

California Housing Dataset

```
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
```

```
df = pd.read_csv('/content/housing[1].csv')
print(df.head())
```

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	\
0	-122.23	37.88	41.0	880.0	129.0	
1	-122.22	37.86	21.0	7099.0	1106.0	
2	-122.24	37.85	52.0	1467.0	190.0	
3	-122.25	37.85	52.0	1274.0	235.0	
4	-122.25	37.85	52.0	1627.0	280.0	

	population	households	median_income	median_house_value	ocean_proximity
0	322.0	126.0	8.3252	452600.0	NEAR BAY
1	2401.0	1138.0	8.3014	358500.0	NEAR BAY
2	496.0	177.0	7.2574	352100.0	NEAR BAY
3	558.0	219.0	5.6431	341300.0	NEAR BAY
4	565.0	259.0	3.8462	342200.0	NEAR BAY

```
print("\nMissing values in dataset:\n", df.isnull().sum())
```

```
Missing values in dataset:
longitude      0
latitude      0
housing_median_age  0
total_rooms    0
total_bedrooms 207
population    0
households    0
median_income  0
median_house_value  0
ocean_proximity  0
dtype: int64
```

```
X = df[['median_income']]
y = df['median_house_value']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
model = LinearRegression()
model.fit(X_train, y_train)
```

LinearRegression ⓘ ?

LinearRegression()

```
y_pred = model.predict(X_test)
```

```
print("\n Model Evaluation:")
print("Mean Squared Error:", mean_squared_error(y_test, y_pred))
print("R2 Score:", r2_score(y_test, y_pred))
```

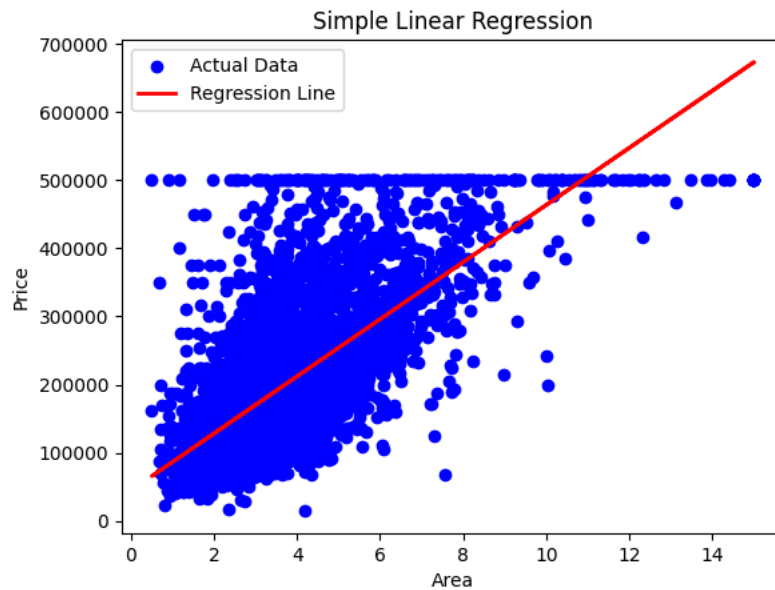
Model Evaluation:

Mean Squared Error: 7091157771.76555

R² Score: 0.45885918903846656

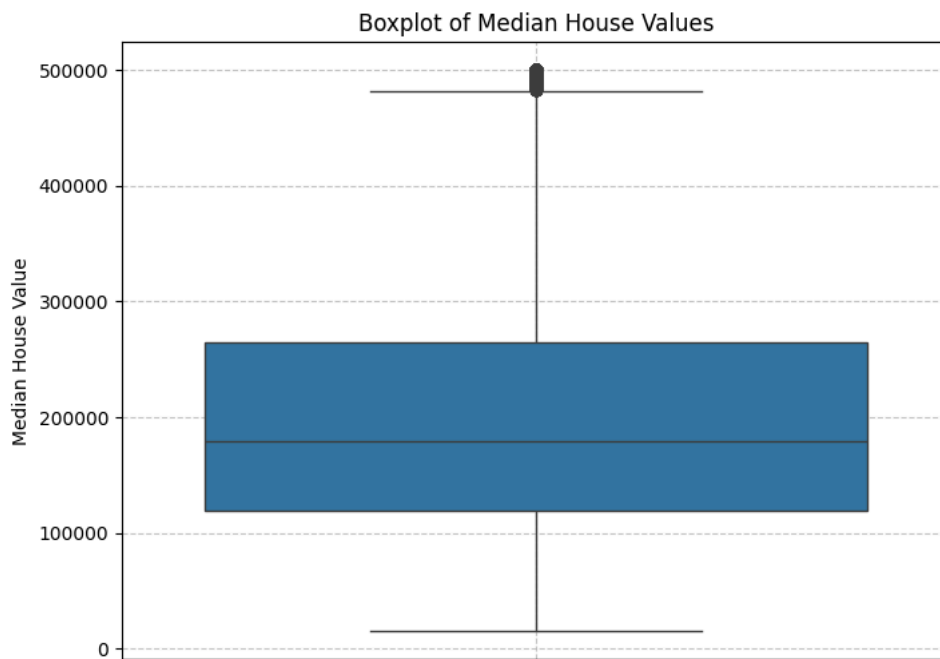
```
plt.scatter(X_test, y_test, color='blue', label='Actual Data')
plt.plot(X_test, y_pred, color='red', linewidth=2, label='Regression Line')
plt.xlabel('Area')
plt.ylabel('Price')
```

```
plt.title('Simple Linear Regression')
plt.legend()
plt.show()
```



```
import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(8, 6))
sns.boxplot(y=df['median_house_value'])
plt.title('Boxplot of Median House Values')
plt.ylabel('Median House Value')
plt.grid(True, linestyle='--', alpha=0.7)
plt.show()
```



```
median_total_bedrooms = df['total_bedrooms'].median()
df['total_bedrooms'].fillna(median_total_bedrooms, inplace=True)
print("Missing values after imputation:")
print(df.isnull().sum())
```

```
Missing values after imputation:
longitude      0
latitude       0
housing_median_age  0
total_rooms     0
total_bedrooms  0
```

```

population      0
households      0
median_income   0
median_house_value 0
ocean_proximity 0
dtype: int64
/tmp/ipython-input-2549053295.py:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through c[...]
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col]

df['total_bedrooms'].fillna(median_total_bedrooms, inplace=True)

```

```

median_total_bedrooms = df['total_bedrooms'].median()
df['total_bedrooms'] = df['total_bedrooms'].fillna(median_total_bedrooms)
print("Missing values after imputation:")
print(df.isnull().sum())

```

```

Missing values after imputation:
longitude      0
latitude      0
housing_median_age 0
total_rooms    0
total_bedrooms 0
population     0
households     0
median_income  0
median_house_value 0
ocean_proximity 0
dtype: int64

```

```

df_encoded = pd.get_dummies(df, columns=['ocean_proximity'], drop_first=False)

print("DataFrame after one-hot encoding:\n", df_encoded.head())

```

```

DataFrame after one-hot encoding:
  longitude  latitude  housing_median_age  total_rooms  total_bedrooms  \
0   -122.23    37.88             41.0         880.0         129.0
1   -122.22    37.86             21.0        7099.0        1106.0
2   -122.24    37.85             52.0        1467.0         190.0
3   -122.25    37.85             52.0        1274.0         235.0
4   -122.25    37.85             52.0        1627.0         280.0

  population  households  median_income  median_house_value  \
0       322.0       126.0         8.3252        452600.0
1      2401.0      1138.0         8.3014        358500.0
2       496.0       177.0         7.2574        352100.0
3       558.0       219.0         5.6431        341300.0
4       565.0       259.0         3.8462        342200.0

  ocean_proximity_<1H  OCEAN  ocean_proximity_INLAND  ocean_proximity_ISLAND  \
0                False      False                  False                  False
1                False      False                  False                  False
2                False      False                  False                  False
3                False      False                  False                  False
4                False      False                  False                  False

  ocean_proximity_NEAR  BAY  ocean_proximity_NEAR  OCEAN
0                True      False
1                True      False
2                True      False
3                True      False
4                True      False

```

```

X = df_encoded.drop('median_house_value', axis=1)
y = df_encoded['median_house_value']

```

```

print("Shape of X:", X.shape)
print("Shape of y:", y.shape)
print("\nFirst 5 rows of X:\n", X.head())
print("\nFirst 5 rows of y:\n", y.head())

```

```

Shape of X: (20640, 13)
Shape of y: (20640,)

```

```

First 5 rows of X:
  longitude  latitude  housing_median_age  total_rooms  total_bedrooms  \
0   -122.23    37.88             41.0         880.0         129.0
1   -122.22    37.86             21.0        7099.0        1106.0
2   -122.24    37.85             52.0        1467.0         190.0
3   -122.25    37.85             52.0        1274.0         235.0

```

```

4      -122.25      37.85      52.0      1627.0      280.0

      population  households  median_income  ocean_proximity_<1H OCEAN \
0           322.0        126.0         8.3252             False
1          2401.0       1138.0         8.3014             False
2           496.0        177.0         7.2574             False
3           558.0        219.0         5.6431             False
4           565.0        259.0         3.8462             False

      ocean_proximity_INLAND  ocean_proximity_ISLAND  ocean_proximity_NEAR BAY \
0                        False                      False                      True
1                        False                      False                      True
2                        False                      False                      True
3                        False                      False                      True
4                        False                      False                      True

      ocean_proximity_NEAR OCEAN
0                        False
1                        False
2                        False
3                        False
4                        False

```

First 5 rows of y:

```

0      452600.0
1      358500.0
2      352100.0
3      341300.0
4      342200.0
Name: median_house_value, dtype: float64

```

```
from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```

print("Shape of X_train:", X_train.shape)
print("Shape of X_test:", X_test.shape)
print("Shape of y_train:", y_train.shape)
print("Shape of y_test:", y_test.shape)

```

```

Shape of X_train: (16512, 13)
Shape of X_test: (4128, 13)
Shape of y_train: (16512,)
Shape of y_test: (4128,)

```

```

model = LinearRegression()
model.fit(X_train, y_train)
print("Linear Regression model trained successfully.")

```

Linear Regression model trained successfully.

```

y_pred = model.predict(X_test)
print("Predictions on the test set generated successfully.")

```

Predictions on the test set generated successfully.

```

print("\nModel Evaluation:")
print("Mean Squared Error:", mean_squared_error(y_test, y_pred))
print("R\u00b2 Score:", r2_score(y_test, y_pred))

```

```

Model Evaluation:
Mean Squared Error: 4908476721.156623
R\u00b2 Score: 0.6254240620553602

```

Visualize Actual vs. Predicted Values

Subtask:

Create a scatter plot comparing the actual house values against the predicted house values to visually assess the model's accuracy and identify any patterns or discrepancies.

```

plt.figure(figsize=(10, 7))
plt.scatter(y_test, y_pred, alpha=0.5, label='Actual vs. Predicted Values')
plt.plot(y_test, y_test, color='red', linestyle='--', label='Perfect Prediction')

```

```
plt.xlabel('Actual House Values')
plt.ylabel('Predicted House Values')
plt.title('Actual vs. Predicted House Values')
plt.legend()
plt.grid(True)
plt.show()
```

