

### 1. What is NB Tree.

Ans:

It's a hybrid learning approach of decision tree & naive bayes classifiers.

NB Tree is a supervised type learning

NB Tree max Posteriori Probability এর উপর base করে কাজ করে।

NB Tree = Decision Tree + NB

Decision tree এর leaf node replaced এর NB model দিয়ে।

NB Tree একটি Pruned tree.

Pruned tree স্বয়ংক্রিয় NB tree majority class নিয়ে কাজ করে।

Leaf node- একটি sub-data set নিয়ে কাজ করে।

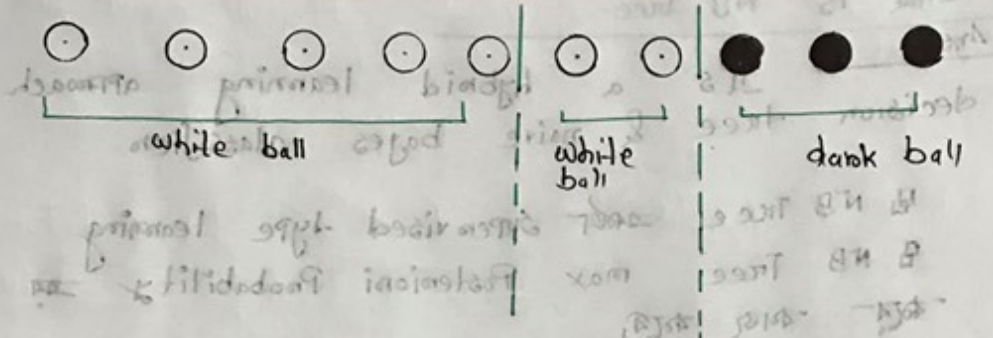
### 2. Why NB Tree is useful?

Ans:

Adaptive NB tree splits the dataset by applying entropy based algorithm & then used standard NB classifiers at the leaf node to handle attribute.

It applies strategy to construct decision tree & replaces leaf node with NB classifiers.

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Joint Probability

For (White ball) =  $\frac{7}{10}$

For (dark - ball) =  $\frac{3}{10}$

5 white balls together =  $\frac{5}{5} = 1$  // Highest Probability  
 2 " " " =  $\frac{2}{2}$   
 3 dark " " =  $\frac{3}{3}$

Main Motivation: We if divide data point, dataset in such manner, the same would be together.



### 3. Define One-R Classifier

Ans:

It's a classification rule. Based on the value of a single Predictor, that generates one rule for each Predictor in the data.

A One-R Classifier  $\Rightarrow$  Construct a frequency table.

### 4. What is Frequency Table

Ans:

It's a table that represents the numbers of occurrences of every unique value in the variable.

### 5. Write the Algorithm of One R Classifier

Ans:

Input:  $D = \{x_1, x_2, \dots, x_n\}$  // Training Data

Output: OneR Model.

Method

```

for each attribute  $A_i \in D$ , do
    for each attribute value  $A_{ij} \in A_i$ , do
        Make a classification rule:
            count how often each class appears
            find the most frequent class
            make the rule assign that class to this  $A_{ij}$ 
    end for
    calculate the error rate of this attribute's  $A_i$  rule
end for
choose the attribute  $A_i \in D$  with the smallest error rate
  
```

**Table:** The playing tennis dataset

Day	Outlook	Temperature	Humidity	Wind	Play
$D_1$	Sunny	Hot	High	Weak	No
$D_2$	Sunny	Hot	High	Strong	No
$D_3$	Overcast	Hot	High	Weak	Yes
$D_4$	Rain	Mild	High	Weak	Yes
$D_5$	Rain	Cool	Normal	Weak	Yes
$D_6$	Rain	Cool	Normal	Strong	No
$D_7$	Overcast	Cool	Normal	Strong	Yes
$D_8$	Sunny	Mild	High	Weak	No
$D_9$	Sunny	Cool	Normal	Weak	Yes
$D_{10}$	Rain	Mild	Normal	Weak	Yes
$D_{11}$	Sunny	Mild	Normal	Strong	Yes
$D_{12}$	Overcast	Mild	High	Strong	Yes
$D_{13}$	Overcast	Hot	Normal	Weak	Yes
$D_{14}$	Rain	Mild	High	Strong	No



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Inner loop for classification rule  
Outer loop for attribute error

### Simulation of One R Based on Playing Tennis Dataset

Attribute Outlook

Item	Attribute Value	Yes	No	Error	Total Error
1	Sunny	Yes = 2/5 = 0.4	No = 3/5 = 0.6	2/5	$\frac{2}{5} + \frac{0}{4} + \frac{2}{5}$
2	Overcast	Yes = 4/4 = 1	No = 0/4 = 0	0/4	$\frac{2+0+2}{5+4+5} = \frac{4}{14}$
3	Rain	Yes = 3/5 = 0.6	No = 2/5 = 0.4	2/5	

\* Error

\* 4/14

total 4 miss-classified

Attribute  $\Rightarrow$  Temperature

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Item	Attribute Value	Yes	No	Error	Total Error
1	Hot	$2/4 = 0.5$	$2/4 = 0.5$	$2/4$	$\frac{2}{4} + \frac{3}{6} + \frac{1}{4}$ $= \frac{2+2+1}{4+6+4}$ $= \frac{5}{14}$
2	Mild	$4/6 = 0.6$	$2/6 = 0.3$	$2/6$	
3	Cool	$9/4 = 0.75$	$1/4 = 0.25$	$1/4$	

ମଧ୍ୟ ସମାନ ବାସ୍ତବ ସୂଚକର ସଂଖ୍ୟା ନିମ୍ନ,  
ଅତଏବ Less Error ହେବା ହେବା ଉଚିତ୍,

Attribute  $\Rightarrow$  Humidity

Item	Attribute Value	Yes	No	Error	Total Error
1	High	$3/7 = 0.4$	$4/7 = 0.5$	$3/7$	$\frac{3}{7} + \frac{1}{7}$ $= \frac{3+1}{7+7}$ $= \frac{4}{14}$
2	Normal	$6/7 = 0.8$	$1/7 = 0.1$	$1/7$	

Attribute  $\Rightarrow$  Wind

Item	Attribute Value	Yes	No	Error	Total Error
1	Weak	$6/8 = 0.7$	$2/8 = 0.2$	$2/8$	$\frac{3}{8} + \frac{3}{6}$ $= \frac{2+3}{8+6}$ $= \frac{5}{14}$
2	Strong	$3/6 = 0.5$	$3/6 = 0.5$	$3/6$	

Error Rate କିମ୍ବା ସଂଖ୍ୟା accuracy (ସଂଖ୍ୟା)



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### 6. What is Support Vector Machine

Ans:

It's a supervised learning model, with associated learning algorithms that analyze data used for classification & regression analysis.

SVM training algorithms build a model that assigns new vectors into one category or the other.

### 7. Write the Working Process of SVM

Ans:

It's a representation of the vectors as point in space, mapped so that the vectors of the separate categories are divided by a clear gap that is as wide as possible.

New vectors are then mapped into that same space & Predicted to belong to a category based on which side of the gap.

ଏହା non-linear Classification Perfromance ବୃଦ୍ଧି ଆଣେ,  
- ଏହାକୁ ଏକ kernel trick.

ଏ kernel trick high-dimensional feature space  
- input କୁ mapping କରାଏ,

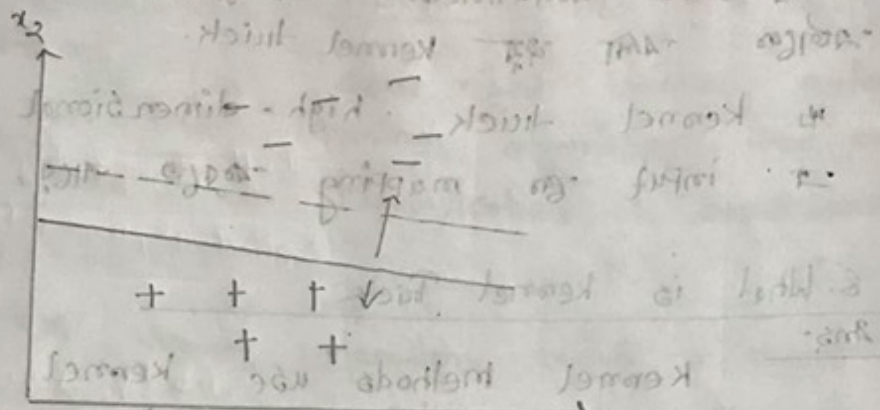
Q. What is kernel Trick

Ans.

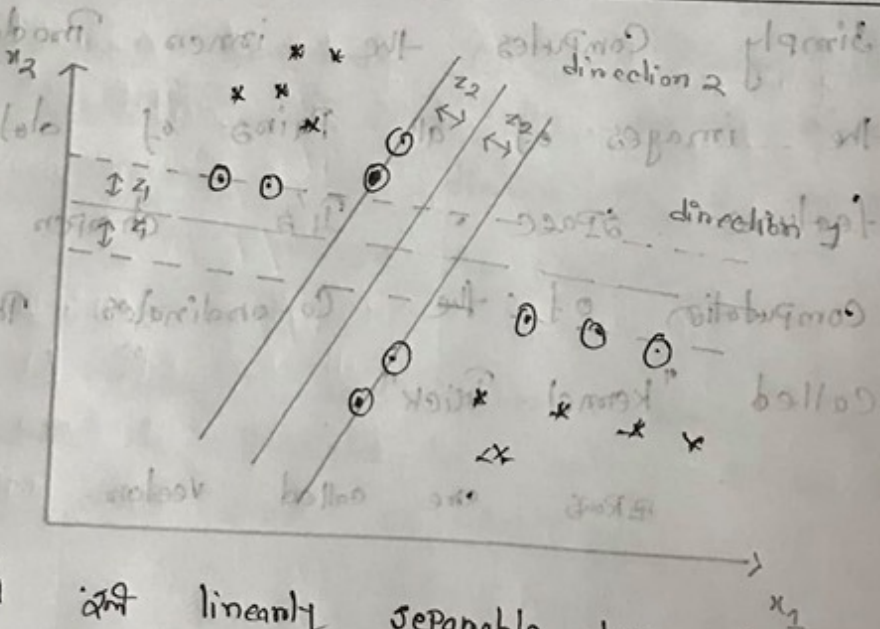
Kernel methods use kernel functions. It enable them to operate in a high-dimensional implicit feature space without even computing the co-ordinates of the data in that space. It simply computes the inner product between the images of all Pairs of data in the feature space. It's cheaper than explicit computation of the co-ordinates. This approach is called "kernel Trick".

Rows are called vectors.





linearly separable two class Problem.



linearly separable two class Problem  
with two possible linear classifiers

→ 0 misclassified instance miss -

9. Show the relation between Vector inner Product & norm with Projection.

Ans:

Let, we have two dimensional vectors

$$u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \quad \& \quad v = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$u^T v = \begin{bmatrix} u_1 & u_2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = v^T u = P \cdot \|u\|$$

$$\|u\| = \sqrt{u_1^2 + u_2^2} \quad [u_1, u_2 \in \mathbb{R}]$$

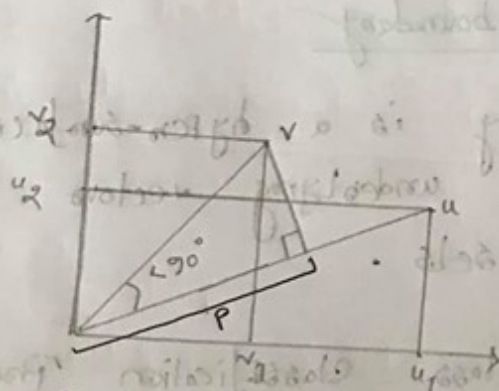


Fig 1: Vector Inner Product of  $u$  &  $v$ , while ( $< 90^\circ$ )

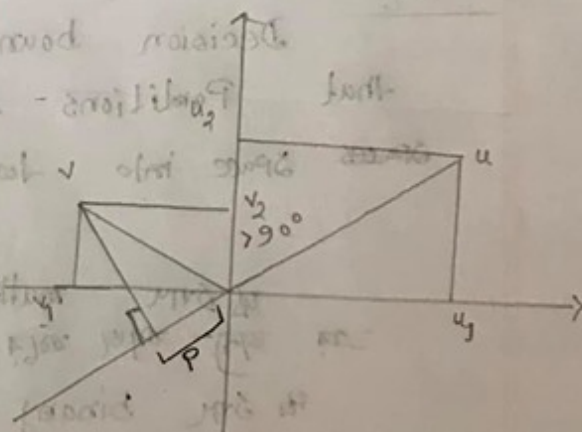


Fig 2: Vector Inner Product of  $u$  &  $v$ , while ( $> 90^\circ$ )

Here

The angles find support vectors.

$P$  means vector inner Product

$u_1 = x$  value of  $u$

$u_2 = y$  value of  $u$

$v_1 = x$  value of  $v$

$v_2 = y$  value of  $v$



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The greater the gap  $P$ , the the gap is maximized

So,

$$U^T V = P \cdot \|U\|$$

$$U^T V = u_1 v_1 + u_2 v_2$$

So,

$$\begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = U$$

$$\begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = V$$

$$U^T V = V^T U$$

$$U^T V = P \cdot \|U\| = V^T U = u_1 v_1 + u_2 v_2 \leftarrow P \in \mathbb{R}$$

Minimizing  $U^T V$ , maximizes  $P$ .

### 1a. Define SVM decision boundary

Ans.

Decision boundary is a hyper-surface that partitions the underlying vector space into two sets.

এ SVM multi-class classification problem  
এক প্রকার করে হয়,

এ SVM binary class - classification  
এক প্রকার করে,

সমস্যা সমাধানের জন্য  
লক্ষ্য হল সঠিকভাবে  
ক্লাস নির্ধারণ করা।  
যে ক্লাসের  $x = 1$ ।  
যে ক্লাসের  $x = 0$ ।  
যে ক্লাসের  $x = -1$ ।  
যে ক্লাসের  $x = 2$ ।

## 11. How to overcome Slow Processing of svm

Ans:

Kernel trick is the solution. In kernel trick mapping the non-linear separable data-set into a higher dimensional space where we can find a hyperplane that can separate the sample.