FF No. 868

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Bansilal Ramnath Agarwal Charitable Trust's VISHWAKARMA INSTITUTE OF TECHNOLOGY, PUNE – 411037.

(An Autonomous Institute Affiliated to Savitribai Phule Pune University) **Examination: ESE**

Year: S.Y. Common

Branch:

Subject: Data science

Subject Code: MD 2201

Max. Marks:60

Total Pages of Question Paper: 1

Day & Date: Wed 22/11/23

Time: 10.30 am -12.30 pm

Instructions to Candidate

1. All questions are compulsory.

2. Neat diagrams must be drawn wherever necessary.

3. Figures to the right indicate full marks.

| Q.No. | CO No | BT No | | 7 : 57 .7., | 5 20 21 18 | i on C | 5.03 | 100 100 100 100 100 100 100 100 100 100 | | Ma |
|--------------|-------|---|---|--|--|-----------------------------------|--------------------|---|--|--------------------------|
| Q.1. | 1 | 1 | Part of an ani | ilial transpa | rangy rong | 4 multi-1; -1 - 1 | | | | mar |
| | | 1 | i O | as shown be | elow – | t published | by a lead | ing multination | onal technology | 12 |
| | | inii Ls | Country | CR_req | CR_ compl in | UD_ req | UD compl | Hemi 2 | HDI | |
| | | 1 | Austria | 21 | 100 | 134 | 32 | Southern | High | |
| | 11 | | Belgium | 10 | 33 | 361 | 73 | Northern | High | |
| | W. H. | | Brazil | 224 | 67 | 703 | 82 | Southern | Medium | |
| | | 1 | Somalia | 104 | 31 | .227 | 61 | Southern | Poor | |
| | 1., | | USA | 92 | 63 | 5950 | 93 | Northern | High | , (17-11-11-1 |
| Q. 2. (A) | 2 | 2 | CR_req: Cont CR_comp: Co UD_req: User UD_compl: U Hemi: Hemisp HDI: Human I Identify each Categorical or What are type minimized wh | ontent remove Data requests of Data requests of Data considered to the Constant of the Constan | val complian sts mpliance in at Index as Discrete stegorical wi c-II errors i | Numerical th justificate hypothes | ion is testino? | Which erro | The state of the s | 4 |
| (B) | 2 | 2 | A sample of 50 week in reading with a standard on this data. | g me newsp | aper. Ine gr | oup in the | sample had | tian average | 6221 | 6 |
| Q. 3. (A) | 3 | 1 | Calculate the di Distance metric | and II. Euc | ndean Distai | nce metric | The state of | | | 4 |
| В) | 3 | 2 | For a given uni local minimum | and global r | nınımum | 1 2 | made of the | | 1 | 4 |
| Q. 4. A) | 4 | 4 22 (2) 2 (2) 3 (2) 4 (2) 4 (2) | As an outcome obtained. SSR a as sum of squaregression fit is | lso known as es due to en | s sum of square squar | ares due to r | egression = | 92 48 SSF a | lso known | 4 |
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| | | . A., | | | .8 | 13 1 2 2 2 2 C | 80 1 | | 1 1 2 1 1 1 1 1 |
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| | | | (1/1+ | e -(-4.07778 + 1.50 | 46 * Savings)). | after Maximum | Likelihood estima | ation is given as Π = | |
| | | | Answer | Based on the | plicant with pplication be prediction | e processed for s made using lo ions would mate | loan disbursemen gistic regression ch with the actua | pproaches the bank t or rejected? for all 10 data, how I loan status? Wha on as a classifier? | v |
| . 5. | 5 | 4 | The train | ning data for | a supervised | d classification i | s as follows - | 1,11 | 6 |
| .) | | | $X_1 = (1.8)$ $X_6 = (7.3)$ Use | $X_{1}=0$, $X_{2}=0$, $X_{3}=0$, $X_{4}=0$, $X_{5}=0$ | (2,1.8, 1), (6.5, 4.2, 2) (6.5, 4.2, 2) (hbor assign | $X_3 = (3.2, 2.4, 10.2), X_8 = (7.0, 4.8, 10.2)$ appropriate classification in $X_1 = (3.2, 2.4, 10.2)$ |), X₄= (2.4,2.6,2). The test datapass to the test poirappropriate class | iss to the test point | |
| | | | 3. | Weighted / N | Modified K | - Nearest Neigl | hbor with $k = 3 t$ | o assign appropria | te |
| | - | 4 | 10 11 | class to the to | hat are calif | t in 3 classes wit | th 12, 28 and 10 s | amples respectivel | ly. 4 |
| 5. | 5 | 4 | Calculate | the Entropy | impurity, | Gini impurity | and Misclassifica | ation impurity at t | he |
| 1 | HIST. | 111/2/14 | node. | | TO THE H | | | Hard and the late of the second | 6 |
| 11/15 | 200 to 11 / 12 / | | | | The second section of the section of | On | The bottom of the late | CONTRACTOR MANUEL CANADA | |
| I Fa | 200 | Sanian, | 16-11-17 | altomate o | ntion for O | OR 5 (A) AND O | 5 (B) together as | O. 5 (C) | |
| nd | | danie, a Life of | | alternate o | | 5 (A) AND Q. | 5 (B) together as | Q. 5 (C) | 10 |
| | 5 | 4 | | alternate o | | | | Q. 5 (C) | 10 |
| | 5 | 4 | | alternate o | | 5 (A) AND Q ng training data | Time | Good Work | v |
| | 5 | 4 | The table | alternate of below gives Worker | the followi | 5 (A) AND Q. ng training data | Time | Good Work Quality? | v |
| | 5 | 4 | The table Sr. No. | alternate of the below gives Worker Sam | Mood Bad | 5 (A) AND Q ng training data Job Painting | Time Morning | Good Work | v l |
| | 5 | 4 | The table | alternate of below gives Worker Sam Sam | Mood Bad Good | 5 (A) AND Q ng training data Job Painting Plumbing | Time | Good Work Quality? Yes | v |
| | 5 | 4 | The table Sr. No. 1 2 3 | alternate of below gives Worker Sam Sam Ashwin | Mood Bad Good Bad | 5 (A) AND Q ng training data Job Painting | Time Morning Evening | Good Work Quality? Yes Yes No No | v l |
| | 5 | 4 | The table Sr. No. 1 2 3 4 | alternate of below gives Worker Sam Sam | Mood Bad Good | 5 (A) AND Q. ng training data Job Painting Plumbing Painting Plumbing Washing | Morning Evening Morning Evening Morning Morning | Good Work Quality? Yes Yes No No Yes | v l |
| | 5 | 4 | The table Sr. No. 1 2 3 | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sam | Mood Bad Good Bad Bad Good Good | 5 (A) AND Q. ng training data Job Painting Plumbing Painting Plumbing Washing Washing | Morning Evening Morning Evening Morning Evening Morning Evening | Good Work Quality? Yes Yes No No Yes Yes | v l |
| | 5 | 4 1.4.10 | The table Sr. No. 1 2 3 4 5 6 7 | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sam Sam | Mood Bad Good Bad Bad Good Good Good | Job Painting Plumbing Painting Plumbing Plumbing Washing Washing Painting | Morning Evening Morning Evening Morning Evening Morning Evening Morning | Good Work Quality? Yes Yes No No Yes Yes Yes Yes | v l |
| | 5 | 4 | The table Sr. No. 1 2 3 4 5 6 7 8 | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sam Sham Sham | Mood Bad Good Bad Bad Good Good Good Good Bad | Job Painting Plumbing Plumbing Plumbing Washing Washing Painting Plumbing | Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening | Good Work Quality? Yes Yes No No Yes Yes Yes Yes No | v l |
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| | 5 | 4 tantin material | The table Sr. No. 1 2 3 4 5 6 7 8 9 10 | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sam Sham Sham Ashwin Ashwin | Bad Good Bad Good Good Good Good Bad Good Good Good Bad Bad Good | Job Painting Plumbing Plumbing Plumbing Washing Washing Plumbing Washing Washing Washing Washing | Morning Evening | Good Work Quality? Yes Yes No No Yes Yes Yes Yes No | v l |
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| | 5 | 4 i.i.a. in ii.a. in iii.a. in | The table Sr. No. 1 2 3 4 5 6 7 8 9 10 11 12 13 | alternate of below gives Worker Sam Sam Ashwin Ashwin Sam Sham Sham Ashwin Ashwin Sham Ashwin Sham Sham Sham Sham Sham Sham | Bad Good Bad Good Good Bad Good Good Good Bad Bad Good Good Bad Bad Good | Job Painting Plumbing Plumbing Plumbing Washing Washing Plumbing Washing | Morning Evening Evening Evening Evening Evening Evening Evening Evening Evening | Good Work Quality? Yes Yes No No Yes Yes Yes Yes No No No Yes Yes No No Yes Yes No Yes | v l |
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| 5. | 6 | 2 | The table Sr. No. 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 15 Using Nal given feat Morning} Explain the second of the se | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sham Sham Ashwin | Mood Bad Good Bad Good Good Bad Bad Good Good Bad Good | graining data Job Painting Plumbing Painting Plumbing Washing Washing Painting Plumbing Washing Washing Washing Washing Washing Washing Washing Washing Washing Painting Washing Washing Washing Washing Washing Painting Washing Painting Washing Plumbing Painting Washing Plumbing Painting Washing Plumbing Painting | Morning Evening Morning Horning Morning Morning Morning Morning Morning | Good Work Quality? Yes Yes No No Yes Yes Yes Yes No No Yes Yes No No Yes Yes No No Yes Yes No The painting, Time = | v (; |
| 5. | | tantin matan ma matan matan matan matan ma matan ma ma ma ma ma ma ma ma ma ma | The table Sr. No. 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 Using Nal given feat Morning} Explain th | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sham Sham Ashwin | mood Bad Good Bad Good Good Bad Good Bad Good Bad Good Good Bad Good Good Bad Good | graining data Job Painting Plumbing Painting Plumbing Washing Washing Painting Plumbing Washing | Morning Evening Morning Horning Morning Morning Morning Morning | Good Work Quality? Yes Yes No No Yes Yes Yes Yes No No Yes Yes No No Yes Yes No The Painting, Time = | v (; |
| 5. | 6 | 2 | The table Sr. No. 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 Using Nal given feat Morning} Explain th | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sam Sham Sham Ashwin Sam Sham Sham Sham Sham Sham Sham Sham | Mood Bad Good Bad Good Good Bad Bad Good Bad Good Bad Good Good Bad Good | graining data Job Painting Plumbing Painting Plumbing Washing Washing Washing Washing Washing Washing Washing Washing Washing Painting Washing Washing Washing Washing Washing Washing Washing Washing Painting Section of the classes | Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening Evening Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening For Good Worning Morning Morning Evening For Good Worning Morning Evening For Good Worning Morning Evening For Good Worning For For For Good Worning For | Good Work Quality? Yes Yes No No Yes Yes Yes Yes No No Yes Yes No Yes Yes No Yes Yes No The painting, Time = | v (; |
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| 5. | 6 | 2 | The table Sr. No. 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 Using Nal given feat Morning} Explain the A Confusiused to de Authentic | alternate of below gives Worker Sam Sam Ashwin Ashwin Ashwin Sam Sham Sham Ashwin Sam Sham Sham Sham Sham Sham Sham Sham | Mood Bad Good Bad Good Good Bad Bad Good Bad Good Bad Good Good Bad Good | graining data Job Painting Plumbing Painting Plumbing Washing Washing Painting Plumbing Washing Painting Painting Red in planning took, is observed and spam message ass | Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening Evening Morning Evening Morning Evening Morning Evening Morning Evening Morning Evening For Good Worning Morning Morning Evening For Good Worning Morning Evening For Good Worning Morning Evening For Good Worning For For For Good Worning For | Good Work Quality? Yes Yes No No No Yes Yes Yes No No Yes Yes No Yes Yes No Yes Yes No Yes Yes No The Painting, Time = sting data sarning algorithm, positive class as — | v (; |

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Bansilal Ramnath Agarwal Charitable Trust's VISHWAKARMA INSTITUTE OF TECHNOLOGY, PUNE – 411037.

(An Autonomous Institute Affiliated to Savitribai Phule Pune University)

Examination: ESE

Year: S.Y. Common

Branch:

Subject: Data science

Subject Code: MD 2201

Max. Marks:60

Total Pages of Question Paper: 1 + 1

Day & Date: Thursday, 11/05/23

Time: 11.00 am -1.00 pm

Instructions to Candidate

1. All questions are compulsory.

2. Neat diagrams must be drawn wherever necessary.

3. Figures to the right indicate full marks.

| Q.No. | CO No | BT No | | Max mark: |
|--------------|-------|----------|---|--------------|
| Q.1. | 1 | 1 | Match the pair and explain each term on written on left side (shown in numbers) with an example – 1. Validity A. Degree to which data is consistent | 8 |
| | | | 2. Accuracy B. Degree to which data conforms to constraints | 1 |
| | har | 1-1 | 3. Uniformity C. Degree to which data is close to true values D. Degree to which data is specified using same Units | |
| Q. 2. (A) | 2 | 2 | What is a 'Choose function' in Binomial distribution. What is the probability of getting a desired outcome of '3', 4 times when | 6 |
| В) | 2 | 2 | A group of researchers want to test the possible effect of an epilepsy medication taken by pregnant mothers on the cognitive development of their children. As evidence, they want to estimate the IQ scores of 3-yr old children born to mothers who were on this medication during of the country provings studies suggest that the SD of IQ scores is 18 points. | 6 |
| | | | How many such children should the researchers sample in order to obtain a 95% confidence interval with a margin of error ≤ 4 points? | |
| . 3. | 3 | I | (Z* for 95% Confidence Interval = 1.96) Calculate the distance between points A (4,5,6) and B (7.2,2.8,6.8) using | 6 |
| 4) | | | i. Manhattan Distance metric and ii. Euclidean Distance metric For a given univariate function $f(x) = 3x^4 - 4x^3 - 12x^2 + 3$ find out the | 4 |
| 3) | 3 | 2 | 1 1 and minimum and global minimum value. | 6 |
|). 4. A) | 4 | 4 | Two wheeler sales of a leading brand are modelled by a Multiple Elliean Regression equation – Sales = - 22036.74 – 1007.61 Unit Cost + 358.88 Ad Exp + 674. 92 Prom Exp. Unit Cost of the Vehicle is in 'Thousand Rs, Ad Exp. is Rs. in Lac and Prom Exp. is also Rs. in Lac. | U |
| 3) | 4 | | Interpret the four coefficients. How is the figure of merit R ² calculated in regression? Explain with a neat sketch. Also justify that value of R ² tending to 1 indicates a good fit of the | 4 |
| 5. | 5 | 4 | The training data for a supervised classification is as follows – $X_1 = (1.8, 1.4, 1), X_2 = (2.2, 1.8, 1), X_3 = (3.0, 2.4, 1), X_4 = (2.2, 2.4, 1), X_5 = (6.2, 4.4, 2), X_6 = (7.2, 4.4, 2), X_7 = (6.6, 4.4, 2), X_8 = (7.0, 4.6, 2).$ The test datapoint is at $(4.4, 3.3)$. Use K-nn approach with $k = 3$ to assign appropriate class to the test point | 6 |

| Q. 5. (B) | 5 | 4 | respect | ively. Cal | lculate tl | re split in 3 cla ne Entropy i at the node. | sses with 8, 2 mpurity, Gir | 0 and 12 samples ni impurity and | 6 |
|--------------|-----------|---------|--------------|---------------|---------------|---|--------------------------------|--|----|
| | + | - | Wilsom | 35Hicution | mparity | OR | | | |
| | | | *alto | ernate optio | on for O.5 | (A) AND Q. | 5 (B) togethe | r as Q. 5 (C) | |
| Q. 5 | 5 | 4 | The tab | le below g | ives the f | ollowing train | ing data – | | 12 |
| (C) | İ | | 1110 1410 | | | | | , | |
| | | | Sr | Cook | Mood | Cuisine | Time | Tasty | |
| | | | No | 1 | d'a | , 1 | | | |
| | | | | Sita | Bad | Indian | Lunch | Yes | |
| | | | 2 | Sita | Good | Mexican | Dinner | Yes | |
| | | | 3 | Asha | Bad | Indian | Lunch | No | |
| | | | 4 | Asha | Bad | Mexican | Dinner | No | |
| , | | | 5 | Asha | Good | Thai | Lunch | Yes | |
| | | | 6 | Sita | Good | Thai | Dinner | Yes | |
| | | - W -41 | 7 | Sham | Good | Indian | Lunch | Yes | |
| | | | 8 | Sham | Bad | Mexican | Dinner | No | |
| | | | 9 | Asha | Bad | Thai | Lunch | No | |
| | | | 10 | | Good | Thai | Dinner | Yes | |
| | | | 11 | | Good | Indian | Dinner | Yes | |
| | | | 12 | | Bad | Thai | Lunch | No | |
| | | | 13 | | Good | Thai | Dinner | Yes | |
| | | | 14 | | Bad | Mexican | Lunch | Yes | |
| | | | 15 | Sham | Good | Indian | Lunch | No . | |
| | 4 | g- | | | . 476 | | in the state of | (T) 1 Southo | |
| | 2 | | Heing N | Jaïve Bays | Classific | cation, estimat | e the class to | r 'Tasty' for the Cuisine = Indian, | |
| | | - 1 | oiven fe | ature vecto | $or X = \{ C$ | look = Sham, N | Mood = Bad, G | Cuisine = Indian, | |
| N. Terri | the great | Nogan. | Time = | Lunch } | 3 | | | | |
| | | | | il - fla folo | l cross va | lidation' appro | oach used in p | planning the | 4 |
| Q. 6. | 6 | 2 | | | | | | | |
| (A) | | 1 | training | and testing | · - Fam o c | lassification e | exercise return | ns the following | 4 |
| (B) | 6 | 3 | | TD - C | 10/2 IN | = 0.912.11 | U.17, 111 | .06. | |
| | | olica | values - | · Accurac | v precisio | on, recall and | f-score | 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1 | |
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| Reg.No. | |

Bansilal Ramnath Agarwal Charitable Trust's VISHWAKARMA INSTITUTE OF TECHNOLOGY, PUNE – 411037.

(An Autonomous Institute Affiliated to Savitribai Phule Pune University)

Examination: ESE

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Subject: Data science

Subject Code: MD 2201

Max. Marks:60

Total Pages of Question Paper: 1

Day & Date: Monday, 19/12/22

Time: 8.30 am -10.30 am

Instructions to Candidate

1. All questions are compulsory.

2. Neat diagrams must be drawn wherever necessary.

3. Figures to the right indicate full marks.

| Q.No. | CO No | BT | | Max marks |
|-------|------------|----|--|--------------|
| | | No | - A L VImpla | 4 |
| 2.1. | 1 | 1 | What is Code book or meta data? Explain with an example. | |
| (A) | 1 | | | 4 |
| (B) | 1 | 1 | Explain with examples the terms – raw data and processed data | 3 |
| Q. 2. | 2 | 2 | Distinguish between point estimate and confidence interval | |
| A) | | | is an along 12 How it regulates the | 6 |
| B) | 2 | 2 | What is the importance of significance level? How it regulates the | |
| , | | | What is the importance of significant and type 2 errors? possibility of occurrence of type 1 and type 2 errors? How are the Margin of Error and Standard error related with each | 3 |
| C) | 2 | 2 | How are the Margin of Error and Standard error rem | |
| | AND IT MAN | 1 | other? Show with the help of an example, L1 | 6 |
| Q. 3. | 3 | • | State the formula for Lp norm. Show What the formul | 4 |
| (A) | | 2 | metric distance is always larger than L2 metric distance metric distance is always larger than L2 metric distance. Draw a typical 'n x n' hessian matrix. How is it used in optimization? Draw a typical 'n x n' hessian matrix. How is it used in optimization? Draw a typical 'n x n' hessian matrix. How is it used in optimization? | 6 |
| (B) | 3 | 4 | Draw a typical 'n x n' hessian matrix. How is it used in optimized by the property of the state of the property of the state of the sta | |
| Q. 4. | 4 | 4 | 1 10 10 CCE = 7 /1. What is the value | 1 |
| (A) | | | What are dichotomous variables in the context of Logistic regression? | 4 🔻 |
| | | | TVI -t ora dichotomous variables in the context of Logistic regions | |
| (B) | 4 | 1 | g: evamples | 6 |
| | | | Give some charge eight points (with (x, y) representing rooms), | |
| Q. 5. | 5 | 4 | Cluster the following eight points (with (x, y) representing locations) into three clusters: A1(2, 10), A2(2, 5), A3(8, 4), A4(5, 8), A5(7, 5), A8(4, 9). Initial cluster centers are: A1(2, 10), A4(5, 10), A3(4, 9). | |
| (A) | | | | |
| | | | 8) and A7(1, 2), A6(3,7) 8) and A7(1, 2). OR Apply K-nn and predict the class for the test point (3,7) for k=3. | , |
| | | | 8) and A/(1, 2). On and predict the class for the test point (3,7) and (| |
| Q. 5. | 5 | 4 | Apply K-nn and predict the cases (x,y,class) Training points with class are (x,y,class) (3.4.1) (1.4.1),(2,5,2),(3,8,1) | |
| (A) | | | Training points with class are (x,y,state) (7,7,2), (7,4,2), (3,4,1), (1,4,1),(2,5,2),(3,8,1) How do you define Genie impurity and entropy impurity? What will How do you define Genie impurity and entropy impurity? Why is | 4 |
| | | | (7,7,2), (7,4,2), (3,3,7) Genie impurity and entropy impurity. | |
| Q.5. | 5 | 3 | How do you define Genie impurity and their values be, for the purest node? their values be, for the purest node? Why is | 4 |
| (B) | | ٠ | How do you define General How do you define General How would you execute the k-fold cross-validation strategy? Why is How would you execute the k-fold cross-validation strategy? Why is how would you execute the k-fold cross-validation strategy? | |
| Q. 6. | 6 | 2 | How would you execute the R 1915 Property of | 6 |
| (A) | | | Leave-one -out most for a classification exercise returns the real real real real real real real rea | - |
| (B) | 6 | 3 | A Confusion matrix for a classification exceeds A Confusion matrix for a classification exceeds Values – TP = 0.962, TN = 0.93, FP = 0.12, FN = 0.07. | |
| (-) | | | values - TP = 0.962, TN = 0.93, The property and f- | |
| | | | Calculate Accuracy, precision, recall, sensitivity, specificity and f- | |
| | | | Calculate Accuracy, F- | |
| | | | score | |