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In [1]: # Experiment No: 9 - Create a linear regression model using housing dataset
In [1]: # Importing Libraries
        import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        from sklearn.model selection import train test split
        from sklearn.linear model import LinearRegression
        from sklearn.metrics import mean squared error, r2 score
In [2]: # Loading the Boston housing dataset from the original source
        data url = "http://lib.stat.cmu.edu/datasets/boston"
        raw df = pd.read csv(data url, sep="\s+", skiprows=22, header=None)
In [3]: # Extracting features and target variable
        # The dataset contains 13 features and the target variable is the median val
        X = np.hstack([raw df.values[::2, :], raw df.values[1::2, :2]])
        y = raw df.values[1::2, 2]
In [4]: # Converting to DataFrame for easier manipulation
        feature names = ['CRIM', 'ZN', 'INDUS', 'CHAS', 'NOX', 'RM', 'AGE', 'DIS',
        X = pd.DataFrame(X, columns=feature names)
In [5]: # Printing data types of the features
        print("\nData types of the features:")
        print(X.dtypes)
       Data types of the features:
       CRIM
                 float64
       7N
                  float64
       INDUS
                float64
                float64
       CHAS
       NOX
                 float64
       RM
                 float64
       AGE
                float64
       DIS
                float64
       RAD
                float64
                 float64
       TAX
       PTRATIO float64
                  float64
       LSTAT
                  float64
       dtype: object
In [6]: # Displaying the head of the dataset
        print("First 5 rows of the Boston Housing dataset:")
        print(X.head())
```

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First 5 rows of the Boston Housing dataset:
                    ZN INDUS CHAS
                                                 AGE
                                                          DIS RAD TAX \
             CRIM
                                      NOX
       0 0.00632 18.0
                        2.31
                                0.0 0.538 6.575 65.2 4.0900 1.0 296.0
                                0.0 0.469 6.421 78.9 4.9671 2.0 242.0
       1 0.02731 0.0 7.07
       2 0.02729 0.0 7.07
                               0.0 0.469 7.185 61.1 4.9671 2.0 242.0
       3 0.03237 0.0 2.18
                                0.0 0.458 6.998 45.8 6.0622 3.0 222.0
       4 0.06905 0.0 2.18
                                0.0 0.458 7.147 54.2 6.0622 3.0 222.0
          PTRATIO
                     B LSTAT
             15.3 396.90
       0
                         4.98
             17.8 396.90 9.14
       1
       2
             17.8 392.83 4.03
       3
             18.7 394.63 2.94
       4
             18.7 396.90 5.33
In [7]: # Splitting the dataset into training and testing sets
        X train, X test, y train, y test = train test split(X, y, test size=0.2, rar)
In [8]: # Creating a linear regression model
        model = LinearRegression()
        model.fit(X train, y train) # Fitting the model
Out[8]:
            LinearRegression •
        LinearRegression()
In [9]: # Making predictions on the test set
        y pred = model.predict(X test)
In [10]: # Evaluating the model's performance
        mse = mean squared error(y test, y pred)
        r2 = r2 \ score(y \ test, y \ pred)
In [11]: # Displaying the results
        print("\nMean Squared Error:", mse)
        print("R^2 Score:", r2)
       Mean Squared Error: 24.291119474973478
       R^2 Score: 0.6687594935356326
In [12]: # Visualizing the predictions
        plt.scatter(y test, y pred)
        plt.xlabel('Actual Prices')
        plt.ylabel('Predicted Prices')
        plt.title('Actual vs Predicted Prices')
        plt.plot([y.min(), y.max()], [y.min(), y.max()], 'r--') # Diagonal line
        plt.show()
```

Actual vs Predicted Prices

