

```
In [3]: import numpy as np

a = np.array([[1,2,3],[4,5,6]])
print("2D array: \n",a)
```

```
2D array:
[[1 2 3]
 [4 5 6]]
```

```
In [8]: a1 = np.array((1,2,3))
print("2D array tuple: \n",a1)
print("type of a1: ",type(a1))
```

```
2D array tuple:
[1 2 3]
type of a1: <class 'numpy.ndarray'>
```

```
In [12]: a2 = np.arange(20).reshape(5,4)
print("Multiple array using arange module: \n",a2)
```

```
Multiple array using arange module:
[[ 0  1  2  3]
 [ 4  5  6  7]
 [ 8  9 10 11]
 [12 13 14 15]
 [16 17 18 19]]
```

```
In [13]: a3 = np.random.randint(20,35,16).reshape(4,-1)
print("matrix: \n",a3)
```

```
matrix:
[[23 26 21 21]
 [23 32 20 31]
 [26 25 20 30]
 [23 30 30 26]]
```

## Pandas Module:

- it is one of the most important modules in data science
- Used to work with data in the form of frames/table
- Stands for Panel Data
- used for:
  - Data analysis, data manipulation and data cleaning
  - Data analysis: Analyzing the data
    - that means the statistics of the data, Volume of the data etc...
  - Data Manipulation: Transforming the data from one format to another format
    - like Merging, Concatenating etc....
  - Data Cleaning: Removing the Null Data.
    - Null Data means NaN(Not a Number)
- Data structures in pandas

## 1. Series:

- sequential data

## 2. Data frame:

- Data in tabular format i.e; rows and columns

## • Installation

- pip install pandas
- import pandas

```
In [3]: import pandas as pd
```

```
In [13]: dic = {1:"mounav",2:"ajay",3:"charan",4:"rizwanullah"}
data = {'Name': ['mounav', 'ajay', 'charan', 'rizwanullah'], 'Roll': [291, '2A1', 268, 299]}
d1 = pd.DataFrame(data)
print("Tabular data: \n",d1)
```

Tabular data:

	Name	Roll
0	mounav	291
1	ajay	2A1
2	charan	268
3	rizwanullah	299

```
In [14]: pd.__version__
```

```
Out[14]: '1.2.4'
```

```
In [15]: # we can create series using str,tuple,list, and dict
```

```
st = "I am a Student of Vresec"
r = pd.Series(st)
print("Series of string: ",r)
```

```
Series of string: 0    I am a Student of Vresec
dtype: object
```

```
In [16]: # try with input().split()
```

```
li = input("enter a statement: ").split()
print("Result: \n",pd.Series(li))
```

```
enter a statement: hello my name is M.sai
Result:
0    hello
1      my
2    name
3      is
4    M.sai
dtype: object
```

```
In [17]: # using tuple
print("Result: \n",pd.Series(range(10,30,6)))
# print("Result_1: \n",pd.Series((1,2,3,4))
```

```
Result:
0    10
1    16
2    22
3    28
dtype: int64
```

```
In [18]: # Using Dictionary
k = {'34':"thirty Four", 12:90, 'tuple':(9,8.2,'sai'), 'list':[1,2,3,4]}
s = pd.Series(k)
print("Tabular form: \n\n",s)
print("index values: ",s.index)
```

Tabular form:

```
34      thirty Four
12      90
tuple   (9, 8.2, sai)
list    [1, 2, 3, 4]
dtype: object
index values: Index(['34', 12, 'tuple', 'list'], dtype='object')
```

```
In [19]: # creating series using numpy arrays
import numpy as np

ar = np.arange(10,40,5)
print("array: ",ar)

ar1 = pd.Series(ar)
print("index values also: \n",ar1)
```

```
array: [10 15 20 25 30 35]
index values also:
0    10
1    15
2    20
3    25
4    30
5    35
dtype: int32
```

```
In [20]: # modifying index values
print("index data: ",ar1.index)
```

```
index data: RangeIndex(start=0, stop=6, step=1)
```

```
In [21]: ar1.index = [4,5,6,7,9,1]
print("user index: \n",ar1)
```

```
user index:
 4    10
 5    15
 6    20
 7    25
 9    30
 1    35
dtype: int32
```

```
In [22]: li2 = [9,3,4,5,6,'hi']
inx = [num for num in range(5,11)]
pd.Series(li2,index=inx)
```

```
Out[22]: 5      9
 6      3
 7      4
 8      5
 9      6
10     hi
dtype: object
```

```
In [23]: # passing the user index to series created by dict
print("dictionary: ",dic," length od dictionary: ",len(dic))
print()

ix = [num for num in range(10,14)]
e = pd.Series(dic,index = ix)
print("table: \n",e)
```

```
dictionary: {1: 'mounav', 2: 'ajay', 3: 'charan', 4: 'rizwanullah'}
length od dictionary: 4

table:
 10    NaN
 11    NaN
 12    NaN
 13    NaN
dtype: object
```

```
In [24]: s.index
```

```
Out[24]: Index(['34', 12, 'tuple', 'list'], dtype='object')
```

```
In [25]: s['tuple']
```

```
Out[25]: (9, 8.2, 'sai')
```

## Data Frames:

- Data is in 2D Format

```
In [26]: # dictionary into dataframe
print("data dictionary: ",data)
pd.DataFrame(data)
```

```
data dictionary: { 'Name': ['mounav', 'ajay', 'charan', 'rizwanullah'], 'Roll': [291, '2A1', 268, 299]}
```

Out[26]:

	Name	Roll
0	mounav	291
1	ajay	2A1
2	charan	268
3	rizwanullah	299

In [27]:

```
s
```

Out[27]:

```
34      thirty Four
12          90
tuple    (9, 8.2, sai)
list     [1, 2, 3, 4]
dtype: object
```

In [28]:

```
df = pd.DataFrame(s)
print("Tabular Form: \n",df)
```

Tabular Form:

```
          0
34      thirty Four
12          90
tuple    (9, 8.2, sai)
list     [1, 2, 3, 4]
```

In [38]:

```
# pd.DataFrame(k)
```

In [30]:

```
k
```

Out[30]:

```
{'34': 'thirty Four', 12: 90, 'tuple': (9, 8.2, 'sai'), 'list': [1, 2, 3, 4]}
```

## Concatination Of 2 Data-Frames:

- the 2 data-frames must and should have same size or dimensions

In [31]: `des = {'int':89,'float':90.67,'str':'M.sai','prgm':'apssdc'}  
pd.DataFrame(des,index = [1,2,3])`

Out[31]:

	int	float	str	prgm
1	89	90.67	M.sai	apssdc
2	89	90.67	M.sai	apssdc
3	89	90.67	M.sai	apssdc

In [32]: `df2 = {'friend': ['prabhuram', 'saiesh', 'manohar', 'srinivas'], 'admin_no': [288, 2C7, 295, 2A4]}  
df4 = pd.DataFrame(df2)  
df4`

Out[32]:

	friend	admin_no
0	prabhuram	288
1	saiesh	2C7
2	manohar	295
3	srinivas	2A4

In [33]: `df3 = pd.DataFrame(data)  
df3`

Out[33]:

	Name	Roll
0	mounav	291
1	ajay	2A1
2	charan	268
3	rizwanullah	299

In [34]: `# Concatinating 2 Data Frames  
pd.concat([df3,df4]) # df3 and df4 should have same size`

Out[34]:

	Name	Roll	friend	admin_no
0	mounav	291	NaN	NaN
1	ajay	2A1	NaN	NaN
2	charan	268	NaN	NaN
3	rizwanullah	299	NaN	NaN
0	NaN	NaN	prabhuram	288
1	NaN	NaN	saiesh	2C7
2	NaN	NaN	manohar	295
3	NaN	NaN	srinivas	2A4

```
In [35]: # import os
# os.system("shutdown /s")
```

## Merging 2 Data-Frames

```
In [36]: # merging 2 data frames
```

```
c = pd.DataFrame({'Int_rate':[2,1,2,3], 'Int_gdp':[50,45,45,67]}, index = [2041,2002,2003,2204])
print("result_1: \n\n",c)
```

result\_1:

	Int_rate	Int_gdp
2041	2	50
2002	1	45
2003	2	45
2204	3	67

```
In [37]: u = pd.DataFrame({'low_tier_hpi':[80,90,70,60], 'Unemployment':[1,3,5,6]}, index = [2001,2003,2003,2005])
print("result_2: \n\n",u)
```

result\_2:

	low_tier_hpi	Unemployment
2001	80	1
2003	90	3
2003	70	5
2005	60	6

```
In [44]: # joining the data-frames
```

```
res = c.join(u)
print("Joining the data- frames: \n\n",res)
```

Joining the data- frames:

	Int_rate	Int_gdp	low_tier_hpi	Unemployment
2002	1	45	NaN	NaN
2003	2	45	90.0	3.0
2003	2	45	70.0	5.0
2041	2	50	NaN	NaN
2204	3	67	NaN	NaN

In [45]: # merging the data-frames having similar column values

```
m = pd.DataFrame({'a':[4,5,6,7], 'b':[30,50,60,70], 'c':['a','b','c','d']}, index=m)
print("Data-Frame: \n\n",m)
```

Data-Frame:

	a	b	c
first	4	30	a
second	5	50	b
third	6	60	c
fourth	7	70	d

Out[45]:

	a	b	c
<b>first</b>	4	30	a
<b>second</b>	5	50	b
<b>third</b>	6	60	c
<b>fourth</b>	7	70	d

In [47]:

```
n = pd.DataFrame({'a':[4,9,6,7], 'b':[30,80,60,70], 'c':['a','b','c','d']}, index=n)
new = m.merge(n)
print("merged frame: \n\n",new)
m.merge(n)
```

merged frame:

	a	b	c
0	4	30	a
1	6	60	c
2	7	70	d

Out[47]:

	a	b	c
<b>0</b>	4	30	a
<b>1</b>	6	60	c
<b>2</b>	7	70	d

```
In [50]: new.columns = ['f','s','th'] # renaming the column values
new.rows = [5,6,7] # updating the row values
print("modified frame: \n\n",new)
new
```

modified frame:

	f	s	th
5	4	30	a
6	6	60	c
7	7	70	d

```
<ipython-input-50-84e120910af3>:2: UserWarning: Pandas doesn't allow columns to
be created via a new attribute name - see https://pandas.pydata.org/pandas-docs/stable/indexing.html#attribute-access (https://pandas.pydata.org/pandas-docs/stable/indexing.html#attribute-access)
new.rows = [5,6,7] # updating the row values
```

Out[50]:

	f	s	th
5	4	30	a
6	6	60	c
7	7	70	d

```
In [51]: new.columns = ['f','s','th'] # renaming the column values
new.index = [5,6,7] # updating the row values
print("modified frame: \n\n",new)
new
```

modified frame:

	f	s	th
5	4	30	a
6	6	60	c
7	7	70	d

Out[51]:

	f	s	th
5	4	30	a
6	6	60	c
7	7	70	d

In [54]: # merging the dataframes based on particular column value

```
h = m.merge(n, on = 'c')
print("particular coulmn: \n\n",h)

m.merge(n, on = 'c')
```

particular coulmn:

	a_x	b_x	c	a_y	b_y
0	4	30	a	4	30
1	5	50	b	9	80
2	6	60	c	6	60
3	7	70	d	7	70

Out[54]:

	a_x	b_x	c	a_y	b_y
0	4	30	a	4	30
1	5	50	b	9	80
2	6	60	c	6	60
3	7	70	d	7	70

In [64]: # data range

```
dates = pd.date_range('2022-01-01', '2022-01-10')
s3 = pd.Series(dates)
print("range of dates: \n",s3)
pd.Series(dates)
```

range of dates:

0	2022-01-01
1	2022-01-02
2	2022-01-03
3	2022-01-04
4	2022-01-05
5	2022-01-06
6	2022-01-07
7	2022-01-08
8	2022-01-09
9	2022-01-10

dtype: datetime64[ns]

Out[64]: 0 2022-01-01

0	2022-01-01
1	2022-01-02
2	2022-01-03
3	2022-01-04
4	2022-01-05
5	2022-01-06
6	2022-01-07
7	2022-01-08
8	2022-01-09
9	2022-01-10

dtype: datetime64[ns]

```
In [65]: dt = pd.date_range('2022-01-10', periods = 5)
s1 = pd.Series(dt)
print("series of calendar dates: \n",s1)

pd.Series(dt)
```

```
series of calendar dates:
 0    2022-01-10
 1    2022-01-11
 2    2022-01-12
 3    2022-01-13
 4    2022-01-14
dtype: datetime64[ns]
```

```
Out[65]: 0    2022-01-10
 1    2022-01-11
 2    2022-01-12
 3    2022-01-13
 4    2022-01-14
dtype: datetime64[ns]
```

```
In [66]: temp = [30,32,35,40,31]
s2 = pd.Series(temp, index = dt)
print("temperatures on days: \n\n",s2)

pd.Series(temp, index = dt)
```

```
temperatures on days:

 2022-01-10    30
 2022-01-11    32
 2022-01-12    35
 2022-01-13    40
 2022-01-14    31
Freq: D, dtype: int64
```

```
Out[66]: 2022-01-10    30
 2022-01-11    32
 2022-01-12    35
 2022-01-13    40
 2022-01-14    31
Freq: D, dtype: int64
```

## Working with .csv file:

- reading .csv located at same location
- Syntax:
  - pd.read\_csv('filename')

```
In [67]: dir(pd) # pd ==> pandas
```

```
plotting ,
'qcut',
'read_clipboard',
'read_csv',
'read_excel',
'read_feather',

'read_fwf',
'read_gbq',
'read_hdf',
'read_html',
'read_json',
'read_orc',
'read_parquet',
'read_pickle',
'read_sas',
'read_spss',
'read_sql',
'read_sql_query',
'read_sql_table',
'read_stata'
```

```
In [71]: marks = pd.read_csv('marks.csv')
marks
```

Out[71]:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.00	95	100
1	2A0	Sai	89	91	93.00	89	99
2	295	manohar	88	71	93.00	49	69
3	291	mounav	85	94	93.00	85	98
4	2A1	ajay	83	98	99.00	99	100
5	2C7	sairesh	99	98	93.00	100	96
6	288	prabhuram	86	92	90.00	88	99
7	268	Charan	100	100	100.00	100	100
8	293	Uday	79	81	92.00	99	59
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99

In [72]: `print("Student marks Table: \n\n",marks)`

Student marks Table:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.00	95	100
1	2A0	Sai	89	91	93.00	89	99
2	295	manohar	88	71	93.00	49	69
3	291	mounav	85	94	93.00	85	98
4	2A1	ajay	83	98	99.00	99	100
5	2C7	sairesh	99	98	93.00	100	96
6	288	prabhuram	86	92	90.00	88	99
7	268	Charan	100	100	100.00	100	100
8	293	Uday	79	81	92.00	99	59
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99

In [76]: `# Data-Processing  
# Select 5 Random Sample  
marks.sample() # Random sample in our data`

Out[76]:

Roll	Name	Maths	Discrete	OS	CO	DS
5	2C7	sairesh	99	98	93.0	100

In [78]: `marks.sample(5) # generates 5 rows with student marks`

Out[78]:

	Roll	Name	Maths	Discrete	OS	CO	DS
7	268	Charan	100	100	100.0	100	100
6	288	prabhuram	86	92	90.0	88	99
2	295	manohar	88	71	93.0	49	69
1	2A0	Sai	89	91	93.0	89	99
4	2A1	ajay	83	98	99.0	99	100

```
In [86]: s4 = marks.head()
print("top of the rows: \n",s4)
print("\ntop 3 rows or samples: \n",marks.head(3))

marks.head()
```

top of the rows:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.0	95	100
1	2A0	Sai	89	91	93.0	89	99
2	295	manohar	88	71	93.0	49	69
3	291	mounav	85	94	93.0	85	98
4	2A1	ajay	83	98	99.0	99	100

top 3 rows or samples:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.0	95	100
1	2A0	Sai	89	91	93.0	89	99
2	295	manohar	88	71	93.0	49	69

Out[86]:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.0	95	100
1	2A0	Sai	89	91	93.0	89	99
2	295	manohar	88	71	93.0	49	69
3	291	mounav	85	94	93.0	85	98
4	2A1	ajay	83	98	99.0	99	100

In [87]: marks.head(3)

Out[87]:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.0	95	100
1	2A0	Sai	89	91	93.0	89	99
2	295	manohar	88	71	93.0	49	69

```
In [84]: s5 = marks.tail()
print("last rows: \n",s5)
print("\nlast 3 rows or samples: \n",marks.tail(3))

marks.tail()
```

last rows:

	Roll	Name	Maths	Discrete	OS	CO	DS
6	288	prabhuram	86	92	90.00	88	99
7	268	Charan	100	100	100.00	100	100
8	293	Uday	79	81	92.00	99	59
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99

last 3 rows or samples:

	Roll	Name	Maths	Discrete	OS	CO	DS
8	293	Uday	79	81	92.00	99	59
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99

Out[84]:

	Roll	Name	Maths	Discrete	OS	CO	DS
6	288	prabhuram	86	92	90.00	88	99
7	268	Charan	100	100	100.00	100	100
8	293	Uday	79	81	92.00	99	59
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99

In [85]: marks.tail(3)

Out[85]:

	Roll	Name	Maths	Discrete	OS	CO	DS
8	293	Uday	79	81	92.00	99	59
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99

In [89]: marks['Roll'] # getting particular column

```
Out[89]: 0      299
1      2A0
2      295
3      291
4      2A1
5      2C7
6      288
7      268
8      293
9      2C8
10     2B2
Name: Roll, dtype: object
```

In [90]: `marks[['Roll', 'Name', 'CO']]`

Out[90]:

	Roll	Name	CO
0	299	Rizwan	95
1	2A0	Sai	89
2	295	manohar	49
3	291	mounav	85
4	2A1	ajay	99
5	2C7	sairesh	100
6	288	prabhuram	88
7	268	Charan	100
8	293	Uday	99
9	2C8	Yesu	100
10	2B2	Shanak	89

In [92]: `help(pd)`

Help on package pandas:

NAME

pandas

DESCRIPTION

pandas - a powerful data analysis and manipulation library for Python

=====

\*\*pandas\*\* is a Python package providing fast, flexible, and expressive data structures designed to make working with "relational" or "labeled" data both easy and intuitive. It aims to be the fundamental high-level building block for doing practical, \*\*real world\*\* data analysis in Python. Additionally, it has the broader goal of becoming \*\*the most powerful and flexible open source data analysis / manipulation tool available in any language\*\*. It is already well on its way toward this goal.

Main Features

-----

Here are just a few of the things that pandas does well:

- Easy handling of missing data in floating point as well as non-floating point data.
- Size mutability: columns can be inserted and deleted from DataFrame and higher dimensional objects
- Automatic and explicit data alignment: objects can be explicitly aligned to a set of labels, or the user can simply ignore the labels and let `Series`, `DataFrame`, etc. automatically align the data for you in computations.
- Powerful, flexible group by functionality to perform split-apply-combine operations on data sets, for both aggregating and transforming data.
- Make it easy to convert ragged, differently-indexed data in other Python and NumPy data structures into DataFrame objects.
- Intelligent label-based slicing, fancy indexing, and subsetting of large data sets.
- Intuitive merging and joining data sets.
- Flexible reshaping and pivoting of data sets.
- Hierarchical labeling of axes (possible to have multiple labels per tick).
- Robust IO tools for loading data from flat files (CSV and delimited), Excel files, databases, and saving/loading data from the ultrafast HDF5 format.
- Time series-specific functionality: date range generation and frequency conversion, moving window statistics, date shifting and lagging.

```
PACKAGE CONTENTS
    _config (package)
    _libs (package)
    _testing
    _typing
    _version
    api (package)
    arrays (package)
    compat (package)
    conftest
    core (package)
    errors (package)
    io (package)
    plotting (package)
    testing
    tests (package)
    tseries (package)
    util (package)

SUBMODULES
    _hashtable
    _lib
    _tslib
    offsets

FUNCTIONS
    __getattr__(name)
        # GH 27101

DATA
    IndexSlice = <pandas.core.indexing._IndexSlice object>
    NA = <NA>
    NaT = NaT
    __docformat__ = 'restructuredtext'
    __git_version__ = '2cb96529396d93b46abab7bbc73a208e708c642e'
    describe_option = <pandas._config.config.CallableDynamicDoc object>
    get_option = <pandas._config.config.CallableDynamicDoc object>
    options = <pandas._config.config.DictWrapper object>
    reset_option = <pandas._config.config.CallableDynamicDoc object>
    set_option = <pandas._config.config.CallableDynamicDoc object>

VERSION
    1.2.4

FILE
    c:\programdata\anaconda3\lib\site-packages\pandas\__init__.py
```

```
In [93]: help(pd.api)
```

Help on package pandas.api in pandas:

NAME

pandas.api - public toolkit API

PACKAGE CONTENTS

extensions (package)  
indexers (package)  
types (package)

FILE

c:\programdata\anaconda3\lib\site-packages\pandas\api\\_\_init\_\_.py

```
In [97]: new1 = pd.read_csv('marks.csv', usecols = ['Name','OS']) # getting particular col  
print(new1)
```

	Name	OS
0	Rizwan	92.00
1	Sai	93.00
2	manohar	93.00
3	mounav	93.00
4	ajay	99.00
5	saiesh	93.00
6	prabhuram	90.00
7	Charan	100.00
8	Uday	92.00
9	Yesu	99.99
10	Shanak	93.00

Out[97]:

	Name	OS
0	Rizwan	92.00
1	Sai	93.00
2	manohar	93.00
3	mounav	93.00
4	ajay	99.00
5	saiesh	93.00
6	prabhuram	90.00
7	Charan	100.00
8	Uday	92.00
9	Yesu	99.99
10	Shanak	93.00

In [98]: `marks.columns`

Out[98]: `Index(['Roll', 'Name', 'Maths', 'Discrete', 'OS', 'CO', 'DS'], dtype='object')`

In [100]: `marks.Roll`

Out[100]:

0	299
1	2A0
2	295
3	291
4	2A1
5	2C7
6	288
7	268
8	293
9	2C8
10	2B2

Name: Roll, dtype: object

In [101]: `marks.index`

Out[101]: `RangeIndex(start=0, stop=11, step=1)`

In [103]: `marks.dtypes # data types of each column`

Out[103]:

Roll	object
Name	object
Maths	int64
Discrete	int64
OS	float64
CO	int64
DS	int64

dtype: object

In [104]: `marks.info # textual information`

Out[104]:

		<bound method DataFrame.info of		Roll	Name	Maths	Discrete	OS
CO	DS	Rizwan	97	100	92.00	95	100	
0	299	Sai	89	91	93.00	89	99	
1	2A0	manohar	88	71	93.00	49	69	
2	295	mounav	85	94	93.00	85	98	
3	291	ajay	83	98	99.00	99	100	
4	2A1	saiesh	99	98	93.00	100	96	
5	2C7	prabhuram	86	92	90.00	88	99	
6	288	Charan	100	100	100.00	100	100	
7	268	Uday	79	81	92.00	99	59	
8	293	Yesu	99	99	99.99	100	99	
9	2C8	Shanak	89	91	93.00	89	99	
10	2B2							>

```
In [108]: w = marks.describe()
print("statistics of the existing data or file: \n\n",w)

marks.describe() # statistics of the data
```

statistics of the existing data or file:

	Maths	Discrete	OS	CO	DS
<b>count</b>	11.000000	11.000000	11.000000	11.000000	11.000000
<b>mean</b>	90.363636	92.272727	94.362727	90.272727	92.545455
<b>std</b>	7.256345	9.034278	3.527504	14.826267	14.334320
<b>min</b>	79.000000	71.000000	90.000000	49.000000	59.000000
<b>25%</b>	85.500000	91.000000	92.500000	88.500000	97.000000
<b>50%</b>	89.000000	94.000000	93.000000	95.000000	99.000000
<b>75%</b>	98.000000	98.500000	96.000000	99.500000	99.500000
<b>max</b>	100.000000	100.000000	100.000000	100.000000	100.000000

Out[108]:

	Maths	Discrete	OS	CO	DS
<b>count</b>	11.000000	11.000000	11.000000	11.000000	11.000000
<b>mean</b>	90.363636	92.272727	94.362727	90.272727	92.545455
<b>std</b>	7.256345	9.034278	3.527504	14.826267	14.334320
<b>min</b>	79.000000	71.000000	90.000000	49.000000	59.000000
<b>25%</b>	85.500000	91.000000	92.500000	88.500000	97.000000
<b>50%</b>	89.000000	94.000000	93.000000	95.000000	99.000000
<b>75%</b>	98.000000	98.500000	96.000000	99.500000	99.500000
<b>max</b>	100.000000	100.000000	100.000000	100.000000	100.000000

In [113]: `help(marks.describe)`

same as the median.

For object data (e.g. strings or timestamps), the result's index will include ``count``, ``unique``, ``top``, and ``freq``. The ``top`` is the most common value. The ``freq`` is the most common value's frequency. Timestamps also include the ``first`` and ``last`` items.

If multiple object values have the highest count, then the ``count`` and ``top`` results will be arbitrarily chosen from among those with the highest count.

For mixed data types provided via a ``DataFrame``, the default is to return only an analysis of numeric columns. If the dataframe consists only of object and categorical data without any numeric columns, the default is to return an analysis of both the object and categorical columns. If ``include='all'`` is provided as an option, the result will include a union of attributes of each type.

The `include` and `exclude` parameters can be used to limit which columns in a ``DataFrame`` are analyzed for the output.

In [112]: `marks.Maths.value_counts() # Maths is a subject name`

Out[112]:

89	2
99	2
97	1
83	1
100	1
85	1
86	1
88	1
79	1

Name: Maths, dtype: int64

In [125]: `marks['Discrete'].value_counts()`

Out[125]:

98	2
100	2
91	2
81	1
99	1
71	1
92	1
94	1

Name: Discrete, dtype: int64

In [126]: `marks['OS'].value_counts()`

```
-----  
KeyError                                                 Traceback (most recent call last)  
C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\base.py in get_l  
oc(self, key, method, tolerance)  
    3079         try:  
-> 3080             return self._engine.get_loc(casted_key)  
    3081         except KeyError as err:  
  
pandas\_libs\index.pyx in pandas._libs.index.IndexEngine.get_loc()  
  
pandas\_libs\index.pyx in pandas._libs.index.IndexEngine.get_loc()  
  
pandas\_libs\hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashT  
able.get_item()  
  
pandas\_libs\hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashT  
able.get_item()  
  
KeyError: 'OS'
```

The above exception was the direct cause of the following exception:

```
KeyError                                                 Traceback (most recent call last)  
<ipython-input-126-f120debf5fa0> in <module>  
----> 1 marks['OS'].value_counts()  
  
C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\frame.py in __getitem__  
(self, key)  
    3022         if self.columns.nlevels > 1:  
    3023             return self._getitem_multilevel(key)  
-> 3024         indexer = self.columns.get_loc(key)  
    3025         if is_integer(indexer):  
    3026             indexer = [indexer]  
  
C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\base.py in get_l  
oc(self, key, method, tolerance)  
    3080             return self._engine.get_loc(casted_key)  
    3081         except KeyError as err:  
-> 3082             raise KeyError(key) from err  
    3083  
    3084         if tolerance is not None:
```

**KeyError**: 'OS'

In [127]: `marks.sort_values('Roll', ascending = False)`

Out[127]:

	Roll	Name	Maths	Discrete	OS	CO	DS
9	2C8	Yesu	99	99	99.99	100	99
5	2C7	sairesh	99	98	93.00	100	96
10	2B2	Shanak	89	91	93.00	89	99
4	2A1	ajay	83	98	99.00	99	100
1	2A0	Sai	89	91	93.00	89	99
0	299	Rizwan	97	100	92.00	95	100
2	295	manohar	88	71	93.00	49	69
8	293	Uday	79	81	92.00	99	59
3	291	mounav	85	94	93.00	85	98
6	288	prabhuram	86	92	90.00	88	99
7	268	Charan	100	100	100.00	100	100

In [128]: `marks.sort_values('Roll', ascending = True)`

Out[128]:

	Roll	Name	Maths	Discrete	OS	CO	DS
7	268	Charan	100	100	100.00	100	100
6	288	prabhuram	86	92	90.00	88	99
3	291	mounav	85	94	93.00	85	98
8	293	Uday	79	81	92.00	99	59
2	295	manohar	88	71	93.00	49	69
0	299	Rizwan	97	100	92.00	95	100
1	2A0	Sai	89	91	93.00	89	99
4	2A1	ajay	83	98	99.00	99	100
10	2B2	Shanak	89	91	93.00	89	99
5	2C7	sairesh	99	98	93.00	100	96
9	2C8	Yesu	99	99	99.99	100	99

In [130]: `marks.sort_values('DS', ascending = False)`

Out[130]:

	Roll	Name	Maths	Discrete	OS	CO	DS
0	299	Rizwan	97	100	92.00	95	100
4	2A1	ajay	83	98	99.00	99	100
7	268	Charan	100	100	100.00	100	100
1	2A0	Sai	89	91	93.00	89	99
6	288	prabhuram	86	92	90.00	88	99
9	2C8	Yesu	99	99	99.99	100	99
10	2B2	Shanak	89	91	93.00	89	99
3	291	mounav	85	94	93.00	85	98
5	2C7	sairesh	99	98	93.00	100	96
2	295	manohar	88	71	93.00	49	69
8	293	Uday	79	81	92.00	99	59

In [138]: `# adding new col into data-frame`

```
marks['Total'] = 100 # modified only in program
marks
```

Out[138]:

	Roll	Name	Maths	Discrete	OS	CO	DS	Total
0	299	Rizwan	97	100	92.00	95	100	100
1	2A0	Sai	89	91	93.00	89	99	100
2	295	manohar	88	71	93.00	49	69	100
3	291	mounav	85	94	93.00	85	98	100
4	2A1	ajay	83	98	99.00	99	100	100
5	2C7	sairesh	99	98	93.00	100	96	100
6	288	prabhuram	86	92	90.00	88	99	100
7	268	Charan	100	100	100.00	100	100	100
8	293	Uday	79	81	92.00	99	59	100
9	2C8	Yesu	99	99	99.99	100	99	100
10	2B2	Shanak	89	91	93.00	89	99	100

In [135]: `marks.Math.sum() # summation of one column data`

Out[135]: 994

In [136]: `marks.sum()`

Out[136]:

	Roll	2992A02952912A12C72882682932C82B2
Name	RizwanSaimanoharmounavajaysaieshprabhuramChara...	994
Maths		1015
Discrete		1037.99
OS		993
CO		1018
DS		1100
Total		
dtype:	object	

In [139]: `marks.Discrete.sum()`

Out[139]: 1015

In [140]: `marks.isnull() # checking for the null values`

Out[140]:

	Roll	Name	Maths	Discrete	OS	CO	DS	Total
0	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False
5	False	False	False	False	False	False	False	False
6	False	False	False	False	False	False	False	False
7	False	False	False	False	False	False	False	False
8	False	False	False	False	False	False	False	False
9	False	False	False	False	False	False	False	False
10	False	False	False	False	False	False	False	False

In [141]: `marks.isna() # same as isnull()`

Out[141]:

	Roll	Name	Maths	Discrete	OS	CO	DS	Total
0	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False
5	False	False	False	False	False	False	False	False
6	False	False	False	False	False	False	False	False
7	False	False	False	False	False	False	False	False
8	False	False	False	False	False	False	False	False
9	False	False	False	False	False	False	False	False
10	False	False	False	False	False	False	False	False

In [142]: `# Finding the sum/count of Null values in our Data`

`marks.isnull().sum()`

Out[142]:

Roll	0
Name	0
Maths	0
Discrete	0
OS	0
CO	0
DS	0
Total	0

`dtype: int64`

In [146]: `# finding the no.of unique values under each col`

`marks.nunique()`

Out[146]:

Roll	11
Name	11
Maths	9
Discrete	8
OS	6
CO	7
DS	6
Total	1

`dtype: int64`

In [144]: # Cleaning of the Data

```
clean = marks.dropna()
print("new data frame without nulls: \n\n",clean)

clean
```

new data frame without nulls:

	Roll	Name	Maths	Discrete	OS	CO	DS	Total
0	299	Rizwan	97	100	92.00	95	100	100
1	2A0	Sai	89	91	93.00	89	99	100
2	295	manohar	88	71	93.00	49	69	100
3	291	mounav	85	94	93.00	85	98	100
4	2A1	ajay	83	98	99.00	99	100	100
5	2C7	sairesh	99	98	93.00	100	96	100
6	288	prabhuram	86	92	90.00	88	99	100
7	268	Charan	100	100	100.00	100	100	100
8	293	Uday	79	81	92.00	99	59	100
9	2C8	Yesu	99	99	99.99	100	99	100
10	2B2	Shanak	89	91	93.00	89	99	100

Out[144]:

	Roll	Name	Maths	Discrete	OS	CO	DS	Total
0	299	Rizwan	97	100	92.00	95	100	100
1	2A0	Sai	89	91	93.00	89	99	100
2	295	manohar	88	71	93.00	49	69	100
3	291	mounav	85	94	93.00	85	98	100
4	2A1	ajay	83	98	99.00	99	100	100
5	2C7	sairesh	99	98	93.00	100	96	100
6	288	prabhuram	86	92	90.00	88	99	100
7	268	Charan	100	100	100.00	100	100	100
8	293	Uday	79	81	92.00	99	59	100
9	2C8	Yesu	99	99	99.99	100	99	100
10	2B2	Shanak	89	91	93.00	89	99	100

```
In [147]: from sklearn.datasets import load_iris  
  
d3 = load_iris()  
d3
```

In [ ]: