

In-class activities/Lab (IS 733)

Task 1: Understanding Naive Bayes and K-nearest neighbors

1a: Manually calculate prediction using **the Naive Bayes Model and K nearest neighbor, K=2; Euclidean Distance** for the test example for the following example:

- Use any random combination to test/report your probability

ID	Contains Link	Contains Money Words	Length	Class
1	Yes	Yes	Long	Spam
2	No	No	Short	Ham
3	Yes	No	Long	Spam
4	No	Yes	Short	Spam
5	Yes	Yes	Short	Spam
6	No	No	Long	Ham
7	Yes	No	Short	Ham
8	No	Yes	Long	Spam
9	Yes	Yes	Long	Spam
10	No	No	Short	Ham

→ k nearest neighbour (k=2) Calculations  
~~Eucledian distance~~

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classwork

Q1 a) Manually calculate prediction using the Naive Bayes model & k nearest neighbour k=2. Eucledian distance for the test example for following example.

Soln -

1] Test example

Contains link = Yes (CL)

Contains Money words = No (CM)

length = long. (L)

2] Naives Bayes Calculations.

a] calculate prior probabilities

$$P(\text{spam}) = \frac{\text{No of Spam Msg}}{\text{Total Msg}} = \frac{6}{10} = 0.6$$

$$P(\text{ham}) = \frac{\text{No of Ham Msg}}{\text{Total Msg}} = \frac{4}{10} = 0.4$$

b] Calculate condition probability.

i]  $P(\text{CL} = \text{yes} | \text{Spam})$

$$P(\text{yes} | \text{spam}) = \frac{\text{Spam with yes}}{\text{Total spam Msg}} = \frac{4}{6} = 0.67$$

ii]  $P(\text{CL} = \text{yes} | \text{ham})$

$$P(\text{yes} | \text{ham}) = \frac{\text{ham with yes}}{\text{Total ham Msg}} = \frac{1}{4} = 0.25$$

$$P(CM=No|spam)$$

$$P(No|spam) = \frac{\text{Spam with No}}{\text{Total spam msg}} = \frac{2}{6} = 0.33$$

$$P(CM=No|ham)$$

$$P(No|ham) = \frac{\text{Ham with No}}{\text{Total ham msg}} = \frac{3}{4} = 0.75$$

$$P(\text{Length}=\text{long}|spam)$$

$$P(\text{long}|spam) = \frac{\text{Spam msg with long}}{\text{Total spam msg}} = \frac{3}{6} = 0.5$$

$$P(\text{Length}=\text{long}|ham)$$

$$P(\text{long}|ham) = \frac{\text{Ham with long}}{\text{Total ham}} = \frac{2}{4} = 0.5$$

c) Apply Bayes Theorem

$$P(\text{spam}|x) = P(\text{spam}) \times P(x|\text{spam})$$

$$P(\text{ham}|x) = P(\text{ham}) \times P(x|\text{ham})$$

$$P(\text{spam}|x) = 0.6 \times 0.67 \times 0.33 \times 0.5 = 0.666$$

$$P(\text{ham}|x) = 0.4 \times 0.25 \times 0.75 \times 0.5 = 0.0375$$

Since  $P(\text{spam}|x) > P(\text{ham}|x)$ , the test example is classified as Spam.

3] K-Nearest Neighbors ( $k=2$ ) using Euclidean distance

→ Each feature is represented as

Contains link ( $\text{yes}=1, \text{no}=0$ )

Contains money word ( $\text{yes}=1, \text{no}=0$ )

Length ( $\text{Long}=1, \text{Short}=0$ )

So, the test example is (1, 0, 1).



i] Computing Euclidean distance.

$$d = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2}$$

calculating the distance from the test example (1,0,0) to all other points

ID	CL	cm	L	Class	Distance
1	1	1	1	Spam	$\sqrt{(1-1)^2 + (0-0)^2 + (0-0)^2} = 0$
2	0	0	0	ham	1.41
3	1	0	1	Spam	0
4	0	1	0	Spam	1.73
5	1	1	0	Spam	1.41
6	0	0	1	ham	1
7	1	0	0	ham	1
8	0	1	1	Spam	1
9	1	1	1	Spam	1
10	0	0	0	ham	1.41

ii] Find the nearest neighbors  
two closest points

- 1 ID 3 (distance = 0.0, Spam)
- 1 ID 1 (distance = 1.0, Spam)

Since Both are spam, test example is classified as Spam.

Final classification

Naive Bayes = Spam

K nearest neighbour = Spam

Thus, test example is Spam.

1b: write code (with AI assistant) to build a naive Bayes and KNN classifier. You can use the hamspam.csv to test it out.

[https://github.com/AryanJ09/IS733\\_Class/blob/main/01272025\\_CW/CW-03-03-2025/CLASSWORK\\_1B.ipynb](https://github.com/AryanJ09/IS733_Class/blob/main/01272025_CW/CW-03-03-2025/CLASSWORK_1B.ipynb)

## Task2: Understanding ROC and AUC

2a: Create a ROC (with AI assistant/Excel ) (Refer to roc\_data.csv)

Step1: Given the threshold (0.95,0.90,0.85,0.80,0.75,0.70), derive True Positive and False Positive

Step2: Calculate the True Positive Rate (TPR) and False Positive Rate (FPR), enter the values into the sheet

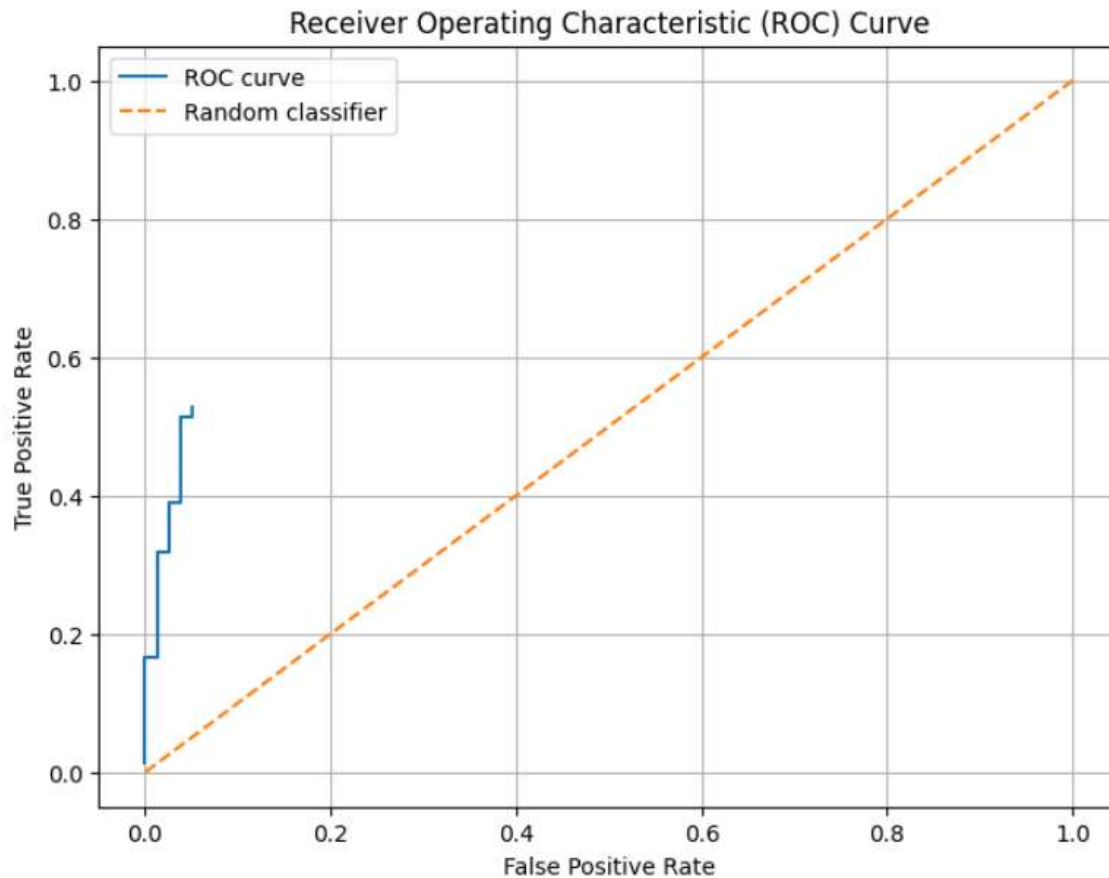
Step3: plot the set points (FRP, TPR) on the ROC diagram

2a]

classmate  
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Threshold value	TP	TN	FP	FN	TPR	FPR
0.95	13	74	4	11	0.5417	0.0513
0.90	15	73	5	9	0.6389	0.0641
0.85	18	73	5	7	0.7083	0.0641
0.80	19	73	5	6	0.7600	0.0641
0.75	20	72	6	6	0.7639	0.0769
0.70	21	72	6	5	0.8056	0.0769

[https://github.com/AryanJ09/IS733\\_Class/blob/main/01272025\\_CW/CW-03-03-2025/733\\_classwork\\_graph\\_2a.ipynb](https://github.com/AryanJ09/IS733_Class/blob/main/01272025_CW/CW-03-03-2025/733_classwork_graph_2a.ipynb)



[https://github.com/AryanJ09/IS733\\_Class/blob/main/01272025\\_CW/CW-03-03-2025/roc\\_dataaryan\\_q2a.csv](https://github.com/AryanJ09/IS733_Class/blob/main/01272025_CW/CW-03-03-2025/roc_dataaryan_q2a.csv)

2b. Write code (with AI assistant) to fit the model using your favorite classifier (NB, KNN, or Decision tree); using the hamspam.csv, ask to output an ROC curve and AUC score. (Hint: if you fit a decision tree, you might want to reduce max\_depth)

[https://github.com/AryanJ09/IS733\\_Class/blob/main/01272025\\_CW/CW-03-03-2025/CLASSWORK\\_2B.ipynb](https://github.com/AryanJ09/IS733_Class/blob/main/01272025_CW/CW-03-03-2025/CLASSWORK_2B.ipynb)

Submission to blackboard

1a and 2a: photocopy of your manual calculation

The rest of the task (1b, 2b): Python Notebook uploaded to GitHub and submit a link