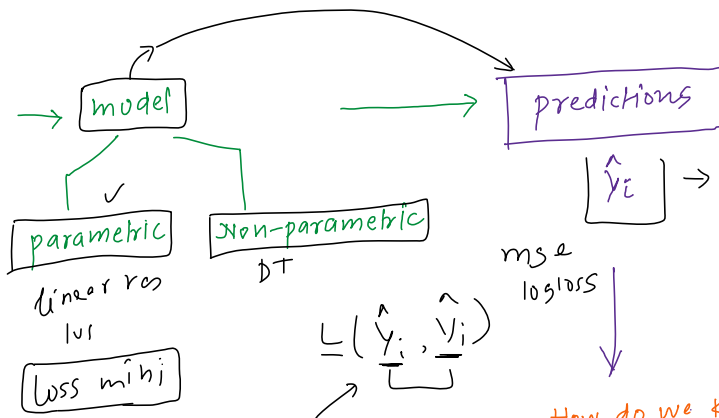


$x_i$	$y_i$
cgpa	iq
7.2	80
8.1	110
5.6	90
7.3	95
package	
	3
	4
	12
	19

Xgboost  $\rightarrow$  2 ideas (1st idea)



Problems with just keeping the loss function

overfitting

(Loss function)  
 $L(\theta)$

Objective function

$$obj(\theta) = L(\theta) + \Omega(\theta)$$

Training Loss  
(measures how well the model fits on training data)

Regularization  
(measures complexity of the model)

- Ridge regression:  $\sum_{i=1}^n (y_i - w^T x_i)^2 + \lambda \|w\|^2$ 
  - Linear model, square loss, L2 regularization
- Lasso:  $\sum_{i=1}^n (y_i - w^T x_i)^2 + \lambda \|w\|_1$ 
  - Linear model, square loss, L1 regularization
- Logistic regression:

$$\left[ \sum_{i=1}^n [y_i \ln(1 + e^{-w^T x_i}) + (1 - y_i) \ln(1 + e^{w^T x_i})] \right] + \lambda \|w\|^2$$

log loss

Helps in Bias Variance Tradeoff

$$L(\theta) + \Omega(\theta)$$

Regularization

parametric models

linear  $\rightarrow$  neural networks

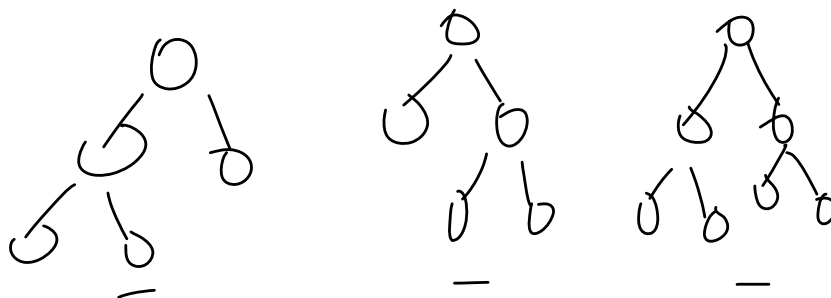
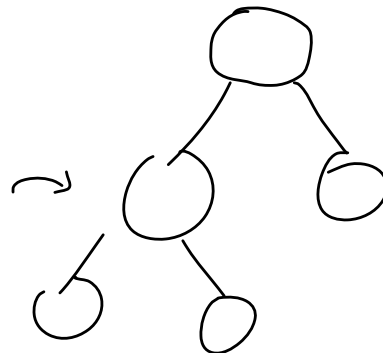
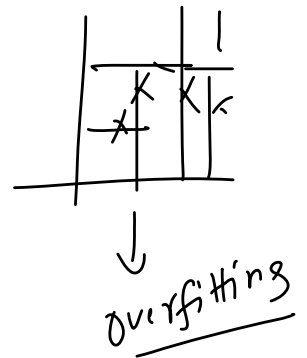
tree based models

cgpa | iq | package

Random forest

2<sup>nd</sup> interim

Decision tree



→ random forest

5 kpa

Stage wise additive modelling

cgpa   iq   package	pred	p r	pred
-	5	-2	-1.5
-	5	8	6
-	5	-1	0
-	5	-3	-4

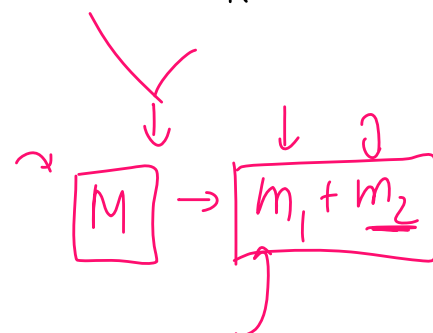
$m_2$

gradient

boosting

xgboost

$$m_1 + m_2$$



$$5 + (-1.5) \rightarrow 3.5$$

$$5 + 6 \rightarrow 11$$



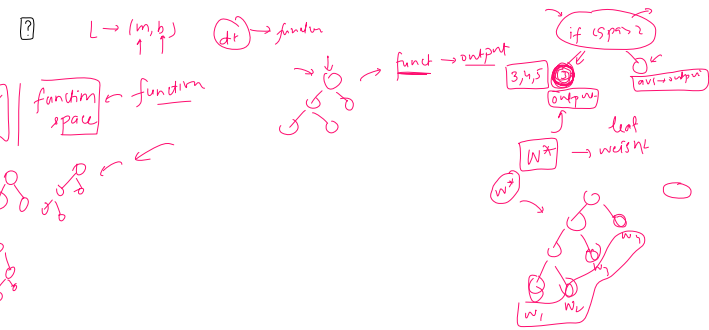
math / objective Object  
↳ regularization

find value of  $w$ 's that minimizes  $L^{(t)}$

back in the eq<sup>6</sup>

→ Similarity score for a tree structure (2)

$$\rightarrow \mathcal{L}_{split} = \frac{1}{2} \left[ \frac{(\sum_{i \in \mathcal{I}_L} g_i)^2}{\sum_{i \in \mathcal{I}_L} h_i + \lambda} + \frac{(\sum_{i \in \mathcal{I}_K} g_i)^2}{\sum_{i \in \mathcal{I}_K} h_i + \lambda} - \frac{(\sum_{i \in \mathcal{I}_P} g_i)^2}{\sum_{i \in \mathcal{I}_P} h_i + \lambda} \right] - \gamma$$



# XGBoost High Level Overview

24 March 2024 07:18



# 1. Lambda (& Alpha)

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## 2. Gamma

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### 3. Shrinkage

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## 4. Max Depth

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## 5. Max Leaves

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## 6. Min Child Weight

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## 7. Subsample

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## 8. Col Subsample(ByTree/ByLevel/ByNode)

24 March 2024 00:37