PSG COLLEGE OF TECHNOLOGY, COIMBATORE - 641 004

Department of Computer Science and Engineering BE CSE - G2 & SEM 6

CONTINUOUS ASSESSMENT TEST 1 Date: 20.1.2025

19Z601 - Machine Learning

Time: 1 Hour 30 minutes.

Maximum Marks: 50

INSTRUCTIONS: 1. Answer ALL questions. Each Question carries 25 Marks. 2. In each question, subdivision a contains 5 questions and the weightage of each

question is one mark, subdivision b(i) and b(ii) carries 5 marks each and subdivision c carries 10 marks each.

3. Subdivisions (a) and (b) will be with no choice and Subdivision (c) may be with choice but not in more than 1 question.

4. Course Outcome Table Qn. 1 CO1 Qn.2 CO2

| 1.a | | T === |
|-----|---|-------|
| i) | You are given a dataset containing austinia (5x1mark=5marks) | BTL |
| | You are given a dataset containing customer information, including their | L2 |
| | age, income, purchase history, and geographic location. The goal is to | |
| | behaviour to tailor marketing strategies | |
| - | Which type of machine learning technique is most suitable for this task? | |
| | ry capervised Learning | |
| | Unsupervised Learning | |
| | C) Reinforcement Learning | |
| | D) Semi-Supervised Learning | |
| i) | How does one-hot encoding handle categorical variables with multiple | 10 |
| | unique values, and what is a potential drawback of using this technique? | L2 |
| | A) It replaces each category with its frequency and can lead to biased | |
| | models. | |
| | B) It assigns a unique integer to each category, which may imply an | |
| | unintended ordinal relationship. | |
| k | Cylt creates a binary column for each category, which can increase | |
| | dimensionality significantly. | |
| | D) It combines similar categories into a six I | |
| | D) It combines similar categories into a single column, which may lose information. | 1676 |

| specified as data points (vector) [2,3,1,4,5,, 2]. Fill the missing data point with mode data imputation. v) The first principal component accounts for the variance in the data, while subsequent components account for the remaining variance in desceding order. | | | e following statemen | ts about types of | missing values in | а | |
|--|-----------------|---|--|---|-------------------------------------|----------|--|
| specified as data points (vector) [2,3,1,4,5, | V LA | 1. MCA data 2. MNA hand unob Which of the S) Both state 3) Both state 3) Only state | is unrelated to both of R (Missing Not at Ralled appropriately, served values. It following is correct? The ements are true. It is true. | observed and und andom) can lead as the missing | observed data. * to biased analysis | if not | |
| The first principal component accounts for the | m | pecified as nissing data | data points (vector) point with mode dat | [2,3,1,4, 5, a imputation. | 2, 2]. Fill the | | |
| Consider the training dataset of 4 instances. It contains the details of the performance of students and their likelihood of getting a job offer or not in their final semester. Apply the Find-S Algorithm. CGPA Interactiveness Practical Knowledge >=9 Yes Excellent Yes >=9 Yes Good Yes >=8 No Good No >=9 Yes Good Yes >=9 Yes Good Yes >=8 No Good No >=9 Yes Good Yes II) You are given a dataset containing the sepal width of Iris flower in centimeters: Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values (i) Min-Max Normalization (ii) Z-score Normalization | v) Th | he first prir ne data, w | ncipal component ac hile subsequent co | counts for the | high variand | ce in L2 | |
| Consider the training dataset of 4 instances. It contains the details of the performance of students and their likelihood of getting a job offer or not in their final semester. Apply the Find-S Algorithm. CGPA Interactiveness Practical Knowledge >=9 Yes Excellent Yes >=9 Yes Good Yes >=8 No Good No >=9 Yes Good Yes ii) You are given a dataset containing the sepal width of Iris flower in centimeters: Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values (i) Min-Max Normalization (ii) Z-score Normalization | b. | | | (2) | 5 marks = 10 ma | rks) | |
| >=9 Yes Excellent Yes >=9 Yes Good Yes >=8 No Good No >=9 Yes Good Yes ii) You are given a dataset containing the sepal width of Iris flower in centimeters: Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values for the above dataset. (i) Min-Max Normalization (ii) Z-score Normalization | in | their final s | emester. Apply the F | Find-S Algorithm. Practical | | | |
| >=9 Yes Good Yes >=8 No Good No >=9 Yes Good Yes ii) You are given a dataset containing the sepal width of Iris flower in centimeters: Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values for the above dataset. (i) Min-Max Normalization (ii) Z-score Normalization | - | >=9 | Yes | | Voc | | |
| >=8 No Good No >=9 Yes Good Yes ii) You are given a dataset containing the sepal width of Iris flower in centimeters: Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values for the above dataset. (i) Min-Max Normalization (ii) Z-score Normalization | 1 | >=9 | Yes | | | | |
| ii) You are given a dataset containing the sepal width of Iris flower in centimeters: Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values for the above dataset. (i) Min-Max Normalization (ii) Z-score Normalization | - | >=8 | No | | | | |
| Sepal Width (cm): [3.5,3,3.2,3.6] Use the following data scaling techniques and provide the scaled values for the above dataset. (i) Min-Max Normalization (ii) Z-score Normalization | | >=9 | Yes | Good | | | |
| Use the following data scaling techniques and provide the scaled values for the above dataset. (i) Min-Max Normalization (ii) Z-score Normalization | | | | | | | |
| | Us for | se the follow the above (i) Min-Ma | ring data scaling tech dataset. x Normalization | niques and provid | de the scaled value | es | |
| marks) | All Street Land | | | (1 x 10 | marks = 40 | | |
| mound, determine the version | | y the Can | didate Elimination M | lethod, determin | the version | ce L4 | |

2-10

(S and G boundaries) after processing the given dataset. Assume that "EnjoySport" is the target concept.

| Example | Sky | AirTemp | Humidity | Wind | Water | Forecast | EnjoySport |
|---------|-------|---------|----------|--------|-------|----------|------------|
| 1 | Sunny | Warm | Normal | Strong | Warm | Same | Yes |
| 2 | Sunny | Warm | High | Strong | Warm | Same | Yes |
| 3 | Rainy | Cold | High | Strong | Warm | Change | No |
| 4 | Sunny | Warm | High | Strong | Cool | Change | Yes |

Analyze the changes in the version space (S and G boundaries) after processing the dataset. Identify any inconsistencies in the data or concepts that might lead to conflicts within the version space

| 2.a | (5x1mark=5marks) | BTL |
|--------|---|-----|
| 2.a i) | Which of the following statements about linear regression models is TRUE? A) Linear regression can only be applied if the relationship between variables is non-linear. The slope coefficient in a simple linear regression represents the average change in the dependent variable for a one-unit increase in the independent variable. C) In multiple linear regression, adding more independent variables always improves the model's accuracy. | L2 |
| | D) Linear regression models do not assume any relationship between the independent and dependent variables. × | |
| ii) | Which of the following is an example of cultural bias in datasets? A) Missing values in the dataset. B) Overrepresentation of one demographic group leading to skewed predictions. C) Using a dataset that includes diverse geographic and cultural populations. D) Applying feature scaling to normalize data. | L3 |
| iii) | The following are the assumptions of linear regression: Assumption 1: Linearity Assumption 2: Heteroscedasticity Which of the following is correct? A) Both Assumptions are true. B) Both Assumptions are false. C) Only Assumption 1 is true. D) Only Assumption 2 is true. | L2 |

| iv) | Loss/Error of a Machine Learning (ML) model is decomposed into | LI |
|-----|--|-----|
| (V) | Min . D. Iannian / a | L1 |
| v) | High variance and low bias in a machine learning model can lead to Overfitting fitting of model. | |
| b. | (2 x 5 marks = 10 marks) | |
| i) | Mathematically derive the estimates of the unknown parameters in a linear regression model using the Ordinary Least Squares (OLS) method. | L5 |
| ii) | Let us assume a binomial logistic regression problem where the classes are pass and fail. The student dataset has entrance mark based on the historic data of those who are selected or not selected. Based on the logistic regression, the values of the learnt parameters are a ₀ = 1 and a ₁ = 8. Assuming marks of x = 60, compute the resultant class. | |
| c. | (1 x 10 marks = 10 marks) | |
| i) | You are given a dataset containing information about the number of hours students spend studying and their corresponding scores on a test. Your task is to perform simple linear regression to predict test scores based on the number of hours studied using the following dataset. | |
| | No.of Hours Studies Test Scores | 1 |
| | | |
| | 2 75 | |
| | 2 75 3 82 | 197 |
| | The state of the s | 17- |
| | 3 82 | |

PSG COLLEGE OF TECHNOLOGY, COIMBATORE - 641 004 Department of Computer Science and Engineering BE CSE - G2 & SEM 6

CONTINUOUS ASSESSMENT TEST 2 Date: 24.2.2025

19Z601 - Machine Learning

Time: 1 Hour 30 minutes.

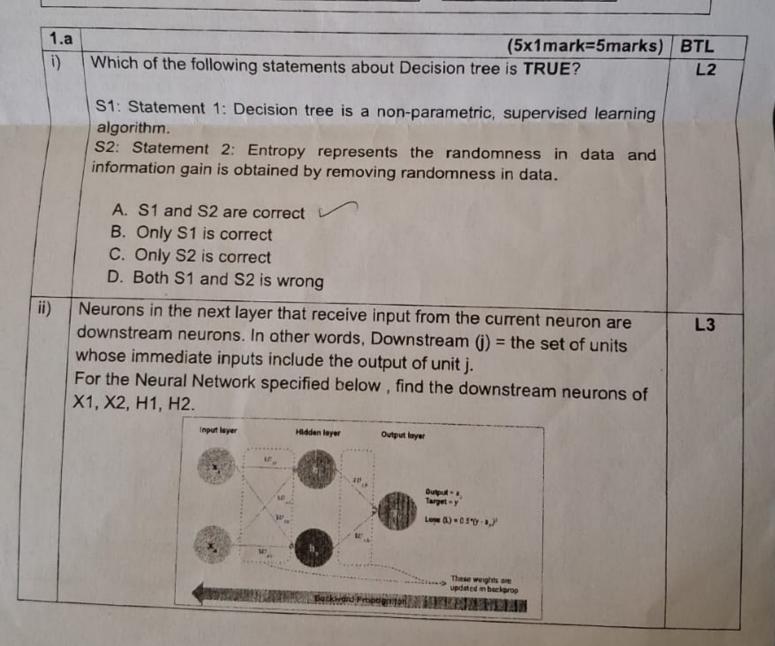
Maximum Marks: 50

INSTRUCTIONS:

- 1. Answer ALL questions. Each Question carries 25 Marks.
- In each question, subdivision a contains 5 questions and the weightage of each question is one mark, subdivision b(i) and b(ii) carries 5 marks each and subdivision c carries 10 marks each.
- 3. Subdivisions (a) and (b) will be with no choice and Subdivision (c) may be with choice but not in more than 1 question.
- 4. Course Outcome Table:

Qn. 1 CO3

Qn.2 CO4



| 122 | Which type of decision tree uses utilizes Gini impurity to identify the ideal | L1 |
|------|--|-----|
| iii) | attribute to split on. | |
| | A) ID3 | |
| | B) C4.5 | |
| | C) CART | |
| | D) None of the Above | |
| : | Percentron rule is suitable for liverally separable data. | L2 |
| iv) | The weight update rule in Perceptron training rule method is mathematically | L2 |
| V) | specified as we will Augo Awiz n(t-0) XI | |
| b. | (2 x 3 marks = 10 marks) | |
| i) | Using a perceptron model, design a two input perceptron that implements | L3 |
| '/ | the Boolean functions OR, NAND, NOR, and XOR. | |
| | a) Graphically represent on a 2D plane. | |
| | b) Determine if each function is linearly separable. | |
| | c) For linearly separable functions, derive appropriate weights and bias | |
| | values for a perceptron that implements them. | |
| | d) Explain why XOR cannot be implemented using a single-layer | |
| | perceptron. | 1.4 |
|) | Derive the backpropagation rule using stochastic gradient descent | L4 |
| | algorithm by considering the output unit case alone. (Squashing activation function can be used). | |
| | function can be used). | |
| 10 | (1 x 10 marks = 10 marks) | |
| | | L4 |
| | Construct a Decision Tree using the ID3 algorithm to classify whether a child will enjoy sport based on the given attributes. Calculate the Information | L4 |
| - | Construct a Decision Tree using the ID3 algorithm to classify whether a child will enjoy sport based on the given attributes. Calculate the Information Gain for each attribute at the root node and choose the best splitting | L4 |
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| | Construct a Decision Tree using the ID3 algorithm to classify whether a child will enjoy sport based on the given attributes. Calculate the Information Gain for each attribute at the root node and choose the best splitting attribute. Draw the resulting Decision Tree. | L4 |
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| | Construct a Decision Tree using the ID3 algorithm to classify whether a child will enjoy sport based on the given attributes. Calculate the Information Gain for each attribute at the root node and choose the best splitting attribute. Draw the resulting Decision Tree. Dry Weather Temperature Mumidily Wind Play? | L4 |

| 2.a | | (5v1mark=F- | | | | |
|------|--|------------------------------------|--|--|--|--|
| i) | Which are the statements true with respect to | (5x1mark=5m o Support Vector Ma | chine L2 | | | |
| | (SVM)? | | | | | |
| | S1: Statement 1 : SVM uses the kernel trick that takes as low | | | | | |
| | dimensional space as input and transforms to higher dimensional | | | | | |
| | space. | | | | | |
| | S2 : Support vectors are datapoints near to the hyperplane (decision | | | | | |
| | boundary) | | | | | |
| | A) Both S1 and S2 are true | | U.S. of the co. | | | |
| | B) Only S1 is true | | 1 1800 | | | |
| | D) Both S1 and S2 are not true | | - PCB DUS | | | |
| ii) | B) Only S1 is true C) Only S2 is true D) Both S1 and S2 are not true The Naïve Bayes algorithm mathematically s For the data points presented in the graph be function is applicable to project it into higher | specified as | - L3 | | | |
| iii) | For the data points presented in the graph by | elow what kind of ke | ernel L2 | | | |
| 111) | function is applicable to project it into higher | dimensional space | ? | | | |
| | | | 100 | | | |
| | 2D Space | | State Inches | | | |
| | | | | | | |
| 1 | | | | | | |
| 1 | | | | | | |
| 1 | | a way to the server. | | | | |
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| | | BAR . | 1 1 200 | | | |
| | A) Radial Basis Function | | | | | |
| | B) Polynomial Function | | | | | |
| | C) Sigmoid Function | | 1000 | | | |
| iv) | D) Linear Kernel function The Naïve part of Naïve Rever describe | | Annual State of the last | | | |
| | aigentilli is | | L1 | | | |
| b. | The hyperplane of SVM is represented math | | L1 | | | |
| i) | Apply the News Barre III | 2 x 5 marks = 10 m | arks) | | | |
| ,, | Apply the Naïve Bayes with Laplace smoothing algorithm to the following email example, predict this email is Spam or Ham: You Buy Valium! | | | | | |
| | Email | Label | | | | |
| | Buy Valium! | Spam | 11 11 11 11 | | | |
| | You good? | Ham | | | | |
| | The state of the s | | 10 10 10 10 10 10 10 10 10 10 10 10 10 1 | | | |
| | Valium help you. | Spam | | | | |
| | Valium help you. Good Valium help. I need Valium for my health condition. | Spam | | | | |

| ii) | A hospital wants to classify whether a patient has a disease (Yes or No) based on medical test results using Support Vector Machine (SVM) and | L5 |
|------|---|------|
| | Naïve Bayes (NB). Compare SVM and Naïve Bayes in terms of: | 12 |
| | Assumptions | 1000 |
| | Handling of numerical medical data | |
| | Performance on small vs. large datasets | 19 |
| 1 12 | Computational complexity | |
| | Sensitivity to missing or noisy data | |
| | Architectures (Mathematical Intuition) | |
| C. | (1 x 10 marks = 10 marks) | L4 |
| 1) | Compute the suitable SVM that accurately discriminates the following two classes. Positively labeled points | |
| | $\binom{3}{1}, \binom{3}{-1}, \binom{6}{1}, \binom{6}{-1}$ | |
| | Negatively labeled points | |
| | $\binom{1}{0}$, $\binom{0}{1}$, $\binom{0}{-1}$, $\binom{-1}{0}$ | |
| | Identify the support vectors, show the SVM architecture, slope and the decision hyperplane. Show all necessary diagrams. | |
| | Consider the point $\binom{3}{1}$. Predict the label. | |
| 1 | What are the interpretations from the obtained final hyperplane ? | 1 |

PSG COLLEGE OF TECHNOLOGY, COIMBATORE - 641 004 Department of Continue and Engineering

Department of Computer Science and Engineering

III BE CSE - G2, 6th Semester CONTINUOUS ASSESSMENT TEST 3

19Z601 - Machine Learning Date: 15.04.2025

Time: 1 Hour 15 Minutes

Maximum Marks: 35

INSTRUCTIONS:

- 1. Answer ALL questions.
- 2. Question No. 1 carries 8 marks and question No. 2 carries 27 marks 3. In question No. 1, subdivision a carries total of 8 marks (one mark for each question).

 4. In question No. 2
- 4. In question No. 2, subdivision a carries total of 7 marks (one mark for each question), subdivisions h(t) and the subdivision as carries total of 7 marks (one mark for each question), subdivisions b(i) and b(ii) carries 5 marks each and subdivision c carries 10 marks.
- 5. Course Outcome Table:

CO 1 to 4 Qn. 1

Qn.2

1. a

(8 x 1 mark = 8 marks)

Which of the following is not a supervised machine learning algorithm? i. a) K-means b) Naïve Bayes c) SVM for classification problems d) Decision tree [CO1 - Remember] model is a generative model used in ML ii. a) Naive Bayes b) Linear Regression c) Logistic Regression d) Support vector machines [CO2 – Understand] How is the model's performance affected by the bias-variance trade-off? iii. a) High bias can lead to underfitting, and high variance can lead to overfitting b) Increasing bias improves model accuracy c) Low bias and low variance lead to better ML models d) Increasing variance improves model generalization [CO2 - Understand]

| iv. | The first principal component accounts for the variance in the data, we subsequent components account for the remaining variance in descending order. |
|-----|--|
| | subsequent components account is an remaining variance in descent |
| | |
| | The second secon |
| | THE RESERVE OF THE PARTY OF THE |
| v. | The Naïve part of Naïve Bayes algorithm is[CO4 – Understand] |
| | CO4 = Chacistans |
| vi. | Neurons in the next layer that receive input from the current neuron are downstream neurons. In other words, Downstream (j) = the set of units whose immediate inputs include the output of unit j. For the Neural Network specified below, find the downstream neurons of X1, X2, H1, H2. |
| | Input layer Hidden layer Output layer |
| | |
| | LO _A k Output ***, |
| | Loss (3.) = 0.5°(y - 0.)* |
| | |
| | These weights are undered in tacknoon |
| | Backward Propagation |
| | Input Layer Neurons: X1, X2; Hidden Layer Neurons: H1, H2; |
| | Output Layer Neurons : Yk |
| | |
| i. | Which type of decision tree uses utilizes Gini impurity to identify the ideal attribute to |
| 300 | split on. |
| | A) ID3 |
| | B) C4.5 |
| | C) CART |
| | D) None of the Above |
| | [CO3 – Remember] |
| i. | Which are the statements true with respect to Support Vector Machine (SVM)? |
| 200 | \$1. Statement 1: SVM uses the kernel trick that takes as low dimensional space as |
| | input and transforms to higher dimensional space. |
| | S2: Support vectors are datapoints near to the hyperplane (decision boundary) |
| - 1 | A) Both S1 and S2 are true |
| | B) Only S1 is true |
| 1 | C) Only S2 is true |
| | D) Both S1 and S2 are not true [CO4- Remember] |
| | $(7 \times 1 \text{ mark} = 7 \text{ mark})$ |
| | Given the two vectors $A = [1, 2]$ and $B = [2, 3]$, what is the cosine distance between |
| _ | |
| 1 | them? |

| | A) 0.07 |
|------|--|
| 1 | B) 0,22 |
| | C) 0.33 |
| | D) 0.45 |
| | |
| ii) | What is the Love 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 |
| | What is the Levenshtein (edit) distance between the words "form" and "fork"? A) 0 |
| | B) 1 |
| | C) 2 |
| | D) 3 |
| | D) S |
| iii) | [CO5-Apply] |
| ш | In Q-learning, what does the Q-value represent? A) The action taken be all |
| 1 | A) The action taken by the agent in a given state. B) The expected future reward for a given state-action pair. C) The discount feature reward for a given state-action pair. |
| | The discount factor used in the learning process |
| 1 | D) The learning rate used in updating the Q-values. |
| iv) | Which of the following statements is to be a second of the following statements in the second of the following statements is to be a second of the following statements in the second of the following statements is to be a second of the following statements in the second of the following statements is to be a second of the following statements in the second of the following statements in the second of the following statements in the second of the s |
| | Which of the following statements is true about k-means clustering? |
| | A) K-means clustering is a supervised learning algorithm. B) K-means clustering requires the |
| 100 | B) K-means clustering requires the number of clusters (k) to be known beforehand. C) K-means clustering can only be used for numerical data. |
| | D) K-means clustering always converges to the global minimum. |
| (v) | Which of the following is not a type of clustering algorithm? |
| 1 | A) K-means clustering B) Hierarchical clustering |
| 1 | C) DBSCAN (Density-Based Spatial Clustering of Applications with Noise) D) K-Nearest Neighbors (KNN) |
| | D) K-Ivearest Iveignoots (KIVIV) |
| vi) | L1 distance measure is otherwise known as [CO5- Remember] |
| vii) | L∞ distance measure is represented mathematically as |
| | [CO5- Understand] |
| | (o o o nacistana) |
| b. | Calculate the SMC and Jacond Similarity C. 55 : 2 marks = 10 marks) |
| i) | Calculate the SMC and Jaccard Similarity Co-efficient for the following two binary vectors. |
| | |
| 1 | $\mathbf{x} = (1, 0, 0, 0, 0, 0, 0, 0, 0, 0)$ $\mathbf{y} = (0, 0, 0, 0, 0, 1, 0, 0, 1)$ |
| | $f_{01} = 2$ the number of attributes where x was 0 and y was 1 |
| | $f_{10} = 1$ the number of attributes where x was 0 and y was 1 |
| 1 | $f_{00} = 7$ the number of attributes where x was 1 and y was 0 |
| | $f_{11} = 0$ the number of attributes where x was 1 and y was 1 |
| | |
| | [CO5- Apply] |
| | a 2 of A |

| | | | 2 0 | | | | | | |
|-----|--|-----------|-----------|--------------------|--|------------|-------------|-----------|--|
| ii) | What is Reinforcemen | t Learnin | ig ! Exp | lain the st | eps invol | ved in Q I | earning. | | |
| | [CO5- Understand] | | | | | | | | |
| | | | | | | 10 | | | |
| | (1 x 10 marks = 10 marks) | | | | | | | | |
| c. | | | - | | | (1 x 10 | marks | ing where | |
| 1) | For the given data, con | mpute tw | o cluste | rs using k | C-means a | algorithm | for cluster | ning when | |
| | initial cluster centers a | re (1.0,1 | .0) and (| 5.0,7.0). | Execute f | or two ite | rations. | | |
| | T | Record N | umber | A | | В | | | |
| | | | 1 | 1.0 | THE RESERVE AND ADDRESS OF THE PERSON NAMED IN | 1.0 | | | |
| | | | 2 | 1.5 | STATE OF TAXABLE PARTY. | 2.0 | | | |
| | | R | 3 | 3. | STATE OF TAXABLE PARTY. | 4.0 | 316 | | |
| | | R4 | | 5. | 0 | 7.0 | | | |
| | | R | 5 | 3. | 5 | 5.0 | 7770 | | |
| | | R | 6 | 4. | 5 | 5.0 | | | |
| | Marine San Control of | R | 7 | 3. | 5 | 4.5 | - 100 | 05- Appl | |
| | | | | | | | ICC |)5 APP | |
| | | | | OR | | i ala li | ak and co | mplete li | |
| () | Use the distance matr | ix in tab | le given | below t | o perform | dandrogr | am. The | dendrogra | |
| | merarchical clustering | g. Show | your res | sults by a | rawing a | deriano | | | |
| | should clearly show th | he order | in which | the point | is are mei | geu. | | | |
| | | 1 | P1 | P2 | Р3 | P4 | P5 | | |
| | I The same of | D1 | 100000 | 0.10 | 0.41 | 0.55 | 0.35 | - | |
| | | P1 | 0,00 | MATERIAL PROPERTY. | | 0.47 | 0.98 | | |
| | - | P2 | 0.10 | 0.00 | 0.64 | | | | |
| | | P3 | 0.41 | 0.64 | 0.00 | 0.44 | 0.85 | | |
| | 1000 | P4 | 0.55 | 0.47 | 0.44 | 0.00 | 0.76 | | |
| | The state of the s | P5 | 0.35 | 0.98 | 0.85 | 0.76 | 0.00 | | |

[CO5-Apply]