

Bronco ID: |0|1|3|4|8|4|6|7|9|

Last Name: Francisco

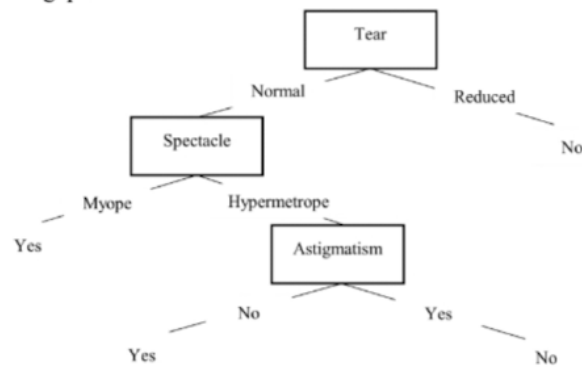
First Name: Serrano

1:

Part a:

A2:

1. [16 points] Considering that ID3 built the decision tree below after analyzing a given training set, answer the following questions:



- a) [12 points] What is the accuracy of this model if applied to the test set below? You must **identify each** True Positive, True Negative, False Positive, and False Negative for full credit. For instance: TP = 1,5 | TN = 2,3 ...

#	Age	Spectacle	Astigmatism	Tear	Lenses (ground truth)
1	Young	Hypermetrope	Yes	Normal	Yes
2	Young	Hypermetrope	No	Normal	Yes
3	Young	Myope	No	Reduced	No
4	Presbyopic	Hypermetrope	No	Reduced	No
5	Presbyopic	Myope	No	Normal	No
6	Presbyopic	Myope	Yes	Reduced	No
7	Prepresbyopic	Myope	Yes	Normal	Yes
8	Prepresbyopic	Myope	No	Reduced	No

— NO
— NO
— NO
— NO
— NO

#	Age	Spectacle	Astigmatism	Tear	Lenses (ground truth)
1	Young	Hypermetrope	Yes	Normal	Yes
2	Young	Hypermetrope	No	Normal	Yes
3	Young	Myope	No	Reduced	No
4	Presbyopic	Hypermetrope	No	Reduced	No
5	Presbyopic	Myope	No	Normal	No
6	Presbyopic	Myope	Yes	Reduced	No
7	Prepresbyopic	Myope	Yes	Normal	Yes
8	Prepresbyopic	Myope	No	Reduced	No

TP = 2,7
FP = 5
TN = 3,4,6,8
FN = 1

$$\text{accuracy} = \frac{TP + TN}{TP + TN + FN + FP} = \frac{2 + 4}{8} = \frac{6}{8} = 0.75$$

Part b:

$$b) \text{ Precision} = \frac{TP}{TP+FP} = \frac{2}{2+1} = \frac{2}{3}$$

$$\text{Recall} = \frac{TP}{TP+FN} = \frac{2}{3}$$

↓ actually relev. but misclassified

$$\text{Recall} = \text{precision} = \text{F1 score} = \frac{2}{3}$$

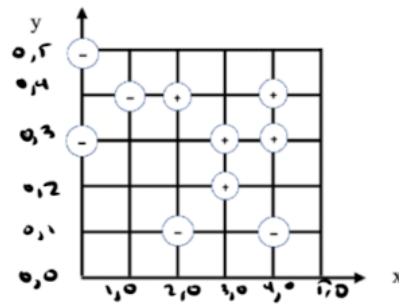
$$\begin{aligned} F1 &= 2 * \left(\frac{(\frac{2}{3})(\frac{2}{3})}{\frac{2}{3} + \frac{2}{3}} \right) \\ &= 2 * \left(\frac{\frac{4}{9}}{\frac{4}{3}} \right) \\ &= 2 * \left(\frac{1}{3} \right) \\ &= \frac{2}{3} \end{aligned}$$

2:

https://github.com/franserr99/cs4210/blob/main/a2/decision_tree_2.py

3:

3. [32 points] Consider the dataset below to answer the following questions:



- a. [4 points] What is the leave-one-out cross-validation error rate (LOO-CV) for 1NN? Use Euclidean distance as your distance measure and the error rate calculated as:

$$\text{error rate} = \frac{\text{number of wrong predictions}}{\text{total number of predictions}}$$

- b. [4 points] What is the leave-one-out cross-validation error rate (LOO-CV) for 3NN?
 c. [4 points] What is the leave-one-out cross-validation error rate (LOO-CV) for 9NN?
 d. [5 points] Draw the **decision boundary** learned by the 1NN algorithm.

points:

i=1	0,5 : ⊖
i=2	0,3 : ⊖
i=3	0,4 : ⊖
i=4	2,4 : ⊕
i=5	2,1 : ⊖
i=6	3,2 : ⊖
i=7	3,3 : ⊕
i=8	4,1 : ⊖
i=9	4,3 : ⊕
i=10	4,4 : ⊕

leave i=1 out (i=1 not test)

distances:

$$\begin{aligned} d_{1,2} &= ((0-0)^2 + (5-3)^2)^{1/2} = 2 \\ d_{1,3} &= ((0-1)^2 + (5-4)^2)^{1/2} = 1.41 \\ d_{1,4} &= ((0-2)^2 + (5-4)^2)^{1/2} = 2.23 \\ d_{1,5} &= ((0-2)^2 + (5-1)^2)^{1/2} = 4.47 \\ d_{1,6} &= ((0-3)^2 + (5-2)^2)^{1/2} = 4.24 \\ d_{1,7} &= ((0-3)^2 + (5-3)^2)^{1/2} = 3.61 \end{aligned}$$

$$d_{1,8} = ((0-4)^2 + (5-1)^2)^{1/2} = 5.65$$

$$d_{1,9} = ((0-4)^2 + (5-3)^2)^{1/2} = 4.47$$

$$d_{1,10} = ((0-4)^2 + (5-4)^2)^{1/2} = 4.12$$

1NN i=1 is closest \Rightarrow i=1 (classified as) ⊖. True label ⊖, no misprediction

3NN: $\{i=3, i=2, i=4\}$

$i=3: \ominus$
 $i=4: \oplus$
 $i=2: \ominus$

we classify as \ominus the label \ominus , no misprediction

9NN: $\{ \text{all except one left out} \}$

$\begin{matrix} 5 \\ 5 \end{matrix} \begin{matrix} \ominus \\ \oplus \end{matrix}$ over all of set. True label is \ominus so $(4 \ominus, 5 \oplus)$

so we classify as \oplus
 but true label is \ominus
 so misprediction = 1

leave $i=2$ out ($i=2$ is far left) (0,3)

distances:

$$d_{2,1} = ((0-0)^2 + (3-5)^2)^{1/2} = 2$$

$$d_{2,3} = ((0-1)^2 + (3-4)^2)^{1/2} = 1.41$$

$$d_{2,4} = ((0-2)^2 + (3-4)^2)^{1/2} = 2.24$$

$$d_{2,5} = ((0-2)^2 + (3-1)^2)^{1/2} = 2.83$$

$$d_{2,6} = ((0-3)^2 + (3-2)^2)^{1/2} = 3.16$$

$$d_{2,7} = ((0-3)^2 + (3-3)^2)^{1/2} = 3$$

$$d_{2,8} = ((0-4)^2 + (3-1)^2)^{1/2} = 4.47$$

$$d_{2,9} = ((0-4)^2 + (3-3)^2)^{1/2} = 4$$

$$d_{2,10} = ((0-4)^2 + (3-4)^2)^{1/2} = 4.12$$

1NN: $\{i=3\}$

$i=3: \ominus$, $i=2$ true label is \oplus so \checkmark no misprediction

3NN: $\{i=3, i=1, i=4\}$

$i=3: \ominus$
 $i=1: \ominus$
 $i=4: \oplus$

majority vote: \ominus , classification \ominus no mispred

9NN:

same as before. even split on classification means
 where we have an tie or the opposite label is
 what we classify as.
 mispredictions = 2

here $i=3$ or ($i=3$ var test)

distances:

$$d_{3,1} = ((1-0)^2 + (4-5)^2)^{1/2} = 1.41$$

$$d_{3,2} = ((1-0)^2 + (4-3)^2)^{1/2} = 1.41$$

$$d_{3,4} = ((1-2)^2 + (4-4)^2)^{1/2} = 1$$

$$d_{3,5} = ((1-2)^2 + (4-1)^2)^{1/2} = 3.16$$

$$d_{3,6} = ((1-3)^2 + (4-2)^2)^{1/2} = 2.83$$

$$d_{3,7} = ((1-3)^2 + (4-3)^2)^{1/2} = 2.24$$

1NN: $i=4$

$i=4: \oplus$, $i=3$ true label \ominus

misprediction = 1

$$d_{3,8} = ((1-4)^2 + (4-1)^2)^{1/2} = 3$$

$$d_{3,9} = ((1-4)^2 + (4-3)^2)^{1/2} = 3.16$$

$$d_{3,10} = ((1-4)^2 + (4-4)^2)^{1/2} = 3$$

3NN: $i=1, i=2, i=4$

$i=4: \oplus$
 $i=1: \ominus$
 $i=2: \ominus$
 classification: \ominus , true label \ominus
 misprediction = 0

9NN: sawed picture.

misprediction = 3

here $i=4$ or ($i=4$ var test)

distances:

$$d_{4,1} = ((2-0)^2 + (4-5)^2)^{1/2} = 2.24$$

$$d_{4,2} = ((2-0)^2 + (4-3)^2)^{1/2} = 2.24$$

$$d_{4,3} = ((2-1)^2 + (4-4)^2)^{1/2} = 1$$

$$d_{4,5} = ((2-2)^2 + (4-1)^2)^{1/2} = 3$$

$$d_{4,6} = ((2-3)^2 + (4-2)^2)^{1/2} = 2.24$$

$$d_{4,7} = ((2-3)^2 + (4-3)^2)^{1/2} = 1.41$$

$$d_{4,8} = ((2-4)^2 + (4-1)^2)^{1/2} = 3.6$$

$$d_{4,9} = ((2-4)^2 + (4-3)^2)^{1/2} = 2.24$$

$$d_{4,10} = ((2-4)^2 + (4-4)^2)^{1/2} = 2$$

1NN: $\{i=3\}$

$i=3: \ominus, i=4$ true label \oplus

misprediction = 2

3NN: $\{i=3, i=7, i=10\}$

$i=3: \ominus$
 $i=7: \oplus$
 $i=10: \oplus$
 classification: \oplus , true label \oplus
 misprediction = 0

9NN: same as before.

misprediction = 4

leave $i=5$ out ($i=5$ is our test)

distances:

$$d_{g,1} = ((2-0)^2 + (1-5)^2)^{1/2} = 4.47$$

$$d_{g,2} = ((2-0)^2 + (1-3)^2)^{1/2} = 2.83$$

$$d_{g,3} = ((2-1)^2 + (1-4)^2)^{1/2} = 3.16$$

$$d_{g,4} = ((2-2)^2 + (1-4)^2)^{1/2} = 3$$

$$d_{g,6} = ((2-3)^2 + (1-2)^2)^{1/2} = 1.41$$

$$d_{g,7} = ((2-3)^2 + (1-3)^2)^{1/2} = 2.24$$

$$d_{r,8} = ((2-4)^2 + (1-1)^2)^{1/2} = 2$$

$$d_{r,9} = ((2-4)^2 + (1-3)^2)^{1/2} = 2.83$$

$$d_{r,10} = ((2-4)^2 + (1-4)^2)^{1/2} = 3.61$$

1NN: $\{i=6\}$

$i=6: \oplus$, predict \oplus

$i=5$ true label = \ominus
 miss!

misprediction = 3

3NN: $\{i=6, i=8, i=7\}$

$i=6: \oplus$
 $i=7: \oplus$
 $i=8: \ominus$
 classification: \oplus , true label \ominus
 misprediction = 1

9NN: same as before.

misprediction = 5

here $i=6$ at ($i=$ test)

distances:

$$d_{6,1} = ((3-0)^2 + (2-5)^2)^{1/2} = 4.24$$

$$d_{6,2} = ((3-0)^2 + (2-3)^2)^{1/2} = 3.14$$

$$d_{6,3} = ((3-1)^2 + (2-4)^2)^{1/2} = 2.83$$

$$d_{6,4} = ((3-2)^2 + (2-4)^2)^{1/2} = 2.24$$

$$d_{6,5} = ((3-2)^2 + (2-1)^2)^{1/2} = 1.41$$

$$d_{6,7} = ((3-3)^2 + (2-3)^2)^{1/2} = 1$$

$$d_{6,8} = ((3-4)^2 + (2-1)^2)^{1/2} = 1.41$$

$$d_{6,9} = ((3-4)^2 + (2-3)^2)^{1/2} = 1.41$$

$$d_{6,10} = ((3-1)^2 + (2-4)^2)^{1/2} = 2.24$$

can pick 2/3 for $i=9, 8, 9$ bc distances are the same

1NN: $i=7$

$i=7$: \oplus
 $i=6$ misclassified \ominus
 good classification!

misclassification = 3

3NN:

2/3 are \ominus and 1/3 is \oplus , here is already 1 \oplus guaranteed so depending on how you pick you can get a diff classification

we will pick the ones in the majority

9NN:
 same,

misclassification = 6

$i=7$: \oplus
 $i=5$: \ominus
 $i=8$: \ominus } \ominus miss!
 misclassification = 2

here $i=7$ at ($i=$ test)

distances:

$$d_{7,1} = ((3-0)^2 + (3-5)^2)^{1/2} = 3.61$$

$$d_{7,2} = ((3-0)^2 + (3-3)^2)^{1/2} = 3$$

$$d_{7,3} = ((3-1)^2 + (3-4)^2)^{1/2} = 2.24$$

$$d_{7,4} = ((3-2)^2 + (3-4)^2)^{1/2} = 1.41$$

$$d_{7,5} = ((3-2)^2 + (3-1)^2)^{1/2} = 2.24$$

$$d_{7,6} = ((3-3)^2 + (3-2)^2)^{1/2} = 1$$

$$d_{7,8} = ((3-4)^2 + (3-1)^2)^{1/2} = 2.24$$

$$d_{7,9} = ((3-4)^2 + (3-3)^2)^{1/2} = 1$$

$$d_{7,10} = ((3-4)^2 + (3-4)^2)^{1/2} = 1.41$$

1NN: can pick $i=6$ or $i=9$, both classify as \oplus

true label is \oplus so no misprediction

3NN: $\left. \begin{matrix} \text{mispredictions} = 3 \\ i=6, i=9, i=4 \end{matrix} \right\}$

$\left. \begin{matrix} i=6: \oplus \\ i=9: \oplus \\ i=4: \oplus \end{matrix} \right\}$ classify as \oplus , true label is \oplus

mispredictions = 2
9NN: mispredictions = 7

here $i=8$ at ($i=8$ now test)

distances:

$$d_{8,1} = ((4-0)^2 + (1-5)^2)^{1/2} = 5.66$$

$$d_{8,2} = ((4-0)^2 + (1-3)^2)^{1/2} = 4.47$$

$$d_{8,3} = ((4-1)^2 + (1-4)^2)^{1/2} = 4.24$$

$$d_{8,4} = ((4-2)^2 + (1-4)^2)^{1/2} = 3.61$$

$$d_{8,5} = ((4-2)^2 + (1-1)^2)^{1/2} = 2$$

$$d_{8,6} = ((4-3)^2 + (1-2)^2)^{1/2} = 1.41$$

$$d_{8,7} = ((4-3)^2 + (1-3)^2)^{1/2} = 2.23$$

$$d_{8,9} = ((4-4)^2 + (1-3)^2)^{1/2} = 2$$

$$d_{8,10} = ((4-4)^2 + (1-4)^2)^{1/2} = 3$$

1NN: $i=6: \oplus$

$i=8$ true label: \ominus

another mis

misprediction = 4

9NN: misprediction = 8

3NN: $\left. \begin{matrix} i=6: \oplus \\ i=5: \ominus \\ i=9: \oplus \end{matrix} \right\}$ guess \oplus

misprediction = 3

case $i=9$ at $(1=9)$ (not test)

distances:

$$d_{9,1} = ((4-0)^2 + (3-1)^2)^{1/2} = 4.47$$

$$d_{9,2} = ((4-0)^2 + (3-3)^2)^{1/2} = 4$$

$$d_{9,3} = ((4-1)^2 + (3-4)^2)^{1/2} = 3.16$$

$$d_{9,4} = ((4-2)^2 + (3-4)^2)^{1/2} = 2.24$$

$$d_{9,5} = ((4-2)^2 + (3-1)^2)^{1/2} = 2.83$$

$$d_{9,6} = ((4-3)^2 + (3-2)^2)^{1/2} = 1.41$$

1NN: $i=7$ or $i=10$

both are \oplus so classify as \oplus

$i=9$ has label \oplus
mis prediction = 4

4NN: mis prediction = 9

$$d_{9,7} = ((4-3)^2 + (3-3)^2)^{1/2} = 1$$

$$d_{9,8} = ((4-4)^2 + (3-1)^2)^{1/2} = 2$$

$$d_{9,10} = ((4-4)^2 + (3-4)^2)^{1/2} = 1$$

3NN: $i=7: \oplus$
 $i=10: \oplus$
 $i=8: \ominus$ } \oplus ✓

mis predictions = 3

case $i=10$ at $(1=10)$ (not test)

distances:

$$d_{10,1} = ((4-0)^2 + (4-1)^2)^{1/2} = 4.12$$

$$d_{10,2} = ((4-0)^2 + (4-3)^2)^{1/2} = 4.12$$

$$d_{10,3} = ((4-1)^2 + (4-4)^2)^{1/2} = 3$$

$$d_{10,4} = ((4-2)^2 + (4-4)^2)^{1/2} = 2$$

$$d_{10,5} = ((4-2)^2 + (4-1)^2)^{1/2} = 3.61$$

$$d_{10,6} = ((4-3)^2 + (4-2)^2)^{1/2} = 2.24$$

$$d_{10,7} = ((4-3)^2 + (4-3)^2)^{1/2} = 1.41$$

$$d_{10,8} = ((4-4)^2 + (4-1)^2)^{1/2} = 3$$

$$d_{10,9} = ((4-4)^2 + (4-3)^2)^{1/2} = 1$$

1NN:

$i=9$ is closest, is \oplus

$i=10$ true label is \oplus

so no miss

mis predictions = 4

9NN:

mis prediction = 10

3NN:

$i=9$: \oplus
 $i=7$: \oplus
 $i=4$: \oplus } \oplus ✓

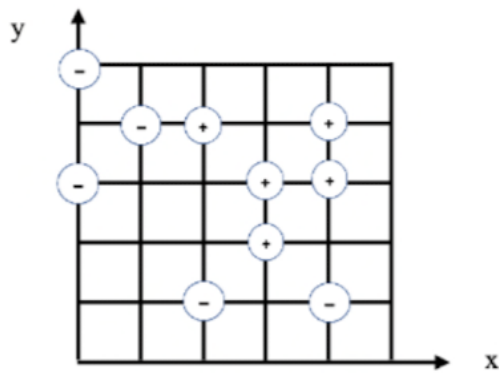
mis predictions = 3

$$1NN \text{ error rate} = \frac{4}{10} = 0.4$$

$$b.) 3NN \text{ error rate} = \frac{3}{10} = 0.3$$

$$c.) 9NN \text{ error rate} = \frac{10}{10}$$

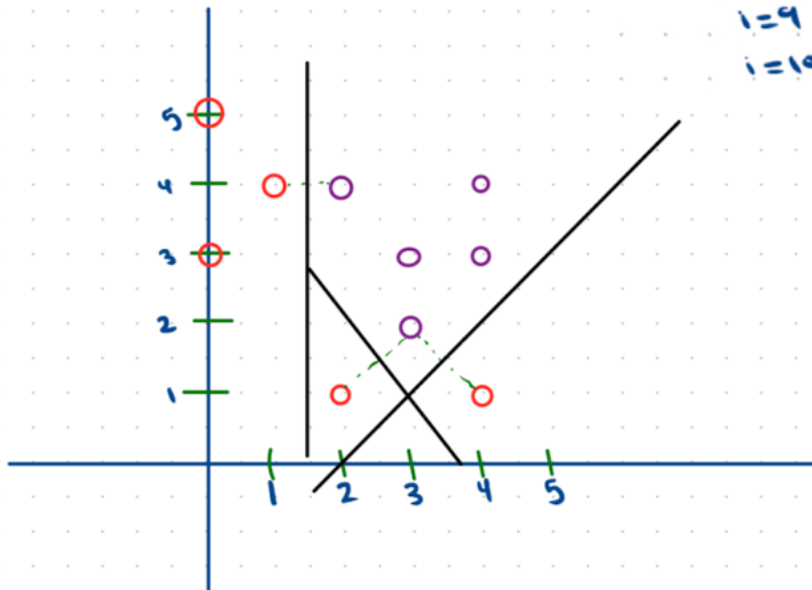
Part d:



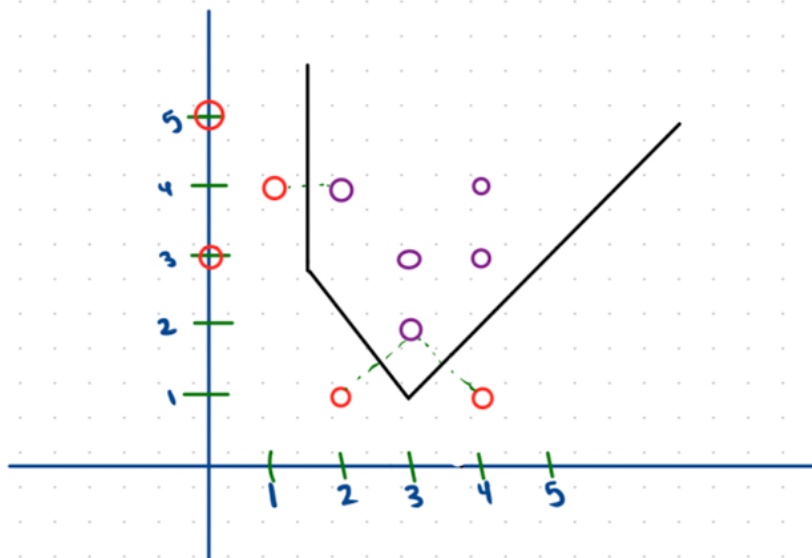
points:

i=1
i=2
i=3
i=4
i=5
i=6
i=7
i=8
i=9
i=10

0,5 : 0
0,3 : 0
5,4 : 0
2,4 : 0
2,1 : 0
3,2 : 0
3,3 : 0
4,1 : 0
4,3 : 0
4,4 : 0



0 = +
0 = -



4:

4. [12 points] Find the class of instance #10 below following the 3NN strategy. Use Euclidean distance as your distance measure. You must **show all your calculations** for full credit.

ID	Red	Green	Blue	Class
#1	220	20	60	1
#2	255	99	21	1
#3	250	128	14	1
#4	144	238	144	2
#5	107	142	35	2
#6	46	139	87	2
#7	64	224	208	3
#8	176	224	23	3
#9	100	149	237	3
#10	154	205	50	?

$$3NN, \text{ Euclidean distance} = \left(\sum_{i=1}^n (q_i - p_i)^2 \right)^{1/2}$$

$$\#1 \quad d_1 = ((154-220)^2 + (205-20)^2 + (50-60)^2)^{1/2} = 196.67$$

$$\#2 \quad d_2 = ((154-255)^2 + (205-99)^2 + (50-21)^2)^{1/2} = 149.26$$

$$\#3 \quad d_3 = ((154-250)^2 + (205-128)^2 + (50-14)^2)^{1/2} = 128.22$$

$$\#4 \quad d_4 = ((154-144)^2 + (205-238)^2 + (50-144)^2)^{1/2} = 100.12 \quad *$$

$$\#5 \quad d_5 = ((154-107)^2 + (205-142)^2 + (50-35)^2)^{1/2} = 80.01 \quad *$$

$$\#6 \quad d_6 = ((154-46)^2 + (205-139)^2 + (50-87)^2)^{1/2} = 131.87$$

$$\#7 \quad d_7 = ((154-64)^2 + (205-224)^2 + (50-208)^2)^{1/2} = 187.83$$

$$\#8 \quad d_8 = ((154-176)^2 + (205-224)^2 + (50-23)^2)^{1/2} = 39.67 \quad *$$

$$\#9 \quad d_9 = ((154-100)^2 + (205-149)^2 + (50-237)^2)^{1/2} = 202.536$$

#4, 5, 8 are selected

#4 → class 2
 #5 → class 2
 #8 → class 3

⇒ #10 is classified as class 2 by majority vote

5:Part a:

5. [25 points] Use the dataset below to answer the next questions:

Day	Outlook	Temperature	Humidity	Wind	PlayTennis
D1	Sunny	Hot	High	Weak	No
D2	Sunny	Hot	High	Strong	No
D3	Overcast	Hot	High	Weak	Yes
D4	Rain	Mild	High	Weak	Yes
D5	Rain	Cool	Normal	Weak	Yes
D6	Rain	Cool	Normal	Strong	No
D7	Overcast	Cool	Normal	Strong	Yes
D8	Sunny	Mild	High	Weak	No
D9	Sunny	Cool	Normal	Weak	Yes
D10	Rain	Mild	Normal	Weak	Yes
D11	Sunny	Mild	Normal	Strong	Yes
D12	Overcast	Mild	High	Strong	Yes
D13	Overcast	Hot	Normal	Weak	Yes
D14	Rain	Mild	High	Strong	No

- a) [10 points] Classify the instance <D15, Sunny, Mild, Normal, Weak> following the Naïve Bayes strategy. **Show all your calculations** until the final normalized probability values.
- b) [15 points] Complete the Python program (naive_bayes.py) that will read the file weather_training.csv (training set) and output the classification of each test instance from the file weather_test (test set) **if the classification confidence is ≥ 0.75** . Sample of output:

Day	Outlook	Temperature	Humidity	Wind	PlayTennis	Confidence
D15	Sunny	Hot	High	Weak	No	0.86
D16	Sunny	Mild	High	Weak	Yes	0.78

$$\begin{aligned}
 a.) \quad P(\text{class} = \text{yes} \mid \text{outlook} = \text{sunny}, \text{temp} = \text{mild}, \text{humidity} = \text{normal}, \text{wind} = \text{Weak}) &= \\
 &= P(\text{class} = \text{yes}) * \prod_i P(A_i = x_i \mid \text{class} = \text{yes}) \\
 &= \frac{9}{14} \left(P(\text{outlook} = \text{sunny} \mid \text{class} = \text{yes}) * (P(\text{temp} = \text{mild} \mid \text{class} = \text{yes})) * \right. \\
 &\quad \left. (P(\text{humidity} = \text{normal} \mid \text{class} = \text{yes})) * P(\text{wind} = \text{Weak} \mid \text{class} = \text{yes})) \right) \\
 &= \frac{9}{14} \left(\frac{2}{9} * \frac{4}{9} * \frac{6}{9} * \frac{6}{9} \right) \\
 &= \frac{2592}{91854} = 0.0282
 \end{aligned}$$

$$\begin{aligned}
 P(\text{class} = \text{No} \mid \text{outlook} = \text{sunny}, \text{temp} = \text{mild}, \text{humidity} = \text{normal}, \text{wind} = \text{Weak}) &= \\
 &= P(\text{class} = \text{no}) * \prod_i P(A_i = x_i \mid \text{class} = \text{no}) \\
 &= \frac{5}{14} \left(P(\text{outlook} = \text{sunny} \mid \text{class} = \text{no}) * (P(\text{temp} = \text{mild} \mid \text{class} = \text{no})) * \right. \\
 &\quad \left. (P(\text{humidity} = \text{normal} \mid \text{class} = \text{no})) * P(\text{wind} = \text{Weak} \mid \text{class} = \text{no})) \right) \\
 &= \frac{5}{14} \left(\frac{3}{9} \right) \left(\frac{2}{9} \right) \left(\frac{1}{9} \right) \left(\frac{2}{9} \right) = \frac{3 \cdot 2 \cdot 2}{14 \cdot 25 \cdot 5} = \frac{12}{1750} = 0.00685714
 \end{aligned}$$

now normalize the values:

$P(\text{class} = \text{yes} \mid \text{outlook} = \text{sunny}, \text{Temp} = \text{mild}, \text{Humidity} = \text{normal}, \text{wind} = \text{Weak})$

$$= \frac{0.0282}{0.0282 + 0.00685714}$$

$$= 0.804$$

$P(\text{class} = \text{no} \mid \text{outlook} = \text{sunny}, \text{Temp} = \text{mild}, \text{Humidity} = \text{normal}, \text{wind} = \text{Weak}) =$

$$\frac{0.00685714}{0.0282 + 0.00685714} = 0.1955$$

$$= 0.196$$

\therefore instance #15 classified as yes.

Part b:

https://github.com/franserr99/cs4210/blob/main/a2/naive_bayes.py