**Objective**: This worksheet introduces *DataFrame* object of Pandas. *Series* object reviewed in the previous worksheet is analogous to one-dimensional array and is helpful to create an object for time series type of data. *DataFrame* is analogous to two-dimensional array and it is helpful to store relational data.

- To elaborate the concept, let us create a data frame using a two dimension NumPy array and **DataFrame()** function of Pandas package.
- First create a two dimensional NumPy array of size 5x3 of random numbers from 0 to 10.

• Let us now create a Pandas data frame from this two dimensional array using *DataFrame()* function.

```
df = pd.DataFrame(twoDdarray)
df

0 1 2
0 6 5 3
1 7 7 3
2 9 2 8
3 2 4 1
4 7 0 2
```

- Notice from the above output there is a name for each row (0 to 4) and a name (0 to 2) for each column.
- Data frames values can be accessed by the *values* property of the object. The returned values is a NumPy array.

• In the data frames, row names are called index and column names are called columns that can also be accessed by *index* and *columns* properties respectively.

```
print (df.index)
print (df.columns)

RangeIndex(start=0, stop=5, step=1)
RangeIndex(start=0, stop=3, step=1)
```

- Notice from the above output that row is ranging from 0 to 5 with a step of 1, while columns are ranging from 0 to 3 with a step of 1. End-values 5 and 3 are exclusive.
- Programmer can also provide row and column names explicitly. In the example below, rows are named from R1 to R5 using and columns are named from C1 to C3.

```
df = pd.DataFrame(twoDdarray,
                   index = ['R1', 'R2', 'R3', 'R4', 'R5'],
                   columns = ['C1', 'C2', 'C3'])
print (df)
    C1
        C2
             C3
         5
R1
     6
              3
         7
R2
     7
              3
         2
R3
     9
              8
     2
         4
              1
R4
R5
     7
              2
```

• If a data frame is already created (let us say in a variable x) with default index and column names, it can be changed later with explicit programmer provide names as shown below:

```
x.index = ['Row-1', 'Row-2', 'Row-3']
x.columns = ['Col-1', Col-2', 'Col-3']
```

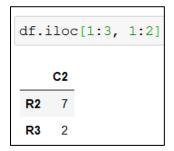
• Individual data element can be accessed using *loc* and *iloc*. For example the data element (value 0) which is present in the 5<sup>th</sup> row and 2<sup>nd</sup> column can be accessed as follows:

```
df.loc['R5', 'C2']
0
df.iloc[4,1]
0
```

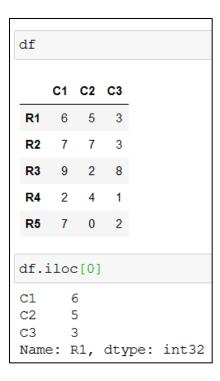
Data elements can also be accessed using slicing. In case of *loc* the end value is inclusive but in case of *iloc* it is exclusive.

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df.loc['R2':'R4', 'C2':'C3']				
		C2	C3	
	R2	7	3	
	R3	2	8	
	R4	4	1	



- Data frames can be considered and viewed as the collection of series. A data frame would have m series where m is the count of rows in the data frames.
- In the example below what is R1 alone? It is a pandas Series object where indices are C1, C2 and C3.



• Similarly, the elements of other rows (R2 to R5) are also Series objects where the element indices are same C1 to C3. This can also be verified checking the data type.

```
type (df.iloc[0])
pandas.core.series.Series
```

- With the same logic each a data frame would have n series where n is the count of columns in the data frames.
- In the example below what is C1 alone? It is pandas Series object where indices are R1 to R5. Notice the subtle difference while slicing the series from a column:

```
df.iloc[:,0]

R1   6
R2   7
R3   9
R4   2
R5   7
Name: C1, dtype: int32

type (df.iloc[:,0])
pandas.core.series.Series
```

A dataframe can be transposed also using the transpose() function.

