

```
In [1]: import pandas as pd
df = pd.read_csv(r'C:\Users\raj20\Downloads\melb_data.csv\melb_data.csv')
df.columns
```

Out[1]: Index(['Suburb', 'Address', 'Rooms', 'Type', 'Price', 'Method', 'SellerG', 'Date', 'Distance', 'Postcode', 'Bedroom2', 'Bathroom', 'Car', 'Landsize', 'BuildingArea', 'YearBuilt', 'CouncilArea', 'Lattitude', 'Longtitude', 'Regionname', 'Propertycount'], dtype='object')

```
In [2]: df.isnull().any() #are there any missing values?
```

Out[2]: Suburb False
Address False
Rooms False
Type False
Price False
Method False
SellerG False
Date False
Distance False
Postcode False
Bedroom2 False
Bathroom False
Car True
Landsize False
BuildingArea True
YearBuilt True
CouncilArea True
Lattitude False
Longtitude False
Regionname False
Propertycount False
dtype: bool

```
In [3]: df.isnull().sum() #no. of missing values
```

Out[3]: Suburb 0
Address 0
Rooms 0
Type 0
Price 0
Method 0
SellerG 0
Date 0
Distance 0
Postcode 0
Bedroom2 0
Bathroom 0
Car 62
Landsize 0
BuildingArea 6450
YearBuilt 5375
CouncilArea 1369
Lattitude 0
Longtitude 0
Regionname 0
Propertycount 0
dtype: int64

```
In [4]: #delete the columns having NaN
```

```
In [5]: cols_with_missing = [col for col in df.columns
                             if df[col].isnull().any()]
print(cols_with_missing)

['Car', 'BuildingArea', 'YearBuilt', 'CouncilArea']
```

```
In [6]: reduced_df = df.drop(cols_with_missing,axis =1) #axis 1 for col, 0 for rows
reduced_df.columns
```

Out[6]: Index(['Suburb', 'Address', 'Rooms', 'Type', 'Price', 'Method', 'SellerG', 'Date', 'Distance', 'Postcode', 'Bedroom2', 'Bathroom', 'Landsize', 'Lattitude', 'Longitude', 'Regionname', 'Propertycount'], dtype='object')

```
In [7]: df.shape
```

Out[7]: (13580, 21)

```
In [8]: reduced_df.shape
```

Out[8]: (13580, 17)

```
In [9]: reduced_df2 = df.dropna()
reduced_df2
```

Out[9]:

	Suburb	Address	Rooms	Type	Price	Method	SellerG	Date	Distance	Postcode	...	Bathroom	Car	Landsize	BuildingArea	YearBuilt	CouncilArea	Lattitude	Longitude	Regionname	Propertycount
1	Abbotsford	25 Bloomburg St	2	h	1035000.0	S	Biggin	4/02/2016	2.5	3067.0	...	1.0	0.0	156.0	79.00	1900.0	Yarra	-37.80790	144.99340	Northern Metropolitan	4019.0
2	Abbotsford	5 Charles St	3	h	1465000.0	SP	Biggin	4/03/2017	2.5	3067.0	...	2.0	0.0	134.0	150.00	1900.0	Yarra	-37.80930	144.99440	Northern Metropolitan	4019.0
4	Abbotsford	55a Park St	4	h	1600000.0	VB	Nelson	4/06/2016	2.5	3067.0	...	1.0	2.0	120.0	142.00	2014.0	Yarra	-37.80720	144.99410	Northern Metropolitan	4019.0
6	Abbotsford	124 Yarra St	3	h	1876000.0	S	Nelson	7/05/2016	2.5	3067.0	...	2.0	0.0	245.0	210.00	1910.0	Yarra	-37.80240	144.99930	Northern Metropolitan	4019.0
7	Abbotsford	98 Charles St	2	h	1636000.0	S	Nelson	8/10/2016	2.5	3067.0	...	1.0	2.0	256.0	107.00	1890.0	Yarra	-37.80600	144.99540	Northern Metropolitan	4019.0
...
12205	Whittlesea	30 Sherwin St	3	h	601000.0	S	Ray	29/07/2017	35.5	3757.0	...	2.0	1.0	972.0	149.00	1996.0	Whittlesea	-37.51232	145.13282	Northern Victoria	2170.0
12206	Williamstown	75 Cecil St	3	h	1050000.0	VB	Williams	29/07/2017	6.8	3016.0	...	1.0	0.0	179.0	115.00	1890.0	Hobsons Bay	-37.86558	144.90474	Western Metropolitan	6380.0
12207	Williamstown	2/29 Dover Rd	1	u	385000.0	SP	Williams	29/07/2017	6.8	3016.0	...	1.0	1.0	0.0	35.64	1967.0	Hobsons Bay	-37.85588	144.89936	Western Metropolitan	6380.0
12209	Windsor	201/152 Peel St	2	u	560000.0	PI	hockingstuart	29/07/2017	4.6	3181.0	...	1.0	1.0	0.0	61.60	2012.0	Stonnington	-37.85581	144.99025	Southern Metropolitan	4380.0
12212	Yarraville	54 Pentland Pde	6	h	2450000.0	VB	Village	29/07/2017	6.3	3013.0	...	3.0	2.0	1087.0	388.50	1920.0	Maribyrnong	-37.81038	144.89389	Western Metropolitan	6543.0

6196 rows × 21 columns

```
In [10]: reduced_df2.shape
```

Out[10]: (6196, 21)

```
In [11]: ml = pd.read_csv(r'C:\Users\raj20\Downloads\melb_data.csv\melb_data.csv')
```

```
In [12]: print(ml.info())
print(ml.shape)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 13580 entries, 0 to 13579
Data columns (total 21 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Suburb          13580 non-null  object
1   Address         13580 non-null  object
2   Rooms           13580 non-null  int64
3   Type            13580 non-null  object
4   Price           13580 non-null  float64
5   Method          13580 non-null  object
6   SellerG         13580 non-null  object
7   Date            13580 non-null  object
8   Distance        13580 non-null  float64
9   Postcode        13580 non-null  float64
10  Bedroom2        13580 non-null  float64
11  Bathroom        13580 non-null  float64
12  Car              13518 non-null  float64
13  Landsize        13580 non-null  float64
14  BuildingArea    7130 non-null   float64
15  YearBuilt       8205 non-null   float64
16  CouncilArea     12211 non-null  object
17  Lattitude       13580 non-null  float64
18  Longtitude      13580 non-null  float64
19  Regionname      13580 non-null  object
20  Propertycount   13580 non-null  float64
dtypes: float64(12), int64(1), object(8)
memory usage: 2.2+ MB
None
(13580, 21)
```

```
In [13]: ml['BuildingArea'].describe()
```

```
Out[13]: count      7130.000000
mean         151.967650
std           541.014538
min            0.000000
25%           93.000000
50%          126.000000
75%          174.000000
max         44515.000000
Name: BuildingArea, dtype: float64
```

```
In [14]: print(ml['BuildingArea'].mean())
print(ml['BuildingArea'].median())
```

```
151.96764988779805
126.0
```

```
In [17]: print(ml.head())
ml['BuildingArea']=ml['BuildingArea'].fillna(ml['BuildingArea'].median())
print(ml.isnull().any())
```

	Suburb	Address	Rooms	Type	Price	Method	SellerG	\
0	Abbotsford	85 Turner St	2	h	1480000.0	S	Biggin	
1	Abbotsford	25 Bloomburg St	2	h	1035000.0	S	Biggin	
2	Abbotsford	5 Charles St	3	h	1465000.0	SP	Biggin	
3	Abbotsford	40 Federation La	3	h	850000.0	PI	Biggin	
4	Abbotsford	55a Park St	4	h	1600000.0	VB	Nelson	

	Date	Distance	Postcode	...	Bathroom	Car	Landsize	BuildingArea	\
0	3/12/2016	2.5	3067.0	...	1.0	1.0	202.0	126.0	
1	4/02/2016	2.5	3067.0	...	1.0	0.0	156.0	79.0	
2	4/03/2017	2.5	3067.0	...	2.0	0.0	134.0	150.0	
3	4/03/2017	2.5	3067.0	...	2.0	1.0	94.0	126.0	
4	4/06/2016	2.5	3067.0	...	1.0	2.0	120.0	142.0	

	YearBuilt	CouncilArea	Lattitude	Longtitude	Regionname	\
0	NaN	Yarra	-37.7996	144.9984	Northern Metropolitan	
1	1900.0	Yarra	-37.8079	144.9934	Northern Metropolitan	
2	1900.0	Yarra	-37.8093	144.9944	Northern Metropolitan	
3	NaN	Yarra	-37.7969	144.9969	Northern Metropolitan	
4	2014.0	Yarra	-37.8072	144.9941	Northern Metropolitan	

	Propertycount
0	4019.0
1	4019.0
2	4019.0
3	4019.0
4	4019.0

[5 rows x 21 columns]	
Suburb	False
Address	False
Rooms	False
Type	False
Price	False
Method	False
SellerG	False
Date	False
Distance	False
Postcode	False
Bedroom2	False
Bathroom	False
Car	True
Landsize	False
BuildingArea	False
YearBuilt	True
CouncilArea	True
Lattitude	False
Longtitude	False
Regionname	False
Propertycount	False
dtype:	bool

In [18]:

ml.head()

Out[18]:

	Suburb	Address	Rooms	Type	Price	Method	SellerG	Date	Distance	Postcode	...	Bathroom	Car	Landsize	BuildingArea	YearBuilt	CouncilArea	Latitude	Longitude	Regionname	Propertycount
0	Abbotsford	85 Turner St	2	h	1480000.0	S	Biggin	3/12/2016	2.5	3067.0	...	1.0	1.0	202.0	126.0	NaN	Yarra	-37.7996	144.9984	Northern Metropolitan	4019.0
1	Abbotsford	25 Bloomburg St	2	h	1035000.0	S	Biggin	4/02/2016	2.5	3067.0	...	1.0	0.0	156.0	79.0	1900.0	Yarra	-37.8079	144.9934	Northern Metropolitan	4019.0
2	Abbotsford	5 Charles St	3	h	1465000.0	SP	Biggin	4/03/2017	2.5	3067.0	...	2.0	0.0	134.0	150.0	1900.0	Yarra	-37.8093	144.9944	Northern Metropolitan	4019.0
3	Abbotsford	40 Federation La	3	h	850000.0	PI	Biggin	4/03/2017	2.5	3067.0	...	2.0	1.0	94.0	126.0	NaN	Yarra	-37.7969	144.9969	Northern Metropolitan	4019.0
4	Abbotsford	55a Park St	4	h	1600000.0	VB	Nelson	4/06/2016	2.5	3067.0	...	1.0	2.0	120.0	142.0	2014.0	Yarra	-37.8072	144.9941	Northern Metropolitan	4019.0

5 rows × 21 columns

In [19]:

ml['CouncilArea'].unique()

Out[19]:

array(['Yarra', 'Moonee Valley', 'Port Phillip', 'Darebin', 'Hobsons Bay',
 'Stonnington', 'Boroondara', 'Monash', 'Glen Eira', 'Whitehorse',
 'Maribyrnong', 'Bayside', 'Moreland', 'Manningham', 'Banyule',
 'Melbourne', 'Kingston', 'Brimbank', 'Hume', nan, 'Knox',
 'Maroondah', 'Casey', 'Melton', 'Greater Dandenong', 'Nillumbik',
 'Whittlesea', 'Frankston', 'Macedon Ranges', 'Yarra Ranges',
 'Wyndham', 'Cardinia', 'Unavailable', 'Moorabool'], dtype=object)

```
In [20]: ml['CouncilArea'].value_counts()
```

```
Out[20]: Moreland          1163
Boroondara          1160
Moonee Valley        997
Darebin              934
Glen Eira            848
Stonnington          719
Maribyrnong          692
Yarra                647
Port Phillip         628
Banyule              594
Bayside              489
Melbourne            470
Hobsons Bay          434
Brimbank             424
Monash               333
Manningham           311
Whitehorse           304
Kingston             207
Whittlesea           167
Hume                 164
Wyndham              86
Knox                 80
Maroondah            80
Melton               66
Frankston            53
Greater Dandenong    52
Casey                38
Nillumbik            36
Yarra Ranges         18
Cardinia              8
Macedon Ranges        7
Unavailable           1
Moorabool             1
Name: CouncilArea, dtype: int64
```

```
In [ ]:
```