



MUTHAYAMMAL ENGINEERING COLLEGE

(An Autonomous Institution)

(Approved by AICTE, New Delhi, Accredited by NAAC & Affiliated to Anna University)
Rasipuram - 637 408, Namakkal Dist., Tamil Nadu.



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| | MUST KNOW CONCEPTS | | MKC |
| DEPT - AI&DS | | | 2021-22 |
| Course Code & Course Name | : | 19ADCO5 / Introduction to Data Science | |
| Year/Sem/Sec | : | II / III | |

| S.No. | Term | Notation (Symbol) | Concept / Definition / Meaning / Units / Equation / Expression | Units |
|------------------------------|---------------------|----------------------|--|-------|
| Unit-I : Introduction | | | | |
| 1. | Data science | | Data science involves gaining the knowledge from gathered data using different methods. | |
| 2. | Data scientist | | As a data scientist, you take a complex problem, research it, gather as a data, and we use to solve the problem. | |
| 3. | Data Acquisition | | It is a process of analysing the real world physical condition and converts into numerical values, which can be manipulated by computer. | |
| 4. | Data preparation | | Data preparation is a act of manipulating raw data into a form that can readily and accurately be analysed. | |
| 5. | Data cleaning | | Data cleaning is a process of identifying and correcting corrupt, incorrect and irrelevant data from reference set or table. | |
| 6. | Data transformation | | Data transformation is a process of converting data from one format or structure into another format or structure. | |
| 7. | Handling outliers | | Outliers are often used for the fraud detection and finding the malicious activities which happens on the field. | |
| 8. | Data integration | | In this, the data scientist ensures the data is accurate and reliable. | |
| 9. | Data reduction | | Data reduction is the transformation of numerical or alphabetical digital information derived experimentally into a corrected and simplified form. | |
| 10. | Data mining | | It is the process of extracting the required information from the larger set of raw data. | |
| 11. | Model building | | In this the process involves setting up ways of collecting data and finding a statistical, mathematical or a stimulation model to gain | |

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| | | | understanding and make predications. | |
| 12. | clustering | | It is a task of dividing the population or data points into number of groups. It is basically a collection of objects on the basis of similarity and dissimilarity between them. | |
| 13. | Essential data science skills | | <ol style="list-style-type: none"> 1. statistical analysis 2. machine learning 3. computer science and programming 4. data storytelling 5. business intuition 6. analytical thinking 7. critical thinking 8. inquistiveness 9. interpersonal skills | |
| 14. | Statistical analysis | | Identify patterns in data. This includes having a keen sense of pattern detection and normally detection. | |
| 15. | Machine learning | | Implement algorithms and statistical models to enable a computer automatically learn from data. | |
| 16. | Computer science and programming | | Applying the principle of AI, database system and software engineering. known to Write the programs like java, python and SQL programming languages. | |
| 17. | Data storytelling | | Data storytelling is the practice of building a narrative around a set of data and its accompanying visualizations. | |
| 18. | Business intuition | | Connect stakeholders to gain a full understanding of the problem they are looking to solve. | |
| 19. | Analytical thinking | | Find analytical solutions to abstract business issues. | |
| 20. | Critical thinking | | Apply objective analysis of facts before coming to a conclusion. | |
| 21. | Inquistiveness | | Look beyond what's on the surface to discover patterns and solutions within the data | |
| 22. | Interpersonal skills | | Communicate across a diverse audience across all levels of an organization. | |
| 23. | Fundamental steps to complete a data analytics project | | Step 1: understand the business Step 2: get your data Step 3: explore and clean your data Step 4: enrich your dataset Step 5: build helpful visualizations Step 6: get predictive Step 7: iterate | |
| 24. | Applications for data science | | <ol style="list-style-type: none"> 1. fraud and risk detection 2. healthcare 3. gaming 4. E-commerce 5. banking 6. transport | |

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| | | | 7. education | |
| 25. | Jobs for data science | | 1. data scientists 2. data analyst 3. data engineering 4. business intelligence specialists 5. data architects | |
| Unit-II : Data Collection and Data Pre-Processing | | | | |
| 26. | Data collection | | Data collection is the process of accumulating data that's required to solve a problem statements. | |
| 27. | Steps to collect the data | | 1. identify a problem statement 2. determine what data type is needed 3. decide on data sources 4. create a timeline 5. collect your data | |
| 28. | Data pre-processing | | Data preparation plays an important role in your workflow. You need to transform the data in a way that a computer would be able to work with it. | |
| 29. | Steps in data pre-processing | | 1. data cleaning <ul style="list-style-type: none"> • Missing data • Noisy data 2. data transformation <ul style="list-style-type: none"> • Normalization • Attribute selection • Discretization • Concept hierarchy generation 3. data reduction <ul style="list-style-type: none"> • Data cube aggregation • Attribute subset selection • Numerosity reduction • Dimensionality reduction | |
| 30. | Missing data | | You may also notice that some important values are missing. These problems arise due to human factor, program errors and other reasons. | |
| 31. | Noisy data | | A large amount of additional meaningless data is called noisy data. | |
| 32. | Normalization | | Normalization is a technique often applied as part preparation for machine learning. It is used while the features have different ranges. | |
| 33. | Attribute selection | | If you construct a new features combining the given features in order make the data mining process more efficient, it is called as attribute selection. | |
| 34. | Discretization | | Data discretization refers to a method of converting continuous data into discrete buckets by grouping it. | |
| 35. | Concept hierarchy | | Concept hierarchy generation based on the | |

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| | generation | | number of distinct values per attribute. | |
| 36. | Aggregation | | In the case of data aggregation, the data is pooled together and presented in a unified format for data analysis. | |
| 37. | Numerosity reduction | | Numerosity reduction is a method of data reduction that replaces the original data by a smaller form of data representation. | |
| 38. | Types of numerosity reduction | | 1. parametric 2. non- parametric | |
| 39. | Dimensionality reduction | | Dimensionality reduction is the transformation of data from a high-dimensional space into low-dimensional space. | |
| 40. | Data cleansing | | Data cleaning is a process of identifying and correcting corrupt, incorrect and irrelevant data from reference set or table. | |
| 41. | Steps involved for data cleansing | | 1. removal of unwanted observations 2. fixing structural errors 3. managing unwanted outliers 4. handling missing data | |
| 42. | Tools for data cleansing | | <ul style="list-style-type: none"> • Openrefine • Trifacta wrangler • TIBCO clarity • Cloudingo • IBM infosphere quality stage | |
| 43. | Components of data integration | | <ul style="list-style-type: none"> ➤ Data migration ➤ Enterprise application integration (EAI) ➤ Master data management ➤ Data aggregation | |
| 44. | Types of data aggregation | | <ul style="list-style-type: none"> ❖ data federation ❖ data warehousing | |
| 45. | Data federation | | Data is combined into virtual database. | |
| 46. | Data warehousing | | Data is combined into a physical database. | |
| 47. | Advantages of data warehousing | | 1. improved business intelligence 2. rapid access to data 3. historical intelligence | |
| 48. | Disadvantages of data warehousing | | 1. cost of scaling 2. maintenance cost | |
| 49. | Challenges associated with MDM strategy | | 1. complexity 2. overlap 3. governance 4. standards | |
| 50. | Categories of data integration | | 1. analytical data integration (AnDI) 2. operational data integration (OnDI) 3. hybrid data integration (HyDI) | |
| Unit III- Exploratory Data Analytics | | | | |
| 51. | Descriptive statistics | | A population is the group to be studied, and population data is a collection of all in the | |

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| | | | population. | |
| 52. | Descriptive measures | | Descriptive measures of population are called parameters and typically using greek letters. The population mean is μ (mu). | μ |
| 53. | Mean | | The arithmetic mean of a variable, often called as average, is computed by adding up all the values and dividing by the total numbers of values. | |
| 54. | Medium | | The median of a variable is the middle of the data set when the data are sorted in order from least to greatest. | |
| 55. | Mode | | The mode is the value that appears frequently in the data set. | |
| 56. | Range | | The range is the difference between the highest and lowest values in a set of numbers. | |
| 57. | Variance | | The variance is the average of the squared differences from the mean. | |
| 58. | Standard deviation | | In statistics, the standard deviation is a measure of the amount of variance or dispersion of set of values. | |
| 59. | Central limit theorem | | In this theorem, the regardless of the shape of our population, the sampling distribution of the sample mean will be normal as the sample size increases. | |
| 60. | Coefficient of variation | | The coefficient of variation (CV) is a measure of relative variability. It is the ratio of the standard deviation to the mean. | |
| 61. | Variability | | Variability refers to how spread out; that is, it refers to the amount of spread of the scores around the mean. | |
| 62. | Graphical representation | | A graph is defined as a chart with statistical data, which represented in the form of curves or lines drawn across the coordinate point plotted on the surface. | |
| 63. | Types of graphical representation | | <ol style="list-style-type: none"> 1. line graphs 2. bar graphs 3. histograms 4. line plots 4. frequency table etc..... | |
| 64. | Advantages of graphical representation | | <ul style="list-style-type: none"> • It makes data more easily understandable • It saves time • It makes the comparison of data more efficient. | |
| 65. | Pie charts | | A pie chart is a circular statistical graphic, which is divided into slices to illustrate numerical proportion. | |
| 66. | Bar charts | | A bar chart is a chart which represent the data in the rectangular box in the vertical position. | |

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| 67. | Histograms | | A histogram is a bar graph like representation of data that buckets a range of outcomes into columns along the x-axis. | |
| 68. | Skewness | | Skewness is a measure of the symmetry of a distribution. | |
| 69. | Types of skewness | | 1. positive skewed or right-skewed 2. negative skewed or left-skewed | |
| 70. | Kurtosis | | Kurtosis refers to the degree of presence of outliers in the distribution. | |
| 71. | Excess kurtosis | | The excess kurtosis is used in statistics and probability theory to compare the kurtosis is coefficient with that normal distribution. It can be positive, negative or near to zero. | |
| 72. | Types of kurtosis | | 1. leptokurtic 2. platykurtic 3. mesokurtic | |
| 73. | Pivot tables | | Pivot table are a technique in data processing. They arrange and rearrange statistics in order to draw attention to useful information. | |
| 74. | Two ways of ANOVA | | 1. two way ANOVA with replication 2. two way ANOVA without replication | |
| 75. | Assumptions for two way ANOVA | | <ul style="list-style-type: none"> • The population must be close to a normal distribution • Samples must be independent • Population variances must be equal • Groups must have equal sample sizes. | |
| UNIT -IV Model Development | | | | |
| 76. | Regression | | It estimates the relationship between variables | |
| 77. | Types of linear regression | | <ul style="list-style-type: none"> • Simple linear regression • Multiple linear regression | |
| 78. | Error function | | It is the distance between current state and ideal state | |
| 79. | Mean Squared Error | $MSE = \frac{RSS}{n}$ | It is the mean of squared residuals and is calculated by dividing RSS by the number of data values | |
| 80. | Root Mean Squared Error | | It is the square root of mean squared error and is more suitable when large errors are particularly undesirable. | |
| 81. | Mean Absolute Error | | It is the measure of errors between paired observations expressing the same phenomenon | |
| 82. | Ordinary Least Squares | | It is a method in linear regression for estimating the unknown parameters by creating a model | |
| 83. | Feature Selection | | Certain features from the dataset are selected as the data is huge and multi dimensional used to better understand the data | |
| 84. | Multi collinearity | | It is a phenomenon in which one feature variable in a regression model is highly linearly correlated with another feature variable | |

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| 85. | Null Hypothesis | | It is a type of hypothesis used in a statistics that proposes that there is no difference between certain characteristics of a population | |
| 86. | Forward selection | | It is a iterative method in which we start with having no feature in the model | |
| 87. | Backward selection | | It is a feature selection technique while building a machine learning model | |
| 88. | Representation Learning | | It is an area of research that focuses on how to learn compact , numerical representations for different sources of signal | |
| 89. | Data Visualization | | It is the process of translating large data sets and metrics into charts , graphs and other visuals | |
| 90. | Data Splitting | | It is the acts of partitioning available data into two portions , usually for cross-validatory purposes | |
| 91. | Data splitting purpose | | There are two portions. One portion is used to develop a Predictive model and another portion is to evaluate the model's performance | |
| 92. | Benefits of Data Visualization | | <ul style="list-style-type: none"> • Increases the speed of decision making • Solves data inefficiencies and absorb vast amounts of data presented in visual formats • Identifies errors and inaccuracies in data quickly • Promotes storytelling and Conveys the right message to the audience • Optimize and instantly retrieve data via tailor-made reports • Explore business insights and achieve business goals | |
| 93. | Data Science Process Flow | | Line Chart , Histogram , piechart , Area plot , Scatter Plots , Hexbins Plot , Heat map , Box plot , Pair Plot , Bar Chart | |
| 94. | Histogram | | It is a graphical representation that organizes a group of data points into specified ranges | |
| 95. | Characteristics of a Histogram | | <ul style="list-style-type: none"> • Used to display Continuous data in a categorical form • No gaps between the bars , Unlike a bar graph • Width of the bins is equal | |
| 96. | Linear Regression | | It is a linear approach for modelling the relationship between a scalar response and one or more explanatory Variables | |
| 97. | Area Under Curve | | It is a measure of the ability of a classifier to distinguish between classes and is used as a summary of ROC curve | |
| 98. | Sensitivity | | It is a metric that evaluates a model's ability to predict true positives of each available category | |
| 99. | Specificity | | It is metric that evaluates a model's ability to predict true negatives of each available category | |
| 100. | Precision | | It indicates the rate at which positive predictions | |

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| | | | are correct | |
| UNIT-V Model Evaluation | | | | |
| 101. | Model Evaluation | | Model Evaluation is the Subsidiary part of the model development process. It is the phase that is decided whether the model performs better. | |
| 102. | Generalization | | It refers to your model's ability to adapt properly to new , previously unseen data | |
| 103. | Bias | | Bias is the average squared difference between prediction and true values. It measure how good your model fits the data | |
| 104. | Variance | | If you train your data on training data and obtain a very low error , upon changing the data and then training the same previous model , you experience a high error , this is variance. | |
| 105. | Regularization | | It is a method to avoid high variance and overfitting as well as to increase generalization | |
| 106. | Confusion Matrix | | Confusion matrix is an N x N matrix , where N represents the number of categories in the target variable | |
| 107. | Cost Of Classification | | Cost of Classification is a measure of computing cost for classification models | |
| 108. | Accuracy | | It is the ratio of correct predicted values over the total predicted values | |
| 109. | True Positive rate | | $TPR = \frac{TP}{TP + FN}$ | |
| 110. | False Negative Rate | | $FNR = \frac{FN}{TP + FN}$ | |
| 111. | True Negative Rate | | $TNR = \frac{TN}{FP + TN}$ | |
| 112. | False Positive Rate | | $FPR = \frac{FP}{FP + TN}$ | |
| 113. | Precision | | $Precision = \frac{TP}{FP + TP}$ It is an evaluation metric which tells us out of all positive predictions , how many are actually positive | |
| 114. | Recall | | $Recall = \frac{TP}{FN + TP}$ | |
| 115. | F1 Score | | $F1 = \frac{1}{\frac{1}{Precision} + \frac{1}{Recall}}$ F1 is the harmonic mean of precision and recall | |

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| 116. | Log Loss | | Log Loss is the negative average of the log of corrected-predicted probabilities for each instance | |
| 117. | AUC-ROC | | Area Under the Curve - Receiver Operating characteristics is an evaluation metric for binary classification which gives trade-off between false positive rate and true positive rate | |
| 118. | Overfitting | | Refers to a model can't generalize or fit well on unseen data set. | |
| 119. | Underfitting | | Refers to a model that can neither model the training dataset nor generalize to new dataset. | |
| 120. | Ridge Regression | | It is a model tuning method that is used to analyze any data that suffers from multicollinearity | |
| 121. | To Prevent Overfitting | | <ul style="list-style-type: none"> You need to add regularization in case of Linear and SVM models. In decision tree models you can reduce the maximum depth. While in Neural Networks, you can introduce dropout layer to reduce overfitting | |
| 122. | To Prevent Underfitting | | <ul style="list-style-type: none"> Increase model complexity Increase the number of features , performing feature engineering Remove noise from the data Increase the number of epochs or increase the duration of training to get better results | |
| 123. | Logistic Regression | | It is a supervised machine learning algorithm used to predict a dependent | |
| 124. | Hyperparameter | | These are parameters whose values control the learning process and determine the values of model parameters that a learning algorithm ends up learning | |
| 125. | Parameter | | It is a function argument that could have one of range of values | |

Placement Questions

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| 126. | Three times the first of three consecutive odd integers is 3 more than twice the third. The third integer is: | | Let the three integers be x , $x + 2$ and $x + 4$. Then, $3x = 2(x + 4) + 3 \Leftrightarrow x = 11$. \therefore Third integer = $x + 4 = 15$. | |
| 127. | Look at this series: 7, 10, 8, 11, 9, 12, ... | | This is a simple alternating addition and subtraction series. In the first pattern, 3 is added; in the second, 2 is subtracted. | |
| 128. | Look at this series: 22, 21, 23, 22, 24, 23, | | In this simple alternating subtraction and addition series; 1 is subtracted, then 2 is added, and so on. | |
| 129. | Look at this series: 53, 53, 40, 40, 27, 27, ... | | In this series, each number is repeated, then 13 is subtracted to arrive at the next number. | |

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| 130. | Look at this series: 1.5, 2.3, 3.1, 3.9, ... | | In this simple addition series, each number increases by 0.8. | |
| 131. | Three times the first of three consecutive odd integers is 3 more than twice the third. The third integer is: | | Let the three integers be x , $x + 2$ and $x + 4$. Then, $3x = 2(x + 4) + 3 \Leftrightarrow x = 11$. \therefore Third integer = $x + 4 = 15$. | |
| 132. | Look at this series: 7, 10, 8, 11, 9, 12, ... | | This is a simple alternating addition and subtraction series. In the first pattern, 3 is added; in the second, 2 is subtracted. | |
| 133. | Look at this series: 22, 21, 23, 22, 24, 23, | | In this simple alternating subtraction and addition series; 1 is subtracted, then 2 is added, and so on. | |
| 134. | $(112 \times 5^4) = ?$ | | $(112 \times 5^4) = 112 \times (10)^4 = 112 \times 10^4 = 1120000 = 7000022^4 16$ | |
| 135. | It was Sunday on Jan 1, 2006. The day of the week Jan 1, 2010 is | | On 31 st December, 2005 it was Saturday. Number of odd days from the year 2006 to the year 2009 = $(1 + 1 + 2 + 1) = 5$ days. \therefore On 31 st December 2009, it was Thursday. Thus, on 1 st Jan, 2010 it is Friday. | |
| 136. | Today is Monday. After 61 days, it will be: | | Each day of the week is repeated after 7 days. So, after 63 days, it will be Monday. \therefore After 61 days, it will be Saturday. | |
| 137. | If 6 th March, 2005 is Monday, The day of the week on 6 th March, 2004 is | | The year 2004 is a leap year. So, it has 2 odd days. But, Feb 2004 not included because we are calculating from March 2004 to March 2005. So it has 1 odd day only. \therefore The day on 6 th March, 2005 will be 1 day beyond the day on 6 th March, 2004. Given that, 6 th March, 2005 is Monday. \therefore 6 th March, 2004 is Sunday (1 day before to 6 th March, 2005). | |
| 138. | The days in x weeks x days? | | x weeks x days = $(7x + x)$ days = $8x$ days. | |
| 139. | On 8 th Feb, 2005 it was Tuesday. The day of the week on 8 th Feb, 2004 is | | The year 2004 is a leap year. It has 2 odd days. \therefore The day on 8 th Feb, 2004 is 2 days before the day on 8 th Feb, 2005. Hence, this day is Sunday. | |
| 140. | The greatest number that will divide 43, 91 and 183 so as to leave the same remainder in each case. | | Required number = H.C.F. of $(91 - 43)$, $(183 - 91)$ and $(183 - 43)$ = H.C.F. of 48, 92 and 140 = 4. | |
| 141. | The H.C.F. of two numbers is 23 and the other two factors of their L.C.M. are 13 and 14. The larger of the two numbers | | Clearly, the numbers are (23×13) and (23×14) . \therefore Larger number = $(23 \times 14) = 322$ | |

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| | is: | | | |
| 142. | $(112 \times 5^4) = ?$ | | $(112 \times 5^4) = 112 \times (10)^4 = 112 \times 10^4 = 1120000 = 7000022^4 16$ | |
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| 149. | The H.C.F. of two numbers is 23 and the other two factors of their L.C.M. are 13 and 14. The larger of the two numbers is: | | Clearly, the numbers are (23×13) and (23×14) . ∴ Larger number = $(23 \times 14) = 322$ | |
| 150. | Two trains running in opposite directions cross a man standing on the platform in 27 seconds and 17 seconds respectively and they cross each other in 23 seconds. The ratio of their speeds is: | | Let the speeds of the two trains be x m/sec and y m/sec respectively. Then, length of the first train = $27x$ meters, and length of the second train = $17y$ meters. $\therefore \frac{27x + 17y}{x + y} = 23$ $\Rightarrow 27x + 17y = 23x + 23y$ $\Rightarrow 4x = 6y$ $\Rightarrow \frac{x}{y} = \frac{3}{2}$ | |

Faculty Team Prepared
1. **Dr.P.Srinivasan**

Signatures

HoD