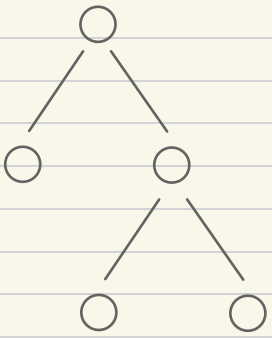
Trees



How do we decide the split ?

For classification

- Gini index
- Entropy

Gini index :

$$G = \sum_{k=1}^{K} \hat{p}_{mk}(1 - \hat{p}_{mk})$$

Entropy :

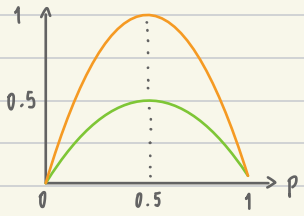
$$D = -\sum_{k=1}^{K} \hat{p}_{mk} \log \hat{p}_{mk}$$

$\hat{p}_{mk}$: proportion of observations in the m^{th} region that are from the k^{th} class

k → class

↳ subset of observations

- a small value in gini index mean the nodes contain mostly the observation from the same class.
- just like gini index, a small value in entropy corresponding to when nodes are pure



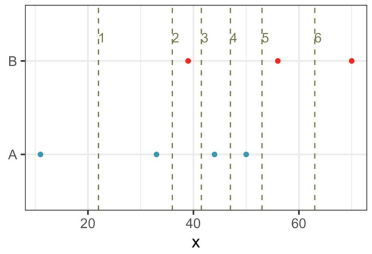
entropy

gini index

Example from the lecture

x	cl
11	A
33	A
39	B
44	A
50	A
56	B
70	B

All possible splits shown by vertical lines



$$D = -\sum_{k=1}^{K} \hat{p}_{mk} \log \hat{p}_{mk}$$

- 2 subset, left or right
- 2 classes, A or B

For split 2:

Left : 2As, 0B

Right : 2As, 3Bs

Step 1 - calculates entropy for each subset

Left :

$\hat{p}_{LA} = 2/2$, $\hat{p}_{LB} = 0/0$

$D_L = -\{[\underbrace{1 \log(1)}_{\hat{p}_{LA}=1}] + [\underbrace{0 \log(0)}_{\hat{p}_{LB}=0}]\} = 0$

Right :

$\hat{p}_{RA} = 2/5$, $\hat{p}_{RB} = 3/5$

$D_R = -\{[\underbrace{0.4 \log(0.4)}_{\hat{p}_{RA}=0.4}] + [\underbrace{0.6 \log(0.6)}_{\hat{p}_{RB}=0.6}]\}$

$= 0.673$

Step 2 · Combine the weighted sum

$D = 2/7\, D_L + 5/7\, D_R$

$= 0.673$

For Split 5 :

Left : 4As, 1B

Right : 0A, 2Bs

Left :

$\hat{p}_{LA} = 4/5$, $\hat{p}_{LB} = 1/5$

$D_L = -\{[0.8 \log(0.8)] + [0.2 \log(0.2)]\} = 0.5$

Right :

$\hat{p}_{RA} = 0/2$, $\hat{p}_{RB} = 2/2$

$D_R = -\{[0 \log(0)] + [1 \log(1)]\}$

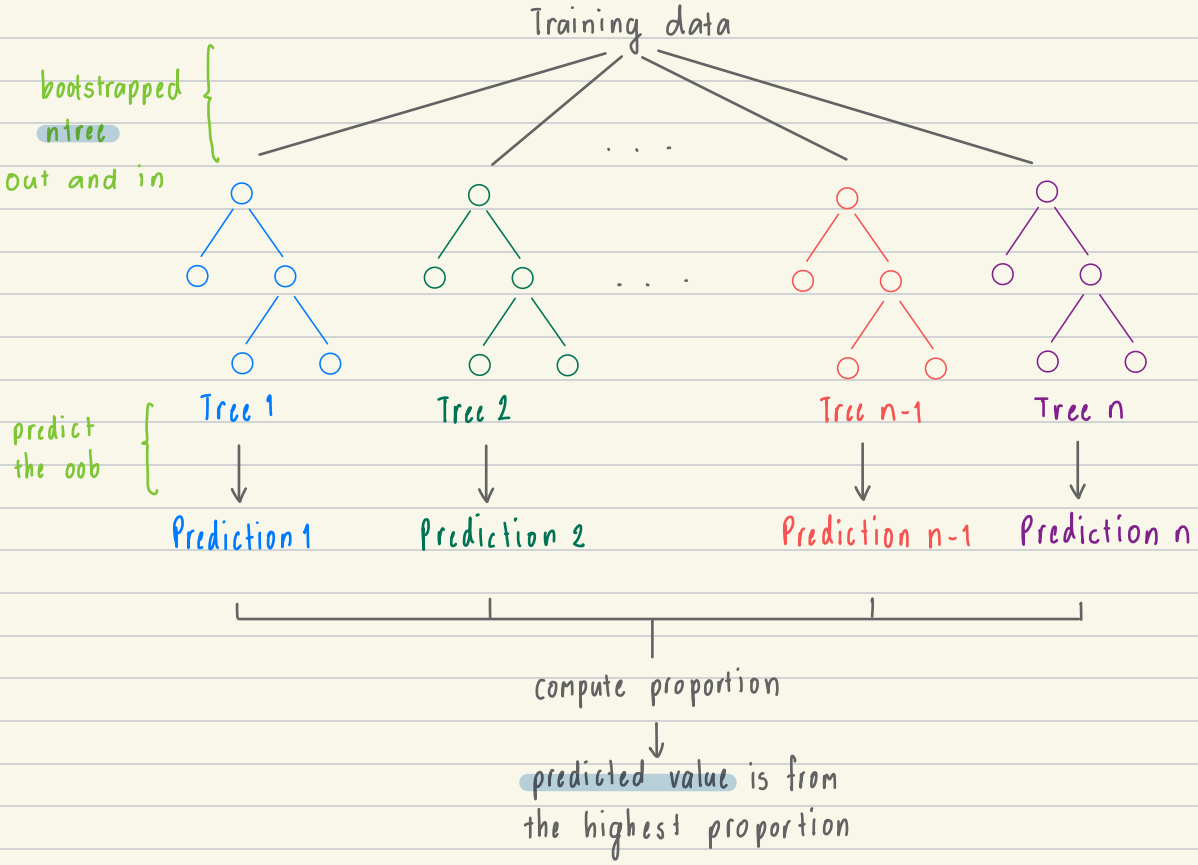
$= 0$

$D = 5/7\, D_L + 2/7\, D_R$

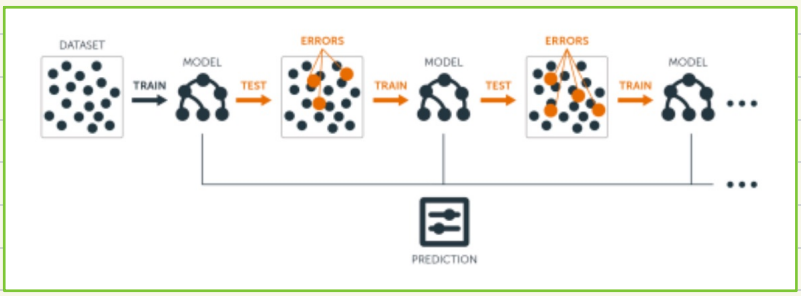
$= 0.3571$

Srathi (2024) Understand Random Forest Algorithms With Examples (Updated 2024)

Random Forest



Boosted trees



- boosted trees can lead to overfitting, while random forest does not lead to overfitting