```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.model_selection import train_test_split, cross_val_score,GridSearchCV
import plotly.express as px
import plotly.graph_objects as go
from sklearn.linear_model import Ridge
df=pd.read_csv('/content/Housing.csv')
\equiv
              price area bedrooms
                                     bathrooms stories mainroad guestroom basement hotwaterheating airconditioning parking prefarea furnishings1
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                     7420
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           1750000 3850
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     545 rows × 13 columns
              Generate code with df
                                       View recommended plots
df.isnull().sum()
→ price
     area
     bedrooms
     bathrooms
                          0
     stories
                          0
     mainroad
                          0
                          0
     guestroom
     basement
     hotwaterheating
     airconditioning
                          0
     parking
                          0
     prefarea
     furnishingstatus
     dtype: int64
#duplicate values
duplicates= df.duplicated().sum()
duplicates
#no duplicates row
→ 0
df.columns
Index(['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad',
             'guestroom', 'basement', 'hotwaterheating', 'airconditioning', 'parking', 'prefarea', 'furnishingstatus'],
           dtype='object')
px.scatter(df, x = 'area', y='bedrooms')
#categorical to integers
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
df['area']=le.fit_transform(df['area'])
df['bedrooms']=le.fit_transform(df['bedrooms'])
df['bathrooms']=le.fit_transform(df['bathrooms'])
df['mainroad']=le.fit_transform(df['mainroad'])
df['basement']=le.fit_transform(df['basement'])
df['parking']=le.fit_transform(df['parking'])
```

```
x = df[['area','bedrooms','bathrooms','mainroad','basement','parking']]
y = df[['price']]
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)

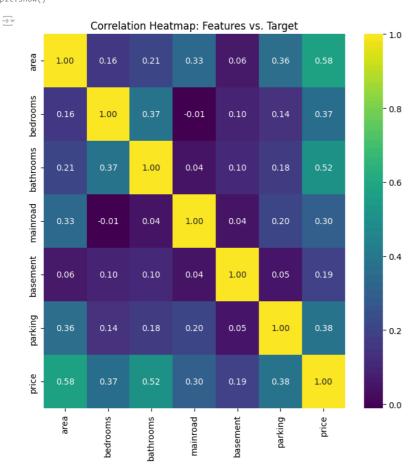
features = df[['area','bedrooms','bathrooms','mainroad','basement','parking']]
target = df[['price']]

# Create a new DataFrame with only the selected columns
data_new= pd.concat([features, target], axis=1)
corr_matrix = data_new.corr()
```

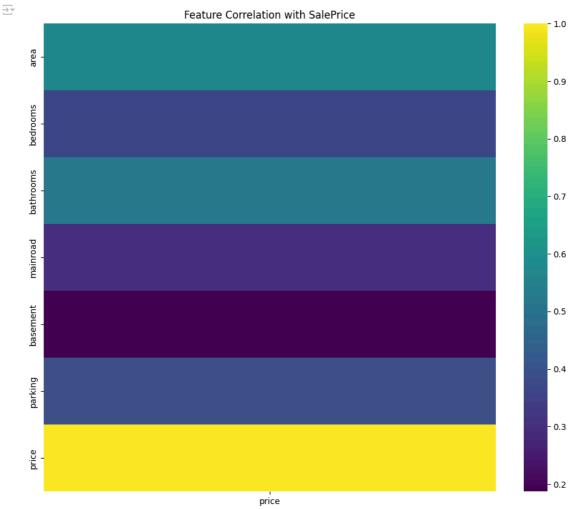
corr\_matrix

$\rightarrow$		area	bedrooms	bathrooms	mainroad	basement	parking	price
	area	1.000000	0.163235	0.211224	0.329097	0.060676	0.360925	0.575107
	bedrooms	0.163235	1.000000	0.373930	-0.012033	0.097312	0.139270	0.366494
	bathrooms	0.211224	0.373930	1.000000	0.042398	0.102106	0.177496	0.517545
	mainroad	0.329097	-0.012033	0.042398	1.000000	0.044002	0.204433	0.296898
	basement	0.060676	0.097312	0.102106	0.044002	1.000000	0.051497	0.187057
	parking	0.360925	0.139270	0.177496	0.204433	0.051497	1.000000	0.384394
	price	0.575107	0.366494	0.517545	0.296898	0.187057	0.384394	1.000000

plt.figure(figsize=(8, 8))
sns.heatmap(corr\_matrix, annot=True, cmap='viridis', fmt=".2f")
plt.title("Correlation Heatmap: Features vs. Target")
plt.show()



#heatmaps
plt.figure(figsize=(12, 10))
sns.heatmap(corr\_matrix[['price']], cmap='viridis')
plt.title("Feature Correlation with SalePrice")
plt.show()



```
#create a linear regression model
model = LinearRegression()
# Fit the model to the training data
model.fit(x\_train,\ y\_train)
     ▼ LinearRegression
     LinearRegression()
# Make predictions on the test data
y_pred = model.predict(x_test)
model.score(x\_train,y\_train)
→ 0.5763925019994292
# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
print(f"Mean Squared Error: {mse:.2f}")
print(f"R-squared: {r2:.2f}")
→ Mean Squared Error: 2498570953172.19
     R-squared: 0.51
model.predict([[7420,4,2,0,1,0]])
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning:
     X does not have valid feature names, but LinearRegression was fitted with feature names
     array([[72871543.50583132]])
plt.scatter(y_test, y_pred)
plt.xlabel("Actual Prices")
plt.ylabel("Predicted Prices")
plt.title("Actual Prices vs. Predicted Prices")
plt.show()
```

