

```
In [2]: import pandas as pd
```

```
In [3]: marks = [80,90,100,40]
labels = ["maths","science","physics","chem"]
```

```
In [4]: ser = pd.Series(marks,labels)
```

```
In [5]: ser
```

```
Out[5]: maths    80
         science   90
         physics  100
         chem      40
         dtype: int64
```

```
In [6]: ser = pd.Series(data = marks,index = labels)
```

```
In [7]: ser
```

```
Out[7]: maths    80
         science   90
         physics  100
         chem      40
         dtype: int64
```

```
In [8]: type(ser)
```

```
Out[8]: pandas.core.series.Series
```

```
In [9]: ser['maths']
```

```
Out[9]: 80
```

```
In [10]: ser['maths']
```

```
Out[10]: 80
```

```
In [11]: import numpy as np  
p = np.arange(16).reshape(4,4)  
labels = ['a','b','c','d']  
label2 = ['A','B','C','D']
```

```
In [12]: p
```

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

```
In [13]: df = pd.DataFrame(p,index = labels,columns = label2)
```

```
In [14]: df
```

```
Out[14]:
```

	A	B	C	D
a	0	1	2	3
b	4	5	6	7
c	8	9	10	11
d	12	13	14	15

```
In [15]: df1 = pd.DataFrame([[1,2,3,4],[4,5,6,7],[8,9,10,11],[12,13,14,15,]], index = "A B C D".split(),columns = "a b c d".split())
```

```
In [16]: "A B C D".split()
```

```
Out[16]: ['A', 'B', 'C', 'D']
```

```
In [17]: "a b c d".split()
```

```
Out[17]: ['a', 'b', 'c', 'd']
```

```
In [18]: df1
```

```
Out[18]:
```

	a	b	c	d
A	1	2	3	4
B	4	5	6	7
C	8	9	10	11
D	12	13	14	15

```
In [19]: df1[['b', 'd']]
```

```
Out[19]:
```

	b	d
A	2	4
B	5	7
C	9	11
D	13	15

```
In [20]: df1['a']
```

```
Out[20]: A      1  
          B      4  
          C      8  
          D     12  
Name: a, dtype: int64
```

```
In [21]: df1.loc['C']
```

```
Out[21]: a      8  
          b      9  
          c     10  
          d     11  
Name: C, dtype: int64
```

```
In [22]: df1.iloc[0]
```

```
Out[22]: a    1  
         b    2  
         c    3  
         d    4  
Name: A, dtype: int64
```

```
In [23]: df1['new'] =df1['a']+df1['b']
```

```
In [24]: df1
```

```
Out[24]:   a   b   c   d  new  
          --  
          A   1   2   3   4   3  
          B   4   5   6   7   9  
          C   8   9  10  11  17  
          D  12  13  14  15  25
```

```
In [25]: df1['new2'] = [1,2,4,6]
```

```
In [26]: df1
```

```
Out[26]:   a   b   c   d  new  new2  
          --  
          A   1   2   3   4   3   1  
          B   4   5   6   7   9   2  
          C   8   9  10  11  17  4  
          D  12  13  14  15  25  6
```

```
In [27]: df1.drop('new2',axis =1)
```

```
Out[27]:
```

	a	b	c	d	new
A	1	2	3	4	3
B	4	5	6	7	9
C	8	9	10	11	17
D	12	13	14	15	25

```
In [28]: df1
```

```
Out[28]:
```

	a	b	c	d	new	new2
A	1	2	3	4	3	1
B	4	5	6	7	9	2
C	8	9	10	11	17	4
D	12	13	14	15	25	6

```
In [29]: df1.drop('new2',axis = 1,inplace = True)
```

```
In [30]: df1
```

```
Out[30]:
```

	a	b	c	d	new
A	1	2	3	4	3
B	4	5	6	7	9
C	8	9	10	11	17
D	12	13	14	15	25

```
df1.drop('D')
```

```
In [31]: df1
```

```
Out[31]:
```

	a	b	c	d	new
A	1	2	3	4	3
B	4	5	6	7	9
C	8	9	10	11	17
D	12	13	14	15	25

```
df1.drop('D', inplace = True)
```

```
In [32]: df1
```

```
Out[32]:
```

	a	b	c	d	new
A	1	2	3	4	3
B	4	5	6	7	9
C	8	9	10	11	17
D	12	13	14	15	25

```
In [55]: df1.reset_index()
```

```
Out[55]:
```

	newindex	a	b	c	d	new
0	CA	1	2	3	4	3
1	NY	4	5	6	7	9
2	WY	8	9	10	11	17
3	DR	12	13	14	15	25

```
In [41]: newind = "CA NY WY DR".split()
```

```
In [42]: newind
```

```
Out[42]: ['CA', 'NY', 'WY', 'DR']
```

In [43]: df1

Out[43]:

	a	b	c	d	new
A	1	2	3	4	3
B	4	5	6	7	9
C	8	9	10	11	17
D	12	13	14	15	25

In [44]: df1["newindex"] = newind

In [45]: df1

Out[45]:

	a	b	c	d	new	newindex
A	1	2	3	4	3	CA
B	4	5	6	7	9	NY
C	8	9	10	11	17	WY
D	12	13	14	15	25	DR

In [49]: df1.set_index("newindex", inplace = True)

In [50]: df1

Out[50]:

	a	b	c	d	new
newindex					
CA	1	2	3	4	3
NY	4	5	6	7	9
WY	8	9	10	11	17
DR	12	13	14	15	25

```
In [52]: df1.head(2) # will return first 5 rows of a dataset
```

```
Out[52]:      a   b   c   d   new
```

	a	b	c	d	new
CA	1	2	3	4	3
NY	4	5	6	7	9

```
In [54]: df1.tail(2)
```

```
Out[54]:      a   b   c   d   new
```

	a	b	c	d	new
WY	8	9	10	11	17
DR	12	13	14	15	25

```
In [56]: df1['a'].unique()
```

```
Out[56]: array([ 1,  4,  8, 12], dtype=int64)
```

```
In [57]: df1['a'].value_counts()
```

```
Out[57]: 12    1  
4     1  
1     1  
8     1  
Name: a, dtype: int64
```

```
In [58]: df1.describe()
```

Out[58]:

	a	b	c	d	new
count	4.000000	4.000000	4.000000	4.000000	4.000000
mean	6.250000	7.250000	8.250000	9.250000	13.500000
std	4.787136	4.787136	4.787136	4.787136	9.574271
min	1.000000	2.000000	3.000000	4.000000	3.000000
25%	3.250000	4.250000	5.250000	6.250000	7.500000
50%	6.000000	7.000000	8.000000	9.000000	13.000000
75%	9.000000	10.000000	11.000000	12.000000	19.000000
max	12.000000	13.000000	14.000000	15.000000	25.000000

```
In [59]: df1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 4 entries, CA to DR
Data columns (total 5 columns):
 a    4 non-null int64
 b    4 non-null int64
 c    4 non-null int64
 d    4 non-null int64
 new  4 non-null int64
dtypes: int64(5)
memory usage: 192.0+ bytes
```

- 1- to replace the missing values with mean
- 2- replace the missingvalues with mode
- 3- replce it with median
- 4- your own value

```
In [65]: df = pd.DataFrame({'A':[1,2,np.NaN],
                           'B':[5,np.NaN,np.NaN],
                           'C':[1,2,3]})
```

```
In [66]: df
```

```
Out[66]:   A    B    C
```

	A	B	C
0	1.0	5.0	1
1	2.0	NaN	2
2	NaN	NaN	3

```
In [67]: df.isnull().any
```

```
Out[67]: <bound method DataFrame.any of      A    B    C  
          0  False  False  False  
          1  False   True  False  
          2   True   True  False>
```

```
In [68]: df.isnull().any()
```

```
Out[68]: A      True  
         B      True  
         C     False  
        dtype: bool
```

```
In [69]: df.isnull().sum()
```

```
Out[69]: A    1  
         B    2  
         C    0  
        dtype: int64
```

```
In [70]: df.dropna()
```

```
Out[70]:   A    B    C
```

	A	B	C
0	1.0	5.0	1

```
In [71]: df.dropna(axis = 1)
```

```
Out[71]:
```

	c
0	1
1	2
2	3

```
In [73]: df.dropna(thresh = 2, axis=1)
```

```
Out[73]:
```

	A	C
0	1.0	1
1	2.0	2
2	NaN	3

```
In [74]: df.fillna(value= 60)
```

```
Out[74]:
```

	A	B	C
0	1.0	5.0	1
1	2.0	60.0	2
2	60.0	60.0	3

```
In [112]: df['A'].fillna(df["A"].mean(), inplace = True)
```

```
In [109]: df['B'].fillna(df["B"].mean(), inplace = True)
```

```
In [110]: df['B'].fillna(df["B"].median(), inplace = True)
```

```
In [78]: p = df['A'].mode()
```

```
In [79]: p
```

```
Out[79]: 0    1.0  
1    2.0  
dtype: float64
```

```
In [81]: p[1]
```

```
Out[81]: 2.0
```

```
In [85]: df['A'].fillna(df["A"].mode()[0])
```

```
Out[85]: 0    1.0  
1    2.0  
2    1.0  
Name: A, dtype: float64
```

```
In [99]: df1 = pd.DataFrame(["Aindia", np.nan, "Aindia", "Australia", "Australia", "us", "canada"], columns = ['country'])
```

```
In [100]: df1
```

```
Out[100]: country  
_____  
0    Aindia  
1      NaN  
2    Aindia  
3   Australia  
4   Australia  
5      us  
6   canada
```

```
In [101]: q = df1['country'].mode()
```

```
In [103]: q[0]
```

```
Out[103]: 'Aindia'
```

```
In [106]: df1['country'].fillna(df1['country'].mode()[1])
```

```
Out[106]: 0      Aindia
1    Australia
2      Aindia
3    Australia
4    Australia
5        us
6    canada
Name: country, dtype: object
```

```
In [113]: df
```

```
Out[113]:   A   B   C
0  1.0  5.0  1
1  2.0  5.0  2
2  1.5  5.0  3
```

mean- ideal condition is if columns data is in a range - age (20,80)
median()- lot of variation salary (1000 to 1000000)
mode - for textual data

create a dataframe with 10 rows and seven columns two columns should be textual
in that make some missing values in four of the columns one in textual column
apply all the functions which are taught

```
In [114]: df2 = pd.DataFrame(np.arange(16).reshape(4,4), "p q r s".split())
```

```
In [115]: df2
```

```
Out[115]:   p   q   r   s
a   0   1   2   3
b   4   5   6   7
c   8   9  10  11
d  12  13  14  15
```

```
In [124]: "a b c d".split()
```

```
Out[124]: ['a', 'b', 'c', 'd']
```

```
In [ ]: pd.DataFrame(data,index list format,columnslist format)
```

```
In [117]: newindex = ["row1","row2","row3","row4"]
```

```
In [118]: df2["newcolumn"] = newindex
```

```
In [119]: df2
```

```
Out[119]:
```

	p	q	r	s	newcolumn
a	0	1	2	3	row1
b	4	5	6	7	row2
c	8	9	10	11	row3
d	12	13	14	15	row4

```
In [122]: df2.set_index("newcolumn",inplace = True)
```

```
In [123]: df2
```

```
Out[123]:
```

	p	q	r	s
newcolumn				
row1	0	1	2	3
row2	4	5	6	7
row3	8	9	10	11
row4	12	13	14	15

```
In [134]: dataset = pd.read_csv(r"D:\pythonbasics\50_Startups.csv")
```

In [127]: dataset

Out[127]:

	R&D Spend	Administration	Marketing Spend	State	Profit
0	165349.20	123456.00	471784.10	New York	192261.83
1	162597.70	151377.59	443898.53	California	191792.06
2	153441.51	101145.55	407934.54	Florida	191050.39
3	144372.41	118671.85	12345.00	New York	182901.99
4	142107.34	91391.77	366168.42	Florida	166187.94
5	131876.90	99814.71	362861.36	New York	156991.12
6	134615.46	147198.87	127716.82	California	156122.51
7	130298.13	145530.06	323876.68	Florida	155752.60
8	120542.52	148718.95	311613.29	New York	152211.77
9	123334.88	108679.17	304981.62	California	149759.96
10	101913.08	110594.11	229160.95	Florida	146121.95
11	100671.96	91790.61	249744.55	California	144259.40
12	93863.75	127320.38	249839.44	Florida	141585.52
13	91992.39	135495.07	252664.93	California	134307.35
14	119943.24	156547.42	256512.92	Florida	132602.65
15	114523.61	122616.84	261776.23	New York	129917.04
16	78013.11	121597.55	264346.06	California	126992.93
17	94657.16	145077.58	282574.31	New York	125370.37
18	91749.16	114175.79	294919.57	Florida	124266.90
19	86419.70	153514.11	0.00	New York	122776.86
20	76253.86	113867.30	298664.47	California	118474.03
21	78389.47	153773.43	299737.29	New York	111313.02
22	73994.56	122782.75	303319.26	Florida	110352.25
23	67532.53	105751.03	304768.73	Florida	108733.99
24	77044.01	99281.34	140574.81	New York	108552.04
25	64664.71	139553.16	137962.62	California	107404.34

	R&D Spend	Administration	Marketing Spend	State	Profit
26	75328.87	144135.98	134050.07	Florida	105733.54
27	72107.60	127864.55	353183.81	New York	105008.31
28	66051.52	182645.56	118148.20	Florida	103282.38
29	65605.48	153032.06	107138.38	New York	101004.64
30	61994.48	115641.28	91131.24	Florida	99937.59
31	61136.38	152701.92	88218.23	New York	97483.56
32	63408.86	129219.61	46085.25	California	97427.84
33	55493.95	103057.49	214634.81	Florida	96778.92
34	46426.07	157693.92	210797.67	California	96712.80
35	46014.02	85047.44	205517.64	New York	96479.51
36	28663.76	127056.21	201126.82	Florida	90708.19
37	44069.95	51283.14	197029.42	California	89949.14
38	20229.59	65947.93	185265.10	New York	81229.06
39	38558.51	82982.09	174999.30	California	81005.76
40	28754.33	118546.05	172795.67	California	78239.91
41	27892.92	84710.77	164470.71	Florida	77798.83
42	23640.93	96189.63	148001.11	California	71498.49
43	15505.73	127382.30	35534.17	New York	69758.98
44	22177.74	154806.14	28334.72	California	65200.33
45	1000.23	124153.04	1903.93	New York	64926.08
46	1315.46	115816.21	297114.46	Florida	49490.75
47	0.00	135426.92	0.00	California	42559.73
48	542.05	51743.15	0.00	New York	35673.41
49	0.00	116983.80	45173.06	California	14681.40

```
In [129]: dataset.isnull().sum()
```

```
Out[129]: R&D Spend      0  
Administration      0  
Marketing Spend    0  
State              0  
Profit             0  
dtype: int64
```

```
In [131]: dataset.tail()
```

```
Out[131]:   R&D Spend  Administration  Marketing Spend  State  Profit  
45      1000.23       124153.04        1903.93  New York  64926.08  
46      1315.46       115816.21        297114.46  Florida  49490.75  
47       0.00       135426.92          0.00  California  42559.73  
48      542.05       51743.15          0.00  New York  35673.41  
49       0.00       116983.80        45173.06  California  14681.40
```

```
In [132]: dataset['State'].unique()
```

```
Out[132]: array(['New York', 'California', 'Florida'], dtype=object)
```

```
In [133]: dataset['State'].value_counts()
```

```
Out[133]: New York    17  
California  17  
Florida     16  
Name: State, dtype: int64
```

```
In [136]: data = pd.DataFrame({'1': ['h', 'e', 'l', 'l', 'o', np.NaN, 'n', 'a', 'i', 't'],
                            '2': ['h', 'e', 'l', 'l', 'o', 'n', 'a', 'i', np.NaN, 't'],
                            '3': ['h', 'e', 'l', 'l', 'o', np.NaN, 'n', 'a', 'i', 't'],
                            '4': ['h', 'e', 'l', 'l', 'o', np.NaN, 'n', 'a', 'i', 't'],
                            '5': ['h', 'e', np.NaN, 'l', 'o', 'n', 'a', 'i', 't', 'k'],
                            '6': ['h', np.NaN, 'l', 'l', 'o', 'k', 'n', 'a', 'i', 't'],
                            '7': ['h', 'e', 'l', 'l', 'o', np.NaN, 'n', 'a', 'i', 't']})
data
```

```
Out[136]:
```

	1	2	3	4	5	6	7
0	h	h	h	h	h	h	h
1	e	e	e	e	e	NaN	e
2	l	l	l	l	l	NaN	l
3	l	l	l	l	l	l	l
4	o	o	o	o	o	o	o
5	NaN	n	NaN	NaN	n	k	NaN
6	n	a	n	n	a	n	n
7	a	i	a	a	i	a	a
8	i	NaN	i	i	t	i	i
9	t	t	t	t	k	t	t

```
In [140]: data['1'].mode()
```

```
Out[140]: 0    1
dtype: object
```

```
In [158]: data['1'].fillna(data['1'].mode()[0], inplace = True)
data['2'].fillna(data['2'].mode()[0], inplace = True)
data['3'].fillna(data['3'].mode()[0], inplace = True)
```

```
In [159]: data
```

```
Out[159]:
```

	1	2	3	4	5	6	7
0	h	h	h	h	h	h	h
1	e	e	e	e	e	NaN	e
2	I	I	I	I	NaN	I	I
3	I	I	I	I	I	I	I
4	o	o	o	o	o	o	o
5	I	n	I	NaN	n	k	NaN
6	n	a	n	n	a	n	n
7	a	i	a	a	i	a	a
8	i	I	i	i	t	i	i
9	t	t	t	t	k	t	t

```
In [143]: d3 = df
```

```
In [144]: d3
```

```
Out[144]:
```

	A	B	C
0	1.0	5.0	1
1	2.0	5.0	2
2	1.5	5.0	3

```
In [145]: d3 = pd.DataFrame({'A':[1,2,np.NaN],  
                           'B':[5,np.NaN,np.NaN],  
                           'C':[1,2,3]})
```

```
In [146]: d3
```

```
Out[146]:
```

	A	B	C
0	1.0	5.0	1
1	2.0	NaN	2
2	NaN	NaN	3

```
In [151]: d3['A'].fillna(d3['A'].mean(), inplace = True)
```

```
In [152]: d3['B'].fillna(d3['B'].median(), inplace = True)
```

```
In [153]: d3
```

```
Out[153]:
```

	A	B	C
0	1.0	5.0	1
1	2.0	5.0	2
2	1.5	5.0	3

```
In [168]: pdataset = pd.read_excel("datasetexcel.xlsx")
```

In [169]: pdataset

Out[169]:

	maths	roll	tutorname	op	std (mean - value)2
0	40.0	1.0		1	3.000000
1	3.0	3.0		3	NaN
2	5.0	NaN		3	55.000000
3	4.0	4.0		3	5.000000
4	3.0	3.0		3	31.000000
5	3.0	3.0		3	35.285714
6	3.0	3.0		3	39.571429
7	NaN	NaN		3	43.857143
8	2.0	2.0		3	48.142857
9	2.0	2.0		3	52.428571
10	2.0	2.0		3	56.714286
11	1.0	1.0		3	61.000000
12	5.0	5.0		3	65.285714
13	6.0	NaN		3	69.571429
14	2.0	2.0		3	73.857143
15	1.0	1.0		3	78.142857
16	3.0	NaN		3	82.428571
17	2.0	2.0		3	86.714286

In []: