

Subject Name: Machine Learning

Subject Code: 3170724

| | QUESTIONS | |
|-----------|--|-------|
| | UNIT NO- 1: | |
| | Introduction to Machine Learning | |
| | TOPIC:1: Human Learning and its Types, Machine Learning and its | |
| | Types | |
| Sr. No | SHORT QUESTIONS/MCQ | Marks |
| 1. | Identify correct groups of machine learning problems: a) Given a transaction, label as fraud or regular, learn fraud detection. b) Given set of movies found on a portal and group them into set of movies with same genre. c) Given database of visitors, automatically discover popularity of holiday place segments and group visitors according to holiday place. d) Find pattern of purchasing electronics during Diwali. [LJIET] | 01 |
| | 1-supervised learning, 2,3,4-unsupervised 1-classification 2-Regression, 3-4 Association rule 1,2- supervised learning 3,4- unsupervised learning 1,2- unsupervised learning 3,4-supervised learning 1,3-supervised learning 2,4-unsupervised learning 1,2,3-supervised learning 4-unsupervised learning | |
| 2. | Training a dog with sound of a bell to pick a ball. Rat learning how to carefully step on Skinner box to release food. Child plays game with friends in recess after observing other adults also play games at home. | 01 |
| 1 | a. Example 1 is type of learning under expert guidance b. Example 2 is type of learning by self c. Example 3 is type of learning under expert guidance d. Example 1,3 both are type of learning under expert guidance e. Example 2,3 both are type of learning by self f. Example 1,2 both are type of learning by self [LJIET] | |
| 3. | True- a,b False-c,d,e,f True- a,e False-b,c,d,f True- b,d False-a,c,e,f True- c,f False-a,b,d,e True-d False-a,b,c,e,f True-b,c False-a,d,e,f Identify correct groups of machine learning problems: | 01 |
| | Given a geographical data, label areas that are hit by flood or earthquake. Given a set of post found on a social networking site, group users into set with similar sentiments. Given database of satellite images, automatically discover places segments where air pollution index is similar. Recognizing handwritten letter and converting into e-mail. [LJIET] | |



L.J. Institute of Engineering & Technology Semester: VII (2021) 1,4-supervised learning, 2,3,-unsupervised 1-classification 2-Regression, 3-4 Association rule 1,2- supervised learning 3,4- unsupervised learning 1,2- unsupervised learning 3,4-supervised learning 1,3-supervised learning 2,4-unsupervised learning 1,2,3-supervised learning 4-unsupervised learning from experience E with respect to some 01 A computer program is said to of task T and performance measure P, if it's measured by performance P improves with . [LJIET] learn, class, performance, task T, experience E. learn, performance, class, task T, experience E learn, class, performance, experience E, task T performance, class, learn, task T, experience E performance, class, learn, experience E, task T performance, learn, class, experience E, task T The choice of model used for solving specific problem is? [LJIET] 01 Human task Machine learning task Abstraction Generalization task Decision tree Mathematical equation Sr. **DESCRIPTIVE QUESTIONS** Marks No 1. Explain how Human learning happens.[LJIET] 07 07 2. Compare different types of Machine Learning. [LJIET] What are the different types of Supervised Learning approaches? Explain by giving 07 examples. [LJIET] What are the different types of Unsupervised Learning approaches? Explain by giving 07 examples. [LJIET] What is generalization? What role does it play in Machine Learning? [LJIET] 07 **TOPIC:2: Well Posed Learning Problem, Applications of Machine** Learning, Issues in Machine Learning **DESCRIPTIVE QUESTIONS** Marks Sr. No State and describe in details any two application areas of Machine Learning. [LJIET] 07 1. 2. Which are the problems that cannot be solved using Machine Learning. [LJIET] 03 UNIT NO- 2: **Preparing to Model TOPIC:1:** Basic Data Types, Exploring numerical data, Exploring categorical data, Exploring relationships between variables SHORT QUESTIONS/MCQ Sr. **Marks** No



L.J. Institute of Engineering & Technology Semester: VII (2021) Divide data into two parts: training and testing b. Find potential issues in data Understand nature and quality of data c. d. Do remediation Train model based on training data Arrange given preprocessing activities in correct order for supervised learning. [LJIET] c, b, d, a, e a, c, d, b, e c, d, b, e, a c, d, b, a, e a, c, b, d, e a, b, c, d, e a. Discrete values can assume finite or count ably infinite number of values 01 b. Pin code is example of Nominal attribute Nominal attribute may or may not have finite number of values. c. Numerical attributes can have count ably infinite values d. Nominal and Ordinal attributes are in general discrete. Which statement is/are false? [LJIET] only a both a,c both b.e all three c, d, e all three a, c, d none 3. is likely to get shifted drastically due to presence of 01 [LJIET] Mean, Outlier Median, Outlier Mean, Variance Mean, Median Variance, Outlier Median, Variance 1. Mechanism for one shot view and Whisker 01 4. understand nature of data 2. Box span from first to third Box Plot b. quartile 3. Can range up to 1.5 times IQR Median from bottom/top of box Denoted by line or band in box Inter quartile range [LJIET] 1-b, 2-d, 3-a, 4-c 1-d, 2-b, 3-a, 4-c 1-b, 2-d, 3-c, 4-a 1-d, 2-b, 3-c, 4-a 1-c, 2-a, 3-d, 4-b 1-c, 2-b, 3-d, 4-a 5. Requires at-least one independent variable 1. 01 2. Requires at-least one dependent variable



L.J. Institute of Engineering & Technology Semester: VII (2021) 3. Determines future outcome of data 4. Determines behavior of data [LJIET] Supervised learning: 1,2,3 Unsupervised learning: 1,4 Supervised learning: 1,3 Unsupervised learning: 2,4 Supervised learning: 2,3 Unsupervised learning: 1,4 Supervised learning: 1,2 Unsupervised learning: 3,4 Supervised learning: 1,3,4 Unsupervised learning: 2 Supervised learning: 2 Unsupervised learning: 1,3,4 What is first and third quartile of data. 01 **44**, **35**, **52**, **35**, **31**, **37**, **49**, **39**, **28 [LJIET]** 33, 46.5 35,49 31,44 31,49 35,44 32.5,45 To explore relationship between variables what can be used? 01 Box plot 2. Histogram 3. Scatter plot 4. Cross tab 5. PCA [LJIET] 3,4 1,2 1,2,3,4 5 1,5 1,2,4 reflects total count of data elements whose value falls The height of _ 01 . [LJIET] specifically _ Bar, bin Bin, bar Histogram, mode Mode, histogram Bar, mode Bin, mode 1. Discrete values can assume only finite values 01 2. Employee ID is example of Nominal attribute 3. Nominal attribute may or may not have finite number of values. 4. Numerical attributes can have count ably infinite values 5. Nominal and Ordinal attributes are in general discrete. Which statement is/are false? [LJIET] only a both a.c both b,e all three c, d, e all three a, c, d none

L.J. Institute of Engineering & Technology Semester: VII (2021) 10 Numerical attributes having lesser possible number of values can be treated as [LJIET] **Categorical** Nominal Ordinal Discrete Interval Ratio 11 1. Larger the value of variance indicates more dispersion in data. 01 2. Deviation between mean and median is significant high, means chances of outliers is less. 3. Larger difference between two quartile values indicates more data spread in respective quarter. Identify false statements [LJIET] All 1,2,3 Both 1.2 Both 2,3 Only 1 Only 2 Only 3 12 What is first and third quartile of data. 01 44, 12, 25, 71, 27, 59, 59, 38, 66 [LJIET] 26,62.5 27,59 27,44 38,59 25.5,68.5 44,59 13 data, if have possibly only two values/ labels is called 'dichotomous'. 01 [LJIET] **Nominal** Interval Ordinal Ratio Discrete **Binary** DESCRIPTIVE QUESTIONS Marks Sr. No What are the main activities involved when you are preparing to start with modeling in 07 Machine Learning? [LJIET] What are the basic data types in Machine Learning? Explain by giving examples of each 07 one of them. [LJIET] Explain in details the different components of Box Plot. State how outliers can be detected 07 using Box Plots. [LJIET]



L.J. Institute of Engineering & Technology Semester: VII (2021) State and explain the two ways in which we can explore the relationship between two variables (attributes). [LJIET] **TOPIC:2: Data Issues and Remediation, Data Pre-processing** SHORT QUESTIONS/MCQ Marks Sr. No Issues of data quality [LJIET] 01 Missing values Outliers Missing values and outliers Error in data collection Error in sample selection Error in data collection and sample selection Techniques for dimensionality reduction PCA stands for [LJIET] 01 Principle Component Analysis Primary Coefficient Analysis Principle Coefficient Analysis Primary Component Analysis Partial Component Aggregation Partial Component Aggregation Dimensionality reduction helps in removing features. [LJIET] 01 and Redundant and ir-relevant Redundant and relevant Noisy and redundant Relevant and rich Noisy and ir-relevant Rich and noisy 4. removing columns which have too many missing values 1. 01 2. removing columns which have high variance in data removing columns with dissimilar data trends which of given statement(s) is/ are correct: [LJIET] Only 1 Only 2 Only 3 Both 1,3 Both 2,3 None of given **DESCRIPTIVE OUESTIONS** Sr. Marks No 1. Explain in details the various ways to address the missing values in a dataset. [LJIET] 07 Define Outliers. Can they be handled in a dataset? If yes, how? [LJIET] 05 07 What are the different techniques for data pre-processing. Explain in brief, dimensionality reduction and feature selection.[LJIET] **UNIT NO- 3: Modelling and Evaluation**



| | TOPIC:1: Selecting a model, Training a model, Model representation | |
|-----------|---|-------|
| | and interpretability | |
| Sr. No | SHORT QUESTIONS/MCQ | Marks |
| 1. | A machine learning problem that does not include target variable is called and that include target variable is called [LJIET] | 01 |
| | descriptive learning, predictive learning predictive learning, descriptive learning | |
| | descriptive learning, reinforcement learning | |
| | predictive learning, reinforcement learning | |
| | reinforcement learning, predictive learning | |
| | reinforcement learning, descriptive learning | |
| 2. | Summarized knowledge representation of raw data is called as? [LJIET] | 01 |
| | Model Abstraction Generalization Training Testing | |
| | Validation | |
| 3. | Full Form of – SRSWS ? [LJIET] | 01 |
| | | |
| | Simple Random Sampling With Replacement | |
| | Sampling Random Simple Without Replacement | |
| | Simple Random Sampling Without Replacement | - |
| | Simple Repeated Sampling With Replacement | |
| | Simple Repeated Sampling Without Replacement | |
| | | |
| 4 | Sampling Random Simple With Replacement | 0.1 |
| 4. | 1 method divides data into 3partitions namely training data, validation | 01 |
| | data, test data 2 is part of input data with known labels | |
| | 3 is used to measure performance of a model | |
| 70. | 4. is used in iteration to refine a model | |
| | 5 is used only once on finalize models to report final performance of a | |
| | model. [LJIET] | |
| | model. [Lott 1] | |
| | Holdout, Test data, Validation Data, Validation Data, Test Data | |
| | Holdout, Test data, Validation Data, Training Data, Test Data | |
| | K-fold CV, Test data, Validation Data, Validation Data, Test Data | |
| | K-fold CV, Test data, Validation Data, Training Data, Test Data | |
| | 3-fold CV, Test data, Validation Data, Validation Data, Test Data | |
| | 3-fold CV, Test data, Validation Data, Training Data, Test Data | |
| 5. | To measure the extent to which model is going wrong in estimating the relationship between Independent Variable and Dependent Variable is called as? [LJIET] Cost function | 01 |
| | Loss function | |
| | Objective function | |
| | No free lunch | |
| | | |
| | Target Function | |
| | AUC | |



L.J. Institute of Engineering & Technology Semester: VII (2021) randomly picks data instance from input data set, with possibility of same data instance being picked up multiple times. [LJIET] **Bootstrapping LOOCV** 10-fold CV k-fold CV Holdout **Stratified Random Sampling** Less than 0.5 – Worse than random prediction 01 0.5-0.7 weak predictive ability 0.7-0.9 good to fair predictive ability 0.9-1.0 excellent predictive ability Given value indications are used with which measure? [LJIET] **AUC ROC** Residual F-measure Kappa-coefficient purity Sr. **DESCRIPTIVE QUESTIONS** Marks No Compare and contrast Descriptive and predictive Models. [LJIET] 05 1. 2. Explain in details about the methods to train a learning model. [LJIET] 07 3. What is underfitting and overfitting in the context of Machine Learning Model? [LJIET] 07 07 4. Write about the bias-variance tradeoffs in context of model fitting. [LJIET] **TOPIC:2:** Model Performance Evaluation, Performance Improvement SHORT QUESTIONS/MCQ Sr. Marks No Follows general principle of machine learning 01 1. a. Also called as non-parametric learning b. Doesn't need to refer back training data c. d. **Decision Tree** k-nearest neighbor [LJIET] Eager Learner—a,c,d Lazy Learners -b,e Eager Learner—a,d Lazy Learners—c,b,e Eager Learner-a,e Lazy Learners b,c,d Eager Learner-b,d Lazy Learners -a,c,e Eager Learner-b,e Lazy Learners -a,c,d Eager Learner-b,c,d Lazy Learners -a,e Model Accuracy [LJIET] 01 (TP+TN)/(TP+FP+FN+TN)(TP+FN)/(TP+FP+FN+TN)(FP+TN)/(TP+FP+FN+TN)(FP+FN)/(TP+FP+FN+TN)(TP+TN)/(TP+FP-FN+TN)

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| | (TP+TN)/(TP-FP+FN-TN) | 0.4 |
| 3. | Sensitivity [LJIET] | 01 |
| | TP | |
| | $\overline{TP + FP}$ | |
| | | |
| | TP + FN | |
| | FP TO THE PROPERTY OF THE PROP | |
| | TP + FP FP | No. |
| | $\frac{1}{TP+FN}$ | b dillion |
| | $\frac{TP}{TP} + TN$ | |
| | TP | - |
| | <u>TN</u> | |
| 4. | TN + FP | 01 |
| 4. | Specificity [LJIET] TP | U1 |
| | $\overline{TP + FP}$ | W |
| | TP | V 74 |
| | $\frac{TP}{TP + FN}$ | |
| | | |
| | <u>FP</u> | |
| | $\overline{TP + FP}$ | |
| / | FP | |
| v | TP + FN | |
| ٠. | TP + TN | |
| , | TP | 0.000 |
| , | TN | |
| | $\frac{TN}{TN + FP}$ | |
| | | |
| 5. | Precision [LJIET] | 01 |
| Λ. | TP | |
| • | $\frac{TP + FP}{TP + FP}$ | |
| | | |
| | TP TP | |
| | TP + FN | |
| | <u>FP</u> | |
| | TP + FP | |
| | FP | |
| | $\overline{TP + FN}$ | |
| | TP + TN | |
| | $\frac{TT + TN}{TP}$ | |
| | | |
| | $\frac{TN}{TN}$ | |
| | 1 T.N. 1 E'D | |
| 6. | $\overline{TN + FP}$ is a part of model preparation activity. [LJIET] | 01 |



| Cross validation | | <u>Semester: VII</u> |
|---|---|----------------------|
| | | |
| Confusion mat | .TIX | |
| ROC curve | | |
| Hold-out | | |
| Ensemble | | |
| 7. Identify false s | statement(s) | 01 |
| 1. Mean i | s impacted if too many data elements are having value closer to the far | r end |
| of the 1 | | |
| | ess/Shape of histogram depends on nature of data. | |
| 3. Height | of bin keeps on decreasing as we move toward right, is called right-sl | kew. |
| | ate relationship can be visualized using scatter-plot and box-plot. | |
| 5. Cross-t | tab allows operations like roll-up and drill-down [LJIET] | |
| 4 only | | |
| 4,5 only | | |
| | | |
| None of given | | |
| 1,2,3 only | | |
| 1,4 only | | |
| All of given | | |
| 8. What is not co | rrect about Eager Learners from following | 0.3 |
| 1 Follow | rs the steps abstraction and generalization. | |
| | o refer back to training data. | |
| | rt Vector Machine | |
| 4. Very li | ttle time in training [LJIET] | |
| 4 1000 | | |
| 2,4 only | | |
| 1,2 only | | |
| 3,4 only | | |
| 1,3,4 only | | |
| All of given | | |
| None of give | | |
| 9. 1. Arises | from simplifying assumptions made by model to make target function | less 0 |
| comple | | |
| | from difference in training data set used to train the model | |
| | under-fitting a model | |
| | over-fitting a model erformance on complex dataset | |
| | in training data set magnified in model [LJIET] | |
| | | |
| There = 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 | ias- 1,3,5 Error due to Variance- 2,4,6 | |
| | ias- 1,4,6 Error due to Variance- 2,3,5 | |
| Error due to bi | 2 4 6 Europe des 4 Marie 2 2 5 | |
| Error due to bi | ias- 2,4,6 Error due to Variance- 1,3,5 | |
| Error due to bi Error due to bi Error due to bi | ias- 2,3,5 Error due to Variance- 1,4,6 | |
| Error due to bi Error due to bi Error due to bi | | |
| Error due to bi Error due to bi Error due to bi Error due to bi | ias- 2,3,5 Error due to Variance- 1,4,6 | |

01

Input variable Output variable

Feature construction

Variable

1.

L.J. Institute of Engineering & Technology Semester: VII (2021) 2. Feature extraction 3. Feature subset selection Discovering missing information about relationship of features a. b. Creating new set of features from original features using functional mapping Deriving most meaningful features from given feature space. [LJIET] c. 1-a, 2-b, 3-c 1-c, 2-a, 3-b 1-a, 2-c, 3-b 1-b, 2-a, 3-c 1-c, 2-b, 3-a 1-b, 2-c, 3-a Text data converted into numerical representation using process called as 01 In this process occurrences of a word belonging to are consolidated in form of ______. **[LJIET]** Vectorization, corpus, bag-of-word Vectorization, bag-of-word, corpus Corpus, bag-of-word, vector Corpus, tokens, normalization Vectorization, token, bag-of-word Corpus, vector, token 4. 1. New set of features are extracted from original set of features which are similar in 01 2. n-dimensional feature space is transformed to m-dimensional space. Where n less than m. 3. New features transformed are orthogonal to each other 4. Sum of variance of new feature is equal to sum of variance of old feature In context of PCA which of following is false [LJIET] 1,2 1,2,4 All of given None of given If P1,P2 and P3 are three principle components 01 What are true among following: P1, P2 and P3 are parallel 1. 2. Variance of P1 is largest 3. Variance of P3 is smallest 4. P1, P2 and P3 are orthogonal 5. Variance of P1 is smallest 6. Variance of P3 is largest [LJIET] 2,3,4 1,2,3 4,5,6 1,5,6 1,2 only 4,5 only

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| 6. | a. PCA | 01 |
| | b. SVD | |
| | c. LDA | |
| | 1. Matrix factorization technique | |
| | 2. Identify dimensions that are independent of each other | |
| | 3. Eigen values and eigen vectors are used | |
| | 4. Singular, left singular and right singular matrix is produced | |
| | 5. Captures variability in data set | |
| | 6. Inter-class and intra-class matrices are used [LJIET] | |
| | | |
| | a-2,3 b-1,4 c-5,6 | |
| | a-1,2 b-5,6 c-3,4 | |
| | a-3,4 b-5,6 c-1,2 | |
| | a-4,5 b-3,6 c-1,2 | |
| | a-2,4 b-1,5 c-3,6 | |
| | a-1,3 b-2,5 c-4,6 | |
| 7. | a. Average information in a variable is | 01 |
| | b. Indicator of feature relevance | |
| | c. Features are ranked in descending order of [LJIET] | |
| | | |
| | Entropy, Mutual Information, Information Gain | |
| | Mutual Information, entropy, Information gain | |
| | Information gain, entropy, Mutual Information | |
| | Entropy, Information Gain, Mutual Information | |
| | Mutual Information, Information gain, entropy | |
| | Information gain, Mutual Information, entropy | |
| 8. | Find the value of the person's correlation coefficient from the following data: | 01 |
| | F1= (43,21,25,42,57,59) | |
| | F2= (99,65,79,75,87,81) [LJIET] | |
| | | |
| | 0.5298 | |
| | 0.9852 | |
| | 0.2599 | |
| 7 | 0.9985 | |
| | -0.5298 | |
| 0 | -0.9852 $n_{11}+n_{00}$ | 0.1 |
| 9. | a. $\frac{n_{11} + n_{00}}{n_{01} + n_{10} + n_{11} + n_{00}}$ | 01 |
| | b. Widely used in text classification | |
| | n_{11} | |
| | $n_{01} + n_{10} + n_{11}$ | |
| | d. $\sqrt[r]{\sum_{i=1}^{n} (F1_i - F2_i)^r}$ [LJIET] | |
| | | |
| | SMC-a, Cosine-b, Jaccard-c, Minkowski-d | |
| | SMC-b, Cosine-a, Jaccard-c, Minkowski-d | |
| | SMC-b, Cosine-c, Jaccard-a, Minkowski-d | |
| | SMC-a, Cosine-d, Jaccard-c, Minkowski-b | |
| | SMC-c, Cosine-b, Jaccard-a, Minkowski-d | |
| 10 | SMC-c, Cosine-d, Jaccard-a, Minkowski-b | 01 |
| 10 | a. Infeature space is expanded.b and are two variants of [LJIET] | 01 |
| | o and are two variants of [LJIE1] | |
| | a-feature construction b-feature construction, feature extraction, feature transformation | ation |



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|------------|---|-----------|----------|
| a-fe | eature transformation b-feature construction, feature extraction, feature subset | | |
| sele | ection | | |
| a-fe | eature transformation b-feature construction, feature extraction, feature transform | nation | |
| | eature construction b-feature selection, feature extraction, feature transformation | | |
| | eature construction b-feature construction, feature selection, feature transformation | | |
| | | IOII | |
| a-te | eature construction b-feature construction, feature extraction, feature selection | | |
| | | | |
| 11 | a Toyt specific feeture construction | | 01 |
| 11 | a. Text specific feature constructionb. Inter-class and intra-class scatter matrix | | UI |
| | c. Eigen values and Eigen-vectors | | |
| | | | |
| | d. Singular value matrix 1- LDA | | |
| | 2- PCA | | |
| | 3- SVD | | |
| | | | |
| 2.4 | 4- Vectorization [LJIET] | | |
| | l, b-1, c-2, d-3 | | |
| | l, b- <mark>2,</mark> c-3, d-1 | | |
| a-4 | , b- <mark>2,</mark> c-1, d-3 | | |
| a-4 | , b- <mark>1,</mark> c-3, d-2 | | |
| a-4 | l, b- <mark>3,</mark> c-2, d-1 | | |
| a-4 | l, b-3, c-1, d-2 | | |
| 12 a. 2 | | | 01 |
| 12 a. Z | $\mathcal{L}_{i=1}^{S_i}$ | | O1 |
| b S | $\sum_{x \in D_i}^{n} (x - m_i)(x - m_i)^T$ | | |
| 0. 2 | $\Delta x \in D_i(x)$ $m_i(x)$ | | |
| cΣ | $\sum_{i=1}^{c} N_i (m_i - m) (m_i - m)^T$ | | |
| 0.2 | | 7 | |
| fro | m above given equations S_W , S_i and S_B relatively are? [LJIET] | | |
| a, t | | | |
| a, c | | | |
| b, a | | | |
| | | | |
| b, c | | | |
| c, a | | | |
| c, t | | | |
| 13 Euc | clidean distance and Manhattan distance respectively for given data is? | | 01 |
| F1= | = (2,3,6,7,8,6,6,7) | | |
| F2= | = (6,5.5,4,2.5,3,5.5,7,6) [LJIET] | | |
| | i, 20.5 | | |
| | 5, 8.6 | | |
| | 8, 8.6 | | |
| | | | |
| | .5, 8.8 | | |
| | 2, 5.5 | | |
| 5.5 | 5, 8.6 | | |
| 14 | used more frequently to calculate the distance between binary vector | s is | 01 |
| Ha | mming distance. [LJIET] | | |
| | | | |
| Mo | anhattan | | |
| | nkowski | | |
| IVII | IINU W SN I | | |



L.J. Institute of Engineering & Technology Semester: VII (2021) Pearson Euclidean Jaccard Cosine 15 a. Filter approach 01 b. Wrapper approach c. Hybrid approach d. Embedded approach 1. Induction algorithm 2. Statistical tests 3. Induction algorithm and statistical tests Simultaneous selection and classification [LJIET] a-2 b-1 c-3 d-4 a-1 b-2 c-3 d-4 a-1 b-2 c-4 d-3 a-2 b-1 c-4 d-3 a-1 b-3 c-2 d-4 a-2 b-3 c-1 d-4 16 What is true about feature selection process 01 1. No new feature is generated 2. Improving efficiency of a learning model 3. Better understanding of underlying model 4. Faster and cost effective model **Used functional mapping [LJIET]** 1,2,3,4 only 1,3,4,5 only 1,2,5 only All of given None of given 1,2 only What is true if groups are well separated in LDA 01 1. Intra-group mean are far away from each other 2. Data points are close to intra-group mean 3. Intra-group mean from grand-mean is large. [LJIET] All of given 1,2 only 2,3 only 1,3 only None of given 1 only 18 What features are ALWAYS candidate for rejection in the process of feature subset selection 01 Redundant features Irrelevant features Weakly relevant features d. Non-redundant features[LJIET] a,b only a,b,c only all of given



L.J. Institute of Engineering & Technology Semester: VII (2021) c only b,c only b,c,d only 19 Which is not a search strategy[**LJIET**] 01 Sequential forward selection Sequential backward selection Bi-directional selection Bi-directional elimination Recursive elimination Recursive Selection **DESCRIPTIVE QUESTIONS** Sr. Marks No 1. What is feature selection? Why is it needed? What are the different approaches for 07 feature selection? [LJIET] 2. Explain the process of feature engineering in context of a text categorization problem. 05 [LJIET] Differentiate between SMC and Jaccard Coefficients. [LJIET] 05 4. Explain in short the three methods for Feature Extraction. [LJIET] 07 5. State and explain the methods to find out the similarity or redundancy aspect of the 07 attributes in a dataset. [LJIET] What is a feature? What is feature Engineering? What are the major elements of Feature 05 Engineering? Explain them. [LJIET] UNIT NO- 5: **Brief overview of Probability TOPIC:1:** Basic concept of Probability, Random Variables, Discrete Distributions, Continuous Distribution, Central Theorem, Monte Carlo Approximations. **DESCRIPTIVE OUESTIONS** Sr. Marks No 1. Define the Bayesian interpretation of probability. [LJIET] 05 2. Define probability of a union of two events with equation. [LJIET] 05 3. What is joint probability? Explain it using appropriate equations. [LJIET] 05 4. What is the chain rule of probability? [LJIET] 05 5. What do you mean by Conditional Probability [LJIET] 05 6. Define Covariance and Correlation [LJIET] 05 What is hypothesis? Give examples [LJIET] 05 **UNIT NO-6: Bayesian Concept Learning** TOPIC:1: Impotence of Bayesian methods, Bayesian theorem, Bayes' theorem and concept learning, Bayesian Belief Network SHORT OUESTIONS/MCO Sr. Marks No In a bolt factory, machines A1, A2 and A3 manufactures respectively 25%, 35% and 01 40% of the total machines. Of these 5%, 4% and 2% are defective bolts. A bolt is drawn at random from the product and is found to be defective. What is probability that it was manufacture by machine A2? [LJIET]

Semester: VII (2021) DESCRIPTIVE OUESTIONS Sr. Marks No 1. Explain following with example: 07 a. Prior probability b. Posterior probability c. Likelihood [LJIET] Explain Naïve Bayes classifier with example and its use in practical life. [LJIET] 07 What are Bayesian Belief Networks? Where are they used? Can they solve all types of 07 problem?[LJIET] How Bayes theorem support the concept learning principle? [LJIET] 07 07 In an airport security checking system, the passenger are checked to find out any intruder. Let I indicate whether somebody is an intruder (I) or not (1) and A indicate alarm whether raised (A=1) or not (A). Alarm will be raised if an intruder is identified with probability P(A|I)=0.98. Alarm will be raised if non-intruder with probability P(A|I)=0.001, which implies error factor. The probability of intruder in population of passenger is P(I) = 0.00001. What is probability that alarm is raised when a person actually is an intruder? [LJIET] For preparation of the exam questions are of types A, B or C. Probability of A, B or C 07 appearing in the exam are 30%, 20% and 50% respectively. During preparation student solved 9 of 10 problems type A, 2 of 10 problems type B, and 6 of 10 problems of type C. What is probability that the student will solve the problem will be of type A? [LJIET] A tumor can be malignant or non-malignant. Only 0.5% of population has malignant 07 tumor. Laboratory report detects malignancy present with 98% accuracy. Also laboratory report shows malignancy not present correctly in 97% of cases. If a Lab report is showing malignancy is present, should we declare this as malignant tumor or not. [LJIET] What is optimal Bayes classifier? Explain. [LJIET] 07 **UNIT NO- 7: Supervised Learning: Classification and Regression: TOPIC:1:** Supervised Learning, Classification Model, Learning steps, Classification algorithms DESCRIPTIVE QUESTIONS Sr. **Marks** No What is supervised learning? Why it is called so? Give three example of supervised 07 learning.[LJIET] Explain in brief, the SVM model. Give its advantage and disadvantage. [LJIET] 07 Write note on: Validation error in kNN, error rate in kNN choosing value of k in kNN, 07 inductive bias in decision tree.[LJIET] Explain in brief: Decision tree algorithm.[LJIET] 07 4. Explain in brief: Random forest model. [LJIET] 5. 07 Explain following w.r.t decision trees: node and leaf, entropy, information gain[LJIET] 07 Discuss SVM model in detail with different scenarios. [LJIET] 07

L.J. Institute of Engineering & Technology Semester: VII (2021) Discuss how to calculate distance between the test data and training data in kNN.[LJIET] 9. Discuss OOB error and variable importance in random forest. [LJIET] 07 10 How to avoid over fitting decision tree? Explain in detail.[LJIET] 07 **TOPIC:2:** Regression, Regression algorithms DESCRIPTIVE OUESTIONS Marks Sr. No 07 What is simple linear regression? What is dependent and independent variable in linear 1. regression?[LJIET] What are condition of a positive slope and conditions of negative slope in linear 07 regression?[LJIET] 3. What is multiple linear regression? Give example.[LJIET] 07 4. What is multicollinearity in regression equation? Explain. [LJIET] 07 5. Explain rise, run, and slope in graph. Find slope if points on line are (-3,-2) and (2,2) 07 respectively.[LJIET] How to improve accuracy of linear regression model?[LJIET] 07 Explain OLS algorithm with steps. [LJIET] 07 Define sum of squares due to error in multiple linear regression.[LJIET] 07 UNIT NO-8: **Unsupervised Learning TOPIC:1:** Supervised vs. Unsupervised Learning, Applications, **Association rules** DESCRIPTIVE OUESTIONS Sr. Marks No How unsupervised learning is different from supervised leaning? Explain with 07 example.[LJIET] Mention few application areas of unsupervised learning.[LJIET] 07 How apriori algorithm helps in reducing the calculation overhead for market basket 07 analysis? Give example. [LJIET] 07 What is Association rule? What are the Applications of Association rule mining? Define support and confidence in Association rule mining.[LJIET] 07 A database has 4 transactions, shown below. TID Date items bought T100 10/15/04 {K, A, D, B} T200 10/15/04 {D, A, C, E, B} T300 | 10/19/04 | {C, A, B, E} T400 10/22/04 {B, A, D} Assuming a minimum level of support min_sup = 60% and a minimum level of confidence min conf = 80%. a. Find all frequent item-sets using the Apriori algorithm. b. List all of the strong association rules, along with their support and confidence values for buys(item1, item2) ⇒ buys(item3) items can be A, B etc.[LJIET]

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6 Explain how Market Basket Analysis uses the concept of association analysis [L.HET]

07

| 6. | Explain how Market Basket Analysis uses the concept of association analysis.[LJIET] | 07 |
|----------------------|--|----------------------|
| 7. | Explain apriori algorithm for association rule learning with an example. [LJIET] | 07 |
| | TOPIC:2: Clustering | |
| Sr. No | SHORT QUESTIONS/MCQ | Marks |
| 1. | In which situation k-meas clustering fails to give good results?[LJIET] | 01 |
| Sr. No | DESCRIPTIVE QUESTIONS | Marks |
| 1. | Describe how quality of clusters are measured in k-means algorithm?[LJIET] | 07 |
| 2. | What are three broad categories of clustering technique? Explain characteristics of each briefly.[LJIET] | 07 |
| 3. | Describe main difference in approach of k-means and k-medoids algorithm with a neat diagram.[LJIET] | 07 |
| 4. | Explain k-means method with step-by-step algorithm.[LJIET] | 07 |
| 5. | Given a set of one-dimensional data point: {5,10,15,20,25,30,35}. Assume that k=2 and first set of random centroid is selected as {15,32} and then it is refined with {12,30} a. Create two clusters with each set of centroid mentioned above following the k-means approach. b. Calculate SSE for each set of centroid.[LJIET] | 07 |
| 6. | How to re-compute cluster centroids in k-means algorithm?[LJIET] | 07 |
| 7. | Explain concept of clustering with neat diagram.[LJIET] | 07 |
| ٠, | UNIT NO- 9 : | $\overline{}$ |
| | Neural Network: | |
| • | TOPIC:1: Introduction to neural network, Biological and Artificial Neurons, Types of Activation functions, Implementation of ANN, | |
| | Architecture, Leaning process, Backpropogation, Deep Learning | |
| Sr. | Til ciniccial c, Ecalling process, Backpropogation, Beep Ecarining | |
| | DESCRIPTIVE QUESTIONS | Marks |
| No 1. | DESCRIPTIVE QUESTIONS What is function of summation junction of a neuron? What is threshold activation | Marks |
| No | What is function of summation junction of a neuron? What is threshold activation function?[LJIET] What is step function? What is the difference of step function with threshold | |
| No 1. | What is function of summation junction of a neuron? What is threshold activation function?[LJIET] | 07 |
| No 1. 2. | What is function of summation junction of a neuron? What is threshold activation function?[LJIET] What is step function? What is the difference of step function with threshold function?[LJIET] | 07 |
| No 1. 2. 3. | What is function of summation junction of a neuron? What is threshold activation function?[LJIET] What is step function? What is the difference of step function with threshold function?[LJIET] Explain McCulloch-Pitts model of neuron.[LJIET] Explain the basic structure of multi-layer perceptron. Explain how it can solve XOR | 07 07 07 |
| 1. 2. 3. 4. | What is function of summation junction of a neuron? What is threshold activation function?[LJIET] What is step function? What is the difference of step function with threshold function?[LJIET] Explain McCulloch-Pitts model of neuron.[LJIET] Explain the basic structure of multi-layer perceptron. Explain how it can solve XOR problem.[LJIET] What is Artificial Neural Network (ANN)? Explain some of salient highlights in different | 07 07 07 07 |
| No 1. 2. 3. 4. | What is function of summation junction of a neuron? What is threshold activation function?[LJIET] What is step function? What is the difference of step function with threshold function?[LJIET] Explain McCulloch-Pitts model of neuron.[LJIET] Explain the basic structure of multi-layer perceptron. Explain how it can solve XOR problem.[LJIET] What is Artificial Neural Network (ANN)? Explain some of salient highlights in different architectural option for ANN.[LJIET] Explain learning process of ANN. With example explain challenges in assigning synaptic | 07 07 07 07 |



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9. Write Short note on: Deep learning, Multi-layer perceptron, Artificial neuron. [LJIET]

