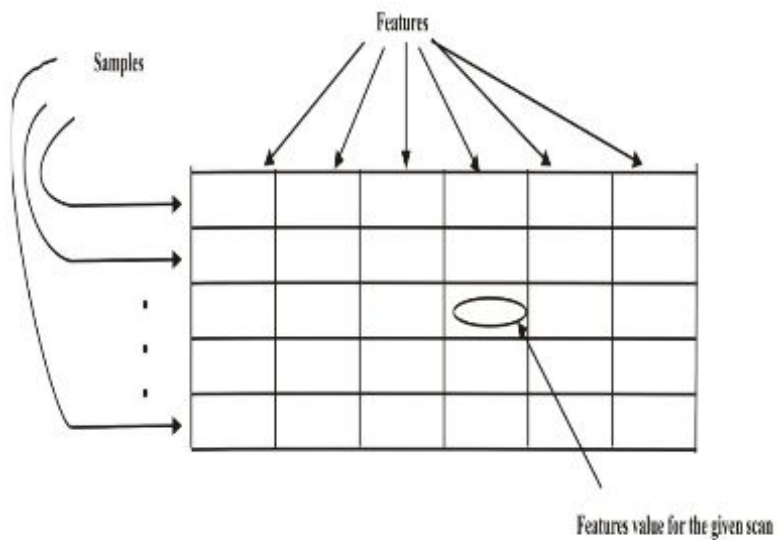
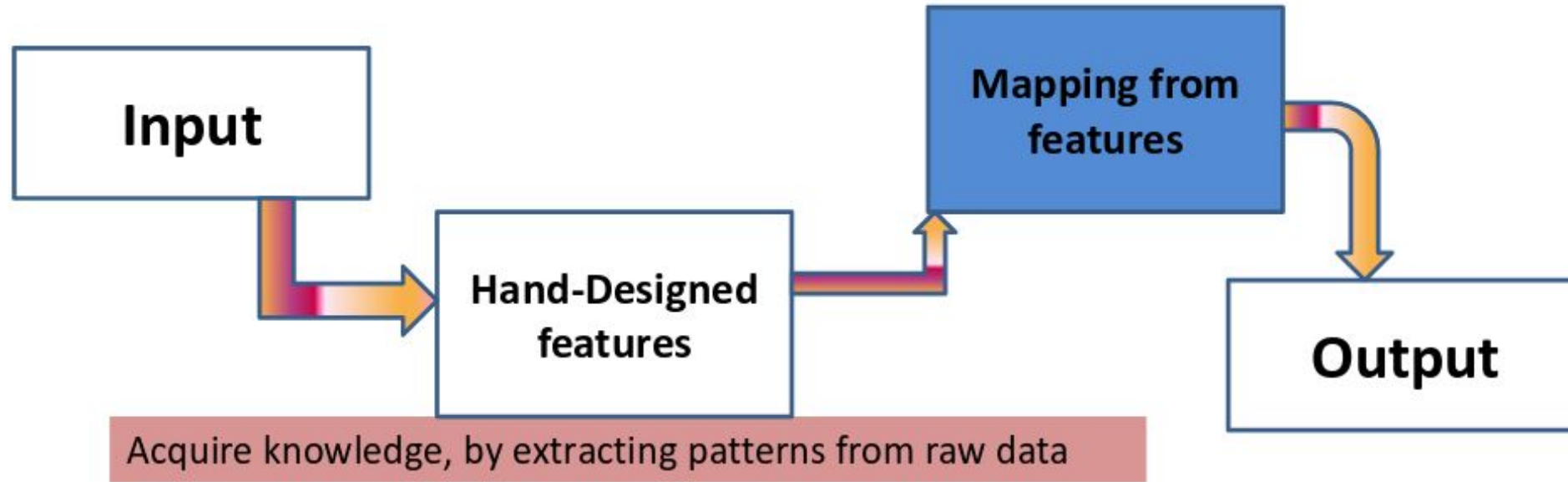


MACHINE LEARNING

What is Machine Learning?

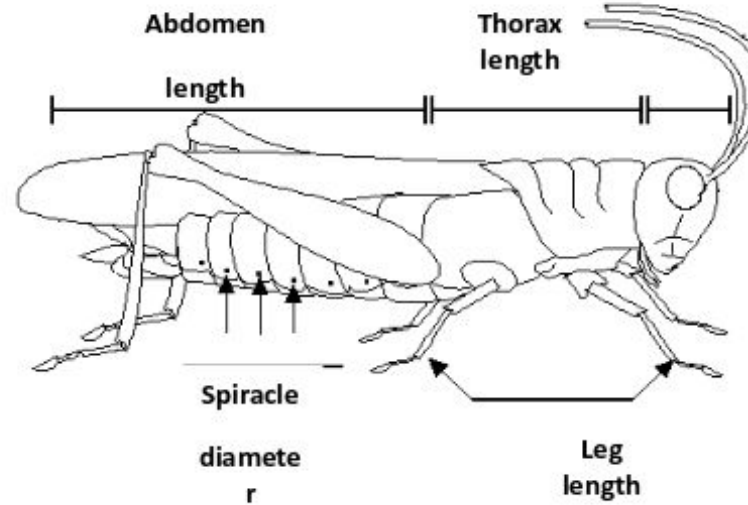
- ❑ A study of algorithms that gives the computers the ability to learn without being explicitly programmed.
- ❑ Reduces human/machine efforts required to perform a task (time optimization).
- ❑ Increases the performance of a task (efficiency optimization).

Classical machine learning



Colour {Green, Brown, Gray, Other}

Has wings?



Challenges in ML

High Dimensional Classification

CalTech 101

Anchor



Joshua Tree



Beaver



Lotus

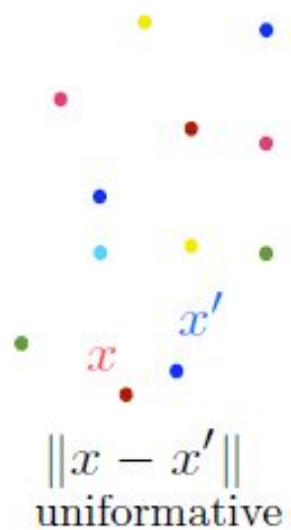


Water Lily



- Considerable variability in each class.
- Euclidean distances are meaningless.
- Need to find Informative Invariants.

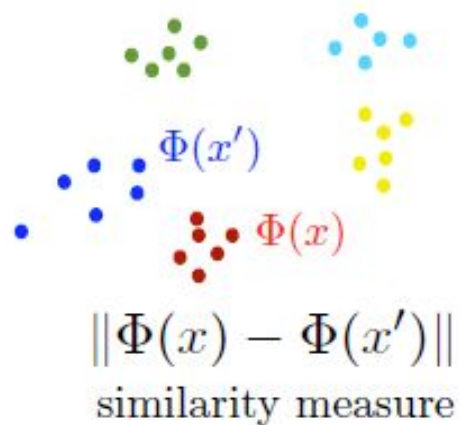
High dimension



Representation
An Art

Φ

- MFCC (audio)
- SIFT (images)
- Gauss. Mixt. Mod.
- Clustering
- Histograms
- Bag of words (parts)
- Deep Neural Networks



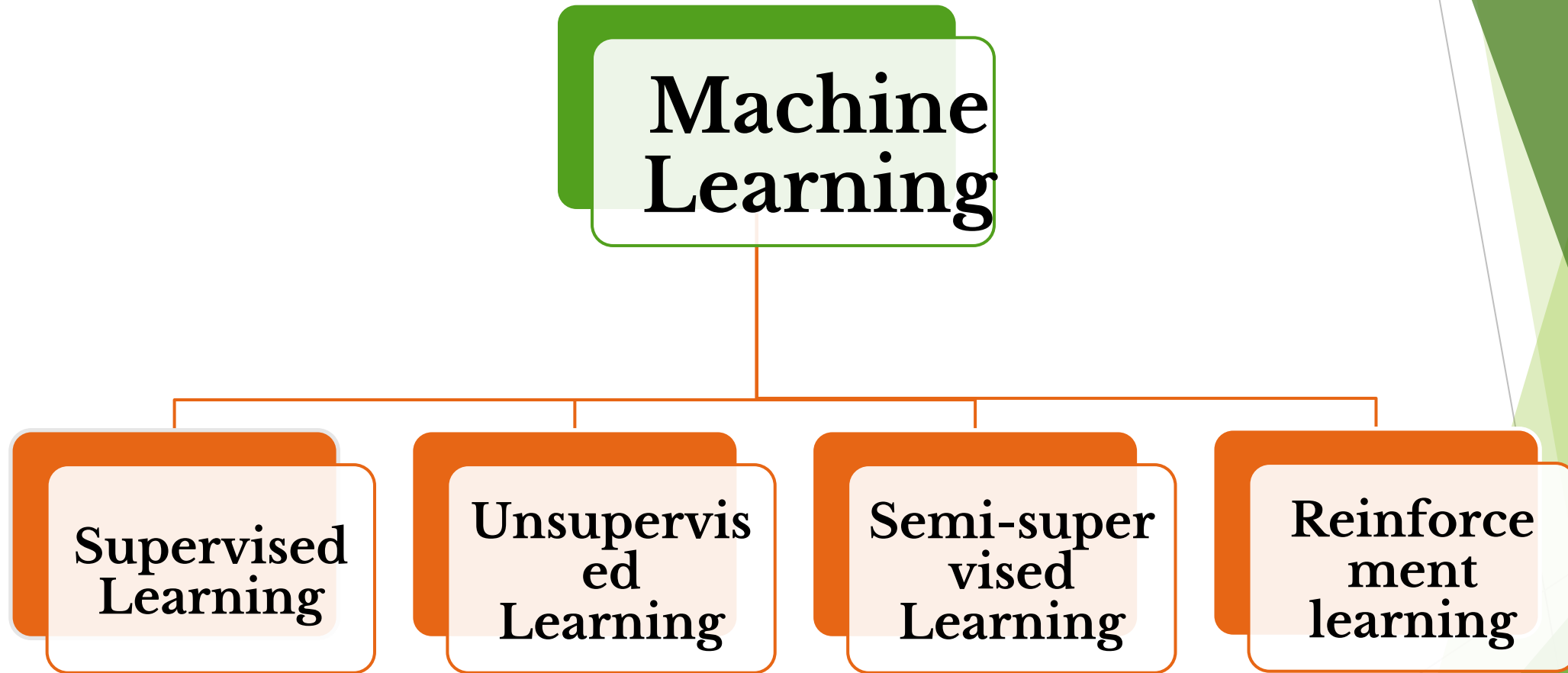
Statistical
Decision Theory

Classifier
 $\longrightarrow y$

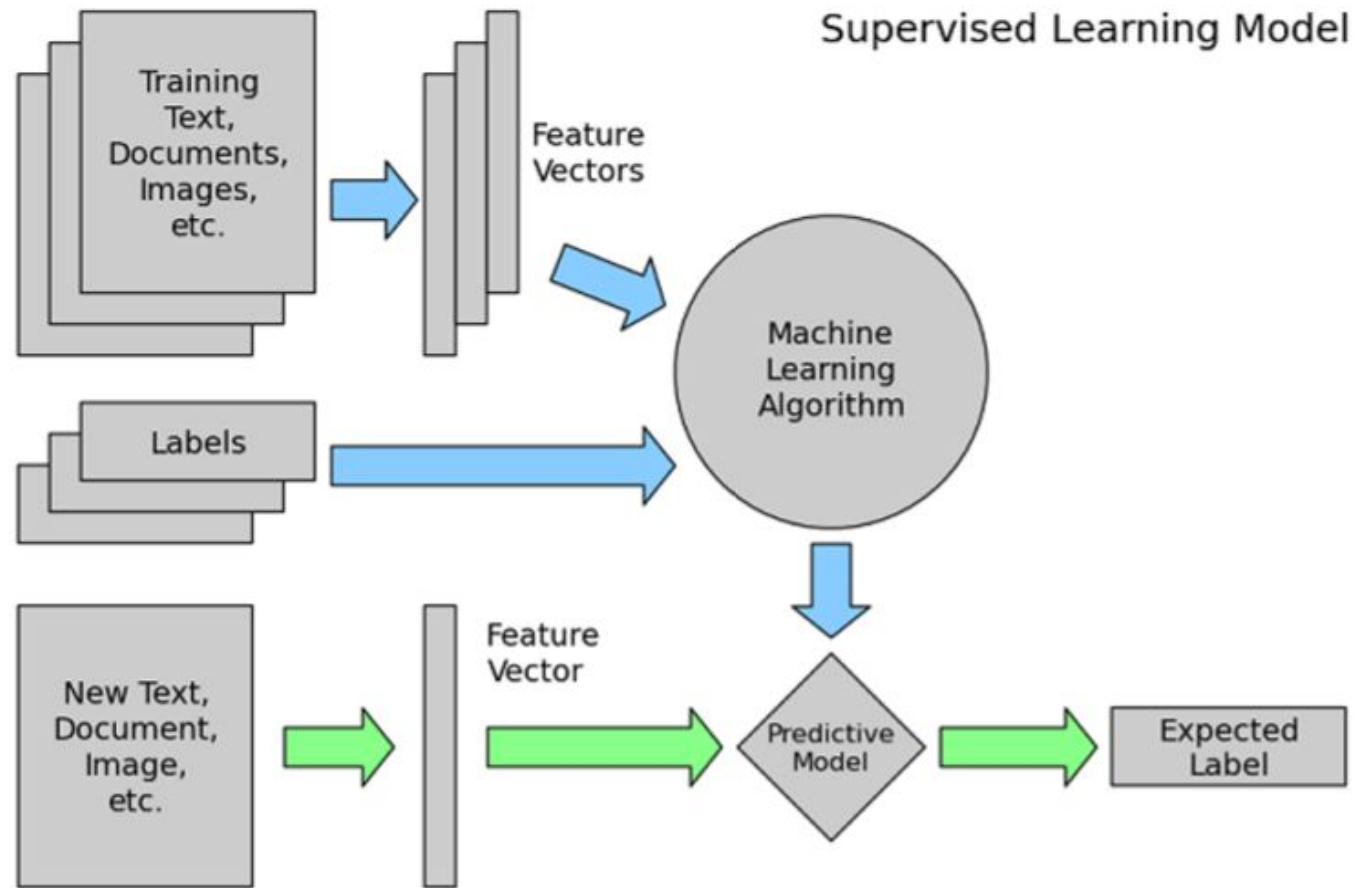
Steps in Machine Learning

- ❑ Collecting data
- ❑ Preparing the data
- ❑ Training a model
- ❑ Evaluating the model
- ❑ Improving the performance

Types of ML ...



Supervised Learning



Regression and Classification



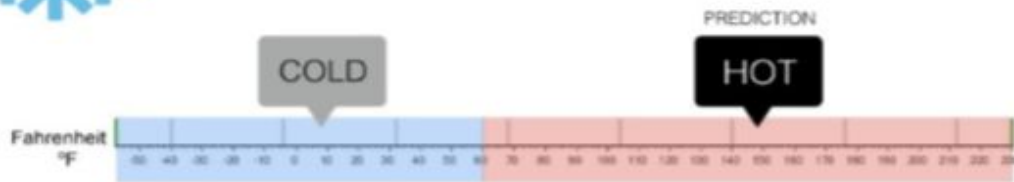
Regression

What is the temperature going to be tomorrow?

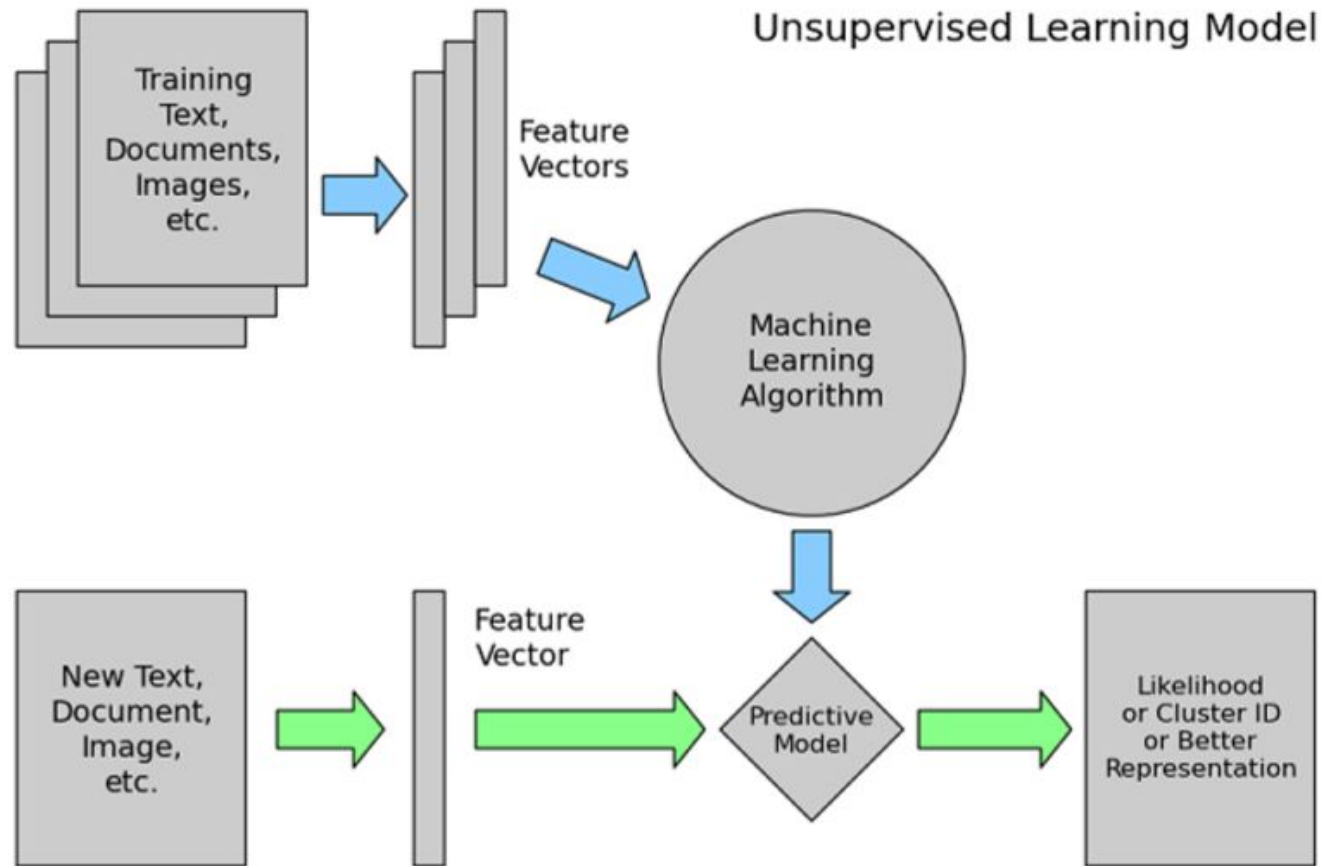


Classification

Will it be Cold or Hot tomorrow?



Unsupervised Learning



Clustering

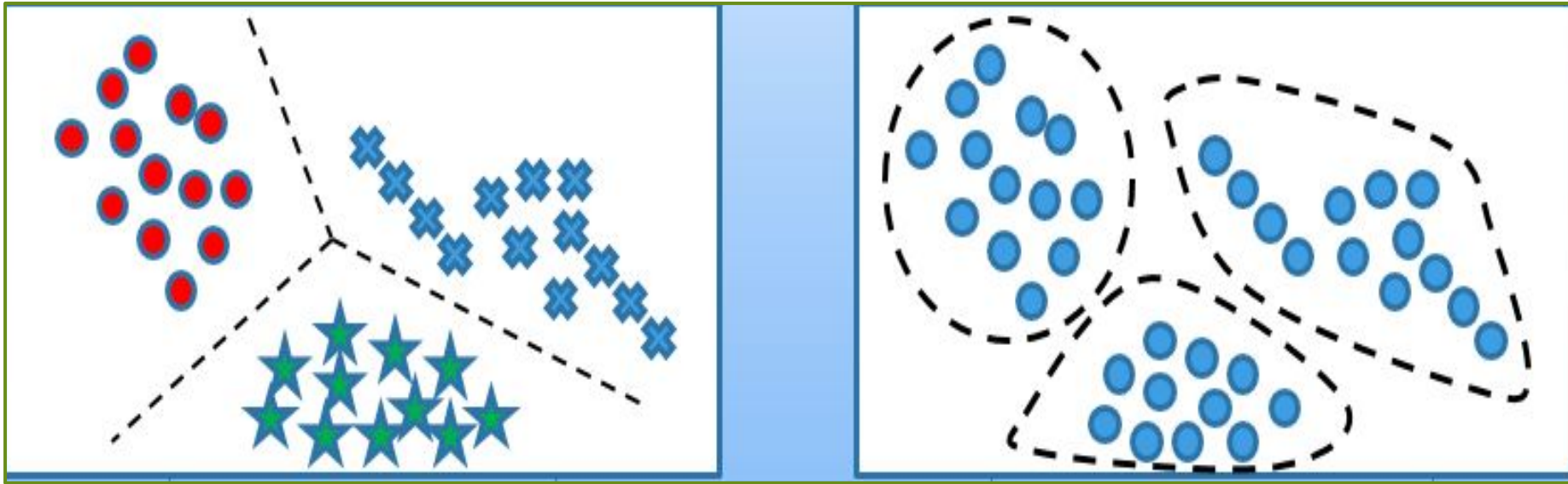
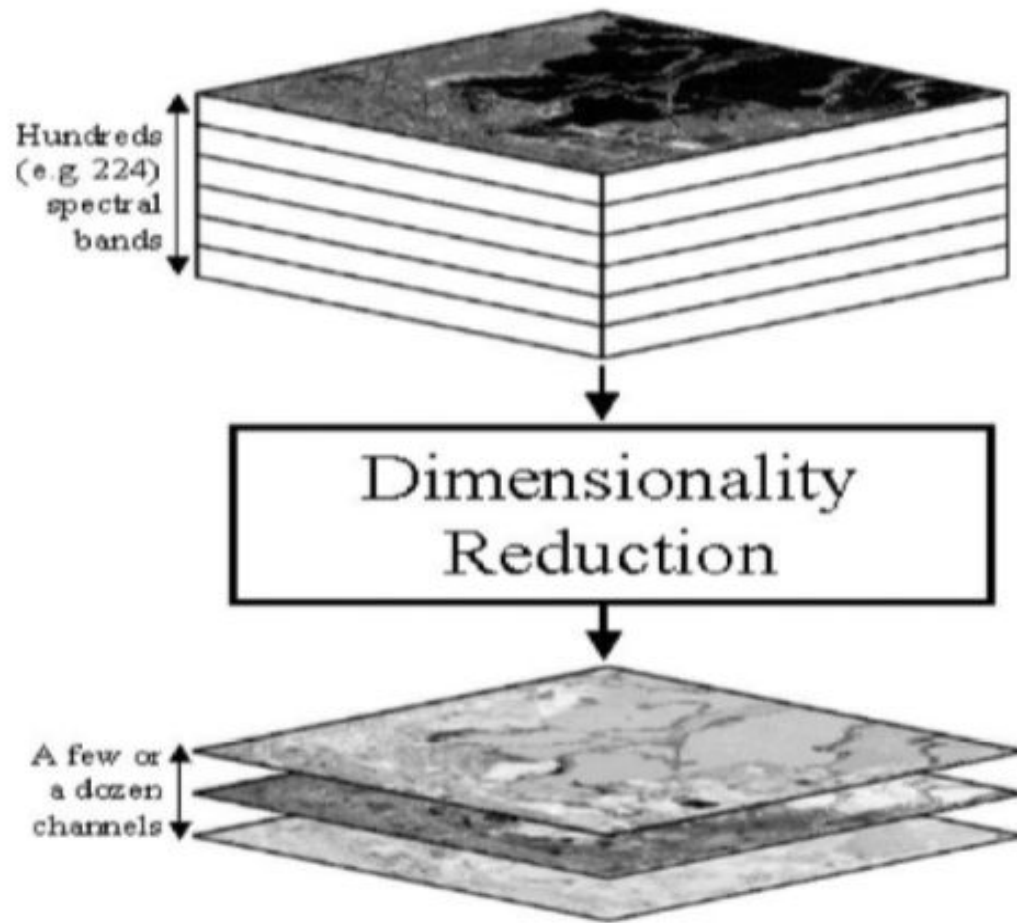


Fig.(a) Supervised learning
Clustering

Fig.(b)

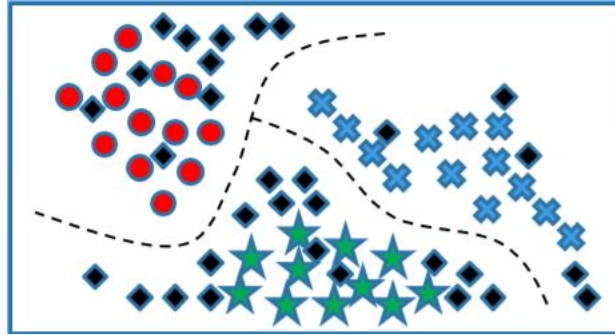
Dimensionality Reduction



source: [http:](http://spie.org/newsroom/3560-dimensionality-reduction-of-multidimensional-satellite-imagery?SS0=1)

[//spie.org/newsroom/3560-dimensionality-reduction-of-multidimensional-satellite-imagery?SS0=1](http://spie.org/newsroom/3560-dimensionality-reduction-of-multidimensional-satellite-imagery?SS0=1)

Semi-supervised Learning & Reinforcement Learning



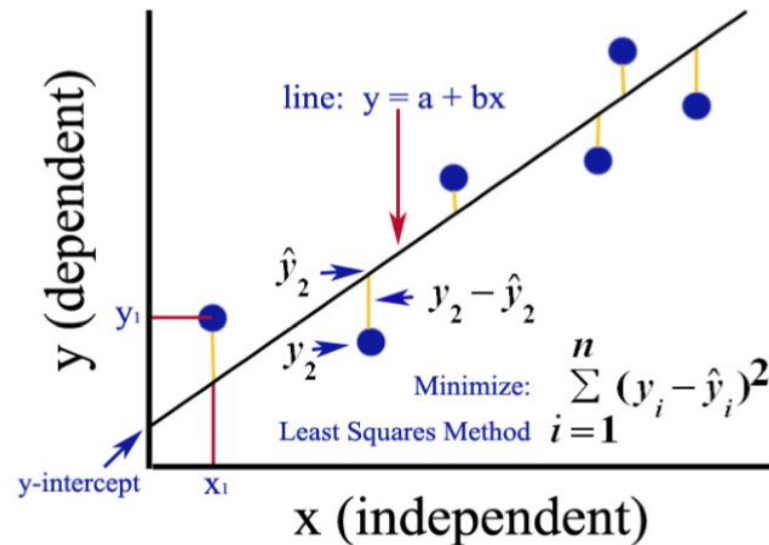
Semi-supervised Learning

- **Reinforcement learning** deals with Decision making (robot, chess machine)

Supervised Learning Techniques

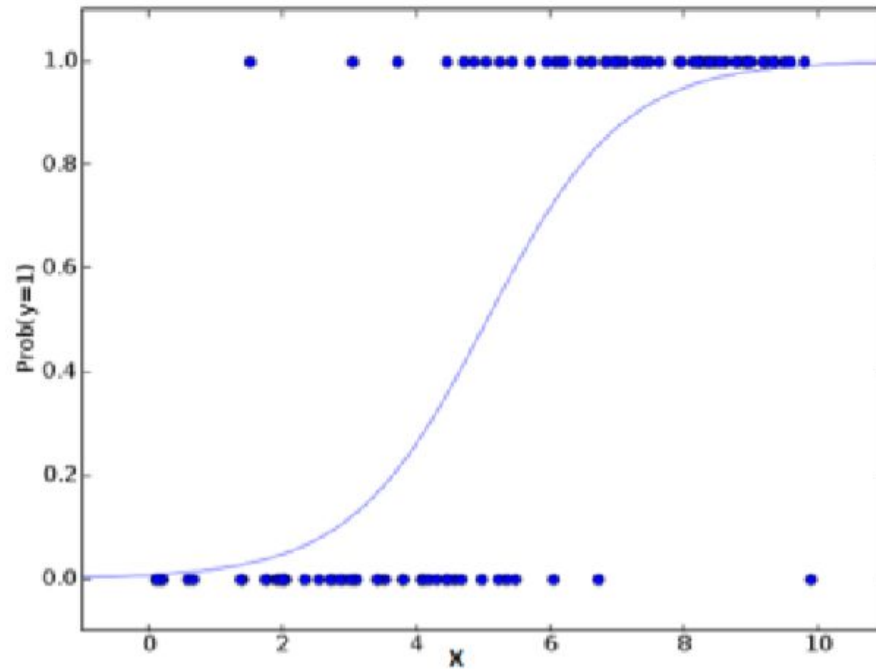
LINEAR REGRESSION

- The core idea is to obtain a line/plane that best fits the data (continuous data).
- Linear regression models a linear relationship between input and output.
- The best fit line/plane is the one for which total prediction errors of all data points are as small as possible.
- Error is the distance between the point to the regression line.

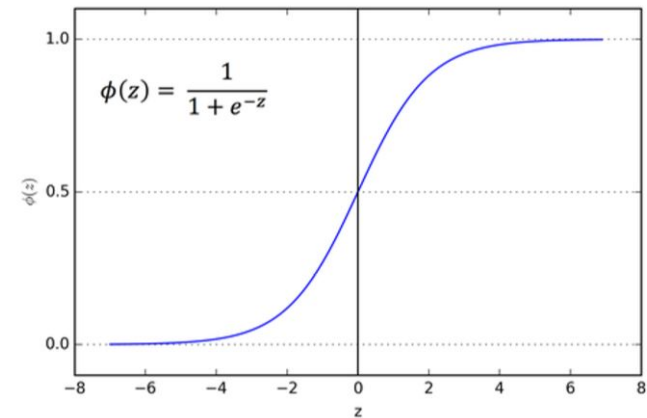


LOGISTIC REGRESSION

- It's a classification algorithm, that is used where the response variable is *categorical* (discrete).
- The idea of Logistic Regression is to find a relationship between features and probability of particular outcome.



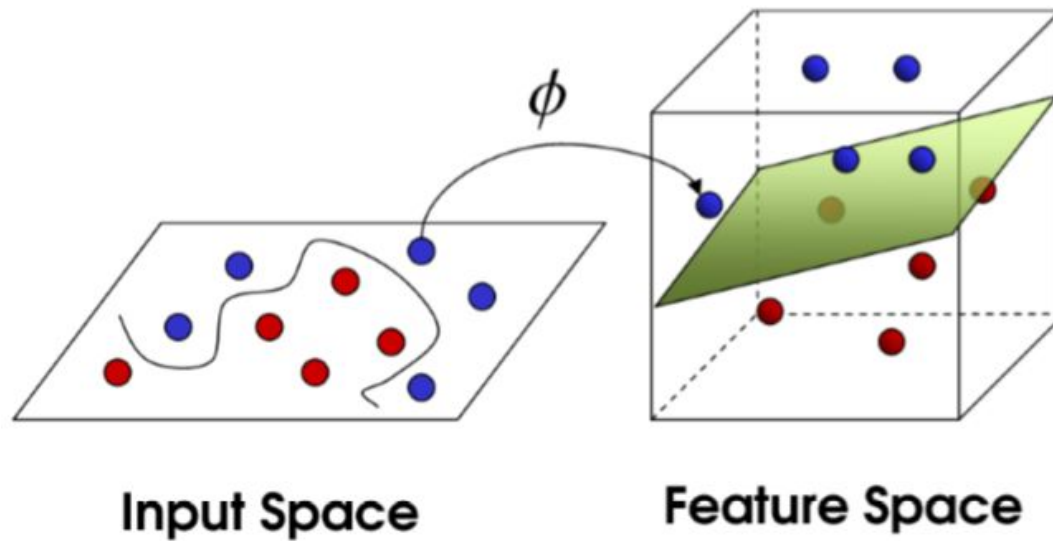
source: solutions4statistics.com



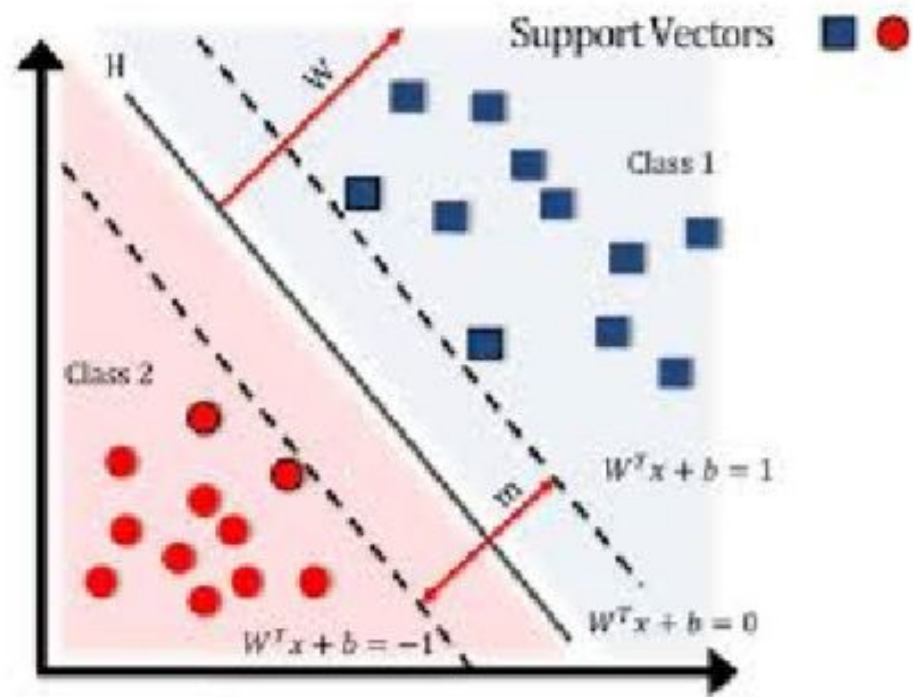
<https://sebastianraschka.com/images/faq/logisticregr-neuralnet/sigmoid.png>

SUPPORT VECTOR MACHINES (SVM)

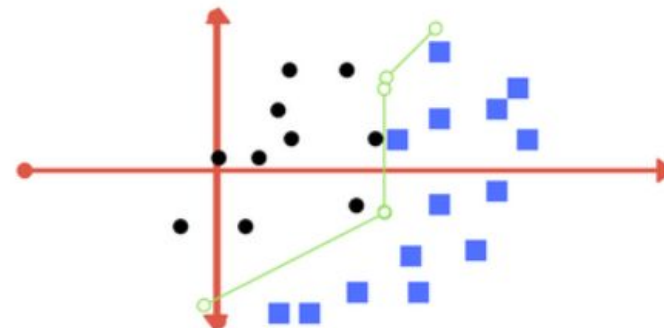
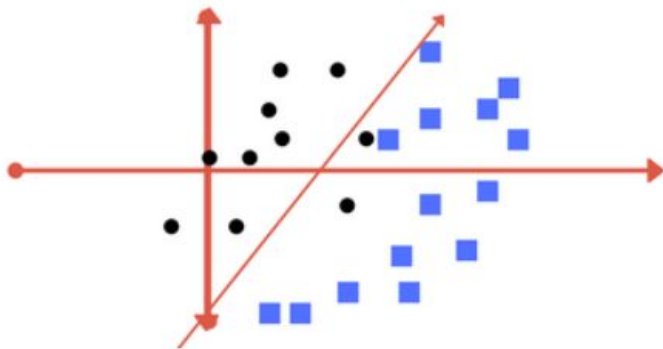
- Discriminative classifier formally defined by a separating hyperplane



source: <https://www.linkedin.com/pulse/support-vector-machine-srinivas-kulkarni/>



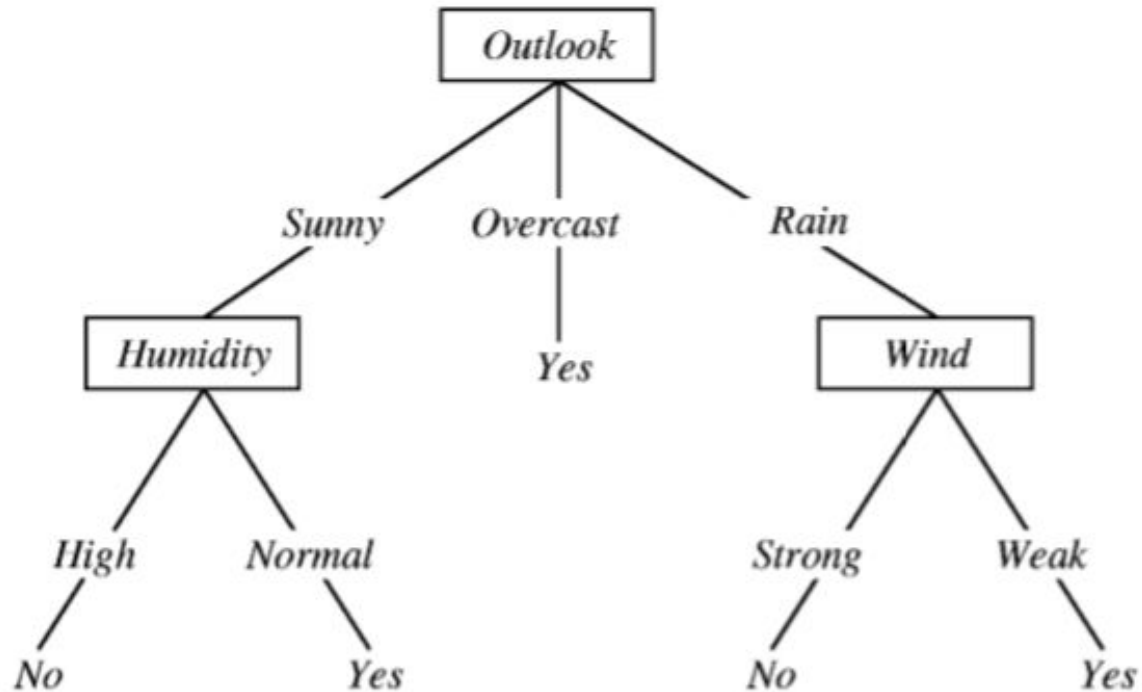
How SVM works?



Left: low regularization value, right: high regularization value

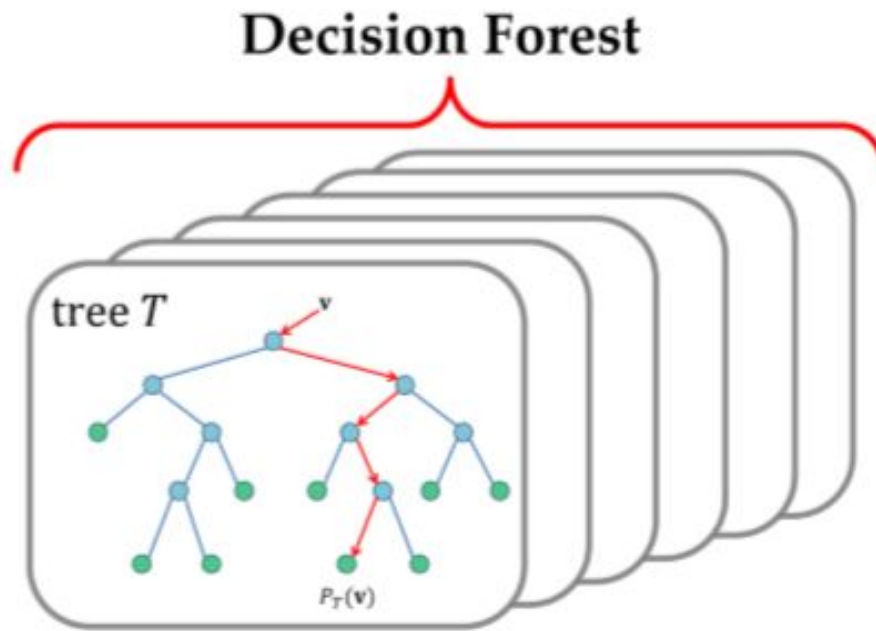
DECISION TREES

- A decision tree is a binary tree.
- Each node partitions data according to a splitting rule.
- Leaf node returns a label for the classification.



RANDOM FOREST TREES

- ▶ Random forest builds multiple decision trees and merges them together to get a more accurate and stable prediction.



NAÏVE BAYES

- Simple "probabilistic classifiers" based on applying Bayes' theorem with strong (naive) independent assumptions between the features.

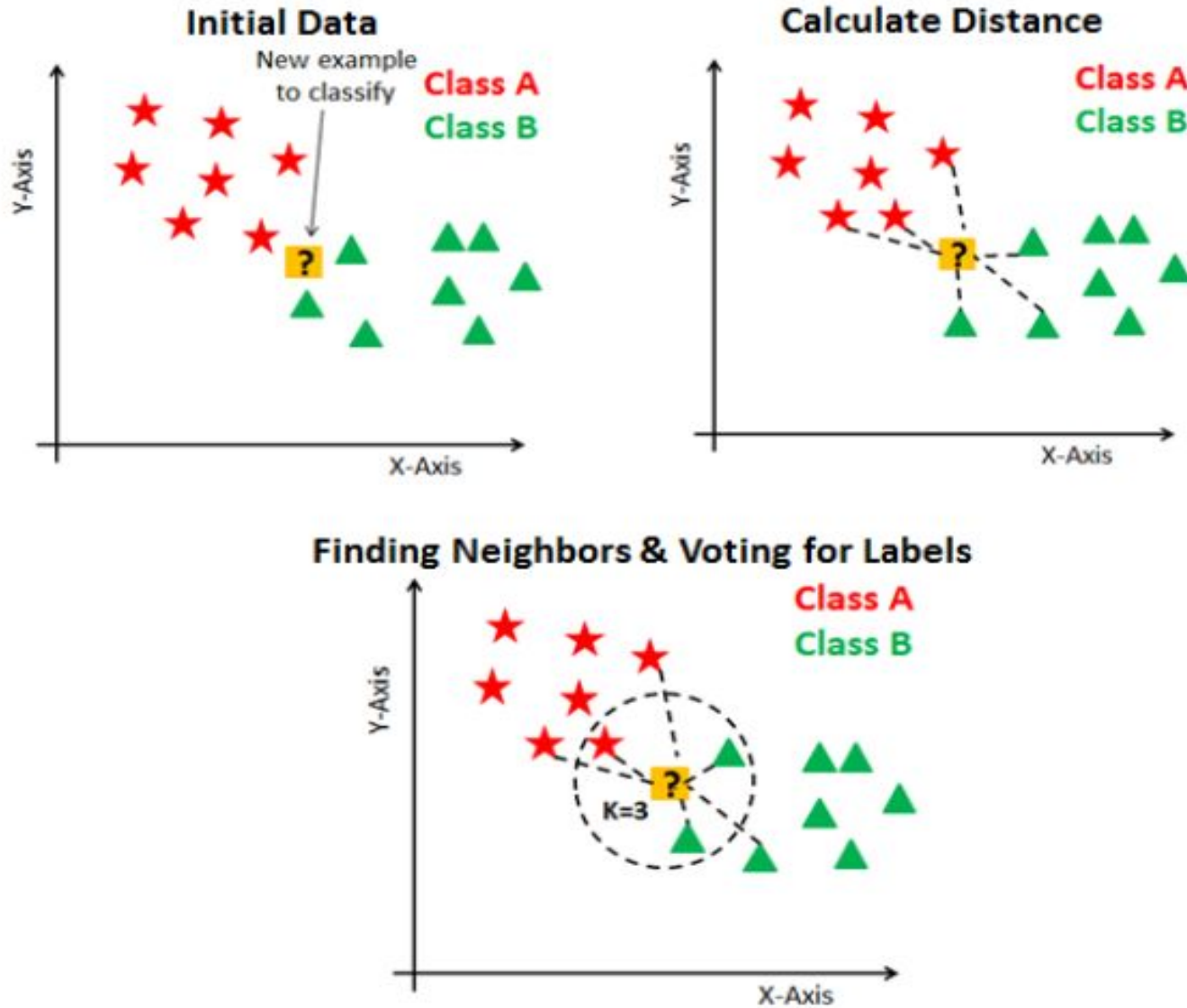
$$P(c | x) = \frac{P(x | c)P(c)}{P(x)}$$

Diagram illustrating the components of Bayes' theorem for Naïve Bayes:

- $P(c | x)$ is labeled **Posterior Probability** (indicated by a downward arrow).
- $P(x | c)$ is labeled **Likelihood** (indicated by an upward arrow).
- $P(c)$ is labeled **Class Prior Probability** (indicated by an upward arrow).
- $P(x)$ is labeled **Predictor Prior Probability** (indicated by a downward arrow).

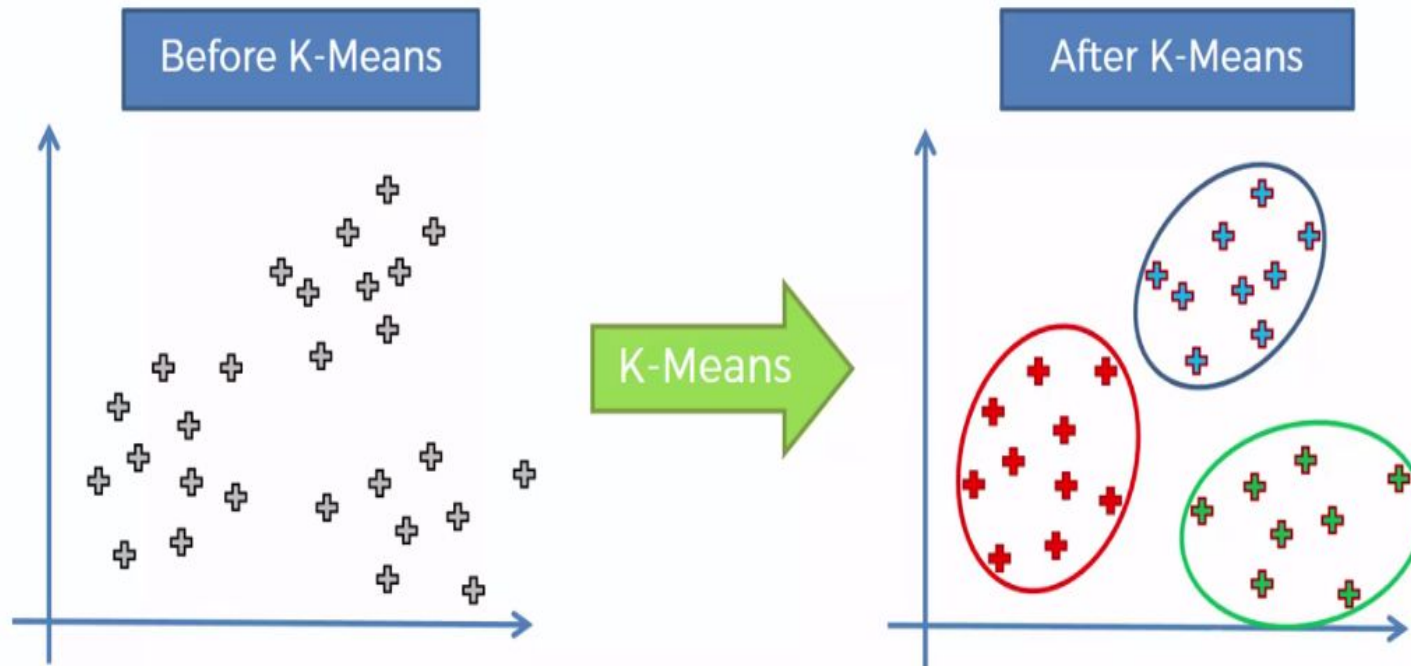
$$P(c | X) = P(x_1 | c) \times P(x_2 | c) \times \dots \times P(x_n | c) \times P(c)$$

K Nearest Neighbors (KNN)



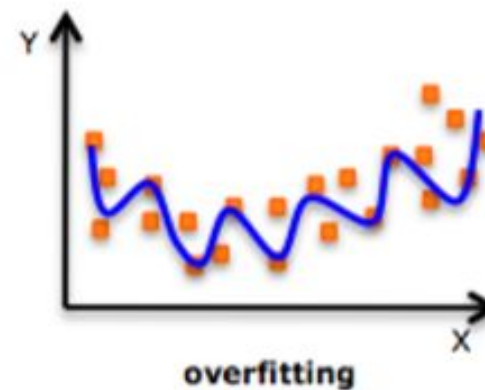
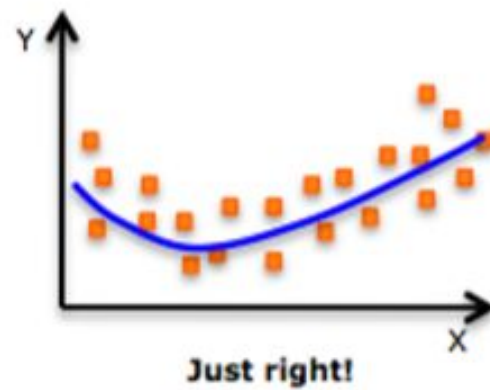
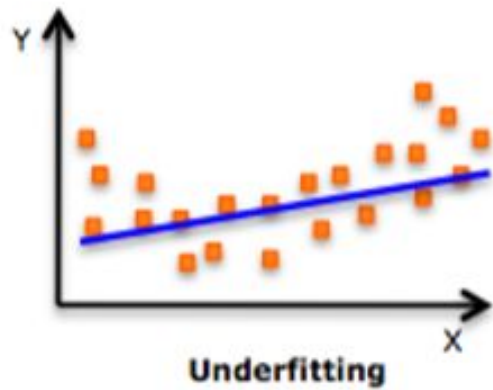
Unsupervised Learning Techniques

K-Means Clustering



Model Evaluation Techniques

❑ Fitting of Data



❑ Performance Metrics

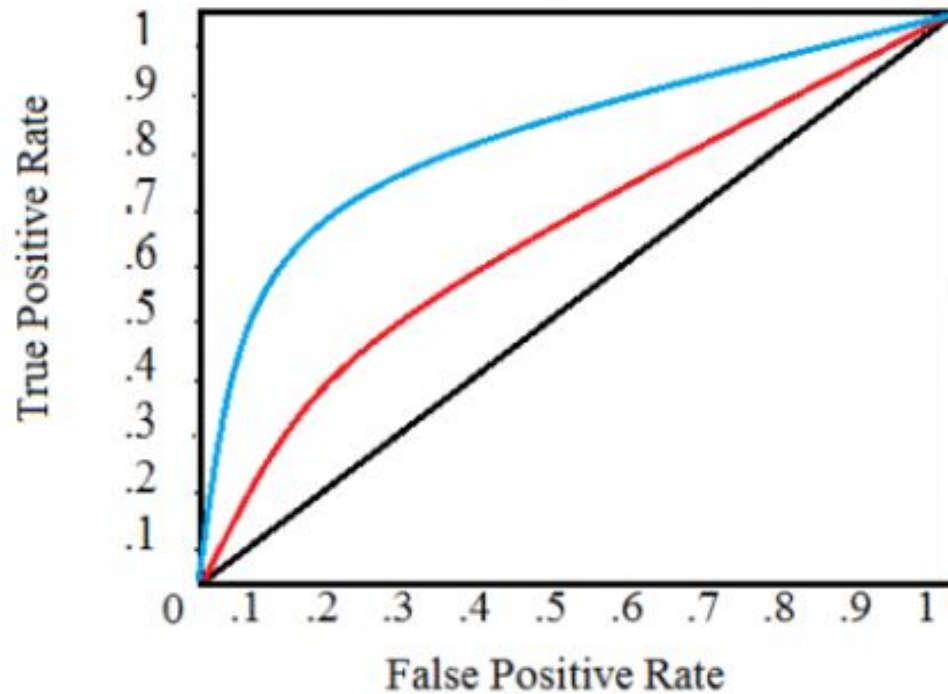
		Actual	
		Positive	Negative
Predicted	Positive	True Positive	False Positive
	Negative	False Negative	True Negative

$$\text{Precision} = \frac{\text{True Positive}}{\text{Actual Results}} \quad \text{or} \quad \frac{\text{True Positive}}{\text{True Positive} + \text{False Positive}}$$

$$\text{Recall} = \frac{\text{True Positive}}{\text{Predicted Results}} \quad \text{or} \quad \frac{\text{True Positive}}{\text{True Positive} + \text{False Negative}}$$

$$\text{Accuracy} = \frac{\text{True Positive} + \text{True Negative}}{\text{Total}}$$

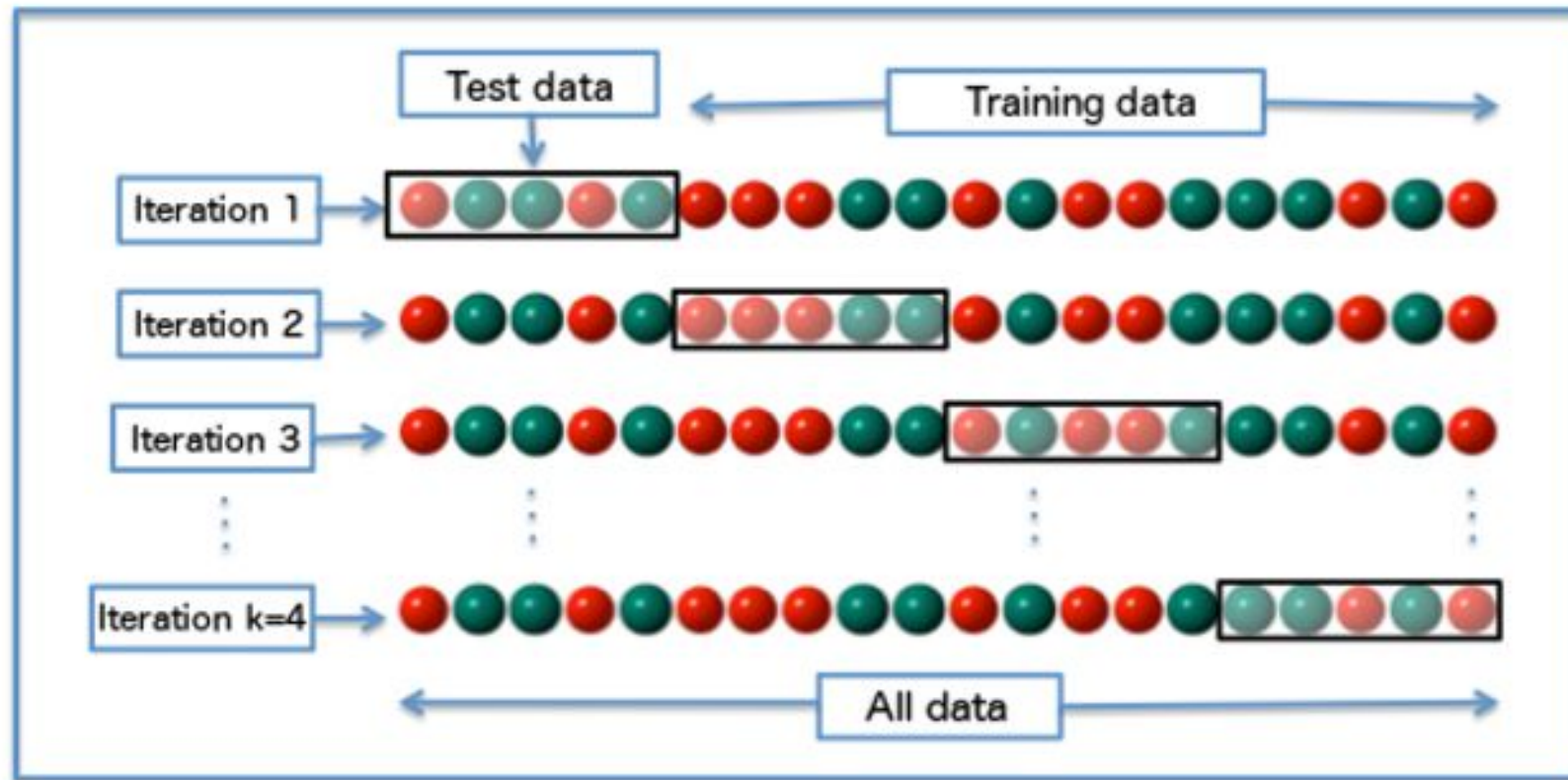
$$\text{F1 Score} = 2 * \frac{\text{Precision} * \text{Recall}}{\text{Precision} + \text{Recall}}$$



Receiver Operating
Characteristics

Performance Enhancement

► Cross Validation



*Thank
you*