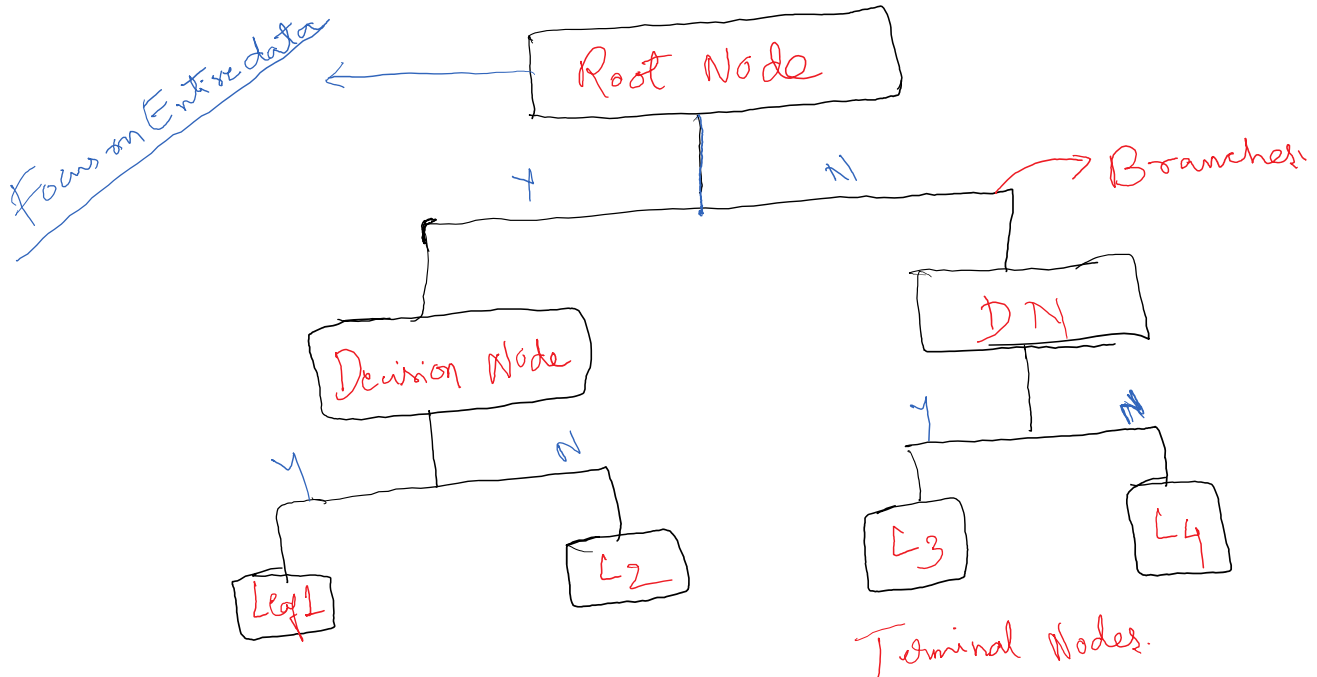


Decision Tree → Divide & Conquer → Most widely used

→ Makes Complex choices from simple set of choices & represent this learning in logical form

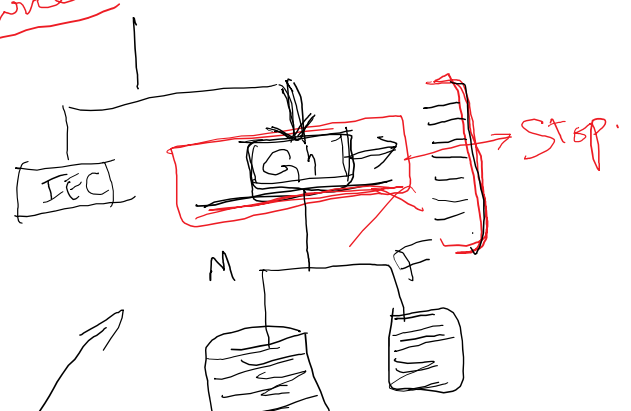


Recursive Partitioning → Enabling a machine to discover/learn something on its own

→ Splits → Data into Homogenous parts → IN/US

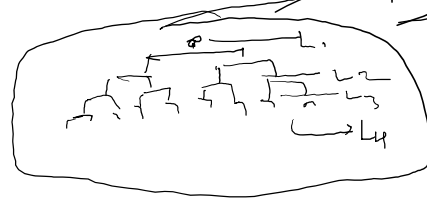
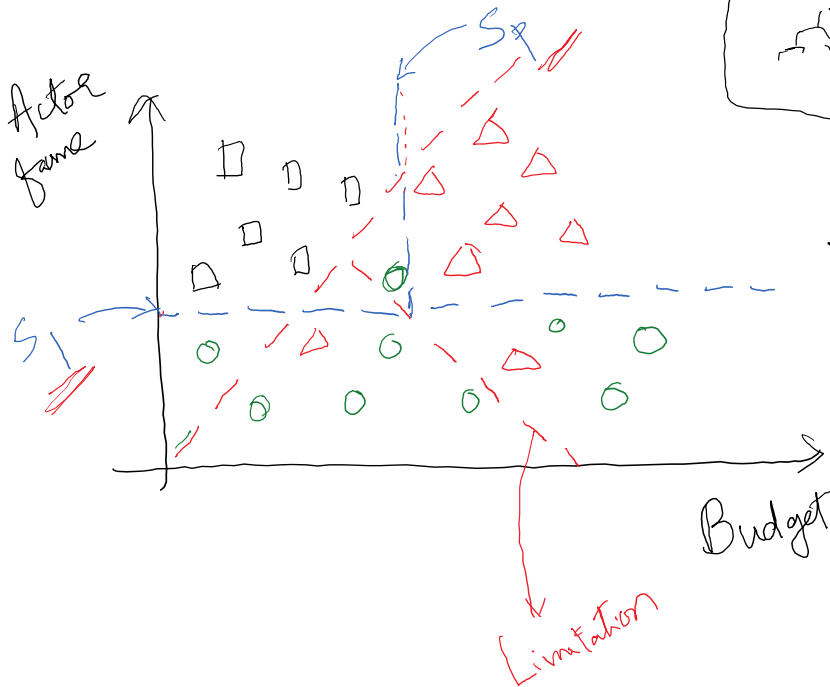
→ Terminates → All nodes have same class.

C.4.5 → Single thread → open source  
C.5.0 → Multi thread → Premium



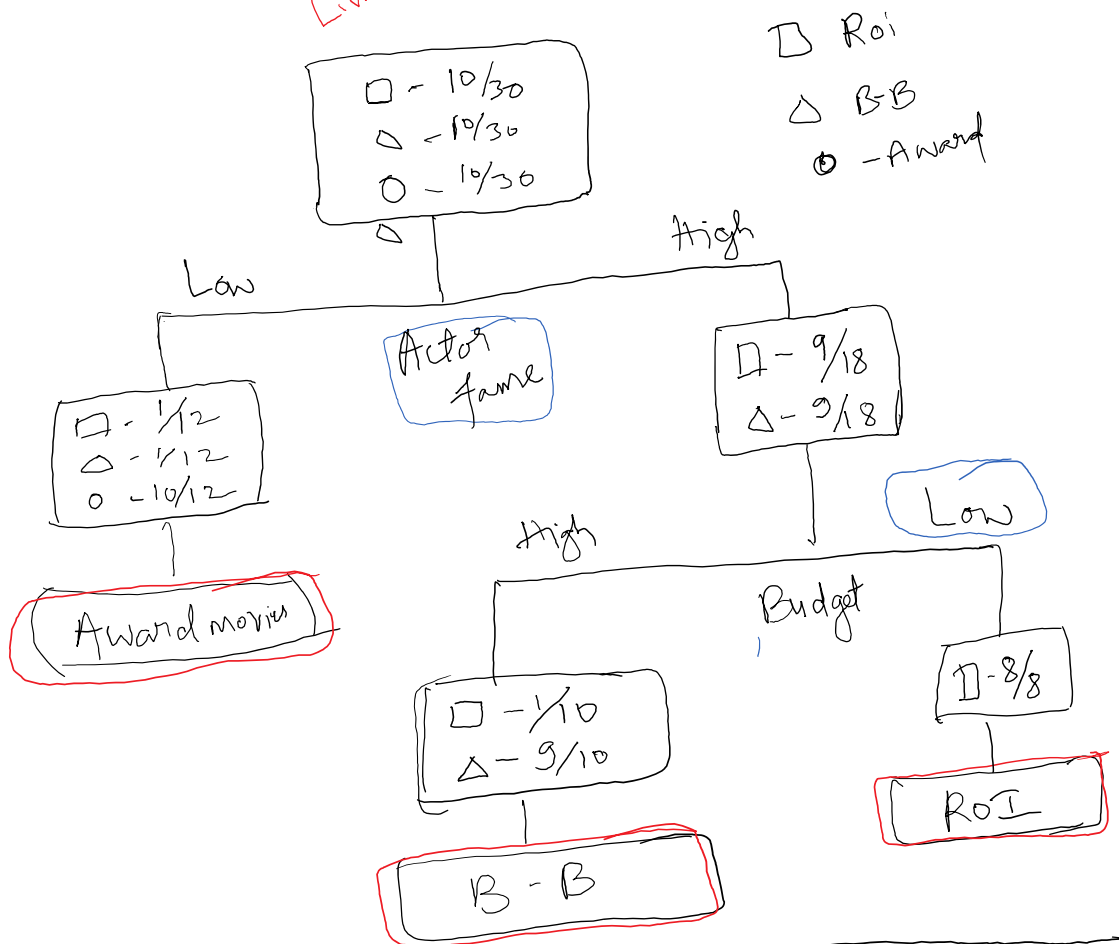
→ No features remaining

→ Has grown to a predicted limit → 4 levels



$\square$  Roi  
 $\Delta$  Blockbuster →  
 $\circ$  Award →

→ Axis-Parallel split



/ ... → Tuna-based (R- Package RWeka)

C5.0 Algorithm / C4.5 → Java based (R-package RWeka)

→ Degree to which the subset contains Homogeneous elements, "PURITY"

→ All are homogeneous → "PURE" class

→ C5.0 uses Entropy → Quantifies the Randomness of a set

Entropy ↑ → Very Diverse

Entropy ↓ → Very "PURE"

2 possible → Entropy btwn 0 & 1

$n$  ————— → 0 to  $\log_2(n)$

$$\text{Entropy}(S) = \sum_{i=1}^C -P_i \log_2(P_i)$$

$S \rightarrow$  Data Segment

$C \rightarrow$  no. of class levels.

$P_i \rightarrow$  proportion of values falling in a class.

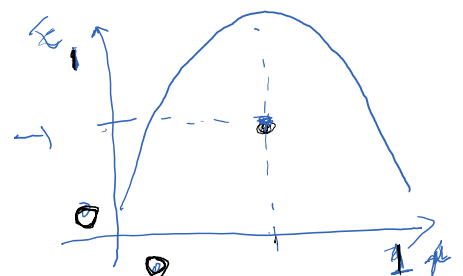
Ex. 2 classes → Red(60%), White(40%)

$$\text{Entropy} = -0.6 \times \log_2(0.6) - 0.4 \times \log_2(0.4) = \underline{\underline{0.97}}$$

↑  
Not PURE

Information gain =  $E(S_1) - E(S_2)$

Info gain ↑      Entropy ↓



→ Other forms of splits → Gini index,  $\chi^2$ , gain Ratio.

Pruning  $\rightarrow$  Trim  $\rightarrow$  Reduce the size to generalise the data.

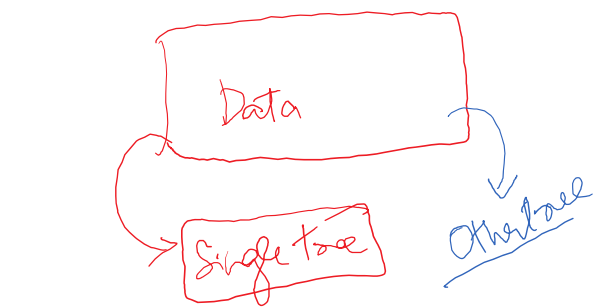
Pre pruning  $\rightarrow$  Early pruning  $\rightarrow$  get result  $\rightarrow$  stop.

post pruning  $\rightarrow$  Allow the tree to grow  $\rightarrow$  Based on need  $\rightarrow$  Trim

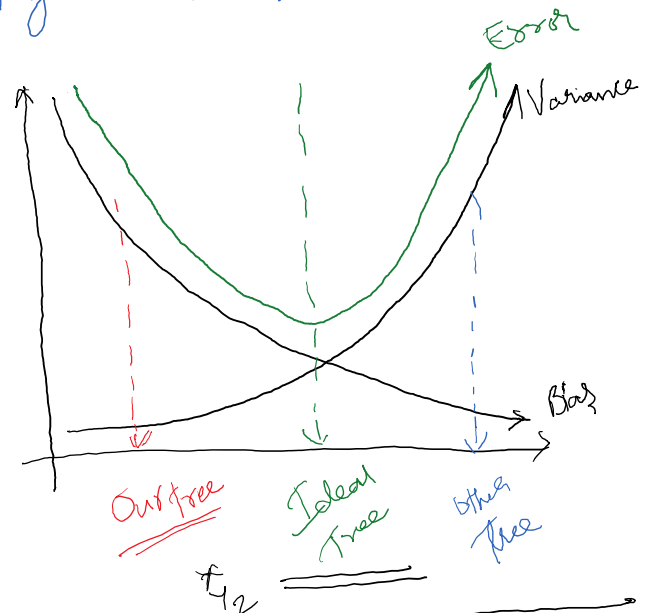
## Bagging, Boosting & Random Forests

Ensemble method  $\rightarrow$  Group methods.

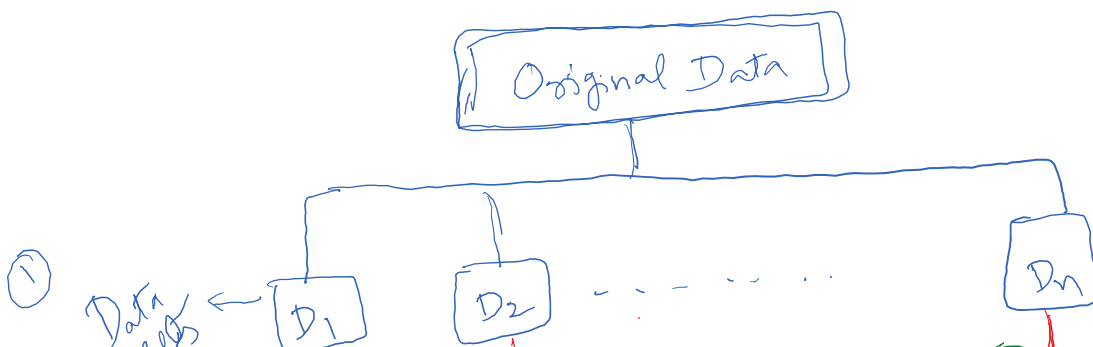
⊗  $\rightarrow$  Group of predictive models to achieve better Accuracy & Model stability  $\rightarrow$  It will give a boost to tree models.



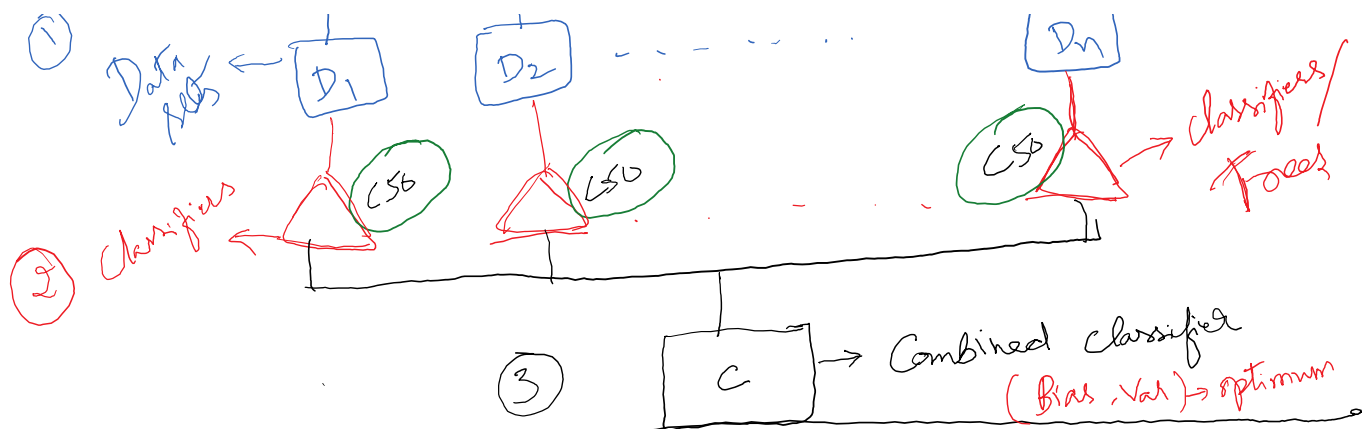
$$E = \text{Noise} + \text{Bias} + \text{Variance}$$



Bagging Reduce Variance of the predictors by Combining Results of Multiple classifiers on different samples of same data.



1 classifier



①  $\rightarrow$  Sample with Replacement  
 $\rightarrow$  Subset selection, Shrinkage

②  $\rightarrow$  Use same classifier.  
 C5.0, Rpart, CART

③  $\rightarrow$  Combine  
 $\rightarrow$  Mean  $\rightarrow$  Mostly  
 $\rightarrow$  Median  
 $\rightarrow$  Mode

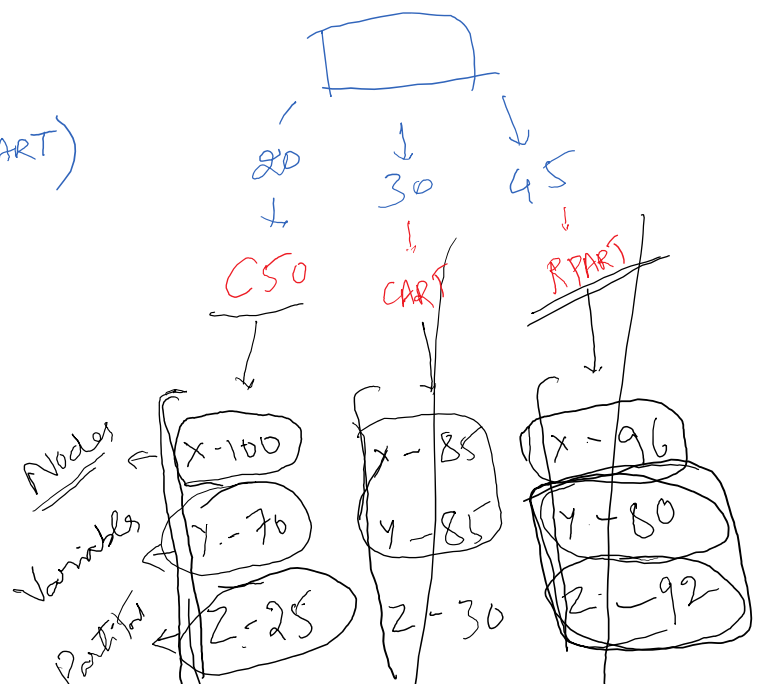
## ⊗ Random Forest

- $\rightarrow$  Apply  $\rightarrow$  Regression & also classification
- $\rightarrow$  Undertakes  $\rightarrow$  Dimensionality Reduction
- $\rightarrow$  It treats Missing values, Outliers, Variable transformation
- $\rightarrow$  Gives the list of Important variables  $\rightarrow$  Contributing %.
- $\rightarrow$  CANNOT Control the mechanism

- $\rightarrow$  Multiple trees  $\Rightarrow$  No pruning
- $\rightarrow$  Different methods  $\rightarrow$  (C5.0, RPART, CART)

$\Rightarrow$  Voting  $\rightarrow$

Var 1	
Var 2	
Var 3	
Var 4	





partition  $\left( \frac{2}{25} \right) \left( \frac{2}{30} \right) \left( \frac{2}{1} \right)$

- $N$  variables →  $M < N$  →  $M$  → Random variables
- Allow the trees to grow fully → No pruning
- Aggregate data of those  $N$  trees

Boosting → Refers to a family of Algorithms to make Weak learners to Strong learners.

(Iterative process)

→ classify S/NS

- Promotional Image → S
- Only Hyperlink → S
- You have won \$... → S
- IEC College → NS
- offer letter Company → NS
- Bank e-statement → NS

Weak learning Rule ①

Rule ②

Rule ③

Rule ④

XGBOOST  
GBM

→ Strong Rule  
No of Iterations

Error

