

Sample Dataset:

You can import using this:

```
"from sklearn.datasets import fetch_california_housing  
import pandas as pd
```

```
# Load the California housing dataset  
california = fetch_california_housing()
```

```
# Create a DataFrame  
df = pd.DataFrame(california.data, columns=california.feature_names)
```

```
# Display the first few rows  
df.head()
```

Tasks:

1. Outlier Detection and Removal

Description:

Steps:

a. Data Loading:

- Provide a dataset (e.g., a housing dataset, sales data, or any other dataset with numerical attributes).
- Load the dataset and display basic statistics using `pandas.describe()`.

b. Task 1: Outlier Detection Using Z-score:

- Instruct students to calculate the Z-scores for a specific numerical feature (e.g., `price` or `size`).
- Set a threshold for detecting outliers (e.g., $Z\text{-score} > 3$ or $Z\text{-score} < -3$).
- Identify and count the number of outliers based on the Z-score threshold.
- Visualize the outliers using scatter plots or box plots.

c. Task 2: Outlier Detection Using IQR:

- Calculate the 25th percentile (Q1) and 75th percentile (Q3) of a specific feature.
- Compute the IQR ($Q3 - Q1$) and identify outliers as values below $Q1 - 1.5 * IQR$ or above $Q3 + 1.5 * IQR$.
- Remove the outliers and compare the dataset before and after outlier removal (e.g., using visualizations or statistical summaries).

2. Feature Scaling Using Standardization and Min-Max Scaling

Steps:

1. Data Loading:

- Provide a dataset (e.g., California Housing, Titanic dataset, etc.).
- Load the dataset and describe the features using summary statistics (`pandas.describe()`).

2. Task 1: Standardization (Z-score Scaling):

- Select one or more numerical features from the dataset (e.g., `price`, `age`, or `salary`).
- Apply Z-score standardization to these features using `sklearn.preprocessing.StandardScaler`.
- Display the transformed features and compare the results to the original data.
- Visualize the original and standardized data using histograms and box plots.

3. Task 2: Min-Max Scaling:

- Apply Min-Max scaling to the same features using `sklearn.preprocessing.MinMaxScaler`.
- Display and compare the transformed features with the original data.
- Visualize the original and Min-Max scaled data.