# **Data Processing with Pandas**

Changing working directory.

```
In [72]:
```

% cd D:\Data Science\GitHub\Python Learning\Python-for-Data-Science\Data Files\Python Learning

D:\Data Science\GitHub\Python Learning\Python-for-Data-Science\Data Files
\Python Learning

### 1. The Series Data Structure

Series is the <u>one-dimensional</u> data structure in *pandas*.

```
In [73]:
```

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

```
In [74]:
```

```
animals = ['Tiger', 'Bear', 'Moose']
pd.Series(animals)
```

### Out[74]:

0 Tiger
1 Bear
2 Moose
dtype: object

#### In [75]:

```
type(animals)
```

#### Out[75]:

list

Handling Missing Values

```
In [76]:
```

```
animals = ['Tiger', 'Bear', None]
pd.Series(animals)
```

#### Out[76]:

```
0 Tiger1 Bear
```

2 None
dtype: object

```
In [77]:
```

```
numbers = [1, 2, None]
pd.Series(numbers)
Out[77]:
```

0 1.0 1 2.0 2 NaN dtype: float64

Here pandas series will turn the missing value to be NaN, which is different from None.

To check if a value is Nan,

```
In [78]:
```

```
np.isnan(pd.Series(numbers)[2])
```

#### Out[78]:

True

#### · Creating Series from Dictionaries

#### In [79]:

```
sports = {'Archery': 'Bhutan',
          'Golf': 'Scotland',
          'Sumo': 'Japan',
          'Taekwondo': 'South Korea'}
s = pd.Series(sports)
S
```

#### Out[79]:

Archery **Bhutan** Golf Scotland Sumo Japan Taekwondo South Korea dtype: object

#### Creating Series with Separate Index Creation

```
In [80]:
```

```
s = pd.Series(['Tiger', 'Bear', 'Moose'], index=['India', 'America', 'Canada'])
S
```

#### Out[80]:

India Tiger America Bear Canada Moose dtype: object

If the number of indexes are not aligned with keys, *pandas* will assign NaN to those missing values.

```
In [81]:
```

#### Out[81]:

Golf Scotland Sumo Japan Hockey NaN dtype: object

### 1.1 Querying a Series

#### In [82]:

#### Out[82]:

Archery Bhutan
Golf Scotland
Sumo Japan
Taekwondo South Korea
dtype: object

Using iloc and loc attribute to query series

```
In [83]:
```

```
s.iloc[3]
Out[83]:
'South Korea'
In [84]:
s.loc['Golf']
```

Out[84]:

'Scotland'

Be careful when using the indexing operator on the series itself. And the safer option is to be more explicit and use the iloc or loc attributes directly.

#### Appending New Values

#### In [85]:

```
s = pd.Series([1, 2, 3])
s.loc['Animal'] = 'Bears'
s
```

#### Out[85]:

0 1
1 2
2 3
Animal Bears
dtype: object

#### • Non-unique Index

#### In [86]:

#### In [87]:

```
print(all_countries)
print(original_sports)
```

Archery Bhutan Golf Scotland Sumo Japan Taekwondo South Korea Cricket Australia Cricket Barbados Pakistan Cricket Cricket England dtype: object Archery **Bhutan** Golf Scotland Sumo Japan South Korea Taekwondo

dtype: object

**Note** that even though we applied .append method to original\_sports, the original series was not changed.

## 2. DataFrame Data Structure

Series is the two-dimensional data structure in pandas.

### 2.1 Query from DataFrame

#### In [88]:

#### Out[88]:

	Name	Item Purchased	Cost
Store 1	Chris	Dog Food	22.5
Store 2	Kevyn	Kitty Litter	2.5
Store 3	Vinod	Bird Seed	5.0

Query element from DataFrame using loc.

#### In [89]:

```
df.loc['Store 3']
```

#### Out[89]:

Name Vinod
Item Purchased Bird Seed
Cost 5
Name: Store 3, dtype: object

• Index can be non-unique

```
In [90]:
```

```
df2 = pd.DataFrame([purchase_1, purchase_2, purchase_3], index = ['Store 1', 'Store 1',
'Store 3'])
df2.loc['Store 1']
```

#### Out[90]:

	Name	Item Purchased	Cost
Store 1	Chris	Dog Food	22.5
Store 1	Kevyn	Kitty Litter	2.5

· Selecting columns

#### In [91]:

```
df.loc[:, ['Name', 'Cost']]
```

#### Out[91]:

	Name	Cost
Store 1	Chris	22.5
Store 2	Kevyn	2.5
Store 3	Vinod	5.0

NOTE: Avoid Query by Chaining Selecting
 This will creating a copy of dataset, which will be quit inefficient.

#### In [92]:

```
df.loc['Store 1']['Cost']
```

Out[92]:

22.5

# 2.2 Dropping & Adding Data from DataFrame

· Dropping Rows

#### In [93]:

```
df.drop('Store 1')
```

#### Out[93]:

	Name	Item Purchased	Cost
Store 2	Kevyn	Kitty Litter	2.5
Store 3	Vinod	Bird Seed	5.0

**NOTE:** The original dataset is untouched.

#### In [94]:

df

#### Out[94]:

	Name	Item Purchased	Cost
Store 1	Chris	Dog Food	22.5
Store 2	Kevyn	Kitty Litter	2.5
Store 3	Vinod	Bird Seed	5.0

• Dropping Columns

#### In [95]:

```
copy_df = df.copy()
del copy_df['Name']
copy_df
```

#### Out[95]:

	Item Purchased	Cost
Store 1	Dog Food	22.5
Store 2	Kitty Litter	2.5
Store 3	Bird Seed	5.0

• Adding Columns

#### In [96]:

```
df['Location'] = None
df
```

#### Out[96]:

	Name	Item Purchased	Cost	Location
Store 1	Chris	Dog Food	22.5	None
Store 2	Kevyn	Kitty Litter	2.5	None
Store 3	Vinod	Bird Seed	5.0	None

# 3. DataFrame Indexing and Loading

**NOTE:** If you want to explicitly use a **copy**, then you should consider calling the **copy method** on the DataFrame for it first.

## 3.1 Creating DataFrame from .csv files.

#### In [97]:

```
%cd D:\Data Science\GitHub\Python Learning\Python-for-Data-Science\Data Files\Python Le
arning

df = pd.read_csv('expense_report.csv', skiprows = 0)
df.head()
```

D:\Data Science\GitHub\Python Learning\Python-for-Data-Science\Data Files
\Python Learning

#### Out[97]:

	TRXN_MONTH	TRXN_DT	TRXN_AMT	MERCHT_NAME	CATG
0	Mar-19	01/03/2019	61.20	VIRGIN ACTIVE	Entertainment
1	Mar-19	04/03/2019	1.31	LOON FUNG LIMITED	Glossary
2	Mar-19	04/03/2019	16.89	LOON FUNG LIMITED	Glossary
3	Mar-19	04/03/2019	20.00	LNK CAPITAL ONE	Other
4	Mar-19	05/03/2019	2.85	SAINSBURY'S SPRMRKTS LT NOTTINGHAM	Glossary

• Show Names of All Columns in the DataFrame.

#### In [98]:

```
df.columns
```

#### Out[98]:

```
Index(['TRXN_MONTH', 'TRXN_DT', 'TRXN_AMT', 'MERCHT_NAME', 'CATG'], dtype
='object')
```

# 4. Querying a DataFrame

# 4.1 Querying DataFrame by Boolean Masking

• Step 1. creating Boolean masking array.

# In [99]:

df['TRXN\_AMT'] > 5

# Out[99]:

	-
0	Т
0	True
1	False
2	True
3	True
4	False
5	True
<i>-</i>	
6	True
7	True
8	True
9	True
10	True
11	
	False
12	True
13	True
14	True
15	True
16	True
17	True
18	True
19	True
20	True
21	False
22	True
23	True
24	False
25	True
26	True
27	False
_,	
20	
28	False
28 29	False
29	False False 
29 1177	False False  True
29	False False 
29 1177 1178	False False  True False
29 1177 1178 1179	False False True False False
1177 1178 1179 1180	False False True False False False
29 1177 1178 1179	False False True False False
29 1177 1178 1179 1180 1181	False False True False False False True
29 1177 1178 1179 1180 1181 1182	False False True False False True False True False
29 1177 1178 1179 1180 1181 1182 1183	False False False False False False True False False False
29 1177 1178 1179 1180 1181 1182	False False True False False True False True False
29 1177 1178 1179 1180 1181 1182 1183 1184	False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185	False True
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186	False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185	False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187	False True False True False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188	False False False False False False False False False True False True False True False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187	False True False True False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188	False False False False False False False False False True False True False True False True
1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190	False False False False False False False False False True False True False True False True False
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191	False False False False False False False False True False True False True True True
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192	False False False False False False False False False True False True False True False True False
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192	False False False False False False False False True False True False True True True
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193	False False False False False False False False True False True False True True True True
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194	False False False False False False False False True False True False True True True True True
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193	False False False False False False False False True False True False True True True True
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195	False False False False False False False False True False True True True True True True True
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196	False False False False False False False False True False True True True True True True True Tru
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197	False False True False False False False False False True False True True True True True True True Tru
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196	False False False False False False False False True False True True True True True True True Tru
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198	False False True False False False False False True False True True True True True True True Tru
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199	False False False False False False False False True False True True True True True True True Tru
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199 1200	False False True False False False False False True False True True True True True True True Tru
29 1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199	False False True False False False False False True False True True True True True True True Tru
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199 1200 1201	False False True False False False False False False True False True True True True True True True Tru
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199 1200 1201 1202	False False True False False False False False False True False True True True True True True True Tru
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199 1200 1201 1202 1203	False False True False False False False False False True False True True True True True True True Tru
29  1177 1178 1179 1180 1181 1182 1183 1184 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199 1200 1201 1202	False False True False False False False False False True False True True True True True True True Tru

1205 True1206 True

Name: TRXN\_AMT, Length: 1207, dtype: bool

• Step 2. overlay the masking on the DataFrame by where function.

#### In [100]:

```
df_500 = df.where(df['TRXN_AMT'] > 5)
df_500.head()
```

#### Out[100]:

	TRXN_MONTH	TRXN_DT	TRXN_AMT	MERCHT_NAME	CATG
0	Mar-19	01/03/2019	61.20	VIRGIN ACTIVE	Entertainment
1	NaN	NaN	NaN	NaN	NaN
2	Mar-19	04/03/2019	16.89	LOON FUNG LIMITED	Glossary
3	Mar-19	04/03/2019	20.00	LNK CAPITAL ONE	Other
4	NaN	NaN	NaN	NaN	NaN

**NOTE:** rows that do not match the condition will give NaN values. To drop those NaN rows.

### In [101]:

```
df_500 = df_500.dropna()
df_500.head()
```

#### Out[101]:

CATG	MERCHT_NAME	TRXN_AMT	TRXN_DT	TRXN_MONTH	
Entertainment	VIRGIN ACTIVE	61.20	01/03/2019	Mar-19	0
Glossary	LOON FUNG LIMITED	16.89	04/03/2019	Mar-19	2
Other	LNK CAPITAL ONE	20.00	04/03/2019	Mar-19	3
Glossary	TESCO STORE 5660 5660TE WELWYN GARDEN C	6.49	06/03/2019	Mar-19	5
Eating Out	SHANGHAI SHANGHAI SHANG NOTTINGHAM	35.40	07/03/2019	Mar-19	6

• However, there is a quicker way to do that.

#### In [102]:

```
df_500 = df[df['TRXN_AMT'] > 5]
df_500.head()
```

#### Out[102]:

CATG	MERCHT_NAME	TRXN_AMT	TRXN_DT	TRXN_MONTH	
Entertainment	VIRGIN ACTIVE	61.20	01/03/2019	Mar-19	0
Glossary	LOON FUNG LIMITED	16.89	04/03/2019	Mar-19	2
Other	LNK CAPITAL ONE	20.00	04/03/2019	Mar-19	3
Glossary	TESCO STORE 5660 5660TE WELWYN GARDEN C	6.49	06/03/2019	Mar-19	5
Eating Out	SHANGHAI SHANGHAI SHANG NOTTINGHAM	35.40	07/03/2019	Mar-19	6

# 5. Indexing DataFrame

# **5.1 Setting Index of DataFrame**

• Re-indexing DataFrame

#### In [103]:

```
df = df.set_index('CATG')
df.head()
```

#### Out[103]:

	TRXN_MONTH	TRXN_DT	TRXN_AMT	MERCHT_NAME
CATG				
Entertainment	Mar-19	01/03/2019	61.20	VIRGIN ACTIVE
Glossary	Mar-19	04/03/2019	1.31	LOON FUNG LIMITED
Glossary	Mar-19	04/03/2019	16.89	LOON FUNG LIMITED
Other	Mar-19	04/03/2019	20.00	LNK CAPITAL ONE
Glossary	Mar-19	05/03/2019	2.85	SAINSBURY'S SPRMRKTS LT NOTTINGHAM

• Reset Index by Number Series

#### In [104]:

```
df = df.reset_index()
df.head()
```

#### Out[104]:

	CATG	TRXN_MONTH	TRXN_DT	TRXN_AMT	MERCHT_NAME
0	Entertainment	Mar-19	01/03/2019	61.20	VIRGIN ACTIVE
1	Glossary	Mar-19	04/03/2019	1.31	LOON FUNG LIMITED
2	Glossary	Mar-19	04/03/2019	16.89	LOON FUNG LIMITED
3	Other	Mar-19	04/03/2019	20.00	LNK CAPITAL ONE
4	Glossary	Mar-19	05/03/2019	2.85	SAINSBURY'S SPRMRKTS LT NOTTINGHAM

## 5.2 Multi-level Index in DataFrame

• Changing the DataFrame to be Multi-level Indexed

#### In [105]:

```
df['TRXN_MONTH'].unique()
df = df[df['TRXN_MONTH'] == 'Oct-19']
df.head()
```

#### Out[105]:

	CATG	TRXN_MONTH	TRXN_DT	TRXN_AMT	MERCHT_NAME
807	Transport	Oct-19	01/10/2019	5.30	TFL TRAVEL CHARGE TFL.GOV.UK/CP
808	Transport	Oct-19	01/10/2019	2.90	TFL TRAVEL CHARGE TFL.GOV.UK/CP
809	Glossary	Oct-19	01/10/2019	2.50	WAITROSE CONVENIENCE NOTTINGHAM
810	Eating Out	Oct-19	01/10/2019	6.55	STARBUCKS TESCO LONDON
811	Transport	Oct-19	01/10/2019	5.70	TFL TRAVEL CHARGE TFL.GOV.UK/CP

```
In [106]:
```

```
df = df.set_index(['TRXN_DT', 'CATG'])
df.head()
```

## Out[106]:

		TRXN_MONTH	TRXN_AMT	MERCHT_NAME
TRXN_DT	CATG			
01/10/2019	Transport	Oct-19	5.30	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	2.90	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Glossary	Oct-19	2.50	WAITROSE CONVENIENCE NOTTINGHAM
	Eating Out	Oct-19	6.55	STARBUCKS TESCO LONDON
	Transport	Oct-19	5.70	TFL TRAVEL CHARGE TFL.GOV.UK/CP

· Querying the data

## In [107]:

```
df.loc[[('01/10/2019', 'Transport'), ('02/10/2019', 'Glossary')]]
```

## Out[107]:

		TRXN_MONTH	TRXN_AMT	MERCHT_NAME
TRXN_DT	CATG			
01/10/2019	Transport	Oct-19	5.30	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	2.90	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	5.70	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	4.40	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	5.30	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	5.80	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	5.70	TFL TRAVEL CHARGE TFL.GOV.UK/CP
02/10/2019	Glossary	Oct-19	3.91	CAPITAL ONE

# 6. Missing Values

# In [111]:

df.fillna
df.head()

# Out[111]:

		TRXN_MONTH	TRXN_AMT	MERCHT_NAME
TRXN_DT	CATG			
01/10/2019	Transport	Oct-19	5.30	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Transport	Oct-19	2.90	TFL TRAVEL CHARGE TFL.GOV.UK/CP
	Glossary	Oct-19	2.50	WAITROSE CONVENIENCE NOTTINGHAM
	Eating Out	Oct-19	6.55	STARBUCKS TESCO LONDON
	Transport	Oct-19	5.70	TFL TRAVEL CHARGE TFL.GOV.UK/CP