## Time : 0

# Correct Marks : 2 Max. Selectable Options : 0

**Question Label : Multiple Select Question** 

Which of the following JavaScript method(s) is/are used to remove data from the session storage?

# **Options :**

6406532734047. 🗸 clear()

6406532734048. < removeItem()

6406532734049. \*\* deleteItem()

6406532734050. \*\* unSet()

# MLT

Section Id :	64065356696
Section Number :	11
Section type :	Online
Mandatory or Optional :	Mandatory
Number of Questions :	17
Number of Questions to be attempted :	17
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 :	Yes
Maximum Instruction Time :	0
Sub-Section Number :	1
Sub-Section Id :	640653118943
Question Shuffling Allowed :	No

Question Number : 318 Question Id : 640653816191 Question Type : MCQ Is Question

Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

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 <u>TOP</u> FOR THE SUBJECTS REGISTERED BY YOU)

**Options :** 

6406532734051. 🗸 YES

6406532734052. **\*** NO

Sub-Section Number :	2
Sub-Section Id :	640653118944
Question Shuffling Allowed :	Yes
Is Section Default? :	null

Question Number : 319 Question Id : 640653816194 Question Type : MCQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

## **Correct Marks : 3**

**Question Label : Multiple Choice Question** 

Imagine a dataset characterized by two features, Feature 1 and Feature 2, demonstrating a perfect negative correlation of -1. When applying k-means clustering with k = 3 to this dataset, what is the most likely arrangement of cluster centers that minimizes the within-cluster sum of squares

(WCSS)?

# **Options :**

6406532734058. \* An equilateral triangle centered around the mean of the data.

6406532734059. V Cluster centers positioned along a straight line.

6406532734060. \* A triangle with two acute angles, positioned strategically within the data distribution.

6406532734061. \* A right-angled triangle with one center at the origin.

# Question Number : 320 Question Id : 640653816197 Question Type : MCQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

## **Correct Marks : 3**

Question Label : Multiple Choice Question

Consider the following two models fitted on a one-dimensional dataset:

Model 1:  $\hat{y} = w_0 + w_1 x$ Model 2:  $\hat{y} = w_1 x^2 + w_2 x + w_3$ 

If both models are trained on the same one-dimensional dataset and evaluated on the same test dataset, which model is more likely to have higher bias and lower variance?

# **Options :**

6406532734067. ✔ Model 1

6406532734068. \* Model 2

6406532734069. \* Both models are equally sensitive to outliers

## 6406532734070. \* Insufficient data

Sub-Section Number :	3
Sub-Section Id :	640653118945
Question Shuffling Allowed :	Yes
Is Section Default? :	null

### Question Number : 321 Question Id : 640653816201 Question Type : MCQ Is Question

# Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

## **Correct Marks : 4**

## **Question Label : Multiple Choice Question**

Consider a logistic regression model for a binary classification problem with two features  $x_1$  and  $x_2$ . The feature vector is  $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  and labels lie in  $\{0, 1\}$ . The threshold for inference is 0.5. The dummy feature and the weight corresponding to it can be ignored for this problem. Let  $x_1$  be the horizontal axis and  $x_2$  be the vertical axis. You are given two feature vectors:

$$\mathbf{x_1} = \begin{bmatrix} 1\\\\\sqrt{3} \end{bmatrix}, \mathbf{x_2} = \begin{bmatrix} -1\\\\\sqrt{3} \end{bmatrix}$$

The weight vector makes an angle of  $\theta$  with the positive  $x_1$  axis (horizontal). Each  $\theta$  corresponds to a different classifier. For what range of values of  $\theta$  are both  $\mathbf{x_1}$  and  $\mathbf{x_2}$  predicted to belong to class-1?

## Hints:

- To draw the weight vector  $\mathbf{w} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$ , plot the point  $(w_1, w_2)$  and draw an arrow starting at the origin to this point.
- $\tan(60^\circ) = \sqrt{3}$

## **Options :**

6406532734074. <br/>  $30^{\circ} < \theta < 150^{\circ}$ 

6406532734075. **\*** 
$$0^{\circ} < \theta < 60^{\circ}$$

6406532734076. **≈** 60° < θ < 180°

Sub-Section Number :	4
Sub-Section Id :	640653118946
Question Shuffling Allowed :	Yes
Is Section Default? :	null

Question Number : 322 Question Id : 640653816193 Question Type : MSQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

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

**Question Label : Multiple Select Question** 

Which of the following statements accurately describes the characteristics of kernel functions? Assume the dataset to be mean-centered.

## **Options :**

6406532734054. Kernel PCA can reconstruct original PCA if the kernel function is  $k(x_i, x_j) = (x_i^T x_j + 1)^2$ .

The dimensionality of the transformed dataset  $\phi(X)$ , computed using the kernel function, is always smaller than the original feature space.

The dimensionality of the transformed dataset  $\phi(X)$ , computed using the kernel 6406532734056.  $\checkmark$  function, can exceed the original feature space.

The dimension of the transformed dataset  $\phi(X)$ , whose inner products the 6406532734057.  $\checkmark$  kernel function computes, can be infinite.

#### Question Number : 323 Question Id : 640653816196 Question Type : MSQ Is Question

Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

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

## **Question Label : Multiple Select Question**

Let X be a data matrix of the shape (d, n) and y be the associated label vector of shape (n, 1). Assume that a linear regression model with loss as the sum of squared error is trained on the data  $\{X, y\}$ . In which of the following cases, the loss on the training data will necessarily be zero? Assume that the solution of the model is obtained by the normal equation that is  $w^* = (XX^T)^{-1}Xy$ .

# **Options :**

6406532734063.  $\checkmark$  If y lies in the space spanned of row vectors of X.

6406532734064. \* If y lies in the space spanned of row vectors of  $X^T$ .

If all the data points satisfy the equality  $x_1 + x_2 + \ldots + x_d = 0$ , where  $x_i$  is 6406532734065.  $\checkmark$  the *i*th feature and y = 0 for all the data points.

If all the data points satisfy the equality  $x_1^3 + x_2^3 + \ldots + x_d^3 = 0$ , where  $x_i$  is 6406532734066. **\*** the *i*th feature and y = 0 for all the data points.

# Question Number : 324 Question Id : 640653816203 Question Type : MSQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

# Correct Marks : 3 Max. Selectable Options : 0

**Question Label : Multiple Select Question** 

Consider the following dataset on which the soft margin SVM is applied.



Which of the following statements is/are true about this dataset? **Options :** 

6406532734085. \* Points {F, A, C, I} are the only support vectors.

6406532734086. ✓ Points {A, N} are a subset of support vectors.

6406532734087. ✓ Points {F, A, N} are a subset of support vectors.

6406532734088. ✓ Points except {F, A, C, I, N} do not play any role in determining optimal weight vector.

6406532734089. \* Points except {F, A, C, I} do not play any role in determining optimal weight vector.

Question Number : 325 Question Id : 640653816205 Question Type : MSQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

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

**Question Label : Multiple Select Question** 

Given a two-dimensional data set where points from class 1 are:

{(-2, 3), (-1, 1), (-1, 2), (-1, 4)}

And points from class 0 are:

## Which of the following statements are true?

# **Options :**

The given data points from classes 1 and 0 can be linearly separated using a 6406532734091.  $\checkmark$  Hard-margin SVM.

A perceptron model and a hard margin SVM can give different decision boundary 6406532734092.  $\checkmark$  for this dataset.

A Soft-margin SVM would be a more robust choice than a Hard-margin SVM for this dataset as the dataset is not linearly separable.

 $\begin{array}{l} \text{The width of the separation between the two supporting hyperplanes is 4.} \\ \text{(Hint:Calculate width using formulae } \frac{2}{||\mathbf{w}||}) \end{array}$ 

Question Number : 326 Question Id : 640653816206 Question Type : MSQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

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

Question Label : Multiple Select Question

Consider the following two-dimensional dataset with two classes: +1 for blue points and -1 for red points. An AdaBoost algorithm was run on this dataset using decision stumps as weak learners.



When training the new weak learner  $h_t(x)$  (decision stump at  $t^{th}$  iteration), we choose the split that minimizes the weighted miss-classification error with respect to current weights  $D_t$  i.e. choose  $h_t$  that minimizes  $\sum_{i=1}^n D_t(i)\mathbb{1}(h_t(x_i) \neq y_i)$ . Based on the above data, answer the below given question.

To train the second decision stump, which pair of points will be assigned equal weights to create the training data-set?

## **Options :**

```
6406532734095. * [2, 2]^T, [1, 2]^T
6406532734096. * [2, 2]^T, [1, 4]^T
6406532734097.  \checkmark [1, 1]^T, [1, 4]^T
6406532734098.  \checkmark [3, 1]^T, [3, 4]^T
Sub-Section Number :
```

Sub-Section Id :

**Question Shuffling Allowed :** 

5 640653118947 Yes

# Question Number : 327 Question Id : 640653816202 Question Type : MSQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

#### Correct Marks : 4 Max. Selectable Options : 0

#### **Question Label : Multiple Select Question**

Consider a soft-margin Support Vector Machine (SVM) for a binary classification problem with a dataset in a two-dimensional space. The optimization problem for the soft-margin SVM is formulated as:

Minimize 
$$\frac{1}{2} \|\mathbf{w}\|^2 + C \sum_{i=1}^N \xi_i$$

subject to the constraints:

$$y_i(\mathbf{w} \cdot \mathbf{x}_i + b) \ge 1 - \xi_i$$
 and  $\xi_i \ge 0$  for all  $i$ 

Where C is a positive constant.

Let  $w^*$ ,  $\xi^*$  be the optimal solutions, and  $\alpha^*$ ,  $\beta^*$  be the optimal dual solutions of the soft margin SVM problem.

Which of the following statements about the soft-margin SVM is correct?

#### **Options :**

6406532734079. \* If  $i^{th}$  data point lies on one of the supporting hyperplanes, then  $\alpha_i^* = 0$ .

If  $i^{th}$  data point lies on the correct supporting hyperplane, it does not pay any 6406532734080.  $\checkmark$  bribes.

A smaller value of C allows for a larger margin, potentially leading to less misclas-6406532734081. **\*** sifications on the training data.

6406532734082.  $\checkmark$  For a dataset with n data-points, there are 2n constraints for soft-margin SVM.

6406532734083.  $\checkmark$  As C approaches  $\infty$  the soft margin SVM is equal to the hard margin SVM.

null

6406532734084.  $\approx$  C can be negative, as long as the bribe( $\xi$ ) each data point pays is non-negative.

Sub-Section Number :	6
Sub-Section Id :	640653118948
Question Shuffling Allowed :	Yes
Is Section Default? :	null

# Question Number : 328 Question Id : 640653816192 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

## **Correct Marks : 3**

## **Question Label : Short Answer Question**

Consider a dataset X in  $\mathbb{R}^3$ . The dataset X consists of 4 samples with 3 features each. The covariance matrix C of this dataset has three non-zero eigenvalues which follow the given linear equations:

 $2\lambda_1 + 3\lambda_2 - \lambda_3 = 5$  $\lambda_1 - 2\lambda_2 + 4\lambda_3 = 8$  $3\lambda_1 + \lambda_2 - 2\lambda_3 = 3$ 

Determine the variance of the given dataset.

Response Type : Numeric

Evaluation Required For SA : Yes

Show Word Count : Yes

Answers Type : Equal

Text Areas : PlainText

**Possible Answers :** 

5

Question Number : 329 Question Id : 640653816195 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

**Correct Marks : 3** 

# Question Label : Short Answer Question

Consider a dataset of *n* observations  $\{x_1, x_2, ..., x_n\}$ , where each  $x_i$  follows a Bernoulli distribution with parameter *p*, i.e.,  $x_i \sim \text{Bernoulli}(p)$  for i = 1, 2, ..., n. However, you have reason to believe that the parameter *p* might differ for two distinct groups within the dataset. You suspect that there are two groups in the dataset, each with its own parameter( $p_1$  and  $p_2$ ). Now, develop an algorithm to estimate the parameters  $p_1$  and  $p_2$  using maximum likelihood estimation. Then, apply your algorithm to a dataset with the following observations and corresponding group labels:  $\{0, 1, 1, 0, 1\}$  and  $\{1, 0, 1, 0, 1\}$  for group 1 and group 2 respectively.

Calculate the maximum likelihood estimates of  $p_1$  and rounded to two decimal places.

# Response Type : Numeric

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

# 0.6

Question Number : 330 Question Id : 640653816198 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

# Correct Marks : 3

# Question Label : Short Answer Question

Consider the following decision tree for a classification problem in which all the data-points are constrained to lie in the unit square in the first quadrant. That is  $0 \ge x_1, x_2 \le 1$ . If a point is picked uniformly at random from the unit square, what is the probability that the decision tree predicts this point as belonging to class 1?



Response Type : Numeric

**Evaluation Required For SA :** Yes

Show Word Count : Yes

Answers Type : Equal

Text Areas : PlainText

**Possible Answers :** 

0.75

Question Number : 331 Question Id : 640653816199 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

**Correct Marks : 3** 

**Question Label : Short Answer Question** 

Consider the following dataset with 6 samples along with the corresponding labels. Each sample has three binary features  $f_1$ ,  $f_2$  and  $f_3$ .

sample	$f_1$	$f_2$	$f_3$	y
$x_1$	1	1	0	1
$x_2$	0	1	0	1
$x_3$	1	1	1	0
$x_4$	0	1	1	0
$x_5$	1	0	1	0
$x_6$	1	1	1	1

Assume that the features are conditionally independent given the label y. Suppose the test sample is  $x_{test} = [0, 1, 0]^T$ .

What is the estimated probability that the test point belongs to class 0 (that is,  $p(y = 0|x_{test}?))$ ?

# Response Type : Numeric

# Evaluation Required For SA : Yes

Show Word Count : Yes

Answers Type : Equal

Text Areas : PlainText

# **Possible Answers :**

0

# Question Number : 332 Question Id : 640653816200 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

# **Correct Marks : 3**

# Question Label : Short Answer Question

Consider a linearly separable binary classification data set with 1000 data points and 100 features. Assume that there exists a w such that ||w|| = 1,  $y_i(w^T x_i) \ge 0.5 \forall i$ . Also assume that  $||x||_2 \le 2 \forall i$  What is the maximum number of mistakes that the Perceptron algorithm can make in this data set?

# Response Type : Numeric

Evaluation Required For SA : Yes

# Show Word Count : Yes

Answers Type : Equal

#### Text Areas : PlainText

#### **Possible Answers :**

# 16

# Question Number : 333 Question Id : 640653816204 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

## **Correct Marks : 3**

# Question Label : Short Answer Question

Consider a single iteration of the AdaBoost algorithm that was run on three sample points, starting with uniform weights on the sample points. The labels are either +1 or -1 In the table below, some values have been omitted.

Data point	True label	Predicted label	Initial weight	Updated weight
$\mathbf{x}_1$	?	1	$\frac{1}{3}$	$\frac{1}{2}$
$\mathbf{x}_2$	-1	-1	$\frac{1}{3}$	?
$\mathbf{x}_3$	-1	?	$\frac{1}{3}$	$\frac{1}{4}$

Based on the above data, what will be the updated weight for point  $x_2$ ?

Response Type : Numeric

Evaluation Required For SA : Yes

Show Word Count : Yes

Answers Type : Equal

Text Areas : PlainText

**Possible Answers :** 

0.25

Question Number : 334 Question Id : 640653816207 Question Type : SA Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

**Correct Marks : 3** 

**Question Label : Short Answer Question** 

Consider a simple neural network with one hidden layer. The network has the following architecture:

Input layer with 3 neurons. Hidden layer with 2 neurons, using the sigmoid activation function. Output layer with 1 neuron, using the linear activation function.

The weights and biases for the network are as follows:

Hidden Layer:

Neuron 1: Weights: [0.5, -0.2, 0.8] Bias: 0.1

Neuron 2: Weights: [0.4, 0, 0.2] Bias: -0.4

Output Layer:

Neuron 1: Weights: [0.2, 0.4]

Bias: -0.3

Assume that the input values are [0.6, 0.3, 0.8].

Calculate output of Neuron 1 in hidden layer

**Response Type :** Numeric

Evaluation Required For SA : Yes

Show Word Count : Yes

Answers Type : Range

Text Areas : PlainText

## **Possible Answers :**

0.70 to 0.80

# BDM

Section Id :	64065356697
Section Number :	12
Section type :	Online
Mandatory or Optional :	Mandatory
Number of Questions :	20
Number of Questions to be attempted :	20
Section Marks :	30
Display Number Panel :	Yes
Section Negative Marks :	0
Group All Questions :	No
Enable Mark as Answered Mark for Review and Clear Response :	Yes
Maximum Instruction Time :	0
Sub-Section Number :	1
Sub-Section Id :	640653118949
Question Shuffling Allowed :	No
Is Section Default? :	null

Question Number : 335 Question Id : 640653816208 Question Type : MCQ Is Question Mandatory : No Calculator : None Response Time : N.A Think Time : N.A Minimum Instruction Time : 0

**Correct Marks : 0** 

Question Label : Multiple Choice Question

THIS IS QUESTION PAPER FOR THE SUBJECT "DIPLOMA LEVEL : BUSINESS DATA MANAGEMENT (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.