

# Data Wrangling I

## 1. Import all the required Python Libraries.

Why do we need pandas, why not our excel?

- **Advantage**

quickly analyse data & gives you insight  
need not to be a programmer

- **Disadvantage**

Can not handle large amount of data  
it may crashes while loading a data  
Missing value, cleaned data involves lots of process

- Pandas developed for data analysis
- support Multiple file format
- Time series analysis
- One Script can be used for similar operation again & again

In [127]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
#so that we can view the graphs inside the notebook
```

In [128]:

```
#from google.colab import drive
#drive.mount('/content/gdrive')
```

### Fundamental Data Types in Pandas

1. Series---- 1 D array with corresponding index
2. Data Frame ---- n D array In [129]:

```
s1 = pd.Series(range(1,10,1))
```

In [130]:

```
s1
```

Out[130]:

```
0    1
1    2
2    3
3    4
4    5
5    6
6    7
7    8 8    9 dtype: int64 In [131]:
```

```
s3 = pd.Series({1:21, 2:13,3:45})
```

In [132]:

```
s3
```

Out[132]:

```
1    21
2    13 3    45 dtype: int64
```

In [133]: s2 = pd.Series([1, 2, 3, 4], index=['p', 'q', 'r', 's'],

```
name='one') In [134]:
```

```
s2
```

Out[134]:

```
p    1
q    2
r    3
s    4
Name: one, dtype: int64
```

In [135]:

```
df1 = pd.DataFrame(s2)
df1
```

Out[135]:

	one
p	1
q	2
r	3
s	4

**2. Locate an open source data from the web (e.g. <https://www.kaggle.com> (<https://www.kaggle.com>)). Provide a clear description of the data and its source (i.e., URL of the web site).**

### 3. Load the Dataset into pandas data frame

Real power- Import from different formats <http://pandas.pydata.org/pandas-docs/version/0.20/io.html> (<http://pandas.pydata.org/pandas-docs/version/0.20/io.html>)

In [136]:

```
df2 = pd.read_csv("/content/sample_data/california_housing_test.csv")  
#dataframe_name = pd.read_<format>(filename) In
```

[137]:

```
df2.head(10)
```

Out[137]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	househ
0	-122.05	37.37	27.0	3885.0	661.0	1537.0	6
1	-118.30	34.26	43.0	1510.0	310.0	809.0	2

2	-117.81	33.78	27.0	3589.0	507.0	1484.0	4
3	-118.36	33.82	28.0	67.0	15.0	49.0	
4	-119.67	36.33	19.0	1241.0	244.0	850.0	2
5	-119.56	36.51	37.0	1018.0	213.0	663.0	2
6	-121.43	38.63	43.0	1009.0	225.0	604.0	2
7	-120.65	35.48	19.0	2310.0	471.0	1341.0	4
8	-122.84	38.40	15.0	3080.0	617.0	1446.0	5
9	-118.02	34.08	31.0	2402.0	632.0	2830.0	6

In [138]:

```
df2.tail(3)
```

Out[138]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	hous
2997	-119.70	36.30	10.0	956.0	201.0	693.0	
2998	-117.12	34.10	40.0	96.0	14.0	46.0	
2999	-119.63	34.42	42.0	1765.0	263.0	753.0	

In [139]:

```
df2['median_house_value_new']=df2['median_house_value']+111
```

In [140]:

```
df2.tail(3)
```

Out[140]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	hous
2997	-119.70	36.30	10.0	956.0	201.0	693.0	
2998	-117.12	34.10	40.0	96.0	14.0	46.0	
2999	-119.63	34.42	42.0	1765.0	263.0	753.0	

In [141]:

```
# write
# <dataframe's name>.to_<file_format>(<file_name>)
```

In [142]:

```
df2.to_json('data1.json')
```

In [143]:

```
#If our age dataset is an year old  
#df[age_now]= df[age]+1  
#df[salary_increment]=df[salary]+5000
```

In [144]:

```
#df1['value'] = df1['num']*2  
# internally for each value in column num perform each_value*2 and save it as the corre  
sponding  
# result in the value column  
#df1
```

In [145]:

```
len(df2['total_rooms'])
```

Out[145]:

3000 In

[146]:

```
df2['total_rooms'].count()
```

Out[146]:

3000

```
In [147]:
```

```
df2['total_rooms'].mean()
```

```
Out[147]:
```

```
2599.578666666667
```

```
In [148]:
```

```
df2['total_rooms'].sum()
```

```
Out[148]:
```

```
7798736.0
```

```
In [149]:
```

```
df2['total_rooms'].median()
```

```
Out[149]:
```

```
2106.0 In
```

```
[150]:
```

```
df2['total_rooms'].std()
```

```
Out[150]:
```

```
2155.59333162558
```

```
In [151]:
```

```
df2['total_rooms'].min()
```

```
Out[151]:
```

```
6.0
```

```
In [152]:
```

```
df2['total_rooms'].max()
```

```
Out[152]:
```

```
30450.0
```

```
In [153]:
```

```
df2['total_rooms'].describe()
```

```
Out[153]:
```

```
count      3000.000000
mean       2599.578667 std
2155.593332 min
6.000000 25%
1401.000000
50%       2106.000000 75%
3129.000000 max      30450.000000
Name: total_rooms, dtype: float64
```

```
In [154]:
```

```
df2['total_rooms'].cumsum()
```

```
Out[154]:
```

```
0          3885.0
1          5395.0
2          8984.0
3          9051.0
4         10292.0
...
2995      7790662.0
2996      7795919.0
2997      7796875.0
2998      7796971.0
2999      7798736.0
Name: total_rooms, Length: 3000, dtype: float64
```

```
In [155]:
```

```
# When you give the whole dataframe, then all numerical columns will be analysis
df2.mean()
```

```
Out[155]:
```

```
longitude          -119.589200
latitude            35.635390
housing_median_age  28.845333
total_rooms        2599.578667
total_bedrooms     529.950667
population         1402.798667
households         489.912000
median_income       3.807272
median_house_value  205846.275000
median_house_value_new 205957.275000
dtype: float64
```

```
In [156]:
```

```
df2.describe()
```

Out[156]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	populati
count	3000.000000	3000.00000	3000.000000	3000.000000	3000.000000	3000.00
mean	-119.589200	35.63539	28.845333	2599.578667	529.950667	1402.79
std	1.994936	2.12967	12.555396	2155.593332	415.654368	1030.54
min	-124.180000	32.56000	1.000000	6.000000	2.000000	5.00
25%	-121.810000	33.93000	18.000000	1401.000000	291.000000	780.00
50%	-118.485000	34.27000	29.000000	2106.000000	437.000000	1155.00
75%	-118.020000	37.69000	37.000000	3129.000000	636.000000	1742.75
max	-114.490000	41.92000	52.000000	30450.000000	5419.000000	11935.00

In [157]: df =

pd.read\_csv("/content/sample\_data/california\_housing\_test.csv") In

[158]:

df.describe()

Out[158]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	populati
count	3000.000000	3000.00000	3000.000000	3000.000000	3000.000000	3000.00
mean	-119.589200	35.63539	28.845333	2599.578667	529.950667	1402.79
std	1.994936	2.12967	12.555396	2155.593332	415.654368	1030.54
min	-124.180000	32.56000	1.000000	6.000000	2.000000	5.00
25%	-121.810000	33.93000	18.000000	1401.000000	291.000000	780.00
50%	-118.485000	34.27000	29.000000	2106.000000	437.000000	1155.00
75%	-118.020000	37.69000	37.000000	3129.000000	636.000000	1742.75
max	-114.490000	41.92000	52.000000	30450.000000	5419.000000	11935.00

In [159]:

df.columns

Out[159]:

```
Index(['longitude', 'latitude', 'housing_median_age', 'total_rooms',  
      'total_bedrooms', 'population', 'households', 'median_income',  
      'median_house_value'],  
      dtype='object')
```



```
In [160]:
```

```
df['longitude']
```

```
Out[160]:
```

```
0      -122.05
1      -118.30
2      -117.81
3      -118.36
4      -119.67
...
2995   -119.86
2996   -118.14
2997   -119.70
2998   -117.12
2999   -119.63
Name: longitude, Length: 3000, dtype: float64
```

```
In [161]:
```

```
df.longitude
```

```
Out[161]:
```

```
0      -122.05
1      -118.30
2      -117.81
3      -118.36
4      -119.67
...
2995   -119.86
2996   -118.14
2997   -119.70
2998   -117.12
2999   -119.63
Name: longitude, Length: 3000, dtype: float64
```

```
In [162]:
```

```
df.iloc[:,1:3]
```

```
Out[162]:
```

	latitude	housing_median_age
0	37.37	27.0
1	34.26	43.0
2	33.78	27.0
3	33.82	28.0
4	36.33	19.0

--	--	--

...	...	...
2995	34.42	23.0
2996	34.06	27.0
2997	36.30	10.0
2998	34.10	40.0
2999	34.42	42.0
3000	rows × 2 columns	

## 4.Data Preprocessing:

check for missing values in the data using pandas , describe() function to get some initial statistics. Filling missing values using fillna(), replace() and interpolate()