Upload the Dataset

```
from google.colab import files
uploaded = files.upload()
```

```
Choose Files Housing.csv.xlsx
• Housing.csv.xlsx(application/vnd.openxmlformats-officedocument.spreadsheetml.sheet) - 40874 bytes, last modified: 5/8/2025 - 100% done
Saving Housing.csv.xlsx to Housing.csv (2).xlsx
```

```
import pandas as pd
from google.colab import files

# Upload the file
uploaded = files.upload()

# Assuming the uploaded file is named 'Housing.csv.xlsx'
df = pd.read_excel('Housing.csv.xlsx')  # No need for sep

# Display the first few rows
df.head()
```

```
Choose Files Housing.csv.xlsx
• Housing.csv.xlsx(application/vnd.openxmlformats-officedocument.spreadsheetml.sheet) - 40874 bytes, last modified: 5/8/2025 - 100% done
Saving Housing.csv.xlsx to Housing.csv (5).xlsx
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furn
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	

Next steps: Generate code with df View recommended plots New interactive sheet

Data Exploration

```
# Display first few rows
df.head()
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furn
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	

Next steps: Generate code with df View recommended plots New interactive sheet

```
# Cell 2 - Display information about the DataFrame
print("Shape:", df.shape)
print("Columns:", df.columns.tolist())
df.info()
df.describe()
```

```
Shape: (545, 13)
Columns: ['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544
Data columns (total 13 columns):
 #   Column            Non-Null Count  Dtype
---  ------            --------------  -----
 0   price             545 non-null    int64
 1   area              545 non-null    int64
```

```
1   area              545 non-null    int64
```

```
# Shape of the dataset
print("Shape:", df.shape)
# Column names
print("Columns:", df.columns.tolist())
# Data types and non-null values
df.info()
# Summary statistics for numeric features
df.describe()
```

```
memory usage: 55.5+ KB
Shape: (545, 13)
Columns: ['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544
Data columns (total 13 columns):
 #   Column            Non-Null Count  Dtype
---  ------            --------------  -----
 0   price             545 non-null    int64
 1   area              545 non-null    int64
 2   bedrooms          545 non-null    int64
 3   bathrooms         545 non-null    int64
 4   stories           545 non-null    int64
 5   mainroad          545 non-null    object
 6   guestroom         545 non-null    object
 7   basement          545 non-null    object
 8   hotwaterheating   545 non-null    object
 9   airconditioning   545 non-null    object
 10  parking           545 non-null    int64
 11  prefarea          545 non-null    object
 12  furnishingstatus  545 non-null    object
dtypes: int64(6), object(7)
memory usage: 55.5+ KB
```

	price	area	bedrooms	bathrooms	stories	parking
count	5.450000e+02	545.000000	545.000000	545.000000	545.000000	545.000000
mean	4.766729e+06	5150.541284	2.965138	1.286239	1.805505	0.693578
std	1.870440e+06	2170.141023	0.738064	0.502470	0.867492	0.861586
min	1.750000e+06	1650.000000	1.000000	1.000000	1.000000	0.000000
25%	3.430000e+06	3600.000000	2.000000	1.000000	1.000000	0.000000
50%	4.340000e+06	4600.000000	3.000000	1.000000	2.000000	0.000000
75%	5.740000e+06	6360.000000	3.000000	2.000000	2.000000	1.000000
max	1.330000e+07	16200.000000	6.000000	4.000000	4.000000	3.000000

Check for Missing Values and Duplicates

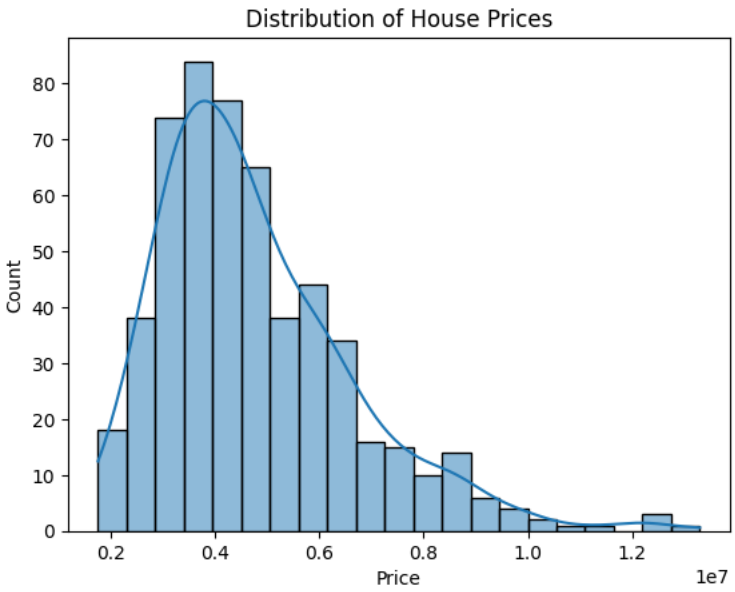
```
# Check for missing values
print(df.isnull().sum())
# Check for duplicates
print("Duplicate rows:", df.duplicated().sum())
```

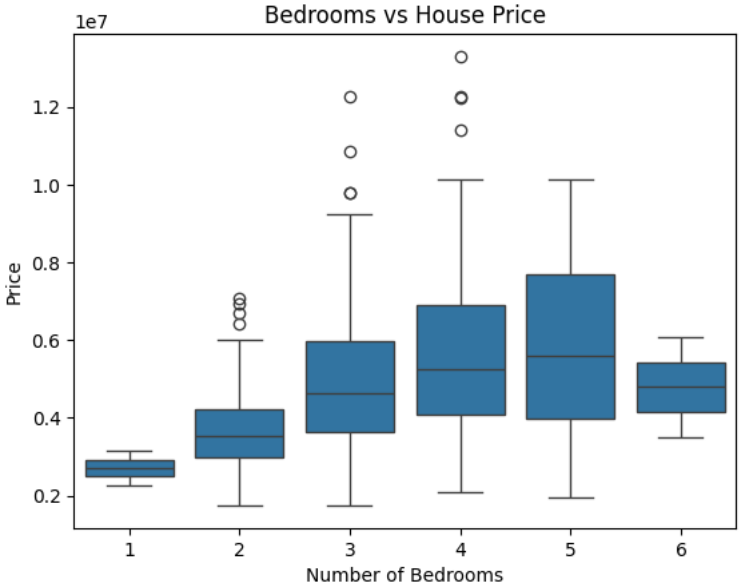
```
price             0
area              0
bedrooms          0
bathrooms         0
stories           0
mainroad          0
guestroom         0
basement          0
hotwaterheating   0
airconditioning   0
parking           0
prefarea          0
furnishingstatus  0
dtype: int64
Duplicate rows: 0
```

Visualize a Few Features

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

```
import matplotlib.pyplot as plt
import pandas as pd

# Distribution of house prices
sns.histplot(df['price'], kde=True)
plt.title('Distribution of House Prices')
plt.xlabel('Price')
plt.show()

# Relationship between number of bedrooms and price
sns.boxplot(x='bedrooms', y='price', data=df)
plt.title('Bedrooms vs House Price')
plt.xlabel('Number of Bedrooms')
plt.ylabel('Price')
plt.show()
```





Identify Target and Features

```
import pandas as pd # Make sure pandas is imported

# ... (Your other code)

target = 'hotwaterheating'

# Reload or recreate the DataFrame if necessary
# df = pd.read_csv('Housing.csv', sep=';')  # Assuming Housing.csv is your data fil
```

```
features = df.columns.drop(target)
```

```
Features: Index(['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad',
       'guestroom', 'basement', 'airconditioning', 'parking', 'prefarea',
       'furnishingstatus'],
      dtype='object')
```

One-Hot Encoding

```
df_encoded = pd.get_dummies('df', drop_first=True)
```

Feature Scaling

```
from sklearn.preprocessing import StandardScaler

# Check that 'price' exists in the DataFrame
# assert 'price' in df_encoded.columns, "'price' column not found in df_encoded"

# Scale the feature columns (excluding the target 'price')
scaler = StandardScaler()
X_scaled = scaler.fit_transform(df_encoded.drop('price', axis=1))

# Extract the target variable
y = df_encoded['price']
```

```
---------------------------------------------------------------------------
KeyError                                  Traceback (most recent call last)
<ipython-input-32-ccdfd5d00768> in <cell line: 0>()
      6 # Scale the feature columns (excluding the target 'price')
      7 scaler = StandardScaler()
----> 8 X_scaled = scaler.fit_transform(df_encoded.drop('price', axis=1))
      9
     10 # Extract the target variable

 3 frames
/usr/local/lib/python3.11/dist-packages/pandas/core/indexes/base.py in drop(self, labels, errors)
   7068         if mask.any():
   7069             if errors != "ignore":
-> 7070                 raise KeyError(f"{labels[mask].tolist()} not found in axis")
   7071             indexer = indexer[~mask]
   7072         return self.delete(indexer)

KeyError: "['price'] not found in axis"
```

Next steps: Explain error

Train-Test Split

```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
# Split data
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)
```

```
---------------------------------------------------------------------------
NameError                                 Traceback (most recent call last)
<ipython-input-33-c0276d08d785> in <cell line: 0>()
      3 from sklearn.metrics import mean_squared_error, r2_score
      4 # Split data
----> 5 X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)

NameError: name 'X_scaled' is not defined
```

Next steps: Explain error

Model Building

```
# Train model
model = LinearRegression()
```

```
model.fit(X_train, y_train)
# Predict
y_pred = model.predict(X_test)
```

```
---------------------------------------------------------------------------
NameError                                 Traceback (most recent call last)
<ipython-input-34-fb6bf145dda5> in <cell line: 0>()
      1 # Train model
      2 model = LinearRegression()
----> 3 model.fit(X_train, y_train)
      4 # Predict
      5 y_pred = model.predict(X_test)

NameError: name 'X_train' is not defined
```

Next steps: Explain error

Evaluation

```
print("MSE:", mean_squared_error(y_test, y_pred))
print("R² Score:", r2_score(y_test, y_pred))
```

```
---------------------------------------------------------------------------
NameError                                 Traceback (most recent call last)
<ipython-input-35-e2d208fc2f54> in <cell line: 0>()
----> 1 print("MSE:", mean_squared_error(y_test, y_pred))
      2 print("R² Score:", r2_score(y_test, y_pred))

NameError: name 'y_test' is not defined
```

Next steps: Explain error

Make Predictions from New Input

```
# Sample input (replace values with any other valid values from the original dataset)
new_student = {
'school': 'GP', # 'GP' or 'MS'
'sex': 'F', # 'F' or 'M'
'age': 17, # Integer
'address': 'U', # 'U' or 'R'
'famsize': 'GT3', # 'LE3' or 'GT3'
'Pstatus': 'A', # 'A' or 'T'
'Medu': 4, # 0 to 4
'Fedu': 3, # 0 to 4
'Mjob': 'health', # 'teacher', 'health', etc.
'Fjob': 'services',
'reason': 'course',
'guardian': 'mother',
'traveltime': 2,
'studytime': 3,
'failures': 0,
'schoolsup': 'yes',
'famsup': 'no',
'paid': 'no',
'activities': 'yes',
'nursery': 'yes',
'higher': 'yes',
'internet': 'yes',
'romantic': 'no',
'famrel': 4,
'freetime': 3,
'goout': 3,
'Dalc': 1,
'Walc': 1,
'health': 4,
'absences': 2,
4/26/25, 12:08 PM sample project.ipynb - Colab
https://colab.research.google.com/drive/1LHSouQeD_tA9J58hn8Q1-yEM77VgZi3U#scrollTo=5BYaJj5jmg8c&printMode=true 9/14
'G1': 14,
'G2': 15
}
```

```
  File "<ipython-input-36-b7b44068fba0>", line 33
    4/26/25, 12:08 PM sample project.ipynb - Colab
                   ^
SyntaxError: leading zeros in decimal integer literals are not permitted; use an 0o prefix for octal integers
```

Next steps: Fix error

Convert to DataFrame and Encode

```
import numpy as np
# Convert to DataFrame
new_df = pd.DataFrame([new_student])
# Combine with original df to match columns
df_temp = pd.concat([df.drop('G3', axis=1), new_df], ignore_index=True)
# One-hot encode
df_temp_encoded = pd.get_dummies(df_temp, drop_first=True)
# Match the encoded feature order
df_temp_encoded = df_temp_encoded.reindex(columns=df_encoded.drop('G3', axis=1).columns, fill_value=0)
# Scale (if you used scaling)
new_input_scaled = scaler.transform(df_temp_encoded.tail(1))
```

```
---------------------------------------------------------------------------
NameError                                 Traceback (most recent call last)
<ipython-input-37-99899f534580> in <cell line: 0>()
      1 import numpy as np
      2 # Convert to DataFrame
----> 3 new_df = pd.DataFrame([new_student])
      4 # Combine with original df to match columns
      5 df_temp = pd.concat([df.drop('G3', axis=1), new_df], ignore_index=True)

NameError: name 'new_student' is not defined
```

Next steps: Explain error

Predict the Final Grade

```
predicted_grade = model.predict(new_input_scaled)
print("  Predicted Final Grade (G3):", round(predicted_grade[0], 2))
```

Deployment-Building an Interactive App

```
!pip install gradio
```

Create a Prediction Function

```
import gradio as gr
```

Create the Gradio Interface

```
inputs = [
gr.Dropdown(['GP', 'MS'], label="School (GP=Gabriel Pereira, MS=Mousinho da Silveira)"),
gr.Dropdown(['M', 'F'], label="Gender (M=Male, F=Female)"),
gr.Number(label="Student Age"),
gr.Dropdown(['U', 'R'], label="Residence Area (U=Urban, R=Rural)"),
gr.Dropdown(['LE3', 'GT3'], label="Family Size (LE3=≤3, GT3=>3 members)"),
gr.Dropdown(['A', 'T'], label="Parent Cohabitation Status (A=Apart, T=Together)"),
gr.Number(label="Mother's Education Level (0-4)"),
gr.Number(label="Father's Education Level (0-4)"),
gr.Dropdown(['teacher', 'health', 'services', 'at_home', 'other'], label="Mother's Job"),
gr.Dropdown(['teacher', 'health', 'services', 'at_home', 'other'], label="Father's Job"),
gr.Dropdown(['home', 'reputation', 'course', 'other'], label="Reason for Choosing School"),
gr.Dropdown(['mother', 'father', 'other'], label="Guardian"),
gr.Number(label="Travel Time to School (1-4)"),
gr.Number(label="Weekly Study Time (1-4)"),
gr.Number(label="Past Class Failures (0-3)"),
gr.Dropdown(['yes', 'no'], label="Extra School Support"),
```

```
gr.Dropdown(['yes', 'no'], label="Family Support"),
gr.Dropdown(['yes', 'no'], label="Extra Paid Classes"),
gr.Dropdown(['yes', 'no'], label="Participates in Activities"),
gr.Dropdown(['yes', 'no'], label="Attended Nursery"),
gr.Dropdown(['yes', 'no'], label="Aspires Higher Education"),
gr.Dropdown(['yes', 'no'], label="Internet Access at Home"),
gr.Dropdown(['yes', 'no'], label="Currently in a Relationship"),
gr.Number(label="Family Relationship Quality (1-5)"),
gr.Number(label="Free Time After School (1-5)"),
gr.Number(label="Going Out Frequency (1-5)"),
gr.Number(label="Workday Alcohol Consumption (1-5)"),
4/26/25, 12:08 PM sample project.ipynb - Colab
https://colab.research.google.com/drive/1LHSouQeD_tA9J58hn8Q1-yEM77VgZi3U#scrollTo=5BYaJj5jmg8c&printMode=true 13/14
It looks like you are running Gradio on a hosted a Jupyter notebook. For the Gradio app to work, sharing must be enabled. Automatically se
Colab notebook detected. To show errors in colab notebook, set debug=True in launch()
* Running on public URL: https://37518063c688a89403.gradio.live
This share link expires in 1 week. For free permanent hosting and GPU upgrades, run `gradio deploy` from the terminal in the working direc
 Student Performance Predictor
Enter academic and demographic info to predict the final grade (G3) of a student.
GP 0
School (GP=Gabriel Pereira, MS=Mousinho da Silveira)  Predicted Final Grade (G3)
gr.Number(label="Weekend Alcohol Consumption (1-5)"),
gr.Number(label="Health Status (1=Very Bad to 5=Excellent)"),
gr.Number(label="Number of Absences"),
gr.Number(label="Grade in 1st Period (G1: 0-20)"),
gr.Number(label="Grade in 2nd Period (G2: 0-20)")
1
```

Upload the Dataset

```
from google.colab import files
uploaded = files.upload()
```

Choose Files Housing.csv.xlsx

- **Housing.csv.xlsx**(application/vnd.openxmlformats-officedocument.spreadsheetml.sheet) - 40874 bytes, last modified: 5/8/2025 - 100% done

```
Saving Housing.csv.xlsx to Housing.csv (6).xlsx
```

Load the Dataset

```
import pandas as pd

# Load the Excel file
df = pd.read_excel('/Housing.csv.xlsx')

# Preview the data
df.head()
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furn
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	
4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	

Next steps: Generate code with df | View recommended plots | New interactive sheet

Data Exploration

```
# Dataset shape
print("Shape of the dataset:", df.shape)

# Column names
print("Columns:", df.columns.tolist())

# Info about data types and missing value
df.info()
```

```
# Summary statistics
df.describe()
```

```
Shape of the dataset: (545, 13)
Columns: ['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning
<class 'pandas.core.frame.DataFrame'
RangeIndex: 545 entries, 0 to 544
Data columns (total 13 columns):
 #   Column            Non-Null Cour   Dtype
---  ------            -------------   -----
 0   price             545 non-null    int64
 1   area              545 non-null    int64
 2   bedrooms          545 non-null    int64
 3   bathrooms         545 non-null    int64
 4   stories           545 non-null    int64
 5   mainroad          545 non-null    object
 6   guestroom         545 non-null    object
 7   basement          545 non-null    object
 8   hotwaterheating   545 non-null    object
 9   airconditioning   545 non-null    object
 10  parking           545 non-null    int64
 11  prefarea          545 non-null    object
 12  furnishingstatus  545 non-null    object
dtypes: int64(6), object(7)
memory usage: 55.5+ KB
```

	price	area	bedrooms	bathrooms	stories	parking
count	5.450000e+02	545.000000	545.000000	545.000000	545.000000	545.000000
mean	4.766729e+06	5150.541284	2.965138	1.286239	1.805505	0.693578
std	1.870440e+06	2170.141023	0.738064	0.502470	0.867492	0.861586
min	1.750000e+06	1650.000000	1.000000	1.000000	1.000000	0.000000
25%	3.430000e+06	3600.000000	2.000000	1.000000	1.000000	0.000000
50%	4.340000e+06	4600.000000	3.000000	1.000000	2.000000	0.000000
75%	5.740000e+06	6360.000000	3.000000	2.000000	2.000000	1.000000
max	1.330000e+07	16200.000000	6.000000	4.000000	4.000000	3.000000

Data Cleaning

```
# Check for missing values
print("Missing values:\n", df.isnull().sum())

# Check for duplicates
print("Duplicate rows:", df.duplicated().sum())
```

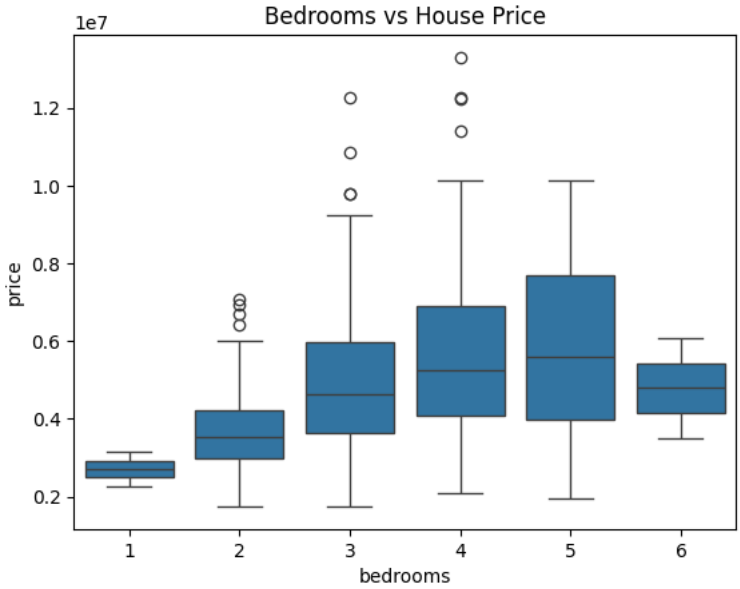
```
Missing values:
 price               0
area                0
bedrooms            0
bathrooms           0
stories             0
mainroad            0
guestroom           0
basement            0
hotwaterheating     0
airconditioning     0
parking             0
prefarea            0
furnishingstatus    0
dtype: int64
Duplicate rows: 0
```

Data Visualization (Modify column names as per your dataset)

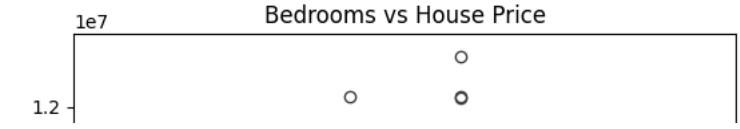
```
import seaborn as sns
import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib.pyplot as plt

# Example: Distribution of target column (e.g., price)
sns.histplot(df['price'], kde=True)
```

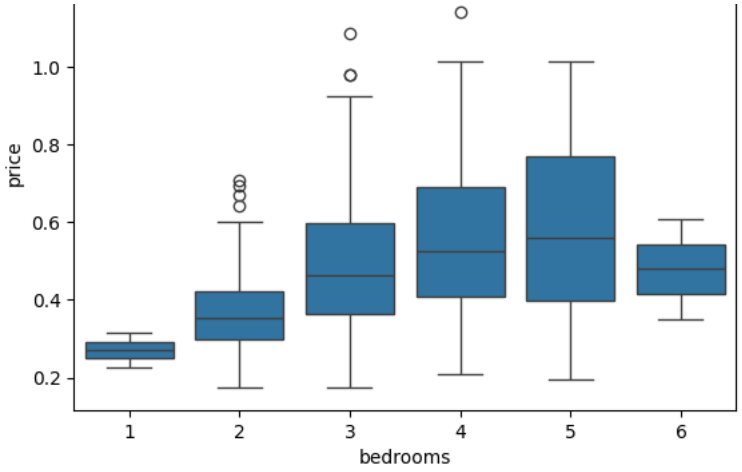
```
plt.title('Distribution of House Prices')
plt.xlabel('Price')
plt.show()

# Example: Boxplot of a feature vs price (adjust columns)
sns.boxplot(x='bedrooms', y='price', data=df)
plt.title('Bedrooms vs House Price')
plt.show()

# Example: Distribution of target column (e.g., price)
sns.histplot(df['price'], kde=True)
plt.title('Distribution of House Prices')
plt.xlabel('Price')
plt.show()

# Example: Boxplot of a feature vs price (adjust columns)
sns.boxplot(x='bedrooms', y='price', data=df)
plt.title('Bedrooms vs House Price')
plt.show()
```


Distribution of House Prices
Count
Price
1e7


Bedrooms vs House Price
1e7
price
bedrooms


Distribution of House Prices
Count
Price
1e7


Bedrooms vs House Price
1e7



Feature Selection and Target Definition

```
# Choose target and features
target = 'price'
features = df.columns.drop(target)

print("Target:", target)
print("Features:", features)
```

```
Target: price
Features: Index(['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad', 'guestroom',
       'basement', 'hotwaterheating', 'airconditioning', 'parking', 'prefarea',
       'furnishingstatus'],
      dtype='object')
```

Encoding Categorical Variables

```
# Identify categorical columns
categorical_cols = df.select_dtypes(include='object').columns
print("Categorical columns:", categorical_cols.tolist())

# Apply one-hot encoding
df_encoded = pd.get_dummies(df, drop_first=True)
```

```
Categorical columns: ['mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning', 'prefarea', 'furnishingstatus']
```

Feature Scaling

```
from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()
X_scaled = scaler.fit_transform(df_encoded.drop(target, axis=1))
y = df_encoded[target]
```

Train-Test Split

```
from sklearn.model_selection import train_test_split

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

Model Building (Linear Regression)

```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_scor

model = LinearRegression()
```

```
model.fit(X_train, y_train)

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

Model Evaluation

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

```
Mean Squared Error: 1754318687330.6675
R² Score: 0.6529242642153177
```

Predict from New Input

```
# Example: Replace with values from your actual dataset structure
new_data = {
    'area': 1500,
    'bedrooms': 3,
    'bathrooms': 2,
    'stories': 2,
    'mainroad': 'yes',
    'guestroom': 'no',
    'basement': 'yes',
    'hotwaterheating': 'no',
    'airconditioning': 'yes',
    'parking': 1,
    'prefarea': 'yes',
    'furnishingstatus': 'semi-furnished'
}

# Convert to DataFrame
new_df = pd.DataFrame([new_data])

# Combine with original for consistent encoding
temp_df = pd.concat([df.drop(columns=[target]), new_df], ignore_index=True)

# One-hot encode
temp_encoded = pd.get_dummies(temp_df, drop_first=True)

# Align with training features
temp_encoded = temp_encoded.reindex(columns=df_encoded.drop(target, axis=1).columns, fill_value=0)

# Scale
new_scaled = scaler.transform(temp_encoded.tail(1))

# Predict
prediction = model.predict(new_scaled)
print(f"  Predicted House Price:  {prediction[0]:,.2f}")
```

```
  Predicted House Price:  6,125,612.95
```

Optional: Try Another Model (Random Forest)

```
from sklearn.ensemble import RandomForestRegressor

rf_model = RandomForestRegressor(random_state=42)
rf_model.fit(X_train, y_train)

rf_pred = rf_model.predict(X_test)

print("Random Forest MSE:", mean_squared_error(y_test, rf_pred))
print("Random Forest R²:", r2_score(y_test, rf_pred))
```

```
Random Forest MSE: 1959406221695.9854
Random Forest R²: 0.6123495913214113
```

Deploy a Gradio App

```
!pip install gradio

import gradio as gr

def predict_house_price(area, bedrooms, bathrooms, stories, mainroad, guestroom,
                        basement, hotwaterheating, airconditioning, parking,
                        prefarea, furnishingstatus):

    input_dict = {
        'area': area,
        'bedrooms': bedrooms,
        'bathrooms': bathrooms,
        'stories': stories,
        'mainroad': mainroad,
        'guestroom': guestroom,
        'basement': basement,
        'hotwaterheating': hotwaterheating,
        'airconditioning': airconditioning,
        'parking': parking,
        'prefarea': prefarea,
        'furnishingstatus': furnishingstatus
    }

    input_df = pd.DataFrame([input_dict])
    combined_df = pd.concat([df.drop(columns=[target]), input_df], ignore_index=True)
    encoded_df = pd.get_dummies(combined_df, drop_first=True)
    encoded_df = encoded_df.reindex(columns=df_encoded.drop(target, axis=1).columns, fill_value=0)

    scaled_input = scaler.transform(encoded_df.tail(1))
    result = model.predict(scaled_input)[0]

    return f" {result:,.2f}"

# Define input fields
inputs = [
    gr.Number(label="Area (sq ft)"),
    gr.Number(label="Bedrooms"),
    gr.Number(label="Bathrooms"),
    gr.Number(label="Stories"),
    gr.Dropdown(['yes', 'no'], label="Main Road Access"),
    gr.Dropdown(['yes', 'no'], label="Guest Room"),
    gr.Dropdown(['yes', 'no'], label="Basement"),
    gr.Dropdown(['yes', 'no'], label="Hot Water Heating"),
    gr.Dropdown(['yes', 'no'], label="Air Conditioning"),
    gr.Number(label="Parking Spots"),
    gr.Dropdown(['yes', 'no'], label="Preferred Area"),
    gr.Dropdown(['furnished', 'semi-furnished', 'unfurnished'], label="Furnishing Status")
]

# Output field
output = gr.Text(label="Predicted House Price")

# Launch interface
gr.Interface(fn=predict_house_price, inputs=inputs, outputs=output,
             title="  House Price Predictor",
             description="Enter housing details to estimate the price.").launch()
```

```
Requirement already satisfied: gradio in /usr/local/lib/python3.11/dist-packages (5.29.1)
Requirement already satisfied: aiofiles<25.0,>=22.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (24.1.0)
Requirement already satisfied: anyio<5.0,>=3.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (4.9.0)
Requirement already satisfied: fastapi<1.0,>=0.115.2 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.115.12)
Requirement already satisfied: ffmpy in /usr/local/lib/python3.11/dist-packages (from gradio) (0.5.0)
Requirement already satisfied: gradio-client==1.10.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (1.10.1)
Requirement already satisfied: groovy~=0.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.1.2)
Requirement already satisfied: httpx>=0.24.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.28.1)
Requirement already satisfied: huggingface-hub>=0.28.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.31.1)
Requirement already satisfied: jinja2<4.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.1.6)
Requirement already satisfied: markupsafe<4.0,>=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.0.2)
Requirement already satisfied: numpy<3.0,>=1.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.0.2)
Requirement already satisfied: orjson~=3.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.10.18)
Requirement already satisfied: packaging in /usr/local/lib/python3.11/dist-packages (from gradio) (24.2)
Requirement already satisfied: pandas<3.0,>=1.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.2.2)
Requirement already satisfied: pillow<12.0,>=8.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (11.2.1)
Requirement already satisfied: pydantic<2.12,>=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.11.4)
Requirement already satisfied: pydub in /usr/local/lib/python3.11/dist-packages (from gradio) (0.25.1)
Requirement already satisfied: python-multipart>=0.0.18 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.0.20)
Requirement already satisfied: pyyaml<7.0,>=5.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (6.0.2)
Requirement already satisfied: ruff>=0.9.3 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.11.9)
Requirement already satisfied: safehttpx<0.2.0,>=0.1.6 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.1.6)
Requirement already satisfied: semantic-version~=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.10.0)
Requirement already satisfied: starlette<1.0,>=0.40.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.46.2)
Requirement already satisfied: tomlkit<0.14.0,>=0.12.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.13.2)
Requirement already satisfied: typer<1.0,>=0.12 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.15.3)
Requirement already satisfied: typing-extensions~=4.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (4.13.2)
Requirement already satisfied: uvicorn>=0.14.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.34.2)
Requirement already satisfied: fsspec in /usr/local/lib/python3.11/dist-packages (from gradio-client==1.10.1->gradio) (2025.3.2)
Requirement already satisfied: websockets<16.0,>=10.0 in /usr/local/lib/python3.11/dist-packages (from gradio-client==1.10.1->gradio) (1
Requirement already satisfied: idna>=2.8 in /usr/local/lib/python3.11/dist-packages (from anyio<5.0,>=3.0->gradio) (3.10)
Requirement already satisfied: sniffio>=1.1 in /usr/local/lib/python3.11/dist-packages (from anyio<5.0,>=3.0->gradio) (1.3.1)
Requirement already satisfied: certifi in /usr/local/lib/python3.11/dist-packages (from httpx>=0.24.1->gradio) (2025.4.26)
Requirement already satisfied: httpcore==1.* in /usr/local/lib/python3.11/dist-packages (from httpx>=0.24.1->gradio) (1.0.9)
Requirement already satisfied: h11>=0.16 in /usr/local/lib/python3.11/dist-packages (from httpcore==1.*->httpx>=0.24.1->gradio) (0.16.0)
Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (3.18.0)
Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (2.32.3)
Requirement already satisfied: tqdm>=4.42.1 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (4.67.1)
Requirement already satisfied: hf-xet<2.0.0,>=1.1.0 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (1
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->gradio) (2.9.0.
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->gradio) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->gradio) (2025.2)
Requirement already satisfied: annotated-types>=0.6.0 in /usr/local/lib/python3.11/dist-packages (from pydantic<2.12,>=2.0->gradio) (0.7
Requirement already satisfied: pydantic-core==2.33.2 in /usr/local/lib/python3.11/dist-packages (from pydantic<2.12,>=2.0->gradio) (2.33
```