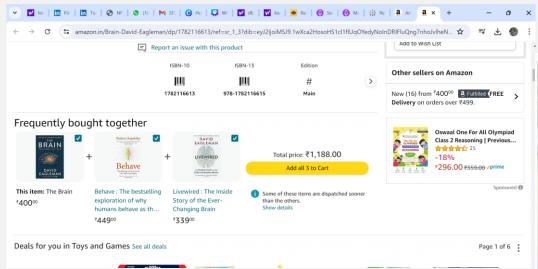




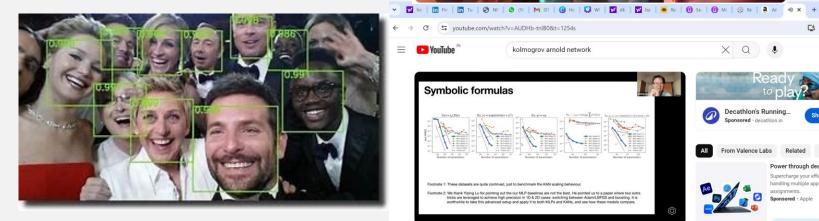
# Introduction to Machine Learning

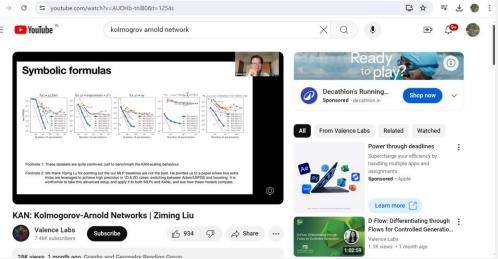
Definition of ML, Types of ML, Why ML?













### Definition of ML

- [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 experience E with respect to some task T and some performance measure P, if its performance on T, as measured by P, improves with experience E.—Tom Mitchell, 1997



# Why to use ML?

- Machine Learning algorithm can often simplify code and perform better.
- Complex problems for which there is no good solution at all using a traditional approach: the best Machine Learning techniques can find a solution.
- Fluctuating environments: a Machine Learning system can adapt to new data.
- Getting insights about complex problems and large amounts of data.



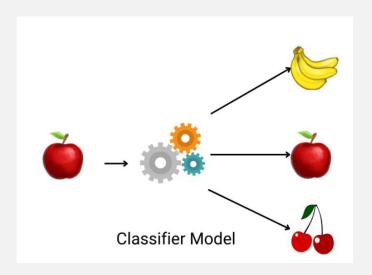
# Types of Machine Learning:

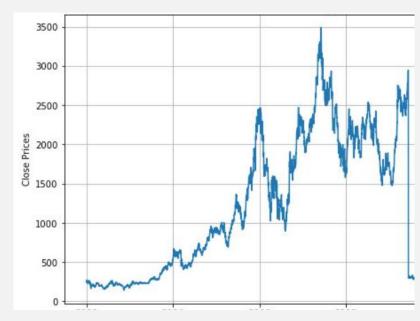
- Supervised learning
- Unsupervised learning
- Semisupervised learning
- Reinforcement Learning.



# Supervised Learning

- The training data you feed to the algorithm includes the desired solutions, called labels.
- A typical supervised learning task is classification.
- Another typical task is to predict a target numeric value, such as the price of a car, given a set of features (mileage, age, brand, etc.) called predictors.



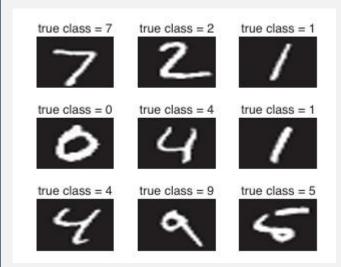


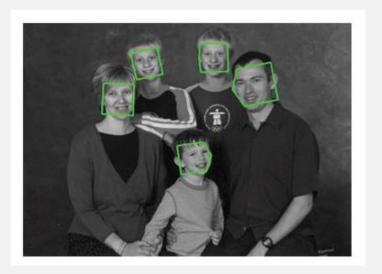


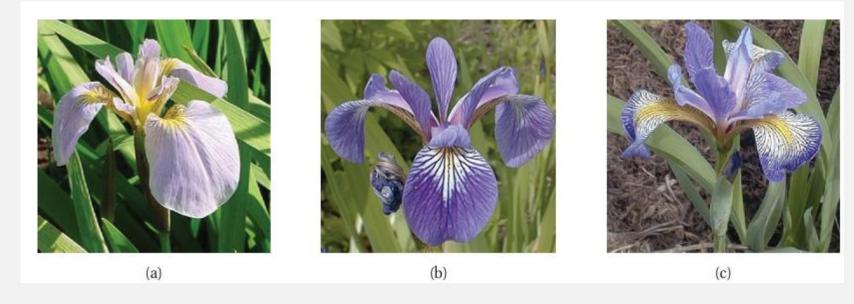
#### Classification Problem

- For the supervised problem the training data  $D = \{(x_1, y_1), (x_2, y_2), ..., (x_N, y_N)\}$  consists of inputs  $x_i \in \mathbb{R}^d$  and output  $y_i$  pair.
- In the case of classification, the output set is categorical i.e.  $y_i \in \{1,2,\ldots,K\}$
- If  $y_i \in \{1,0\}$ , then the classification problem is called binary classification problem.
- If output set contains more than two class then it is called multi-class classification.











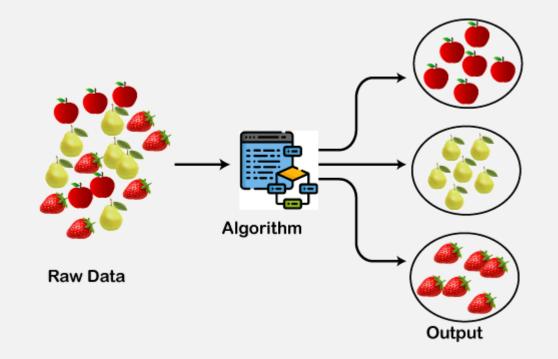
## Regression Problem

- For regression problem the output or target variable is continuous.
- Input  $x_i \in \mathbb{R}^d$  and the output  $y_i \in \mathbb{R}$ .
- Stock price prediction
- Temperature prediction



# Unsupervised Learning

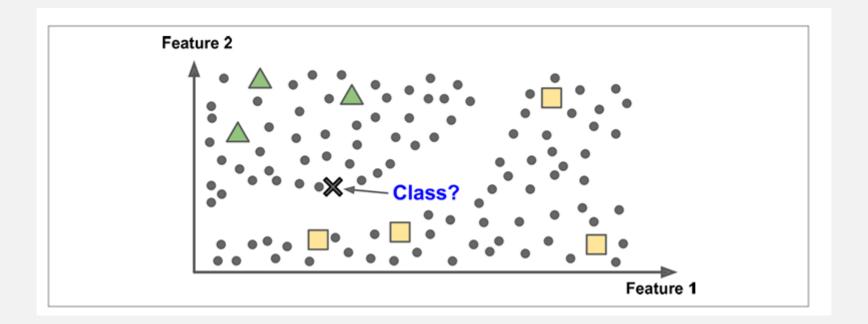
 The training data is unlabeled. The system tries to learn without a teacher.





## Semi-supervised Learning

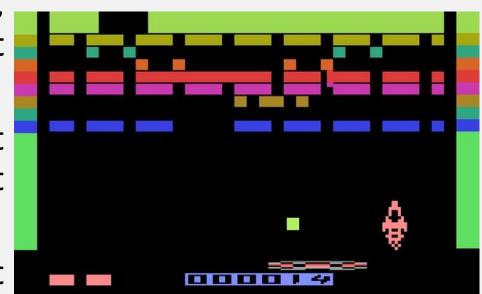
 Some algorithms can deal with partially labeled training data, usually a lot of unlabeled data and a little bit of labeled data. This is called semisupervised learning.





# Reinforcement Learning

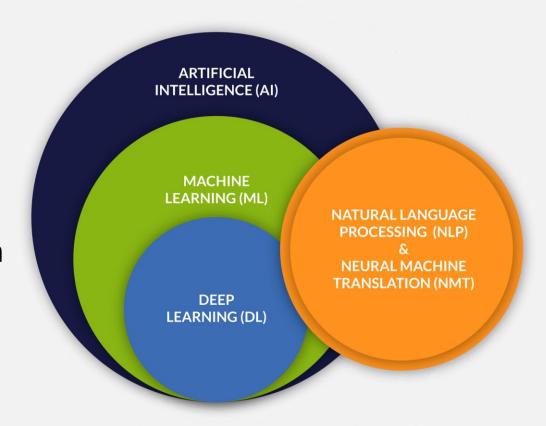
- The learning system, called an agent in this context, can observe the environment, select and perform actions, and get rewards in return.
- It must then learn by itself what is the best strategy, called a policy, to get the most reward over time.
- A policy defines what action the agent should choose when it is in a given situation.





## AI, ML, and DL

- Artificial intelligence, or AI, is technology that enables computers and machines to simulate human intelligence and problem-solving capabilities.
- Machine Learning is a subset of artificial intelligence that helps you build Al-driven applications.
- Deep Learning is a subset of machine learning that uses vast volumes of data and complex algorithms to train a model.



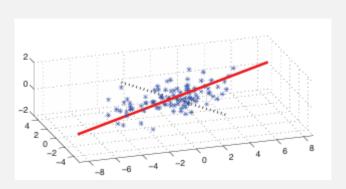


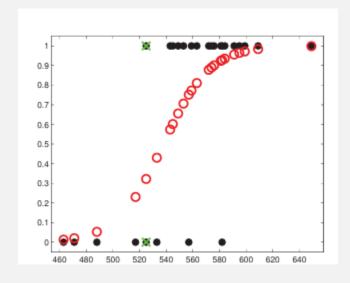
#### Model as a function

• A predictor is a function that, when given a particular input example, produces an output. For simplicity we are considering here that the output is a single number i.e. a real number or scalar value.

• 
$$f(x): \mathbb{R}^{N \times d} \to \mathbb{R}$$

• 
$$f(x) = \beta_0 + \beta x$$







# Model as Probability Distribution

- Machine learning techniques adopt a probabilistic approach, treating the target variable as a distribution of possible functions rather than a single function.
- $P(y|x) = P(y|f(x)) = P(y|x,\beta)$



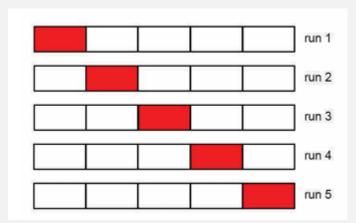
## Learning in Machine Learning

"A computer is said to learn from experience E with respect to some class of task T and performance measure P, if its performance at tasks in T, as measured by P, improves with experience E."



#### Model Selection

- When we have a variety of models of different complexity, how should we pick the right one?
- The most conventional method is to select the model with minimum error.
- Train-Test-Validation Split
- Cross Validation





#### Reference

- Machine Learning-A Probabilistic Perspective-Kelvin P. Murphy
- Hands-on Machine Learning with Scikit-Learn, Keras and Tensorflow-Aurelien Geron