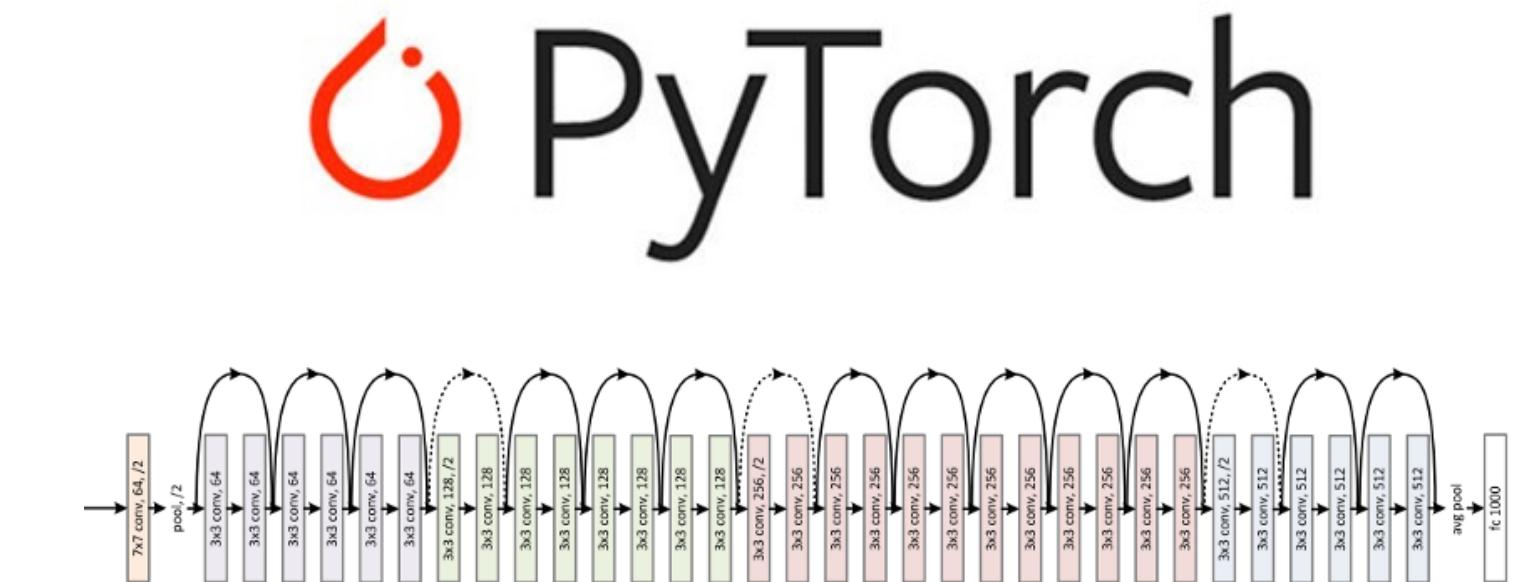
## Hardware

Progress in hardware  
such as Graphical  
Processing Unit (GPU)



## Data

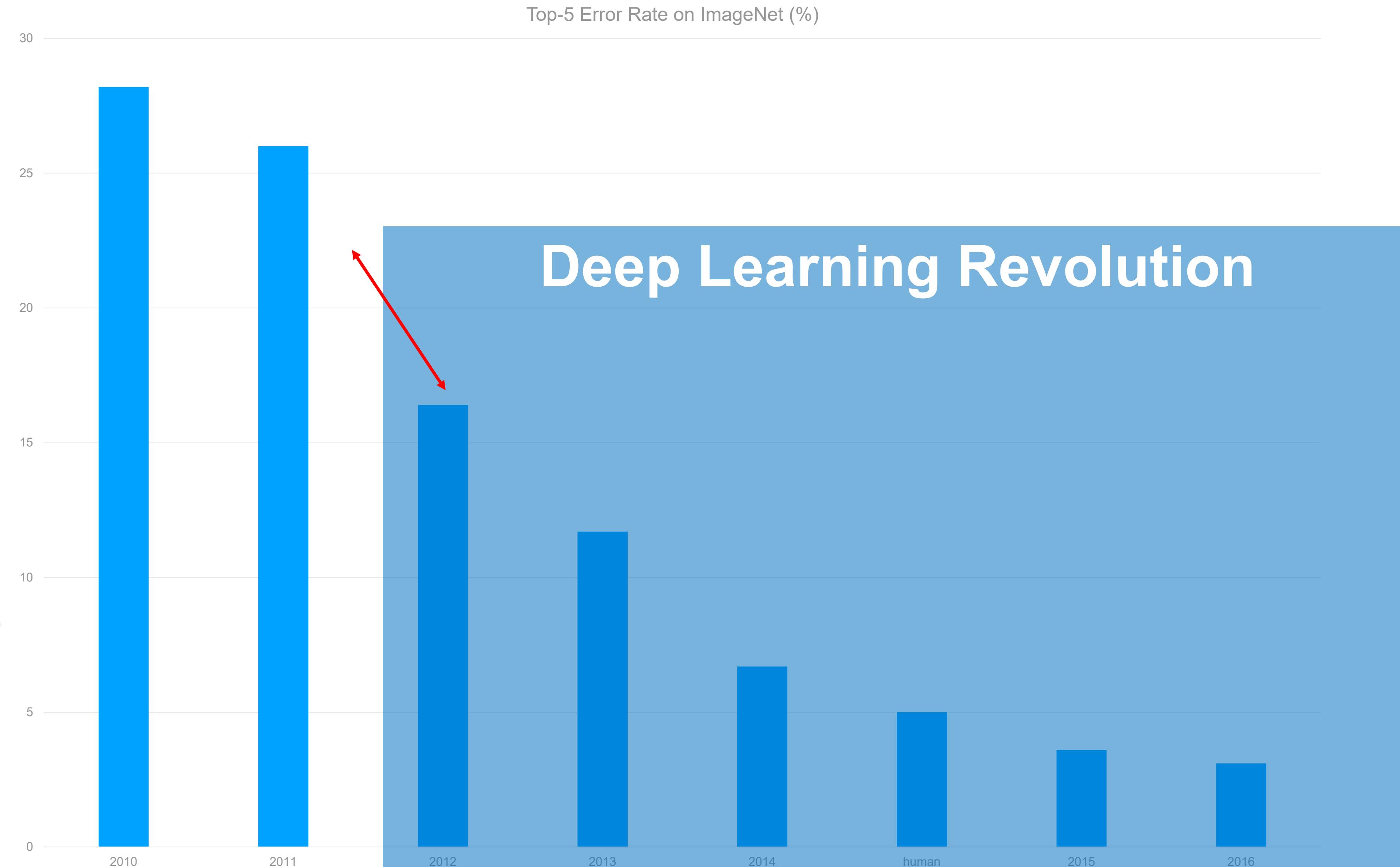
Massive amount of  
labeled data



## Open Science/Software

Shared models/frameworks

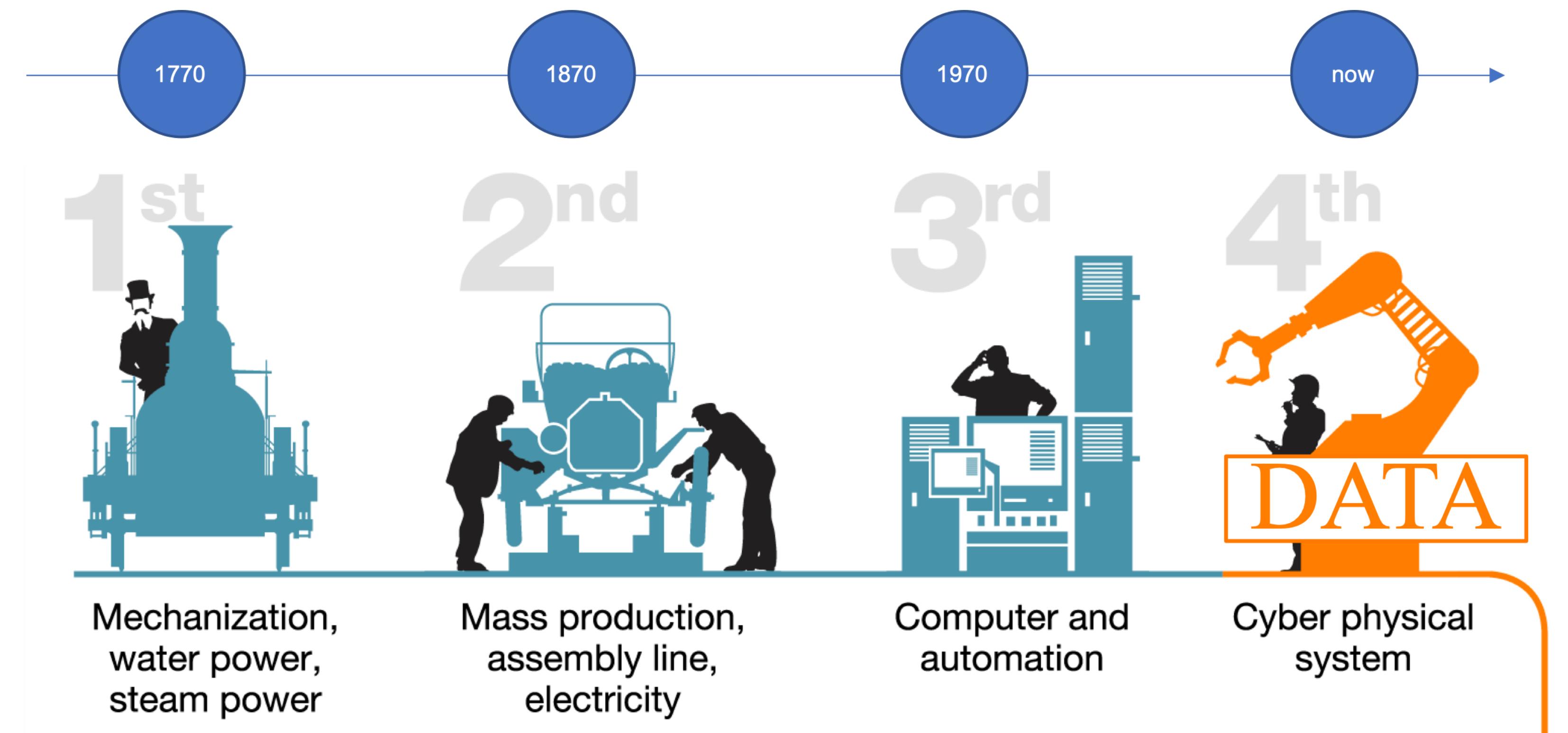
# Power of depth



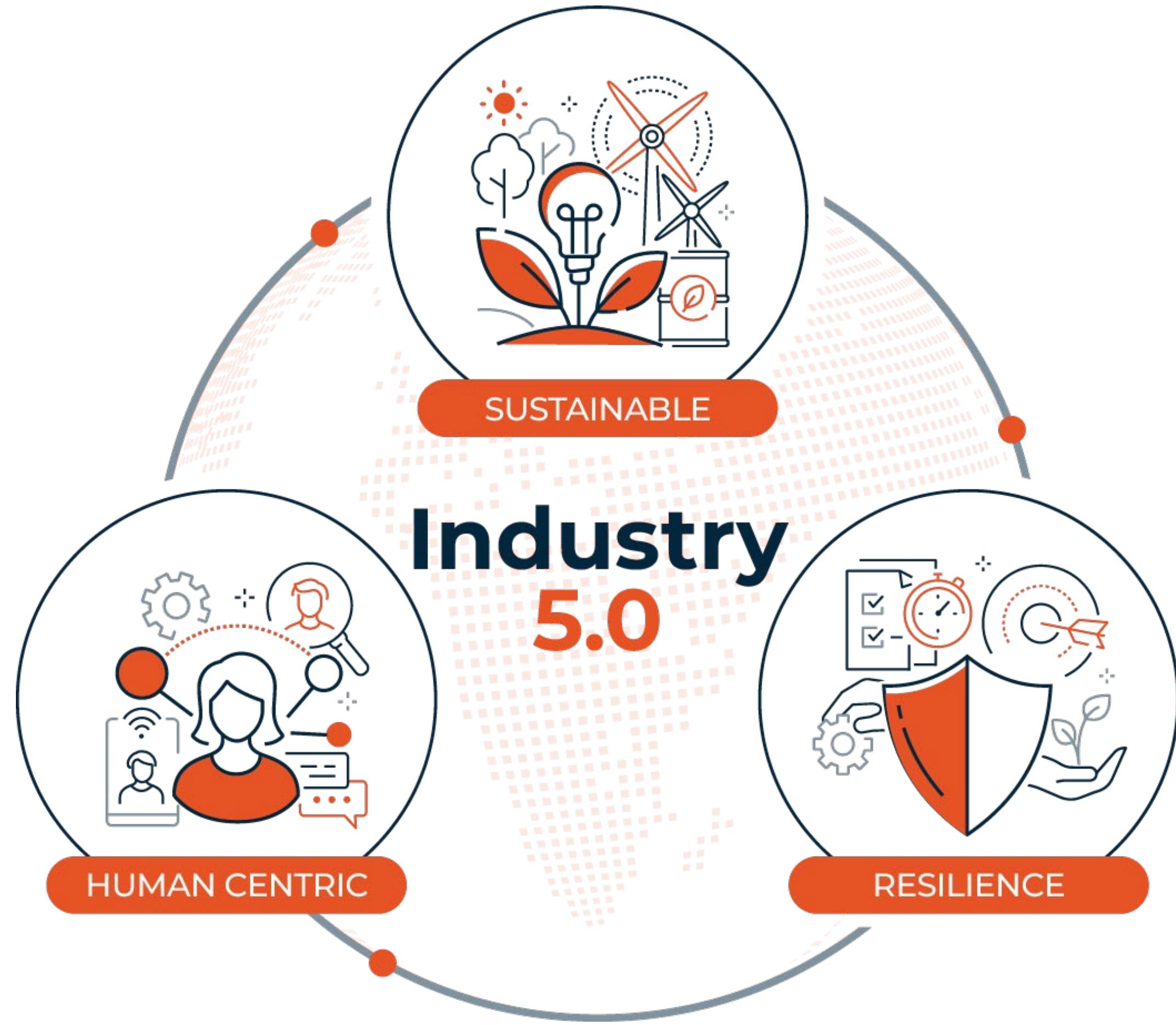
# Data revolution

# Data revolution

## 4<sup>th</sup> industrial revolution

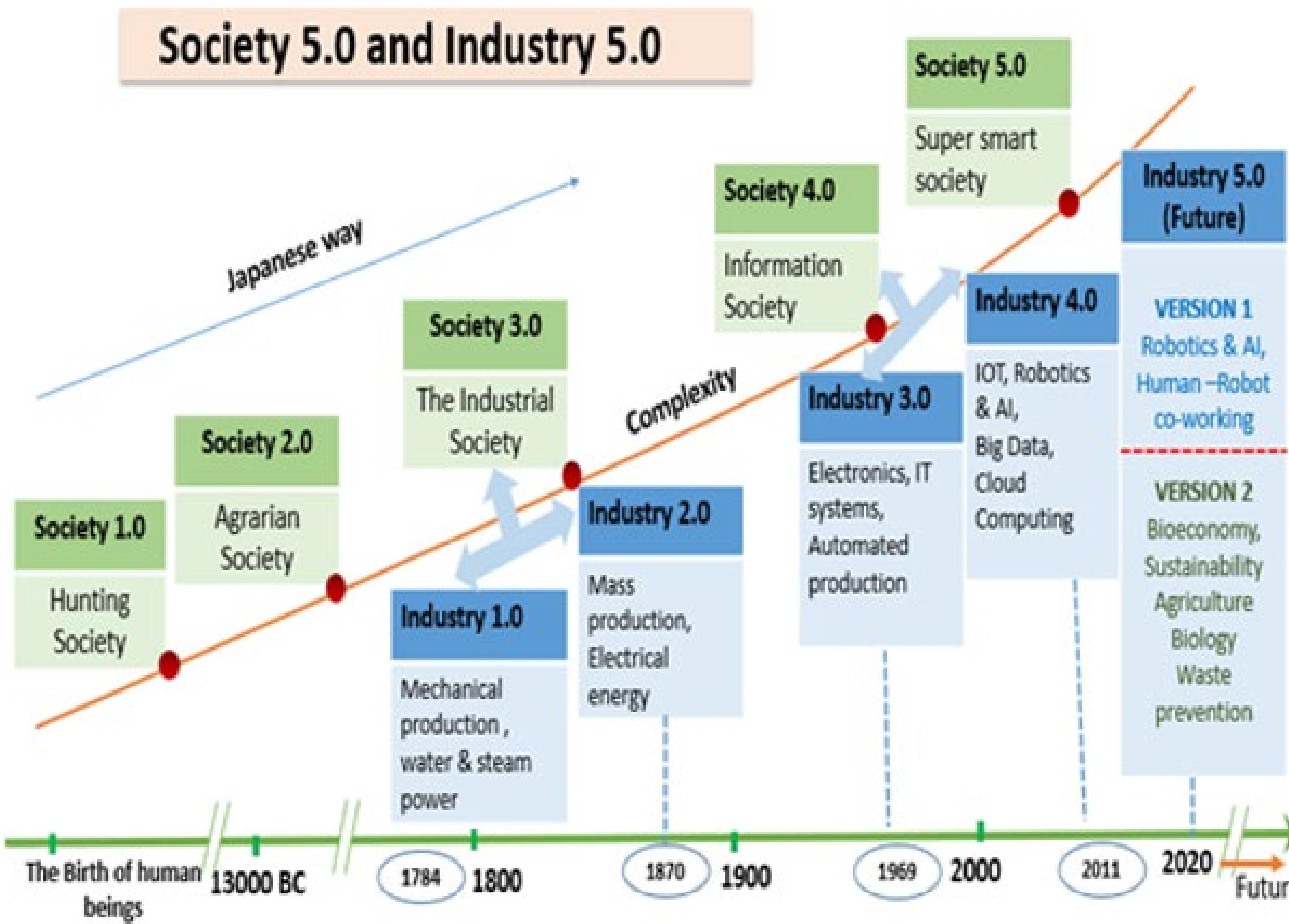


# From Industry 4.0 to Industry 5.0



Source: Momenta Partners

# A new industrial/societal revolution?



# Source: SCHUMPETER CIRCLE

# Data revolution

## Amount of data



Data is everywhere

... and Machine Learning is a tool to learn from Big Data.

# ML revolution

# What is ML? Is it AI?

## AI vs ML



Artificial Intelligence

# What is ML? Is it AI?

## AI vs ML



Artificial Intelligence



Machine Learning

# What is ML? Is it AI?

## AI vs ML



Artificial Intelligence



Machine Learning

What is the difference between AI and machine learning?

Machine learning is considered as a part of AI

Other methods for achieving AI include

- rule-based systems,
- evolutionary computation
- expert systems...

# What is ML?

- Instead of listing out the rules, let machines automatically learn how input data is correlated with a given task/objective/outcome/output.
- Let machines learn from experience/examples/data/feedbacks...

Machine learning research is part of research on artificial intelligence, seeking to provide knowledge to computers through data, observations and interacting with the world. That acquired knowledge allows computers to correctly generalize to new settings.

Defintion by Yoshua Bengio, Université de Montréal

# Examples of ML

# Is it a ML example?



Video credit: Tesla  
<https://www.tesla.com/autopilot>

# Examples of ML

## Research example - GAN

The following people don't exist... They are generated by an AI

Try yourself at: <https://thispersondoesnotexist.com>



# Examples of ML

## Research example - GAN

GAN can do much more.

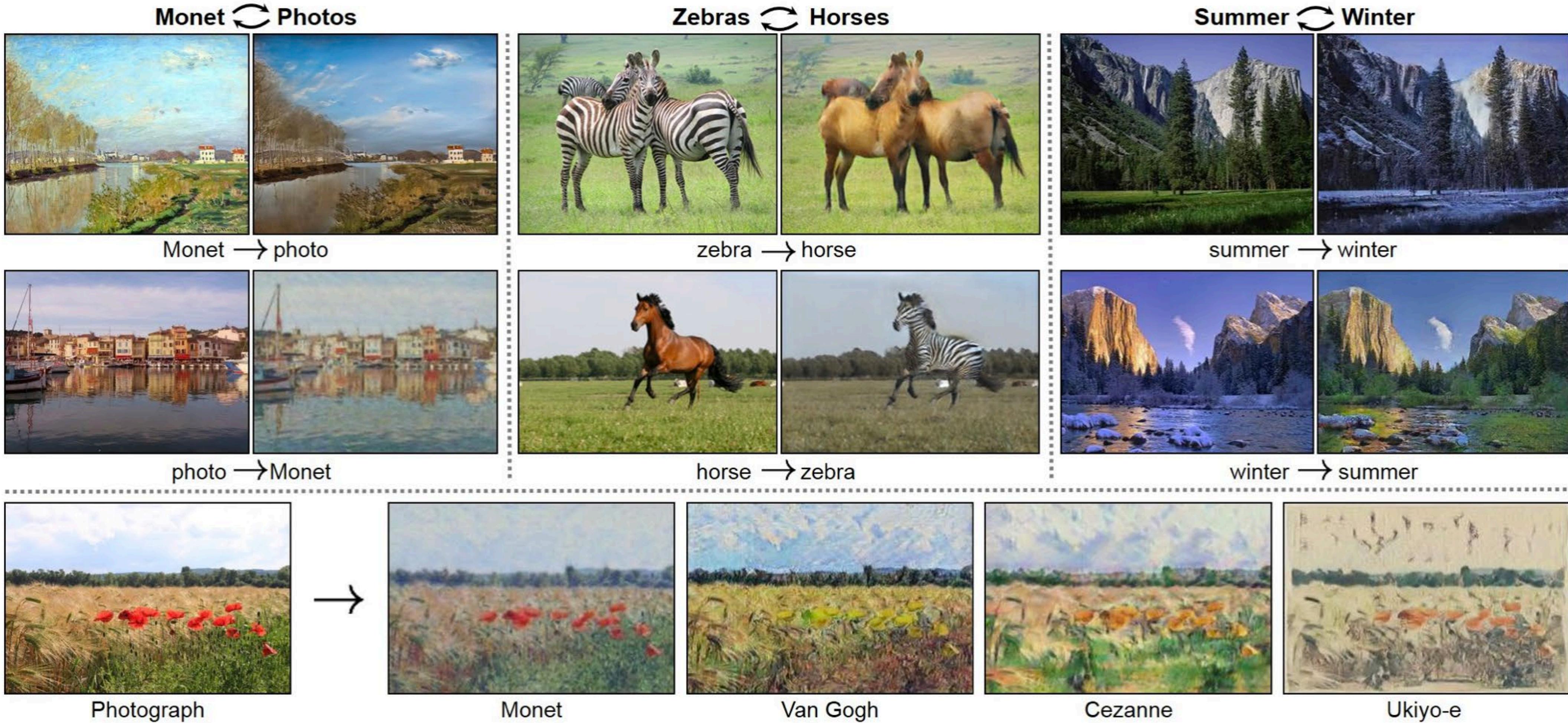


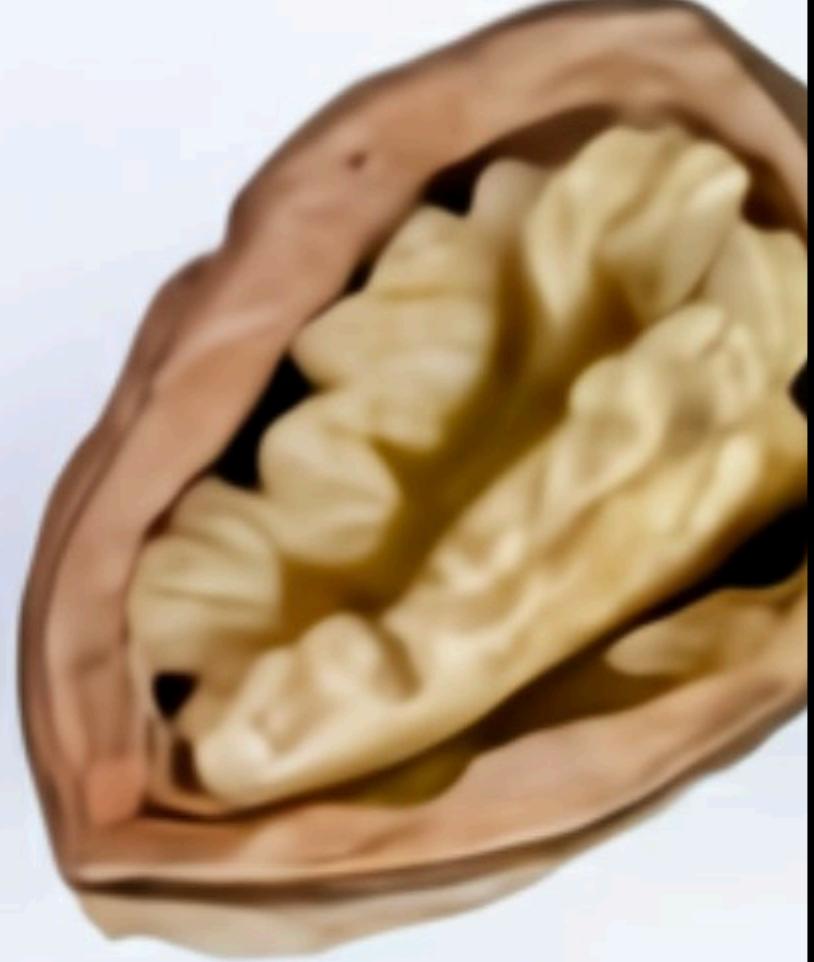
Image credit: Zhu et al., [Unpaired Image-to-Image Translation using Cycle-Consistent Adversarial Networks](#), 2017

# Examples of ML

## Research example - Text to image

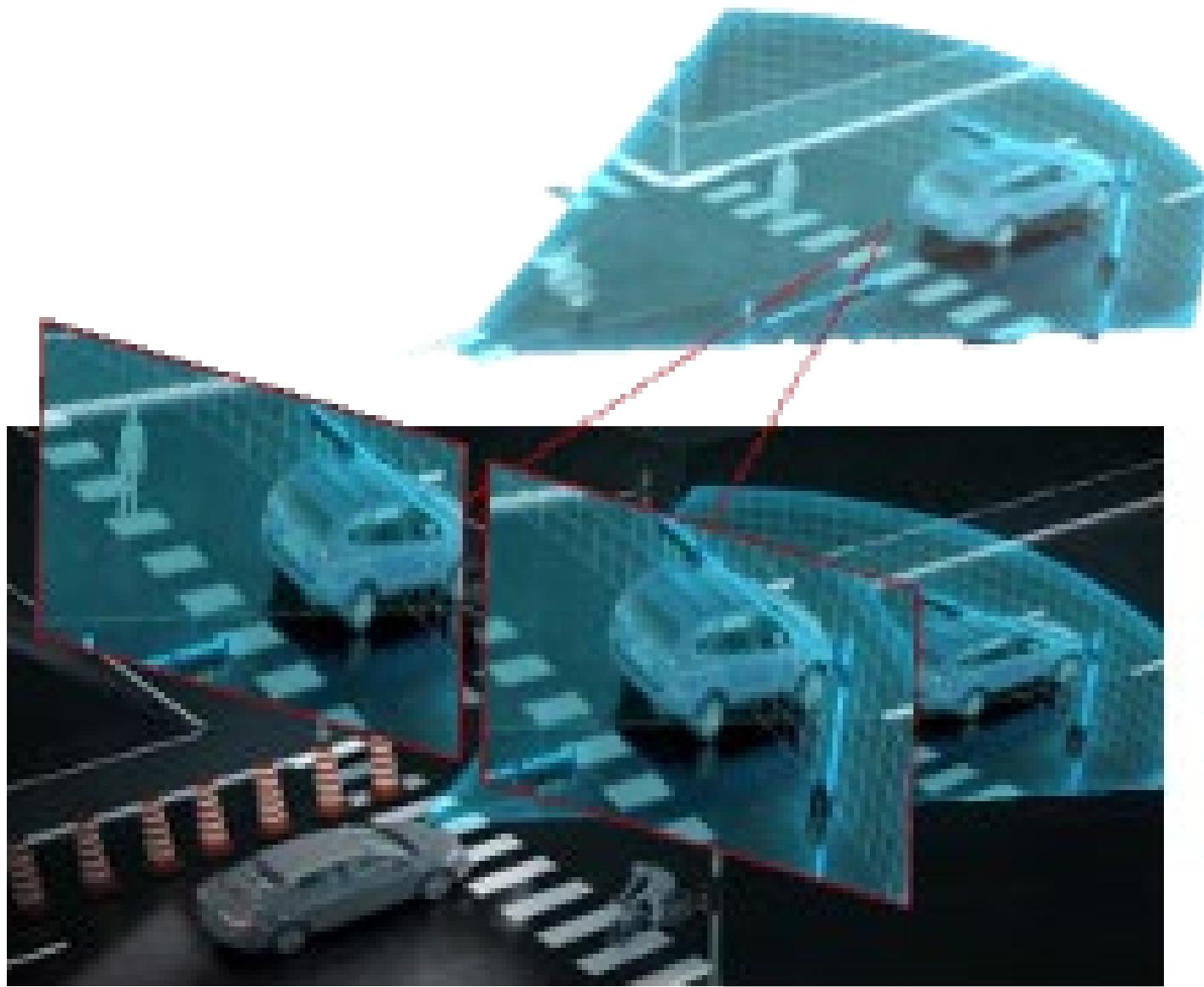
The following image were created from scratch based on the input text.

Try yourself at: <https://openai.com/blog/dall-e/>

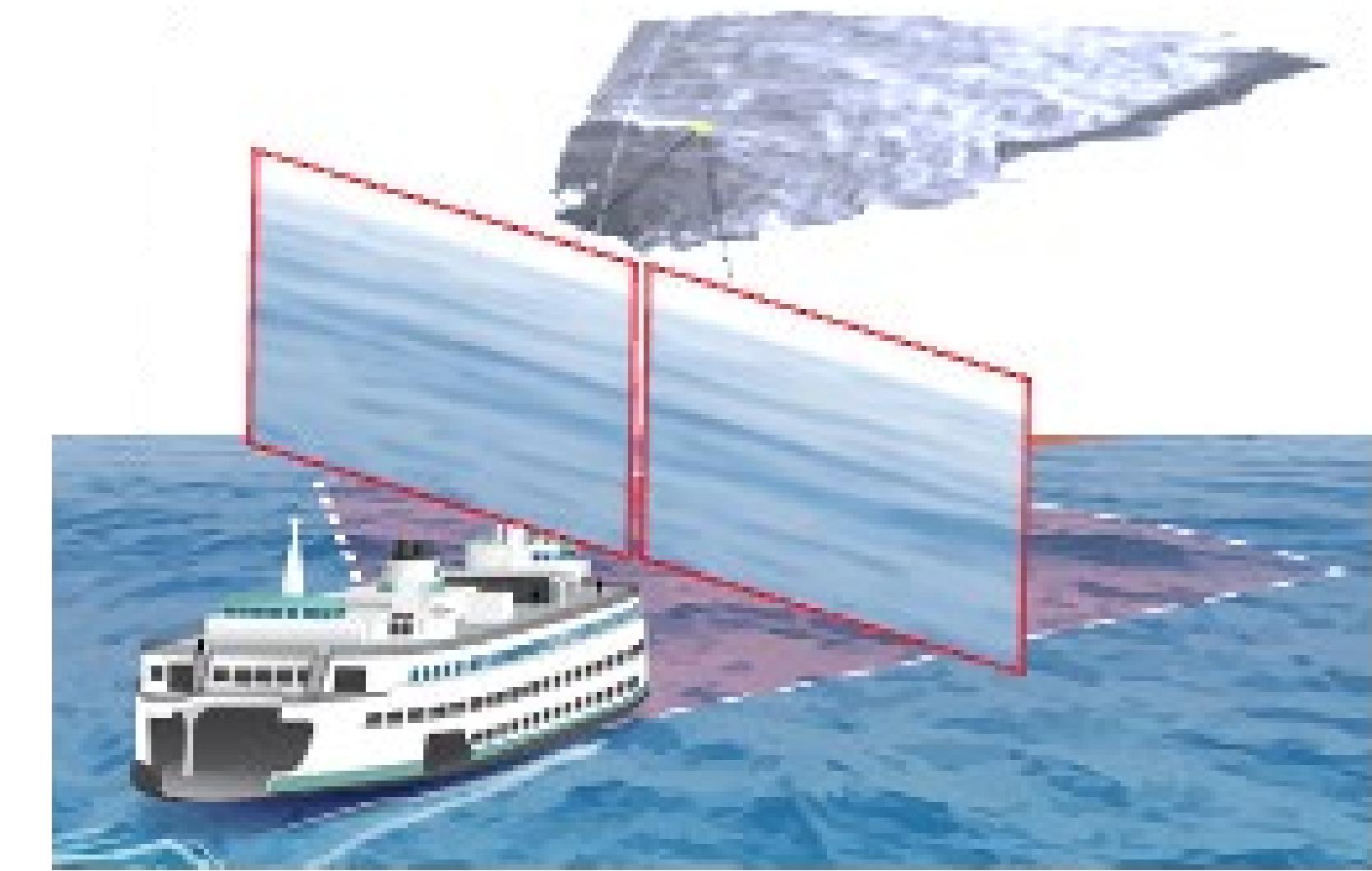
Input:	“An arm chair in the shape of an avocado”	“A collection of glasses is sitting on a table”	“A cross-section view of a walnut”
Output:			

# Within EPFL...

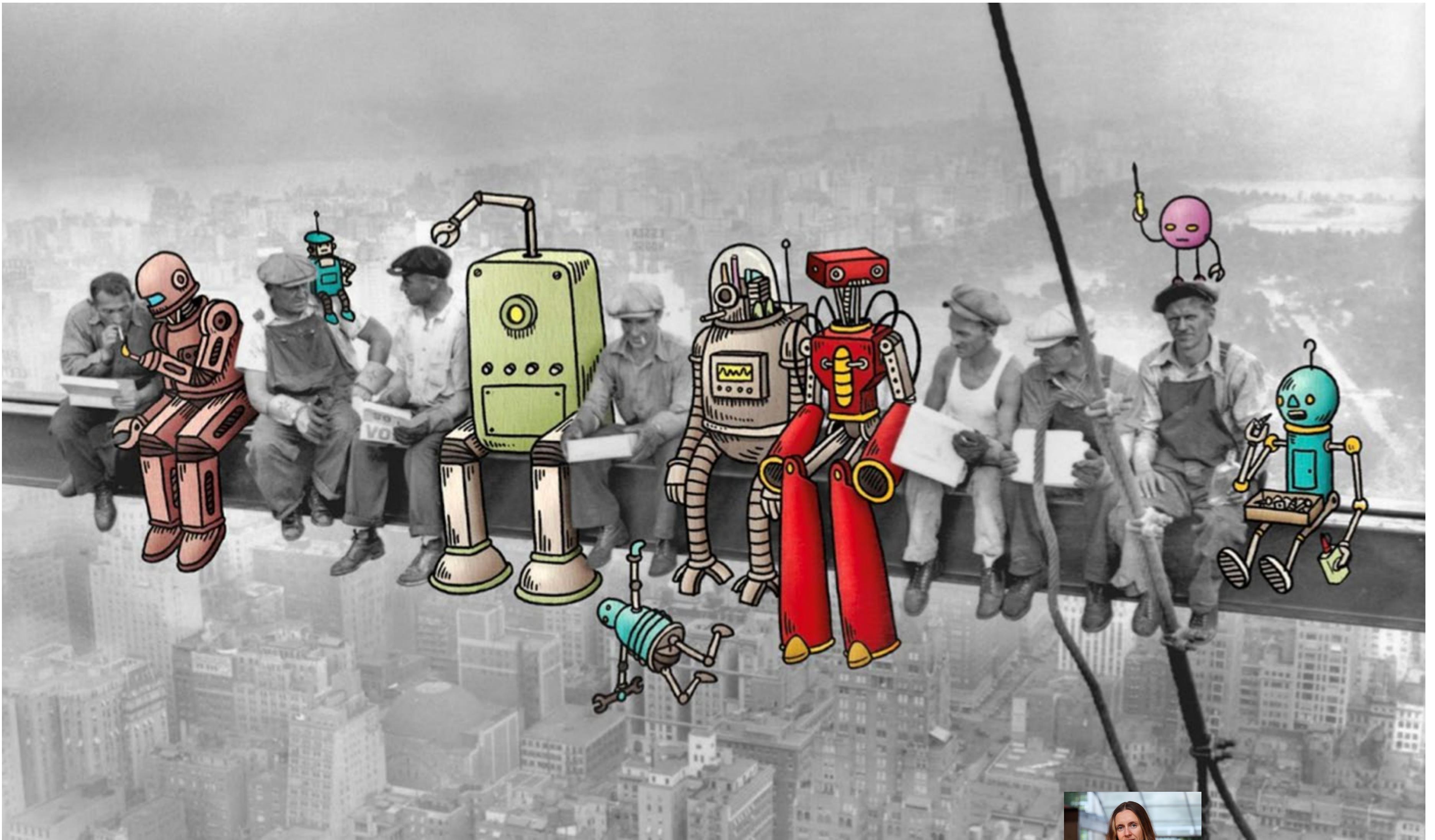
- 3D reconstruction  
of traffic participants



- 3D reconstruction  
of ocean waves



**EERL lab**



CRCL lab



ICE lab



**Luts lab**

# To recap

# ML for civil engineers

In the following weeks, you will learn how to

- Predict concrete strength
- Determine building damage following an earthquake
- Predict slope stability
- Detect steel defects
- Predict hot water usage
- Predict the number of vehicles on a bridge
- Predict traffic flows
- ...



# Typical steps to follow in a machine learning (ML) project (1/3)

## 1. Problem Definition and Understanding

- Clearly articulate the problem you aim to solve.
- Identify the objectives and success criteria.
- Ask the right questions: Identify the questions your data can answer to solve the business problem.
- Understand the data: Learn the context, limitations, and opportunities within the available data.

## 2. Identify the Value:

- Determine the potential value and impact of solving the problem (both from business and from the scientific perspective)
- Assess how the solution will benefit stakeholders and align with business goals.

## 3. Collect Data:

- Gather relevant data from various sources.
- Make sure that your data is representative for the test data (application data) → be aware of the variability of the operating conditions (→ domain shift)
- Ensure data quality and completeness.
- Acquire labels for your data if possible (ensure the quality of labels)

## 4. Explore and Understand the Data:

- Perform exploratory data analysis.
- Visualize data to understand patterns and relationships.
- Understand the data distribution (Explore the statistical properties, distributions, and trends in the data)
- Identify data quality issues (Look for missing values, outliers, and inconsistencies that may affect analysis).
- Generate hypotheses (Form hypotheses based on patterns observed during data exploration).

# Typical steps to follow in a machine learning (ML) project (2/3)

## 5. Prepare Data:

- Clean the data (handle missing values, outliers).
- Feature engineering (create meaningful features, transform existing ones).
- Split data into training, validation, and test sets.
- Scaling and normalization (Prepare the data for modeling by applying appropriate scaling techniques)
- Address class imbalances: If needed, use techniques like oversampling, undersampling, or synthetic data generation to balance the dataset.

## 6. Select and Train Models:

- Select appropriate models (Choose models based on the problem type (regression, classification, clustering, etc.) and data characteristics)
- Split the dataset (Use train-test splits (or cross-validation) to ensure your model's generalization to unseen data)
- Baseline model (Start with a simple model as a baseline to compare more complex models).
- Tune Hyperparameters (Optimize model hyperparameters for better performance).
- Iterate and improve (Experiment with different models and tuning hyperparameters)
- Avoid overfitting (Implement regularization techniques or apply cross-validation to avoid overfitting the training data).

## 7. Evaluate Models:

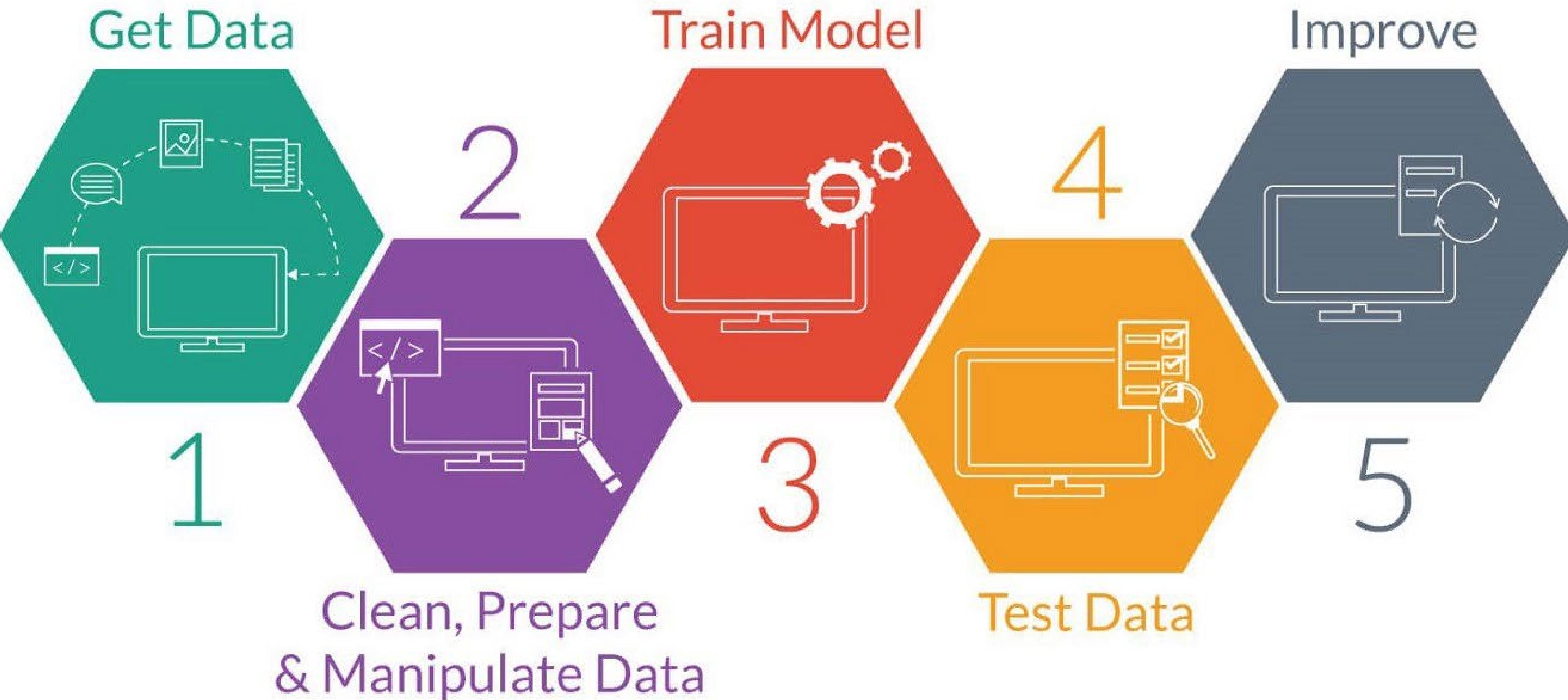
- Choose appropriate metrics (Select evaluation metrics that are relevant to the problem (e.g., accuracy, precision, recall, F1-score, RMSE))
- Validate on unseen data (Always validate your model on a test set or through cross-validation to assess its performance)
- Compare models: Compare different models based on performance metrics and business value.

# Typical steps to follow in a machine learning (ML) project (3/3)

8. Model Interpretation
  - What do the results mean?
  - Explainability (Ensure that you understand how your model works, and use XAI techniques such as SHAP values or feature importance to explain the results).
  - Check assumptions (Ensure that the model's assumptions hold and are consistent with the problem and data context)
9. Test Model:
  - Evaluate the final model on the test dataset to gauge its real-world performance.
10. Deploy Model:
  - Integrate the model into a production environment.
11. Monitor and Maintain:
  - Continuously monitor model performance.
  - Update the model as needed based on new data and changing conditions.
  - Monitor how your model is actually used
12. Documentation and Communication
  - Document thoroughly (Keep detailed notes on every step, including data sources, preprocessing steps, model choices, evaluation, and deployment procedures)
  - Communicate results (Present findings to stakeholders in a clear, actionable manner. Tailor communication based on the audience (e.g., technical team vs. business executives).)

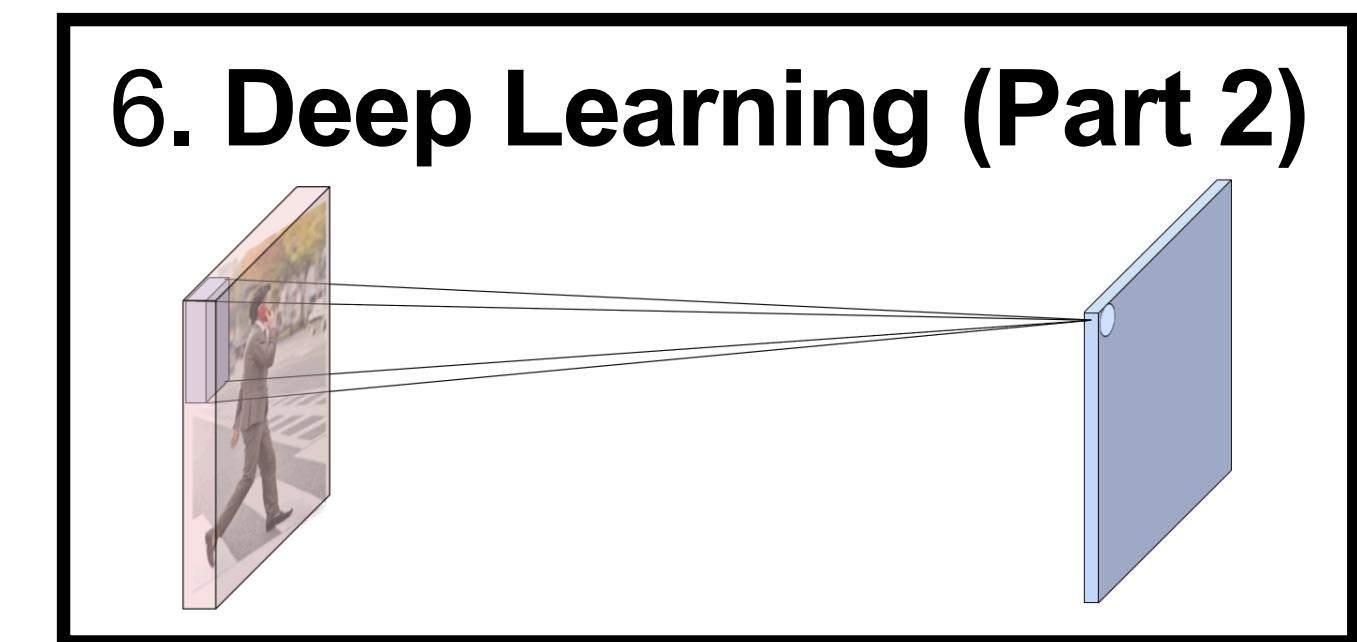
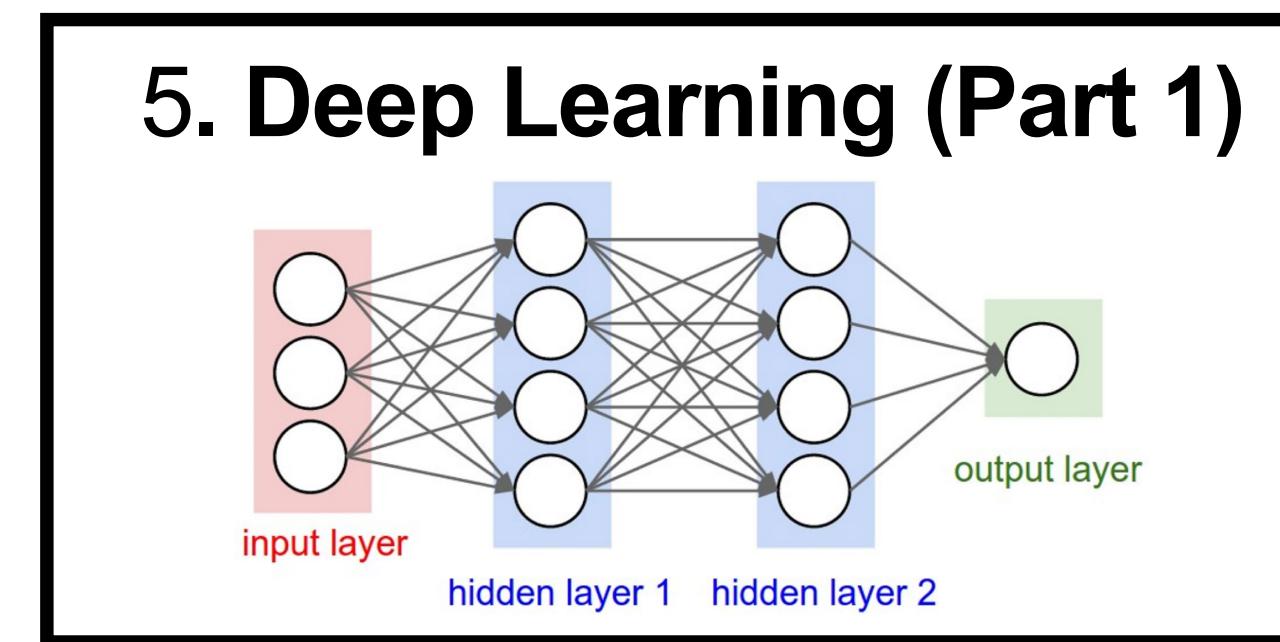
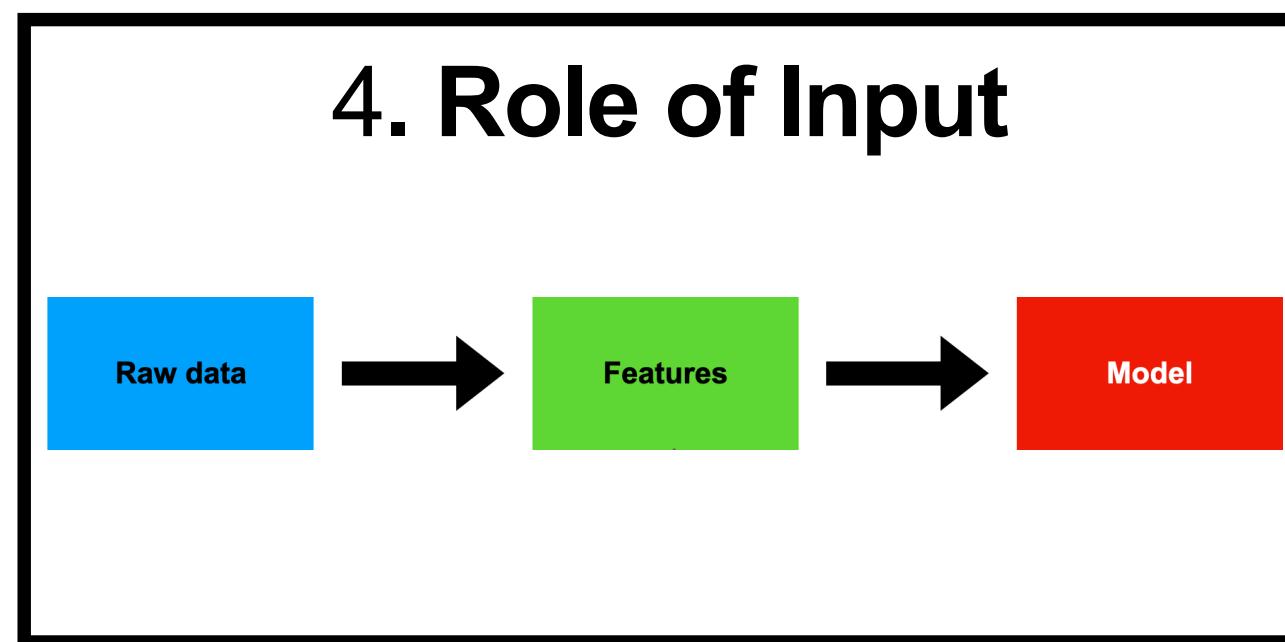
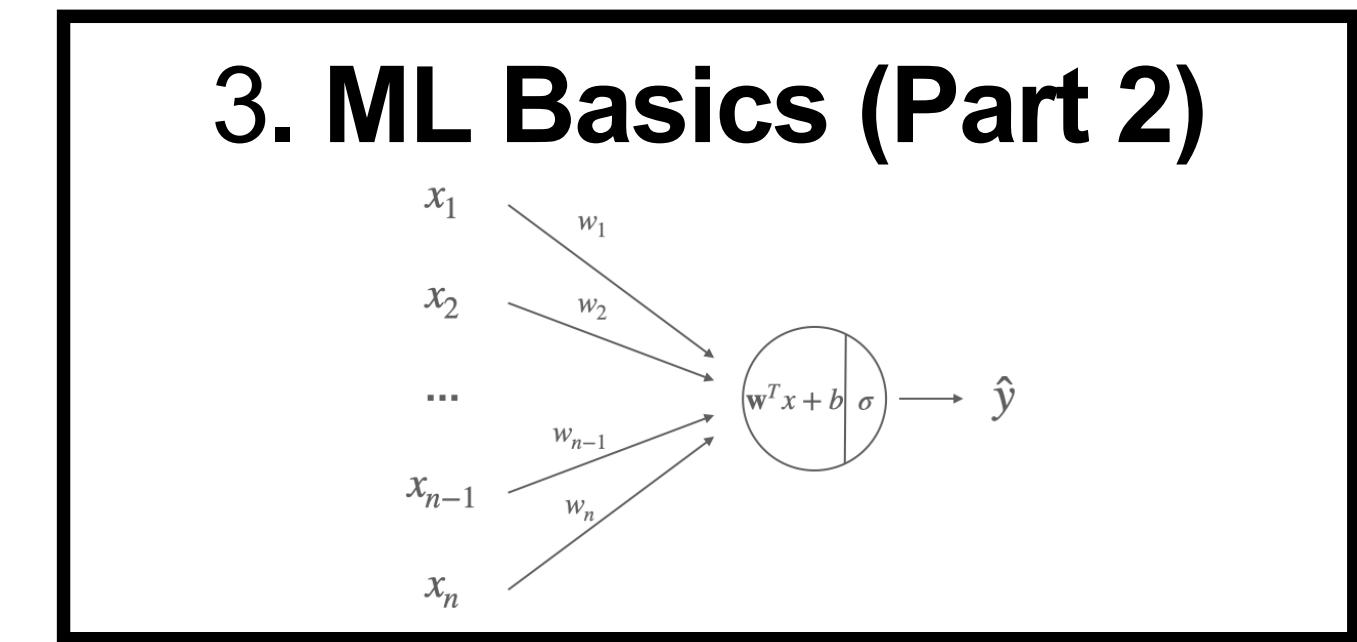
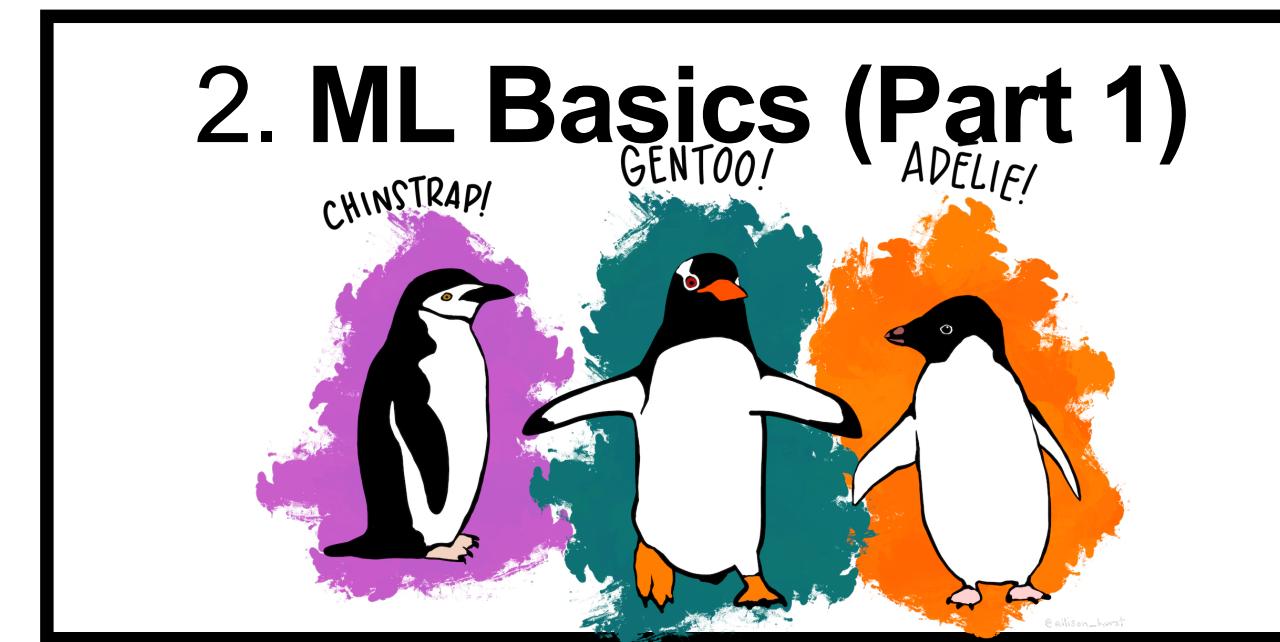
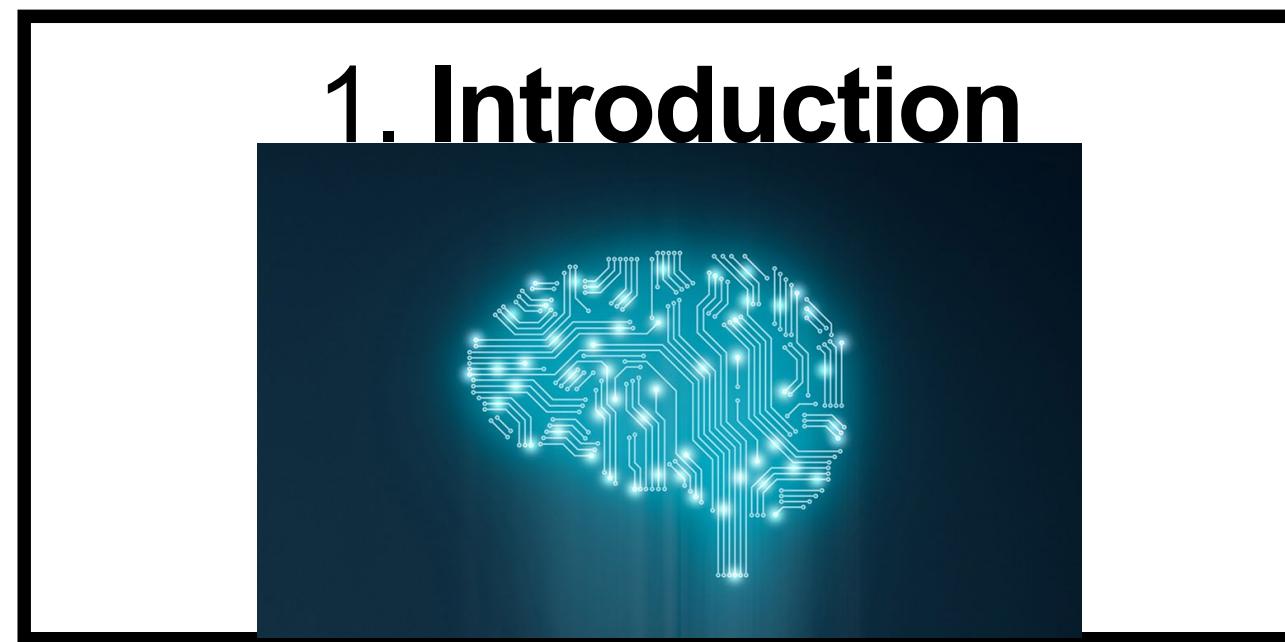
# What is ML?

## The process



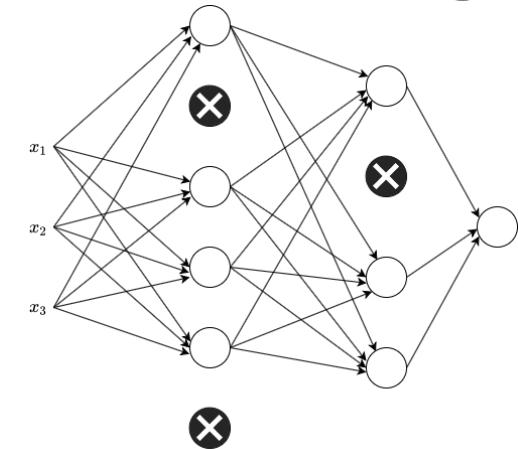
# Road map

## Lectures

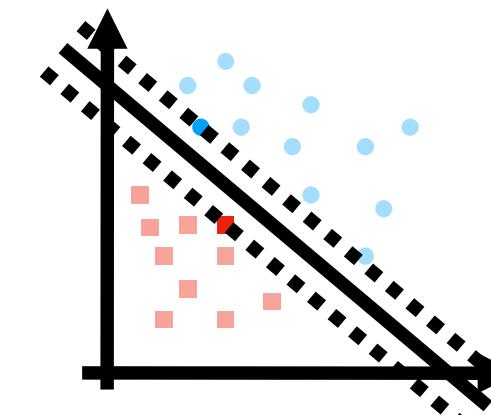


# Road map Lectures

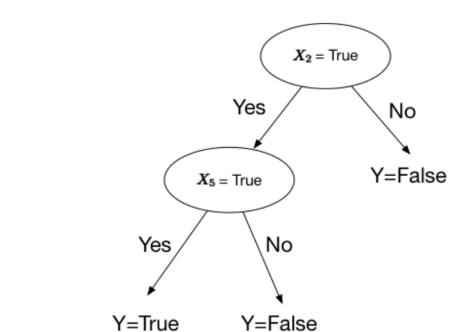
## 7. Deep Learning (Part 3)



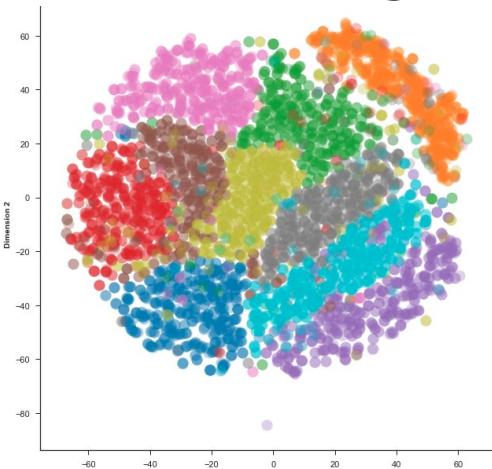
## 8. Support Vector Machine



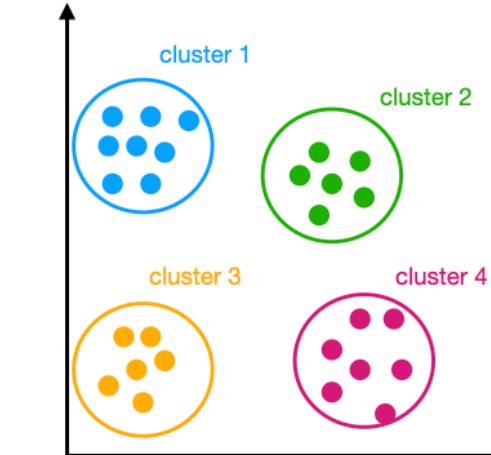
## 9. Other Supervised Learning Models



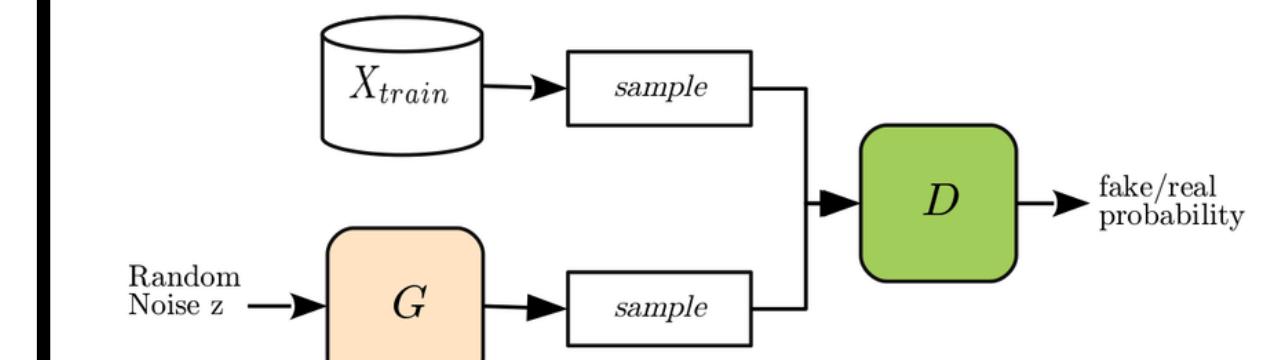
## 10. Dimensionality Reduction



## 11. Clustering



## 12. Other Learning Examples



# At the end of the course

You will:

- Know the most common ML techniques
- Know how to implement ML techniques in Python using popular data science libraries
- Have all the tools needed to start a ML project on your own
- Be able to read ML literature

## Papers on ML in Civil Engineering:

- Chang, Lin, Chang, [Applications of neural network models for structural health monitoring based on derived modal properties](#), 2018
- Dong, Chen, Guan, [Cost Index Predictions for Construction Engineering Based on LSTM Neural Networks](#), 2020
- Huang, Huang, Lyu, [An Improved KNN-Based Slope Stability Prediction Model](#), 2020
- Silva, Moita, Arruda, [Machine learning techniques to predict the compressive strength of concrete](#), 2020
- Feng et al., [Machine learning-based compressive strength prediction for concrete: An adaptive boosting approach](#), 2020
- Li, Zhao, [Image-Based Concrete Crack Detection Using Convolutional Neural Network and Exhaustive Search Technique](#), 2019
- Gao, Kong, Mosalam, [Deep leaf-bootstrapping generative adversarial network for structural image data augmentation](#), 2019
- ...

# Resources

## To go further

- « What is Artificial Intelligence Exactly? » by Dagogo Altraide:  
<https://www.youtube.com/watch?v=kWmX3pd1f10&list=>