

```
import pandas as pd
import numpy as np
```

Double-click (or enter) to edit

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



Choose Files house\_price.csv

- **house\_price.csv**(text/csv) - 29981 bytes, last modified: 4/27/2025 - 100% done  
Saving house\_price.csv to house\_price (1).csv

```
df = pd.read_csv('house_price.csv')
```

## ✓ 1. Total number of houses and features

```
print(f"Rows: {df.shape[0]}, Columns: {df.shape[1]}")
```



Rows: 545, Columns: 13

## ✓ 2. Display first 5 houses using .sample()

```
print(df.sample(5))
```



	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
50	7420000	7440	3	2	4	yes	no	no	
441	3220000	4370	3	1	2	yes	no	no	
530	2240000	1950	3	1	1	no	no	no	
195	4970000	4410	4	3	2	yes	no	yes	
76	6650000	6420	3	2	3	yes	no	no	

  

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
50	no	no	1	yes	unfurnished
441	no	no	0	no	unfurnished
530	yes	no	0	no	unfurnished
195	no	no	2	no	semi-furnished
76	no	yes	0	yes	furnished

### ✓ 3. Display last 5 houses with selected columns

```
print(df.tail(5)[['area', 'price', 'basement']])
```

```

↵
   area  price basement
540 3000 1820000      yes
541 2400 1767150       no
542 3620 1750000       no
543 2910 1750000       no
544 3850 1750000       no

```

### ✓ 4. List all feature names sorted alphabetically

```
print(sorted(df.columns))
```

```

↵ ['airconditioning', 'area', 'basement', 'bathrooms', 'bedrooms', 'furnishingstatus', 'guestroom', 'hotwaterheating', 'mainroad', 'parking', 'prefarea', '

```

### ✓ 5. Check data types and count them

```
print(df.dtypes.value_counts())
```

```

↵ object    7
   int64     6
   Name: count, dtype: int64

```

### ✓ 6. Find columns having more than 10 missing values

```
missing_cols = df.isnull().sum()
print(missing_cols[missing_cols > 10])
```

```

↵ Series([], dtype: int64)

```

### ✓ \*7. find the correlation between the price and the area \*

```
correlation = np.corrcoef(df['price'], df['area'])[0, 1]
print(f"Correlation between price and area: {correlation}")
```

Correlation between price and area: 0.5359973457780796

## 8. Find the average price of houses for each number of bedrooms.

```
avg_price_per_bedroom = df.groupby('bedrooms')['price'].mean()
print(avg_price_per_bedroom)
```

bedrooms

1	2.712500e+06
2	3.632022e+06
3	4.954598e+06
4	5.729758e+06
5	5.819800e+06
6	4.791500e+06

Name: price, dtype: float64

## 9. Calculate the average price of houses based on the number of stories

```
avg_price_by_stories = df.groupby('stories')['price'].mean()
print(avg_price_by_stories)
```

stories

1	4.170659e+06
2	4.764074e+06
3	5.685436e+06
4	7.208450e+06

Name: price, dtype: float64

## 10. Find the percentage of houses that have access to the main road (assuming mainroad is a binary feature: 1 for access, 0 for no access).

```
mainroad_percentage = np.mean(df['mainroad'] == 1) * 100
print(f"Percentage of houses with main road access: {mainroad_percentage}%")
```

➦ Percentage of houses with main road access: 0.0%

## ✓ 11. Compare the price of houses with and without a guestroom.

```
price_with_guestroom = df[df['guestroom'] == 1]['price'].mean()
price_without_guestroom = df[df['guestroom'] == 0]['price'].mean()

print(f"Average price of houses with guestroom: {price_with_guestroom}")
print(f"Average price of houses without guestroom: {price_without_guestroom}")
```

➦ Average price of houses with guestroom: nan  
Average price of houses without guestroom: nan

## ✓ 12. Calculate average SalePrice rounded to 2 decimals

```
print(round(df['price'].mean(), 2))
```

➦ 4766729.25

## ✓ 13. Find houses with SalePrice > 1.5 times mean using np.where

```
high_price = np.where(df['price'] > 1.5 * df['price'].mean(), True, False)
print(df[high_price])
```

➦

18	no	yes	2	no	furnished
19	no	yes	1	yes	semi-furnished
20	yes	no	2	no	semi-furnished
21	no	yes	2	no	unfurnished
22	no	yes	1	no	furnished
23	no	yes	1	no	furnished
24	no	yes	2	no	furnished
25	no	yes	2	yes	furnished
26	no	yes	0	yes	semi-furnished
27	no	no	1	no	semi-furnished
28	yes	no	2	no	unfurnished
29	no	yes	1	yes	semi-furnished
30	no	yes	2	no	unfurnished
31	no	yes	2	no	semi-furnished
32	no	yes	1	yes	furnished
33	no	no	1	no	unfurnished
34	no	yes	1	no	furnished
35	no	yes	2	no	furnished
36	yes	no	1	yes	furnished
37	no	yes	2	no	furnished
38	no	yes	2	no	unfurnished
39	no	yes	1	no	semi-furnished
40	no	yes	0	yes	furnished
41	no	yes	0	yes	furnished
42	no	yes	2	no	unfurnished
43	no	no	2	no	semi-furnished
44	no	yes	1	no	furnished
45	no	yes	0	no	semi-furnished
46	no	yes	1	no	furnished
47	no	yes	3	yes	furnished
48	no	no	1	no	unfurnished
49	no	yes	0	yes	semi-furnished
50	no	no	1	yes	unfurnished
51	no	yes	1	no	unfurnished
52	no	yes	1	no	furnished
53	no	yes	2	no	semi-furnished
54	no	yes	1	no	semi-furnished
55	no	yes	1	no	unfurnished
56	no	no	1	yes	semi-furnished
57	no	yes	1	yes	furnished
58	no	yes	1	no	semi-furnished
59	no	yes	1	no	furnished

## ✓ 14. Average SalePrice per Area (descending)

```
print(df.groupby('area')['price'].mean().sort_values(ascending=False))
```

```

↗ area
7420    12355000.0
8960    12250000.0

```

```

9960    12250000.0
7500    11532500.0
16200   10150000.0
...
2400    1933575.0
1700    1890000.0
3649    1890000.0
2990    1855000.0
3620    1750000.0
Name: price, Length: 284, dtype: float64

```

## ✓ 15. Create a new column 'TotalArea' with apply() row-wise

```

df['TotalArea'] = df.apply(lambda row: row['stories'] + row['bathrooms'], axis=1)
print(df[['furnishingstatus', 'TotalArea']])

```

```

↗
furnishingstatus  TotalArea
0      furnished         5
1      furnished         8
2  semi-furnished         4
3      furnished         4
4      furnished         3
..          ...
540  unfurnished         2
541  semi-furnished         2
542  unfurnished         2
543  furnished          2
544  unfurnished         3

```

[545 rows x 2 columns]

## ✓ 16. Find correlation between important features

```

print(df[['area', 'stories', 'price']].corr())

```

```

↗
      area  stories  price
area    1.000000  0.083996  0.535997
stories  0.083996  1.000000  0.420712
price    0.535997  0.420712  1.000000

```

## ✓ 17. Calculate standard deviation of SalePrice manually

```
saleprice_std = np.sqrt(np.mean((df['price'] - df['price'].mean())**2))
print(f"Standard Deviation of price: {saleprice_std:.2f}")
```

Standard Deviation of price: 1868722.83

## 18. Descriptive stats for numerical columns using .agg()

```
print(df.select_dtypes(include=[np.number]).agg(['mean', 'std', 'min', 'max']))
```

```

price      area  bedrooms  bathrooms  stories  parking \
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
max   1.330000e+07  16200.000000  6.000000  4.000000  4.000000  3.000000

TotalArea
mean    3.091743
std     1.135501
min     2.000000
max     8.000000

```

## 19. Calculate the average price for houses with and without air conditioning.

```
price_with_ac = df[df['airconditioning'] == 1]['price'].mean()
price_without_ac = df[df['airconditioning'] == 0]['price'].mean()

print(f"Average price of houses with air conditioning: {price_with_ac}")
print(f"Average price of houses without air conditioning: {price_without_ac}")
```

Average price of houses with air conditioning: nan  
Average price of houses without air conditioning: nan

## 20. Find the number of houses for each parking space category.

```
parking_counts = df['parking'].value_counts()
print(parking_counts)
```

```

parking
0      299
1      126

```

```

2    108
3     12
Name: count, dtype: int64

```

## 21. Analyze how the number of bedrooms and bathrooms affect the price.

```

avg_price_by_bed_bath = df.groupby(['bedrooms', 'bathrooms'])['price'].mean().unstack()
print(avg_price_by_bed_bath)

```

```

bathrooms      1          2          3          4
bedrooms
1      2.712500e+06      NaN      NaN      NaN
2      3.596852e+06  4.194750e+06      NaN      NaN
3      4.463484e+06  6.379569e+06  6807500.0      NaN
4      4.710583e+06  6.358990e+06  7250250.0  12250000.0
5      4.996250e+06  5.405750e+06  8295000.0      NaN
6      3.500000e+06  6.083000e+06      NaN      NaN

```

## 22. Analyze the effect of having multiple stories on house price.

```

avg_price_by_stories = df.groupby('stories')['price'].mean()
print(avg_price_by_stories)

```

```

stories
1      4.170659e+06
2      4.764074e+06
3      5.685436e+06
4      7.208450e+06
Name: price, dtype: float64

```

## 23. Find the number of missing values in each column.

```

missing_values = df.isnull().sum()
print(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
TotalArea     0
dtype: int64

```

## 24. Calculate the minimum and maximum price for houses based on their furnishing status.

```

furnishing_price_range = df.groupby('furnishingstatus')['price'].agg(['min', 'max'])
print(furnishing_price_range)

```

```

furnishingstatus      min      max
furnished             1750000 13300000
semi-furnished        1767150 12250000
unfurnished           1750000 10150000

```

## 25. Find the correlation between various numerical columns like price, area, bedrooms, etc.

```

correlation_matrix = df[['price', 'area', 'bedrooms', 'bathrooms', 'stories']].corr()
print(correlation_matrix)

```

```

price      price      area  bedrooms  bathrooms  stories
price      1.000000  0.535997  0.366494  0.517545  0.420712
area       0.535997  1.000000  0.151858  0.193820  0.083996
bedrooms   0.366494  0.151858  1.000000  0.373930  0.408564
bathrooms  0.517545  0.193820  0.373930  1.000000  0.326165
stories    0.420712  0.083996  0.408564  0.326165  1.000000

```

## 26. Find out the percentage of houses with each type of furnishing status (e.g., fully furnished, semi-furnished).

```
furnishing_percentage = df['furnishingstatus'].value_counts(normalize=True) * 100
print(furnishing_percentage)
```

```
furnishingstatus
semi-furnished    41.651376
unfurnished       32.660550
furnished         25.688073
Name: proportion, dtype: float64
```

## 27. Count the total number of houses that have a guestroom.

**bold text**

```
df['guestroom'] = df['guestroom'].map({'Yes': 1, 'No': 0})
```

```
# Now you can calculate the total number of houses with a guestroom
guestroom_count = df['guestroom']
print(f"Total number of houses with a guestroom: {guestroom_count}")
```

```
Total number of houses with a guestroom: 0      NaN
1      NaN
2      NaN
3      NaN
4      NaN
..
540    NaN
541    NaN
542    NaN
543    NaN
544    NaN
Name: guestroom, Length: 545, dtype: float64
```

