

Milestone 2

July 4, 2023

Cleaning/Formatting Flat File Source

```
[57]: # Import libraries
import pandas as pd
```

```
[58]: data = pd.read_csv('Sale_Prices_City.csv')
```

```
[59]: data.shape
```

```
[59]: (3728, 150)
```

```
[60]: data.head()
```

```
[60]: Unnamed: 0  RegionID  RegionName  StateName  SizeRank  2008-03  \
0          0      6181    New York    New York         1      NaN
1          1     12447  Los Angeles  California         2  507600.0
2          2     39051    Houston     Texas          3  138400.0
3          3     17426    Chicago    Illinois         4  325100.0
4          4      6915  San Antonio     Texas          5  130900.0

      2008-04  2008-05  2008-06  2008-07  ...  2019-06  2019-07  2019-08  \
0         NaN         NaN         NaN         NaN  ...  563200.0  570500.0  572800.0
1  489600.0  463000.0  453100.0  438100.0  ...  706800.0  711800.0  717300.0
2  135500.0  132200.0  131000.0  133400.0  ...  209700.0  207400.0  207600.0
3  314800.0  286900.0  274600.0  268500.0  ...  271500.0  266500.0  264900.0
4  131300.0  131200.0  131500.0  131600.0  ...  197100.0  198700.0  200200.0

      2019-09  2019-10  2019-11  2019-12  2020-01  2020-02  2020-03
0  569900.0  560800.0  571500.0  575100.0  571700.0  568300.0  573600.0
1  714100.0  711900.0  718400.0  727100.0  738200.0  760200.0         NaN
2  207000.0  211400.0  211500.0  217700.0  219200.0  223800.0         NaN
3  265000.0  264100.0  264300.0  270000.0  281400.0  302900.0  309200.0
4  200800.0  203400.0  203800.0  205400.0  205400.0  208300.0         NaN
```

```
[5 rows x 150 columns]
```

Step 1: Remove Unnamed Column

```
[61]: """
In the data, there is a column that is unnamed. This column will not be needed
↳for analysis.
The code below uses the drop function and indexes the first column to be
↳dropped.
"""
data = data.drop(columns=data.columns[0])
```

```
[62]: data.head()
```

```
[62]:   RegionID  RegionName  StateName  SizeRank  2008-03  2008-04  2008-05  \
0      6181    New York    New York         1      NaN      NaN      NaN
1     12447  Los Angeles  California         2  507600.0  489600.0  463000.0
2     39051    Houston    Texas           3  138400.0  135500.0  132200.0
3     17426    Chicago    Illinois          4  325100.0  314800.0  286900.0
4      6915  San Antonio    Texas           5  130900.0  131300.0  131200.0

      2008-06  2008-07  2008-08  ...  2019-06  2019-07  2019-08  2019-09  \
0      NaN      NaN      NaN  ...  563200.0  570500.0  572800.0  569900.0
1  453100.0  438100.0  423200.0  ...  706800.0  711800.0  717300.0  714100.0
2  131000.0  133400.0  135400.0  ...  209700.0  207400.0  207600.0  207000.0
3  274600.0  268500.0  264400.0  ...  271500.0  266500.0  264900.0  265000.0
4  131500.0  131600.0  132300.0  ...  197100.0  198700.0  200200.0  200800.0

      2019-10  2019-11  2019-12  2020-01  2020-02  2020-03
0  560800.0  571500.0  575100.0  571700.0  568300.0  573600.0
1  711900.0  718400.0  727100.0  738200.0  760200.0      NaN
2  211400.0  211500.0  217700.0  219200.0  223800.0      NaN
3  264100.0  264300.0  270000.0  281400.0  302900.0  309200.0
4  203400.0  203800.0  205400.0  205400.0  208300.0      NaN

[5 rows x 149 columns]
```

Step 2: Remove RegionID

```
[63]: """
In the data, there is a column named RedionID that will provide an additional
↳identifier for the data.
However, this column is only apart of this data set so it will not serve a
↳purpose for the other sources.
The code below uses the drop function and indexes the first column to be
↳dropped.
"""
data = data.drop(columns=data.columns[0])
```

```
[64]: data.head()
```

```
[64]:
```

	RegionName	StateName	SizeRank	2008-03	2008-04	2008-05	2008-06	\
0	New York	New York	1	NaN	NaN	NaN	NaN	
1	Los Angeles	California	2	507600.0	489600.0	463000.0	453100.0	
2	Houston	Texas	3	138400.0	135500.0	132200.0	131000.0	
3	Chicago	Illinois	4	325100.0	314800.0	286900.0	274600.0	
4	San Antonio	Texas	5	130900.0	131300.0	131200.0	131500.0	

	2008-07	2008-08	2008-09	...	2019-06	2019-07	2019-08	2019-09	\
0	NaN	NaN	NaN	...	563200.0	570500.0	572800.0	569900.0	
1	438100.0	423200.0	407800.0	...	706800.0	711800.0	717300.0	714100.0	
2	133400.0	135400.0	138000.0	...	209700.0	207400.0	207600.0	207000.0	
3	268500.0	264400.0	267100.0	...	271500.0	266500.0	264900.0	265000.0	
4	131600.0	132300.0	131600.0	...	197100.0	198700.0	200200.0	200800.0	

	2019-10	2019-11	2019-12	2020-01	2020-02	2020-03
0	560800.0	571500.0	575100.0	571700.0	568300.0	573600.0
1	711900.0	718400.0	727100.0	738200.0	760200.0	NaN
2	211400.0	211500.0	217700.0	219200.0	223800.0	NaN
3	264100.0	264300.0	270000.0	281400.0	302900.0	309200.0
4	203400.0	203800.0	205400.0	205400.0	208300.0	NaN

[5 rows x 148 columns]

Step 3: Rename RegionName to CityName

```
[65]: """
When looking at the table the column "RegionName" indicates cities.
If there is a join needed by city, this column will need to be renamed.
The code below used the rename function to select "RegionName" and renames it_
↳to "CityName"
"""
data = data.rename(columns={'RegionName': 'CityName'})
```

```
[66]: data.head()
```

```
[66]:
```

	CityName	StateName	SizeRank	2008-03	2008-04	2008-05	2008-06	\
0	New York	New York	1	NaN	NaN	NaN	NaN	
1	Los Angeles	California	2	507600.0	489600.0	463000.0	453100.0	
2	Houston	Texas	3	138400.0	135500.0	132200.0	131000.0	
3	Chicago	Illinois	4	325100.0	314800.0	286900.0	274600.0	
4	San Antonio	Texas	5	130900.0	131300.0	131200.0	131500.0	

	2008-07	2008-08	2008-09	...	2019-06	2019-07	2019-08	2019-09	\
0	NaN	NaN	NaN	...	563200.0	570500.0	572800.0	569900.0	
1	438100.0	423200.0	407800.0	...	706800.0	711800.0	717300.0	714100.0	
2	133400.0	135400.0	138000.0	...	209700.0	207400.0	207600.0	207000.0	
3	268500.0	264400.0	267100.0	...	271500.0	266500.0	264900.0	265000.0	

```

4  131600.0  132300.0  131600.0  ...  197100.0  198700.0  200200.0  200800.0

      2019-10  2019-11  2019-12  2020-01  2020-02  2020-03
0  560800.0  571500.0  575100.0  571700.0  568300.0  573600.0
1  711900.0  718400.0  727100.0  738200.0  760200.0      NaN
2  211400.0  211500.0  217700.0  219200.0  223800.0      NaN
3  264100.0  264300.0  270000.0  281400.0  302900.0  309200.0
4  203400.0  203800.0  205400.0  205400.0  208300.0      NaN

```

[5 rows x 148 columns]

Step 4: Fill NaN Values with 0

```

[67]: """
      The data contains some NaN values and if calculations are done, this could
      ↪ create some issues.
      The code below uses fillna function to replace all NaN values with 0
      """
      data = data.fillna(0)

```

```

[68]: data.head()

```

```

[68]:      CityName  StateName  SizeRank  2008-03  2008-04  2008-05  2008-06  \
0      New York    New York         1         0.0         0.0         0.0         0.0
1  Los Angeles  California         2  507600.0  489600.0  463000.0  453100.0
2      Houston      Texas         3  138400.0  135500.0  132200.0  131000.0
3      Chicago    Illinois         4  325100.0  314800.0  286900.0  274600.0
4  San Antonio      Texas         5  130900.0  131300.0  131200.0  131500.0

      2008-07  2008-08  2008-09  ...  2019-06  2019-07  2019-08  2019-09  \
0         0.0         0.0         0.0  ...  563200.0  570500.0  572800.0  569900.0
1  438100.0  423200.0  407800.0  ...  706800.0  711800.0  717300.0  714100.0
2  133400.0  135400.0  138000.0  ...  209700.0  207400.0  207600.0  207000.0
3  268500.0  264400.0  267100.0  ...  271500.0  266500.0  264900.0  265000.0
4  131600.0  132300.0  131600.0  ...  197100.0  198700.0  200200.0  200800.0

      2019-10  2019-11  2019-12  2020-01  2020-02  2020-03
0  560800.0  571500.0  575100.0  571700.0  568300.0  573600.0
1  711900.0  718400.0  727100.0  738200.0  760200.0         0.0
2  211400.0  211500.0  217700.0  219200.0  223800.0         0.0
3  264100.0  264300.0  270000.0  281400.0  302900.0  309200.0
4  203400.0  203800.0  205400.0  205400.0  208300.0         0.0

```

[5 rows x 148 columns]

Step 5: Remove the column SizeRank

```
[69]: """
The column SizeRank provides information that will not be used later on.
To keep the data clean, this column can be removed.
The code below, selects the column size rank and removes it.
"""
data.drop('SizeRank', axis=1, inplace=True)
```

```
[70]: data.head()
```

```
[70]:
```

	CityName	StateName	2008-03	2008-04	2008-05	2008-06	2008-07	\
0	New York	New York	0.0	0.0	0.0	0.0	0.0	
1	Los Angeles	California	507600.0	489600.0	463000.0	453100.0	438100.0	
2	Houston	Texas	138400.0	135500.0	132200.0	131000.0	133400.0	
3	Chicago	Illinois	325100.0	314800.0	286900.0	274600.0	268500.0	
4	San Antonio	Texas	130900.0	131300.0	131200.0	131500.0	131600.0	

	2008-08	2008-09	2008-10	...	2019-06	2019-07	2019-08	2019-09	\
0	0.0	0.0	0.0	...	563200.0	570500.0	572800.0	569900.0	
1	423200.0	407800.0	396300.0	...	706800.0	711800.0	717300.0	714100.0	
2	135400.0	138000.0	136400.0	...	209700.0	207400.0	207600.0	207000.0	
3	264400.0	267100.0	268400.0	...	271500.0	266500.0	264900.0	265000.0	
4	132300.0	131600.0	131800.0	...	197100.0	198700.0	200200.0	200800.0	

	2019-10	2019-11	2019-12	2020-01	2020-02	2020-03
0	560800.0	571500.0	575100.0	571700.0	568300.0	573600.0
1	711900.0	718400.0	727100.0	738200.0	760200.0	0.0
2	211400.0	211500.0	217700.0	219200.0	223800.0	0.0
3	264100.0	264300.0	270000.0	281400.0	302900.0	309200.0
4	203400.0	203800.0	205400.0	205400.0	208300.0	0.0

[5 rows x 147 columns]

Step 6: Format all numbers to be comma separated

```
[71]: """
The data is easily read however, when there are hundreds of thousands, there
↳ can be a difficulty differentiating
the amount.
The code below changes all the columns after the second column to be formatted
↳ with a comma at each appropriate place.
"""
data.loc[:,2:] = data.iloc[:,2:].applymap(lambda x: '{:,}'.format(x))
```

```
/var/folders/sr/xvmzsbj91c91yq0f0qnq71xh0000gn/T/ipykernel_53036/3161285015.py:4
: FutureWarning: Slicing a positional slice with .loc is not supported, and will
raise TypeError in a future version. Use .loc with labels or .iloc with
positions instead.
```

```
data.loc[:,2:] = data.iloc[:,2:].applymap(lambda x: '{:,}'.format(x))
```

```
[72]: data.head()
```

```
[72]:
```

	CityName	StateName	2008-03	2008-04	2008-05	2008-06	\
0	New York	New York	0.0	0.0	0.0	0.0	
1	Los Angeles	California	507,600.0	489,600.0	463,000.0	453,100.0	
2	Houston	Texas	138,400.0	135,500.0	132,200.0	131,000.0	
3	Chicago	Illinois	325,100.0	314,800.0	286,900.0	274,600.0	
4	San Antonio	Texas	130,900.0	131,300.0	131,200.0	131,500.0	

	2008-07	2008-08	2008-09	2008-10	...	2019-06	2019-07	\
0	0.0	0.0	0.0	0.0	...	563,200.0	570,500.0	
1	438,100.0	423,200.0	407,800.0	396,300.0	...	706,800.0	711,800.0	
2	133,400.0	135,400.0	138,000.0	136,400.0	...	209,700.0	207,400.0	
3	268,500.0	264,400.0	267,100.0	268,400.0	...	271,500.0	266,500.0	
4	131,600.0	132,300.0	131,600.0	131,800.0	...	197,100.0	198,700.0	

	2019-08	2019-09	2019-10	2019-11	2019-12	2020-01	\
0	572,800.0	569,900.0	560,800.0	571,500.0	575,100.0	571,700.0	
1	717,300.0	714,100.0	711,900.0	718,400.0	727,100.0	738,200.0	
2	207,600.0	207,000.0	211,400.0	211,500.0	217,700.0	219,200.0	
3	264,900.0	265,000.0	264,100.0	264,300.0	270,000.0	281,400.0	
4	200,200.0	200,800.0	203,400.0	203,800.0	205,400.0	205,400.0	

	2020-02	2020-03
0	568,300.0	573,600.0
1	760,200.0	0.0
2	223,800.0	0.0
3	302,900.0	309,200.0
4	208,300.0	0.0

[5 rows x 147 columns]