

Q1 1

5 Points

A classifier that gets 100% accuracy on the training set and 65% accuracy on test set is better than a classifier that attains 85% accuracy on the training set and 75% accuracy on the test set. (True/False) Why?

False

The second classifier has better test accuracy.

The second classifier reflects the true accuracy.

The first classifier is overfitting.

Q2 2

5 Points

A and B are two events. If $P(A,B)$ decreases while $P(A)$ increases, how $P(B|A)$ acts (increase or decrease) Why?

decrease

$$P(A,B) = P(A|B) P(B) = P(B|A) P(A)$$

$$P(B|A) = (P(A|B) P(B)) / P(A)$$

When $P(A, B)$ decreases, $P(B|A)$ decreasesWhen $P(A)$ increases, $P(B|A)$ decreasestherefore $P(B|A)$ is reduced**Q3 3**

10 Points

Suppose we want to compute 10-Fold Cross-Validation error on 100 training examples. We need to compute error N_1 times, and the Cross-Validation error is the average of the errors. To compute each error, we need to build a model with data of size N_2 , and test the model on the data of size N_3

☒ $N_1 = 10, N_2 = 90, N_3 = 10$

☐ $N_1 = 1, N_2 = 90, N_3 = 10$

☐ $N_1 = 10, N_2 = 100, N_3 = 10$

☐ $N_1 = 10, N_2 = 100, N_3 = 100$

Q4 4

20 Points

Suppose we are given the following dataset, where A, B, C are input binary random variables, and y is binary output whose value we want to predict.

How would a naive Bayes classifier predict y given this input: A = 1, B=0, C=1.

A	B	C	y
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	1
1	1	0	1
1	1	1	1
1	0	1	0
1	0	1	1

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(Q4)

$$P(Y=0) = \frac{4}{9} \quad P(Y=1) = \frac{5}{9}$$

$$P(A=1 | Y=0) = \frac{1}{4} \quad P(B=0 | Y=0) = \frac{3}{4} \quad P(C=1 | Y=0) = \frac{2}{4}$$

$$P(A=1 | Y=1) = \frac{4}{5} \quad P(B=0 | Y=1) = \frac{2}{5} \quad P(C=1 | Y=1) = \frac{3}{5}$$

$$P(A=1 | Y=0) P(B=0 | Y=0) P(C=1 | Y=0) P(Y=0) = \frac{1}{4} \times \frac{3}{4} \times \frac{2}{4} \times \frac{4}{9}$$

$$= \frac{1}{24}$$

$$P(A=1 | Y=1) P(B=0 | Y=1) P(C=1 | Y=1) P(Y=1) = \frac{4}{5} \times \frac{2}{5} \times \frac{3}{5} \times \frac{5}{9}$$

$$= \frac{2}{75}$$

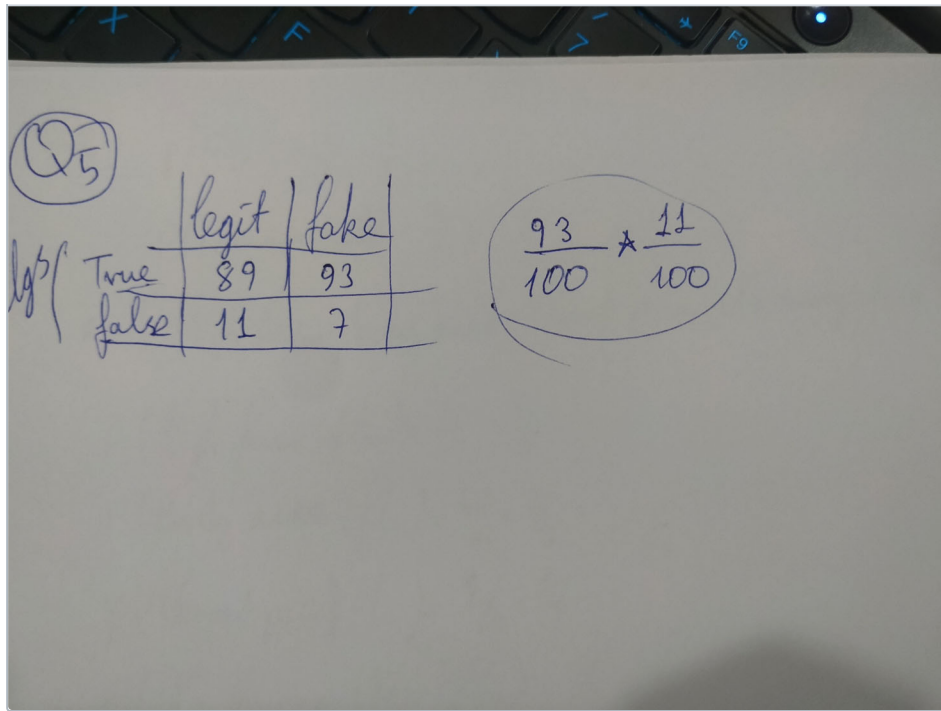
Q5 5

30 Points

Imagine that we develop an algorithm to predict fake reviews. Based on the previous experience, we know that 95% of the reviews are legit and 5% are fake. If a review is fake, there is a 93% chance that the algorithm predicts it as fake. If a review is legit, the algorithm classifies it as legit with 89% chance. What is the chance that the review is actually fake if the algorithm suggests it is fake.

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Q6 6

15 Points

Consider a naive Bayes classifier trained on the dataset given below. A new patient comes who has x =[High Fever, Body Ache and Throat Pain, but NO runny nose]. Calculate the posterior probability of having Flu given these symptoms

	Fever	Body Ache	Runny Nose	Throat Pain	Disease
1	High	Yes	No	Yes	Flu
2	High	Yes	No	No	Flu
3	High	No	Yes	No	Flu
4	Medium	Yes	No	No	Flu
5	Medium	No	No	No	Flu
6	High	Yes	No	Yes	Flu
7	Low	No	Yes	Yes	Common cold
8	Low	No	Yes	Yes	Common cold
9	Low	Yes	No	No	Common cold
10	Medium	No	Yes	Yes	Common cold

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Q6

$$P(\text{Flu}) = \frac{6}{10} = \frac{3}{5}$$

$$P(\text{common cold}) = \frac{2}{5}$$

$x = [\text{High fever, body aches, Throat pain, No runny nose}]$

$$P(\text{High fever} | \text{Flu}) = \frac{4}{6} = \frac{2}{3}$$

$$P(\text{Body aches} | \text{Flu}) = \frac{4}{6} = \frac{2}{3}$$

$$P(\text{Throat pain} | \text{Flu}) = \frac{2}{6} = \frac{1}{3}$$

$$P(\text{No Runny Nose} | \text{Flu}) = \frac{5}{6}$$

$$P(x | \text{Flu}) = \underbrace{P(\text{Flu})}_{3/5} \cdot \frac{2}{3} \cdot \frac{2}{3} \cdot \frac{1}{3} \cdot \frac{5}{6} \Rightarrow \frac{4}{54} = \frac{2}{27}$$

Q7 7

15 Points

Fill the given functions for the k-fold cross validation. (You can use `train_knn`, and `predict_knn` functions in your code)

```
def cross_validation(data,k)
```

```
...
```

```
return average_accuracy
```

```

(23) def cross_validation (data, k):
    acc_list = []
    piece = len(data) // k
    for i in range(1, k+1):
        # test data split
        test_data = data[(i-1)*piece : i*piece]
        piece_before_test_data = data[: (i-1)*piece]
        piece_after_test_data = data[i*piece : ]
        train_data = np.concatenate (piece_before_test_data,
                                     piece_after_test_data)
        sim_matrix = calc_distance (train, test)
        accuracy = calc_acc calc_acc ()
        acc_list.append (accuracy)

    average_accuracy = sum (acc_list) / len (acc_list)
    return average_accuracy

```

Quiz-2

● GRADED

STUDENT

MEHMET TAHA USTA

TOTAL POINTS

75 / 100 pts

QUESTION 1

1	5 / 5 pts
QUESTION 2	
2	5 / 5 pts
QUESTION 3	
3	10 / 10 pts
QUESTION 4	
4	20 / 20 pts
QUESTION 5	
5	5 / 30 pts
QUESTION 6	
6	15 / 15 pts
QUESTION 7	
7	15 / 15 pts