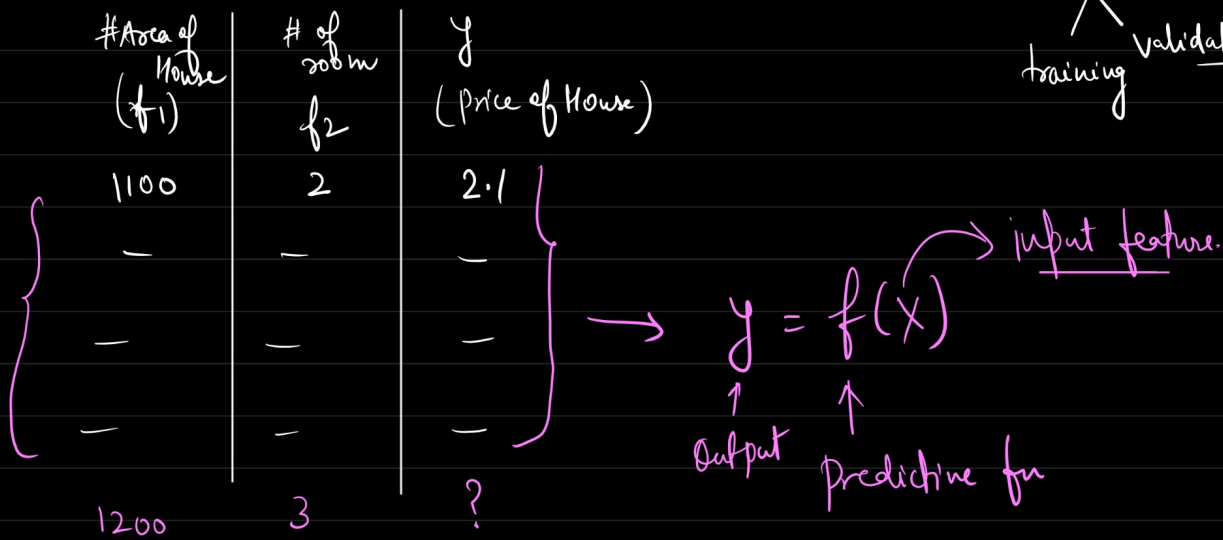
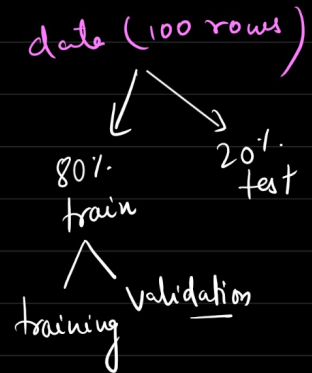


- ① train - test split
- ② Model training and performance
- ③ Overfitting and underfitting
- ④ Bias and Variance.

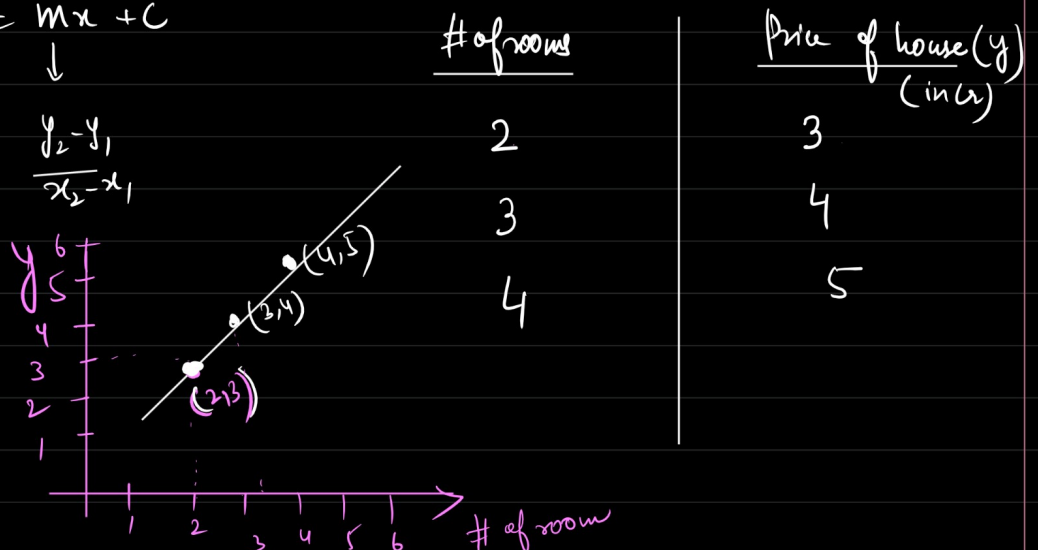
train - model training.

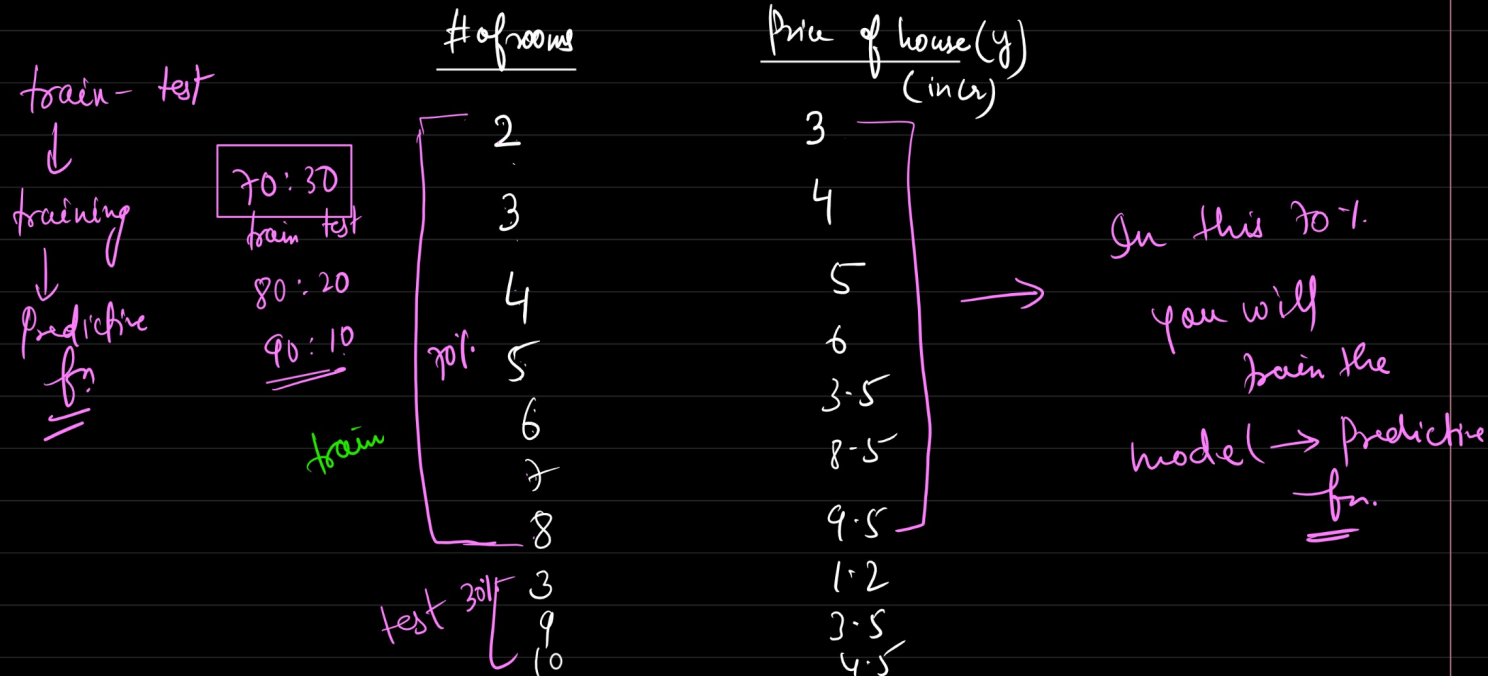


* Model training → given the data estimate the prediction function by minimising the error.

eqn of line → $y = mx + c$

$y = mx + c$

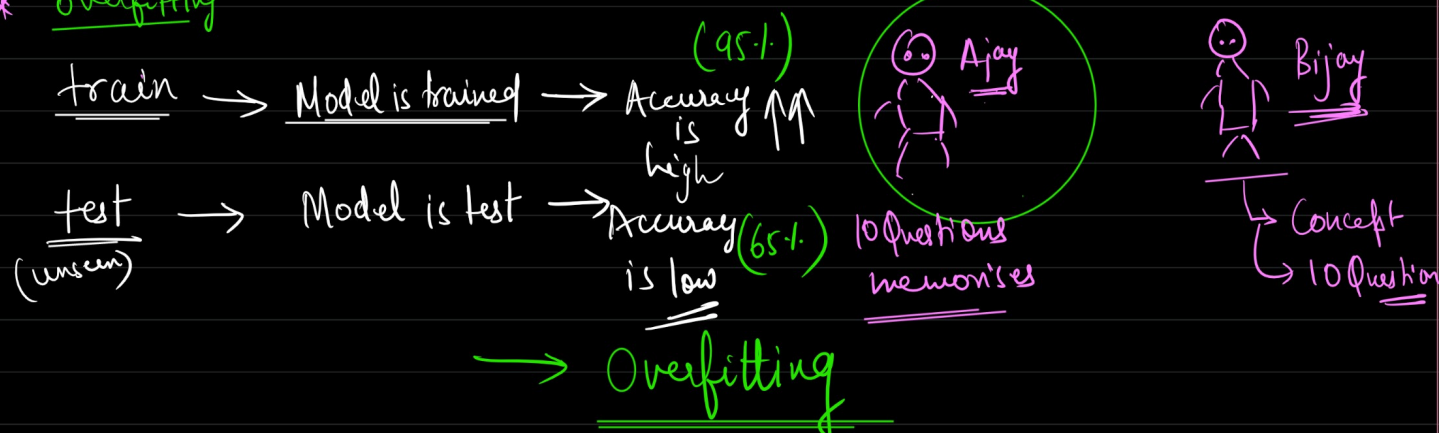




* Model performance ↑ (Accuracy should be high)

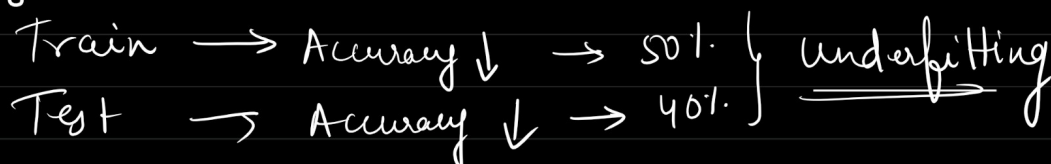
Viswa sir (final)

* Overfitting

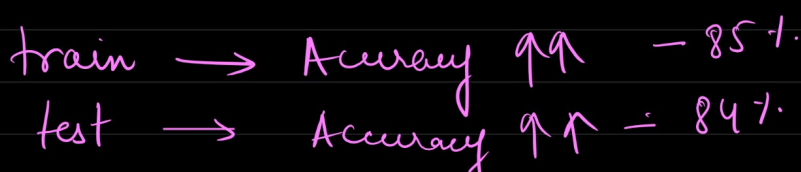


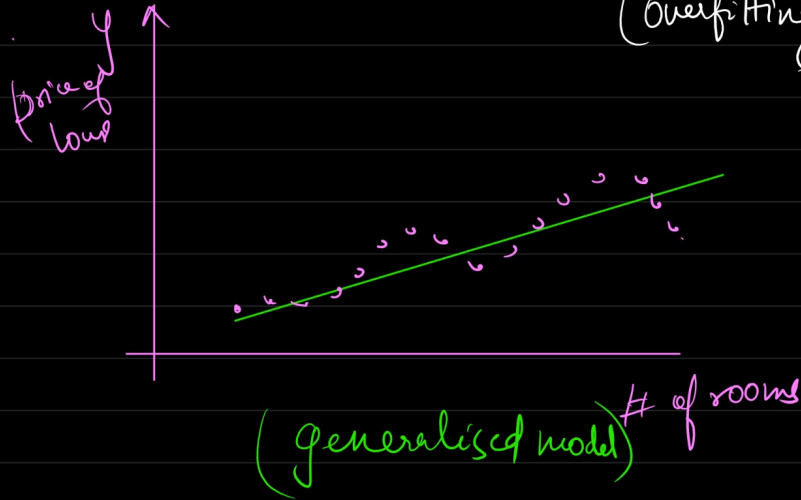
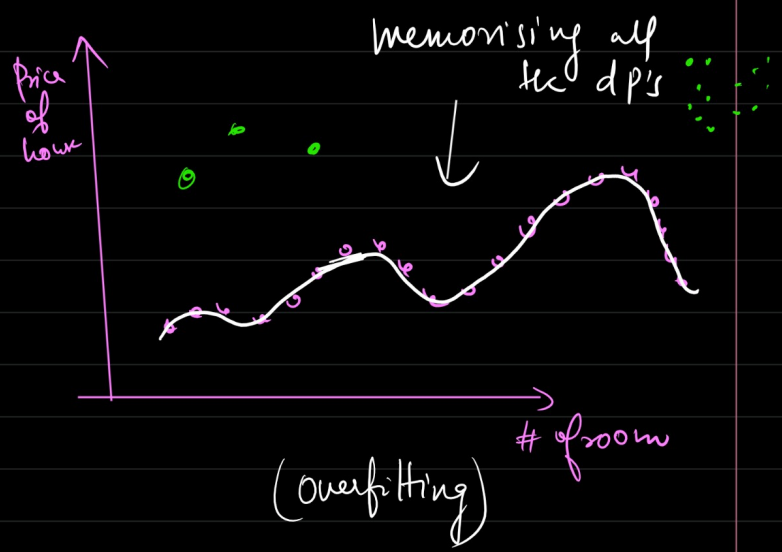
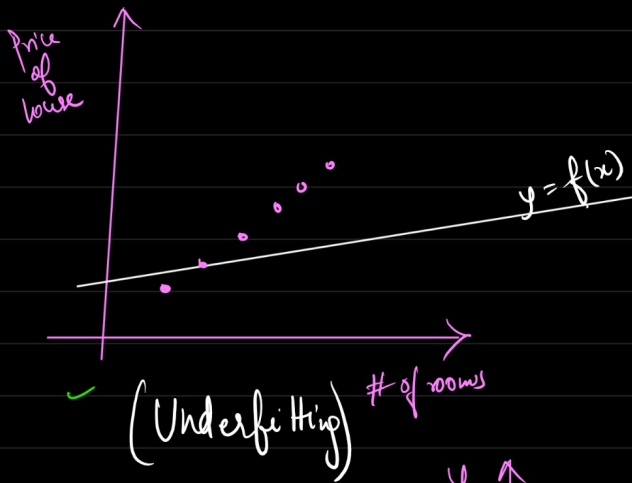
* Model performs well on train data, but worse on test data.

* Underfitting



* Generalised model





* Bias/Variance



# of rooms	Price of house y (in cr)	$y = f(x)$ train	y_{train} Prediction	Errors
2	3		2.9	3-2.9
3	4		4.1	4-4.1
4	5		3.2	5-3.2
5	6		6.1	
6	3.5		2.3	
7	8.5		2.2	
8	9.5		3.5	
3	1.2		3	1.2-3
9	3.5		4	3.5-4
10	1.1		5	4-5-5

training error

testing error

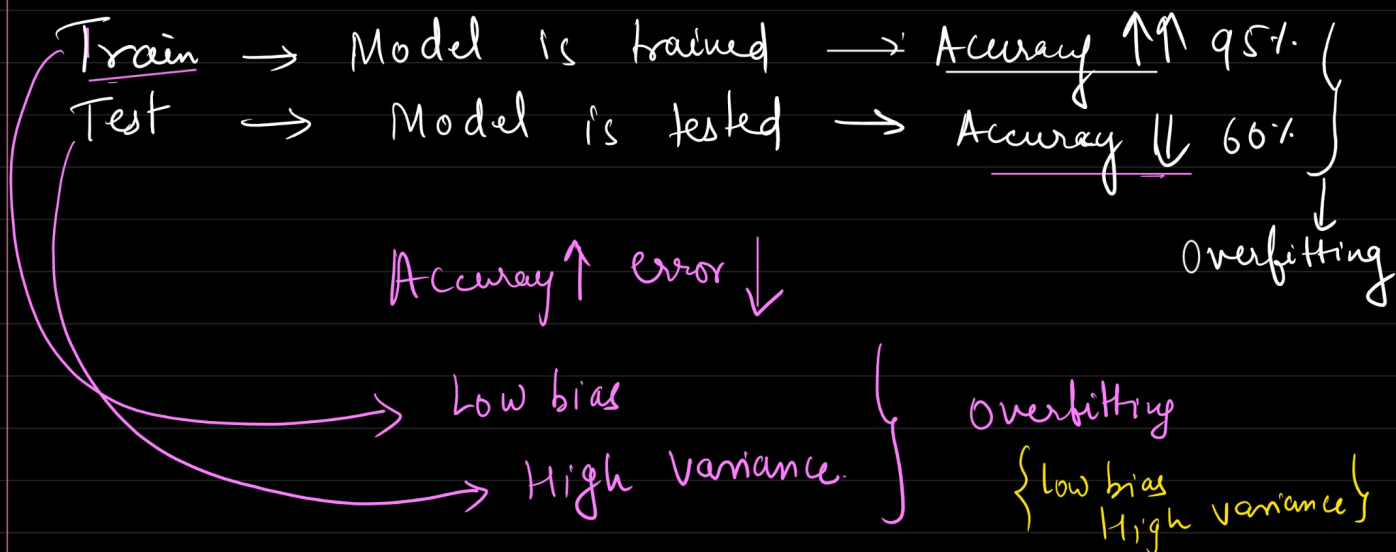
* training error is also known as bias

High training error means high bias.

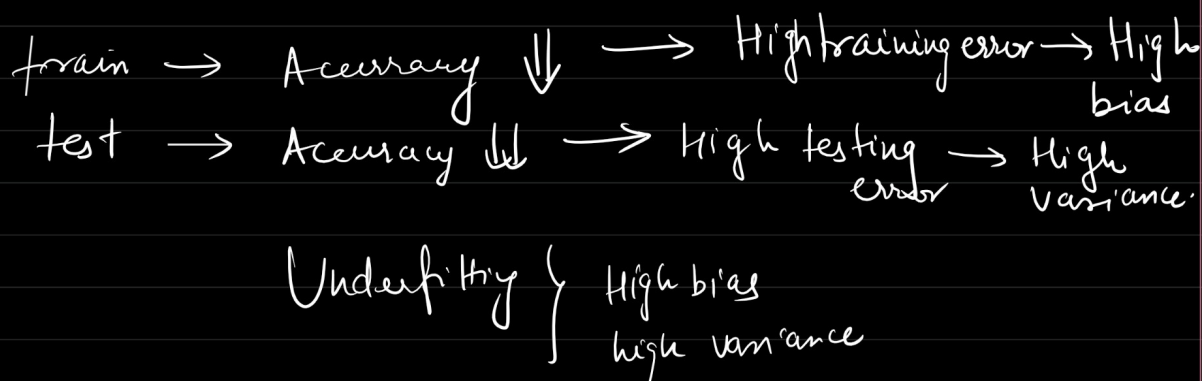
* testing error is also known as variance.

High testing error means high variance.

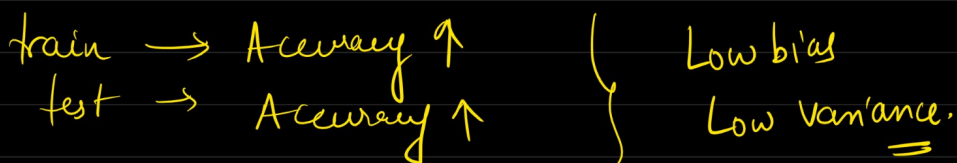
Overfitting



Underfitting



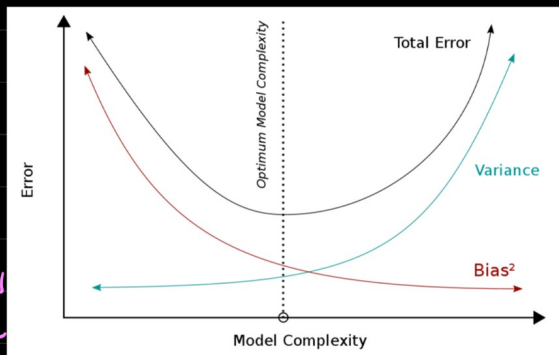
* Generalised model



Bias/variance trade off.

Underfitting

- Bias is high
- Var is high
- Some more training data
- Use some more features
- Use some other ML Algorithms.



Error

Bias is high

Variance is low

Total Error

Variance

{ high variance }
{ low bias }

Bias²

low bias, low variance

Overfitting

- Select relevant feature.
- Don't learn noise
- Use some Algorithms.