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Module 2: Data Wrangling

Question 2

Drop the columns "id" and "Unnamed: 0" from axis 1 using the method `drop()`, then use the method `describe()` to obtain a statistical summary of the data. Take a screenshot and submit it, make sure the `inplace` parameter is set to `True`

```
[23]: # drop the 'id' and 'Unnamed: 0' columns
df.drop(['id', 'Unnamed: 0'], axis=1, inplace=True)

# describe the data
df.describe()
```

```
[23]:
```

	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	grade	sqft_above
count	2.161300e+04	21600.000000	21603.000000	21613.000000	2.161300e+04	21613.000000	21613.000000	21613.000000	21613.000000	21613.000000	21613.000000
mean	5.400881e+05	3.372870	2.115736	2079.899736	1.510697e+04	1.494309	0.007542	0.234303	3.409430	7.656873	1788.390691

Simple 0 1 Pyolite | Idle










Mode: Command Ln 1, Col 1 House_Sales_in_King_Count_USA.jupyterlite.ipynb

Support/Feedback

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[23]:

	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	grade	sqft_above
count	2.161300e+04	21600.000000	21603.000000	21613.000000	2.161300e+04	21613.000000	21613.000000	21613.000000	21613.000000	21613.000000	21613.000000
mean	5.400881e+05	3.372870	2.115736	2079.899736	1.510697e+04	1.494309	0.007542	0.234303	3.409430	7.656873	1788.390691
std	3.671272e+05	0.926657	0.768996	918.440897	4.142051e+04	0.539989	0.086517	0.766318	0.650743	1.175459	828.090978
min	7.500000e+04	1.000000	0.500000	290.000000	5.200000e+02	1.000000	0.000000	0.000000	1.000000	1.000000	290.000000
25%	3.219500e+05	3.000000	1.750000	1427.000000	5.040000e+03	1.000000	0.000000	0.000000	3.000000	7.000000	1190.000000
50%	4.500000e+05	3.000000	2.250000	1910.000000	7.618000e+03	1.500000	0.000000	0.000000	3.000000	7.000000	1560.000000
75%	6.450000e+05	4.000000	2.500000	2550.000000	1.068800e+04	2.000000	0.000000	0.000000	4.000000	8.000000	2210.000000
max	7.700000e+06	33.000000	8.000000	13540.000000	1.651359e+06	3.500000	1.000000	4.000000	5.000000	13.000000	9410.000000

We can see we have missing values for the columns `bedrooms` and `bathrooms`

[72]:

```
print("number of NaN values for the column bedrooms :", df['bedrooms'].isnull().sum())
print("number of NaN values for the column bathrooms :", df['bathrooms'].isnull().sum())
```

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Support/Feedback



number of NaN values for the column bedrooms : 13
number of NaN values for the column bathrooms : 10

We can replace the missing values of the column 'bedrooms' with the mean of the column 'bedrooms' using the method `replace()`. Don't forget to set the `inplace` parameter to `True`

```
[25]: mean=df['bedrooms'].mean()  
df['bedrooms'].replace(np.nan,mean, inplace=True)
```

We also replace the missing values of the column 'bathrooms' with the mean of the column 'bathrooms' using the method `replace()`. Don't forget to set the `inplace` parameter to `True`

```
[26]: mean=df['bathrooms'].mean()  
df['bathrooms'].replace(np.nan,mean, inplace=True)
```

```
[27]: print("number of NaN values for the column bedrooms :", df['bedrooms'].isnull().sum())  
print("number of NaN values for the column bathrooms :", df['bathrooms'].isnull().sum())
```

number of NaN values for the column bedrooms : 0
number of NaN values for the column bathrooms : 0

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