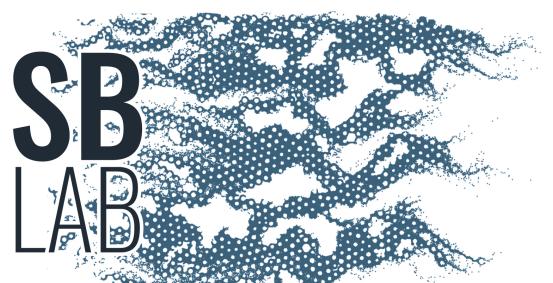

Introduction to machine learning in Hydrology

Lazaro J. Perez & Marc Berghouse

contact: lazaro.perez@dri.edu



Outline

- Introduction
- What is machine learning?
- Matlab

Introduction

♦ Ice-breaker

Name

Research topic

What you hope to get out of the class

Class structure

♦ Location

Computer lab in the CRVB @ DRI Campus

♦ Plan

Tuesday: Lecture/Theory/Principles/Research examples

Thursday: Coding/Work on example

♦ Contact

Lazaro - lazaro.perez@dri.edu - Response time less than a day

Office hours - Monday 10-14

Marc -marc.berghouse@dri.edu - ?

Class structure

♦ Tools

Python (Jupiter notebooks)

Install Anaconda. If you need help work with Marc

Use Colab or Kaggle as workaround

Matlab (Regression Learner App // Code)

♦ Grading

You will grade yourself

Homework - Five models applied to specific cases - (60 %)

Project - Develop a predictive model to specific data - (35 %)

Project presentations start May 4th

Participation - 5%

Homework

♦ Objectives

Assignments designed to reinforce the methods shown in class

Develop a workflow useful for your research

Learn by failure: Reproduce wrong predictions and detect error

Investigate the performance of different models for single/multiple tasks

Plan

Date	Topic	Assignment
01/24	Intro & ML Basics (Laz)	
01/31	Applications of ML in Hydrology (Laz)	
02/02	Matlab & Random Forests (Laz)	H1
02/09	Python (Marc)	H2
02/14	K-Nearest Neighbor (Marc)	
02/21	Stochastic Gradient Descent (Marc)	H3
02/28	Support Vector Machine (Laz)	



ML Basics

- ◆ What is machine learning?

Humans learn from experience



Machines follow orders

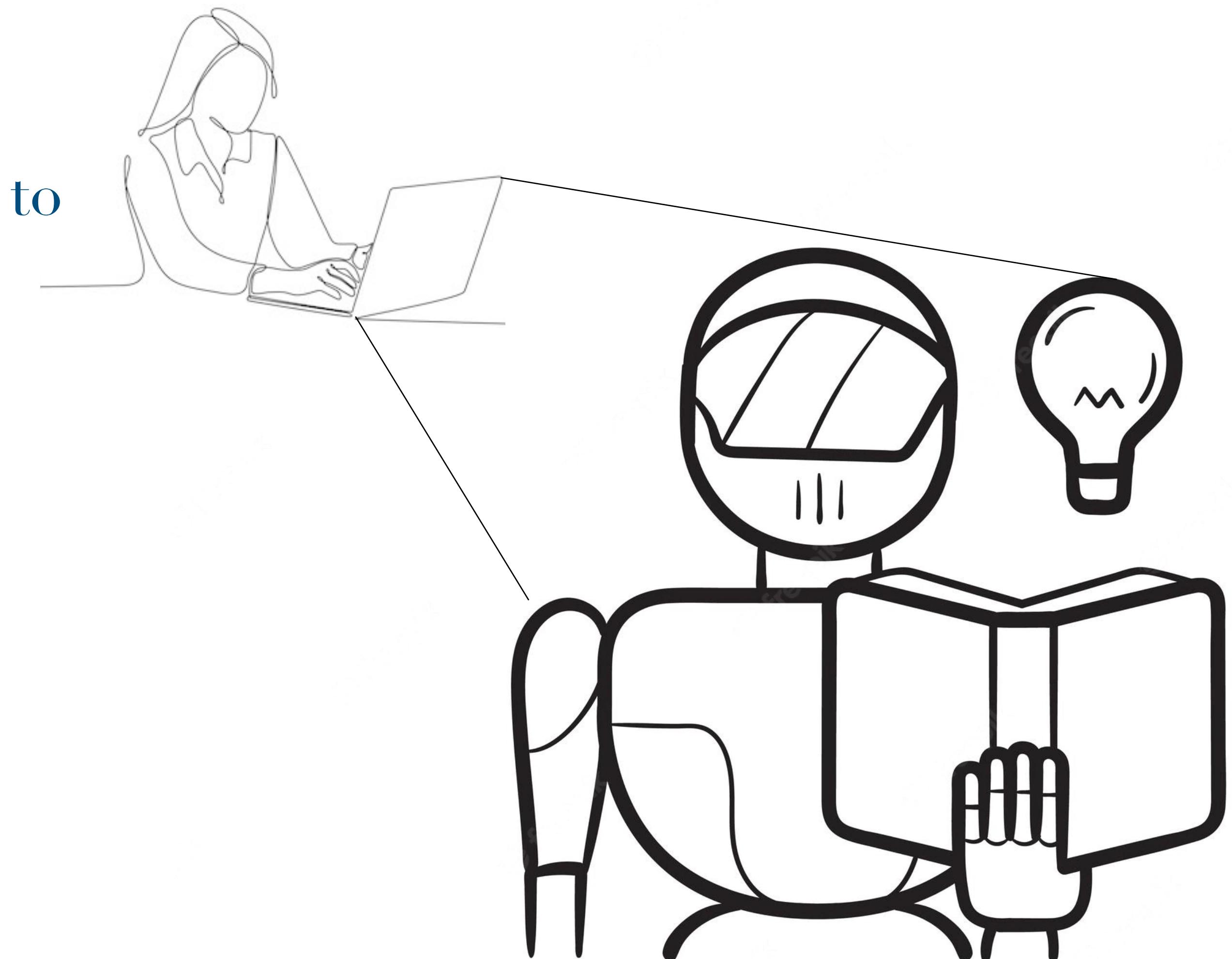


ML Basics

- ♦ What is machine learning?

What if humans can train machines to learn from past... data

Machine Learning!



ML Basics

♦ What is machine learning?

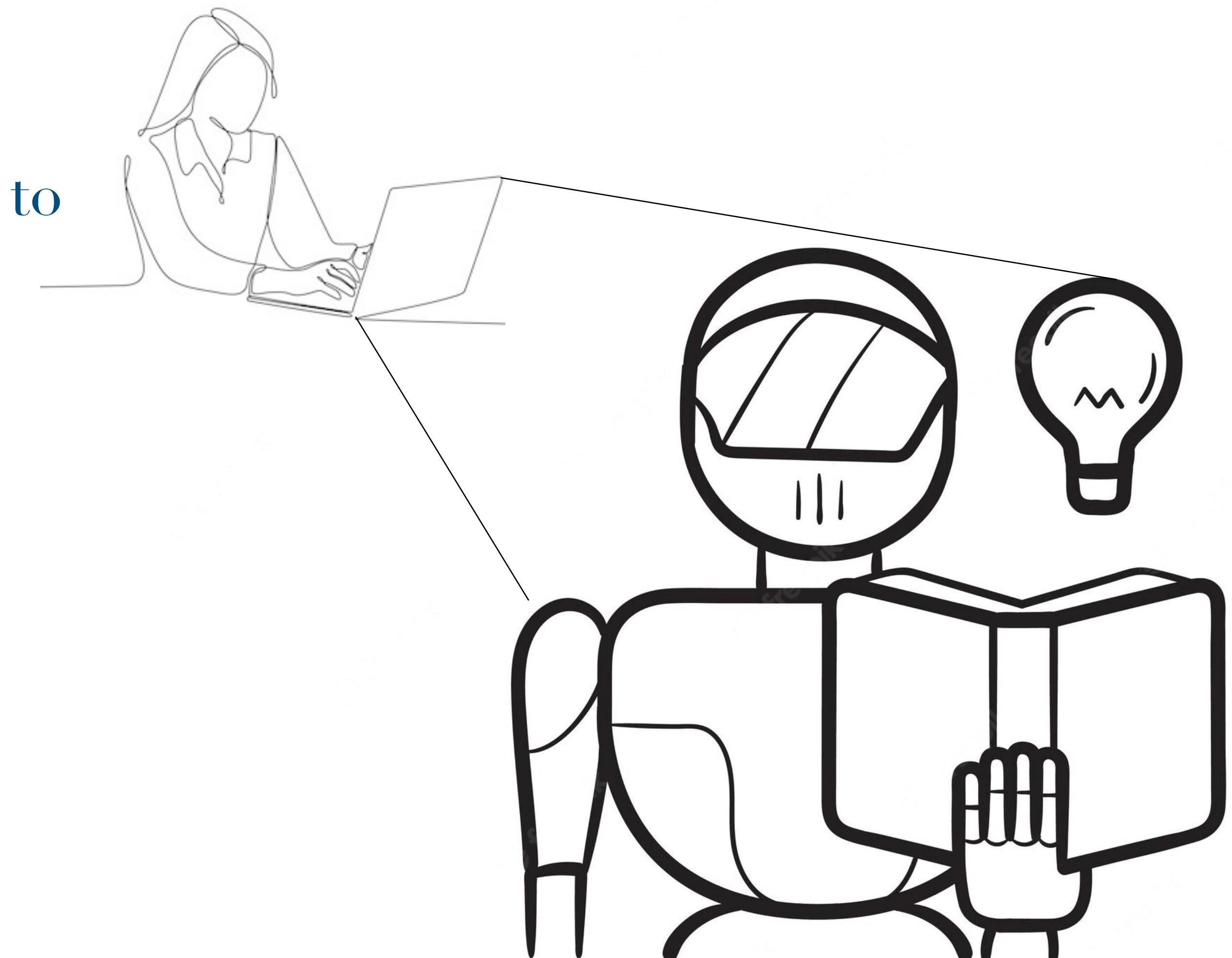
What if humans can train machines to learn from past... data

Machine Learning!

But it is not as simple as learning...

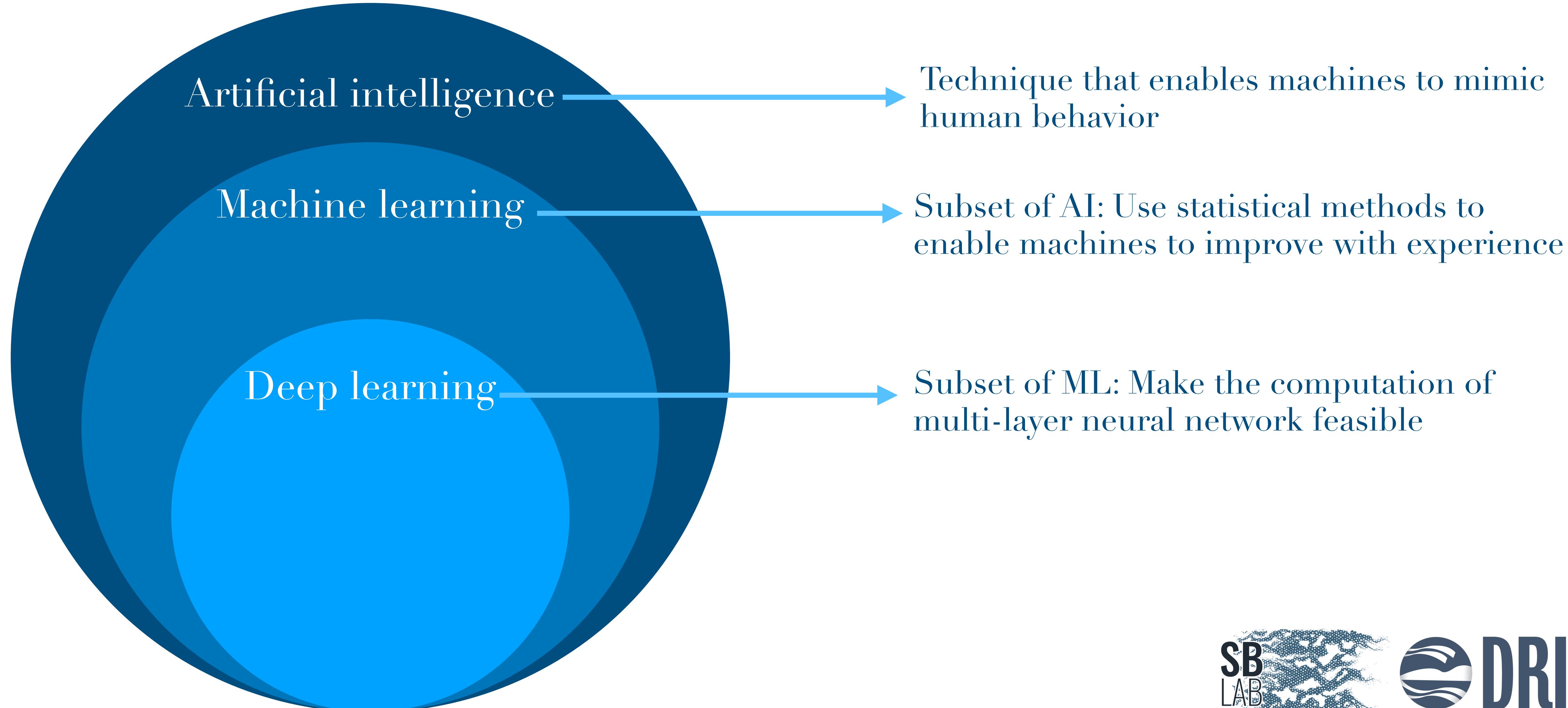
Understanding

Reasoning



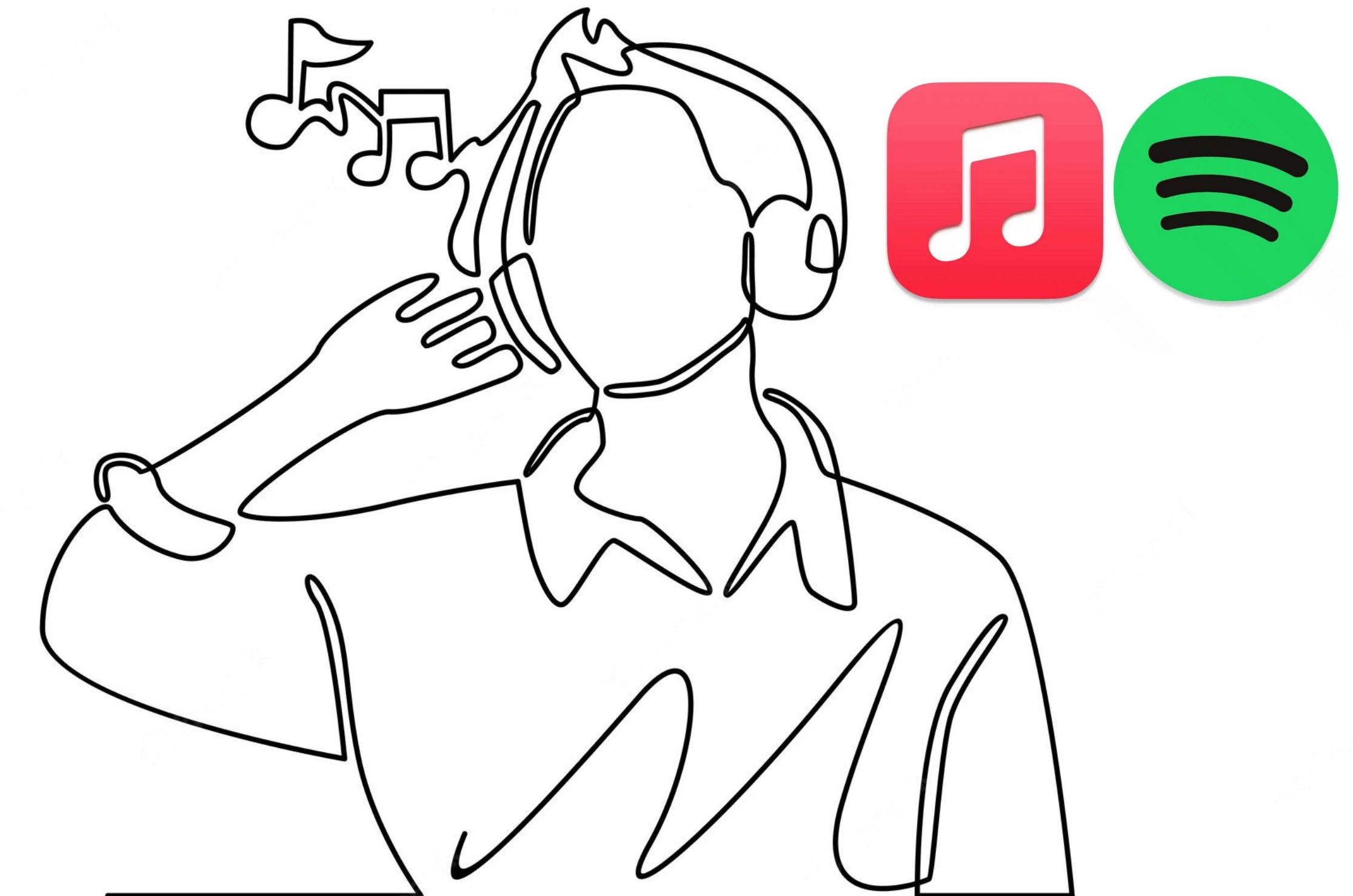
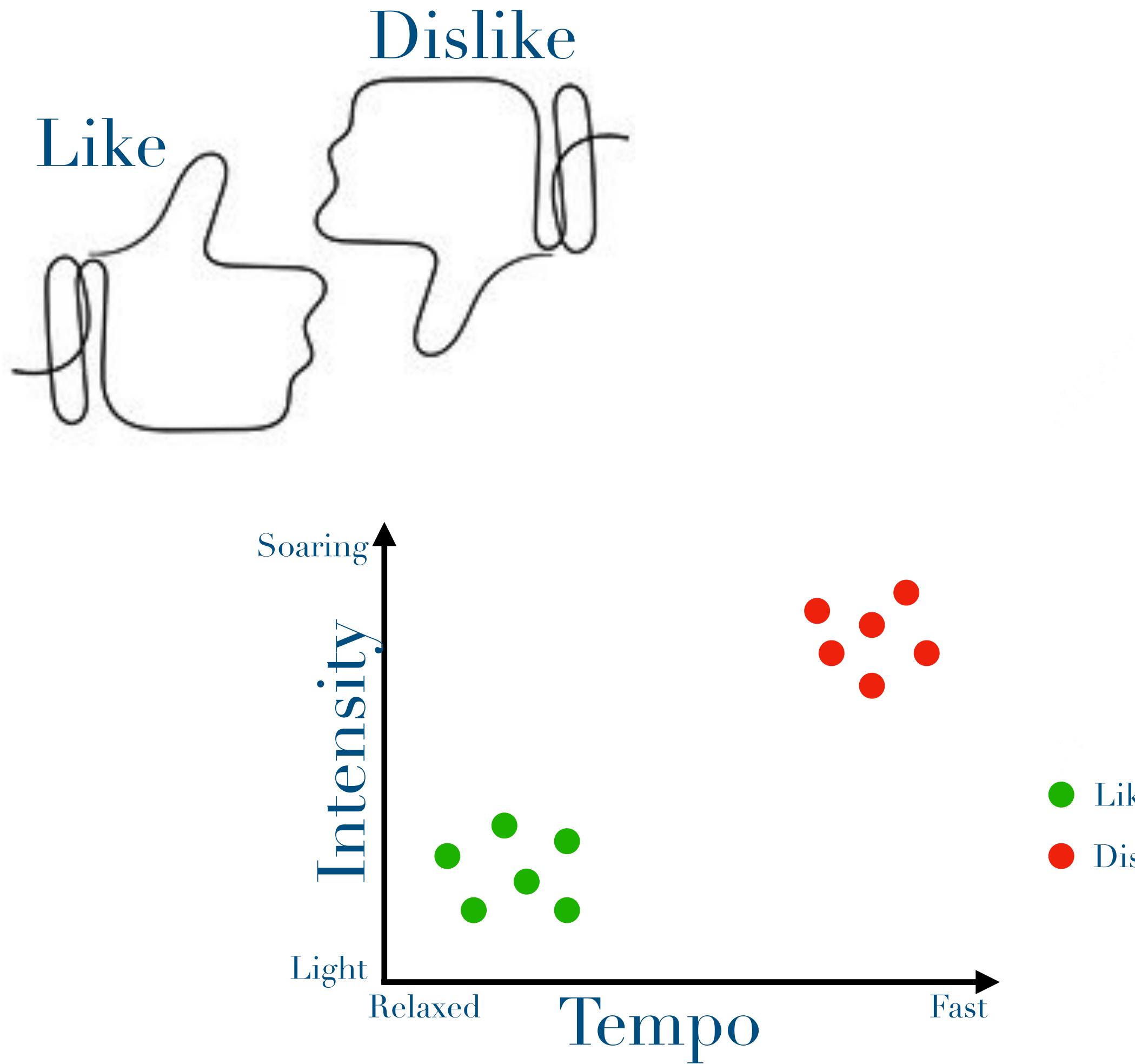
ML Basics

♦ What is machine learning?



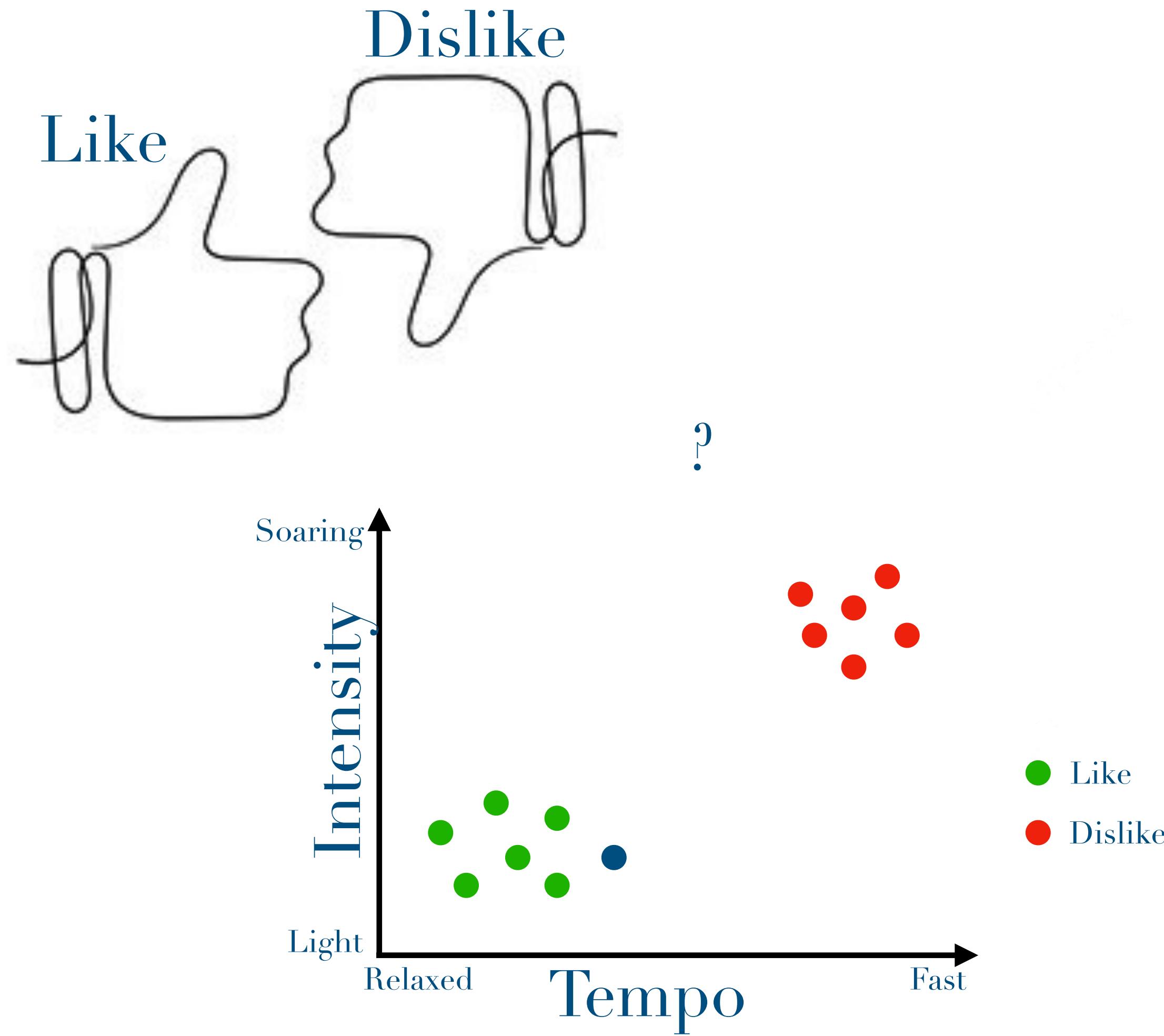
ML Basics: Examples in daily life

Classification



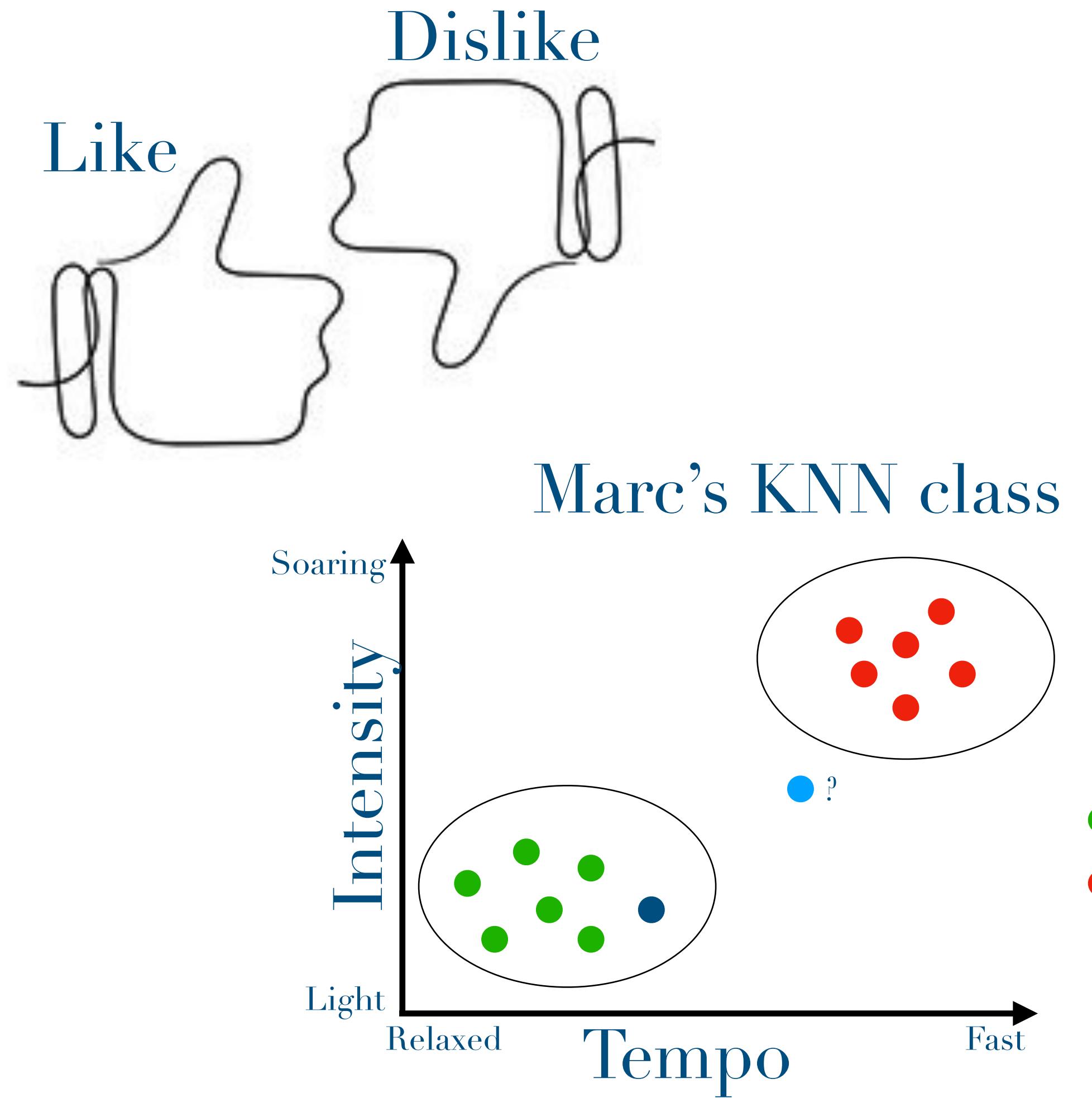
ML Basics: Examples in daily life

Classification



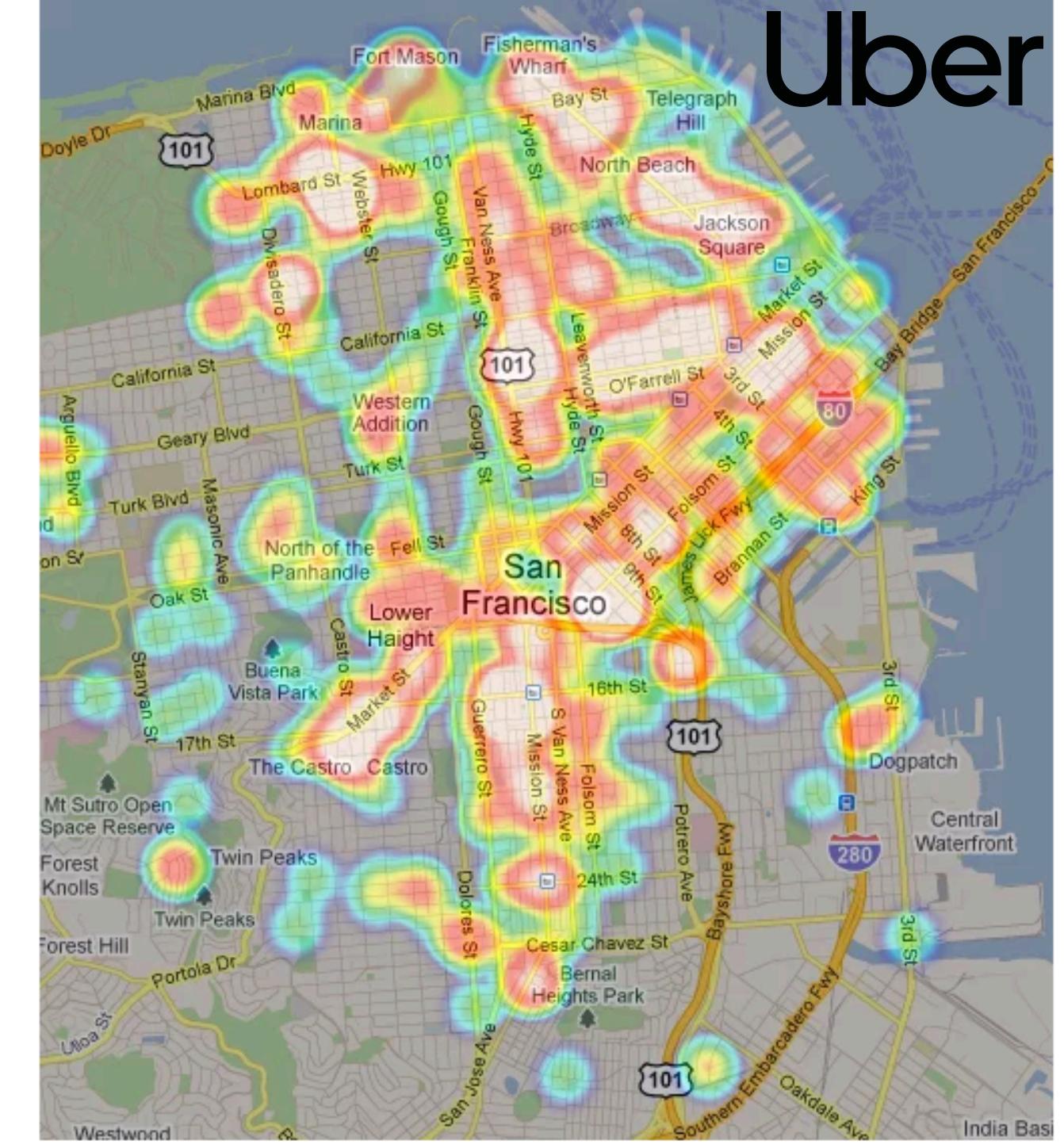
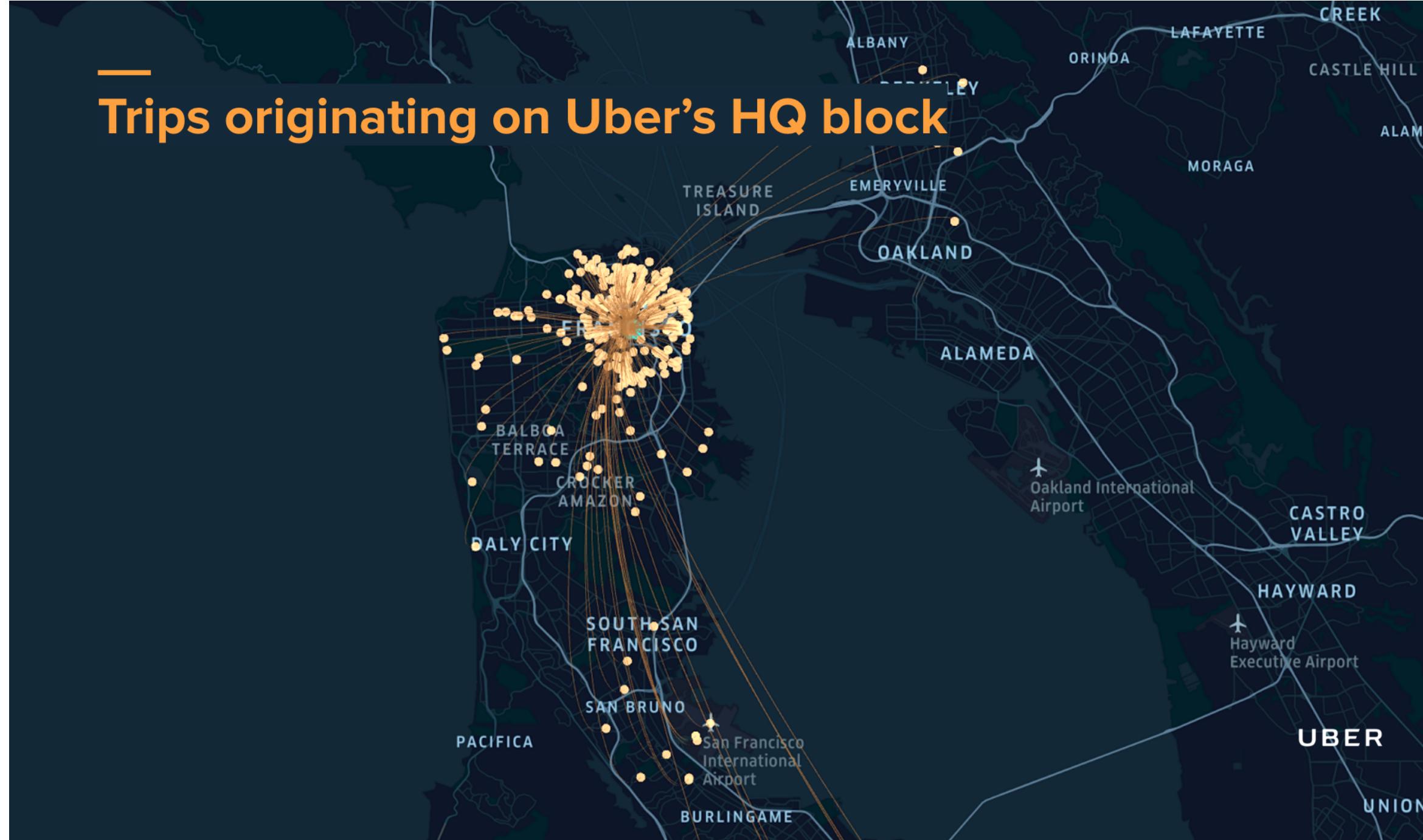
ML Basics: Examples in daily life

Classification



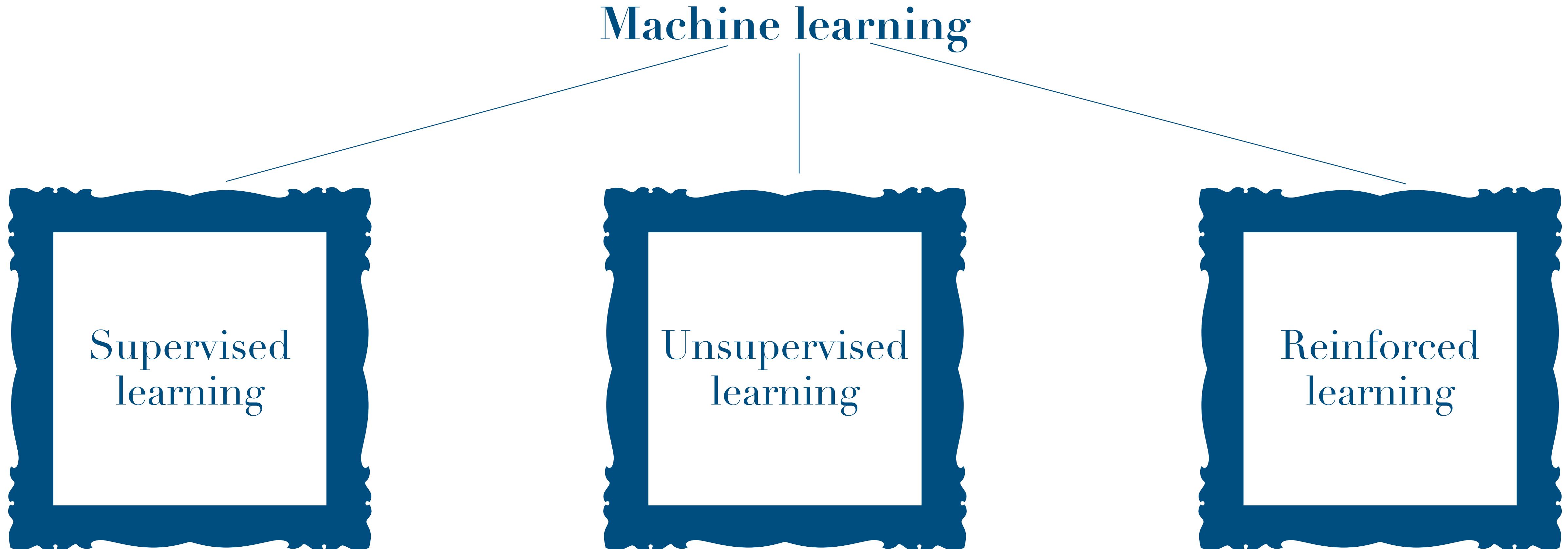
ML Basics: Examples in daily life

Regression



tpep_pickup_datetime	tpep_dropoff_datetime	passenger_count	trip_distance	pickup_longitude	pickup_latitude	RateCodeID	store_and_fwd_flag	dropoff_longitude	dropoff_latitude	payment_type	fare_amount	extra	mta_tax	tip_amount	tolls_amount	improvement_surcharge	total_amount
1/15/15 14:00	1/15/15 14:13	1	3	-73.964272	40.7730217	1	N	-73.965775	40.8046684	2	12	0	0.5	0	0	0.3	12.8
1/15/15 14:00	1/15/15 14:05	1	0.67	-73.970932	40.7959213	1	N	-73.970169	40.7891235	1	5	0	0.5	1	0	0.3	6.8
1/7/15 14:58	1/7/15 15:06	1	0.98	-73.948692	40.7777824	1	N	-73.955284	40.786869	1	7	0	0.5	1.4	0	0.3	9.2
1/7/15 14:58	1/7/15 15:12	3	4.39	-73.988663	40.7227058	1	N	-73.987221	40.6944084	2	15.5	0	0.5	0	0	0.3	16.3

ML Basics: How do machines learn?



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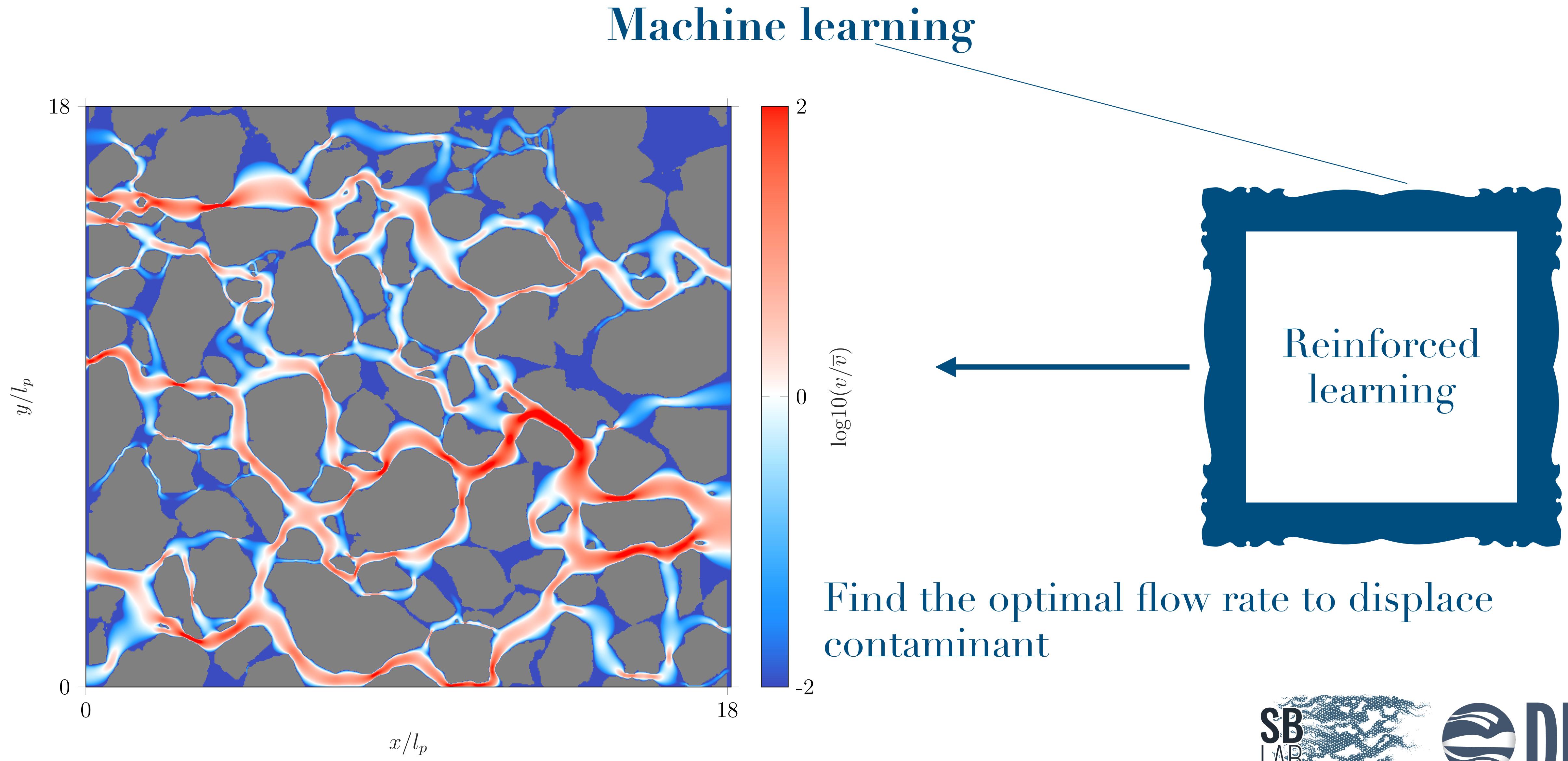


Machine learning



Reinforced
learning

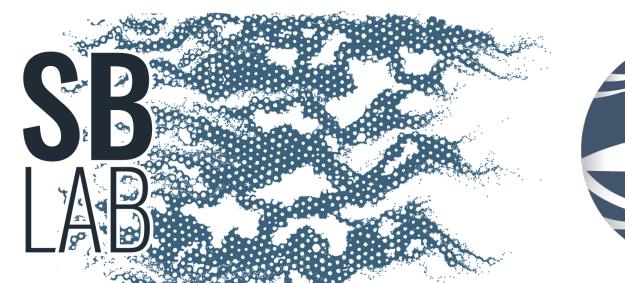
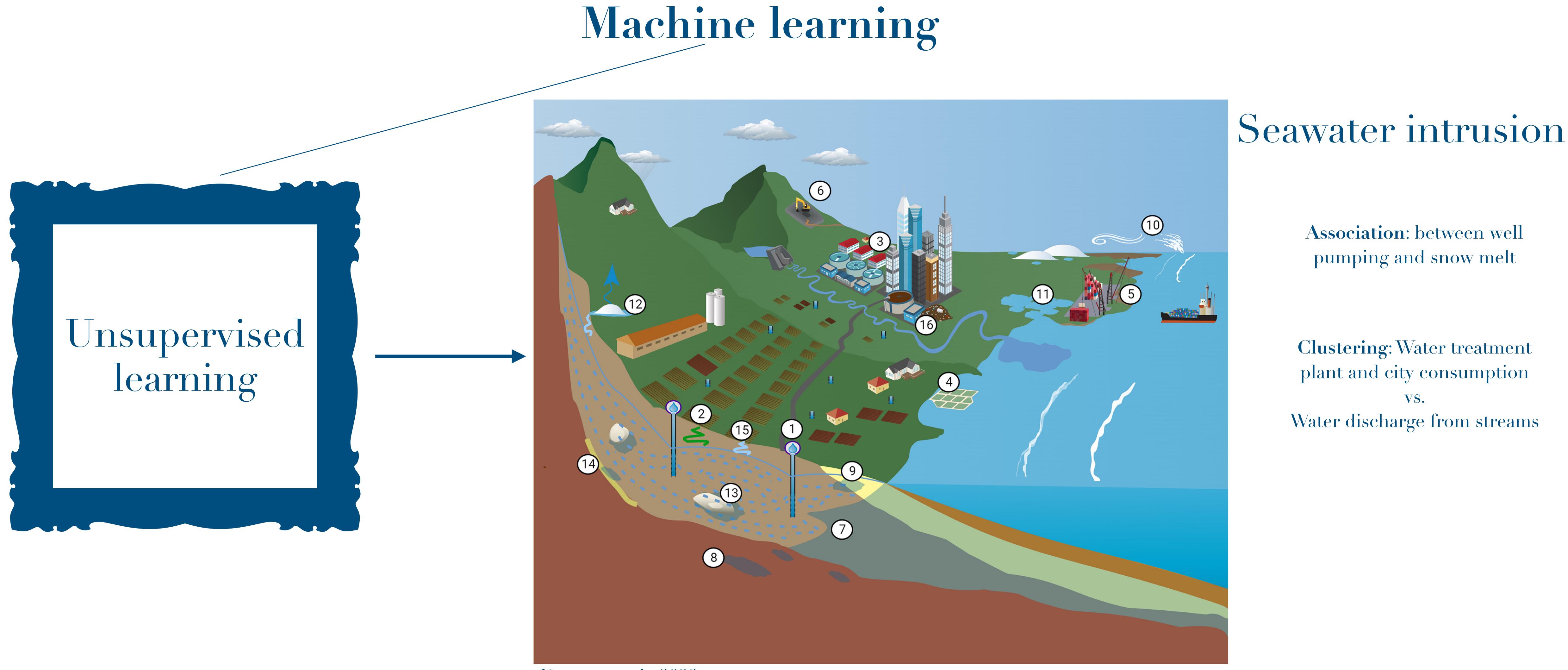
ML Basics: How do machines learn?



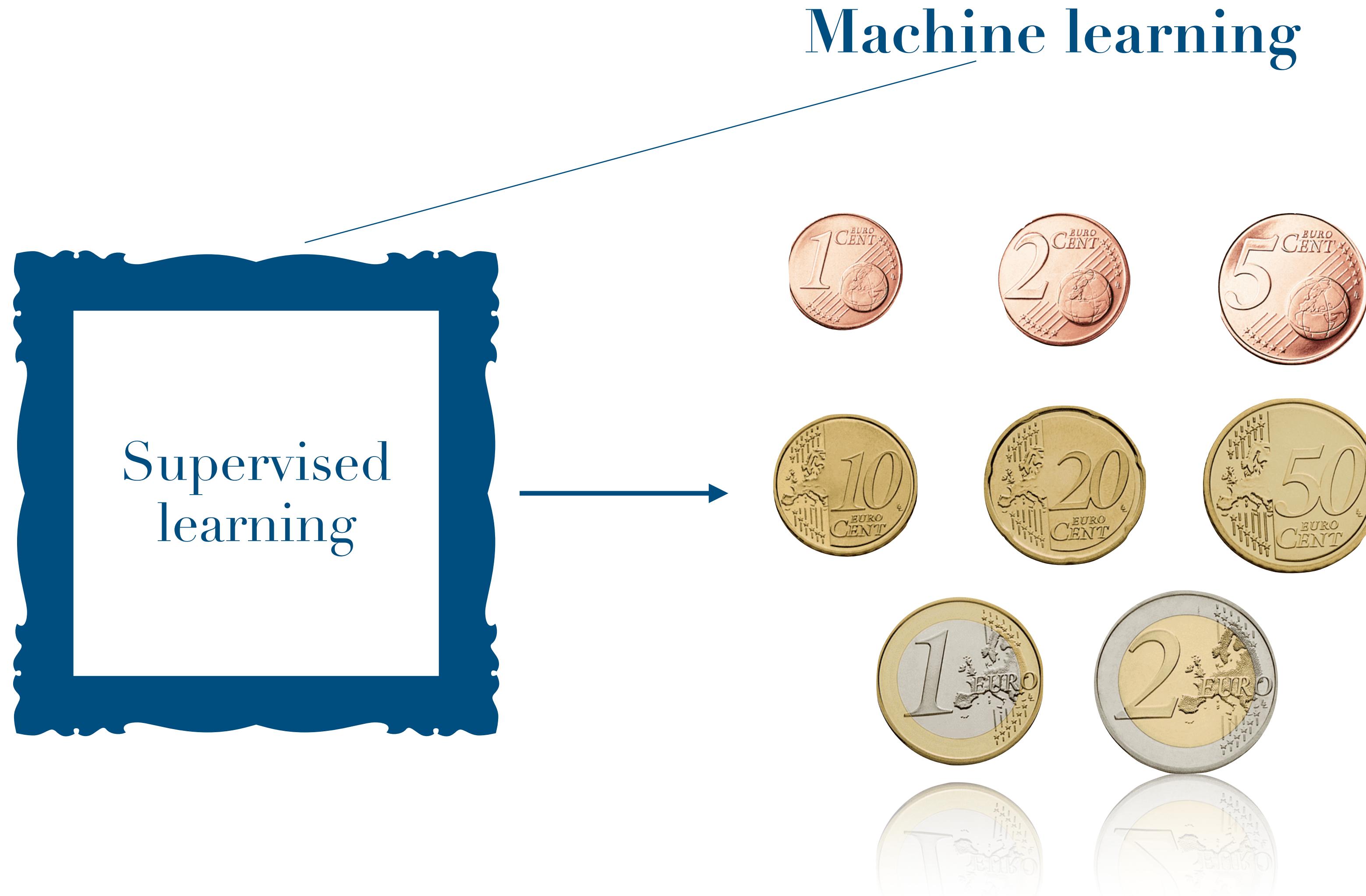
ML Basics: How do machines learn?



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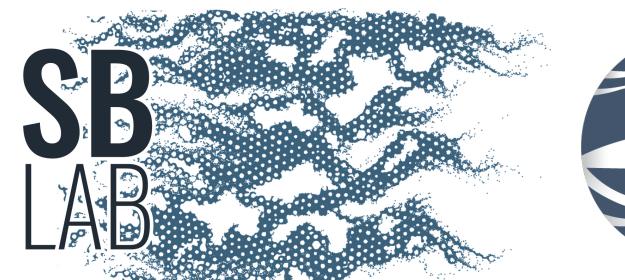
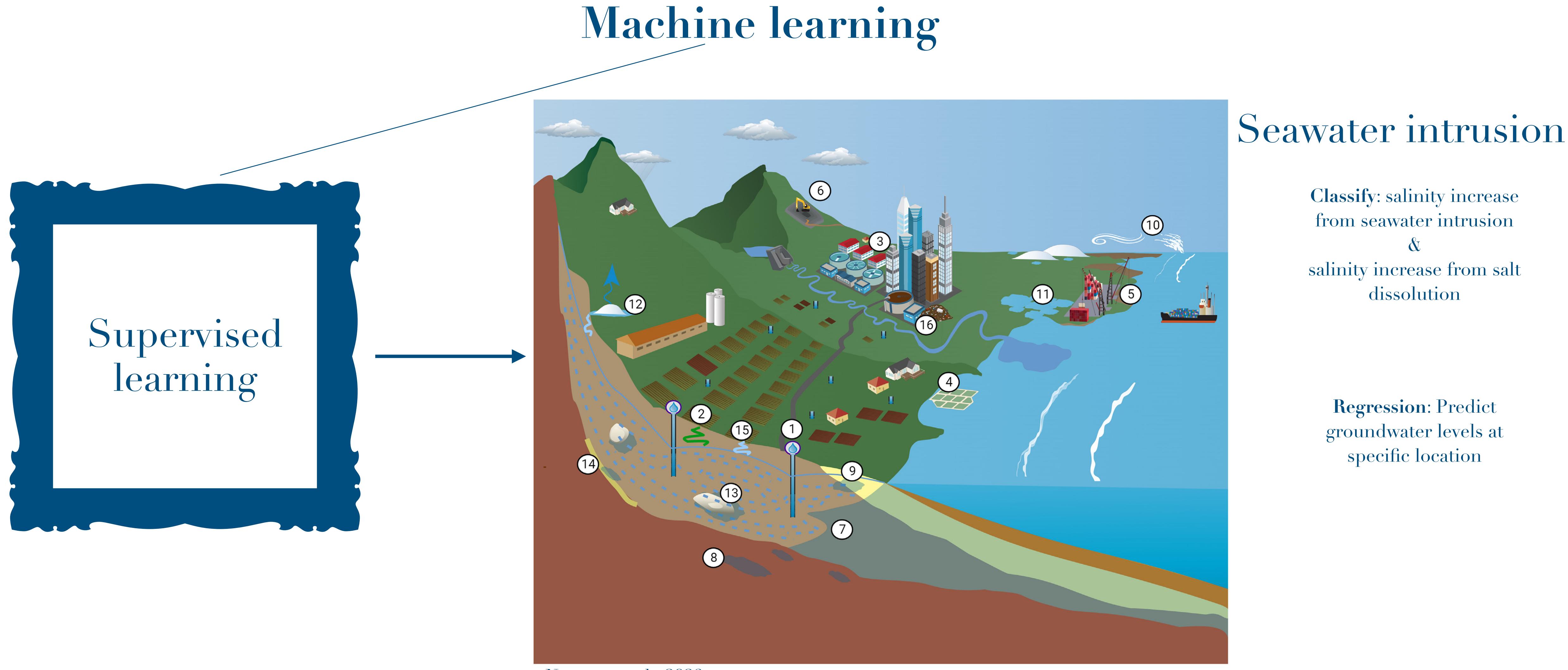


ML Basics: How do machines learn?



Value	Weight (g)	Diameter (mm)
0.01	2.3	16.25
0.02	3.05	18.75
0.05	3.9	21.25
0.1	4.1	19.75
0.2	5.75	22.25
0.5	7.8	24.25
1	7.5	23.25
2	8.5	25.75

ML Basics: How do machines learn?



Quiz: Supervised vs. Unsupervised vs. Reinforced?

Daily life

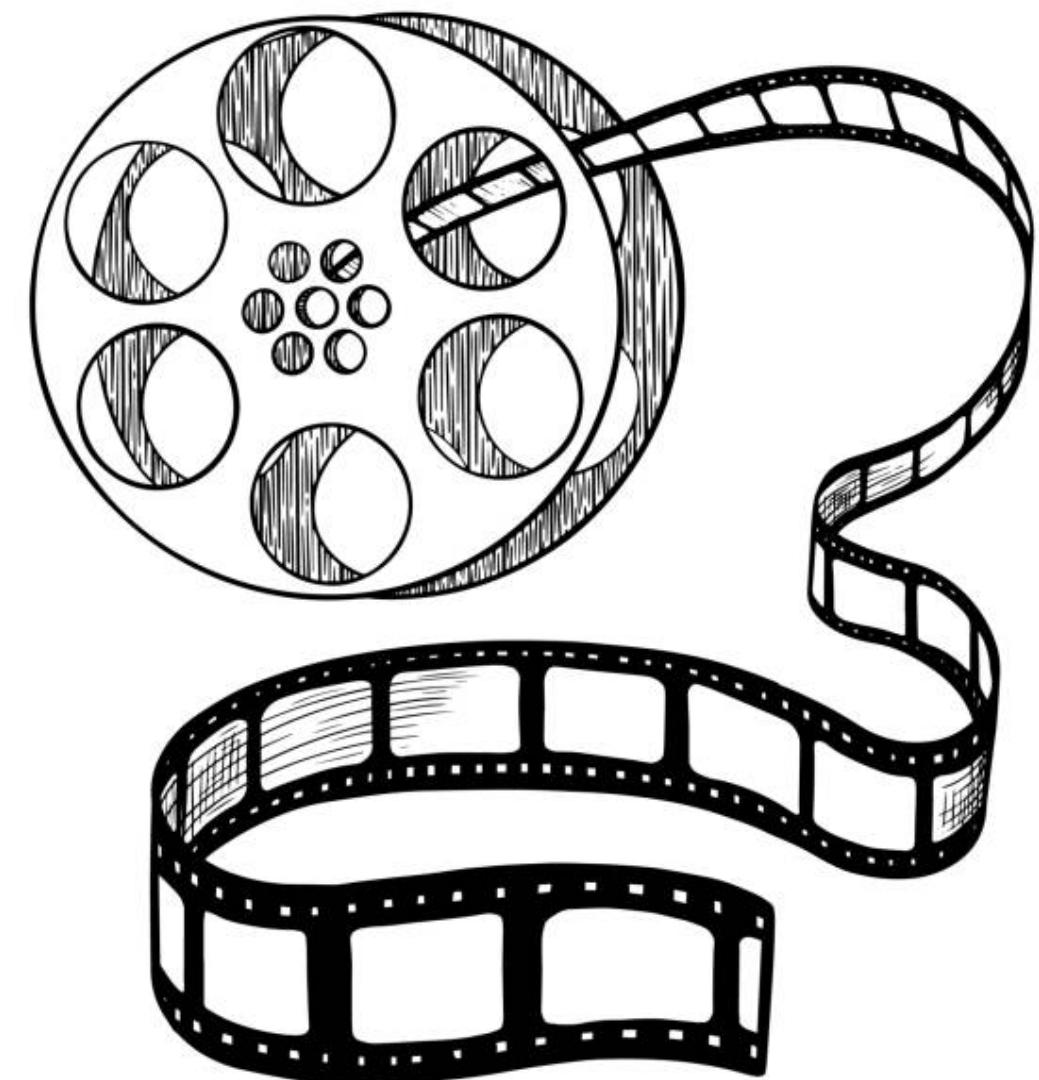
Scenario 1: Facebook

Face recognition



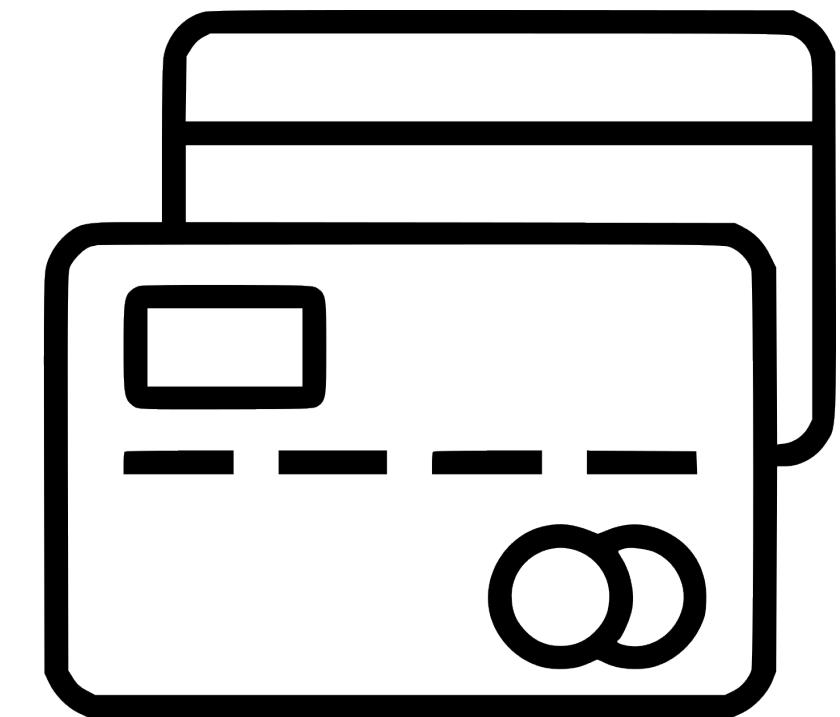
Scenario 2: Netflix

Movie recommendation



Scenario 3: Fraud detection

Credit card transactions

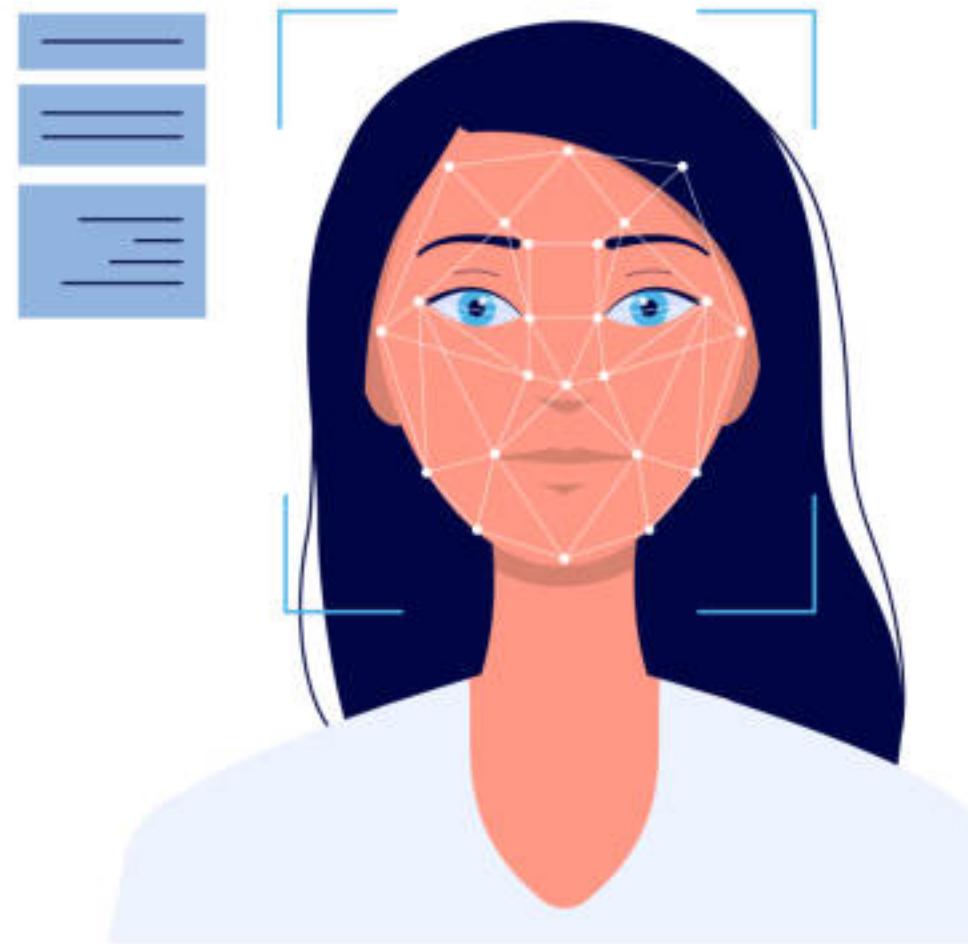


Quiz: Supervised vs. Unsupervised vs. Reinforced?

Daily life

Scenario 1: Facebook

Face recognition

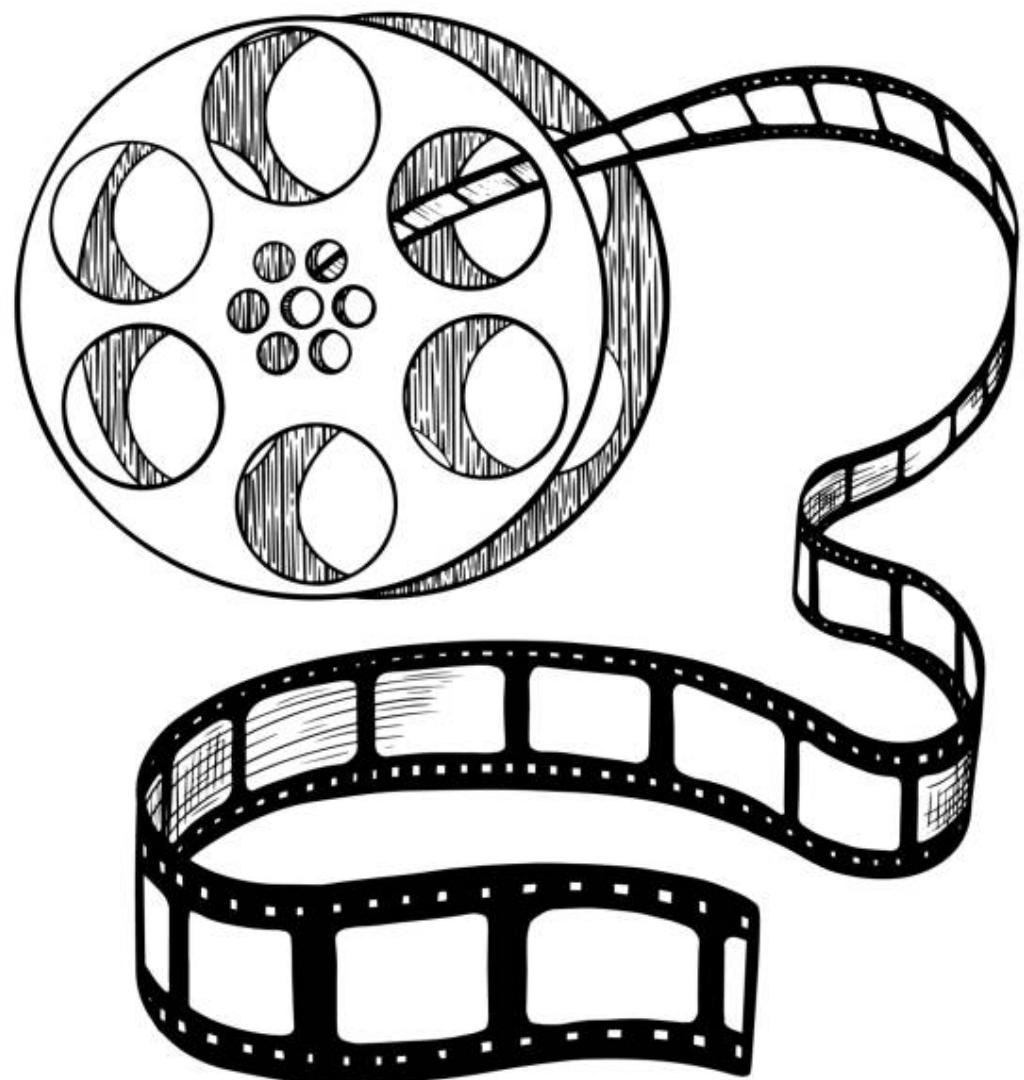


It is supervised learning!

Here Facebook is using tagged photos to recognize the person. Therefore, the tagged photos become the labels of the pictures.

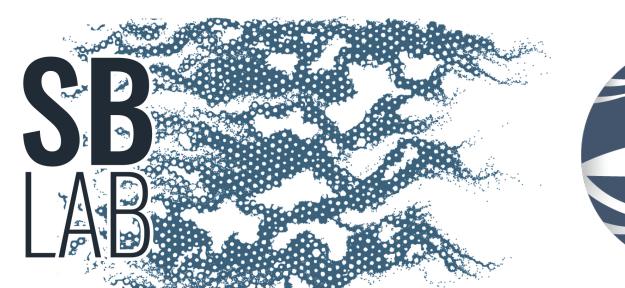
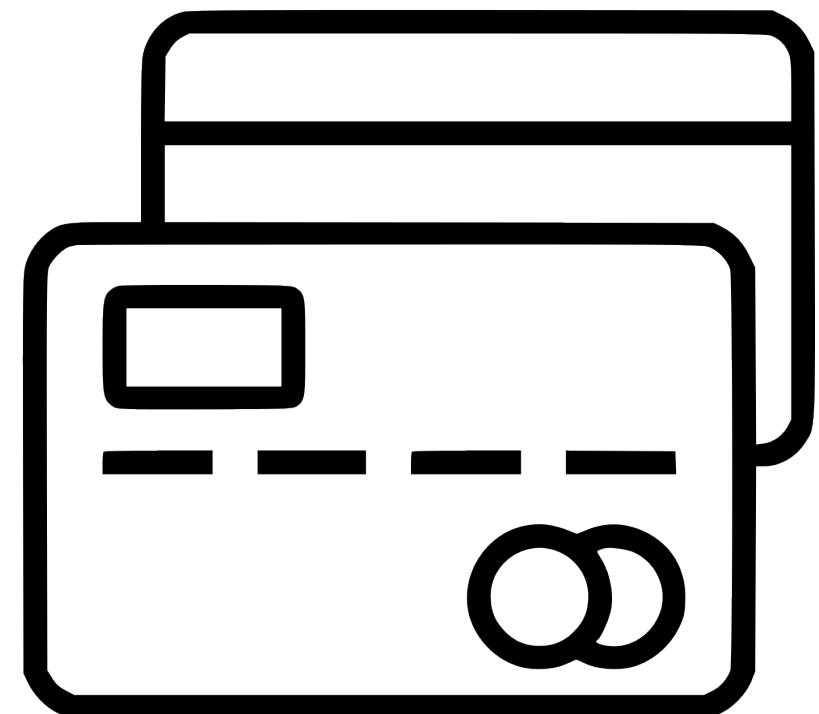
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Scenario 3: Fraud detection

Credit card transactions

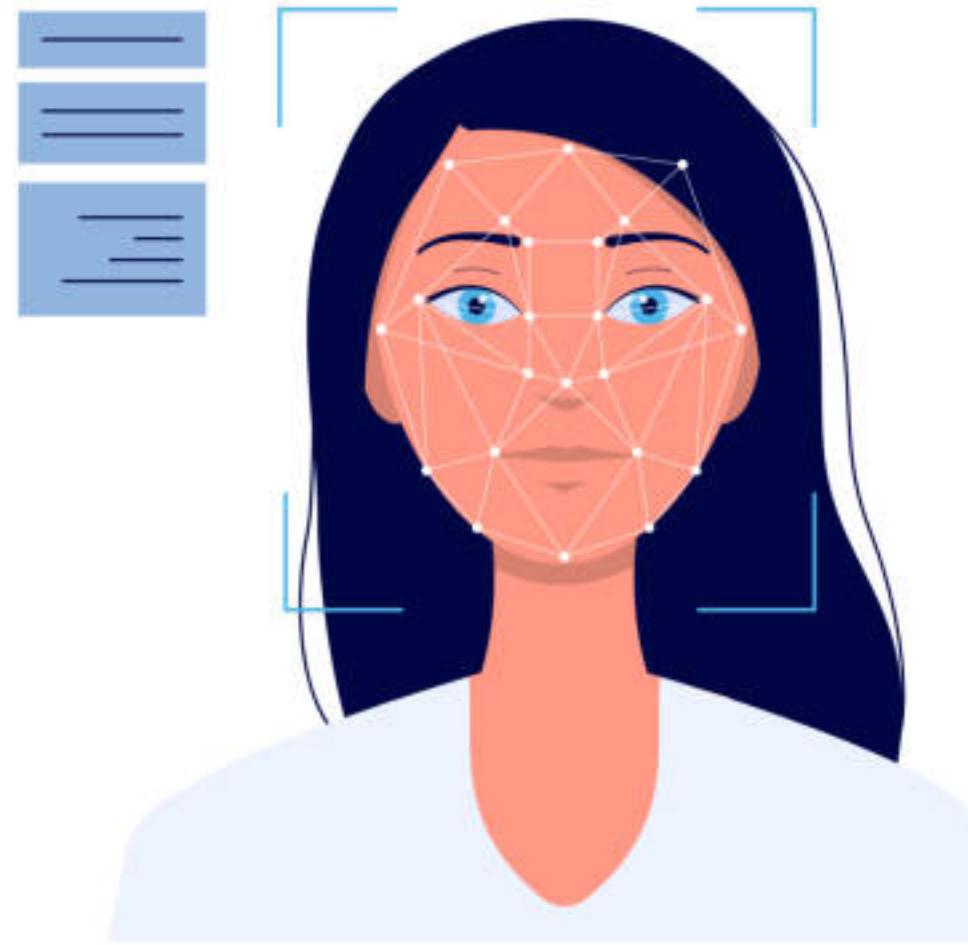


Quiz: Supervised vs. Unsupervised vs. Reinforced?

Daily life

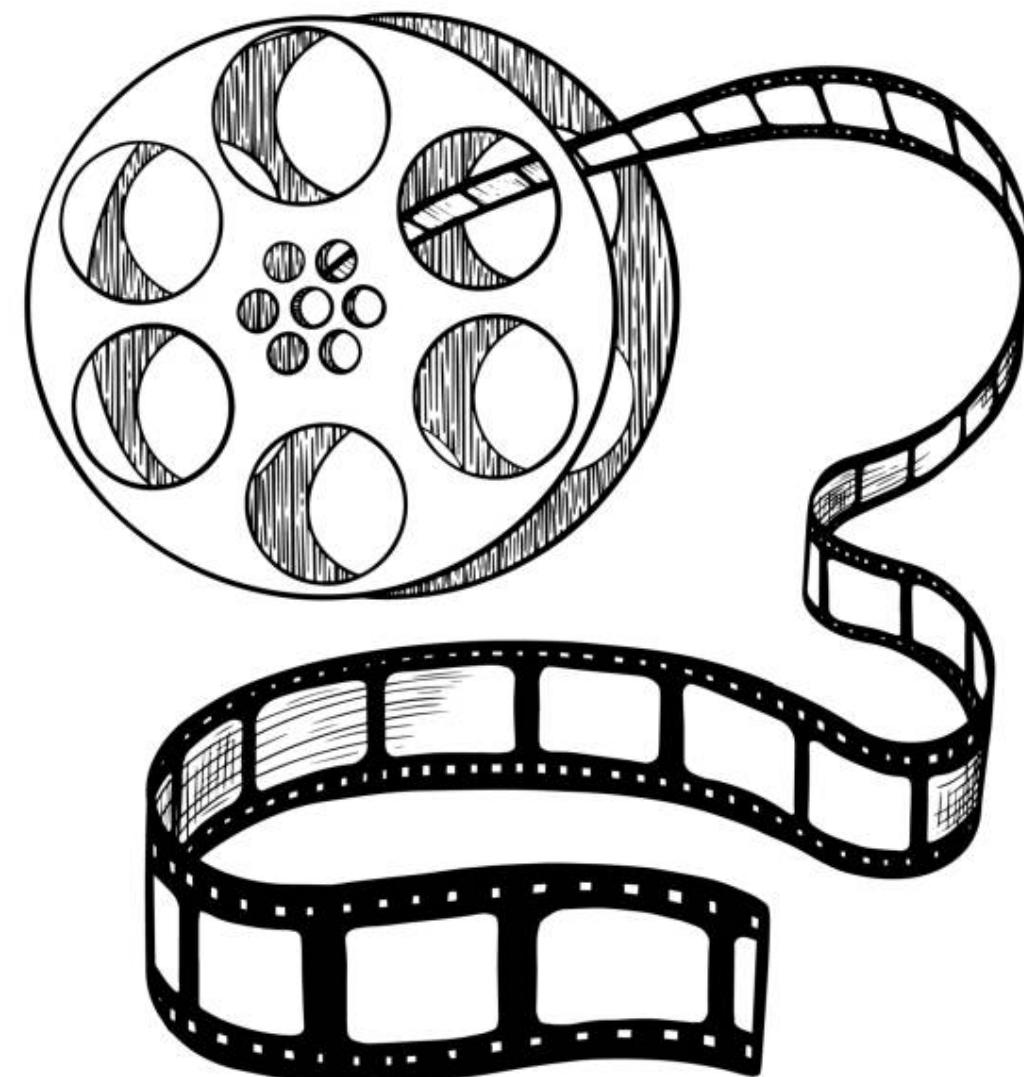
Scenario 1: Facebook

Face recognition



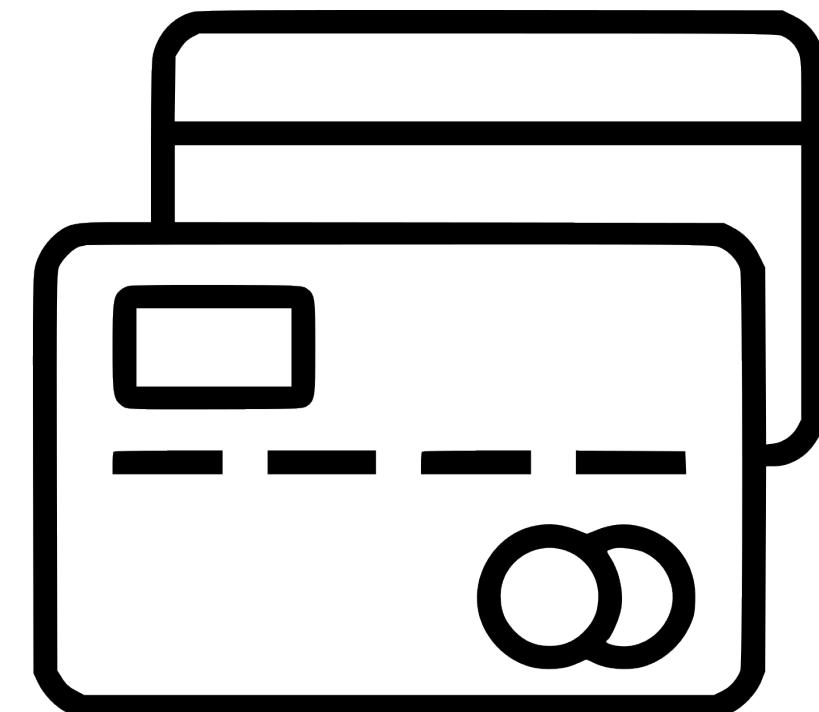
Scenario 2: Netflix

Movie recommendation



Scenario 3: Fraud detection

Credit card transactions



It is supervised learning!

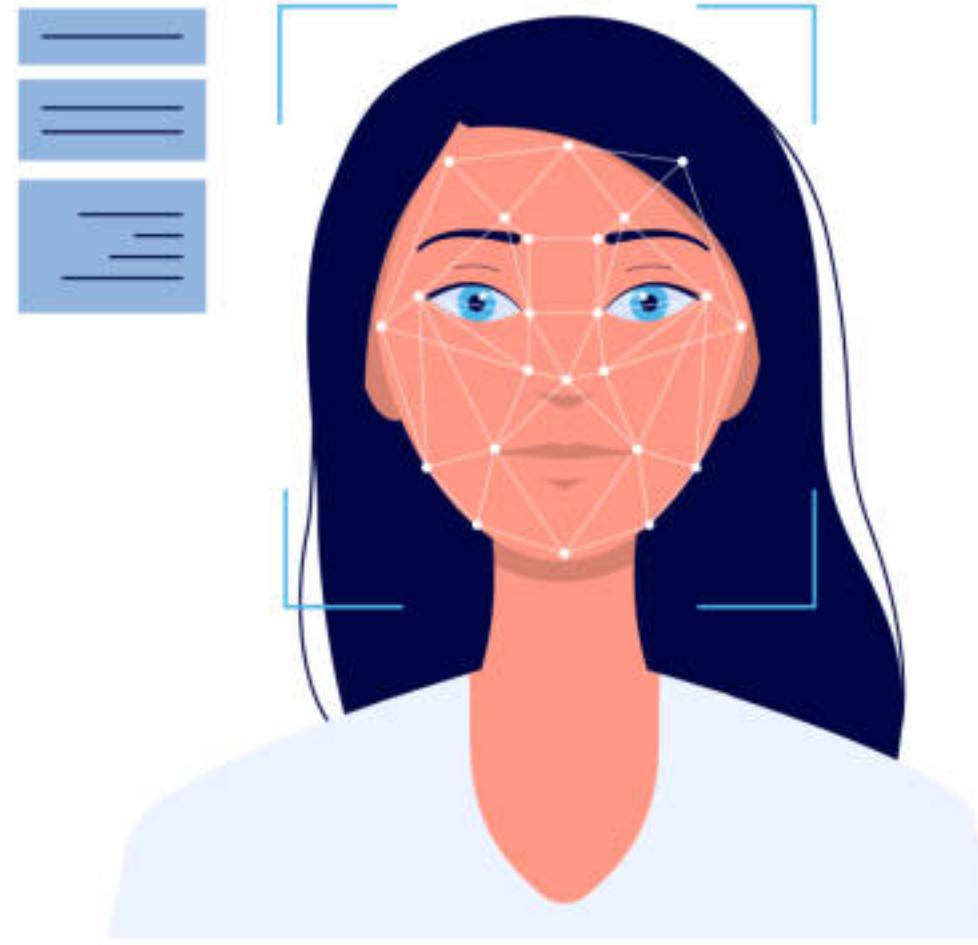
The model trains a classifier on pre-existing labels (genres of movies). Evaluate the features based on your likes/dislikes and then recommend new movies based on similar features.

Quiz: Supervised vs. Unsupervised vs. Reinforced?

Daily life

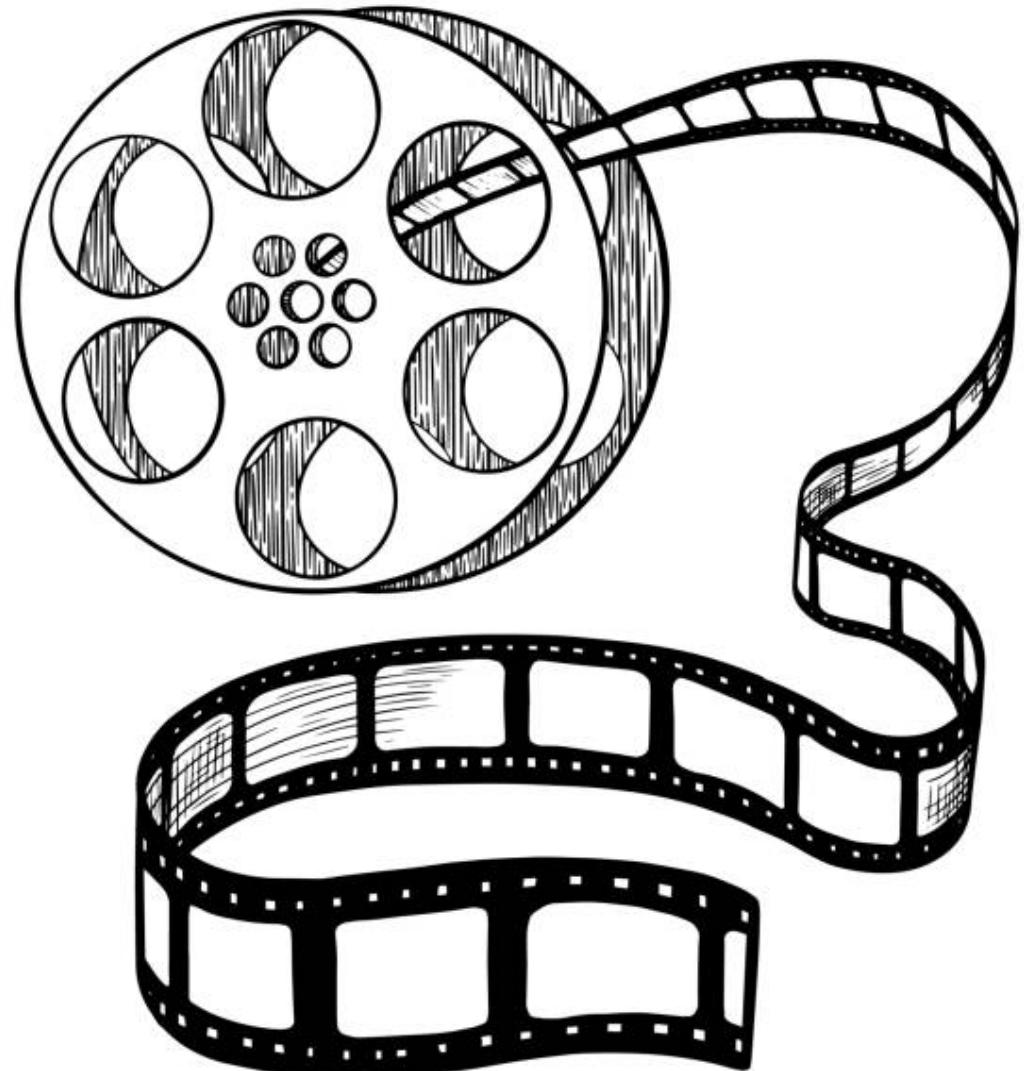
Scenario 1: Facebook

Face recognition



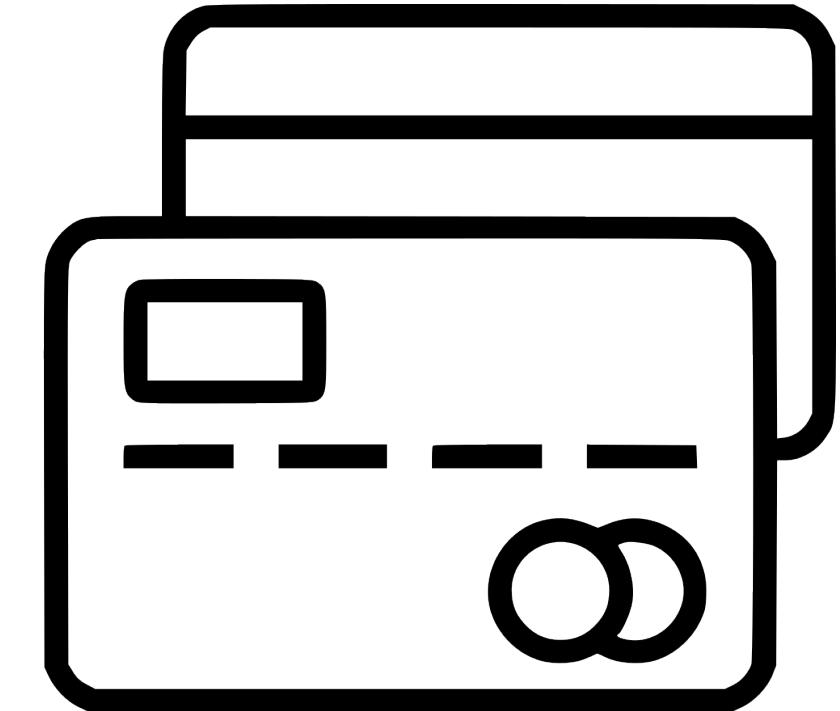
Scenario 2: Netflix

Movie recommendation



Scenario 3: Fraud detection

Credit card transactions



It is unsupervised learning!

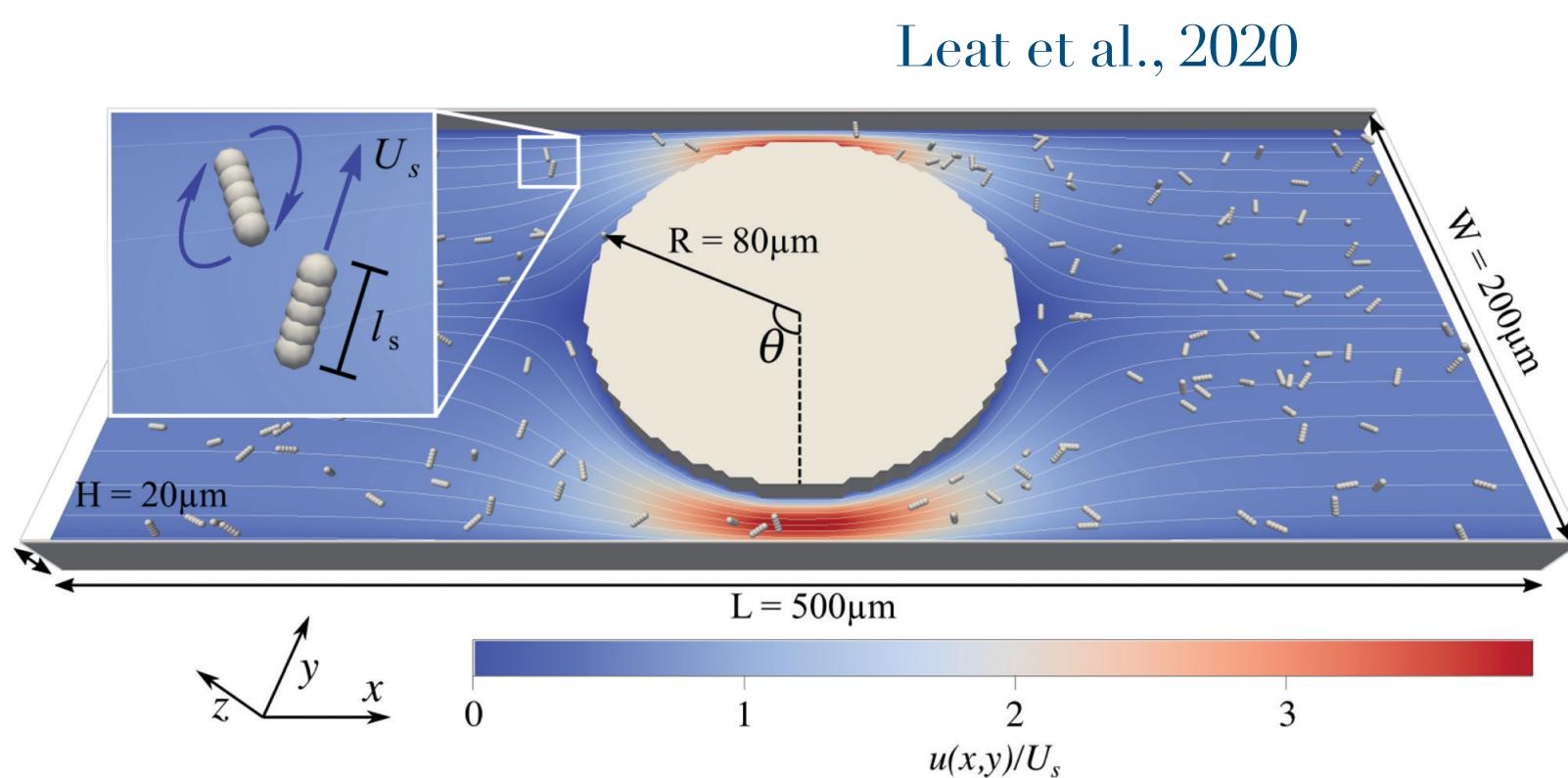
Suspicious transactions are not defined, hence there are no labels of "fraud" and "not fraud". The model tries to identify outliers by looking at anomalous transactions and flags them as 'fraud'.

Quiz: Supervised vs. Unsupervised vs. Reinforced?

Hydrology

Scenario 1: Bioremediation

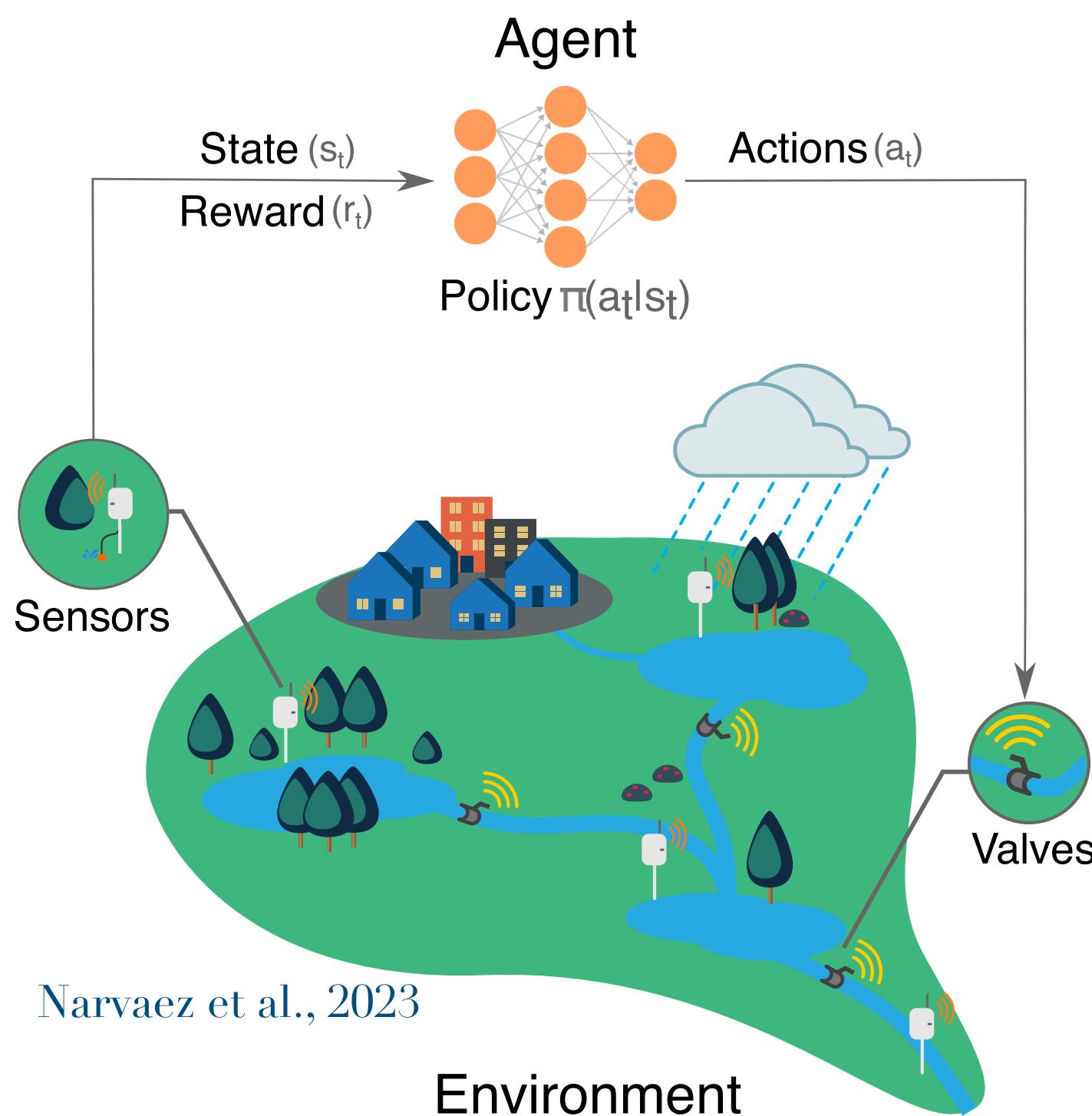
Determine best species



Leat et al., 2020

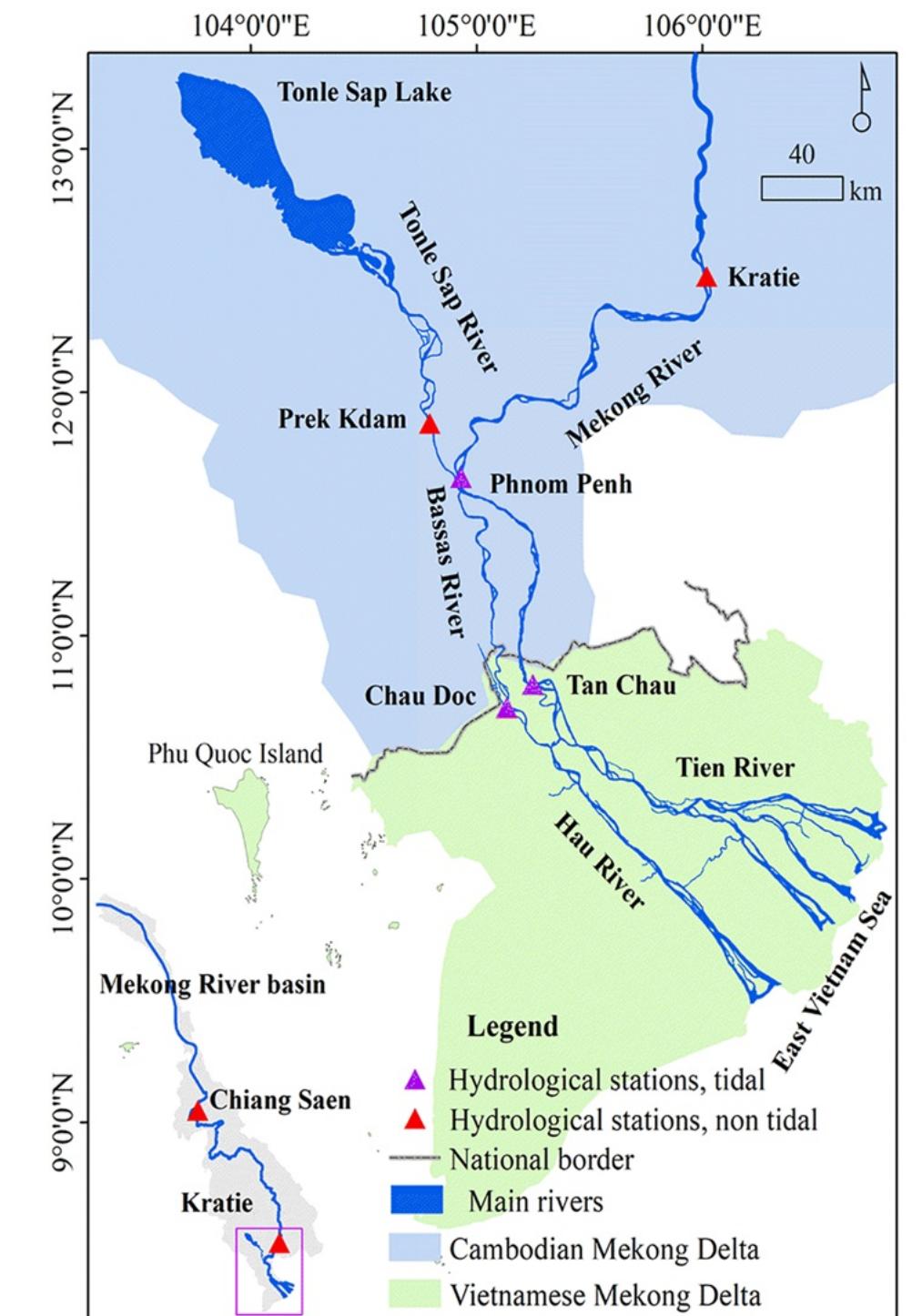
Scenario 2: Storms

Systems adapt to storms



Narvaez et al., 2023

Scenario 3: River discharge



Than et al., 2021