Statistics in Data Science

- 1. Write a Python script to analyze a dataset and demonstrate:
 - a. The impact of outliers on the mean, median, and mode.
 - b. How these measures can be used to summarize data effectively.

Steps

- a. Create a dataset with outliers.
- b. Calculate and compare the central tendency measures before and after outlier removal.

Measures of Central Tendency

2. Use a dataset containing test scores to compute the mean, median, and mode using Python. Also, visualize these statistics on a histogram.

Steps

- a. Use Python's statistics module for calculations.
- b. Use matplotlib or seaborn for the histogram.

Measures of Dispersion

- 3. Write a Python program to calculate:
 - a. The range, interquartile range (IQR), variance, and standard deviation of a dataset.
 - b. Visualize the spread using a boxplot.

Steps

- a. Use numpy for dispersion measures.
- b. Use matplotlib for the boxplot.

Hypothesis Testing

4. Write a Python program to perform a paired t-test to evaluate the effectiveness of a study program. Use two datasets: scores before and after the program.

Steps

- a. Use scipy.stats.ttest_rel() for the t-test.
- b. Interpret the p-value to determine significance.

Types of Errors

- 5. Write a Python function to simulate a hypothesis test with the following:
 - a. A scenario that results in a Type I error (rejecting a true null hypothesis).
- b. A scenario that results in a Type II error (failing to reject a false null hypothesis).

Steps

- a. Use synthetic data and statistical tests.
- b. Highlight conditions leading to each error type.

Regression Analysis

- 6. Use a dataset of years of experience vs. salary to:
 - a. Fit a simple linear regression model.
 - b. Predict the salary for 5 years of experience.

Steps:

- a. Use sklearn.linear model.LinearRegression.
- b. Visualize the regression line over the scatterplot.

Underfitting and Overfitting

- 7. Demonstrate underfitting and overfitting using polynomial regression:
 - a. Generate a dataset from a quadratic equation with noise.
 - b. Fit a linear model (underfitting) and a high-degree polynomial (overfitting).
 - c. Visualize both models and their performance on training and testing sets.

Steps

- a. Use numpy to generate synthetic data.
- b. Use sklearn for regression models.

Types of Regression Analysis

- 8. Compare the performance of:
 - a. Linear regression
 - b. Polynomial regression
 - c. Lasso regression

Use a real or synthetic dataset to fit these models and compare their mean squared errors.

Steps

- a. Use sklearn for implementation.
- b. Use matplotlib to visualize results.

Correlation and Regression

- 9. Create a scatterplot to visualize the relationship between two variables (e.g., hours studied vs. marks scored).
 - a. Calculate the Pearson correlation coefficient.
 - b. Fit a linear regression model and overlay the regression line on the scatterplot.

Steps

- a. Use numpy for correlation.
- b. Use sklearn for regression and matplotlib for visualization.

Note: After completing the code, submit it in .ipynb format along with a brief explanation for each part in your own words.