```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from \ sklearn.preprocessing \ import \ StandardScaler
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
data = pd.read_csv('/content/housing.csv')
data.head()
\rightarrow
         longitude \ latitude \ housing\_median\_age \ total\_rooms \ total\_bedrooms \ population
      0
            -122.23
                        37.88
                                                           880.0
                                                                            129.0
                                                                                         322.0
                                               41.0
            -122.22
                                                                                        2401.0
      1
                        37 86
                                               21 0
                                                          7099 0
                                                                           1106 0
            -122.24
      2
                        37.85
                                               52.0
                                                          1467.0
                                                                            190.0
                                                                                         496.0
      3
            -122.25
                        37.85
                                               52.0
                                                          1274.0
                                                                            235.0
                                                                                         558.0
            -122.25
                        37.85
                                               52.0
                                                          1627.0
                                                                            280.0
                                                                                         565.0
      4
                                          View recommended plots
 Next steps:
              Generate code with data
data.describe()
\rightarrow
                longitude
                                latitude housing_median_age
                                                                total rooms total bedrooms
      count 20640.000000 20640.000000
                                                 20640.000000
                                                               20640.000000
                                                                                20433.000000 2
               -119.569704
                                                                                   537.870553
                               35 631861
                                                    28 639486
                                                                2635 763081
      mean
                  2.003532
                                2.135952
                                                    12.585558
                                                                2181.615252
                                                                                  421.385070
       std
               -124.350000
                               32.540000
                                                     1.000000
                                                                    2.000000
                                                                                     1.000000
       min
       25%
               -121.800000
                               33.930000
                                                    18.000000
                                                                1447.750000
                                                                                   296.000000
       50%
               -118.490000
                               34.260000
                                                    29.000000
                                                                2127.000000
                                                                                   435.000000
       75%
               -118.010000
                               37.710000
                                                    37.000000
                                                                                   647.000000
                                                                3148.000000
               -114.310000
                               41.950000
                                                    52.000000
                                                               39320.000000
                                                                                  6445.000000 3
       max
data.isnull().sum()
     longitude
                               0
₹
     latitude
                               0
     housing_median_age
                               0
                               0
     total_rooms
     total bedrooms
                             207
     population
                               0
     households
                               0
     median income
                               0
     median_house_value
                               0
     ocean_proximity
                               0
     dtype: int64
data = data.dropna()
X = data.drop('median_house_value', axis=1)
y = data['median_house_value']
X = pd.get_dummies(X)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
model = LinearRegression()
model.fit(X_train, y_train)
     ▼ LinearRegression
     LinearRegression()
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)
r2 = r2_score(y_test, y_pred)
print(f'Mean Squared Error: {mse}')
print(f'Root Mean Squared Error: {rmse}')
print(f'R^2 Score: {r2}')
→ Mean Squared Error: 4802173538.604162
     Root Mean Squared Error: 69297.71669113032
     R^2 Score: 0.6488402154431991
plt.figure(figsize=(10, 6))
plt.scatter(y\_test, \ y\_pred, \ edgecolors=(0, \ 0, \ 0))
plt.plot([y\_test.min(), y\_test.max()], [y\_test.min(), y\_test.max()], 'k--', lw=3)
plt.xlabel('True Prices')
plt.ylabel('Predicted Prices')
plt.title('True vs Predicted Prices')
plt.show()
```



## True vs Predicted Prices

