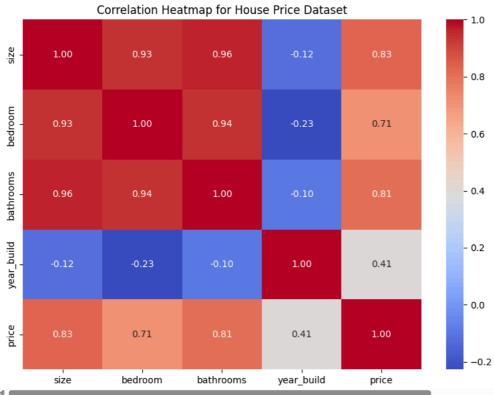
```
from google.colab import files
uploaded = files.upload()
Ð₹
     Choose Files No file chosen
                                        Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
import pandas as pd
# Load the uploaded dataset
df = pd.read_csv("house_price.csv")
df.head()
print("shape:",df.shape)
print("columns:",df.columns.tolist())
df.info()
df.describe()
→ shape: (20, 6)
     columns: ['size', 'bedroom', 'bathrooms', 'location', 'year_build', 'price']
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 20 entries, 0 to 19
     Data columns (total 6 columns):
      # Column
                      Non-Null Count Dtype
      0
          size
                      20 non-null
                                       int64
          bedroom
                      20 non-null
                                       int64
      1
          bathrooms
                      20 non-null
                                       float64
      2
      3
          location
                      20 non-null
                                       object
          year_build 20 non-null
                                       int64
          price
                      20 non-null
                                       int64
     dtypes: float64(1), int64(4), object(1)
     memory usage: 1.1+ KB
                    size bedroom bathrooms
                                               year build
                                                                   price
      count
               20.000000 20.00000
                                    20.000000
                                                 20.000000
                                                                20.000000
                           2.85000
            1615.000000
                                     1.950000 1999.200000 247750.000000
      mean
              328.913682
                           0.74516
                                     0.666886
                                                 13.213231
                                                             54179.890424
       std
       min
             1100.000000
                           2.00000
                                     1.000000 1970.000000 170000.000000
      25%
             1337.500000
                           2.00000
                                     1.500000
                                              1993.750000
                                                           203750.000000
      50%
             1625.000000
                           3.00000
                                     2.000000 2002.500000 245000.000000
             1862 500000
                           3 00000
                                     2 500000 2008 250000 292500 000000
      75%
      max
             2200.000000
                           4.00000
                                     3.000000 2018.000000 340000.000000
     4 =
# Check missing values print(df.isnull().sum())
df_cleaned = df.dropna() # removes rows with missing values
print(df cleaned)
<del>_</del>
         size
               bedroom bathrooms
                                       location year_build
                                                              price
     0
         1500
                     3
                              2.0
                                       suburb A
                                                       1995
                                                             225000
     1
         1200
                              1.0
                                   city center
                                                             180000
         2000
                               2.5
                                     Rural area
                                                        1980
                                                              250000
     3
         1800
                     3
                              2.0
                                      suburb B
                                                       2010
                                                             310000
     4
         1400
                              1.5 city center
                                                       1998
                                                              210000
                     2
     5
         2200
                     4
                              3.0
                                    Rural area
                                                       1975
                                                              280000
         1600
                                                       2005
                                                              240000
     6
                     3
                              2.0
                                       suburb A
                     2
                                                       1990
                                                              170000
         1100
                              1.0
                                       suburb C
                                                              330000
     8
         1900
                     3
                              2.5
                                    Rural area
                                                       2015
     9
         1700
                     3
                              2.0
                                       suburb B
                                                       2008
                                                              290000
     10
         1300
                     2
                              1.5
                                   city center
                                                        2001
                                                              200000
     11
         2100
                     4
                              3.0
                                       suburb A
                                                       1970
                                                              260000
     12
         1550
                              2.0
                                       suburb C
                                                       1997
                                                              230000
                                                        2004
     13
         1250
                              1.0
                                     Rural area
                                                              190000
     14
        1850
                     3
                              2.5
                                       suburb B
                                                        2012
                                                              320000
     15
         1650
                                                        2007
                                                              250000
                     3
                              2.0
                                   city center
                                                              175000
         1150
                                                        1985
     16
                              1.0
                                       suburb A
                                                              340000
     17
         1950
                     4
                              3.0
                                     Rural area
                                                        2018
                                                              300000
     18
        1750
                     3
                              2.0
                                       suburb C
                                                        2009
     19
        1350
                              1.5
                                       suburb B
                                                        2003
                                                             205000
import pandas as pd
data = {
        "Size (sqft)" : [1500,2000,1200,1800,1000,2200,1600],
        "Bedrooms "
                      : [ 3,4,2,3,2,4,3],
        "Bathrooms"
                      : [2,3,1,2,1,3,2],
```

```
"Location"
                      :["suburb","city","suburb","city","rural","suburb","City"],
        "Year Built" : [2005,2010,1995,2008,1980,2015,2012],
        "Price"
                      : [250000,400000,180000,350000,120000,420000,300000]
   }
df = pd.DataFrame(data)
print(df)
₹
        Size (sqft) Bedrooms
                                Bathrooms Location Year Built
                                                                 Price
                                                          2005 250000
               1500
                             3
                                        2
                                            suburb
     1
               2000
                             4
                                        3
                                             city
                                                          2010 400000
     2
               1200
                             2
                                        1
                                            suburb
                                                          1995
                                                                180000
                                             city
     3
               1800
                                                          2008 350000
     4
               1000
                                                          1980
                                                                120000
                                        1
                                             rural
     5
               2200
                                        3
                                            suburb
                                                          2015 420000
                                                          2012 300000
               1600
                                              Citv
     6
                             3
high Price = df[df["Price"] > 120000]
print(high_Price)
<del>___</del>
        Size (sqft) Bedrooms
                                Bathrooms Location Year Built
                                                                 Price
                                                          2005 250000
               1500
                             3
                                       2
                                            suburb
     1
               2000
                             4
                                        3
                                              city
                                                          2010 400000
     2
               1200
                             2
                                        1
                                            suburb
                                                          1995
                                                                180000
     3
               1800
                             3
                                        2
                                              city
                                                          2008 350000
     5
               2200
                             4
                                        3
                                            suburb
                                                          2015
                                                                420000
                                                          2012 300000
                                              City
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df["Location"] = le.fit_transform(df["Location"])
print(df)
                               Bathrooms Location Year Built
₹
       Size (sqft) Bedrooms
                                                                 Price
                                                           2005 250000
     a
              1500
                             3
                                       2
                                                  3
                                                                 400000
     1
               2000
                             4
                                        3
                                                  1
                                                           2010
     2
               1200
                             2
                                        1
                                                  3
                                                           1995
                                                                 180000
     3
               1800
                                        2
                                                           2008
                                                                 350000
     4
               1000
                             2
                                        1
                                                  2
                                                           1980
                                                                 120000
               2200
                                        3
                                                           2015
     6
               1600
                                                           2012
                                                                 300000
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
# Load the dataset
df = pd.read_csv('house_price.csv')
# Clean column names
df.columns = df.columns.str.strip().str.lower() # Standardize names
# Display column names to verify
print("Columns in dataset:", df.columns.tolist())
# Select only numeric columns for correlation
numeric_df = df.select_dtypes(include=['int64', 'float64'])
# Check if there are enough numerical columns
if numeric_df.shape[1] < 2:</pre>
   print("Not enough numeric columns to generate a correlation heatmap.")
else:
   # Generate correlation matrix
   correlation_matrix = numeric_df.corr()
   # Plot heatmap
   plt.figure(figsize=(8, 6))
    sns.heatmap(correlation_matrix, annot=True, cmap="coolwarm", fmt=".2f")
   plt.title("Correlation Heatmap for House Price Dataset")
    plt.tight_layout()
    plt.show()
```

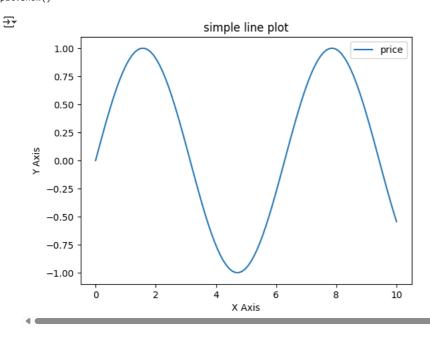
Tolumns in dataset: ['size', 'bedroom', 'bathrooms', 'location', 'year_build', 'price']



```
import matplotlib.pyplot as plt
import numpy as np

x = np.linspace(0, 10, 100)
y = np.sin(x)

plt.plot(x,y,label="price")
plt.xlabel("X Axis")
plt.ylabel("Y Axis")
plt.title("simple line plot")
plt.legend()
plt.show()
```



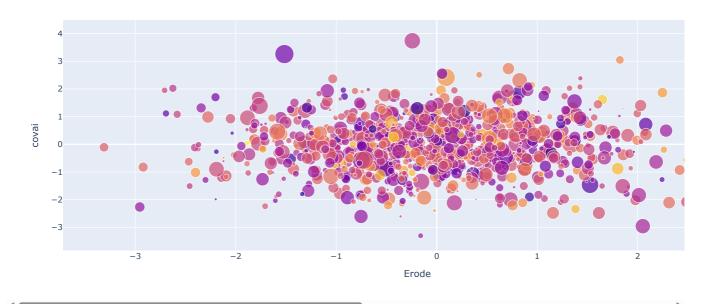
```
import pandas as pd

Data = {
    "Location": ["Erode", "covai", "Madurai", "salem"],
    "price": [42000, 30000, 35000, 18000]
}
df = pd.DataFrame(Data) # Corrected: 'DataFrame', not 'DataFrame'

def performance_category(price):
    if price >= 42000:
```

```
return "High" # Corrected: Use standard quotes
   elif price >= 30000:
       return "medium" # Corrected: Use standard quotes, consistent indentation
   else:
       return "low" # Corrected: Use standard quotes, consistent indentation
₹
      Location price performance
    a
        Erode
               42000
                           High
    1
         covai
               30000
                          medium
      Madurai 35000
                         medium
        salem 18000
                            low
!pip install plotly
import plotly.express as px
import pandas as pd
import numpy as np
# Creating sample data
data = np.random.randn(1000, 4)
# Create the DataFrame
df = pd.DataFrame(data, columns=["Erode", "covai", "Madurai", "salem"])
# Ensure 'Madurai' column (used for size) has only positive values
df['Madurai'] = abs(df['Madurai']) # Taking the absolute value
# Create the bubble chart
fig = px.scatter(df, x="Erode", y="covai", size="Madurai",
               color="salem", # Use 'salem' for color variation
               hover_name="Erode", # Display 'Erode' on hover
               title="Rooms Available")
fig.show()
Requirement already satisfied: plotly in /usr/local/lib/python3.11/dist-packages (5.24.1)
    Requirement already satisfied: tenacity>=6.2.0 in /usr/local/lib/python3.11/dist-packages (from plotly) (9.1.2)
    Requirement already satisfied: packaging in /usr/local/lib/python3.11/dist-packages (from plotly) (24.2)
```

Rooms Available



```
# prompt: create a high available house list
import plotly.express as px

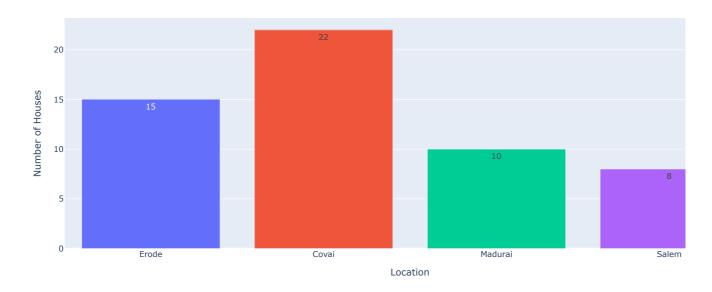
# Sample data (replace with your actual data)
data = {
    "Location": ["Erode", "Covai", "Madurai", "Salem"],
    "Available Houses": [15, 22, 10, 8] # Number of houses available in each location
}

df = pd.DataFrame(data)

# Create the bar chart
fig = px.bar(df, x="Location", y="Available Houses",
```



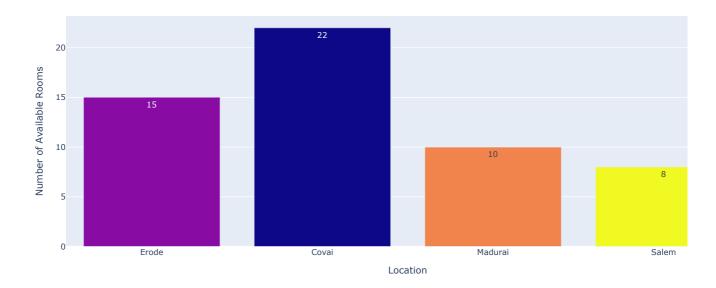
House Availability by Location



```
# prompt: create a low cost available rooms
import pandas as pd
import plotly.express as px
# Sample data (replace with your actual data)
    "Location": ["Erode", "Covai", "Madurai", "Salem"],
    "Available Rooms": [15, 22, 10, 8], # Number of rooms available
    "Price": [1000, 1200, 800, 700] # Price per room
df = pd.DataFrame(data)
# Calculate affordability score (example: lower price = higher score)
df['Affordability'] = 1 / df['Price']
# Create the bar chart with affordability as color
fig = px.bar(df, x="Location", y="Available Rooms",
            title="Low-Cost Available Rooms by Location",
            color="Affordability", # Color bars by affordability
            text="Available Rooms",
            hover_data=["Price"] # Show price on hover
fig.update_layout(xaxis_title="Location", yaxis_title="Number of Available Rooms")
fig.show()
```



Low-Cost Available Rooms by Location



```
# prompt: create a high cost available rooms list
# Sample data (replace with your actual data)
data = {
    "Location": ["Erode", "Covai", "Madurai", "Salem"],
    "Available Rooms": [15, 22, 10, 8], # Number of rooms available
    "Price": [1000, 1200, 800, 700] # Price per room
}
df = pd.DataFrame(data)
# Define a threshold for "high cost"
high_cost_threshold = 900 # Example: Rooms with price above 900 are high cost
# Filter for high-cost available rooms
high_cost_rooms = df[df["Price"] > high_cost_threshold]
# Create the bar chart for high-cost rooms
fig = px.bar(high_cost_rooms, x="Location", y="Available Rooms",
            title="High-Cost Available Rooms by Location",
             color="Price", # Color bars by price
             text="Available Rooms",
            hover_data=["Price"] # Show price on hover
            )
fig.update_layout(xaxis_title="Location", yaxis_title="Number of Available Rooms")
fig.show()
```



High-Cost Available Rooms by Location



```
# prompt: create a lexgury rooms hotel names
import pandas as pd
# Sample data (replace with your actual data)
data = {
    "Room Type": ["Presidential Suite", "Luxury Suite", "Deluxe Room", "Executive Suite", "Royal Suite"],
    "Price": [2000, 1500, 1000, 1200, 1800], # Price per night
    "Features": ["Ocean view, private pool", "City view, jacuzzi", "Comfortable bed, balcony", "Business amenities, large workspace", "!
}
luxury_hotel_df = pd.DataFrame(data)
# Print the DataFrame
luxury_hotel_df
```

_				
₹		Room Type	Price	Features
	0	Presidential Suite	2000	Ocean view, private pool
	1	Luxury Suite	1500	City view, jacuzzi
	2	Deluxe Room	1000	Comfortable bed, balcony
	3	Executive Suite	1200	Business amenities, large workspace
	4	Royal Suite	1800	Spacious living area, butler service

```
# prompt: create a lower cost and foods provider
# Sample data (replace with your actual data)
data = {
    "Food Provider": ["Provider A", "Provider B", "Provider C", "Provider D"], \[ \]
    "Average Cost per Meal": [8, 6, 10, 7], # Average cost of a meal "Customer Rating": [4.2, 4.5, 3.8, 4.0], # Customer rating out of 5
    "Delivery Time (minutes)": [30, 25, 40, 35] # Average delivery time
}
food_providers_df = pd.DataFrame(data)
# Calculate a composite score (example: higher rating, lower cost, faster delivery = higher score)
food_providers_df['Composite Score'] = (
    food_providers_df['Customer Rating'] * 0.5 + # Weighting for customer rating
    (1 / food_providers_df['Average Cost per Meal']) * 0.3 + \# Weighting for cost (inverse)
    (1 / food_providers_df['Delivery Time (minutes)']) * 0.2 # Weighting for delivery time (inverse)
)
# Sort by composite score (descending) to find the best providers
sorted_providers = food_providers_df.sort_values(by='Composite Score', ascending=False)
\ensuremath{\text{\#}} Print the sorted DataFrame
print(sorted_providers)
```

```
# Create the bar chart for high-cost rooms
fig = px.bar(sorted_providers, x="Food Provider", y="Composite Score",
             title="Food Providers Ranking by Composite Score",
             color="Average Cost per Meal", # Color bars by average cost
             text="Customer Rating", # Show customer ratings
hover_data=["Delivery Time (minutes)"] # Show delivery time on hover
fig.update_layout(xaxis_title="Food Provider", yaxis_title="Composite Score")
fig.show()
<del>_</del>→
      Food Provider Average Cost per Meal Customer Rating \
          Provider B
                                             6
                                                             4.5
     0
          Provider A
                                             8
                                                             4.2
          Provider D
     3
                                                             4.0
                                                             3.8
     2
          Provider C
                                            10
        Delivery Time (minutes) Composite Score
                               25
                                           2.308000
     0
                               30
                                           2.144167
     3
                               35
                                           2.048571
                                           1.935000
```

Food Providers Ranking by Composite Score

