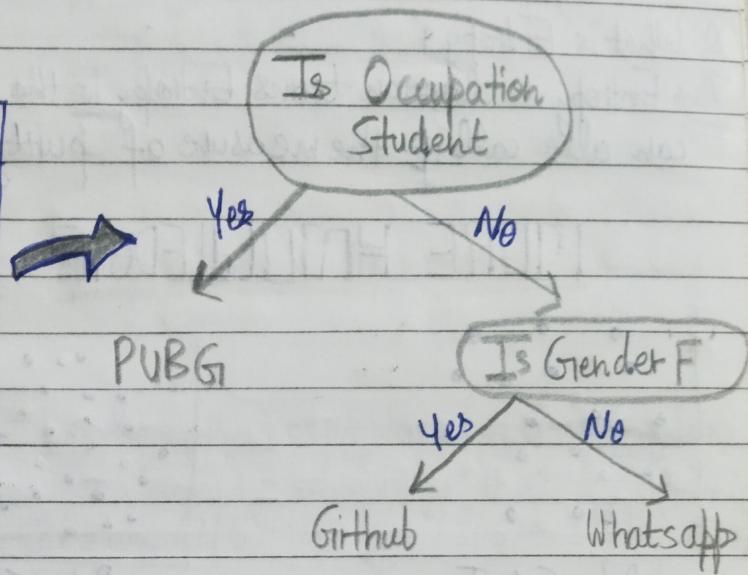


Date.....

Decision Tree

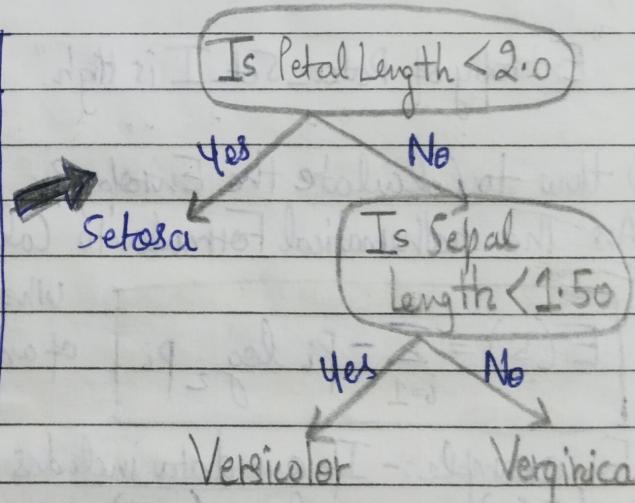
Example 1:-

Gender	Occupation	Suggestion
F	Student	PUBG
F	Programmer	Github
M	Programmer	Whatsapp
F	Programmer	Github
M	Student	PUBG
M	Student	PUBG

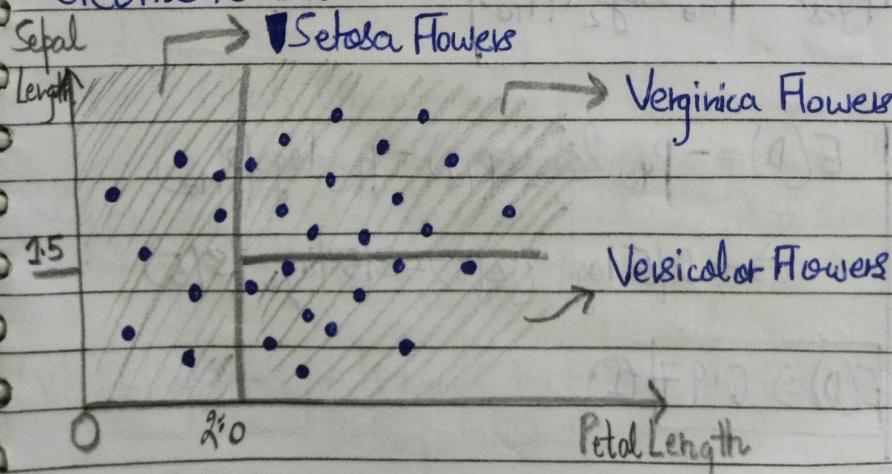


Example 2:-

Petal Length	Sepal Length	Type
1.34	0.34	Setosa
3.45	1.45	Versicolor
1.69	0.98	Setosa
2.56	1.79	Virginica
3.00	1.13	Versicolor
1.3	0.86	Setosa



Geometric Intuition



Spiral

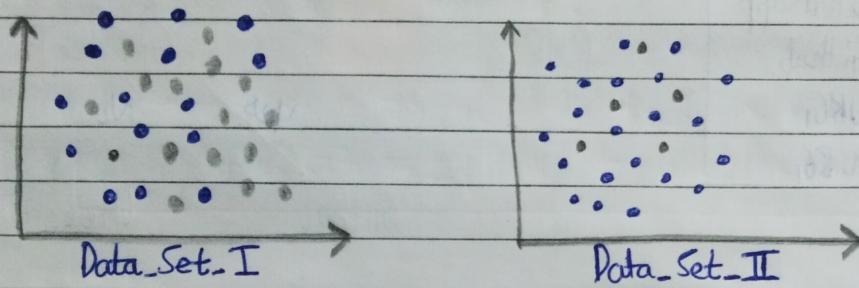
Date.....

Entropy in Decision Trees

Q What is Entropy?

Ans In layman terms, Entropy is the measure of disorder. Or you can also call it the measure of purity/impurity. Let's see an example:-

MORE KNOWLEDGE LESS ENTROPY



"Entropy of Data_Set_I is High."

Q How to Calculate the Entropy?

Ans The mathematical formula to calculate Entropy is :-

$$E(S) = \sum_{i=1}^k -P_i \log_2 P_i$$

Where 'Pi' is simply the frequent probability of an element / class 'i' in our data.

For example:- If our data includes 2 classes only. "Yes" and "No".

$$E(D) = -P_{Yes} \log_2 (P_{Yes}) - P_{No} \log_2 (P_{No})$$

Example:- ①

Salary	Age	Purchase
20000	21	Yes
10000	45	No
60000	27	Yes
15000	31	No
12000	18	No

Data_Set_I

$$E(D) = -P_{Yes} \log_2 (P_{Yes}) - P_{No} \log_2 (P_{No})$$

$$\Rightarrow -2/5 \log_2 (2/5) - 3/5 \log_2 (3/5)$$

$$E(D) \Rightarrow 0.97 \text{ ①}$$

Data Set II

Date.....

Salary	Age	Purchase	$E(D) = -P_{Yes} \log_2(P_{Yes}) - P_{No} \log_2(P_{No})$
34000	31	No	$\Rightarrow -1/5 \log_2(1/5) - 4/5 \log_2(4/5)$
15000	25	No	
69000	57	Yes	$E(D) \Rightarrow 0.72$ (I)
25000	21	No	
32000	28	No	

From (I) and (II), Entropy of (I) is High....

Example:- (2)

Salary	Age	Purchase	$E(D) = -P_{Yes} \log_2(P_{Yes}) - P_{No} \log_2(P_{No}) - P_{Maybe} \log_2(P_{Maybe})$
20000	21	Yes	$\Rightarrow -2/8 \log_2(2/8) - 3/8 \log_2(3/8) - 3/8 \log_2(3/8)$
10000	45	No	
60000	47	Yes	$E(D) \Rightarrow 1.56$
15000	31	No	
30000	19	Maybe	
12000	40	No	
40000	35	Maybe	
20000	29	Maybe	

Data Set III

Observations.....

- (i) As the Knowledge decreases, the Entropy Increases.
- (ii) For 2 Class Classification, the minimum Value of Entropy will be "0" and the Maximum Value of Entropy will be "1".
- (iii) For 3 Class Classification, the minimum Value of Entropy will be "0" and the Maximum Value of Entropy is greater than > 1 .
- (iv) $Entropy = 0 \rightarrow$ It means that the Data Set has too much Knowledge to predict the new input...
- (v) $Entropy = 1 \rightarrow$ It means that the Data Set values let's say "Yes" and "No" Both probabilities are equal...