

## **PANDAS**

Pandas stand for Panel Data System

Pandas is an open source library for data analysis, Data manipulation and Data Visualization.

(OR) Pandas provide powerful data structures for data analysis, time series and statistics.

Pandas works on the top numpy and matplotlib.

### **Features of pandas**

1. Handling huge amount data
2. Missing Data
3. Cleaning up data
4. Alignment and indexing
5. Merging and joining
6. Grouping and Visualizing data
7. Time Series Functionality
8. Allows to load data from multiple file formats
9. Input and Output Tools

Pandas library is used by scikit-learn for ML

### **Applications of Pandas**

1. Recommendation Systems
2. Stock Prediction
3. Big Data and Data Science
4. NLP (Natural Language Processing)
5. Statistics and Analytics
6. Neuroscience

### **Important data structures of Pandas are,**

1. Series
2. DataFrame

Q: What is data analysis?

Data analysis is process of collecting, transforming, cleaning and modeling the data with goal of discovering required information.

Data analysis process consists of the following steps.

1. Data Requirement Specifications
2. Data Collection
3. Data Processing
4. Data Cleaning

5. Data Analysis
6. Communication

### **What is Series?**

Pandas series is a one dimensional array object, this object can hold data of any type. It can be integers, floats, string or python objects.

Pandas series represents or equal to a column in any data base (MsExcel, Oracle, MySQL, SQLServer,...)

### **What is DataFrame?**

DataFrame is a two dimensional array object or data structure. Data stored tabular format, which is rows and columns.

The Dataframe consist of 3 components.

1. Data
2. Rows
3. Columns

### **How to install pandas?**

Other than jupyter and googlecolab, it is required to install pandas lib.

pip install pandas

### **Pandas Series**

Series is single dimension array like object with homogeneous or heterogeneous data.

Series object can be created in different ways.

1. Using array
2. Using Dictionary
3. Using Scalar values
4. Using other iterables

Series is name of the class or type which is used to construct Series object.

### **Syntax: Series(data,index,dtype)**

Data : the source using which series object is created

Index : index values must hashable and must be unique

dtype: type of the series is defined using dtype.

### **Creating Empty Series**

```

▶ import pandas as pd
import numpy as np
s1=pd.Series(dtype=np.int8)
print(s1)

```

```

↳ Series([], dtype: int8)

```

## Creating Series using List object

```

▶ s2=pd.Series([10,20,30,40,50])
print(s2)
s3=pd.Series([10,20,30,40,50],index=['a','b','c','d','e'])
print(s3)

```

```

↳ 0    10
   1    20
   2    30
   3    40
   4    50
dtype: int64
a    10
b    20
c    30
d    40
e    50
dtype: int64

```

## Creating Series using ndarray

```

▶ a=np.ndarray(shape=(5,))
i=0
for value in range(10,60,10):
    a[i]=value
    i+=1
print(a)
print(type(a))
s=pd.Series(a)
print(s)

```

```

↳ [10. 20. 30. 40. 50.]
<class 'numpy.ndarray'>
0    10.0
1    20.0
2    30.0
3    40.0
4    50.0
dtype: float64

```

## Creating Series Using Dictionary

We can create series using dictionary (OR) we can pass the dictionary object to series.

Series object is using dictionary values as data and dictionary keys as index labels.

```
sales_dict={2018:50000,2019:60000,2020:75000}
s=pd.Series(sales_dict)
print(s)
emp_dict={'naresh':5000,'suresh':6000,'kishore':9000}
s=pd.Series(emp_dict)
print(s)
```

```
2018    50000
2019    60000
2020    75000
dtype: int64
naresh     5000
suresh     6000
kishore    9000
dtype: int64
```

## Creating Series using Scalar values

If the series is created using scalar values we must define index. This index defines the length of series.

```
s=pd.Series(15,index=[0,1,2,3,4])
print(s)
```

```
0    15
1    15
2    15
3    15
4    15
dtype: int64
```

## Accessing Data from Series

Series is index based collection, we can read and manipulate data using index.

This index starts with 0.

```

s1=pd.Series([100,200,300,400,500])
print(s1)
print(s1[0],s1[1],s1[2],s1[3],s1[4])
s2=pd.Series([1000,2000,3000,4000,5000],index=['a','b','c','d','e'])
print(s2['a'],s2['b'],s2['c'],s2['d'],s2['e'])
print(s2[0],s2[1],s2[2],s2[3],s2[4])

```

```

0    100
1    200
2    300
3    400
4    500
dtype: int64
100 200 300 400 500
1000 2000 3000 4000 5000
1000 2000 3000 4000 5000

```

## Reading multiple elements/values from series

Series allows reading multiple elements by defining index labels within list.

```

s1=pd.Series(range(100,1000,100))
print(s1)
print(s1[[0,3,6,8]])
s2=pd.Series([100,200,300,400,500],index=['a','b','c','d','e'])
print(s2)
print(s2[['a','c','e']])

```

```

0    100
1    200
2    300
3    400
4    500
5    600
6    700
7    800
8    900
dtype: int64
0    100
3    400
6    700
8    900
dtype: int64
a    100
b    200
c    300
d    400
e    500

```

✓ 0s completed at 7:01 PM

Series allows slicing, to read multiple elements/values.

```

▶ s1=pd.Series(range(100,1000,100))
print(s1)
print(s1[:3])
print(s1[-3:])
print(s1[-1::-1])

```

```

❏ 0    100
   1    200
   2    300
   3    400
   4    500
   5    600
   6    700
   7    800
   8    900
dtype: int64
0    100
1    200
2    300
dtype: int64
6    700
7    800
8    900
dtype: int64
8    900
7    800
6    700
5    600

```

✓ Done - completed at 7:05 PM

## DataFrame

DataFrame is two dimensional array object with heterogeneous data. In DataFrame data is stored in the form of rows and columns.

### How to create DataFrame?

DataFrame can be created in different ways.

1. Series
2. Lists
3. Dictionary
4. Numpy array
5. From another dataframe
6. Data can read from files or database

“DataFrame” is type or class name, to create dataframe object

### Syntax:

DataFrame(data,index,columns,dtype)

data : data is taken from various sources

Index : row labels

columns : columns labels

dtype: data type of each column

## Creating empty dataframe

```
import pandas as pd
#creating empty dataframe
df=pd.DataFrame()
print(df)
```

```
Empty DataFrame
Columns: []
Index: []
```

## Creating DataFrame using dictionary

Dictionary consist of key and values.

Dictionary keys as columns headers and values are columns values

```
d={'empno':[1,2,3,4,5], 'ename':['naresh', 'suresh', 'rajesh', 'kishore', 'raman'], 'sal':[5000,6000,7000,9000,6000]}
df=pd.DataFrame(d)
print(df)
```

	empno	ename	sal
0	1	naresh	5000
1	2	suresh	6000
2	3	rajesh	7000
3	4	kishore	9000
4	5	raman	6000

## Create DataFrame using List

A nested list represents the content of dataframe.

Each list within list is represented as row.

```
person_list=[['naresh',50],['suresh',45],['kishore',35]]
df=pd.DataFrame(person_list,columns=['name','age'],dtype=float)
print(df)
```

	name	age
0	naresh	50.0
1	suresh	45.0
2	kishore	35.0

## DataFrame created with missing data

Missing data is identified with NaN(Not a Number)

```
data=[['naresh',45],['suresh',56],['kishore',65],['rajesh']]
df=pd.DataFrame(data,columns=['name','age'])
print(df)
```

	name	age
0	naresh	45.0
1	suresh	56.0
2	kishore	65.0
3	rajesh	NaN

```
data=[{'name':'naresh','age':45},{'name':'kishore'},{'name':'suresh'},{'age':50},{}]
df=pd.DataFrame(data,index=['p1','p2','p3','p4','p5'])
print(df)
```

```
name age
p1 naresh 45.0
p2 kishore NaN
p3 suresh NaN
p4 NaN 50.0
p5 NaN NaN
```

## Selecting Data

1. Row Selection
2. Column Selection

### Column Selection

Selecting columns from DataFrame can be done using column header.

```
data=[{'name':'naresh','age':45},{'name':'kishore'},{'name':'suresh'},{'age':50},{}]
df=pd.DataFrame(data)
print(df)
c1=df['name']
c2=df['age']
print(type(c1),type(c2))
print(c1,c2)
```

```
name age
0 naresh 45.0
1 kishore NaN
2 suresh NaN
3 NaN 50.0
4 NaN NaN
<class 'pandas.core.series.Series'> <class 'pandas.core.series.Series'>
0 naresh
1 kishore
2 suresh
3 NaN
4 NaN
Name: name, dtype: object 0 45.0
1 NaN
2 NaN
3 50.0
4 NaN
Name: age, dtype: float64
```

## Reading multiple columns from DataFrame

In order to read multiple columns, the column names must be defined as a list. It return multiple columns as a dataframe.

When we single column it read as a series.



```
data={'a':[1,2,3,4,5], 'b':[100,200,300,400,500], 'c':[1000,2000,3000,4000,5000], 'd':[10000,20000,30000,40000,50000]}
df=pd.DataFrame(data)
print(df)
print(df[['a','c']])
r=df[['a','c']]
print(r)
print(type(r))
```

```

a      b      c      d
0      1    100   1000  10000
1      2    200   2000  20000
2      3    300   3000  30000
3      4    400   4000  40000
4      5    500   5000  50000

a      c
0      1  1000
1      2  2000
2      3  3000
3      4  4000
4      5  5000

a      c
0      1  1000
1      2  2000
2      3  3000
3      4  4000
4      5  5000
<class 'pandas.core.frame.DataFrame'>
```

## Column Addition

Adding new column to the existing DataFrame.

```
data={'col1':pd.Series([1,2,3]),
      'col2':pd.Series([10,20,30])}
df=pd.DataFrame(data)
print(df)
df['col3']=pd.Series([100,200,300])
print(df)
df['col4']=df['col2']+df['col3']
print(df)
```

```

col1  col2
0      1   10
1      2   20
2      3   30

col1  col2  col3
0      1   10  100
1      2   20  200
2      3   30  300

col1  col2  col3  col4
0      1   10  100  110
1      2   20  200  220
2      3   30  300  330
```

## Column Deletion

The column deletion is done using del keyword.

It allows deleting one or more than one columns.

The column is deleted with column name or column labels.

```

import pandas as pd
l=[[ 'naresh',45],[ 'suresh',50],[ 'ramesh',60]]
df=pd.DataFrame(l,columns=[ 'name', 'age' ])
print(df)
del df['name']
print(df)

```

```

name age
0  naresh  45
1  suresh  50
2  ramesh  60
age
0  45
1  50
2  60

```

## Row Selection, Addition and Deletion

Each row is identified with index or label. We can read rows from dataframe using index or label.

DataFrame provide two methods to perform this operation.

1. loc
2. iloc

loc() is used to read the rows using label

iloc() is used to read the rows using index

```

student_data={'rno':[1,2,3,4,5],
              'name':['naresh','suresh','ramesh','rajesh','kiran']}
df=pd.DataFrame(student_data,index=['s1','s2','s3','s4','s5'])
print(df)
print(df.loc['s1'])
row=df.loc['s1']
print(row)
print(type(row))
print(row[0],row[1])

```

```

rno name
s1  1  naresh
s2  2  suresh
s3  3  ramesh
s4  4  rajesh
s5  5  kiran
rno      1
name  naresh
Name: s1, dtype: object
rno      1
name  naresh
Name: s1, dtype: object
<class 'pandas.core.series.Series'>
1  naresh

```

```

▶ student_data={'rno':[1,2,3,4,5],
                'name':['naresh','suresh','ramesh','rajesh','kiran']}
df=pd.DataFrame(student_data,index=['s1','s2','s3','s4','s5'])
print(df)
print(df.iloc[0])
print(df.iloc[1])

```

```

▶
   rno  name
s1    1 naresh
s2    2 suresh
s3    3 ramesh
s4    4 rajesh
s5    5 kiran
rno    1
name  naresh
Name: s1, dtype: object
rno    2
name  suresh
Name: s2, dtype: object

```

Slicing is used to read more than one row

```

▶ student_data={'rno':[1,2,3,4,5],
                'name':['naresh','suresh','ramesh','rajesh','kiran']}
df=pd.DataFrame(student_data,index=['s1','s2','s3','s4','s5'])
print(df)
print(df[0:3])
print(df[0::2])
df1=df[0:3]
print(type(df1))

```

```

▶
   rno  name
s1    1 naresh
s2    2 suresh
s3    3 ramesh
s4    4 rajesh
s5    5 kiran
   rno  name
s1    1 naresh
s2    2 suresh
s3    3 ramesh
   rno  name
s1    1 naresh
s3    3 ramesh
s5    5 kiran
<class 'pandas.core.frame.DataFrame'>

```

## Append Row

After creating data frame we can add a new row using append method. This method will add row at the end of dataframe.

`dataframe.append(row)`

Row is represented as a dataframe.

```

l=[[ 'naresh',45],[ 'suresh',50],[ 'ramesh',60]]
df=pd.DataFrame(l,columns=[ 'name', 'age' ])
df1=pd.DataFrame([[ 'rajesh',60],[ 'kishore',60]],columns=[ 'name', 'age' ])
df2=df.append(df1)
print(df2)
print(df2.iloc[0])
print(df2.iloc[3])

```

```

name age
0  naresh  45
1  suresh  50
2  ramesh  60
0  rajesh  60
1  kishore  60
name    naresh
age      45
Name: 0, dtype: object
name    rajesh
age      60
Name: 0, dtype: object

```

## Deletion of rows

Deletion of rows are done using a method drop().

It delete only one row.

Deleting is done using row labels/index.

It row labels are duplicated it remove multiple rows.

```

l=[[ 'naresh',45],[ 'suresh',50],[ 'ramesh',60]]
df=pd.DataFrame(l,columns=[ 'name', 'age' ])
df1=pd.DataFrame([[ 'rajesh',60],[ 'kishore',60]],columns=[ 'name', 'age' ])
df2=df.append(df1)
print(df2)
df3=df2.drop(0)
df4=df2.drop(1)
print(df3)
print(df4)
print(df2)

```

```

name age
0  naresh  45
1  suresh  50
2  ramesh  60
0  rajesh  60
1  kishore  60
name age
1  suresh  50
2  ramesh  60
1  kishore  60
name age
0  naresh  45
2  ramesh  60
0  rajesh  60
name age

```

## head and tail methods of DataFrame

head and tail are the methods of DataFrame object.

head() returns first n number of rows  
tail() returns last n number of rows

```
▶ person_dict={'name':pd.Series(['naresh','ramesh','kishore','ramesh']),  
              'grade':pd.Series([45,67,88,34])}  
df=pd.DataFrame(person_dict)  
print(df)  
df1=df.head(2)  
df2=df.tail(2)  
print(df1)  
print(df2)
```

```
↵   name  grade  
0  naresh    45  
1   ramesh    67  
2  kishore    88  
3   ramesh    34  
   name  grade  
0  naresh    45  
1   ramesh    67  
   name  grade  
2  kishore    88  
3   ramesh    34
```

## Other Operations of DataFrame

sum(): This function return sum

```
▶ import pandas as pd  
df=pd.DataFrame({'sales':[10000,2000,3000,4000,5000,60000]})  
print(df)  
s=df.sum()  
print("Total is",s)
```

```
↵   sales  
0  10000  
1   2000  
2   3000  
3   4000  
4   5000  
5  60000  
Total is sales    84000  
dtype: int64
```

```

import pandas as pd
df=pd.DataFrame({'sales':[10000,2000,3000,4000,5000,60000]})
print(df)
s=df.sum()
print("Total is",s)
df=pd.DataFrame({'name':['naresh','suresh','rajesh'],'age':[45,40,35]},columns=['name','age'])
print(df)
s=df.sum()
print(s)

```

```

sales
0    10000
1     2000
2     3000
3     4000
4     5000
5    60000
Total is sales      84000
dtype: int64
   name  age
0  naresh  45
1  suresh  40
2  rajesh  35
name    nareshsureshrajesh
age              120
dtype: object

```

**describe():** This function perform statistical operations on dataframe.

```

df=pd.DataFrame({'sales':[1000,2000,3000,4000,5000,6000,7000]})
print(df)
print(df.describe())
x=df.describe()
print(type(x))
print(x.iloc[0])
print(x.loc['mean'])

```

```

sales
0    1000
1    2000
2    3000
3    4000
4    5000
5    6000
6    7000

   sales
count    7.000000
mean    4000.000000
std     2160.246899
min     1000.000000
25%     2500.000000
50%     4000.000000
75%     5500.000000
max     7000.000000
<class 'pandas.core.frame.DataFrame'>
sales      7.0

```

## Pandas: Function Application

We can apply customized functions from library or userdefined.

This functions are applied based on the application requirement on rows, columns or element wise.

**pipe()** : table based

**apply()** : row based or column based

**applymap()**: element based

```

def total(a,b):
    return a+b
df=pd.DataFrame({'col1':[10,20,30,40,50], 'col2':[100,200,300,400,500]})
print(df)
df.pipe(total,10)

```

```

col1  col2
0    10   100
1    20   200
2    30   300
3    40   400
4    50   500

```

```

col1  col2
0    20   110
1    30   210
2    40   310
3    50   410
4    60   510

```

**apply()** this function is used apply a function to rows or columns.

```

import numpy as np
df=pd.DataFrame({'col1':[1,2,3,4,5], 'col2':[10,20,30,40,50]})
print(df)
print(df.apply(np.sqrt))
print(df.apply(np.sum,axis=0))
print(df.apply(np.sum,axis=1))

```

```

col1  col2
0     1    10
1     2    20
2     3    30
3     4    40
4     5    50

```

```

col1  col2
0  1.000000  3.162278
1  1.414214  4.472136
2  1.732051  5.477226
3  2.000000  6.324555
4  2.236068  7.071068

```

```

0     11
1     22
2     33
3     44
4     55
dtype: int64

```

**applymap()** : This function is used to apply a function to individual element in dataframe.

```
df=pd.DataFrame({'c1':[1,2,3],'c2':[4,5,6]})
print(df)
print(df.applymap(str))
df=pd.DataFrame({'c1':['aaa','bbb','ccc']})
print(df)
print(df.applymap(str.upper))
```

```

c1 c2
0  1  4
1  2  5
2  3  6
c1 c2
0  1  4
1  2  5
2  3  6
c1
0  aaa
1  bbb
2  ccc
c1
0  AAA
1  BBB
2  CCC
```

## Pandas : Missing Data

Pandas library provide different functions for cleaning missing values or data.

NaN which is defined as missing value.

The fillna() function fill object with NaN values

```
import pandas as pd
list1=[[1,2,3],[4,5,6],[],[1,2]]
df=pd.DataFrame(list1)
print(df)
df1=df.fillna(0)
print(df1)
df2=df.fillna(1)
print(df2)
```

```

0  1  2
0  1.0  2.0  3.0
1  4.0  5.0  6.0
2  NaN  NaN  NaN
3  1.0  2.0  NaN
0  1  2
0  1.0  2.0  3.0
1  4.0  5.0  6.0
2  0.0  0.0  0.0
3  1.0  2.0  0.0
0  1  2
0  1.0  2.0  3.0
1  4.0  5.0  6.0
2  1.0  1.0  1.0
3  1.0  2.0  1.0
```

## Fill NaN forward and backward

We can fill the values in the different directions over the object.



```

▶ list1=[[1,2,3],[4,5,6],[7,8,9],[10,11,12],[13,14,15]]
df=pd.DataFrame(list1,index=['a','c','e','f','h'],columns=['col1','col2','col3'])
print(df)
df=df.reindex(['a','b','c','d','e','f','g','h'])
print(df)
df1=df.fillna(method="pad")
print(df1)
df2=df.fillna(method="bfill")
print(df2)

```

```

↳   col1  col2  col3
a      1     2     3
c      4     5     6
e      7     8     9
f     10    11    12
h     13    14    15
   col1  col2  col3
a   1.0   2.0   3.0
b   NaN   NaN   NaN
c   4.0   5.0   6.0
d   NaN   NaN   NaN
e   7.0   8.0   9.0
f  10.0  11.0  12.0
g   NaN   NaN   NaN
h  13.0  14.0  15.0

```

```

▶   col1  col2  col3
a   1.0   2.0   3.0
↳ b   NaN   NaN   NaN
c   4.0   5.0   6.0
d   NaN   NaN   NaN
e   7.0   8.0   9.0
f  10.0  11.0  12.0
g   NaN   NaN   NaN
h  13.0  14.0  15.0
   col1  col2  col3
a   1.0   2.0   3.0
b   1.0   2.0   3.0
c   4.0   5.0   6.0
d   4.0   5.0   6.0
e   7.0   8.0   9.0
f  10.0  11.0  12.0
g  10.0  11.0  12.0
h  13.0  14.0  15.0
   col1  col2  col3
a   1.0   2.0   3.0
b   4.0   5.0   6.0
c   4.0   5.0   6.0
d   7.0   8.0   9.0
e   7.0   8.0   9.0
f  10.0  11.0  12.0
g  13.0  14.0  15.0
h  13.0  14.0  15.0

```

```
list1=[[1,2,3],[4,5,6],[7,8,9]]
df=pd.DataFrame(list1,index=[i for i in range(3)],columns=["col"+str(i) for i in range(1,4)])
print(df)
df=df.reindex([i for i in range(10)])
print(df)
```

```
col1 col2 col3
0    1    2    3
1    4    5    6
2    7    8    9
col1 col2 col3
0    1.0  2.0  3.0
1    4.0  5.0  6.0
2    7.0  8.0  9.0
3    NaN  NaN  NaN
4    NaN  NaN  NaN
5    NaN  NaN  NaN
6    NaN  NaN  NaN
7    NaN  NaN  NaN
8    NaN  NaN  NaN
9    NaN  NaN  NaN
```

## Drop Missing Values

We can drop or exclude missing elements/values using a predefined method called `dropna()`.

If any row value is NaN the complete row is excluded.

```
l=[[1,2,3],[4,5],[6]]
df=pd.DataFrame(l)
print(df)
df1=df.dropna()
print(df1)
df2=df.dropna(axis=1)
print(df2)
```

```
0    1    2
0    1  2.0  3.0
1    4  5.0  NaN
2    6  NaN  NaN
0    1    2
0    1  2.0  3.0
0
0    1
1    4
2    6
```

## Replacing Missing or Generic values

Dataframe provide replace method, this method is used to replace missing value or generic value with any other value or specific value.

```
l=[[1,2,3],[4,5],[6]]
df=pd.DataFrame(l)
print(df)
df1=df.replace({float('NaN'):0,5.0:10.0})
print(df1)
```

```
0  1  2
0  1  2.0  3.0
1  4  5.0  NaN
2  6  NaN  NaN
0  1  2
0  1  2.0  3.0
1  4  10.0  0.0
2  6  0.0  0.0
```

```
d1=[[ 'naresh',45],[ 'kishore',50],[ ]]
df=pd.DataFrame(d1,columns=[ 'name', 'age' ])
print(df)
df1=df.replace({float('NaN'):0})
print(df1)
```

```
name age
0  naresh  45.0
1  kishore  50.0
2    None  NaN
name age
0  naresh  45.0
1  kishore  50.0
2      0    0.0
```

## Pandas Groupby

Group by is one of the important concept in processing data in data science.

We can create group of categories and apply functions to that categories. Group by involving the following steps.

1. Splitting
2. Applying
3. Combining

Pandas datasets/dataframe can be split on any axis. There are multiple ways or methods are used to split data.

1. DataFrame.groupby(key)
2. DataFrame.groupby(key,axis=1)

### 3. DataFrame.groupby(key1,key2)

#### Grouping data using one or more than one key

In order to group data using one key, we pass one key argument to groupby method. Key is field name or column name. in order to group multiple columns, we need provide more than one column name in groupby.

	empno	ename	job	deptno	salary
0	1	aaa	manager	10	1000
1	2	bbb	clerk	20	2000
2	3	ccc	manager	10	3000
3	4	ddd	hr	10	4000
4	5	eee	clerk	20	5000
5	6	fff	manager	20	6000
6	7	ggg	manager	30	1000
7	8	hhh	clerk	10	3000
8	9	iii	hr	20	4000
9	10	jjj	hr	30	8000

deptno :10

0	1	aaa	manager	10	1000
2	3	ccc	manager	10	3000
3	4	ddd	hr	10	4000
7	8	hhh	clerk	10	3000

job : manager

0	1	aaa	manager	10	1000
2	3	ccc	manager	10	3000
job: clerk					
7	8	hhh	clerk	10	3000
job: hr					
3	4	ddd	hr	10	4000

```
import pandas as pd
empdict={'empno':[1,2,3,4,5,6,7,8,9,10],
        'ename':['aaa','bbb','ccc','ddd','eee','fff','ggg','hhh','iii','jjj'],
        'job':['manager','clerk','manager','hr','clerk','manager','manager','clerk','hr','hr'],
        'deptno':[10,20,10,10,20,20,30,10,20,30],
        'salary':[1000,2000,3000,4000,5000,6000,1000,3000,4000,8000]}
empdf=pd.DataFrame(empdict,columns=['empno','ename','job','deptno','salary'])
print(empdf)
print(empdf.groupby('deptno'))
print(empdf.groupby('deptno').groups)
print(empdf.groupby(['deptno','job']).groups)
```

empno	ename	job	deptno	salary
1	aaa	manager	10	1000
2	bbb	clerk	20	2000
3	ccc	manager	10	3000
4	ddd	hr	10	4000
5	eee	clerk	20	5000
6	fff	manager	20	6000
7	ggg	manager	30	1000
8	hhh	clerk	10	3000
9	iii	hr	20	4000
10	jjj	hr	30	8000

```
pandas.core.groupby.generic.DataFrameGroupBy object at 0x7f742c580090>
0: [0, 2, 3, 7], 20: [1, 4, 5, 8], 30: [6, 9]]
10, 'clerk'): [7], (10, 'hr'): [3], (10, 'manager'): [0, 2], (20, 'clerk'): [1, 4], (20, 'hr'): [8], (20, 'manager'): [5], (30, 'hr'): [9], (30, 'manager'): [6]]
```

#### Iterating through groups

We can read the content of group by object. This can be done using for loop.

```

import pandas as pd
empdict={'empno':[1,2,3,4,5,6,7,8,9,10],
        'ename':['aaa','bbb','ccc','ddd','eee','fff','ggg','hhh','iii','jjjj'],
        'job':['manager','clerk','manager','hr','clerk','manager','manager','clerk','hr','hr'],
        'deptno':[10,20,10,10,20,20,30,10,20,30],
        'salary':[1000,2000,3000,4000,5000,6000,1000,3000,4000,8000]}
empdf=pd.DataFrame(empdict,columns=['empno','ename','job','deptno','salary'])
deptgroup=empdf.groupby('deptno')
for name,group in deptgroup:
    print(name)
    print(group)

```

```

10
empno  ename  job  deptno  salary
0      1  aaa  manager    10    1000
2      3  ccc  manager    10    3000
3      4  ddd    hr      10    4000
7      8  hhh  clerk     10    3000
20
empno  ename  job  deptno  salary
1      2  bbb  clerk     20    2000
4      5  eee  clerk     20    5000
5      6  fff  manager    20    6000
8      9  iii    hr      20    4000
30
empno  ename  job  deptno  salary
6      7  ggg  manager    30    1000

```

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`get_group(name)` : This method is used to get the selected group from grouped objects (OR) select single group.

```

import pandas as pd
empdict={'empno':[1,2,3,4,5,6,7,8,9,10],
        'ename':['aaa','bbb','ccc','ddd','eee','fff','ggg','hhh','iii','jjjj'],
        'job':['manager','clerk','manager','hr','clerk','manager','manager','clerk','hr','hr'],
        'deptno':[10,20,10,10,20,20,30,10,20,30],
        'salary':[1000,2000,3000,4000,5000,6000,1000,3000,4000,8000]}
empdf=pd.DataFrame(empdict,columns=['empno','ename','job','deptno','salary'])
deptgroup=empdf.groupby('deptno')
print(deptgroup.get_group(10))
print(deptgroup.get_group(20))

```

```

empno  ename  job  deptno  salary
0      1  aaa  manager    10    1000
2      3  ccc  manager    10    3000
3      4  ddd    hr      10    4000
7      8  hhh  clerk     10    3000
empno  ename  job  deptno  salary
1      2  bbb  clerk     20    2000
4      5  eee  clerk     20    5000
5      6  fff  manager    20    6000
8      9  iii    hr      20    4000

```

## Aggregations

The aggregate function return single aggregated value for each group. Once the group by object is created we can perform different/several aggregate operations.

The aggregate operations are applied on group using 'agg' method

```

import pandas as pd
import numpy as np
empdict={'empno':[1,2,3,4,5,6,7,8,9,10],
        'ename':['aaa','bbb','ccc','ddd','eee','fff','ggg','hhh','iii','jjjj'],
        'job':['manager','clerk','manager','hr','clerk','manager','manager','clerk','hr','hr'],
        'deptno':[10,20,10,10,20,20,30,10,20,30],
        'salary':[1000,2000,3000,4000,5000,6000,1000,3000,4000,8000]}
empdf=pd.DataFrame(empdict,columns=['empno','ename','job','deptno','salary'])
deptgroup=empdf.groupby('deptno')
print(deptgroup.agg(np.size))
deptgroup=empdf[['deptno','salary']].groupby('deptno')
print(deptgroup.get_group(10))
print(deptgroup.agg(np.sum))
deptgroup=empdf[['deptno','job','salary']].groupby(['deptno','job'])
for name,group in deptgroup:
    print(name)
    print(group)
print(deptgroup.get_group((10,'manager')))
print(deptgroup.agg(np.sum))

```

```

empno  ename  job  salary
deptno
10      4      4      4
20      4      4      4
30      2      2      2
deptno  salary
0      10    1000
2      10    3000
3      10    4000
7      10    3000
salary
deptno
10      11000
20      17000
30       9000
(10, 'clerk')
deptno  job  salary
7      10  clerk   3000
(10, 'hr')
deptno  job  salary
3      10  hr     4000
(10, 'manager')
deptno  job  salary
0      10  manager  1000
2      10  manager  3000
(20, 'clerk')
deptno  job  salary
1      20  clerk   2000
4      20  clerk   5000
(20, 'hr')
deptno  job  salary

```

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## Transformations

Transformation on a group or a column return an object that is indexed the same size of that is being grouped. The transformation should return result the same size of that group.

```

import pandas as pd
import numpy as np
empdict={'empno':[1,2,3,4,5,6,7,8,9,10],
         'ename':['aaa','bbb','ccc','ddd','eee','fff','ggg','hhh','iii','jjjj'],
         'job':['manager','clerk','manager','hr','clerk','manager','manager','clerk','hr','hr'],
         'deptno':[10,20,10,10,20,20,30,10,20,30],
         'salary':[1000,2000,3000,4000,5000,6000,1000,3000,4000,8000]}
empdf=pd.DataFrame(empdict,columns=['empno','ename','job','deptno','salary'])
print(empdf)
deptgroup=empdf[['deptno','salary']].groupby("deptno")
incr=lambda s:s+100
print(deptgroup.transform(incr))
df=deptgroup.transform(incr)
print(type(df))

```

```

empno  ename  job  deptno  salary
0      1   aaa  manager    10    1000
1      2   bbb   clerk    20    2000
2      3   ccc  manager    10    3000
3      4   ddd    hr      10    4000
4      5   eee   clerk    20    5000
5      6   fff  manager    20    6000
6      7   ggg  manager    30    1000
7      8   hhh   clerk    10    3000
8      9   iii    hr      20    4000
9     10  jjjj    hr      30    8000

```

```

salary
0      1100
1      2100
2      3100
3      4100
4      5100
5      6100
6      1100
7      3100
8      4100
9      8100

```

```
<class 'pandas.core.frame.DataFrame'>
```

## Filtration

Filtration filters the data of dataset based on condition or test.

filter() function is used to filter data. This function returns subset of data which is again one dataframe.

### filter(condition)

```

import pandas as pd
empdict={'empno':[1,2,3,4,5,6,7,8,9,10],
        'ename':['aaa','bbb','ccc','ddd','eee','fff','ggg','hhh','iii','jjjj'],
        'job':['manager','clerk','manager','hr','clerk','manager','manager','clerk','t
        'deptno':[10,20,10,10,20,20,30,10,20,30],
        'salary':[1000,2000,3000,4000,5000,6000,1000,3000,4000,8000]}
empdf=pd.DataFrame(empdict,columns=['empno','ename','job','deptno','salary'])
print(empdf)
empgroup=empdf.groupby('job')
print(empgroup.groups)
for name,group in empgroup:
    print(name)
    print(group)
print(empgroup.filter(lambda x:x['job']=='manager'))

```

## Merging and joining

Merging and joining DataFrame objects.

Pandas provide a single function called merge which is used to perform join operation between dataframe objects.

**pandas.merge(left, right, how='inner', on=None, left\_on=None, right\_on=None, left\_index=False, right\_index=False, sort=False, suffixes=('\_x', '\_y'), copy=True, indicator=False, validate=None)**

left → A dataframe object

right → A dataframe object

on → column names to join on. The column names must be exists in data frame object of left and right.

how → the value of this can be left,right,inner

sort → sort the result of dataframe



```

emp=pd.DataFrame({'empid':[1,2,3],
                  'ename':['abc','xyx','bca'],
                  'sal':[4000,5000,6000],
                  'deptno':[10,20,10]})
dept=pd.DataFrame({'deptno':[10,20,30],
                  'dname':['sales','HR','Accounts']})

print(emp)
print(dept)
pd.merge(emp,dept,on='deptno')

```

```

empid  ename  sal  deptno
0      1   abc  4000      10
1      2   xyx  5000      20
2      3   bca  6000      10
deptno  dname
0      10   sales
1      20     HR
2      30  Accounts

empid  ename  sal  deptno  dname
0      1   abc  4000      10   sales
1      3   bca  6000      10   sales
2      2   xyx  5000      20     HR

```

```

import pandas as pd
left=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['abc','aaa','acb','abb','acc'],
                  'subject_id':['sub1','sub2','sub4','sub6','sub5']})
right=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['bbb','bca','bac','bab','bba'],
                  'subject_id':['sub2','sub4','sub3','sub6','sub5']})

print(left)
print(right)
pd.merge(left,right,on=['id','subject_id'])

```

```

id  name  subject_id
0   1   abc         sub1
1   2   aaa         sub2
2   3   acb         sub4
3   4   abb         sub6
4   5   acc         sub5
id  name  subject_id
0   1   bbb         sub2
1   2   bca         sub4
2   3   bac         sub3
3   4   bab         sub6
4   5   bba         sub5

id  name_x  subject_id  name_y
0   4   abb         sub6   bab
1   5   acc         sub5   bba

```

## Merging using “how” argument

The how argument of merge specifies which keys are included in the returning result dataframe/table. If key combination does not exists/appear either of dataframes/tables, the value joined table is NaN.

Merge	SQL Terminology	Description
Left	Left Outer Join	Use keys from left object
Right	Right Outer Join	Use keys from right object
OuterFull	outer join	Use Union of keys
Inner	Inner Join	Use intersection Keys

## Left join

```
import pandas as pd
left=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['abc','aaa','acb','abb','acc'],
                  'subject_id':['python','java','sub4','sub6','sub5',]})
right=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['bbb','bca','bac','bab','bba'],
                  'subject_id':['java','sub4','sub3','sub6','sub5']})

print(left)
print(right)
pd.merge(left,right,on='subject_id',how='left')
```

```
id  name subject_id
0   1   abc     python
1   2   aaa       java
2   3   acb     sub4
3   4   abb     sub6
4   5   acc     sub5
id  name subject_id
0   1   bbb       java
1   2   bca     sub4
2   3   bac     sub3
3   4   bab     sub6
4   5   bba     sub5
```

	id_x	name_x	subject_id	id_y	name_y
0	1	abc	python	NaN	NaN
1	2	aaa	java	1.0	bbb
2	3	acb	sub4	2.0	bca
3	4	abb,	sub6	4.0	bab
4	5	acc	sub5	5.0	bba

## Right outer join

```
import pandas as pd

left=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['abc','aaa','acb','abb','acc'],
                  'subject_id':['python','java','sub4','sub6','sub5']})

right=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['bbb','bca','bac','bab','bba'],
                  'subject_id':['java','sub4','sub3','sub6','sub5']})

print(left)
print(right)
pd.merge(left,right,on='subject_id',how='right')
```

```
id  name subject_id
0  1   abc    python
1  2   aaa      java
2  3   acb    sub4
3  4   abb,   sub6
4  5   acc    sub5
id  name subject_id
0  1  bbb      java
1  2  bca    sub4
2  3  bac    sub3
3  4  bab    sub6
4  5  bba    sub5
```

```
id_x  name_x  subject_id  id_y  name_y
```

id_x	name_x	subject_id	id_y	name_y
0	2.0	aaa	1	bbb
1	3.0	acb	2	bca
2	NaN	NaN	3	bac
3	4.0	abb,	4	bab
4	5.0	acc	5	bba

## Full outer join

```

import pandas as pd
left=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['abc','aaa','acb','abb','acc'],
                  'subject_id':['sub1','sub2','sub4','sub6','sub5',]})
right=pd.DataFrame({'id':[1,2,3,4,5],
                  'name':['bbb','bca','bac','bab','bba'],
                  'subject_id':['sub2','sub4','sub3','sub6','sub5']})

print(left)
print(right)
pd.merge(left,right,how='outer',on='subject_id')

```

```

id  name  subject_id
0  1    abc      sub1
1  2    aaa      sub2
2  3    acb      sub4
3  4    abb      sub6
4  5    acc      sub5
id  name  subject_id
0  1    bbb      sub2
1  2    bca      sub4
2  3    bac      sub3
3  4    bab      sub6
4  5    bba      sub5

```

	id_x	name_x	subject_id	id_y	name_y
0	1.0	abc	sub1	NaN	NaN
1	2.0	aaa	sub2	1.0	bbb
2	3.0	acb	sub4	2.0	bca
3	4.0	abb,	sub6	4.0	bab
4	5.0	acc	sub5	5.0	bba
5	NaN	NaN	sub3	3.0	bac

**How to load data into dataframe from different source?**

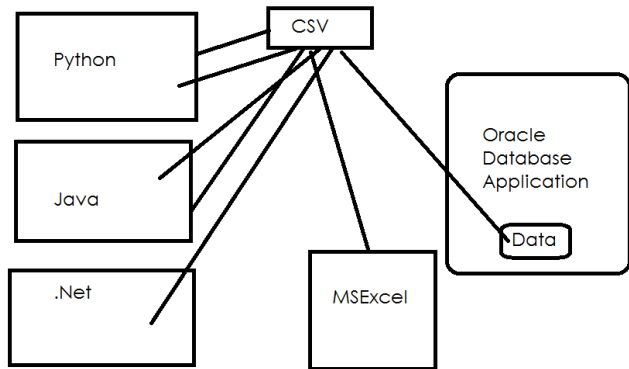
1. Loading from CSV file
2. Loading from excel file
3. Loading from database table
4. Loading from JSON file
5. SAS
6. SPSS
7. STATA

## CSV File

CSV stands for comma separated values.

CSV is standard or protocol. This standard is used to exchange data between two different applications.

CSV standard import and export format used by database applications and many programming languages.



+ Code + Text

```
import pandas as pd
df=pd.read_csv("employee.csv")
print(df)
```

	empno	ename	job	salary
0	1	naresh,rao	manager	50000
1	2	suresh	clerk	20000
2	3	kishore	hr	40000
3	4	ramesh	manager	50000
4	5	rajesh	clerk	20000

## Reading Excel Sheet

We can read and load the data from excel sheet data frame.

This is done using a function `pandas.read_excel(path)`

```
df=pd.read_excel("product.xlsx")
print(df)
```

	pname	price
0	mouse	100
1	keyboard	2000
2	monitor	5000
3	disk	15000
4	cdrw	2000

## Reading JSON file

JSON stands Java Script Object Notation. It is standard which is used to exchange data between two different applications. JSON is a text file. It is used in internet applications exchange data between client and server or between the servers.

`pandas.read_json()`

This is used to read data from json file.

## Reading Specific worksheet from workbook(Excel)

```
import pandas as pd
df1=pd.read_excel("Book1.xlsx",sheet_name="student")
print(df1)
df2=pd.read_excel("Book1.xlsx",sheet_name="employee")
print(df2)
```

```
rollno name course
0      1   aaa  python
1      2   bbb   java
2      3   ccc  python
3      4   ddd     c
4      5   eee    c++
empid  ename salary
0      1  naresh  8000
1      2  suresh  5000
2      3  kishore  9000
3      4  rajesh  7000
4      5   raman  4500
```

## How to read specific Columns from worksheet?

`read_excel` function provide an optional argument called `usecols`, by default it read all the columns from worksheet.

```
import pandas as pd
df1=pd.read_excel("Book1.xlsx",sheet_name="student",usecols=['rollno','course'])
print(df1)
```

```
rollno course
0      1  python
1      2   java
2      3  python
3      4     c
4      5    c++
```

## How to read data from worksheet without header row?

```
import pandas as pd
df1=pd.read_excel("Book1.xlsx",sheet_name="student",header=None)
print(df1)
```

```
0      0      1      2
0  rollno  name  course
1      1   aaa  python
2      2   bbb   java
3      3   ccc  python
4      4   ddd     c
5      5   eee    c++
```

## pandas.read\_table()

this function is used to read data from text file. It will read text file and return data in table format.

```
df=pd.read_table("student.csv",sep=",")  
df
```

```
rollno  name  course  
0      1  suresh  python  
1      2  rajesh   java  
2      3  ramesh  oracle  
3      4  kishore    c  
4      5   kiran   cpp
```

```
df=pd.read_table("emp.txt",sep=" ")  
print(df)
```

```
empno  ename  salary  
0      1  ramesh   5000  
1      2  suresh   6000  
2      3  kishore  9000
```

```
df=pd.read_table("emp.txt",sep=" ",nrows=2)  
print(df)  
df=pd.read_table("emp.txt",sep=" ",skipfooter=2,engine="python")  
print(df)
```

```
empno  ename  salary  
0      1  ramesh   5000  
1      2  suresh   6000  
empno  ename  salary  
0      1  ramesh   5000
```