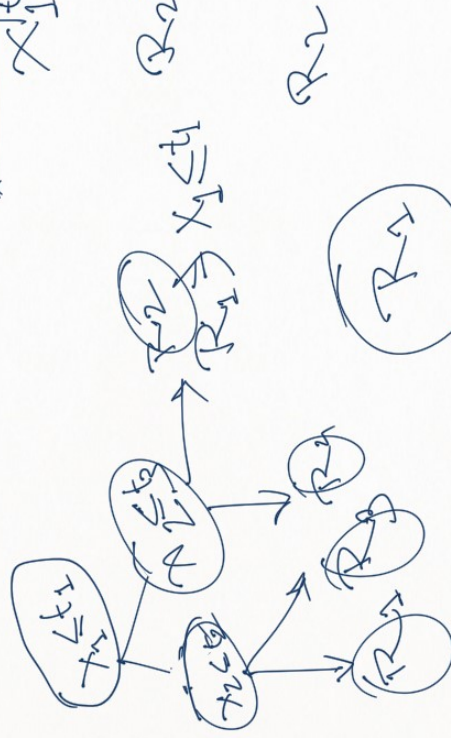
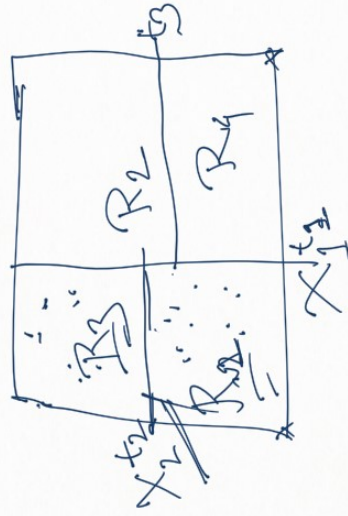
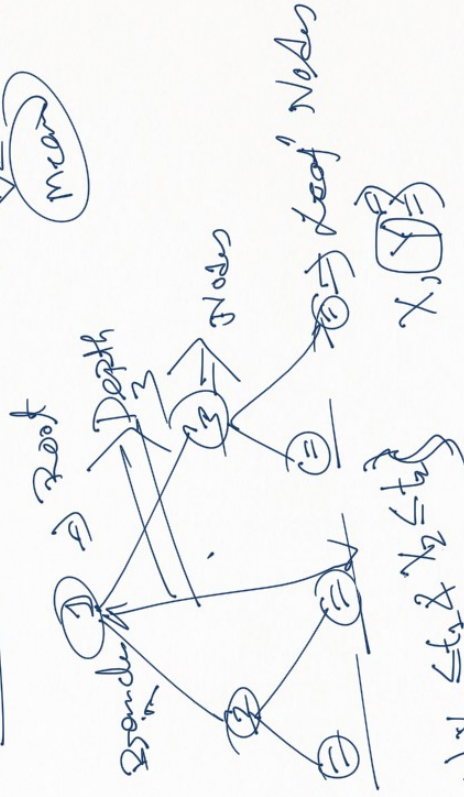


CART \Rightarrow Classification & Regression Trees.

Regression



\Rightarrow Recursively Split-
Binary Tree



X_{train}, y_{train}
 \Rightarrow mean

$R_1 \in \{x | x_1 \leq t_1 \& x_2 \leq t_2\}$
 $R_2 \Rightarrow$

Output = $\sum_{i=1}^J R_i$

$$\sum_{i=1}^J \sum_{n \in R_i} (\bar{y}_{R_i}) \Rightarrow \text{Output}$$

Based on feature values \Downarrow divide input space

$$(\bar{y}_{R_i}) \Rightarrow$$

$$\sum_{i=1}^J \sum_{n \in R_i} (y_i - \hat{y}_{R_i})^2$$

$$MSE \Rightarrow \sum_{j=1}^J R_j$$

$$R_1, R_2 = MSE \Rightarrow$$

$$\sum_{i \in R_1} (y_i - \hat{y}_{R_1})^2 + \sum_{i \in R_2} (y_i - \hat{y}_{R_2})^2$$

Dividing input space into K -part \Rightarrow Cross-validation

Cost Complexity Pruning \Rightarrow Prune a decision tree \Rightarrow

$$S \Rightarrow I_A - \sum_{j=1}^n \sum_{i \in R_j} \left(\frac{y_i - y_{R_j}}{x} \right) + \alpha \frac{|T|}{|I|} \Rightarrow \text{Depth of a tree}$$

Observed Parameter $i \in R_j$ y_{R_j}

K-Class

$$\text{Entropy} = - \sum_{k=1}^K \frac{P_k \log(P_k)}{P_k}$$

$$\text{Gini Index} = \sum_{k=1}^K P_k (1 - P_k)$$

$P_k = \text{Probability of } x_i \in \text{Class } k$

$$\frac{P(A) + P(B)}{P(A) \log P(A) + (1 - P(A)) \log (1 - P(A))} = 1 \Rightarrow P(B) = 1 - P(A)$$

$$\underline{0.1} \Rightarrow P(0) + P(1) = 1 = P(0) = 1 - P(1)$$

$$\underline{0.1} \Rightarrow P(0) + P(1) = 1 = P(0) = 1 - P(1)$$

$$-P_0 \log P_0 - P_1 \log P_1 = -P_0 \log P_0 - (1-P_0) \log (1-P_0)$$

$$-0.6 \log 0.6 - 0.4 \log 0.4$$

$$P(0) = 1 = \alpha_i \in \text{class } 0$$

all values

class 0 \Rightarrow all values

$$\frac{P_1}{0} \Rightarrow \text{Punk}$$

$$\text{class } 1 \Rightarrow \frac{P_0 = P_1 = \frac{1}{2} = 2^{-1}}{2} = \frac{1}{2} \log 2$$

$$1 \Rightarrow \text{maximal from } 0.5 \Rightarrow \text{maximal from } 0.5$$

$$0.5 \Rightarrow \text{maximal from } 0.5$$

$$I \Rightarrow -\frac{1}{2} \log 2 = 1$$

$$= \frac{1}{2} + \frac{1}{2} = 1$$

$$= 1 \log 1$$

$$\frac{P_0}{0} \Rightarrow \text{Punk}$$

$$\frac{P_1}{0} \Rightarrow \text{Punk}$$

$$\frac{P_2}{0} \Rightarrow \text{Punk}$$

$$\frac{P_3}{0} \Rightarrow \text{Punk}$$

$$0.4$$

$$= 0.6 \times 0.737 + 0.4 \times 1.327$$

$$= 0.97$$

$$P_0 = 1, P_1 = 0$$

$$= 0$$

$$\begin{aligned}
 \text{Gini Index} &\Rightarrow \sum_{k=1}^K P_k (1 - P_k) \\
 &\Rightarrow P_0(1 - P_0) + P_1(1 - P_1) = P_0(1 - P_0) + (1 - P_0)(1 - 1 + P_0) \\
 &= 2P_0(1 - P_0) \\
 P_0 &= 0.5 & 2 \times 0.9 \times 0.1 &= 0.18 \\
 P_0 &= 0.7 & 2 \times 0.5 \times 0.5 &= 0.5 \\
 & & 2 \times 0.7 \times 0.3 &= 0.42 \\
 P_0 &= 1 & & \\
 P_0 &= 0.9 & & = 2 \times 1 \times (1 - 1) = 0
 \end{aligned}$$

$$K=3 \quad P(A), P(B), P(C)$$

$$\text{Entropy} = - \frac{P(A) \log P(A) - P(B) \log P(B) - P(C) \log P(C)}{P(A)(1-P(A)) + P(B)(1-P(B)) + P(C)(1-P(C))}$$

$$\text{Gini} \Rightarrow \frac{3 \times 0.33 \times 0.67}{2 \times 0.67}$$

$$P(A) = P(B) = P(C) = 0.33$$

$$P(A) = 0, P(B) = 0, P(C) = 1$$

$$P(A) = 0$$

$$P(A) + P(B) + P(C) = 1$$

$$P(C) = 1 - P(A) - P(B)$$

Ensemble methods \Rightarrow

Combining multiple trials.

Combining multiple Diff. bags of input data

Bagging \Rightarrow Taking

Out of Bag Error \Rightarrow

Out of Bag score

$10 \Rightarrow$ $10 \Rightarrow$ 10
 $\textcircled{10}$ $\textcircled{15}$

$10 \Rightarrow$ Bag 1
 $\textcircled{10} \Rightarrow$ Out of Bag 1

Random Forest \Rightarrow

Input data bootstrapped.



Estimators.

Gender Age Class