

Regularisation

① Underfitting

② Overfitting

Bias & Variance \Rightarrow Tradeoff \Rightarrow Low Bias & Variance

↓
Stability of the model to
learn properly \Rightarrow assumptions / Inductive bias

① low bias \Rightarrow fewer assumptions \Rightarrow

② High bias \Rightarrow ~~few~~ more assumptions \Rightarrow Underfitting



Linear Regress
Assumptions
① linearity b/w
data

reduce high bias

\Rightarrow Neural Nets

* More complex

* No. of features

* Increase training size.

Variance: Variance is the measure of spread
from its mean.

Variance in the model is how much sensitive
it is to variations in my training data.
... how much can my model adjust to

i.e., more
variance in my train

low variance :- less sensitive to change in
my train \Rightarrow high bias

high variance :- Overfitting highly sensitive to
change in my train set / Overfitting

reducing high variance :-

- (1) Cross Val - Scores & tune param
- \Rightarrow (2) feature selection / relevant features
- (3) Regularization
- (4) Ensemble learning \Rightarrow RF - DT
- (5) Early Stopping \rightarrow N.N
- (6) Simplifying the model \rightarrow N.N

Error

Optimal solution

Underfitting
zone

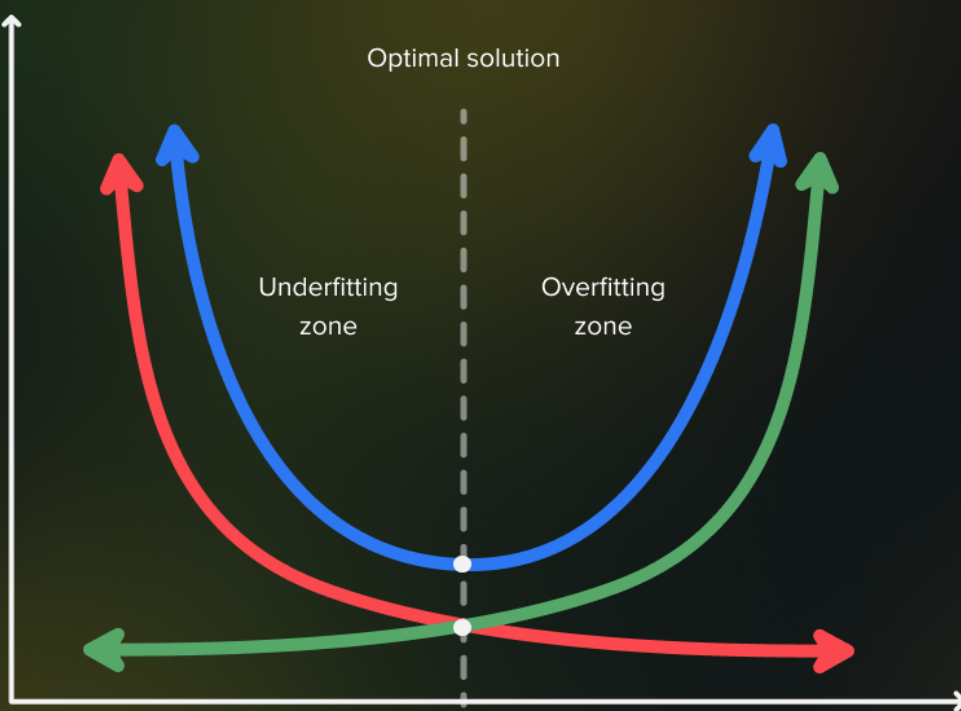
Overfitting
zone

● Bias²

● Total error

● Variance

Model Complexity



Regularization:

ML \Rightarrow Critical Thinking & Creativity

Overfitting \Rightarrow Optimization algorithm \Rightarrow Parameter $\Rightarrow \min(J(\theta_1, \theta_2))$

reduce.

Means \Rightarrow Model Lacks generalization

Adjust my Error/Cost in train to get better results in val?

I compromise Score \Rightarrow

Train \Rightarrow Achieve better perfor

Val/Test \Rightarrow Generalization

Regular:

Penalize my cost fu for

L_1 -lasso

$L_2 \rightarrow$ Ridge

Elastic Net

Least Absolute Shrinkage & Selection Operator.

$$\arg\min \left(\frac{1}{2} \|y - \theta * X\|^2 + \lambda \|\theta\|_1 \right)$$

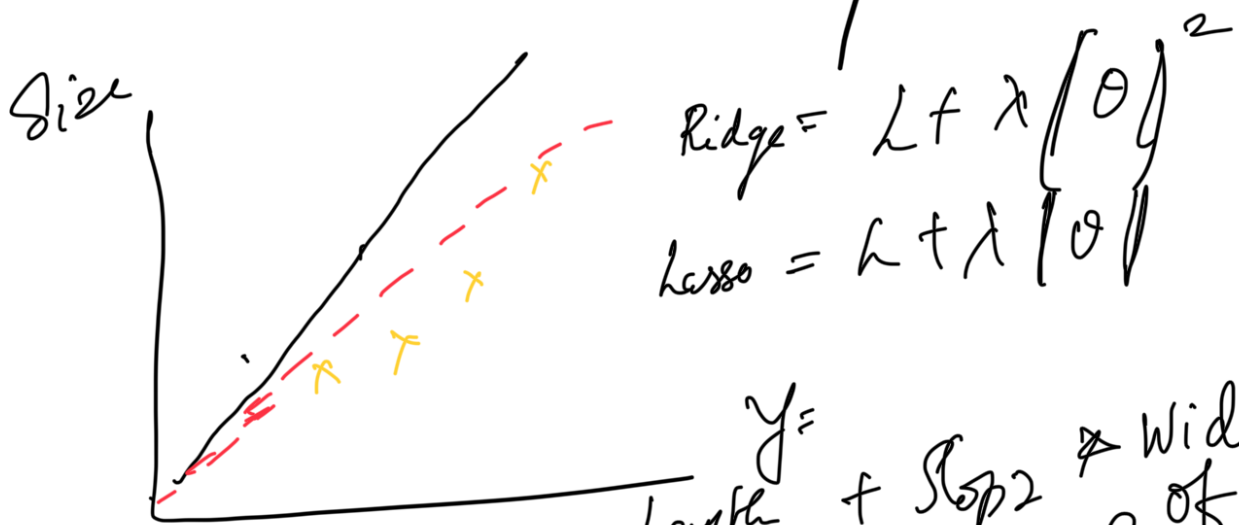
$$L + \lambda \left(\frac{1}{2} \|\theta\|^2 + \alpha \|\theta\|_1 \right)$$

$$\arg\min \left(\frac{1}{2} \|y - \theta * X\|^2 + \lambda \|\theta\|_1 \right)$$

Slope & Intercepts

$$h = \arg\min \left(\frac{1}{2} \|y - \theta * X\|^2 \right)$$

params



$$\text{Ridge} = L + \lambda (\theta)^2$$

$$\text{Lasso} = L + \lambda |\theta|$$

$\theta = \text{Slope}_1 \times \text{height} + \text{Slope}_2 \times \text{width of the tree}$
 $+ \text{place} \times \text{Slope}_3 + \text{fall} \times \text{Slope}_4$
 $+ \text{Slope}_5 \times \text{day of the week}$

Ridge

$$L + \lambda (\text{Slope})^2$$

Lasso

$$L + \lambda |\text{Slope}|$$

$$\lambda = 0.5$$

$$\lambda = 5000$$

Ridge



2, 0, 0, 0, 2, ...

Slope n^2



Ridge \Rightarrow Slope values $\rightarrow 0 \neq 0$

Lasso \Rightarrow Slopes values $\rightarrow 0 = 0$

