

o Internship in AI/ML



Boston House price prediction

```
In [1]: import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        import numpy as np
        from sklearn.model_selection import train_test_split
        from sklearn.impute import SimpleImputer
        from sklearn.preprocessing import StandardScaler
        from sklearn.linear_model import LinearRegression
        from sklearn.metrics import mean_squared_error, r2_score
In [2]: df = pd.read_csv(r"C:\Users\Ashish Mishra\OneDrive\Desktop\ShadowFox\Shadowfox_AI-M
        df.head()
Out[2]:
             CRIM
                    ZN INDUS CHAS
                                       NOX
                                               RM AGE
                                                            DIS RAD TAX PTRATIO
                                                                                         ВΙ
        0 0.00632 18.0
                           2.31
                                       0.538 6.575
                                                    65.2 4.0900
                                                                       296
                                                                                15.3 396.90
        1 0.02731
                           7.07
                                                                       242
                    0.0
                                   0.0 0.469 6.421
                                                   78.9 4.9671
                                                                   2
                                                                                17.8 396.90
        2 0.02729
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                                   0.0 0.469 7.185
                                                   61.1 4.9671
                                                                       242
                                                                                17.8 392.83
        3 0.03237
                    0.0
                           2.18
                                   0.0 0.458 6.998
                                                   45.8 6.0622
                                                                       222
                                                                                18.7 394.63
        4 0.06905
                    0.0
                           2.18
                                   0.0 0.458 7.147 54.2 6.0622
                                                                   3 222
                                                                                18.7 396.90
In [3]: print(df.info())
        print(df.describe())
```

file:///C:/Users/Ashish Mishra/Downloads/Task Level-2.html

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 506 entries, 0 to 505
Data columns (total 14 columns):
     Column
              Non-Null Count Dtype
     -----
              -----
     CRIM
              486 non-null
                               float64
 0
 1
     ΖN
              486 non-null
                               float64
 2
     INDUS
              486 non-null
                               float64
 3
     CHAS
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     NOX
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     RM
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                               float64
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     DIS
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     RAD
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 9
     TAX
              506 non-null
                               int64
     PTRATIO
              506 non-null
 10
                               float64
              506 non-null
 11
     В
                               float64
 12 LSTAT
              486 non-null
                               float64
 13 MEDV
              506 non-null
                               float64
dtypes: float64(12), int64(2)
memory usage: 55.5 KB
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                                                         506.000000
                                                                      506.000000
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                                 11.083992
                                               0.069959
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                                                            0.554695
                                                                        6.284634
std
         8.720192
                     23.388876
                                  6.835896
                                               0.255340
                                                            0.115878
                                                                        0.702617
min
         0.006320
                      0.000000
                                  0.460000
                                               0.000000
                                                            0.385000
                                                                        3.561000
                                  5.190000
                                               0.000000
25%
         0.081900
                      0.000000
                                                            0.449000
                                                                        5.885500
                                  9.690000
                      0.000000
                                               0.000000
                                                                        6.208500
50%
         0.253715
                                                            0.538000
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                     12.500000
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mean
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                      3.795043
                                             408.237154
                                                           18.455534
                                                                      356.674032
std
        27.999513
                      2.105710
                                  8.707259
                                             168.537116
                                                            2.164946
                                                                       91.294864
min
         2.900000
                      1.129600
                                  1.000000
                                             187.000000
                                                           12.600000
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max
            LSTAT
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count
       486.000000
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        12.715432
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         7.155871
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         7.125000
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50%
        11.430000
                     21.200000
75%
        16.955000
                     25.000000
max
        37.970000
                     50.000000
```

In [4]: print("Missing values:", df.isnull().sum())

```
Missing values: CRIM
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         INDUS
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                        20
         CHAS
         NOX
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         RM
                         0
         AGE
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         DIS
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         RAD
                         0
         TAX
                         0
         PTRATIO
                         0
                         0
         LSTAT
                        20
         MEDV
                         0
         dtype: int64
In [5]: plt.figure(figsize=(12, 8))
           sns.heatmap(df.corr(), annot=True, cmap="coolwarm", fmt=".2f")
           plt.show()
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         LSTAT
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                          INDUS CHAS
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                                                                   RAD
                                                                         TAX PTRATIO
                                                                                       В
                     ΖN
                                               RM
In [6]: X = df.drop('MEDV', axis=1) # Features
           y = df['MEDV'] # Target variable
In [7]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
In [8]: imputer = SimpleImputer(strategy='median')
```

```
In [9]: X_train_imputed = imputer.fit_transform(X_train)
         X_test_imputed = imputer.transform(X_test)
In [10]: model = LinearRegression()
         model.fit(X_train_imputed, y_train)
Out[10]:
             LinearRegression
         LinearRegression()
In [11]: y_pred = model.predict(X_test_imputed)
In [12]: mse = mean_squared_error(y_test, y_pred)
         rmse = np.sqrt(mse)
         r2 = r2_score(y_test, y_pred)
In [13]: print(f"Mean Squared Error (MSE): {mse}")
         print(f"Root Mean Squared Error (RMSE): {rmse}")
         print(f"R-squared: {r2}")
        Mean Squared Error (MSE): 24.983445419912112
        Root Mean Squared Error (RMSE): 4.998344267846314
        R-squared: 0.659318743105165
In [14]: plt.figure(figsize=(6, 6))
         plt.scatter(y_test, y_pred)
         plt.xlabel("Actual Prices")
         plt.ylabel("Predicted Prices")
         plt.title("Actual vs Predicted House Prices")
         plt.show()
```

Actual vs Predicted House Prices

