

Xgboost ML Algorithm (Classification) →

- Both step help us to find out which feature should we select, whether salary is the best feature to select and split my decision tree.

$\hat{y} = 0.5$ → Binary Classification

Dataset	Salary	Credit	Approval	R1	\hat{y}	R2
→	<=50K	B	0	-0.5	0.52	-0.48
	<=50K	G	1	0.5	0.58	0.42
	<=50K	G	1	0.5	—	—
	>50K	B	0	-0.5	—	—
	>50K	G	1	0.5	—	—
	>50K	N	1	0.5	—	—
	<=50K	N	0	-0.5	—	—

Steps

- Construct a base Model
- Construct a Decision Tree with root.

- Calculate Similarity Weight

$$= \frac{(\sum \text{Residual})^2}{n}$$

$$\text{Cover Value} \leftarrow \frac{\sum P_r(1-P_r) + \lambda}{2}$$

- Calculate Gain

Hyper parameter

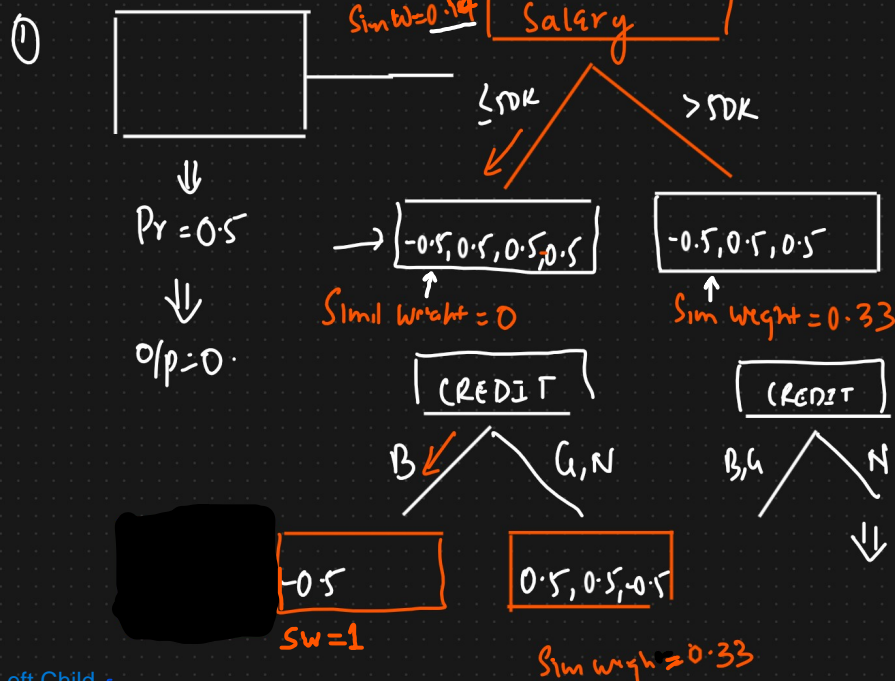
$[-0.5, 0.5, 0.5, -0.5, -0.5, 0.5, 0.5]$

For New Test data :-
1st of all it pass through base model. We can't pass probability directly, so use $\log(\text{odds})$

$$\log(\text{odds}) = \log\left(\frac{P}{1-P}\right)$$

For 1st record :-

$$\log(\text{odds}) = \log\left(\frac{0.5}{0.5}\right) = 0$$



Left Child

$$\text{Similarity}(L) = \frac{0.25}{0.25} = 1$$

Right Child

$$\text{Similarity}(R) = \frac{0.25}{0.75} = 0.33$$

$$\text{Gain} = 1 + 0.33 - 0 = 1.33$$

- Cover value = $\sum P_r(1-P_r)$
Whenever Similarity Weight is less than cover value, we slice down the decision tree.

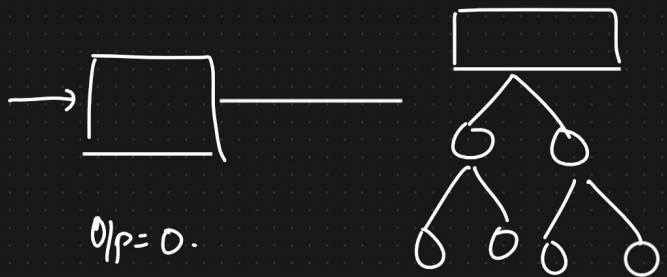
$$\text{Similarity weight (dc)} = \frac{\sum (\text{Residual})^2}{\sum \text{Pr}(1 - \text{Pr})}$$

$$= \frac{[(-0.5 + 0.5 + 0.5 - 0.5)^2]}{0.75} = 0$$

$$1 \leftarrow [0.5(1 - 0.5) + 0.5(1 - 0.5) + 0.5(1 - 0.5) + 0.5(1 - 0.5)]$$

$$\text{Gain} = 0 + 0.33 - 0.14 = 0.21$$

Final o/p



$$\text{New Data} \rightarrow = \sigma \left(0 + \overset{\text{Similarity weight}}{\alpha(1)} \right) \quad \alpha = 0.1 \text{ ; Learning Rate}$$

Logistic function \leftarrow Sigmoid Activation

$$= \sigma(0 + (0.1)(1))$$

$$= \frac{1}{1 + e^{-0.1}} = \underline{\underline{0.52}}$$

Second Record

$$\text{o/p} = \sigma(0 + \alpha(0.33))$$

$$= \sigma(0 + 0.1(0.33))$$

$$= \frac{1}{1 + e^{-0.033}} = \underline{\underline{0.508}}$$

Xgboost classifier



$$O/p = \sigma \left(\underset{\text{Score}}{\text{Base learner}} + f_1(DT1) + f_2(DT2) + f_3(DT3) \right).$$