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

Choose Files No file chosen enable.

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Saving Housing csv to Housing csv

import pandas as pd
df = pd.read\_csv('Housing.csv')
df.head()

<b>→</b>		price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea
	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

```
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', 'airconditior <class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544

Data columns (total 13 columns):

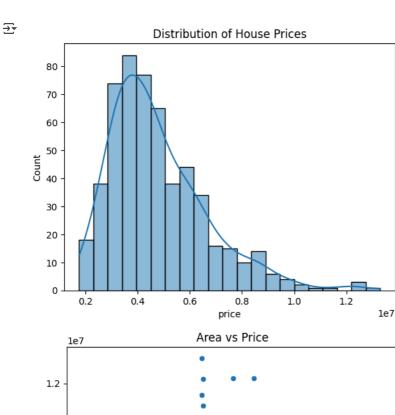
Ducu	coramiis (cocar is	coramiis).	
#	Column	Non-Null Count	Dtype
0	price	545 non-null	int64
1	area	545 non-null	int64
2	bedrooms	545 non-null	int64
3	bathrooms	545 non-null	int64
4	stories	545 non-null	int64
5	mainroad	545 non-null	object
6	guestroom	545 non-null	object
7	basement	545 non-null	object
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

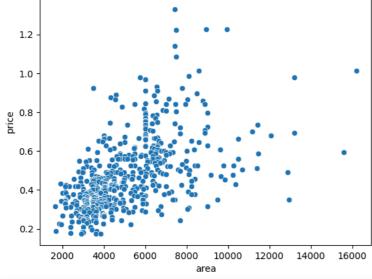
dtypes: int64(6), object(7)
memory usage: 55.5+ KB

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

```
print("Missing Values:\n", df.isnull().sum())
print("Duplicate Rows:", df.duplicated().sum())
```

→ Missing Values: price 0 area 0 bedrooms 0 bathrooms 0 0 stories mainroad 0 guestroom a basement 0 hotwaterheating 0 airconditioning 0





categorical\_cols = df.select\_dtypes(include=['object']).columns
df\_encoded = pd.get\_dummies(df, drop\_first=True)

```
from sklearn.preprocessing import StandardScaler

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

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)
```

```
from sklearn.linear_model import LinearRegression
model = LinearRegression()
model.fit(X_train, y_train)
₹
      ▼ LinearRegression ① ?
     LinearRegression()
from sklearn.metrics import mean_squared_error, r2_score
y_pred = model.predict(X_test)
print("MSE:", mean_squared_error(y_test, y_pred))
print("R2 Score:", r2_score(y_test, y_pred))
→ MSE: 1754318687330.6675
     R<sup>2</sup> Score: 0.6529242642153177
# Replace with actual column names from your dataset
new house = {
    'area': 3000,
    'bedrooms': 3,
    'bathrooms': 2,
    'stories': 2,
    'mainroad': 'yes',
    'guestroom': 'no',
    'basement': 'yes',
'hotwaterheating': 'no',
    'airconditioning': 'yes',
    'parking': 1,
    'prefarea': 'no',
    'furnishingstatus': 'furnished'
new_df = pd.DataFrame([new_house])
df_temp = pd.concat([df.drop('price', axis=1), new_df], ignore_index=True)
df_temp_encoded = pd.get_dummies(df_temp, drop_first=True)
\label{eq:df_emp_encoded} \texttt{df\_temp\_encoded.reindex(columns=df\_encoded.drop('price', axis=1).columns, fill\_value=0)}
new_input_scaled = scaler.transform(df_temp_encoded.tail(1))
predicted_price = model.predict(new_input_scaled)
print(" § Predicted House Price:", round(predicted_price[0], 2))

→ S Predicted House Price: 5976557.41
!pip install gradio
Collecting semantic-version~=2.0 (from gradio)
       Downloading semantic_version-2.10.0-py2.py3-none-any.whl.metadata (9.7 kB)
     Collecting starlette<1.0,>=0.40.0 (from gradio)
```

```
Downloading gradio-5.30.0-py3-none-any.whl (54.2 MB)
                                                  54.2/54.2 MB 16.4 MB/s eta 0:00:00
     Downloading gradio_client-1.10.1-py3-none-any.whl (323 kB)
                                                  323.1/323.1 kB 18.5 MB/s eta 0:00:00
     Downloading aiofiles-24.1.0-py3-none-any.whl (15 kB)
     Downloading fastapi-0.115.12-py3-none-any.whl (95 kB)
                                                 - 95.2/95.2 kB <mark>5.2 MB/s</mark> eta 0:00:00
     Downloading groovy-0.1.2-py3-none-any.whl (14 kB)
     Downloading python_multipart-0.0.20-py3-none-any.whl (24 kB)
     \label{lownloading ruff-0.11.10-py3-none-manylinux_2_17_x86_64. manylinux 2014_x86_64. whl (11.6 MB) \\
                                                  11.6/11.6 MB 73.5 MB/s eta 0:00:00
     Downloading safehttpx-0.1.6-py3-none-any.whl (8.7 kB)
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     Downloading starlette-0.46.2-py3-none-any.whl (72 kB)
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     Downloading tomlkit-0.13.2-py3-none-any.whl (37 kB)
     Downloading uvicorn-0.34.2-py3-none-any.whl (62 kB)
                                                 - 62.5/62.5 kB 5.8 MB/s eta 0:00:00
     Downloading ffmpy-0.5.0-py3-none-any.whl (6.0 kB)
     Downloading pydub-0.25.1-py2.py3-none-any.whl (32 kB)
     Installing collected packages: pydub, uvicorn, tomlkit, semantic-version, ruff, python-multipart, groovy, ffmpy, aiofiles, starlet
     Successfully installed aiofiles-24.1.0 fastapi-0.115.12 ffmpy-0.5.0 gradio-5.30.0 gradio-client-1.10.1 groovy-0.1.2 pydub-0.25.1 r
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import StandardScaler
import gradio as gr
# Load the data
df = pd.read_csv("/content/drive/MyDrive/Housing.csv")
# One-hot encoding
df_encoded = pd.get_dummies(df, drop_first=True)
# Features and target
X = df_{encoded.drop("price", axis=1)}
y = df_encoded["price"]
# Scale features
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
# Train-test split
 \textbf{X\_train, X\_test, y\_train, y\_test = train\_test\_split(X\_scaled, y, test\_size=0.2, random\_state=42) } 
# Train model
model = LinearRegression()
model.fit(X_train, y_train)
# Save column structure
columns = X.columns
# Define prediction function
def predict_price(area, bedrooms, bathrooms, stories, mainroad, guestroom, basement,
                  hotwaterheating, airconditioning, parking, prefarea, furnishingstatus):
    # Form input dict
    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
    }
    # Convert to DataFrame and encode
    input_df = pd.DataFrame([input_dict])
    temp_df = pd.concat([df.drop("price", axis=1), input_df], ignore_index=True)
    temp_encoded = pd.get_dummies(temp_df, drop_first=True)
    temp_encoded = temp_encoded.reindex(columns=columns, fill_value=0)
    # Scale and predict
    input_scaled = scaler.transform(temp_encoded.tail(1))
    prediction = model.predict(input_scaled)[0]
    return round(prediction, 2)
```

```
# Define interface inputs
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 Spaces"),
    gr.Dropdown(["yes", "no"], label="Preferred Area"),
gr.Dropdown(["furnished", "semi-furnished", "unfurnished"], label="Furnishing Status")
# Create the interface
gr.Interface(
    fn=predict_price,
     inputs=inputs,
    outputs=gr.Number(label="Predicted House Price"),
     title="♠ House Price Prediction App",
     description="Enter the house features to estimate its price."
).launch()
```

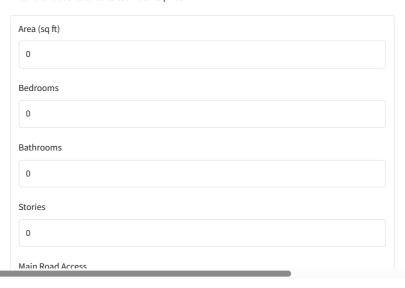
🚁 It looks like you are running Gradio on a hosted a Jupyter notebook. For the Gradio app to work, sharing must be enabled. Automatica

Colab notebook detected. To show errors in colab notebook, set debug=True in launch() \* Running on public URL: <a href="https://025ee1efbdcda63a76.gradio.live">https://025ee1efbdcda63a76.gradio.live</a>

This share link expires in 1 week. For free permanent hosting and GPU upgrades, run `gradio deploy` from the terminal in the working

## **★** House Price Prediction App

Enter the house features to estimate its price.



Start coding or generate with AI.