

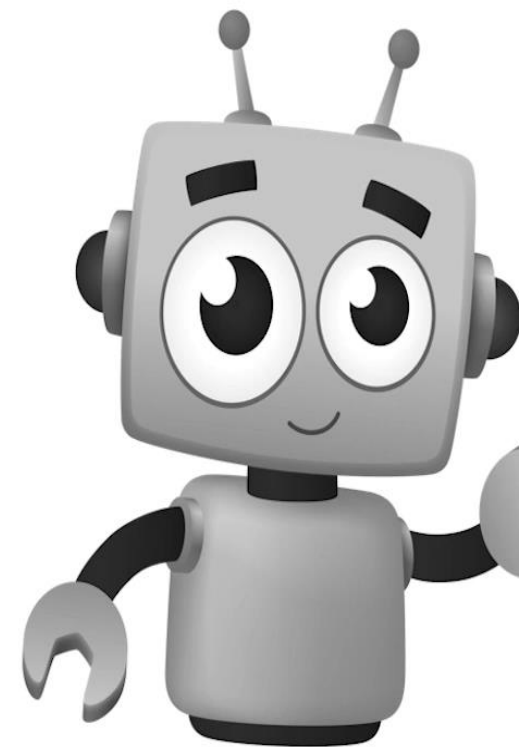
Evaluation Measures of an Algorithm

Zahoor Tanoli (PhD)
CUI, Attock

WHY IS ACCURACY NOT A GOOD METRIC TO EVALUATE CLASSIFIERS?

130 cr
D → 4 (130 cr - 4) MNIST → 6 1 2 → True
7 9 $\frac{4}{5} = 80\%$
Dumb.

- We saw accuracy as a metric to evaluate classifiers
- Accuracy is not always the best way to evaluate classifiers (especially in skewed datasets)
- Evaluating a classifier is trickier than evaluating a regressor



WHAT IS A CONFUSION MATRIX?

- It is a much better way to evaluate classifiers
- How does it work?
- Just count the number of times instances of class A are classified (or confused) as class B

	Predicted Neg	Predicted Positive	
Neg Actual	8 5	6 5	9 7
Positiv Actual	2 2	2 2	

$$\begin{bmatrix} 4 & 3 \\ 2 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 7 & 1 & 0 \\ 0 & 2 & 2 \end{bmatrix}$$

HOW TO CREATE CONFUSION MATRIX?

- ✓ Get the set of predictions, so they can be compared to the actual targets.
- ✓ Each row in a confusion matrix represents an actual class, while each column represents a predicted class
- ✓ First row is actual negative class & second row is actual positive class.
- ✓ First column is predicted negative class & second column is predicted positive class.
- Best classifier is the one having only true positives and true negatives, ie. confusion matrix would have nonzero values only on its main diagonal (top left to bottom right):

$$y_{\text{target}} = [0 \ 1 \ 0 \ 1 \ 1 \ 0]$$

$$y_{\text{predicted}} = [1 \ 1 \ 0 \ 1 \ 0 \ 1]$$

0 predicted 1 predicted
 2
 2

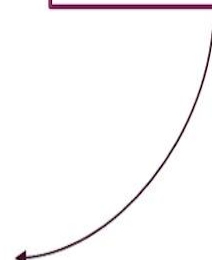
0 Actual
 1 Actual

row → ↓ column =

CONFUSION MATRIX

	Predicted negative <i>classifier ke hisaab se 2 nahi hai</i> ↓	Predicted positive <i>classifier ke hisaab se 2 hai</i> ↓
Actual negative <i>2 nahi hai →</i>	7 9 7 (TN) 1 5	8 (FP)
Actual positive <i>2 hai →</i>	2 2 (FN)	2 2 2 (TP)

Let us say we want to detect hand written 2s



CONFUSION MATRIX

	Predicted negative <i>classifier ke hisaab se 2 nahi hain</i> ↓	Predicted positive <i>classifier ke hisaab se 2 hain</i> ↓
Actual negative <i>2 nahi hain →</i>	7 9 7 7 1 5 (TN) count = 5	8 (FP) Count = 1
Actual positive <i>2 hain →</i>	2 2 Count = 2 (FN)	2 Count = 3 2 (TP) 2

Let us say we want to detect hand written 2s

$$\frac{3}{4}$$

WHAT IS PRECISION?

10 → 2

- Precision in simple terms means – What percent of positive predictions made were correct?
- In Mathematical terms,

$$\text{Precision} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$$

$$\frac{+ TP}{TP + FP}$$

WHAT IS RECALL?

- ✓ Recall in simple terms means – What percent of Actual positive values were correctly classified by your classifier?
- In Mathematical terms,

$$\text{Recall} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$$

$$= \frac{TP}{TP + FN}$$

PRECISION & RECALL FROM CONFUSION MATRIX

<u>Rows = Actual Values</u> ✓ <u>Columns = Predicted Values</u> ✓	Predicted negative	Predicted positive
Actual negative	7 9 7 1 5	8
Actual positive	2 2	2 2 2

Let us say we want to detect hand written 2s

$$\text{Precision} = \frac{TP}{TP + FP}$$

$$\text{Recall} = \frac{TP}{TP + FN}$$

FI- SCORE

99.9%

0.1%

- It is convenient to combine the performance of a classifier (precision and recall) into a single metric called the F1 score.
- F1-Score is the harmonic mean of precision and recall
- In Mathematical terms,

$$F1\text{-Score} = \frac{2 \times \text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

f_1

$$= \frac{\frac{1}{\text{prec}} + \frac{1}{\text{recall}}}{2}$$

10,000

Quality

Ambience

Precision

Recall

f_1



Thank You