

Gradient Boosting

→ It is a supervised machine learning algorithm which is used to solve both classification and regression problem by using boosting technique.

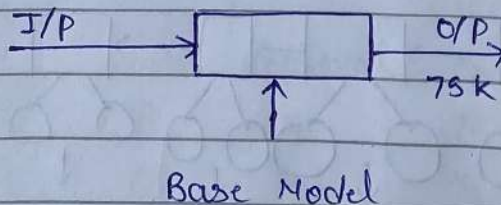
* working of Gradient Boosting :

Dataset :

x_1	x_2	y
Exp	Degree	Salary
2	B-E	50k
3	Master's	70k
5	Master's	80k
6	PHD	100k

① Create a Base Model :

$$\text{Average of output} = \frac{[50k + 70k + 80k + 100k]}{4} = 75k$$



② Compute the Residual Error :

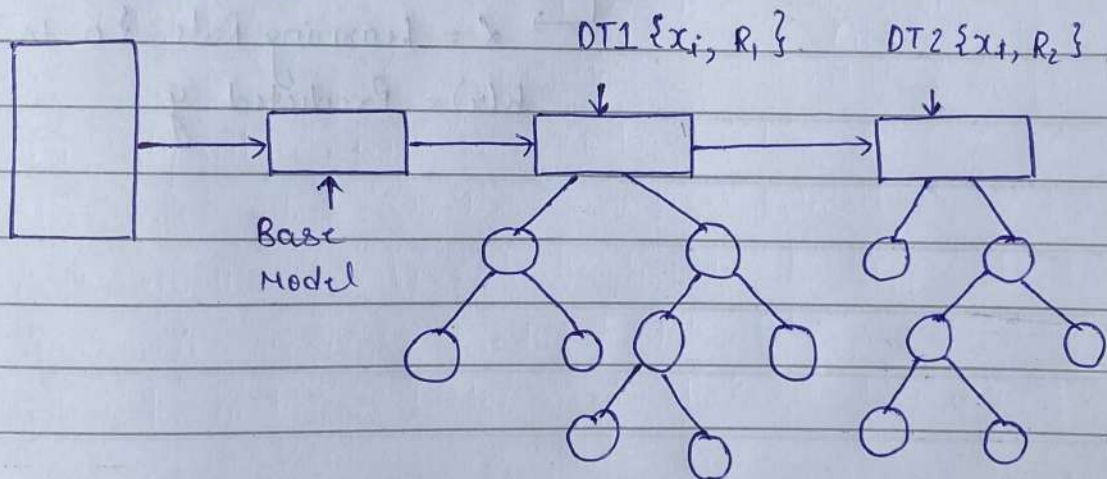
$$\text{Residual Error} = y - \hat{y}$$

Average / Base Model Prediction

Exp	Salary	$(y - \hat{y})$ R_1	Predicted R_2	$(\hat{y} + R_2)$ \hat{y}	$(y - \hat{y})$ R_3
2	60k	-25k	-23k	74.77	-24.77
3	70k	-5k	-3k	74.97	-4.97
5	80k	5k	3k	75.03	4.97
6	100k	25k	20k	75.2	24.8
	$\bar{y} = 75k$				

Predicted from DT1

③ Construct a decision tree. considers the input x_1 and output R_2 :



Predicted output : Base M o/p + DT o/p
 $= 75 + (-23) = 52 \rightarrow \text{very close \{overfitting\}}$

Therefore, Predicted output : BM o/p + α DT o/p
 \downarrow learning rate $\{0 \text{ to } 1\}$
 0.01

Records	Predicted output
1	$75 + 0.01(-23k) = 75 - 0.23k = 74.77k$
2	$75 + 0.01(-3k) = 75 - 0.03k = 74.97k$
3	$75 + 0.01(3k) = 75 + 0.03k = 75.03k$
4	$75 + 0.01(20k) = 75 + 0.2k = 75.2k$

Note: This process of creating decision tree and predicting value then calculating residual errors will keep on repeating until prediction become close or accurate.

$$f = \alpha_0 h_0(x) + \alpha_1 h_1(x) + \alpha_2 h_2(x) + \dots + \alpha_n h_n(x)$$

$$f = \sum_{i=0}^n \alpha_i h_i(x)$$

$\alpha =$ Learning Rate $\{0 \text{ to } 1\}$
 $h(x) =$ Predicted y

