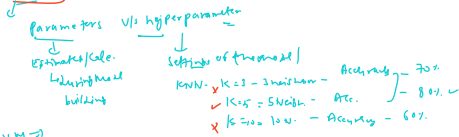
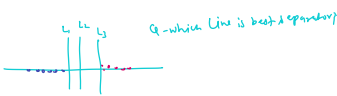
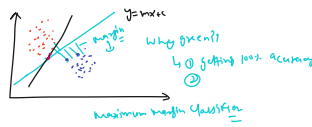


$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots$ ① - Log/Linear

$y = f(x)$ - NB/Knn - non-parametric model

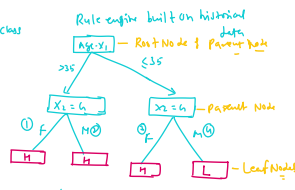


SVM



Decision Tree \Rightarrow it comes up with tree on our dataset so that we can take decisions

Age X_1	Gender X_2	Y - Income class
30	F	H
35	M	L
30	M	H
40	F	H
37	M	?? H
35	M	L

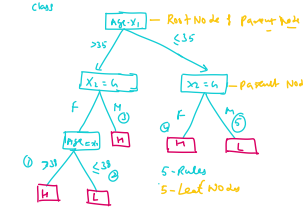


Split Methods - ??

- Gini Score
- variance
- Cutoff
- Entropy

4-Rules

Depth of the tree - 2



Depth of the tree - 3

Answer - 3

Random forest \Rightarrow

total - 52 Cards

$P(R1) = \frac{26}{52} = 0.5 = 1/2 \rightarrow 10$

$P(R2|R1) = \frac{25}{51}$ - without putting the Card back to deck

I have put the Card back to the deck

$P(R2|R1) = \frac{26}{52} = 0.5$

Apply Sampling - $N=1000$

$\frac{1}{1000} n_1 \in N = 10 \times \frac{100}{1000} = 1$ = good

$\frac{1}{1000} n_2 \in N = 10 \times \frac{100}{1000} = 1$

$\frac{1}{1000} n_3 \in N = 10 \times \frac{100}{1000} = 1$

Sampling with replacement

\hookrightarrow Bootstrap Sampling

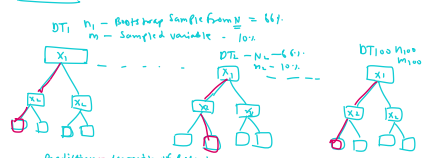
Decision Tree - When the dataset is small - they work well using h/p training
 \hookrightarrow they overfit - they are greedy algo.

$N = 1MV$ - DT - too much time to train - inefficient
 $(Xs) m = 100$

RF \Rightarrow $(N = 1MV, m = 100)$ - Random Sample from N - DT100
 Random sample from m - DT100

Independent trees in parallel

$N = 1MV$
 $m = 100$



Prediction = Majority of Prediction

$DT_1 = H, DT_2 = H, DT_{100} = L$

$60 \times - DT = H$

$40 \times - DT = L$

$N = 10$ total data points

DT_1, DT_2, DT_3
 $n_1 \in N, n_2 \in N, n_3 \in N$
 $10 = n_1 \sim 64\% = 6.4$
 $10 = n_2 \sim 64\% = 6.4$
 $10 = n_3 \sim 64\% = 6.4$

$P(\text{single Point}) = 1/10$

$P(2 \times P) = 1/10$

3.6 security (repeated)