

Lecture 20: Model Evaluation

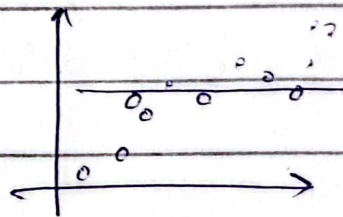
why :-
(i) to check the performance of model
(ii) model can have problems

underfit
overfit

underfit → training & testing is less.
→ simple algorithm

solution (make it complex)

model fails to capture the underlying pattern of the data.

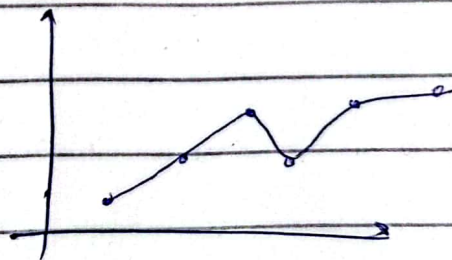


Trying to fit a straight curve the data that has clear curve.

Overfitting

↳ understand training data too much & have poor performance in unseen data.

- memorization not generalization
- poor performance on new data.



⇒ good model is balance b/w underfitting & overfitting

our Goal \Rightarrow appropriate fitting

• Trade of b/w optimizer & generalization

consequences (Both lead to unreliable predictions)

underfitting \Rightarrow high bias model consistency gets it wrong
overfitting \Rightarrow high variance performance wildly depends on specific data.

The metrics

① Accuracy \rightarrow (not good when data is not balanced)
 \downarrow useful when you have balanced dataset (ratio)

Evaluation for Classifier Metrics

① Accuracy

② Confusion Matrix

(describe complete performance of model)

	+ve	-ve	
+ve	TP (i)	FP (ii)	Predicted
-ve	FN (iii)	TN (iv)	
	actual		

TP \Rightarrow (actual & predicted True)

FP \Rightarrow (actual True Model predicted false)

FN \Rightarrow actual false model predicted true

TN \Rightarrow (given -ve model predict -ve)

(2.1)

Precision

نسبة الإيجابيات الحقيقية (Positive class) إلى الإيجابيات المتوقعة

$$\text{Precision} = \frac{\text{no. of True Positive}}{\text{True positive} + \text{false +ve}}$$

(minimize false positive)

2.2

Recall

($\frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$ (actual +ve))

(minimize false negatives)

$$\text{Recall} = \frac{\text{True positives}}{\text{True positive} + \text{False Negative}}$$