1. What is Hadoop?

Hadoop is an open-source framework for distributed storage and processing of large datasets across clusters of commodity hardware. Its main components are HDFS & MapReduce Programming Model.

2. List Components of Hadoop.

- 1. Hadoop Common
- 2. Hadoop Distributed File System (HDFS)
- 3. MapReduce
- 4. YARN (Yet Another Resource Negotiator)
- 5. Hadoop MapReduce
- 6. Hadoop Distributed Copy (DistCp)
- 7. Hadoop Archives (HAR)
- 8. Hadoop Common Utilities
- 9. Hadoop Streaming
- 10. Hadoop Hive
- 11. Hadoop Pig
- 12. Hadoop HBase
- 13. Hadoop Spark

3. What are the key components of Cloudera Hadoop

- 1. Hadoop Distributed File System (HDFS)
- 2. MapReduce
- 3. YARN (Yet Another Resource Negotiator)
- 4. HBase
- 5. Hive
- 6. Pig
- 7. Impala
- 8. Sqoop
- 9. Flume
- 10. Kafka
- 11. Oozie
- 12. Hue
- 13. Spark
- 14. Sentry
- 15. Navigator

4. what is Apache Hive?

Apache Hive is a data warehouse infrastructure on Hadoop, facilitating SQL-like queries and analysis of large datasets stored in distributed environments.

5. What is Apache Hbase?

Apache HBase is a distributed, scalable, and NoSQL database built on Hadoop, offering real-time, random access to large volumes of data.

6. What is Apache spark?

Apache Spark is an open-source, distributed computing framework that provides fast and flexible data processing capabilities, supporting batch processing, interactive querying, and stream processing tasks on

large-scale data sets.

7. Give syntax and example of HDFS get and put commands

The 'hdfs put' and 'hdfs get' commands are used to copy from the local file system and the Hadoop Distributed File System (HDFS).

hdfs dfs -put <localSrc> <dst>

- '<localSrc>': Path to the source file or directory in the local file system.
- '<dst>': Destination path in HDFS.

hdfs dfs -put /path/to/local/file.txt /user/hadoop/data/file.txt

hdfs dfs -get <src> <localDst>

- '<src>': Path to the source file or directory in HDFS.
- '<localDst>': Destination path in the local file system.

hdfs dfs -get /user/hadoop/data/file.txt /path/to/local/file.txt

8. What is the default replication factor in HDFS.

The default replication factor in HDFS (Hadoop Distributed File System) is 3.

9. What is MapReduce?

MapReduce is a programming model and parallel processing framework used in Hadoop for distributed processing of large datasets. It divides computational tasks into two phases: Map, where data is processed and transformed into intermediate key-value pairs, and reduce, where intermediate results are aggregated and combined to produce the final output.

10. Write 3 Advantages of Data Replication in Hadoop.

Fault Tolerance

Data Locality and Performance

Resilience to Network and Hardware Failures

11. What is Big Data?

Big data refers to large volumes of structured, semi-structured, or unstructured data that cannot be processed effectively using traditional database and software techniques.

12. List Characteristics of Big Data

- 1. Volume
- 2. Velocity
- 3. Variety
- 4. Variability
- 5. Veracity
- 6. Value
- 7. Complexity

13. List 7 Vs of Big Data

- 1. Volume
- 2. Velocity
- 3. Variety
- 4. Variability
- 5. Veracity

- 6. Value
- 7. Visualization

14. What are the common imports for MapReduce Program In Java

import java.io.IOException;

import org.apache.hadoop.io.LongWritable;

import org.apache.hadoop.io.Text;

import org.apache.hadoop.mapreduce.Mapper;

import org.apache.hadoop.mapreduce.Reducer;

import org.apache.hadoop.mapreduce.Job;

import org.apache.hadoop.mapreduce.lib.input.TextInputFormat;

import org.apache.hadoop.mapreduce.lib.output.TextOutputFormat;

15. What are advantages of Hadoop?

- 1. Scalability
- 2. Fault tolerance
- 3. Cost-effectiveness
- 4. Flexibility
- 5. Parallel processing
- 6. Support for diverse data types
- 7. Easy integration with existing systems

16. What is the latest Release of Apache Hadoop?

3.4.0 (March 2024)

17. Who created Hadoop

Hadoop was created by Doug Cutting and Mike Cafarella in 2005 at Yahoo.

18. At present who develops Hadoop

Hadoop is currently developed and maintained by the Apache Software Foundation and a diverse community of contributors worldwide. This community-driven development model ensures continuous improvement, innovation, and support for the Hadoop ecosystem.

19. Who is the chief architect of cloudera?

Doug Cutting serves as Chief Architect of Cloudera

20. What is Namenode in Hadoop

The NameNode in Hadoop is a core component that manages the file system namespace and metadata for files stored in Hadoop Distributed File System (HDFS), keeping track of file locations and replication. It serves as the centralized point of contact for client applications to access and manage data stored in the Hadoop cluster.

21. What is Datanode in Hadoop?

Datanode in Hadoop is a node in the cluster responsible for storing and managing data blocks, as well as serving read and write requests from clients. It communicates with the NameNode to report its status and perform data replication and recovery tasks in Hadoop Distributed File System (HDFS).

22. What is Jobtracker?

The JobTracker is a master node in Hadoop's MapReduce framework responsible for coordinating and managing MapReduce jobs submitted to the Hadoop cluster. It tracks the progress of job execution, schedules tasks on TaskTracker nodes, handles task failures and retries, and ensures overall job completion within the cluster.

23. What is Tasktracker?

TaskTracker is a node-level component in Hadoop MapReduce responsible for executing and monitoring Map and Reduce tasks assigned by the JobTracker. It manages task execution and status reporting within the cluster.

24. What is Metadata In HDFS

This metadata includes details such as file names, file sizes, permissions, file locations, replication factors,

timestamps, and other attributes associated with data stored in Hadoop. Metadata is crucial for managing and accessing data efficiently, enabling functions such as file system operations, data replication, fault tolerance, and job scheduling within the Hadoop cluster.

25. Single Namenode can communicate with how many Datanodes?

A single NameNode can manage and coordinate communication with hundreds to thousands of DataNodes

26. What is fsimage?

The 'fsimage' (File System Image) is a persistent, centralized metadata image that contains the namespace and metadata information about the Hadoop Distributed File System (HDFS). It includes details such as file and directory structures, permissions, ownership, replication factors, and block mappings. The 'fsimage' is stored on the NameNode and is used to bootstrap the file system namespace during NameNode startup, providing a snapshot of the file system's state at a given point in time.

27. How fsimage is used in Hadoop?

- 1. Backup of File System State: The `fsimage` serves as a backup of the file system's namespace and metadata, providing a snapshot of the file system's state at a given point in time.
- 2. NameNode Recovery: In the event of a NameNode failure, the `fsimage` is used to restore the file system namespace and metadata during the recovery process, ensuring minimal downtime and data loss.
- 3. Bootstrap NameNode: During NameNode startup, the 'fsimage' is loaded into memory to initialize the file system namespace, providing an efficient way to bootstrap the file system without needing to scan all data directories.
- 4. Checkpointing: The `fsimage` is periodically checkpointed along with the EditLog to create consistent snapshots of the file system's state. These snapshots serve as recovery points and help reduce the time required to restore the file system in case of failure.
- 5. Cluster Maintenance: The `fsimage` can be used for cluster maintenance tasks such as migrating or upgrading Hadoop clusters, ensuring that the file system state is preserved during these operations.

28. What is rack aware data storage in Hadoop?

Rack-aware storage in Hadoop optimizes data placement and replication across racks in a cluster, considering the physical network topology. It reduces network traffic and improves fault tolerance by ensuring data replicas are stored across multiple racks, enhancing data locality and resilience within the cluster's infrastructure. This approach enhances performance and reliability by minimizing network hops and mitigating the impact of rack-level failures in Hadoop clusters.

29. How data node communicated with Name node?

By sending Heart bits and Block reports.

30. What is Data explosion?

Data explosion refers to the exponential growth of digital data generated and collected by organizations, individuals, and devices, driven by factors such as the proliferation of digital technologies, the internet, and connected devices, leading to challenges in data storage, management, and analysis.

32. What is the task of Map Function in MapReduce?

The task of the map function in MapReduce is to process input data and generate intermediate key-value pairs, typically by applying a transformation or computation to each input record independently. This function is applied in parallel across multiple input splits, allowing for distributed and scalable processing of large datasets in Hadoop clusters.

33. What is the task of Reduce Function in MapReduce?

The task of the reduce function in MapReduce is to process and aggregate intermediate key-value pairs generated by the map function, typically by grouping together all values associated with the same key and performing a specified computation or operation on them. This function consolidates and summarizes the intermediate results produced by the map phase, producing the final output of the MapReduce job.

34. Write steps in data analytics lifecycle

- 1. Data Collection
- 2. Data Preprocessing
- 3. Data Exploration and Analysis
- 4. Data Modeling and Algorithms
- 5. Data Visualization and Interpretation
- 6. Model Evaluation and Validation
- 7. Deployment and Implementation
- 8. Monitoring and Maintenance

35. List few data preprocessing Techniques?

- 1. Data Cleaning
- 2. Missing Value Imputation
- 3. Outlier Detection and Removal
- 4. Data Transformation (Normalization/Standardization)
- 5. Handling Imbalanced Data

36. What is data cleaning?

Data cleaning is the process of identifying, correcting, and removing errors, inconsistencies, duplicates, and irrelevant information from raw datasets to ensure accuracy, completeness, and consistency, thus preparing the data for analysis, modeling, or other downstream tasks.

37.List Data Cleaning techniques.

- 1. Handling Missing Values
- 2. Handling Outliers
- 3. Handling Duplicates
- 4. Handling Inconsistent Data
- 5. Handling Incomplete Data
- 6. Handling Irrelevant Data
- 7. Handling Noisy Data
- 8. Handling Encoding Issues

37. What is error correction in data?

Error correction in data refers to the process of identifying and rectifying inaccuracies, inconsistencies, or mistakes in the data to ensure its accuracy, reliability, and consistency for analysis or usage. This involves employing various techniques such as data cleaning, validation, imputation, and transformation to detect and correct errors, improving the quality and integrity of the data.

38. List Data error correction Techniques

- 1. Missing Value Imputation
- 2. Outlier Detection and Removal
- 3. Duplicate Detection and Removal
- 4. Error Correction and Standardization
- 5. Inconsistent Data Handling
- 6. Data Imputation for Incomplete Data

- 7. Irrelevant Data Removal
- 8. Noise Reduction

39. What is data Transformation?

Data transformation is the systematic process of converting, modifying, or reformatting data from its original state to a desired format, structure, or representation to facilitate analysis, visualization, or other data-related tasks.

40. List different data transformation Techniques.

- 1. Normalization: Scaling data to a standard range (e.g., between 0 and 1).
- 2. Standardization: Transforming data to have a mean of 0 and standard deviation of 1.
- 3. Encoding: Converting categorical variables into numerical representations.
- 4. Aggregation: Combining data points into summary statistics or aggregated values.
- 5. Feature Engineering: Creating new features from existing ones.
- 6. Binning or Discretization: Grouping continuous variables into bins.
- 7. Text Preprocessing: Cleaning and transforming text data.
- 8. Date and Time Transformation: Extracting useful information from date and time variables.
- 9. Dimensionality Reduction: Reducing the number of features in a dataset.
- 10. Feature Scaling: Bringing all features to a similar scale.

41. What is data Subset?

A data subset refers to a portion or a subset of a larger dataset that is selected or extracted based on specific criteria, such as certain attributes, conditions, or sampling methods. It contains a smaller, more manageable portion of the original dataset, typically used for analysis, modeling, or other data-related tasks. Data subsets are often created to focus on specific aspects of the data or to reduce computational complexity when working with large datasets.

42. What are the different ways to obtain data subsets.

- 1. Random Sampling: Selecting data points randomly from the dataset.
- 2. Stratified Sampling: Randomly sampling from homogeneous groups within the dataset.
- 3. Systematic Sampling: Selecting data points at regular intervals from the dataset.
- 4. Cluster Sampling: Randomly selecting entire clusters from the dataset.
- 5. Convenience Sampling: Selecting data points based on convenience or availability.
- 6. Filtering: Extracting data points that meet specific criteria from the dataset.
- 7. Partitioning: Dividing the dataset into predefined segments based on certain attributes.
- 8. Feature Selection: Selecting a subset of features (columns) from the dataset.

43. What is data shape?

Data shape refers to the dimensions of a dataset, typically represented as the number of rows and columns it contains, providing insight into its size and organization

44. What is Data Reshaping?

Data reshaping is the process of reorganizing the structure of a dataset, typically by changing its dimensions or rearranging its rows and columns, to meet specific requirements or analysis needs, ensuring compatibility with different tools or models. It involves transforming data from one shape to another while preserving its integrity and meaning.

Reshaping data with Numpy Example array with shape (2, 3)

```
import numpy as np
data = np.array([[1, 2, 3],
[4, 5, 6]])
```

```
# Reshape the array to have shape (3, 2)
reshaped_data = np.reshape(data, (3, 2))
print("Original Data:")
print(data)
print("\nReshaped Data:")
print(reshaped_data)
# Reshaping with pandas
import pandas as pd
```

```
# Example DataFrame
data = pd.DataFrame({
  'A': [1, 2, 3],
  'B': [4, 5, 6]
})
# Reshape the DataFrame using the melt function
reshaped data = data.melt()
print("Original Data:")
print(data)
print("\nReshaped Data:")
print(reshaped data)
Original Data:
 A B
0 1 4
1 2 5
2 3 6
Reshaped Data:
 variable value
0
     Α
           1
           2
1
     Α
2
     A 3
3
     В
         4
           5
4
     В
5
     B
           6
```

45. What is HiveQL?

HiveQL, or Hive Query Language, is a SQL-like language used for querying and analyzing large datasets in Apache Hive. It provides a familiar syntax for users accustomed to SQL, allowing them to perform data manipulation and analysis tasks on structured data stored in Hive tables. HiveQL queries are translated into MapReduce jobs or other execution engine tasks to process data stored in Hadoop Distributed File System (HDFS) or other compatible storage systems.

46. What is data Integration?

Data Integration Example

Data integration refers to the process of combining data from different sources, formats, or systems into a unified and coherent view, enabling seamless access, sharing, and analysis of data across an organization. It involves harmonizing data structures, resolving inconsistencies, and ensuring data quality to create a comprehensive and reliable dataset that supports various business processes, analytics, and decision-making activities.

0 1 Alice 25 1 2 Bob 30

47. What are the steps in model building with data sets?

- 1. Data Preprocessing: Clean, transform, and prepare the dataset for modeling.
- 2. Feature Engineering: Create or select relevant features from the dataset.
- 3. Model Selection: Choose appropriate algorithms or techniques.
- 4. Model Training: Fit the selected models to the dataset.
- 5. Model Evaluation: Assess the performance of the trained models.
- 6. Hyperparameter Tuning: Optimize the model's hyperparameters.
- 7. Model Deployment: Deploy the trained models into production environments.

48. What are the different statistical models?

- 1. **Linear Regression:** Models the relationship between independent variables and a continuous dependent variable.
- 2. Logistic Regression: Models the probability of a binary outcome based on independent variables.
- 3. Time Series Models: Analyze and forecast time-dependent data patterns.
- 4. Survival Analysis: Analyzes time-to-event data, such as failure or death.
- 5. Multivariate Analysis: Examines relationships between multiple variables simultaneously.
- 6. **Generalized Linear Models (GLMs):** Extend linear regression for non-normally distributed data or non-linear relationships.
- 7. **Mixed Effects Models:** Incorporate fixed and random effects for nested data structures.
- 8. **Bayesian Models:** Represent uncertainty using probability distributions and update beliefs based on evidence.

49. What is use of NumPy Library in Python?

NumPy is a fundamental library for numerical computing in Python, providing support for multidimensional arrays, mathematical functions, linear algebra operations, random number generation, and more. It is widely used in scientific computing, data analysis, machine learning, and other domains where efficient numerical operations are required.

50. List few Numpy Library Functions?

- 1. np.array(): Create a NumPy array from a Python list or tuple.
- 2. np.zeros(): Create an array filled with zeros of a specified shape.
- 3. np.ones(): Create an array filled with ones of a specified shape.
- 4. np.arange(): Create an array with evenly spaced values within a specified range.
- 5. np.reshape(): Reshape an array into a new shape.
- 6. np.concatenate(): Join arrays along an existing axis.
- 7. np.sum(): Compute the sum of array elements.
- 8. np.mean(): Compute the mean of array elements.
- 9. np.std(): Compute the standard deviation of array elements.
- 10. np.min(): Find the minimum value in an array.
- 11. np.max(): Find the maximum value in an array.
- 12. 14. np.linalg.inv(): Compute the inverse of a square matrix.
- 13. np.random.rand(): Generate random numbers from a uniform distribution.
- 14. np.random.shuffle(): Shuffle the elements of an array in place.
- 15. np.random.choice(): Generate random samples from a given 1-D array.

51. What is the use of pandas Library in Python.

The pandas library in Python is primarily used for data manipulation and analysis tasks, providing high-performance, easy-to-use data structures and tools for working with structured (tabular) data. It offers functionalities for data cleaning, transformation, exploration, and visualization, making it a powerful tool for data scientists, analysts, and developers working with datasets in various domains.

52. List Few functions in Pandas Library.

- **1. pd.read** csv(): Read data from a CSV file into a DataFrame.
- **2. DataFrame.head():** Return the first n rows of a DataFrame.
- 3. DataFrame.describe(): Generate descriptive statistics of a DataFrame's numeric columns.
- **4. DataFrame.info():** Print a concise summary of a DataFrame's information.
- **5. DataFrame.dropna():** Drop rows or columns containing missing values from a DataFrame.
- **6. DataFrame.fillna():** Fill missing values in a DataFrame with specified values.
- 7. DataFrame.groupby(): Group DataFrame rows based on specified criteria for aggregation or analysis.
- **8. DataFrame.merge():** Merge DataFrame objects by performing a database-style join operation.
- 9. DataFrame.plot(): Create plots and visualizations of data stored in a DataFrame.
- 10. DataFrame.to csv(): Write a DataFrame to a CSV file.

53. What is data Visualization?

Data visualization is the graphical representation of data using charts, graphs, plots, or other visual elements to convey insights, patterns, trends, and relationships within the data. It enables users to explore, understand, and communicate complex datasets more effectively, facilitating data-driven decision-making and storytelling.

54. What are the different data visualization Techniques?

- 1. **Scatter Plots:** Display individual data points as markers on a two-dimensional plane to show relationships between two variables.
- 2. Line Charts: Plot data points connected by lines to visualize trends or changes over time.
- 3. **Bar Charts:** Represent categorical data using rectangular bars of varying lengths to compare values across different categories.
- 4. **Histograms:** Display the distribution of numerical data by dividing it into bins and showing the frequency of data points within each bin.
- 5. **Pie Charts:** Present data as a circular chart divided into slices to show the proportion of each category relative to the whole.
- 6. **Heatmaps:** Use color gradients to represent data values in a matrix, making it easier to identify patterns and trends.
- 7. **Box Plots:** Show the distribution of data along with key statistical measures such as median, quartiles, and outliers.
- 8. **Area Charts:** Plot data points as continuous lines with the area below the line filled to represent the cumulative effect of the data.
- 9. **Bubble Charts:** Display data points as bubbles of varying sizes to represent additional dimensions of the data, such as a third numerical variable.
- 10. **Treemaps:** Represent hierarchical data structures using nested rectangles to visualize the distribution of values across different levels.
- 11. **Parallel Coordinates:** Plot multivariate data by representing each variable as a vertical axis and connecting data points with lines to show relationships between variables.
- 12. **Choropleth Maps:** Use color shading or patterns to represent data values for different geographical regions on a map.

55. What is the use of Matplotlib and seaborn Library in python?

Matplotlib and Seaborn are two popular libraries in Python used for data visualization:

- 1. Matplotlib: Matplotlib is a comprehensive library for creating static, interactive, and animated visualizations in Python. It provides a wide range of plotting functions and customization options, making it suitable for various visualization tasks. Matplotlib is highly customizable but may require more code to create complex visualizations.
- **2. Seaborn:** Seaborn is built on top of Matplotlib and provides a high-level interface for creating attractive and informative statistical graphics. It simplifies the process of creating complex visualizations by offering built-in themes, color palettes, and statistical functions. Seaborn is particularly useful for exploratory data analysis and generating publication-quality plots with less code compared to Matplotlib.

56. List few functions of matplotlib and seaborn library in python Matplotlib:

- 1. plt.plot(): Create a line plot.
- 2. plt.scatter(): Create a scatter plot.
- 3. plt.bar(): Create a bar plot.
- 4. plt.hist(): Create a histogram.
- 5. plt.boxplot(): Create a box plot.
- 6. plt.imshow(): Display an image.
- 7. plt.pie(): Create a pie chart.
- 8. plt.contour(): Create contour plots.
- 9. plt.errorbar(): Create error bar plots.
- 10. plt.annotate(): Annotate points on a plot.

Seaborn:

- 1. **sns.lineplot():** Create a line plot with optional semantic grouping.
- 2. **sns.scatterplot():** Create a scatter plot with optional semantic grouping.
- 3. sns.barplot(): Create a bar plot with optional aggregation of values.
- 4. sns.histplot(): Create a histogram with optional kernel density estimation.
- 5. **sns.boxplot():** Create a box plot with optional grouping by categorical variables.
- 6. sns.heatmap(): Create a heatmap to visualize matrix-like data.
- 7. **sns.pairplot():** Create a grid of pairwise plots for exploring relationships in a dataset.
- 8. **sns.lmplot():** Create a scatter plot with linear regression line fit.
- 9. sns.jointplot(): Create a joint plot for visualizing joint distributions between two variables.
- 10. sns.catplot(): Create categorical plots (e.g., bar plots, box plots) with options for faceting.

57. What is data Warehouse?

A data warehouse is a centralized repository that integrates data from multiple sources into a unified and consistent format, optimized for querying and analysis. It stores historical and current data to support decision-making processes, business intelligence, and reporting activities within an organization, typically organized in a dimensional or relational schema for easy access and retrieval.

58. What is data mining?

Data mining is the process of discovering meaningful patterns, trends, and insights from large datasets using statistical, mathematical, and machine learning techniques. It involves extracting valuable knowledge and information from raw data to support decision-making, prediction, and knowledge discovery in various domains such as business, science, and healthcare.

59. What is pydoop?

pydoop: A Python API for Hadoop MapReduce that allows developers to write MapReduce applications in Python.

60. What is Tableau?

Tableau is a powerful data visualization and business intelligence tool that allows users to create interactive and shareable dashboards, reports, and visualizations from various data sources. It provides a user-friendly interface for analyzing and visualizing large datasets without requiring extensive programming knowledge. Tableau supports connecting to a wide range of data sources, including databases, spreadsheets, cloud services, and big data platforms, making it popular among data analysts, business users, and organizations for exploring and communicating insights from their data.

61. Write features of Tableau.

- 1. **Interactive Dashboards:** Create interactive dashboards for data exploration.
- 2. **Data Connection:** Connect to various data sources effortlessly.
- 3. **Data Blending:** Blend data from multiple sources seamlessly.
- 4. **Visual Analytics:** Analyze data with a wide range of visualizations.
- 5. Advanced Calculations: Perform complex calculations with ease.
- 6. Geospatial Analysis: Visualize and analyze geographic data.
- 7. **Collaboration:** Share insights and collaborate with team members.
- 8. Mobile Compatibility: Access dashboards on mobile devices.
- 9. **Integration:** Integrate with other applications and platforms effortlessly.
- 10.**Scalability:** Scale deployments to meet organizational needs.

62. Who developed Tableau?

Tableau was developed by a company called Tableau Software, which was founded by Chris Stolte, Pat Hanrahan, and Christian Chabot in 2003. Since then, Tableau has grown to become one of the leading data visualization and business intelligence platforms in the industry.

63. What is one dimensional data visualization?

In Tableau, one-dimensional visualization refers to the representation of data along a single axis or dimension. This typically involves visualizations such as histograms, bar charts, line charts, or pie charts, where data is displayed along one axis (x-axis or y-axis) without considering relationships between multiple variables. One-dimensional visualizations are useful for exploring distributions, trends, or patterns within a single variable or category of data.

e.g. Consider a dataset containing the ages of individuals. We can create a histogram in Tableau where the x-axis represents the age range (bins), and the y-axis represents the frequency or count of individuals falling within each age range. Each bar in the histogram represents a specific age range, and its height corresponds to the number of individuals within that age range.

64. What is 2-D Data visualization?

In Tableau, two-dimensional visualization refers to the representation of data using two axes or dimensions. This involves visualizations such as scatter plots, bubble charts, and stacked bar charts, where data is displayed along both the x-axis and y-axis simultaneously. Two-dimensional visualizations enable users to explore relationships, correlations, and patterns between two variables or categories of data, providing insights into how they interact or influence each other.

e.g. we have a dataset containing information about the height and weight of individuals. We can create a scatter plot in Tableau where the height is plotted along the x-axis, and the weight is plotted along the y-axis. Each data point on the scatter plot represents an individual, with their height and weight values determining its position on the plot.

65. What is 3D Visualization?

3D (volumetric) data visualization refers to the representation of data in three-dimensional space, allowing users to explore relationships and patterns among three variables simultaneously. While Tableau primarily focuses on two-dimensional visualizations, certain techniques can be used to create pseudo-3D effects or layered visualizations.

Consider the "Adult" dataset, which contains demographic and income information about individuals. We

can create a pseudo-3D visualization in Tableau by representing three variables, such as age, education level, and income.

We could use a scatter plot where the x-axis represents age, the y-axis represents education level (e.g., years of education), and the color or size of the data points represents income level. While this visualization is technically two-dimensional, the addition of color or size encoding creates the illusion of a third dimension, allowing us to explore relationships among age, education, and income simultaneously.

66. What is multi dimensional Visualization?

In Tableau, multidimensional visualization involves representing data that encompasses multiple dimensions or variables simultaneously. Using the "Adult" dataset, which contains demographic and income information about individuals, we can create multidimensional visualizations to explore relationships among various attributes.

An example of multidimensional visualization in Tableau using the "Adult" dataset could be a parallel coordinates plot. In this visualization, each line represents an individual, and each vertical axis represents a different attribute or dimension, such as age, education level, occupation, and income.

Create a parallel coordinates plot in Tableau using the "Adult" dataset:

- 1. Drag the desired dimensions (e.g., age, education, occupation) onto the Rows shelf to create vertical axes.
- 2. Drag the measure (e.g., income) onto the Columns shelf to represent the value along each axis.
- 3. Tableau will automatically connect the data points (individuals) with lines, creating a parallel coordinates plot.

By examining the parallel coordinates plot, we can visually analyze the relationships among multiple dimensions simultaneously. For example, we can identify how income varies across different levels of education, occupations, and age groups, allowing for a comprehensive exploration of the dataset's multidimensional aspects.

67. What is Temporal Data Visualization?

Temporal data visualization in Tableau involves visualizing data that varies over time. While the "Adult" dataset doesn't inherently contain temporal data, we can still create an example using derived temporal information, such as the year of birth or the date of an event (if applicable).

For this example, let's say we have a modified version of the "Adult" dataset that includes the year of birth of individuals. We can create a temporal data visualization to analyze the distribution of births over different years.

Temporal data visualization in Tableau using the "Adult" dataset:

- 1. Load the "Adult" dataset into Tableau.
- 2. Create a calculated field to extract the year from the date of birth column (assuming it exists in the dataset).
- 3. Drag the calculated year field onto the Columns shelf to represent time.
- 4. Drag a measure onto the Rows shelf to represent the count of individuals born in each year.
- 5. Choose a suitable visualization type, such as a line chart or bar chart, to visualize the distribution of births over time.

By examining the temporal data visualization, we can analyze trends and patterns in the distribution of births over different years. While this example may not directly relate to the original "Adult" dataset, it demonstrates how temporal data visualization can be applied to explore data that varies over time in Tableau.

68. What is Tree/Hierarchical data visualization?

Tree or hierarchical data visualization in Tableau involves representing data organized in a hierarchical structure, such as organizational charts, directory structures, or hierarchical relationships. While the "Adult" dataset doesn't inherently contain hierarchical data, we can create an example using derived hierarchical information, such as job roles within different industries.

For this example, let's say we have a modified version of the "Adult" dataset that includes information about job roles and industries. We can create a hierarchical data visualization to analyze the distribution of job roles within different industries.

Hierarchical data visualization in Tableau using the "Adult" dataset:

- 1. Load the "Adult" dataset into Tableau.
- 2. Create a calculated field to combine the job role and industry information into a hierarchical structure.
- 3. Drag the hierarchical field onto the Rows or Columns shelf to represent the hierarchy.
- 4. Drag a measure onto the Rows or Columns shelf to represent the count of individuals in each job role within each industry.
- 5. Choose a suitable visualization type, such as a treemap or sunburst chart, to visualize the hierarchical data structure.

By examining the hierarchical data visualization, we can analyze the distribution of job roles within different industries and explore the hierarchical relationships between them. While this example may not directly relate to the original "Adult" dataset, it demonstrates how hierarchical data visualization can be applied to explore hierarchical structures in Tableau.

69. What is Network data visualization?

Network data visualization in Tableau involves representing interconnected data using nodes and edges, such as social networks, organizational structures, or network traffic. While the "Adult" dataset doesn't inherently contain network data, we can create an example using derived relational information, such as relationships between individuals based on shared attributes.

For this example, let's say we have a modified version of the "Adult" dataset that includes information about relationships between individuals, such as familial relationships or professional connections. We can create a network data visualization to analyze the connections between individuals based on these relationships.

Network data visualization in Tableau using the "Adult" dataset:

- 1. Load the "Adult" dataset into Tableau.
- 2. Create a calculated field to represent the connections between individuals based on shared attributes or relationships.
- 3. Define the nodes (individuals) and edges (connections) in the network data structure.
- 4. Use Tableau's built-in network visualization features or custom visualizations to represent the network graph.
- 5. Customize the visualization to highlight important nodes, visualize network centrality, or analyze network clusters.

By examining the network data visualization, we can analyze the connections between individuals in the dataset and explore the network structure. While this example may not directly relate to the original "Adult" dataset, it demonstrates how network data visualization can be applied to explore relational data structures in Tableau.

70. What is Web Scraping? Write sample code for web scraping.

Web scraping is the process of extracting data from websites. It involves using automated tools or scripts to navigate web pages, retrieve the desired content, and store it in a structured format for further analysis or use. Web scraping can be used to gather a wide range of information from the web, such as product prices, news articles, weather data, or contact information, among others. It is commonly used in various fields, including data analysis, market research, content aggregation, and competitive intelligence.

```
import requests
from bs4 import BeautifulSoup
def scrape amazon reviews(url):
  # Send a GET request to the URL
  response = requests.get(url)
  # Parse the HTML content using BeautifulSoup
  soup = BeautifulSoup(response.text, 'html.parser')
  # Find the container that holds all the reviews
  reviews = soup.find all('div', class ='review')
  # Iterate through each review and extract information
  for review in reviews:
    # Extract review text
    review text = review.find('span', class ='review-text').get text(strip=True)
    # Extract review rating
     review rating = review.find('i', class ='review-rating').get text(strip=True)
    # Extract review tags
    review tags = review.find all('span', class = 'review-tag')
     tags = [tag.get text(strip=True) for tag in review tags]
     # Extract customer name
    customer name = review.find('span', class ='review-user').get text(strip=True)
     # Print the extracted information
     print("Review:", review text)
    print("Rating:", review rating)
     print("Tags:", tags)
    print("Customer Name:", customer name)
    print()
# Example URL for Amazon product reviews
amazon url = 'https://www.amazon.com/product-reviews/B07C25XC4R'
# Scrape reviews from the provided URL
scrape amazon reviews(amazon url)
Sample output:
Review: This product is amazing! I love the design and functionality.
Rating: 5 out of 5 stars
Tags: ['Verified Purchase', 'Great product', 'Highly recommended']
Customer Name: John Doe
Review: Not impressed with this product. It stopped working after a week.
Rating: 2 out of 5 stars
Tags: ['Verified Purchase', 'Not recommended']
```

Customer Name: Jane Smith