

Assignment Code: DS-AG-005

Statistics Basics | Assignment

Instructions: Carefully read each question. Use Google Docs, Microsoft Word, or a similar tool to create a document where you type out each question along with its answer. Save the document as a PDF, and then upload it to the LMS. Please do not zip or archive the files before uploading them. Each question carries 20 marks.

Total Marks: 200

Question 1: What is the difference between descriptive statistics and inferential statistics? Explain with examples.

Answer:

1. Descriptive Statistics

- Definition: Summarizes and organizes collected data so we can understand it easily.
- Purpose: To describe what the data shows.
- Tools: Mean, Median, Mode, Standard Deviation, Graphs, Charts, Tables.

Example: - Marks of 50 students in a class:

- Average marks = 68
- Highest marks = 95
- Lowest marks = 32
- Standard deviation = 10

*Here, we only describe the collected data, no prediction about a larger population.

2. Inferential Statistics

- Definition: Makes predictions or conclusions about a population based on a sample.
- Purpose: To generalize findings beyond the data we have.
- Tools: Hypothesis testing, Confidence intervals, Regression, Correlation.

Example:- You survey 100 voters in a city of 10,000:

 60% of sample prefer Party A → You estimate that roughly 60% of the entire city supports Party A.

*Here, we infer population trends from a sample.

Question 2: What is sampling in statistics? Explain the differences between random and stratified sampling.

Answer:

Sampling: Selecting a small group (sample) from a large population to study and make conclusions.

• Why? Studying the whole population is time-consuming, costly, or impossible.

Example:

A school has 2000 students. Instead of checking marks of all students, you randomly select 200 students and

study their performance.

1. Random Sampling

- Every individual in the population has an equal chance of being selected.
- Simple and unbiased method.

Example: Pick 200 students randomly from 2000 students' roll numbers.

2. Stratified Sampling

- Population is divided into groups/strata (like gender, class, age).
- A proportional sample is selected from each group to ensure representation.

Example:

2000 students → 1000 boys & 1000 girls.



Question 3: Define mean, median, and mode. Explain why these measures of central tendency are important.

Answer:

These are statistical tools that indicate the center or average of a dataset.

- 1. Mean (Arithmetic Average) :- Sum of all observations divided by the number of observations.
 - Formula: Mean=Sum of observations/Number of observations
 - Example: Marks = 10, 20, 30 \rightarrow Mean = (10+20+30)/3 = 20
- 2. Median :-Middle value of data when arranged in ascending or descending order.
 - Example: 10, 20, 30 → Median = 20
 If data is even: 10, 20, 30, 40 → Median = (20+30)/2 = 25
- 3. Mode :- Most frequently occurring value in the dataset.
 - Example: 2, 4, 4, 6, 7 → Mode = 4

Importance:-

- 1. Summarizes data by representing it with a single value.
- 2. Shows the central point of the data distribution.
- 3. Helps in comparison between different datasets (e.g., average marks of classes).
- 4. Each measure has specific uses:
 - Mean: Useful for overall average (salary, marks).
 - Median: Useful when there are extreme values (income distribution).
 - Mode: Useful to find the most common value (fashion size, exam scores).

Question 4: Explain skewness and kurtosis. What does a positive skew imply about the data?

Answer:

1. Skewness: - Skewness measures the asymmetry of a data distribution around its mean.

- Types:
 - Positive skew (right skew): Tail stretches to the right. Most data values are clustered on the left, with few high values on the right.
 - Negative skew (left skew): Tail stretches to the left. Most data values are clustered on the right, with few low values on the left.
- Example:
 - \circ Positive skew: Income of people in a city \to most earn low/average, few earn very high.

- \circ Negative skew: Age at retirement \rightarrow most retire around 60, few retire early.
- Implication of Positive Skew:
 - Mean > Median > Mode
 - Data is concentrated at lower values with some extremely high values pulling the tail to the right.

2. Kurtosis:- Kurtosis measures the peakedness or flatness of a distribution compared to a normal distribution.

- Types:
 - Leptokurtic: High peak, heavy tails (more extreme values).
 - Platykurtic: Flat distribution, light tails (less extreme values).
 - Mesokurtic: Normal peak (similar to normal distribution).
- Example:
 - Leptokurtic: Test scores where most students score around average but some score extremely high/low.
 - Platykurtic: Uniform distribution like rolling a fair die.

Question 5: Implement a Python program to compute the mean, median, and mode of a given list of numbers.

numbers = [12, 15, 12, 18, 19, 12, 20, 22, 19, 19, 24, 24, 24, 26, 28]

(Include your Python code and output in the code box below.)

Answer:



```
#code:-
# Import required modules
from statistics import mean, median, mode
# Given list of numbers
numbers = [12, 15, 12, 18, 19, 12, 20, 22, 19, 19, 24, 24, 24, 26, 28]
# Compute mean, median, and mode
mean_value = mean(numbers)
median value = median(numbers)
mode_value = mode(numbers)
# Print results
print("Numbers:", numbers)
print("Mean:", mean_value)
print("Median:", median value)
print("Mode:", mode value)
OUTPUT:-
Numbers: [12, 15, 12, 18, 19, 12, 20, 22, 19, 19, 24, 24, 24, 26, 28]
Mean: 19.6
Median: 19
Mode: 12
```

Question 6: Compute the covariance and correlation coefficient between the following two datasets provided as lists in Python:

```
list y = [15, 25, 35, 45, 60]
```

(Include your Python code and output in the code box below.)

Answer:

Paste your code and output inside the box below:

```
# Import required modules
import numpy as np
# Given datasets
list_x = [10, 20, 30, 40, 50]
list_y = [15, 25, 35, 45, 60]
# Convert lists to numpy arrays
x = np.array(list_x)
y = np.array(list_y)
# Compute covariance
cov_matrix = np.cov(x, y) # covariance matrix
cov_xy = cov_matrix[0, 1] # covariance between x and y
# Compute correlation coefficient
corr_matrix = np.corrcoef(x, y)
corr_xy = corr_matrix[0, 1] # correlation coefficient
# Print results
print("List X:", list_x)
print("List Y:", list_y)
print("Covariance:", cov_xy)
print("Correlation Coefficient:", corr_xy)
Output :-
List X: [10, 20, 30, 40, 50]
List Y: [15, 25, 35, 45, 60]
Covariance: 275.0
Correlation Coefficient: 0.995893206467704
```

Question 7: Write a Python script to draw a boxplot for the following numeric list and identify its outliers. Explain the result:

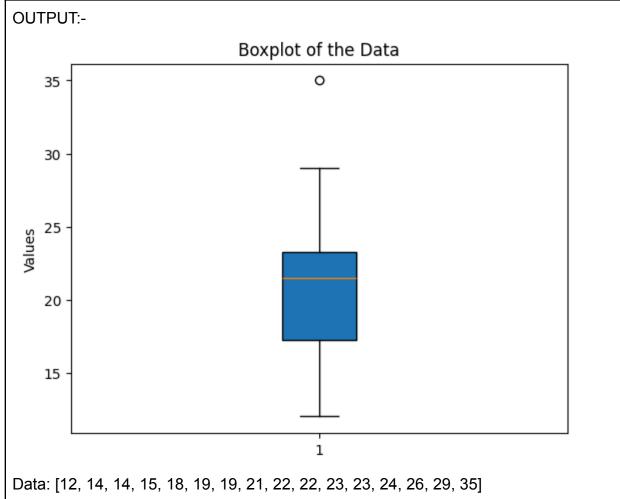
data = [12, 14, 14, 15, 18, 19, 19, 21, 22, 22, 23, 23, 24, 26, 29, 35]

Answer:



```
# Import required libraries
import matplotlib.pyplot as plt
# Given dataset
data = [12, 14, 14, 15, 18, 19, 19, 21, 22, 22, 23, 23, 24, 26, 29, 35]
# Draw boxplot
plt.boxplot(data, vert=True, patch artist=True)
plt.title("Boxplot of the Data")
plt.ylabel("Values")
plt.show()
# Identify outliers using IQR method
Q1 = np.percentile(data, 25) # 1st quartile
Q3 = np.percentile(data, 75) # 3rd quartile
IQR = Q3 - Q1
                          # Interquartile range
lower bound = Q1 - 1.5 * IQR
upper bound = Q3 + 1.5 * IQR
# Find outliers
outliers = [x \text{ for } x \text{ in data if } x < \text{lower bound or } x > \text{upper bound}]
print("Data:", data)
print("Q1:", Q1, "Q3:", Q3, "IQR:", IQR)
print("Lower Bound:", lower bound, "Upper Bound:", upper bound)
print("Outliers:", outliers)
```

3



Q1: 17.25 Q3: 23.25 IQR: 6.0

Lower Bound: 8.25 Upper Bound: 32.25

Outliers: [35]

Question 8: You are working as a data analyst in an e-commerce company. The marketing team wants to know if there is a relationship between advertising spend and daily sales.

- Explain how you would use covariance and correlation to explore this relationship.
- Write Python code to compute the correlation between the two lists:

advertising_spend = [200, 250, 300, 400, 500]

daily_sales = [2200, 2450, 2750, 3200, 4000]

(Include your Python code and output in the code box below.)

Answer:

Covariance:

- Measures how two variables vary together.
- Positive covariance → when advertising spend increases, sales tend to increase.
- Negative covariance → when advertising spend increases, sales tend to decrease.
- Limitation: Not standardized; hard to compare magnitude.

Correlation Coefficient (r):

- Standardized version of covariance.
- Ranges from -1 to 1.
 - \circ r ≈ 1 \rightarrow strong positive relationship
 - $r \approx -1 \rightarrow \text{strong negative relationship}$
 - \circ r ≈ 0 \rightarrow no linear relationship
- Helps marketing team understand strength and direction of the relationship between ad spend and sales.

CODE:-

Import required library

import numpy as np

Given data

advertising_spend = [200, 250, 300, 400, 500] daily_sales = [2200, 2450, 2750, 3200, 4000]

Convert to numpy arrays

```
x = np.array(advertising_spend)
y = np.array(daily_sales)
# Compute covariance
cov_matrix = np.cov(x, y)
cov_xy = cov_matrix[0, 1]
# Compute correlation coefficient
corr_matrix = np.corrcoef(x, y)
corr_xy = corr_matrix[0, 1]
# Print results
print("Covariance:", cov_xy)
print("Correlation Coefficient:", corr_xy)
OUTPUT:-
Covariance: 84875.0
Correlation Coefficient: 0.9935824101653329
```

Question 9: Your team has collected customer satisfaction survey data on a scale of 1-10 and wants to understand its distribution before launching a new product.

- Explain which summary statistics and visualizations (e.g. mean, standard deviation, histogram) you'd use.
- Write Python code to create a histogram using Matplotlib for the survey data:



4

Answer:

Summary Statistics

- 1. Mean \rightarrow Average satisfaction score.
- 2. Median \rightarrow Middle value; useful if there are extreme scores.
- 3. Mode → Most common satisfaction score.
- 4. Standard Deviation (SD) \rightarrow Measures how spread out the scores are.
- 5. Range / Min / Max \rightarrow Gives the overall spread of scores.

Visualizations:-

- 1. Histogram → Shows how frequently each score occurs; visualizes distribution.
- 2. Boxplot → Helps identify median, quartiles, and any outliers.

Why?

- Helps marketing team see the overall trend of customer satisfaction before product launch.
- Can identify if most customers are satisfied (high scores) or if there are concerns (low scores).

CODE:-

Import required library

import matplotlib.pyplot as plt

