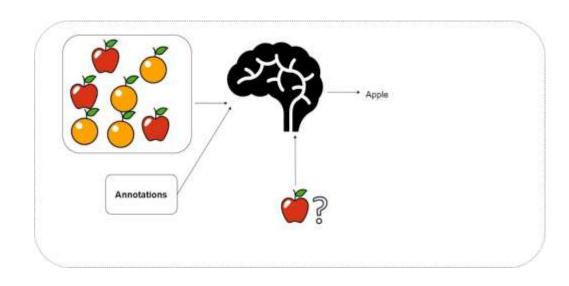
## The Machine Learning Landscape

- Dr. Sifat Momen (SfM1)



#### Learning goals

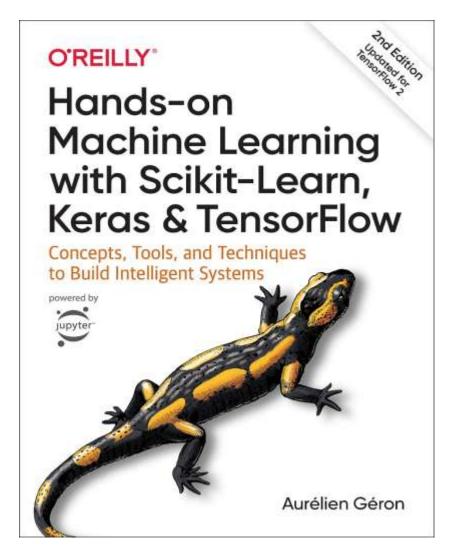
- After this presentation, you should be able to
  - Explain what machine learning is
  - Understand the different machine learning systems
  - Know different jargons related to machine learning
  - Understand the main challenges in machine learning
  - Understand the ethical concerns in machine learning

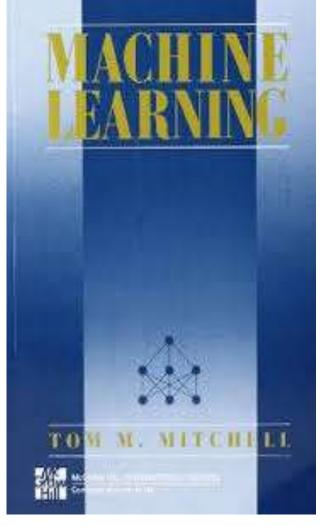
#### CSE 445: Machine Learning

- Welcome to CSE 445: Machine Learning Spring 2025
- I am offering one section this semester
- Please note that you are <u>not</u> allowed to take classes (and specifically exams) in other sections
- This is a senior level course
  - It is expected that students should be able to demonstrate high degree of maturity
  - This includes reading and learning on own from existing literature, writing codes, conducting experiments, analyzing results and writing good technical paper(s).

#### Recommended Materials

- Lecture Materials
- Scholarly Papers





#### Project

- The objective of the project work is for you to
  - Experience a hands on skills in ML.
- Students will have to do innovative projects.
- I am not expecting the projects to have commercial value but it must address a realistic problem and should solve it.
- It should be ideally publishable in a reputed conference/journal.

## What is expected of you?

- Work hard
- Have an inquisitive mind
- Continued effort (You will not be able to do anything magically overnight)
- Consistency is the key to do well in this course

You need to have some backgrounds in

<u>Linear Algebra</u>

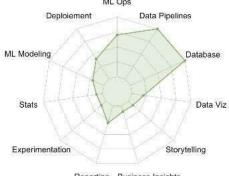
<u>Calculus</u>

<u>Statistics</u>

<u>Algorithms</u>

#### Types of Data Professionals

#### **Data Engineer**



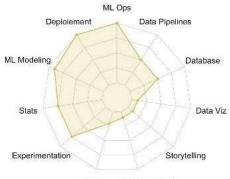
Reporting Business Insights

**Data Scientists** ML Ops

#### **Data Scientist**



#### ML Engineer



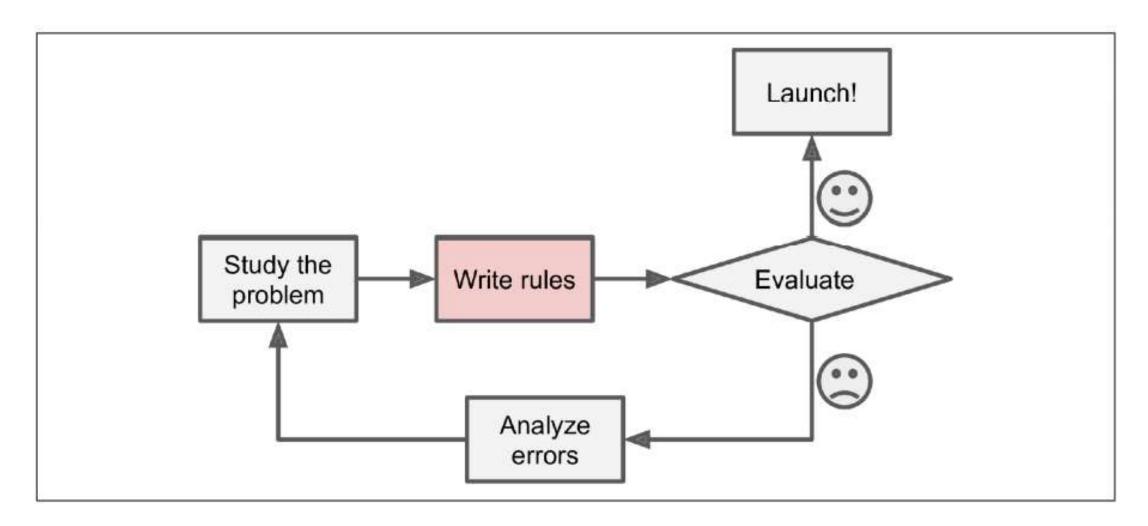
Reporting Business Insights

Data Analysts ML Ops

#### **Data Analyst**



## Traditional Approach to Solving Problems



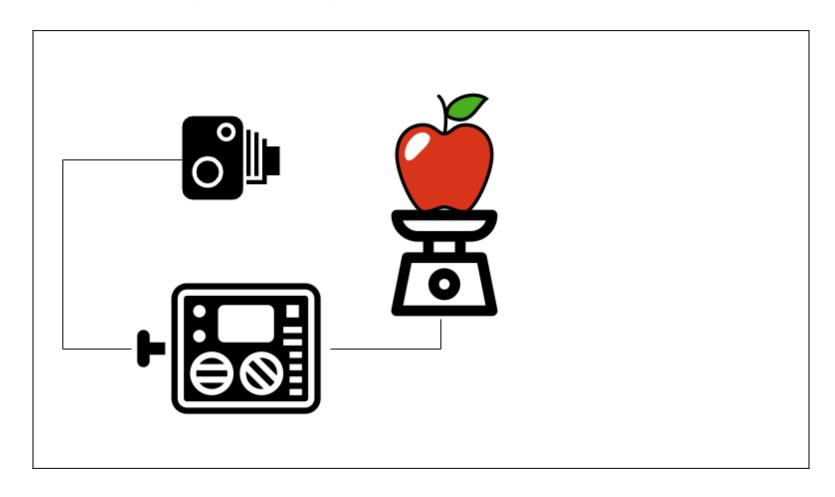
#### What is Machine Learning?

The best way to understand what a machine learning is and appreciate the notion of it is to start with an example.

## A Thought Experiment

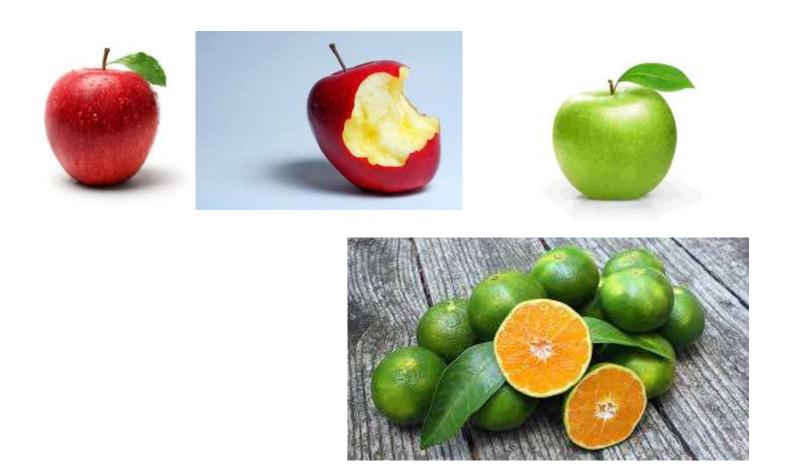
- Imagine we have a physical machine that can take the side-view picture
  of an object placed on it and measure the weight of the object.
- Object placed is either an <u>orange</u> or an <u>apple</u>
- The measurable features in this thought experiment are color and weight

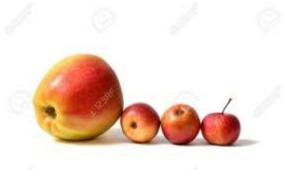
## A Thought Experiment



Now, you are tasked to write a program that determines if the object placed is an apple or an orange. How would you do that?

## Real World





## What about recognizing a cat from a picture?





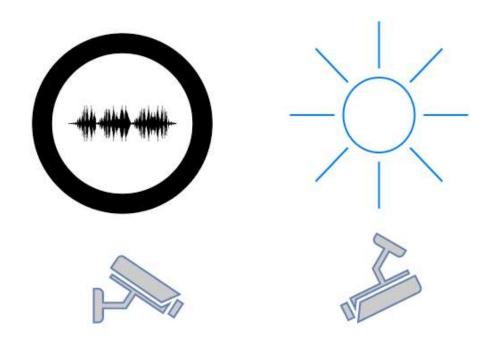
#### Real World Data

Noisy

Illumination

Occlusion

Camera angle



Exists a wide range of variation (in terms of features such as color, texture, weight etc..)

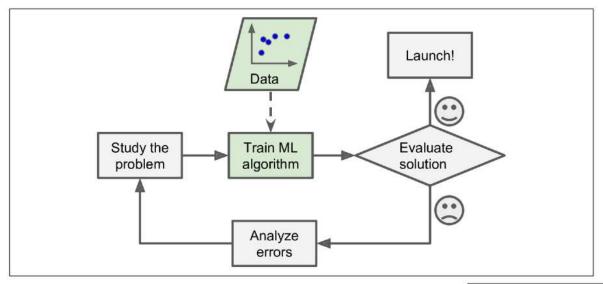
#### Real World Data (cont...)

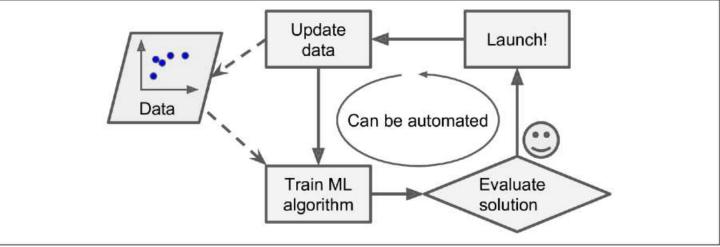
Consequently, it is impossible to write a rule based programming to recognize a cat or differentiate between fruits (such as orange and apple)

A shift in the paradigm is required

- Data driven Computing (Welcome to Machine Learning)
- We use data to create model
- Data is at the core of ML
- Without data, you will not be able to run a ML project.

## Machine Learning Approach

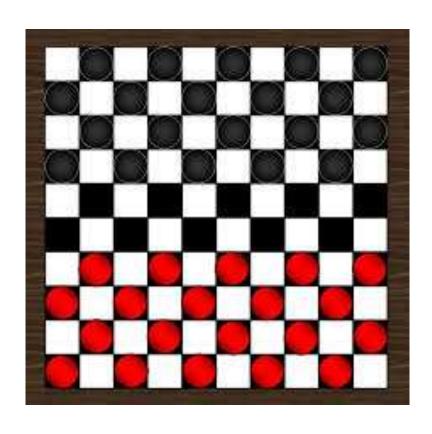




## Well posed learning problems

- Machine Learning
  - It is a science that grew up from AI
  - The term, ML, was coined in 1959, by Arthur Samuel
- Machine Learning is the field of study that gives computers the ability to learn without being explicitly programmed (Arthur Samuel, 1959)
- A computer program is said to learn from <u>experience E</u>, with respect to some <u>task T</u> and <u>performance measure P</u> if its performance at <u>task T</u>, as measured by <u>P</u>, improves with <u>experience E</u>. (Tom M. Mitchell)

#### Checkers learning problem



Task *T*: playing checkers

Performance measure, P: % of games won against opponents

Training Experience, *E*: Playing practice games against itself

#### Handwriting Recognition Problem

Task T: Recognizing and classifying handwritten words/characters within images

Performance measure P: % of words/characters correctly classified

Training Experience, *E*: Predicting the word/character from a collection of handwritten words/characters with given classifications

- MNIST Dataset (Yann Lecunn et al.)
- BanglaLekha Isolated (Nabeel Mohammed, Sifat Momen et al.)
  - https://data.mendeley.com/datasets/hf6sf8zrkc/2
  - Dataset comprises of 50 Bangla basic characters, 10 bangla numerals and 24 selected compound characters (total of 1,66,105 handwritten character images)

## Robot driving learning problem

Task T: driving on public highways using vision sensors

Performance measure, P: Average distance travelled before an error (as judged by a human observer)

Training Experience, *E*: A sequence of images and steering commands recorded while observing a hums driver.

# Classifying an incoming email as either a spam or not spam

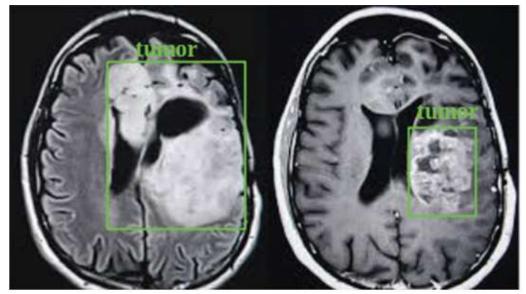
Task T: Correctly classify an email as either spam or not spam

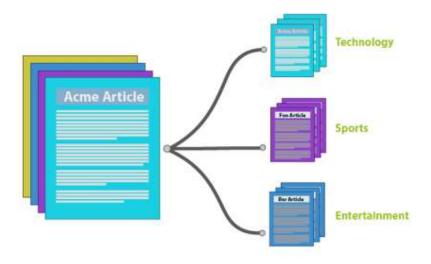
Performance measure, P: % of emails correctly classified

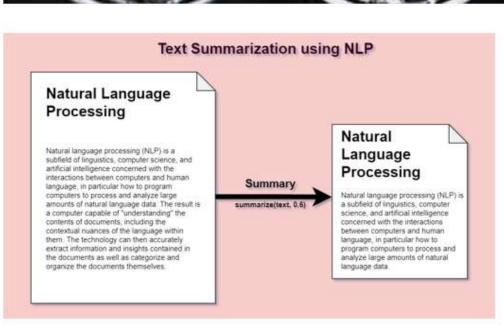
Training Experience, E: A repertoire of emails with proper labeling as either spam or not.

## Some Applications of Machine Learning

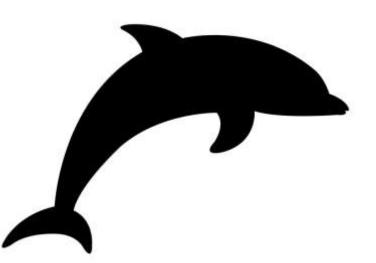












Write a story with dolphins and turtles as characters and include a moral in the story



Once upon a time, in a beautiful coral reef deep in the ocean, there lived a group of dolphins and a group of turtles. The dolphins were known for their intelligence and speed, while the turtles were known for their wisdom and long lives. The two groups lived separately and rarely interacted with each other. The dolphins were always busy swimming and playing, while the turtles were content to take their time and enjoy the peacefulness of the reef. One day, a group of humans came to the reef to fish. They used big nets to catch fish, not caring if they caught anything else in the process. In the process, they caught many of the reef's inhabitants, including the dolphins and turtles. The dolphins, being faster and more agile, managed to escape the nets, but the turtles were not as lucky. Many of them were caught and taken away. The dolphins felt guilty for not being able to help their friends and decided to come up with a plan to save the turtles.

They used their intelligence and speed to come up with a strategy to chase the fishing boat away and free the trapped turtles. The plan was successful and the boat was chased away, but not before many of the turtles were harmed. The dolphins and turtles were able to free the remaining turtles and bring them back to the reef. After that experience, the dolphins and turtles realized the importance of working together. They learned that each of them had unique strengths and weaknesses, and that by combining their abilities, they could protect the reef and all its inhabitants better. From then on, the dolphins and turtles worked together, and the reef was safe and prosperous once more. The moral of the story is that everyone is unique, and we all have different strengths and weaknesses, but when we work together, we can achieve so much more. Unity is strength and cooperation leads to protection of environment and preserving lives.



Write a rap about North South University

#### Verse 1:

Listen up y'all, I got a story to tell About a uni that's doing well North South University, the name you should know Where education and knowledge both flow

#### **Chorus:**

North South, North South
We're the best, no doubt
North South, North South
We're reaching new heights, without a doubt

#### Verse 2:

From BBA to Engineering, we got it all We're making sure you're ready for the call Professors who care, and a campus so fine We're creating leaders, one student at a time

#### Chorus:

North South, North South
We're the best, no doubt
North South, North South
We're reaching new heights, without a doubt

#### Verse 3:

Research and innovation, that's our game
We're putting Bangladesh on the map, with our name
We're not just a uni, we're a community
We're here to empower, and set you free

#### Chorus:

North South, North South
We're the best, no doubt
North South, North South
We're reaching new heights, without a doubt

#### Outro:

So come and check us out, and take a tour North South University, where your future is secure.

## Types of Machine Learning Systems

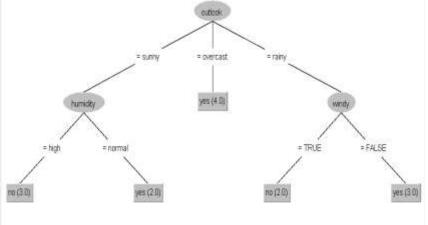
- Various Types of Machine Learning Systems based on the following criteria:
  - Whether or not they are trained with human supervision (supervised, unsupervised, semi-supervised, reinforcement learning)
  - Whether or not they can learn incrementally on the fly (online versus batch learning)
  - Whether they work by simply comparing new data points, or instead by detecting patterns in the training data and building a predictive model(instance-based versus model-based learning)

#### Supervised Learning

In supervised learning, the algorithm builds a mathematical model from a set of data that contains inputs and labels (desired outputs) <X, y>.

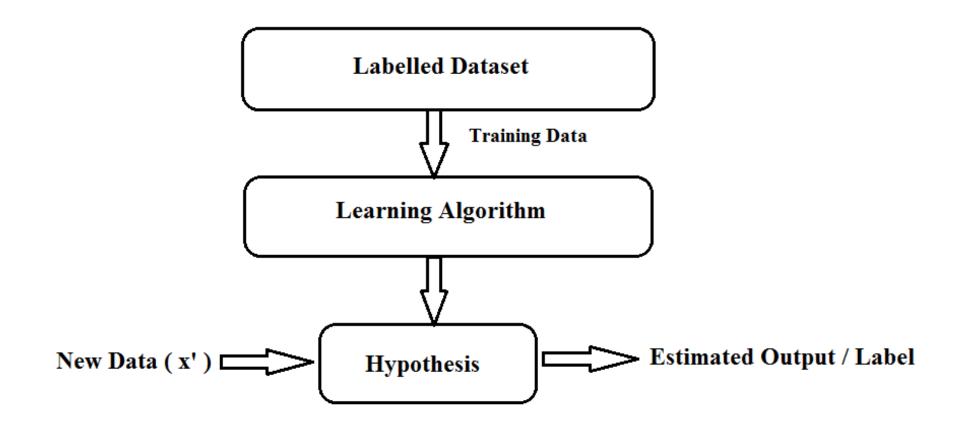
No.	1: outlook	2: temperature	3: humidity	4: windy	5: play
	Nominal	Nominal	Nominal	Nominal	Nominal
1	sunny	hot	high	FALSE	no
2	sunny	hot	high	TRUE	no
3	overcast	hot	high	FALSE	yes
4	rainy	mild	high	FALSE	yes
5	rainy	cool	normal	FALSE	yes
6	rainy	cool	normal	TRUE	no
7	overcast	cool	normal	TRUE	yes
8	sunny	mild	high	FALSE	no
9	sunny	cool	normal	FALSE	yes
10	rainy	mild	normal	FALSE	yes
11	sunny	mild	normal	TRUE	yes
12	overcast	mild	high	TRUE	yes
13	overcast	hot	normal	FALSE	yes
14	rainy	mild	high	TRUE	no

The data contains 14 instances and 4 attributes. The target



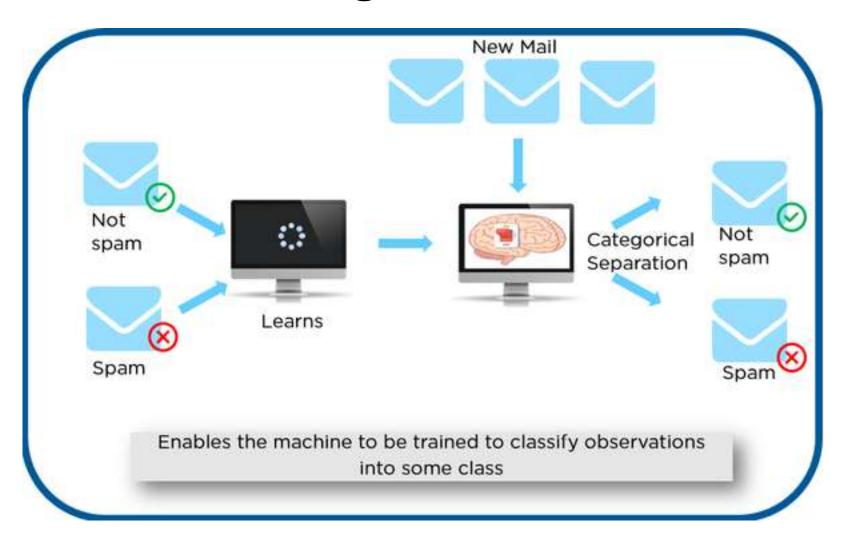
A Decision Tree Model

## Supervised Learning

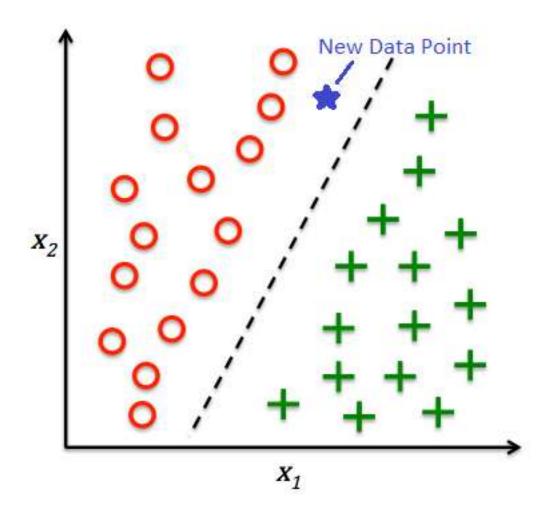


#### **SUPERVISED LEARNING**

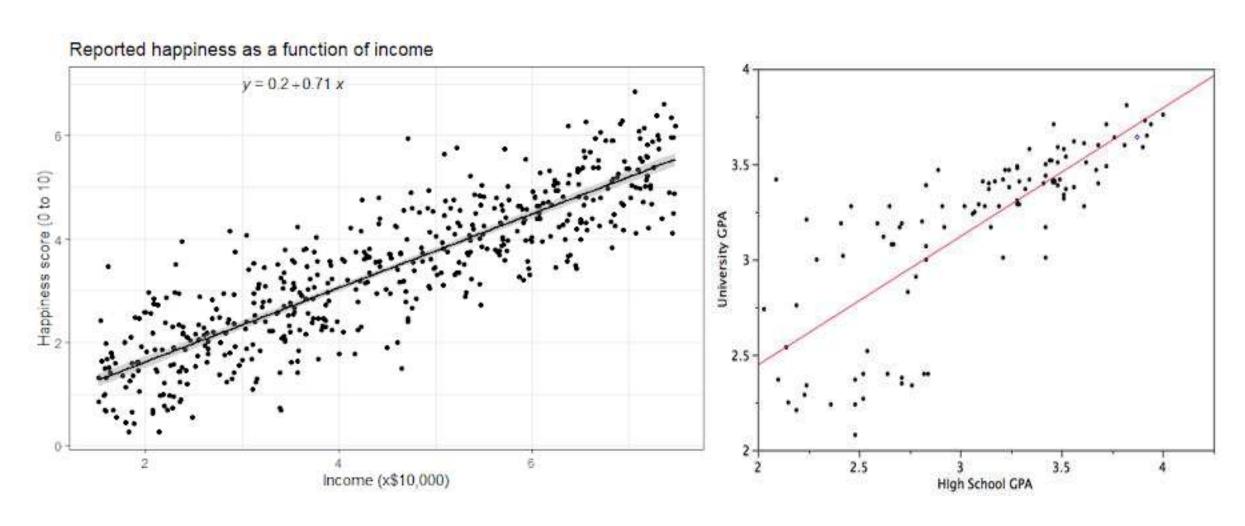
## Supervised Learning - Classification



## Supervised learning - Classification



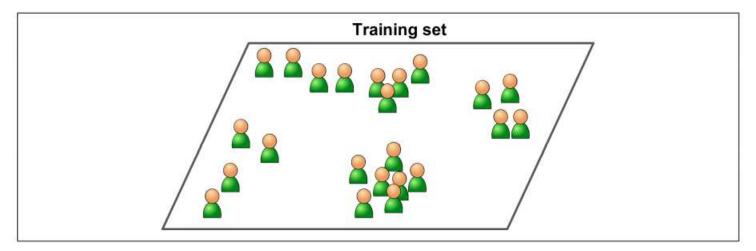
## Regression

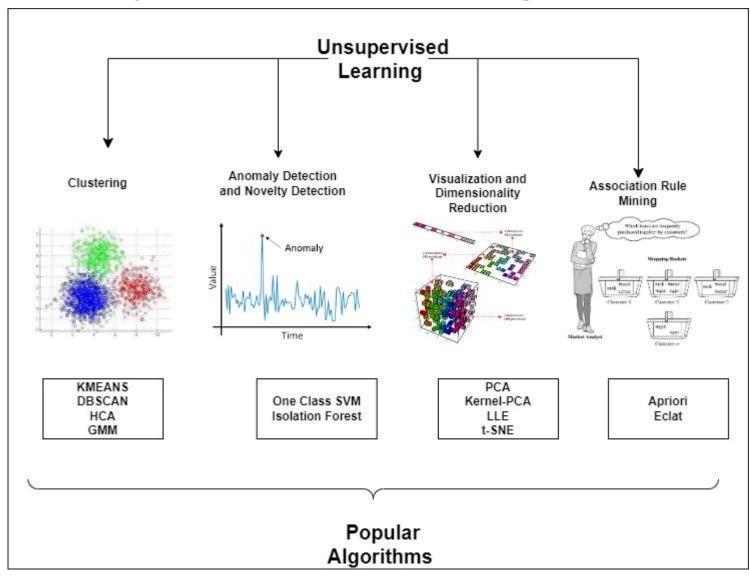


## **Examples of Supervised Learning Algorithms**

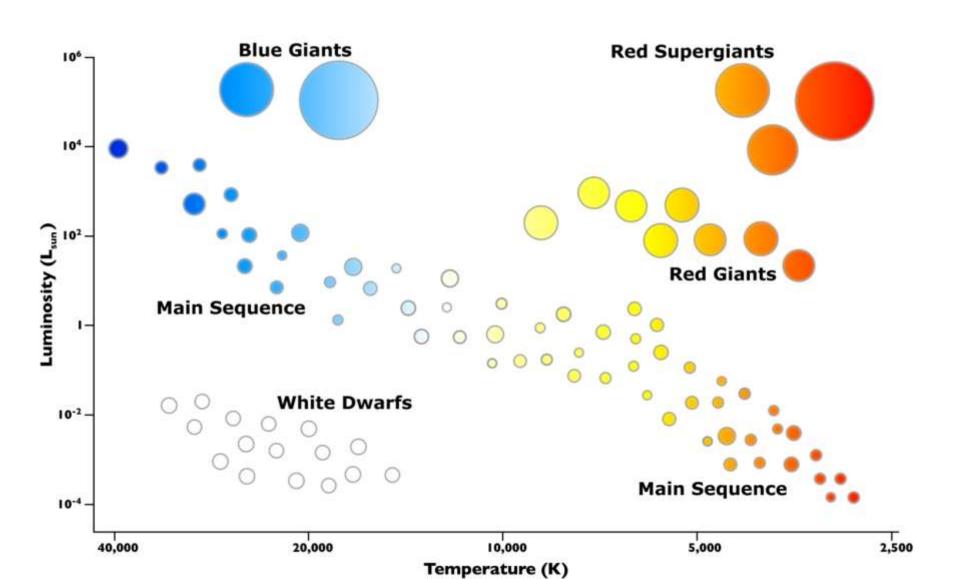
- K-Nearest Neighbors
- Linear Regression
- Logistic Regression
- Support Vector Machines
- Decision Trees
- Random Forests
- Neural Networks
- ..

In Unsupervised Learning, the training data is <u>unlabeled</u>. The system tries to learn without a teacher.





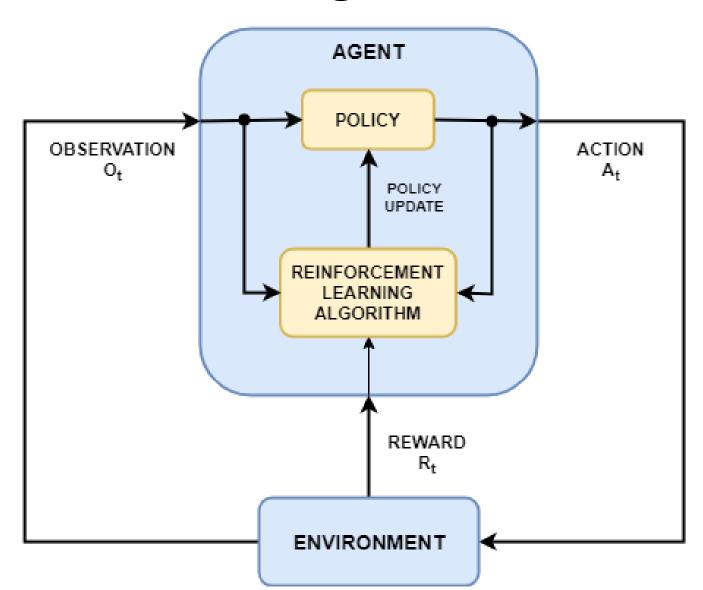
1 H																	2 He
3 Li	4 Be											5 B	6 C	7 N	8 O	9 F	10 Ne
11 Na	12 Mg											13 Al	14 Si	15 P	16 S	17 CI	18 Ar
19 K	20 Ca	21 Sc	22 Ti	23 V	24 Cr	25 Mn	26 Fe	27 Co	28 Ni	29 Cu	30 Zn	31 Ga	32 Ge	33 As	34 Se	35 Br	36 Kr
37 Rb	38 Sr	39 Y	40 Zr	41 Nb	42 Mo	43 Tc	44 Ru	45 Rh	46 Pd	47 Ag	48 Cd	49 In	50 Sn	51 Sb	52 Te	53 I	54 Xe
55 Cs	56 Ba	57 La	72 Hf	73 Ta	74 W	75 Re	76 Os	77 Ir	78 Pt	79 Au	80 Hg	81 TI	82 Pb	83 Bi	84 Po	85 At	86 Rn
87 Fr	88 Ra	89 Ac	104 Rf	105 Db	106 Sg	107 Bh	108 Hs	109 Mt	110 Ds	111 Rg	112 Cn	113 Nh	114 Fl	115 Mc	116 Lv	117 Ts	118 Og
			58 Ce	59 Pr	60 Nd	61 Pm	62 Sm	63 Eu	64 Gd	65 Tb	66 Dy	67 Ho	68 Er	69 Tm	70 Yb	71 Lu	
			90 Th	91 Pa	92 U	93 Np	94 Pu	95 Am	96 Cm	97 Bk	98 Cf	99 Es	100 Fm	101 Md	102 No	103 Lr	



## Semi-Supervised Learning

- Supervised learning works on labeled data
- Unsupervised learning works on unlabeled data
- In Semi-Supervised learning, the training data contains both labeled and unlabeled data
- Scenario
  - We have a large dataset. Manual labeling of the entire dataset is laborious and expensive
  - One solution could be to label a sample of the dataset and train the labeled portion to create a model. This would however mean that we are not fully utilizing the larger dataset we have and thus the model that we create may be less robust
  - A potential solution: (1) Label a sample of the large dataset, (2) train a model using this labeled portion, (3) use the model to predict the unlabeled portion (pseudo-labeling), (4) train using the entire dataset

# Reinforcement Learning



### Batch and Online Learning

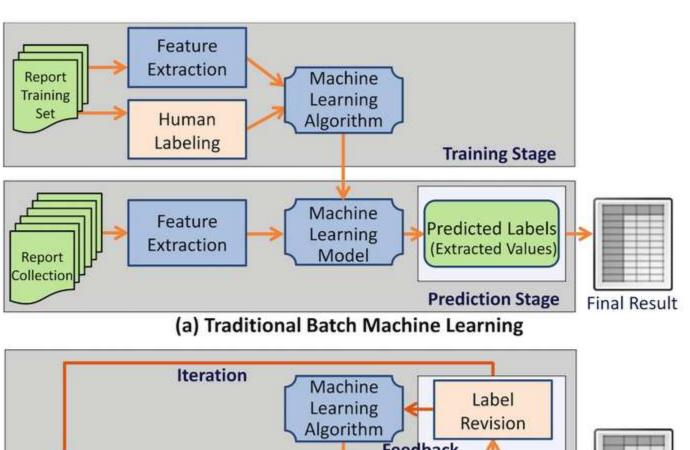
 Another criterion used to classify Machine Learning Systems is whether or not the system can learn incrementally from a stream of incoming data

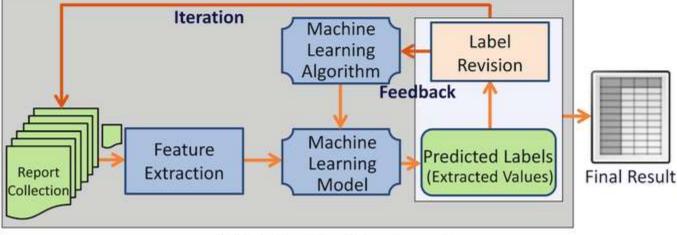
#### Batch Learning

- Incapable of learning incrementally
- It must be trained using all available data
- Generally takes a lot of time and computing resources so it is typically done offline
- Hence, also known as offline learning

#### **Batch Versus Online Learning**

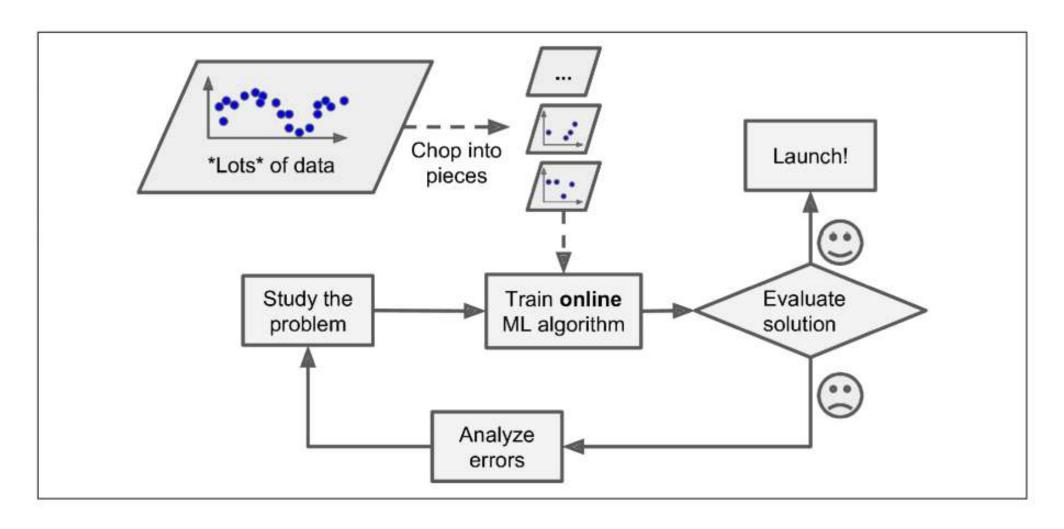
- Online Learning is great for systems that receive data as continuous flow (e.g. stock prices) and need to adapt autonomously.
- It is also a good option if you have limited computing resources – Once an online learning system has learned about new instances, it does not need them anymore, so you can discard them
- Online learning algorithms can also be used to train systems on huge datasets that cannot fit in one machine's main memory (this is called *out of core* learning). The algorithm loads part of the data, runs a training step on that data, and repeats the process until it has run all of the data





(b) Online Machine Learning

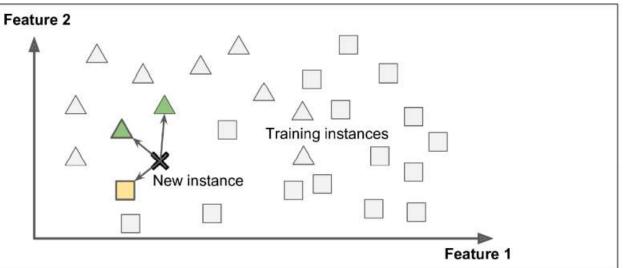
### Online Learning to Handle Large Dataset



#### Instance Based Versus Model Based Learning

- One more way to classify machine learning algorithms is by how they generalize.
- Instance Based Learning
  - The most trivial form of learning is rote learning (to learn by heart)
  - System learns the examples by heart, then generalizes to new case by using a similarity measure to compare them to the learned examples (or a subset of

them)



#### Instance Based Versus Model Based Learning

- Model Based Learning
  - Another way to generalize from a set of examples is to build a model of these examples and then use the model to make predictions

