# Introduction to Machine Learning



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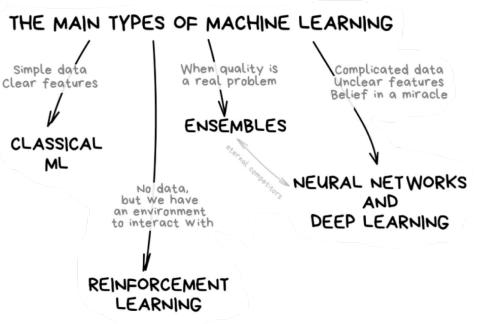
#### **Needs for Machine Learning**

- Consists of understanding that humans can use a computer to recognize algorithms or patterns.
- Computers are far more capable than humans to recognize, analyze and comprehend data and the hidden patterns that lie within them.
- This realization birthed the creation of machine learning.
- The single true goal of machine learning is to predict results based on incoming data. THAT IS IT.

### **Different types of Machine learning**

The idea of machine learning can be broken up into 4 main categories.

- Classical Learning
- Reinforcement Learning
- Neural Nets And Deep Learning
- Ensemble Methods

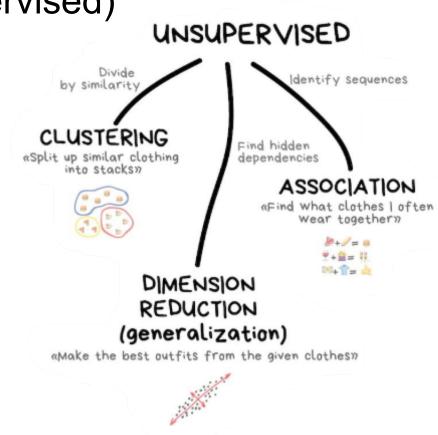


### **Classical Machine Learning**

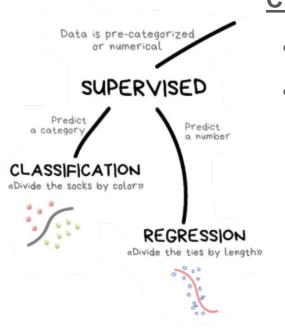
- Encompasses Supervised and Unsupervised learning
  - $\circ$   $\,$  As well as all the subcategories that fall within those  $\,$
- Also includes Ensemble Learning Methods
  - Several classical ML models, aggregate their predictions to get final prediction
- Is one of the earliest forms of general machine learning because it is a basic concept.
  - Has been around since the 50's
  - Came from solving pure math tasks like pattern in numbers and proximity of data points

### Classical Learning (Unsupervised)

- Unsupervised learning occurs when a machine learning model is given data without labels.
- The model simply looks for patterns within the features of the data.
- Ex.
  - Machine is given a dataset of people who do and don't have cancer. The machine only has the different features of each person to recognize patterns between them that may associate to cancer.



## Classical Learning (Supervised)

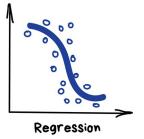


#### **Classification**

- **Binary** two outcomes (yes/no)
  - Ex. Does patient have cancer?
- Multi-class more than two possibilities for the variable being predicted
  - Ex. outcome of an election

#### **Regression**

- Linear Regression Models
  - Used when the variable you are looking to predict is continuous and can be plotted on a line
  - Ex. crop yields
- Polynomial Regression Models
  - Used when the variable can be predicted on a curved line





Classification

### **Binary Classification**

#### Accuracy

Out of all ground truth labels, how many were predicted right?

\*Dumb classifiers that have good accuracies = data imbalance or data skew

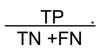
#### Precision

Out of all positive predictions , how many actually are positive ?

#### Recall

Out of positive values in the data set, how many are predicted positive?

\*Reverse functions to focus on negatives (different classifier)



TP+TN

**TP+FN+FP+TN** 



#### Confusion Matrix- How Accurate is Your Model?

True Positive	False Negative
TP	FN
False Positive	True Negative
FP	TN

TP - correctly classified as 1 TN - correctly classified as 0 FP - falsely classified as 1 (correct is 0) FN - falsely classified as 0 (correct is 1)

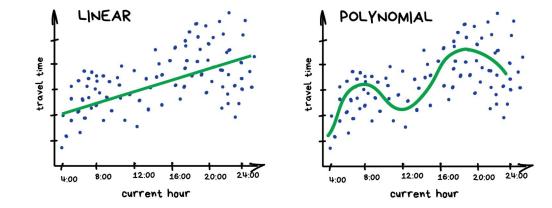
Ground Truth Label	Output of Hypothesis function h(x)~y	
1	0	FN
0	0	TN
0	1	FP
1	1	TP

### **Multi-Class Classification**

- Predicts which individual will win an election (multi-class problem)
  - Data put into the model would be features about prior elections and other factors, and the variable would be who won the election.
- Binary Class: F: for the election (one k-dataset)
  - How many candidates stood
  - Year of election
  - How many from the democrats party
  - How many from republicain
  - $\circ$  Label: won
- The easiest way to create a multi-class classification model is to break it down into multiple binary classification models in a one vs. all classification
- Take the predictions from the majority of the binary classifiers to determine overall decision

#### **Regression Models**

- Regression models are used to predict variables that are continuous and can be put on a number line somewhere.
  - Ex. Price of a stock, Crop yields, Real estate prices
- Simple way to think of it is to predict a variable based on a condition, like traffic based on time of day.
  PREDICT TRAFFIC JAMS
- Computer looks for correlation between features but can take in many more variables then a human and is much more precise in its prediction.



REGRESSION

### Reinforcement Learning (A brief overview)

- Reinforcement learning is a category of machine learning separate from classical learning.
- In RL, the model is not given any data to work with...
- Instead, the model has an environment that it can interact within where it is given a reward system for performing well towards its goal.
  - Ex. Model is tasked with trying to balance a bat on a point.
  - Simulation is run over and over again and the model slowly learns over time what actions prove most fruitful for it (how long the bat stays balanced)
  - Eventually the model becomes proficient at balancing a bat because it has learned what actions and reactions work best for it.