

# Conclusion

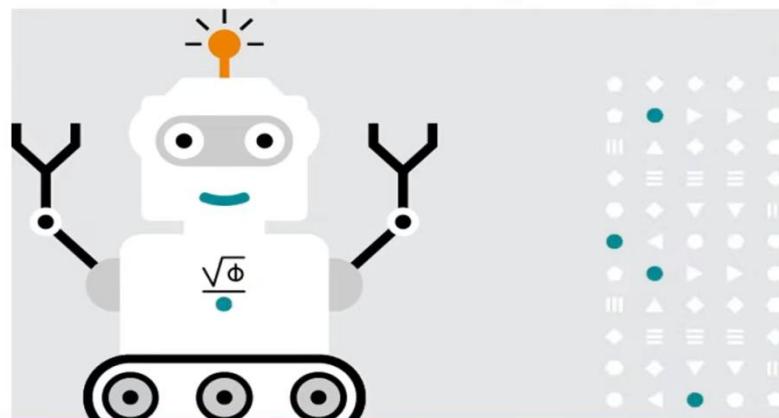
- ❑ In Supervised learning, you train the machine using data which is well "labeled."
- ❑ Regression and Classification are two types of supervised machine learning techniques.
- ❑ The biggest challenge in supervised learning is that if your training set doesn't have examples that you want to have in a class, could give inaccurate results.



The screenshot shows a video player interface. At the top, it displays the title "Machine Learning Full Course with Practical (6 Hours) | Become a Machine Learning Engineer in 2023" with a fire emoji. On the right side of the title bar are icons for a clock and a share button. Below the title, the main content area has a dark blue header with the text "What is Unsupervised Machine Learning". The main body of the video features a purple background with a 3D rendering of a laptop displaying binary code (010, 011) and a network of lines and dots, representing data and connections. The WsCUBE TECH logo is visible in the bottom right corner of the video frame. At the very bottom of the screen, there is a dark footer bar with video control icons (play, pause, volume) and the text "2:51:11 / 6:45:57 • What is Unsupervised Machine Learning? Association & Clustering Algorithms in Machine Learning >".

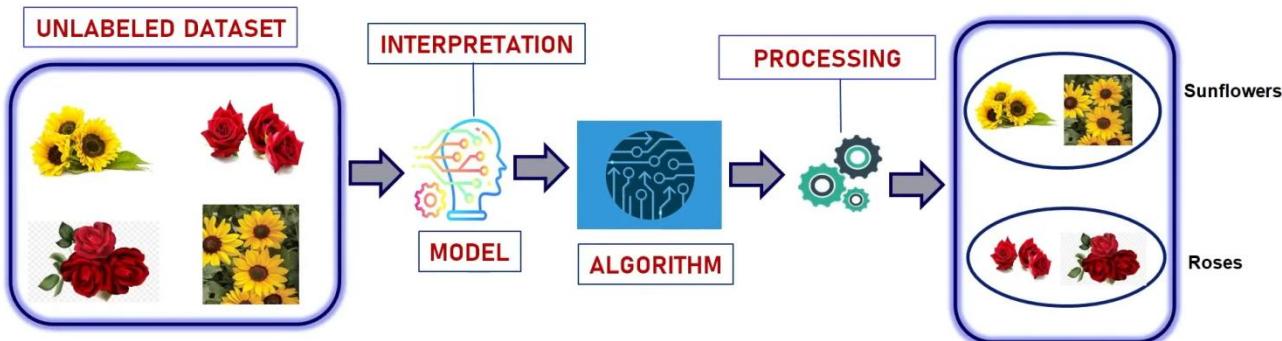
# What is Unsupervised Learning?

"In unsupervised learning, you train the machine using data which is "unlabeled" and models itself find the hidden patterns and insights from the given data."



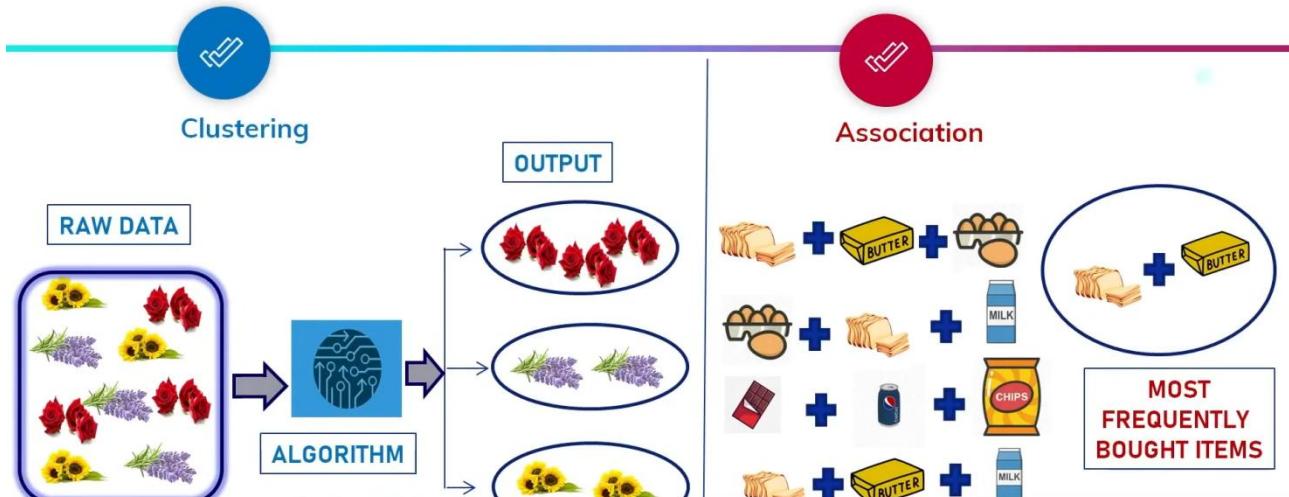
## How Does Unsupervised Learning Works?

The goal of Unsupervised Learning is to group unlabelled data according to the similarities, patterns and differences without any prior training of data



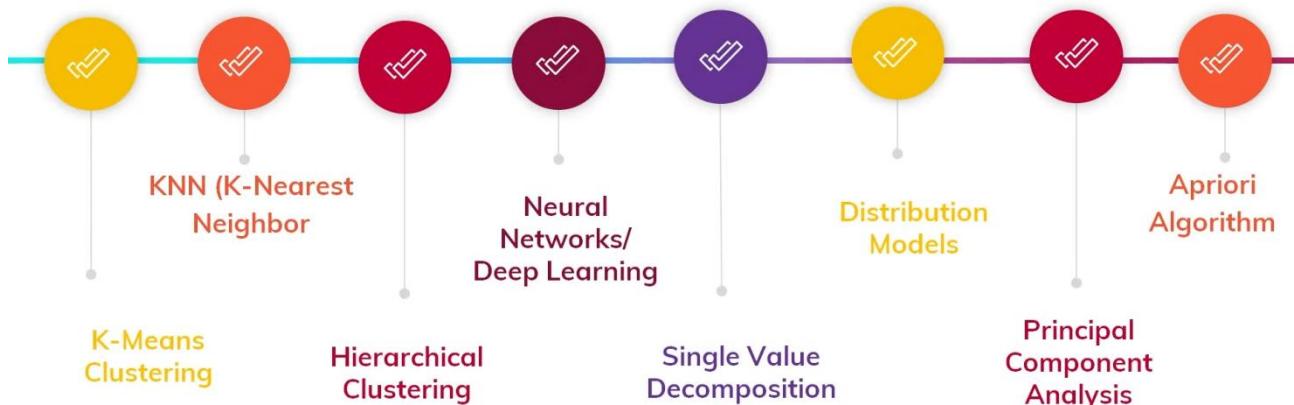
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## Types of Unsupervised ML Algorithms



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# Unsupervised Learning Algorithms



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## Advantages of Unsupervised Learning

-  Unsupervised learning is used for more complex tasks
-  It's helpful in finding patterns in data
-  Saves lot of manual work and expense

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## Disadvantages of Unsupervised Learning

-  Less Accuracy
-  Time Consuming
-  More the features, More the Complexity

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# Conclusion

- ❑ In Unsupervised learning, you train the machine using data which is “unlabeled.”
- ❑ Clustering and Association are two types of unsupervised machine learning techniques.
- ❑ The biggest drawback in unsupervised learning is that it might result in less accuracy.

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## Training & Testing Data in Machine Learning

### Term Insurance Dataset

	Age	Premium
Training Data (80%)	25	18000
	30	32000
	35	42000
	40	47000
	45	55000
Testing Data (20%)		

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Time – 3.19

# Linear Regression in Machine Learning

## What is Linear Regression?

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System For Satisfaction

Linear Regression is a machine learning algorithm based on supervised learning.

It is a statistical method that is used for predictive analysis. Linear regression makes predictions for continuous/real or numeric variables such as cost, age, sales, temperature, product price, etc.

## What is Linear Regression?

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System For Satisfaction

Linear Regression is a machine learning algorithm based on supervised learning.

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Linear Regression with Single Variable



Linear Regression with Multiple Variables

# Linear Regression Real World Application

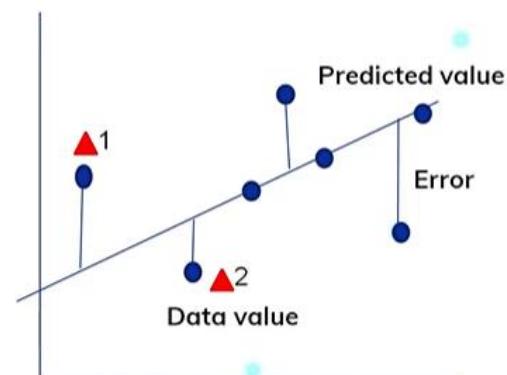
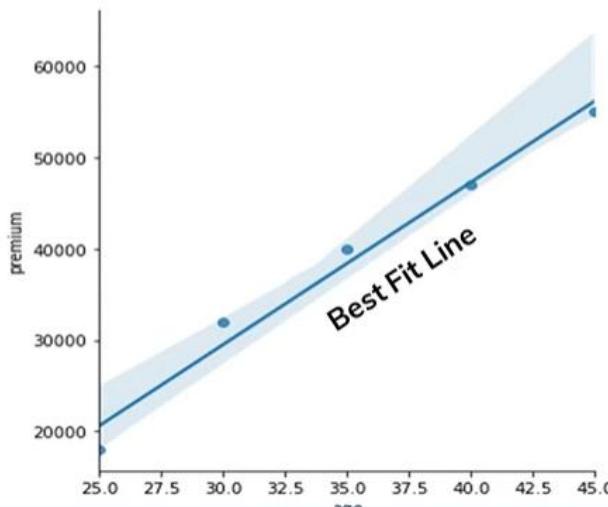


## Term Insurance Dataset

Age	Premium
25	18000
30	32000
35	42000
40	47000
45	55000

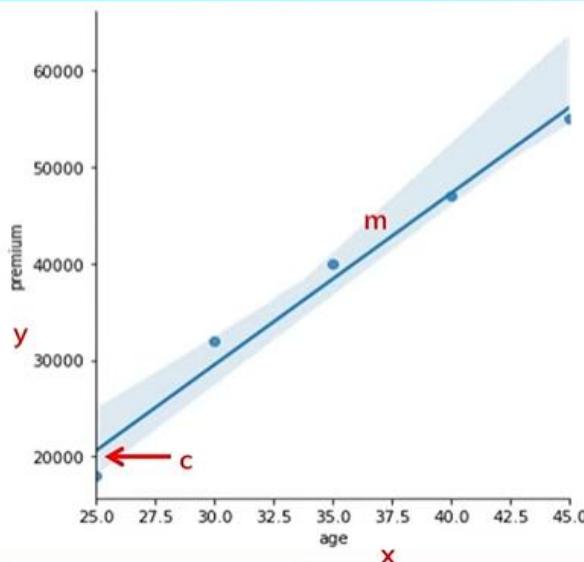
We'll find out the Premium whose age is 21 and 50

# What is Best-Fit Line?



The best fit line will have the least error.

## Independent & Dependent Variable



Linear Equation:  
 $y=mx+c$

where  
 y -dependent variable  
 x –independent variable  
 m-slope/gradient/coefficient  
 c-intercept

$$\text{premium} = m * \text{age} + c$$

**Go on Jupyter**

## Validating our Model

Linear Equation:  $y=mx+c$

$$\text{premium} = m * \text{age} + c$$

where  
 y -dependent variable  
 x –independent variable  
 m-slope/gradient/coefficient  
 c-intercept

Premium whose age is 21

$$\begin{aligned} y &= m * x + c \\ &= 1780 * 21 + (-23900) \\ &= 37,380 - 23900 \\ &= 13480 \end{aligned}$$

Premium whose age is 50

$$\begin{aligned} y &= m * x + c \\ &= 1780 * 50 + (-23900) \\ &= 89,000 - 23900 \\ &= 65100 \end{aligned}$$

# Linear Regression Multiple Variables

## What is Linear Regression?

Linear Regression is a machine learning algorithm based on supervised learning.

It is a statistical method that is used for predictive analysis. Linear regression makes predictions for continuous/real or numeric variables such as cost, age, sales, temperature, product price, etc.



Linear Regression with Single Variable



Linear Regression with Multiple Variables

# Linear Regression Real World Application



## Term Insurance Dataset

Age	height	weight	Premium
25	162.56	70	18000
30	172.72	95	38000
35	167.64	78	38000
40		110	60000
45	157.48	85	70000

We'll find out the Premium whose age is 27 height is 165.56 and weight is 60  
 whose age is 60, height is 165.10 and weight is 80

## Formulae

Linear Equation:  $y = m_1*x_1 + m_2*x_2 + m_3*x_3 + c$

$$\text{premium} = m_1 * \underset{\text{age}}{\text{age}} + m_2 * \underset{\text{height}}{\text{height}} + m_3 * \underset{\text{weight}}{\text{weight}} + c$$

Where,

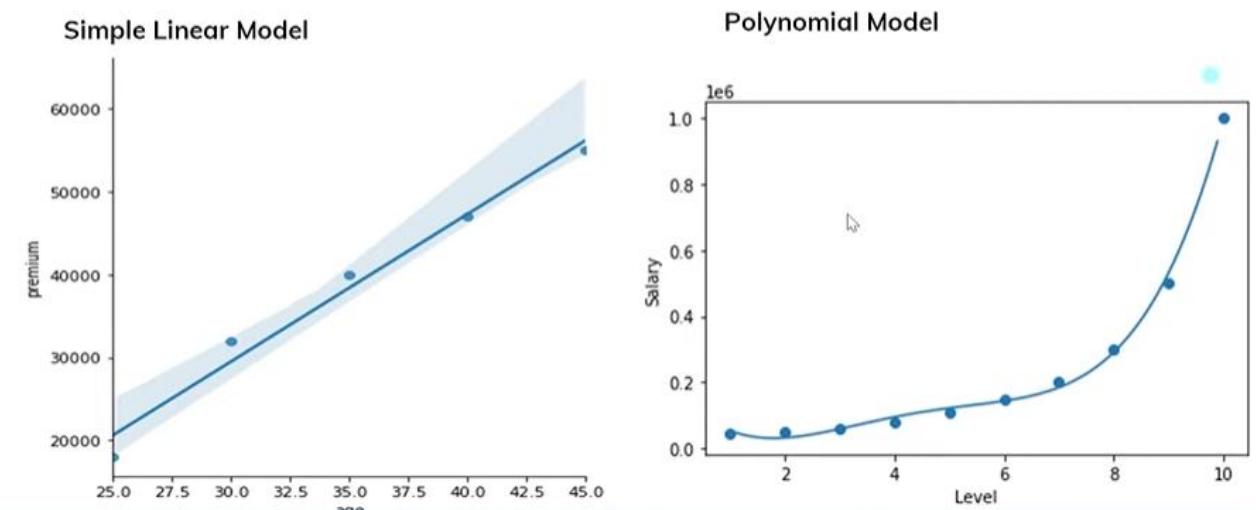
- y- dependent variable (premium)
- $x_1$  – independent variable (age)
- $x_2$  – independent variable (height)
- $x_3$  – independent variable (weight)
- $m_1, m_2, m_3$ - slope/gradient/coefficient
- c-intercept

Time – 3.57

# Machine Learning Polynomial Regression Explained



## What is Polynomial Regression?



## Degree of Polynomial

Linear Equation with Single Variable

$$y = m_1 x_1 + c$$

Linear Equation with Multiple Variables

$$y = m_1 * x_1 + m_2 * x_2 + m_3 * x_3 + c$$

### Degree of Polynomial

0 Degree Polynomial

$$y = \text{constant}$$

1 Degree Polynomial

$$y = mx^1 + c$$

2 Degree Polynomial

$$y = ax^2 + bx + c$$

### Generalized Polynomial Equation

$$y = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + \dots + a_n x^n$$

Time – 4.14

# Logistic Regression [Binary Classification]



## What is Logistics Regression?

Logistic Regression is a machine learning algorithm based on supervised learning.

It is a statistical method that is used for predicting probability of target variable.  
Logistic Regression makes probability for classification problems that are discrete in nature.

Example: English or Hindi, True or False, 1 or 0, Right or Wrong, cat or dog or goat.

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Binary Classification    Ex: win or loss, dead or alive



Multiclass Classification    Ex: Onion or Potato or Sweet Potato,  
Lily or Sunflower or Rose

# Logistic Regression Real Life Example



Heart Attack Prediction  
Tumour Prediction



Credit-Card Fraud



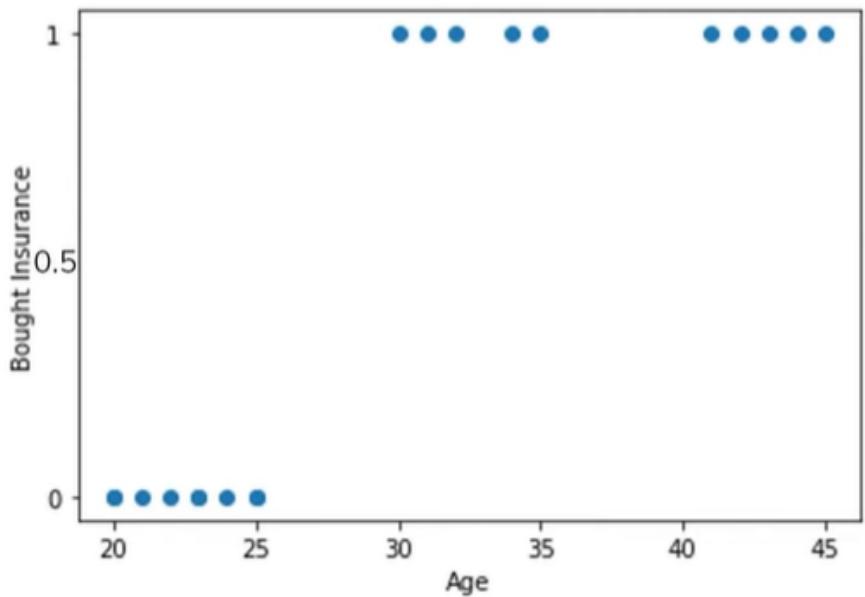
Spam Detection

## Term Insurance Dataset

Age	Bought Insurance	Bought Insurance
21	no	0
48	yes	1
32	yes	1
41	yes	1
20	no	0
35	yes	1
20	no	0
23	no	0

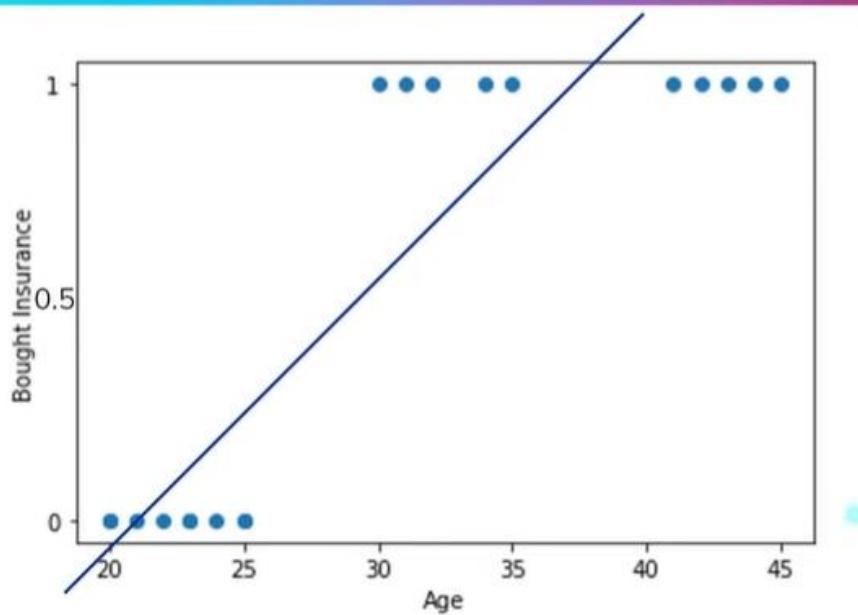
# Linear Regression Graph

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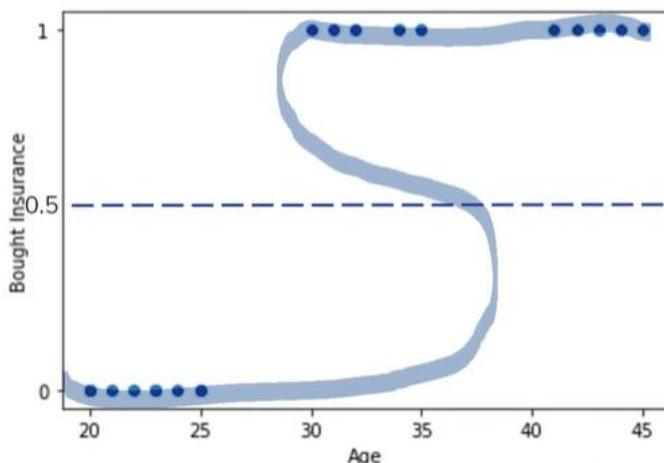


# Linear Regression Graph

---



# Logistic Regression Graph



$$y = 1 / (1 + e^{-x})$$

Sigmoid Function  
Converts input ( $x$ ) into range 0 or 1

$y$  - dependent variable (Bought Insurance)

$x$  - independent variable (age)

e - Euler's Constant ~ 2.71828

Time – 4.32

# Logistic Regression [Multiclass Classification]

# What is Logistics Regression?

Logistic Regression is a machine learning algorithm based on supervised learning.

It is a statistical method that is used for predicting probability of target variable.  
Logistic Regression makes probability for classification problems that are discrete in nature.

Example: English or Hindi, True or False, 1 or 0, Right or Wrong, cat or dog or goat.



Binary Classification    Ex: win or loss, dead or alive



Multiclass Classification Ex: Onion or Potato or Sweet Potato,  
Lily or Sunflower or Rose

## Multiclass Classification Real Life Example



Identity Types of Animal



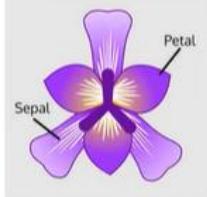
Identify Type of Vehicle



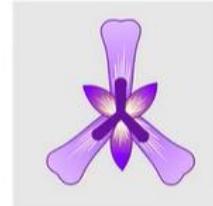
News Categorization

## Iris Dataset

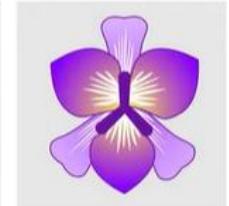
SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
6.8	3.2	5.9	2.3	Iris-virginica
6.9	3.1	5.1	2.3	Iris-virginica
4.9	3.0	1.4	0.2	Iris-setosa
5.6	3.0	4.5	1.5	Iris-versicolor
4.8	3.1	1.6	0.2	Iris-setosa
5.8	2.8	5.1	2.4	Iris-virginica
7.2	3.6	6.1	2.5	Iris-virginica
5.1	3.5	1.4	0.3	Iris-setosa
4.7	3.2	1.6	0.2	Iris-setosa
6.6	3.0	4.4	1.4	Iris-versicolor



Iris Versicolor



Iris Setosa



Iris Virginica

Time- 4.47

# Decision Tree In Machine Learning

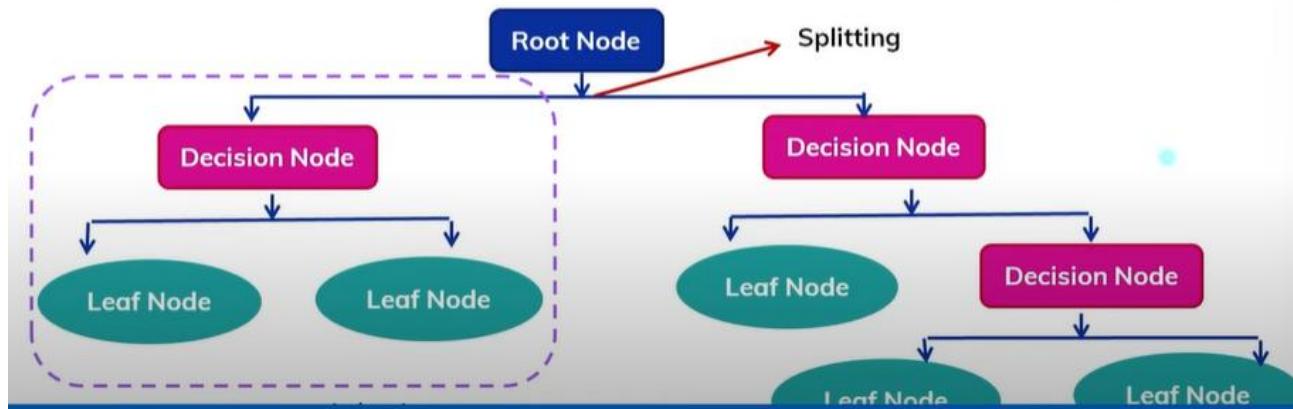
Time – 4.45

## What is Decision Tree in ML?

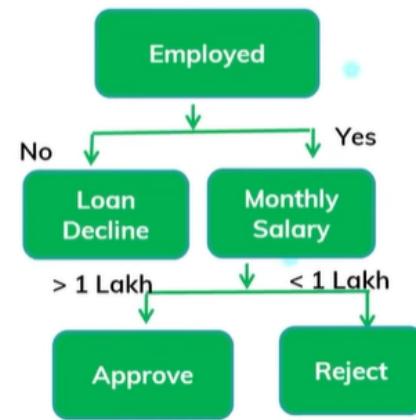
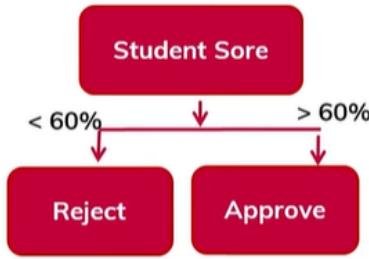
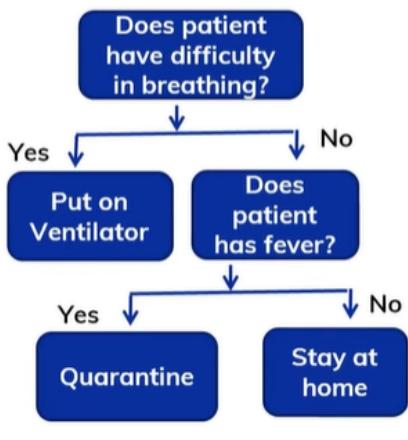


Decision Tree is a machine learning algorithm based on supervised learning, that can be used for both regression and classification problems.

The goal is to build up a classifier model that can predict the class or value of the target variable.

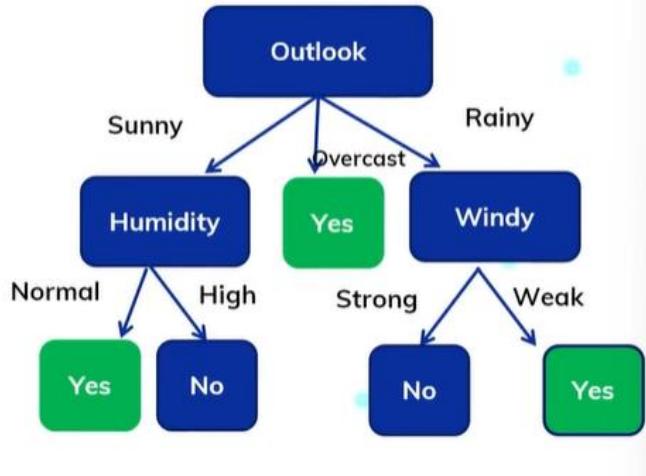


# Decision Tree Real Life Example

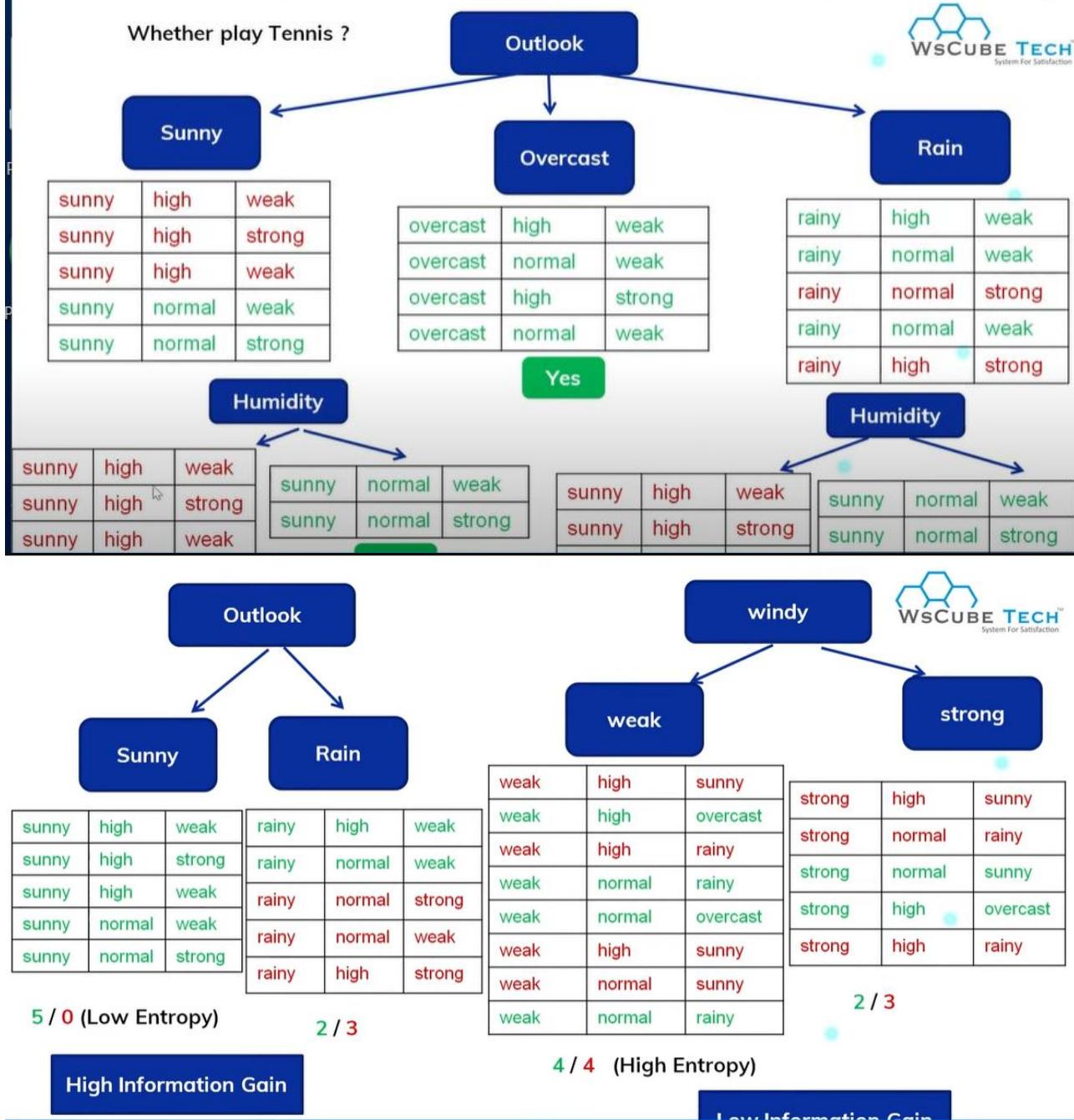


## How we create a Decision Tree?

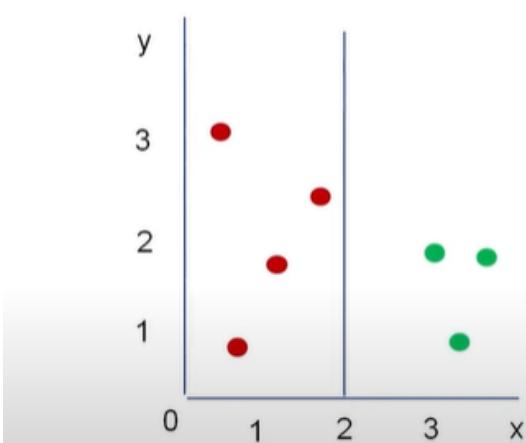
	outlook	humidity	windy	play
1	sunny	high	weak	no
2	sunny	high	strong	no
3	overcast	high	weak	yes
4	rainy	high	weak	yes
5	rainy	normal	weak	yes
6	rainy	normal	strong	no
7	overcast	normal	weak	yes
8	sunny	high	weak	no
9	sunny	normal	weak	yes
10	rainy	normal	weak	yes
11	sunny	normal	strong	yes
12	overcast	high	strong	yes
13	overcast	normal	weak	yes
14	rainy	high	strong	no



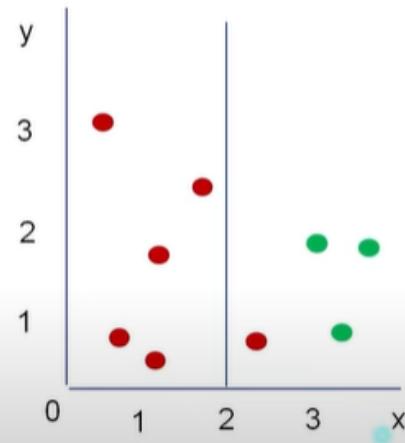
Whether play Tennis ?



# Gini Impurity



A perfect split



A imperfect split

## Random Forest Algorithm

### What is Random Forest Algorithm ?

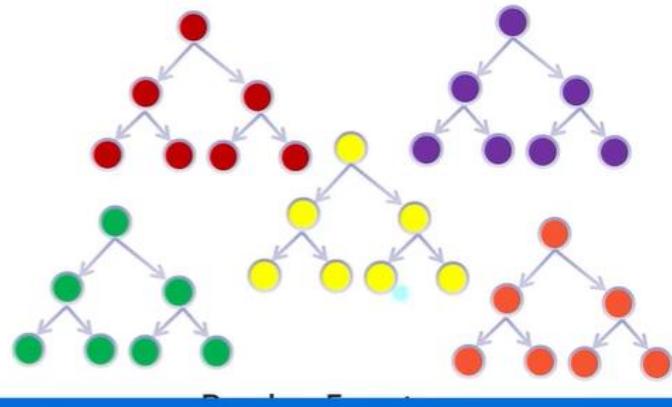
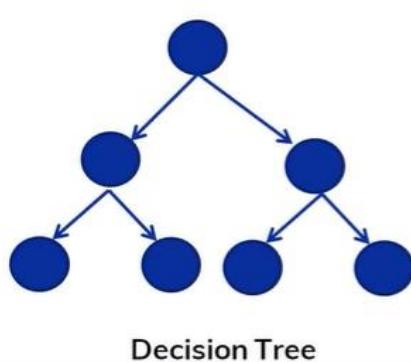
Random Forest is a machine learning algorithm based on supervised learning, that can be used for both regression and classification problems.

It is a collection of multiple random decision trees, which is called the forest.

# What is Random Forest Algorithm ?

Random Forest is a machine learning algorithm based on supervised learning, that can be used for both regression and classification problems.

It is a collection of multiple random decision trees, which is called the forest.

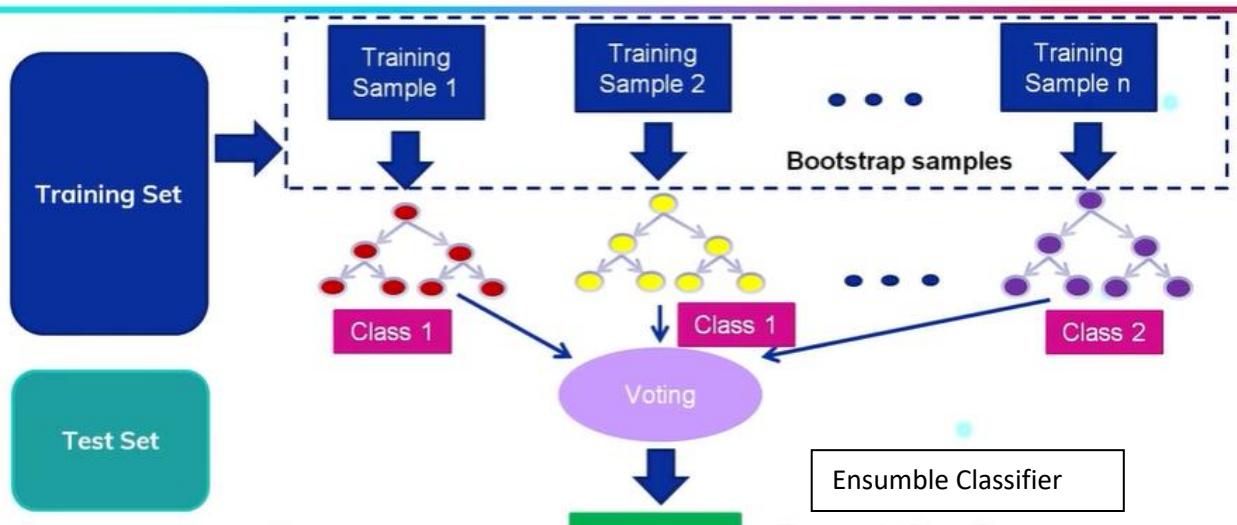


## Real Life Example of Random Forest Algorithm



Considering opinions of many while

## How does the algorithm works ?



# Iris Dataset

SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
6.8	3.2	5.9	2.3	Iris-virginica
6.9	3.1	5.1	2.3	Iris-virginica
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6.6	3.0	4.4	1.4	Iris-versicolor



**TIME – 5.40**

# Naïve Bayes Classifier Algorithm



## What is Naïve Bayes Classifier Algo ?

Naïve Bayes Classifier is a machine learning algorithm based on supervised learning, that can be used for solving classification problems.

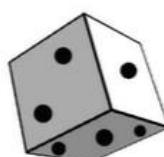
It is a probabilistic classifier, which means it predicts on the basis of the probability of an object.



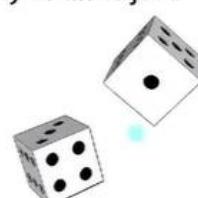
Sample Space:  
{H,T}



Sample Space:  
{HH,HT,TH,TT}



Sample Space:  
{1,2,3,4,5,6}



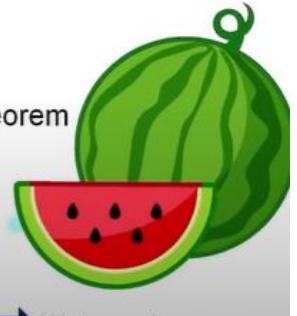
Sample Space:  
{(1,1),(1,2),(1,3),(1,4),(1,5),(1,6),  
(2,1),(2,2),(2,3),(2,4),(2,5),(2,6),  
(3,1),(3,2),(3,3),(3,4),(3,5),(3,6),  
(4,1),(4,2),(4,3),(4,4),(4,5),(4,6),  
(5,1),(5,2),(5,3),(5,4),(5,5),(5,6)}

# Why the Name Naïve Bayes ?

It comprises of two words Naïve and Bayes:

**Naïve:** It's called naive because it makes the assumption that all attributes are independent of each other.

**Bayes:** It's called bayes because it depends on the principle of Baye's Theorem



Fruit = {green, oval, sweet} → Watermelon

## What is Conditional Probability ?

Playing Card Example



Playing Card Suits

4 Suits = Hearts, Clubs, Diamonds, Spades

13 cards = Ace, 2, 3, 4, 5, 6, 7, 8, 9, 10, Jack, Queen, King  
Each of the 13 cards has all 4 suits.

If you pick a card from the deck, can you guess the probability of getting a king given the card is a spade?

Total spade=13

King=1

$P(\text{king}/ \text{spade}) = 1/13 = 0.076$

$P(A|B)$ = probability of A occurring given that B has already occurred.

## What is Bayes Theorem ?

The Naive Bayes classifier works on the principle of conditional probability, as given by the Bayes theorem.

Where,

$$P(A | B) = \frac{P(B | A) P(A)}{P(B)}$$

A,B= Events

$P(A|B)$ = Probability of A given that B is true

$P(B|A)$ = Probability of B given that A is true

$P(A)$ = Probability of event A

$P(B)$ = Probability of event B

# Bayes Theorem [contd...]

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

$$P(\text{ king} | \text{ spade}) = \frac{P(\text{ spade} | \text{ king}) * P(\text{ king})}{P(\text{ spade})}$$

$$= \frac{1/4 * 1/13}{1/4} = 1 / 13 = 0.076$$

$$P(\text{ spade} / \text{ king}) = 1/4$$

$$P(\text{ king}) = 1/13$$

$$P(\text{ spade}) = 1/4$$

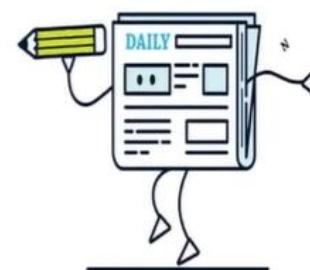
## Where Naïve Bayes is Used?



Face Recognition



Weather Forecast



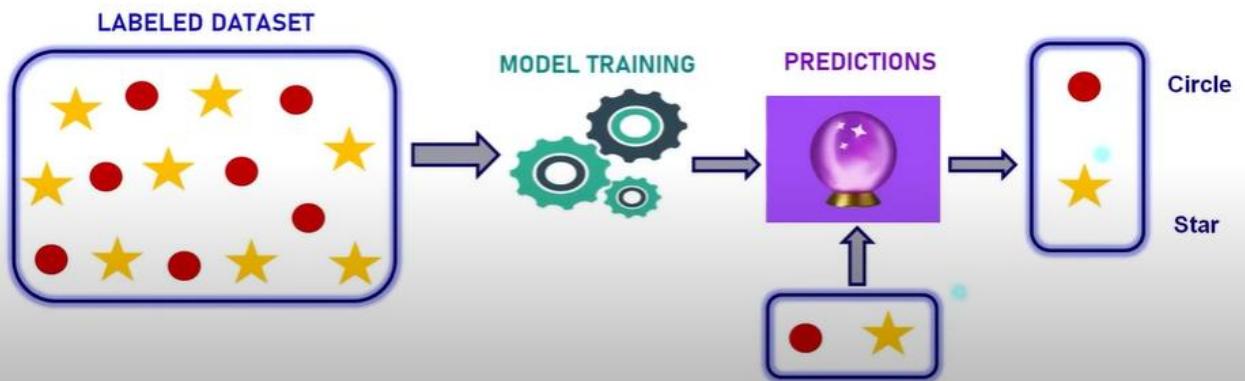
News Categorization

**TIME = 6.02**

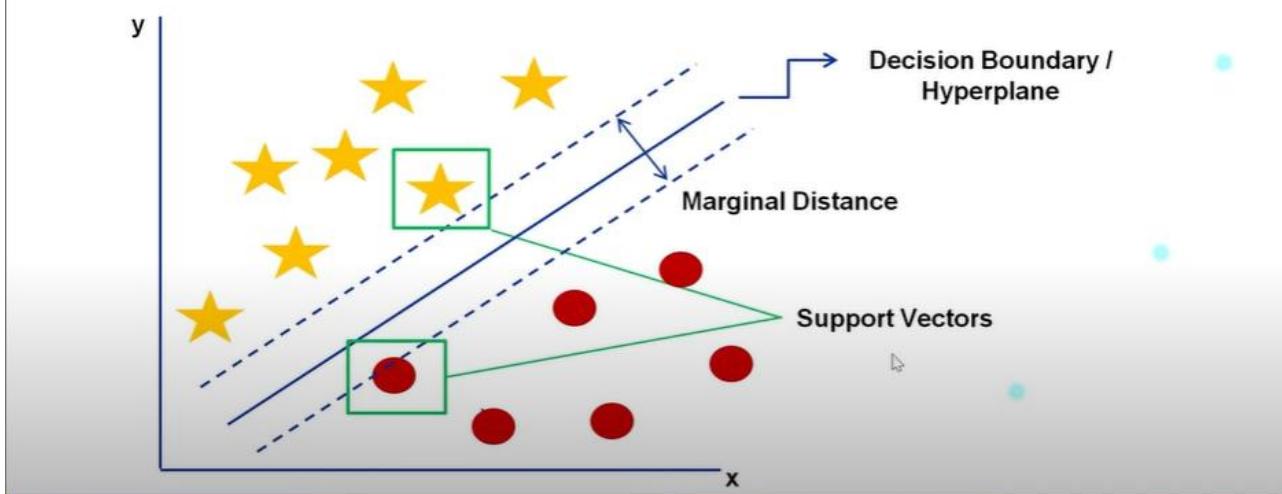
# Support Vector Machine Algorithm

## What is Support Vector Machine ?

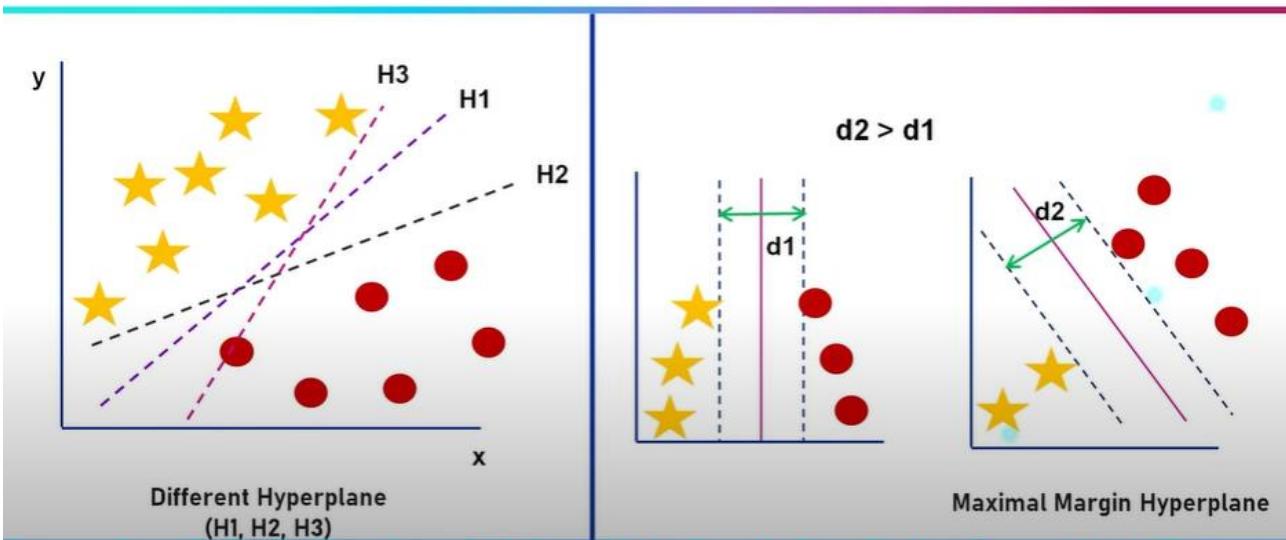
Support Vector Machine is a machine learning algorithm based on supervised learning, that can be used for both regression and classification problems.



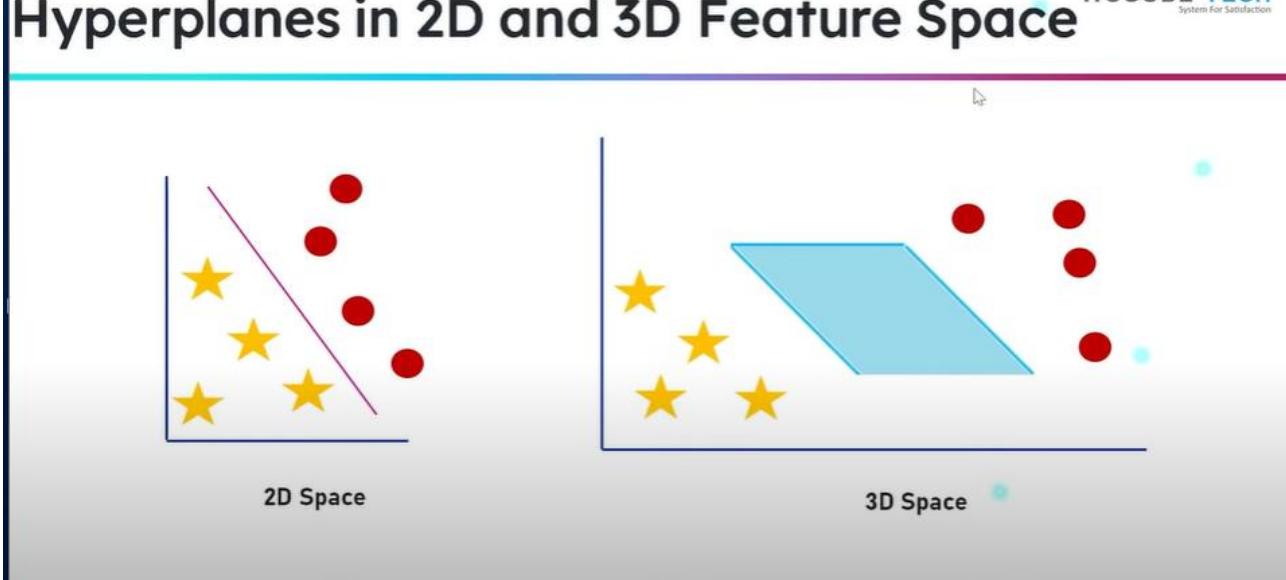
## Support Vector Machine Terminologies



# Which Hyperplane to Select ?



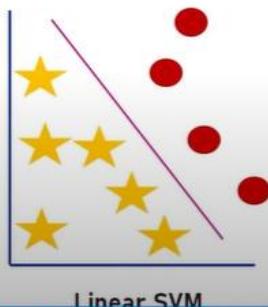
## Hyperplanes in 2D and 3D Feature Space



# Types of SVM



Linear SVM is used when dataset can be classified into 2 classes using a straight line



Linear SVM

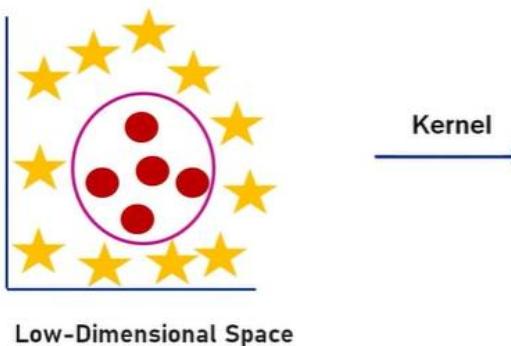
Non-Linear SVM is used when the dataset cannot be classified into 2 classes using a straight line



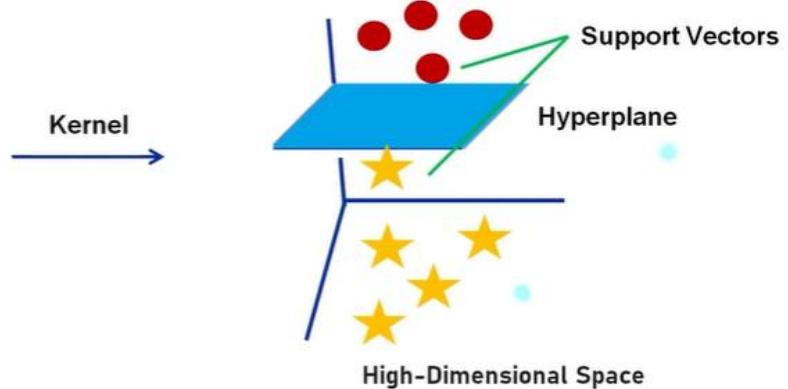
Non-Linear SVM

## Kernel Function

Kernel functions takes low dimensional input space and transform it into a higher-dimensional space, i.e., it converts not separable problem to separable problem.



Low-Dimensional Space



High-Dimensional Space

**TIME – 6.22**

Confusion  
Matrix



# What is Confusion Matrix ?

A confusion matrix is an  $N \times N$  matrix used for evaluating the performance of a classification model, where  $N$  is the number of target classes.  
 This compares the actual target values with predicted target values.

		Actual Values		
		Positive	Negative	
Predicted Values	Positive	TP	FP	Type 1 Error
	Negative	FN	TN	Type 2 Error

## Confusion Matrix for Binary Classification Problem

n=200	Predicted: No	Predicted: Yes	
Actual: No	TN=100	FP=15	115
Actual: Yes	FN=5	TP=80	85
		105	95

**True Negatives (TN):** The patients actually don't have disease and our model also says that these patients don't have disease

**True Positives (TP):** These are cases in which, The patients actually have disease and our model also says that these patients do have disease

**False Positives (FP):** These are cases in which, The patients actually don't have disease but our model says that these patients do have disease

**False Negatives (FN):** These are cases in which, The patients actually have disease but our model says that these patients don't have disease

## Accuracy, Error Rate, Precision, Recall

### ACCURACY

The no of predictions that the model got right

$$\text{Accuracy} = (\text{TP} + \text{TN}) / (\text{TP} + \text{TN} + \text{FP} + \text{FN}) = 80 + 100 / 200 = 0.9$$

### Error Rate

The no of predictions that the model predicted wrongly

$$\text{Error Rate} = (\text{FP} + \text{FN}) / (\text{TP} + \text{TN} + \text{FP} + \text{FN}) = 15 + 5 / 200 = 0.1$$

### Precision

When the model predicts the positives, how often is it right?

$$\text{Precision} = \text{TP} / (\text{TP} + \text{FP}) = 80 / 95 = 0.86$$

### Recall

When it's actually yes, how often does it predict yes?

$$\text{Recall} = (\text{TP}) / (\text{TP} + \text{FN}) = 80 / 85 = 0.94$$

		Predicted: No	Predicted: Yes	
n=200	Actual: No	TN=100	FP=15	115
Actual: Yes	FN=5	TP=80		85
		105	95	

**TIME – 6.32**

# Underfitting and Overfitting



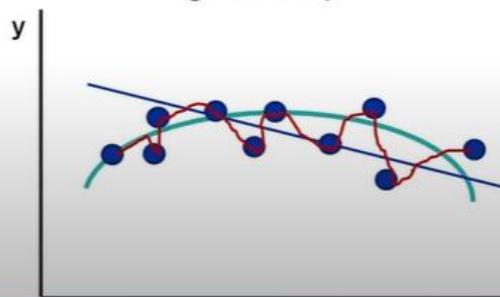
## Underfitting and Overfitting

Model's performance is evaluated based upon Accuracy and Generalisation.

Accuracy: It states how well a model predicts the right output

Generalisation: It states how well model behaves on new data set

A model is said to be best when it behaves nearly same way on training as well as test data with high accuracy

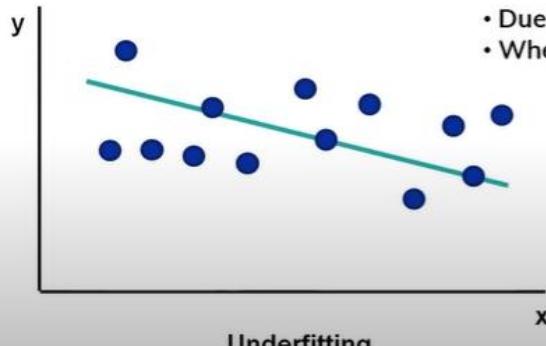


# Underfitting

Underfitting means model has low accuracy score on training as well as on test data

Underfitting occurs:

- Due to lack of data
- When we try to build a linear model with a non-linear data.

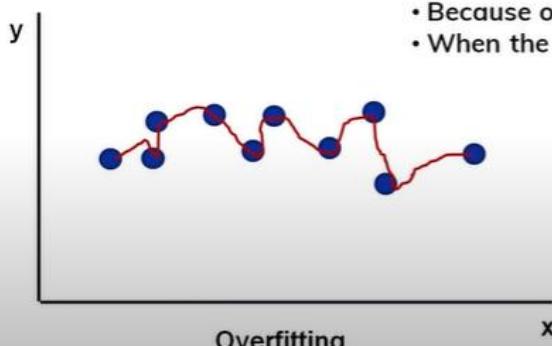


# Overfitting

Overfitting means model has high accuracy score on training data but low score on test data

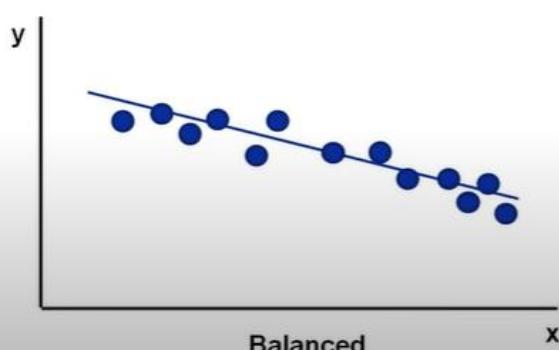
Overfitting occurs:

- Because of a lot more of data
- When the model fits the data too well.



# Right Fit

Ideally, the case when the model makes the predictions with 0 error, is said to have a good fit on the data.



# Real Life Example of Underfitting

Underfitting Scenario



Single Feature:  
**Bristle size : 0.15 mm**



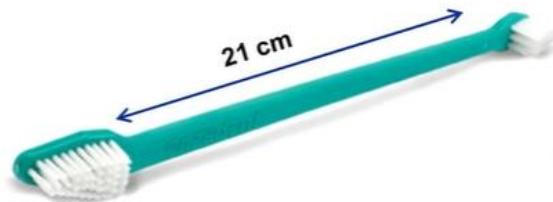
**It's a Toothbrush**

# Real Life Example of Overfitting

Overfitting Scenario



Feature:  
**bristle size : 0.15 mm**  
**bristle type: soft**  
**used in: mouth**  
**handle : 16 cm**



**It's Not a Toothbrush**

# Thanks!

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