

## MLT

Section Id :	64065386938
Section Number :	4
Section type :	Online
Mandatory or Optional :	Mandatory
Number of Questions :	16
Number of Questions to be attempted :	16
Section Marks :	50
Display Number Panel :	Yes
Section Negative Marks :	0
Group All Questions :	No
Enable Mark as Answered Mark for Review and Clear Response :	No
Section Maximum Duration :	0
Section Minimum Duration :	0
Section Time In :	Minutes
Maximum Instruction Time :	0
Sub-Section Number :	1
Sub-Section Id :	640653189869
Question Shuffling Allowed :	No

Question Number : 73 Question Id : 6406531231404 Question Type : MCQ

Correct Marks : 0

Question Label : Multiple Choice Question

**THIS IS QUESTION PAPER FOR THE SUBJECT "DIPLOMA LEVEL : MACHINE LEARNING TECHNIQUES (COMPUTER BASED EXAM)"**

**ARE YOU SURE YOU HAVE TO WRITE EXAM FOR THIS SUBJECT?**

**CROSS CHECK YOUR HALL TICKET TO CONFIRM THE SUBJECTS TO BE WRITTEN.**

**(IF IT IS NOT THE CORRECT SUBJECT, PLS CHECK THE SECTION AT THE TOP FOR THE SUBJECTS REGISTERED BY YOU)**

Options :

6406534160347.  YES

6406534160348.  NO

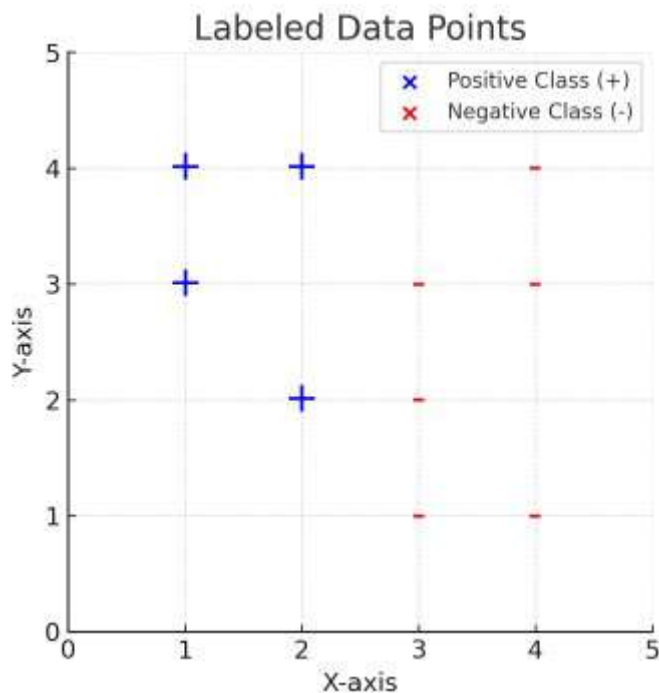
Sub-Section Number :	2
Sub-Section Id :	640653189870
Question Shuffling Allowed :	Yes

Question Number : 74 Question Id : 6406531231405 Question Type : MCQ

Correct Marks : 3

Question Label : Multiple Choice Question

For a given dataset, a 1-Nearest Neighbor (1-NN) and a 3-Nearest Neighbor (3-NN) classifier are applied. Which classifier is likely to exhibit a higher Leave-One-Out Cross Validation (LOOCV) error? In case of tie-breaker, assign a positive class (+) to the data point.



Options :

6406534160349. ✓ 1-NN

6406534160350. ✗ 3-NN

6406534160351. ✗ Both have the same error.

Question Number : 75 Question Id : 6406531231406 Question Type : MCQ

Correct Marks : 3

Question Label : Multiple Choice Question

Consider the following three weight vectors obtained by minimizing the ridge regression objective with penalty parameters  $\lambda = 0.1, 10, 50$ .

$$\theta_1 = [0.5 \quad 0.56 \quad 2.5]^T$$

$$\theta_2 = [0.05 \quad 0.1 \quad 1.23]^T$$

$$\theta_3 = [1.2 \quad 0.84 \quad 3.15]^T$$

Select the most appropriate match for each weight vector corresponding to penalty parameter  $\lambda$  from the following options:

Options :

6406534160352. ✗  $\theta_1$  corresponds to  $\lambda = 50$ ,  $\theta_2$  to  $\lambda = 0.1$ , and  $\theta_3$  to  $\lambda = 10$ .

6406534160353. ✗  $\theta_1$  corresponds to  $\lambda = 0.1$ ,  $\theta_2$  to  $\lambda = 10$ , and  $\theta_3$  to  $\lambda = 50$ .

6406534160354. ✓  $\theta_1$  corresponds to  $\lambda = 10$ ,  $\theta_2$  to  $\lambda = 50$ , and  $\theta_3$  to  $\lambda = 0.1$ .

Sub-Section Number :

Sub-Section Id :

Question Shuffling Allowed :

3

640653189871

Yes

Question Number : 76

Question Id : 6406531231407

Question Type : MCQ

Correct Marks : 4

Question Label : Multiple Choice Question

Consider the following dataset with two features and the corresponding labels:

$x_1$	$x_2$	$y$
1	0	1.5
2	2	4
3	0	4.5
4	2	7

Fit the linear regression model  $y = w_1x_1 + w_2x_2$  using the normal equation obtained from the squared error loss.

Hint: The normal equation for linear regression is:

$$w = (XX^T)^{-1}Xy$$

Options :

6406534160355. ✖  $y = 2x_1 + x_2$
6406534160356. ✔  $y = 1.5x_1 + 0.5x_2$
6406534160357. ✖  $y = 1.8x_1 + 1.45x_2$
6406534160358. ✖  $y = x_1 + 2x_2$

Question Number : 77

Question Id : 6406531231408

Question Type : MCQ

Correct Marks : 4

Question Label : Multiple Choice Question

Consider a dataset with 4 datapoints:

$$\{(x_1, y_1), (x_2, y_2), (x_3, y_3), (x_4, y_4)\},$$

where  $y_i \in \{+1, -1\}$  and  $x_i \in \mathbb{R}^2$ . In the first iteration of the AdaBoost algorithm, suppose a decision stump  $f_1$  is chosen, which correctly classifies the first three data points and incorrectly classifies the last data point. Assume the initial distribution of the dataset assigns equal weights to all data points, i.e.,  $D_0(i) = \frac{1}{4}$ , for  $i = 1, 2, 3, 4$ . What will be the updated distribution of the weights of the data points after the first iteration?

**Options :**

$$D_1(i) = \begin{cases} \frac{1}{3}, & i = 1, 2, 3 \\ 0, & i = 4 \end{cases}.$$

6406534160359. ✖

$$D_1(i) = \begin{cases} \frac{1}{6}, & i = 1, 2, 3 \\ \frac{1}{2}, & i = 4 \end{cases}.$$

6406534160360. ✔

$$D_1(i) = \begin{cases} \frac{1}{4}, & i = 1, 2, 3, 4 \\ 0, & \text{otherwise} \end{cases}.$$

6406534160361. ✖

$$D_1(i) = \begin{cases} \frac{1}{5}, & i = 1, 2, 3 \\ \frac{2}{5}, & i = 4 \end{cases}.$$

6406534160362. ✖

**Sub-Section Number :**

4

**Sub-Section Id :**

640653189872

**Question Shuffling Allowed :**

Yes

**Question Number : 78 Question Id : 6406531231409 Question Type : MSQ**

**Correct Marks : 3 Max. Selectable Options : 0**

Question Label : Multiple Select Question

Which of the following statements correctly differentiates PCA and Kernel PCA?

**Options :**

6406534160363. ✔ PCA maximizes variance in the original feature space, while Kernel PCA maximizes variance in a higher-dimensional transformed space.

6406534160364. ✔ PCA finds principal components using linear transformations in the original space, while Kernel PCA uses non-linear transformations to find principal components in a higher-dimensional space.

6406534160365. ✔ Kernel PCA can capture non-linear patterns in data, making it useful when PCA fails to represent complex structures in a linear space.

6406534160366. ✖ PCA and Kernel PCA always yield identical results regardless of the dataset structure.

**Question Number : 79 Question Id : 6406531231410 Question Type : MSQ**

**Correct Marks : 3 Max. Selectable Options : 0**

Question Label : Multiple Select Question

Consider the following formulation of the soft margin SVM:

$$\min_{w, \epsilon} \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \epsilon_i, \quad C \geq 0$$

$$\text{subject to } (w^T x_i) y_i + \epsilon_i \geq 1, \quad \forall i \\ \epsilon_i \geq 0, \quad \forall i.$$

Which of the following statements is/are correct?

**Options :**

6406534160367. ✖ When  $C = 0$ , the optimal value of the objective function is  $\infty$ .

6406534160368. ✔ When  $C = 0$ , the optimal value of the objective function is 0.

6406534160369. ✖ As  $C$  approaches 0, the soft-margin SVM is equivalent to the hard-margin SVM.

6406534160370. ✔ As  $C$  approaches  $\infty$ , the soft-margin SVM is equivalent to the hard-margin SVM.

6406534160371. ✔ A smaller value of  $C$  will create larger margin.

**Sub-Section Number :**

5

**Sub-Section Id :**

640653189873

**Question Shuffling Allowed :**

Yes

**Question Number : 80 Question Id : 6406531231411 Question Type : MSQ**

**Correct Marks : 2 Max. Selectable Options : 0**

Question Label : Multiple Select Question

Which of the following statements are true for bagging?

**Options :**

6406534160372. ✔ The final model has lesser variance than the individual learners.

6406534160373. ✖ The final model has a higher variance than the individual learners.

6406534160374. ✔ Estimators in bagging can be trained parallelly.

6406534160375. ✖ If the number of data points is large, typically two-third of the data points remain unselected in bags.

**Question Number : 81 Question Id : 6406531231412 Question Type : MSQ**

**Correct Marks : 2 Max. Selectable Options : 0**

Question Label : Multiple Select Question

What is the effect of increasing the regularization parameter  $\lambda$  in Lasso regression?

**Options :**

6406534160376. ✔ It penalizes large coefficients to reduce overfitting.

6406534160377. ✖ It shrinks the coefficients but does not set them to zero.

6406534160378. ✔ It forces more coefficients to be exactly zero, performing feature selection.

6406534160379. ✖ It has no effect on the regression model.

**Sub-Section Number :** 6

**Sub-Section Id :** 640653189874

**Question Shuffling Allowed :** Yes

**Question Number : 82 Question Id : 6406531231413 Question Type : SA**

**Correct Marks : 2**

Question Label : Short Answer Question

For a decision tree, each node has exactly two child nodes (balanced tree). If the tree has a depth of 3, how many leaf nodes are there?

**Response Type :** Numeric

**Evaluation Required For SA :** Yes

**Show Word Count :** Yes

**Answers Type :** Equal

**Text Areas :** PlainText

**Possible Answers :**

8

**Sub-Section Number :** 7

**Sub-Section Id :** 640653189875

**Question Shuffling Allowed :** Yes

**Question Number : 83 Question Id : 6406531231414 Question Type : SA**

**Correct Marks : 3**

Question Label : Short Answer Question



Consider the following dataset in  $\mathbb{R}^2$ :

$$\{(1, 2), (3, 4), (5, 4), (7, 8), (9, 10)\}$$

Lloyd's algorithm ( $K$ -means) is run on this data set with points  $(1, 2)$  and  $(9, 10)$  as the initial cluster centers. Let  $(c_1, d_1)$  and  $(c_2, d_2)$  be the final cluster centers upon convergence. What is the value of the product  $c_1 \times c_2$ ?

**Response Type :** Numeric

**Evaluation Required For SA :** Yes

**Show Word Count :** Yes

**Answers Type :** Equal

**Text Areas :** PlainText

**Possible Answers :**

24

**Question Number : 84 Question Id : 6406531231415 Question Type : SA**

**Correct Marks : 3**

Question Label : Short Answer Question

Let  $X_1, X_2, \dots, X_n$  be i.i.d. samples from a Uniform distribution on the interval  $[0, \theta]$ , where  $\theta$  is an unknown parameter. The probability density function (PDF) of continuous uniform distribution is given by:

$$f(x; \theta) = \begin{cases} \frac{1}{\theta}, & 0 \leq x \leq \theta \\ 0, & \text{otherwise} \end{cases}.$$

Find the Maximum Likelihood Estimate (MLE) of  $\theta$  based on a given sample 10, 15, 12, 20, 17.

**Response Type :** Numeric

**Evaluation Required For SA :** Yes

**Show Word Count :** Yes

**Answers Type :** Equal

**Text Areas :** PlainText

**Possible Answers :**

20

**Question Number : 85 Question Id : 6406531231416 Question Type : SA**

**Correct Marks : 3**

Question Label : Short Answer Question

You are building a Naïve Bayes classifier to determine whether a person has a certain medical condition (Yes or No) based on three features:  $f_1$  (Age),  $f_2$  (Blood pressure level), and  $f_3$  (Smoking status, Yes/No). Given that the features are conditionally independent for a given class, the continuous features  $f_1$  and  $f_2$  are modeled using a Gaussian distribution, while the binary feature  $f_3$  follows a Bernoulli distribution. Determine the total number of parameters that need to be estimated for classification using this Naïve Bayes model.

**Response Type :** Numeric

**Evaluation Required For SA :** Yes

**Show Word Count :** Yes

**Answers Type :** Equal

**Text Areas :** PlainText

**Possible Answers :**

11

**Sub-Section Number :**

8

**Sub-Section Id :**

640653189876

**Question Shuffling Allowed :**

No

**Question Id : 6406531231417 Question Type : COMPREHENSION Sub Question Shuffling Allowed : No Group Comprehension Questions : No Question Pattern Type : NonMatrix**

**Question Numbers : (86 to 87)**

**Question Label :** Comprehension

Consider the following data set with three data points:

$$D = \left\{ \left( \begin{bmatrix} 1 \\ 1 \end{bmatrix}, -1 \right), \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix}, +1 \right), \left( \begin{bmatrix} -1 \\ -1 \end{bmatrix}, +1 \right) \right\}.$$

Based on the above data, answer the given subquestions.

**Sub questions**

**Question Number : 86 Question Id : 6406531231418 Question Type : SA**

**Correct Marks : 3**

**Question Label :** Short Answer Question

Find the squared length of the updated weight vector after one iteration (one pass through all the data points) of the perceptron algorithm, assuming  $w_0 = [0 \ 0]^T$ . While looking for mistakes, cycle through the data points from left to right.



**Response Type :** Numeric

**Evaluation Required For SA :** Yes

**Show Word Count :** Yes

**Answers Type :** Equal

**Text Areas :** PlainText

**Possible Answers :**

1

**Question Number :** 87 **Question Id :** 6406531231419 **Question Type :** MCQ

**Correct Marks :** 2

Question Label : Multiple Choice Question

Will the algorithm converge after this update?

**Options :**

6406534160385. ✓ YES

6406534160386. ✗ NO

**Question Id :** 6406531231420 **Question Type :** COMPREHENSION **Sub Question Shuffling**

**Allowed :** No **Group Comprehension Questions :** No **Question Pattern Type :** NonMatrix

**Question Numbers :** (88 to 89)

Question Label : Comprehension

Consider the transformation  $\phi : \mathbb{R}^2 \rightarrow \mathbb{R}^6$  associated with a polynomial kernel of degree 2:

$$\phi(x) = [1 \quad x_1^2 \quad x_2^2 \quad \sqrt{2}x_1x_2 \quad \sqrt{2}x_1 \quad \sqrt{2}x_2]^T$$

A kernel-SVM is trained on a dataset with the above kernel. The optimal weight vector is  $w = [-25 \quad 1 \quad 2 \quad 0 \quad 0 \quad 0]^T$ . Assume that the dataset is linearly separable in the transformed space.

Based on the above data, answer the given subquestions.

**Sub questions**

**Question Number :** 88 **Question Id :** 6406531231421 **Question Type :** MCQ

**Correct Marks :** 2

Question Label : Multiple Choice Question

What is the shape of the decision boundary in  $\mathbb{R}^2$ ?

**Options :**

6406534160387. ✗ It is a circle.

6406534160388. ✗ It is a straight line.

6406534160389. ✓ It is an ellipse.

6406534160390. ✗ It is a parabola.

**Question Number : 89 Question Id : 6406531231422 Question Type : MSQ**

**Correct Marks : 3 Max. Selectable Options : 0**

Question Label : Multiple Select Question

Which of the following training data points are certainly **not** support vectors?

**Options :**

6406534160391. ✓  $\begin{bmatrix} 1 & 2 \end{bmatrix}^T$

6406534160392. ✗  $\begin{bmatrix} 2\sqrt{2} & 3 \end{bmatrix}^T$

6406534160393. ✓  $\begin{bmatrix} 4 & 2 \end{bmatrix}^T$

6406534160394. ✗  $\begin{bmatrix} 3 & 2\sqrt{2} \end{bmatrix}^T$

**Question Id : 6406531231423 Question Type : COMPREHENSION Sub Question Shuffling**

**Allowed : No Group Comprehension Questions : No Question Pattern Type : NonMatrix**

**Question Numbers : (90 to 91)**

Question Label : Comprehension

Consider a single-layer neural network with two neurons in the hidden layer.

The weight parameters of the network are given as follows:

$$w_1^{(1)} = \frac{1}{2}, \quad w_2^{(1)} = \frac{1}{2}$$

$$w_1^{(2)} = \frac{1}{2}, \quad w_2^{(2)} = -\frac{1}{2}$$

$$w^{out} = \begin{bmatrix} 1 & -1 \end{bmatrix}^T,$$

where  $w_i^{(j)}$  represents the weight associated with the  $j$ -th neuron for the  $i$ -th input feature. Assume that we are solving a binary classification problem.

Based on the above data, answer the given subquestions.

**Sub questions**

**Question Number : 90 Question Id : 6406531231424 Question Type : SA**

**Correct Marks : 3**

Question Label : Short Answer Question

The output layer of the neural network will return a probability  $p$  for the input  $x_{test} = [2 \ 4]^T$ . The sigmoid function is used as the activation function in both the hidden and the output layer of the neural network. Find the value of  $p$ . Enter the answer correct to two decimal places.

**Response Type :** Numeric  
**Evaluation Required For SA :** Yes  
**Show Word Count :** Yes  
**Answers Type :** Range  
**Text Areas :** PlainText  
**Possible Answers :**  
0.64 to 0.68

**Question Number :** 91 **Question Id :** 6406531231425 **Question Type :** SA  
**Correct Marks :** 2  
**Question Label :** Short Answer Question  
If the model predicts the label as 1 for  $p$  greater than 0.5 and 0 otherwise, find the predicted label for  $x_{test}$ .

**Response Type :** Numeric  
**Evaluation Required For SA :** Yes  
**Show Word Count :** Yes  
**Answers Type :** Equal  
**Text Areas :** PlainText  
**Possible Answers :**  
1

## Appdev2

Section Id :	64065386939
Section Number :	5
Section type :	Online
Mandatory or Optional :	Mandatory
Number of Questions :	31
Number of Questions to be attempted :	31
Section Marks :	100
Display Number Panel :	Yes