

```
In [3]: # Pandas
# - panel data
# - core library for data manipulation and data analysis
#- single and multidimensional data sturcture for data manipulation
# single-dimensional --> series object
# multi-dimensional--> data frame
```

```
In [7]: #----- pandas series object(one-dimensinonal)-----

import pandas as pd
a=pd.Series([1,2,3,4])
a
```

```
Out[7]: 0    1
        1    2
        2    3
        3    4
dtype: int64
```

```
In [11]: #--- changing index ---

b=pd.Series([1,2,3,4],index=['a','b','c','d'])
b
```

```
Out[11]: a    1
         b    2
         c    3
         d    4
dtype: int64
```

```
In [33]: # series object from dictionary

c=pd.Series({'a':1,'b':2,'c':3,'d':4})
c
```

```
Out[33]: a    1
         b    2
         c    3
         d    4
dtype: int64
```

```
In [36]: c1=pd.Series({'a':1,'b1':2,'c':3,'d1':4})
          # opertsion between Series
c+c1
```

```
Out[36]: a    2.0
         b    NaN
         b1   NaN
         c    6.0
         d    NaN
         d1   NaN
dtype: float64
```

```
In [17]: # index position change

c=pd.Series({'a':1,'b':2,'c':3,'d':4},index=['b','c','a','d'])
c
```

```
Out[17]: b    2
         c    3
         a    1
```

```
d      4
dtype: int64
```

```
In [28]: # extracting a singel element
          #0 1 2 3 4 5
d=pd.Series([1,2,3,4,5,6])
d[2]
```

```
Out[28]: 3
```

```
In [34]: d[:4]      # sequence of element
```

```
Out[34]: 0      1
          1      2
          2      3
          3      4
dtype: int64
```

```
In [36]: d[-3:]     # extracting elements from back
```

```
Out[36]: 3      4
          4      5
          5      6
dtype: int64
```

```
In [52]: # assign value to the elements(change value)
d[3]=10
d[4]='nikhil'
d
```

```
Out[52]: 0      1
          1      2
          2      3
          3     10
          4   nikhil
          5      6
dtype: object
```

```
In [57]: # defining Series from numpy array and other Series

import numpy as np
arr=np.array([1,2,3,4])
e=pd.Series(arr)
e
```

```
Out[57]: 0      1
          1      2
          2      3
          3      4
dtype: int32
```

```
In [65]: arr[2]=-2    #change value
e
```

```
Out[65]: 0      1
          1      2
          2     -2
          3      4
dtype: int32
```

```
In [80]: # basic math operation on Series
          # Adding a scalar value to series elements
```

```
f=pd.Series([1,2,3,4,5])
f1=f+10 # (+, -, *, /)
f1
```

```
Out[80]: 0    11
         1    12
         2    13
         3    14
         4    15
dtype: int64
```

```
In [84]: f+2 # (+, -, *, /)
```

```
Out[84]: 0    3
         1    4
         2    5
         3    6
         4    7
dtype: int64
```

```
In [85]: np.log(f) # Log
```

```
Out[85]: 0    0.000000
         1    0.693147
         2    1.098612
         3    1.386294
         4    1.609438
dtype: float64
```

```
In [77]: #adding two Series
```

```
g=pd.Series([5,6,7,8,9])
f1=f+g
f1
```

```
Out[77]: 0    6
         1    8
         2   10
         3   12
         4   14
dtype: int64
```

```
In [40]: # evaluating values
import pandas as pd
g=pd.Series([1,0,2,1,2,3],index=['white','white','blue','green','green','yellow'])
g
```

```
Out[40]: white    1
         white    0
         blue     2
         green     1
         green     2
         yellow    3
dtype: int64
```

```
In [41]: g.unique() # unique value()
```

```
Out[41]: array([1, 0, 2, 3], dtype=int64)
```

```
In [42]: g.value_counts() # return unique values and calculate occurrences within a Series
```

```
Out[42]: 2    2
         1    2
         3    1
         0    1
         dtype: int64
```

```
In [51]: g.isin([0,3]) # values are contained within the data structure
```

```
Out[51]: white    False
         white    True
         blue     False
         green    False
         green    False
         yellow   True
         dtype: bool
```

```
In [22]: # NaN(Not a Number) values

import numpy as np
h=pd.Series([3,-6,np.NaN,14])
h
```

```
Out[22]: 0    3.0
         1   -6.0
         2    NaN
         3   14.0
         dtype: float64
```

```
In [24]: # isnull()
         # notnull()
         # - identify the indexes without a value

h.isnull()
```

```
Out[24]: 0    False
         1    False
         2     True
         3    False
         dtype: bool
```

```
In [26]: h.notnull()
```

```
Out[26]: 0     True
         1     True
         2    False
         3     True
         dtype: bool
```

```
In [30]: h[h.notnull()]
```

```
Out[30]: 0    3.0
         1   -6.0
         3   14.0
         dtype: float64
```

```
In [32]: h[h.isnull()]
```

```
Out[32]: 2    NaN
         dtype: float64
```

```
In [68]: #-----Data frame() 2 dimensional-----
```

```
import pandas as pd
a=pd.DataFrame({"Name":["nikhil","rahul","kk"],"marks":[80,75,78],"class":["TY","SY","FY"]})
a
```

Out[68]:

	Name	marks	class
0	nikhil	80	TY
1	rahul	75	SY
2	kk	78	FY

In [64]: `a=pd.DataFrame(a,columns=['Name','marks'])` *# columns show specific columns*
a

Out[64]:

	Name	marks
0	nikhil	80
1	rahul	75
2	kk	78

In [17]: `import pandas as pd
import numpy as np
b=pd.DataFrame(np.arange(16).reshape((4,4)),
 index=['red','blue','yellow','green'],
 columns=['ball','pen','pencil','paper'])`
b

Out[17]:

	ball	pen	pencil	paper
red	0	1	2	3
blue	4	5	6	7
yellow	8	9	10	11
green	12	13	14	15

In [18]: *# selecting elements*

`b.columns`

Out[18]: Index(['ball', 'pen', 'pencil', 'paper'], dtype='object')

In [19]: `b.index`

Out[19]: Index(['red', 'blue', 'yellow', 'green'], dtype='object')

In [21]: `b.values`

Out[21]: array([[0, 1, 2, 3],
[4, 5, 6, 7],
[8, 9, 10, 11],
[12, 13, 14, 15]])

In [23]: `# slicing`

`b[:3]`

Out[23]:

	ball	pen	pencil	paper
red	0	1	2	3
blue	4	5	6	7
yellow	8	9	10	11

In [24]: `# convert column to row`

`b.transpose()`

Out[24]:

	red	blue	yellow	green
ball	0	4	8	12
pen	1	5	9	13
pencil	2	6	10	14
paper	3	7	11	15

In [25]: `# adding new column`

`b['price']=[5,7,8,10]`
`b`

Out[25]:

	ball	pen	pencil	paper	price
red	0	1	2	3	5
blue	4	5	6	7	7
yellow	8	9	10	11	8
green	12	13	14	15	10

In [26]: `b1=np.arange(1,8,2)` `# set new value for price`
`b1`

Out[26]: `array([1, 3, 5, 7])`

In [27]: `b['price']=b1`
`b`

Out[27]:

	ball	pen	pencil	paper	price
red	0	1	2	3	1
blue	4	5	6	7	3
yellow	8	9	10	11	5
green	12	13	14	15	7

```
In [59]: # membership of a value

b.isin([3,'b'])    # value present or not
```

```
Out[59]:
```

	ball	pen	pencil	paper	price
red	False	False	False	True	False
blue	False	False	False	False	True
yellow	False	False	False	False	False
green	False	False	False	False	False

```
In [63]: #delete column

del b['pen']
b
```

```
Out[63]:
```

	ball	pencil	price
red	0	2	1
blue	4	6	3
yellow	8	10	5
green	12	14	7

```
In [70]: #filtering
#- apply the filtering through the application of certain condition(<,>,>=,<=)

b[b<10]
# b[b>10]
# b[b<=10]
# b[b>=10]
```

```
Out[70]:
```

	ball	pencil	price
red	0.0	2.0	1
blue	4.0	6.0	3
yellow	8.0	NaN	5
green	NaN	NaN	7

```
In [12]: # # The index objects
# - show index

c=pd.Series([5,0,3,8,4],index=['red','blue','yellow','white','green'])
c.index
```

```
Out[12]: Index(['red', 'blue', 'yellow', 'white', 'green'], dtype='object')
```

```
In [27]: # methods on index
```

```
c.idxmax()      #----- for max value
```

```
Out[27]: 'white'
```

```
In [29]: c.idxmin()      # for min value
```

```
Out[29]: 'blue'
```

```
In [34]: # index with duplicate labels

c=pd.Series(range(7),index=['red','blue','yellow','white','green','red','blue'])
c
```

```
Out[34]: red      0
blue      1
yellow    2
white     3
green     4
red       5
blue     6
dtype: int64
```

```
In [36]: c['red']
```

```
Out[36]: red      0
red       5
dtype: int64
```

```
In [37]: # other functionalities on indexes
# 1) reindexing
# 2) dropping
# 3) alignment
```

```
In [52]: # 1) reindexing
# to fill missing value(in order)

d=pd.Series([2,5,7,4],index=['one','two','three','four']) # normal series
d
```

```
Out[52]: one      2
two       5
three     7
four      4
dtype: int64
```

```
In [80]: d.reindex(['three','four','five','eee'])
```

```
Out[80]: three     7.0
four      4.0
five      NaN
eee       NaN
dtype: float64
```

```
In [61]: # 2) dropping
# for delete

d.drop(['one','three'])
```

```
Out[61]: two      5
four      4
```


dtype: int64

```
In [67]: # 3) airthmetic and data alignment

d1=pd.Series([1,6,7,8,9],index=['one','brown','three','black','two'])
d1
```

```
Out[67]: one      1
brown    6
three    7
black    8
two      9
dtype: int64
```

```
In [81]: d+d1    # addition of d+d1 , means one(2)+one(1)=3
```

```
Out[81]: black      NaN
brown      NaN
four       NaN
one        3.0
three     14.0
two       14.0
dtype: float64
```

```
In [86]: # operation between data structure

###--- flexible arithmetic methods
# add()  sub()  div()  mul()

d.mul(d1)    # div()  sub()  add()
```

```
Out[86]: black      NaN
brown      NaN
four       NaN
one        2.0
three     49.0
two       45.0
dtype: float64
```

```
In [104... #operation between DataFrame and Series

import numpy as np
b=pd.DataFrame(np.arange(16).reshape((4,4)),
               index=['red','blue','yellow','green'],      # dataframe
               columns=['ball','pen','pencil','paper'])

c=pd.Series([5,0,3,8,4],index=['ball','blue','pen','pencil','paper']) # seri

c-b          # +,*,/
```

```
Out[104...      ball  blue  paper  pen  pencil
red      5   NaN     1    2      6
blue     1   NaN    -3   -2      2
yellow  -3   NaN    -7   -6     -2
```

	ball	blue	paper	pen	pencil
green	-7	NaN	-11	-10	-6

```
In [4]: # function application and mapping

b=pd.DataFrame(np.arange(16).reshape((4,4)),
               index=['red','blue','yellow','green'],
               columns=['ball','pen','pencil','paper'])
b
```

```
Out[4]:
```

	ball	pen	pencil	paper
red	0	1	2	3
blue	4	5	6	7
yellow	8	9	10	11
green	12	13	14	15

```
In [7]: # sqrt() ----- calculate square root

np.sqrt(b)
```

```
Out[7]:
```

	ball	pen	pencil	paper
red	0.000000	1.000000	1.414214	1.732051
blue	2.000000	2.236068	2.449490	2.645751
yellow	2.828427	3.000000	3.162278	3.316625
green	3.464102	3.605551	3.741657	3.872983

```
In [12]: # function by row or column-----

# define function

#lambda function:-block of code that can be passed as an argument to to a function call
# used when you need a function for a short period time
# anonymous function
# calculate the range covered by the elements in an array

b=lambda x:x.max()-x.min()
b
```

```
Out[12]: <function __main__.<lambda>(x)>
```

```
In [46]: def f(x):
          return x.max()-x.min()
          f
```

```
Out[46]: <function __main__.f(x)>
```

```
In [71]: z1=pd.DataFrame(np.arange(49).reshape((7,7)),
```

```

index=['red','blue','yellow','green','black','purple','white'],
columns=['ball','pen','pencil','paper','book','textbook','notebook'])
z1

```

```

Out[71]:

```

	ball	pen	pencil	paper	book	textbook	notebook
red	0	1	2	3	4	5	6
blue	7	8	9	10	11	12	13
yellow	14	15	16	17	18	19	20
green	21	22	23	24	25	26	27
black	28	29	30	31	32	33	34
purple	35	36	37	38	39	40	41
white	42	43	44	45	46	47	48

```

In [72]: # apply() -- apply the function just defined on the DataFrame

z1.apply(f,axis=0)

```

```

Out[72]:

```

	ball	pen	pencil	paper	book	textbook	notebook
min	0	1	2	3	4	5	6
max	42	43	44	45	46	47	48

```

In [74]: # DataFrame in-build function
# 1) head()--> 1st five records
# 2) tail()---> Last 5 records
# 3) shape()--> how much records presents
# 4) describe()--> information about records

#1) head()
z1.head()

```

```

Out[74]:

```

	ball	pen	pencil	paper	book	textbook	notebook
red	0	1	2	3	4	5	6
blue	7	8	9	10	11	12	13
yellow	14	15	16	17	18	19	20
green	21	22	23	24	25	26	27
black	28	29	30	31	32	33	34

```

In [94]: # 2) tail()
z1.tail()

```

```

Out[94]: <bound method NDFrame.tail of

```

	ball	pen	pencil	paper	book	textbook	notebook
red	0	1	2	3	4	5	6
blue	7	8	9	10	11	12	13
yellow	14	15	16	17	18	19	20
green	21	22	23	24	25	26	27
black	28	29	30	31	32	33	34

```

purple    35    36    37    38    39    40    41
white     42    43    44    45    46    47    48>

```

```

In [91]: # 3) shape()

z1.shape

```

```
Out[91]: (7, 7)
```

```

In [95]: # 4) describe()
z1.describe()

```

```
Out[95]:
```

	ball	pen	pencil	paper	book	textbook	notebook
count	7.000000	7.000000	7.000000	7.000000	7.000000	7.000000	7.000000
mean	21.000000	22.000000	23.000000	24.000000	25.000000	26.000000	27.000000
std	15.121728	15.121728	15.121728	15.121728	15.121728	15.121728	15.121728
min	0.000000	1.000000	2.000000	3.000000	4.000000	5.000000	6.000000
25%	10.500000	11.500000	12.500000	13.500000	14.500000	15.500000	16.500000
50%	21.000000	22.000000	23.000000	24.000000	25.000000	26.000000	27.000000
75%	31.500000	32.500000	33.500000	34.500000	35.500000	36.500000	37.500000
max	42.000000	43.000000	44.000000	45.000000	46.000000	47.000000	48.000000

```

In [7]: # sorting and Ranking
# sort
# ascending order
import pandas as pd
s=pd.Series([4,1,2,5,9],index=['a','b','c','d','e'])
s.sort_values()

```

```

Out[7]: b    1
       c    2
       a    4
       d    5
       e    9
dtype: int64

```

```

In [13]: # Ranking

# -- it will know which value is at which rank
s.rank()

```

```

Out[13]: a    3.0
       b    1.0
       c    2.0
       d    4.0
       e    5.0
dtype: float64

```

```

In [7]: # assigning a NaN value

#(Not a Number)
import pandas as pd
import numpy as np

```

```
n=pd.Series([0,1,2,np.NaN],index=['red','black','white','blue'])
n
```

```
Out[7]: red      0.0
black    1.0
white    2.0
blue     NaN
dtype: float64
```

```
In [37]: n['black']=None
n
```

```
Out[37]: red      0.0
black    NaN
white    2.0
blue     NaN
dtype: float64
```

```
In [39]: # filtering out NaN values

n.dropna()      # eliminated all NaN values
```

```
Out[39]: red      0.0
white    2.0
dtype: float64
```

```
In [41]: n[n.notnull()]      # another way to drop Nan values
```

```
Out[41]: red      0.0
white    2.0
dtype: float64
```

```
In [44]: #filling in NaN Occurrences

n.fillna(3)      # fill NaN value
```

```
Out[44]: red      0.0
black    3.0
white    2.0
blue     3.0
dtype: float64
```

```
In [8]: n.fillna({'black':5,'blue':4})
```

```
Out[8]: red      0.0
black    1.0
white    2.0
blue     4.0
dtype: float64
```

```
In [14]: # hierachical indexing and Leveling

h=pd.Series(np.random.rand(9),
            index=[['white','white','white','blue','blue','blue','red','red','red'],
                  ['up','down','right','up','down','right','up','down','left']])
h
```

```
Out[14]: white up      0.928982
          down    0.504775
          right   0.060468
blue     up      0.036114
          down    0.828382
          right   0.413168
```

```

red    up    0.458447
      down  0.998822
      left  0.029433
dtype: float64

```

In [16]: `h.index`

```

Out[16]: MultiIndex([( 'white',  'up'),
                    ( 'white',  'down'),
                    ( 'white',  'right'),
                    (  'blue',   'up'),
                    (  'blue',   'down'),
                    (  'blue',   'right'),
                    (   'red',    'up'),
                    (   'red',    'down'),
                    (   'red',    'left')],
                  )

```

In [19]: `h['white']`

```

Out[19]: up    0.928982
        down  0.504775
        right 0.060468
dtype: float64

```

In [21]: `h[:, 'up']`

```

Out[21]: white    0.928982
        blue     0.036114
        red      0.458447
dtype: float64

```

In [23]: `h['white', 'up']`

Out[23]: 0.9289817809959987

```

In [24]: # unstack()
#--converts the Series with hierachical index in a simple DataFrame,
# where the second set of indexes is converted into a new set of columns

# convert series to dataframe
h.unstack()

```

```

Out[24]:

```

	down	left	right	up
blue	0.828382	NaN	0.413168	0.036114
red	0.998822	0.029433	NaN	0.458447
white	0.504775	NaN	0.060468	0.928982

```

In [31]: h1=pd.DataFrame([1,2,3,4,5],index=['a','b','c','d','e'])
        h1.stack() # convert dataframe to series

```

```

Out[31]: a  0    1
        b  0    2
        c  0    3
        d  0    4
        e  0    5
dtype: int64

```

```
In [9]: # index option and column option
import pandas as pd
import numpy as np
ic=pd.DataFrame(np.random.rand(16).reshape(4,4),
                 index=[['white','white','red','red'],['up','down','up','down']],
                 columns=[['pen','pen','paper','paper'],[1,2,1,2]])
ic
```

```
Out[9]:
```

		pen		paper	
		1	2	1	2
white	up	0.061587	0.832907	0.386859	0.020504
	down	0.680455	0.693956	0.825806	0.842088
red	up	0.604831	0.461955	0.946557	0.936826
	down	0.313891	0.813661	0.456050	0.527880

```
In [10]: # Recordering and sorting Levels

ic.columns.names=['objects','id']
ic.index.names=['colors','status']
ic
```

```
Out[10]:
```

objects		pen		paper	
	id	1	2	1	2
colors	status				
white	up	0.061587	0.832907	0.386859	0.020504
	down	0.680455	0.693956	0.825806	0.842088
red	up	0.604831	0.461955	0.946557	0.936826
	down	0.313891	0.813661	0.456050	0.527880

```
In [11]: # return new object with two levels

ic.swaplevel('colors','status')
```

```
Out[11]:
```

objects		pen		paper	
	id	1	2	1	2
status	colors				
up	white	0.061587	0.832907	0.386859	0.020504
	down	0.680455	0.693956	0.825806	0.842088
down	up	0.604831	0.461955	0.946557	0.936826
	down	0.313891	0.813661	0.456050	0.527880

```
In [25]: ic.sort_values(by='colors')
```

```
Out[25]:
```

objects		pen		paper	
---------	--	-----	--	-------	--

		object	1	pen	1	paper
colors	status	1	2	1	2	
colors	status					
red	up	0.604831	0.461955	0.946557	0.936826	
	down	0.313891	0.813661	0.456050	0.527880	
white	up	0.061587	0.832907	0.386859	0.020504	
	down	0.680455	0.693956	0.825806	0.842088	

```
In [31]: ic.sort_index()
```

Out[31]:

	objects		pen		paper	
	id		1	2	1	2
colors	status					
red	down	0.313891	0.813661	0.456050	0.527880	
	up	0.604831	0.461955	0.946557	0.936826	
white	down	0.680455	0.693956	0.825806	0.842088	
	up	0.061587	0.832907	0.386859	0.020504	

```
In [34]: # summary Statistic by level

ic.sum(level='colors')
```

Out[34]:

	objects		pen		paper	
	id		1	2	1	2
colors						
white	up	0.742042	1.526863	1.212664	0.862592	
	down	0.918722	1.275616	1.402606	1.464707	

```
In [39]: ic.sum(level='id',axis=1) # for column (axis=1)
```

Out[39]:

	id		1	2
	colors	status		
white	up	0.448446	0.853412	
	down	1.506261	1.536044	
red	up	1.551388	1.398781	
	down	0.769941	1.341541	

In []: