

pandasNotes

December 19, 2022

1 Pandas

- is Python library used to analyze data
- has functions for analyzing, cleaning, exploring, and manipulating data
- Pandas refer to “Python Data Analysis”

Why Pandas?

- allows us to analyze big data and make conclusions based on statistical theories.
- can clean messy data sets, and make them readable and relevant.

1.1 Series

- is like a column in a table.
- It is a one-dimensional array holding data of any type.

Labels (*index*) - If nothing else is specified, the values are labeled with their index number. First value has index 0, second value has index 1 etc.

```
[ ]: import pandas as pd

data = pd.Series([0.25, 0.5, 0.75, 1.0])

print(data)
print('-----')
print(type(data))
print('-----')
print(data.values)
print('-----')
print(data.index)
print('-----')
print(data.keys)
print('-----')
```

```
0    0.25
1    0.50
2    0.75
3    1.00
dtype: float64
-----
```

```

<class 'pandas.core.series.Series'>
-----
[0.25 0.5  0.75 1.  ]
-----
RangeIndex(start=0, stop=4, step=1)
-----
<bound method Series.keys of 0    0.25
1    0.50
2    0.75
3    1.00
dtype: float64>
-----

```

describe()

- method returns description of the data in the DataFrame or Series.

count - The number of **not**-empty values.
mean - The average (mean) value.
std - The standard deviation.
min - the minimum value.
25% - The 25% percentile*.
50% - The 50% percentile*.
75% - The 75% percentile*.
max - the maximum value.

```

[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data_description = data.describe()

print(type(data_description))
print('-----')
print(data_description)
print('-----')

```

```

<class 'pandas.core.series.Series'>
-----
count      11.000000
mean       5.363636
std        2.292280
min        2.000000
25%        3.500000
50%        5.000000
75%        7.000000
max        9.000000

```

```
dtype: float64
```

```
-----
```

```
agg()
```

- the same as describe but it doesn't show all description just the information passed for it

```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data_description = data.agg(['min', 'max', 'mean', 'std'])

print(data_description)
print('-----')
```

```
min      2.000000
max      9.000000
mean     5.363636
std      2.292280
dtype: float64
```

```
-----
```

Accessing & Slicing

- here we can access data with index and also slice it as lists

```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

print(data[1:3])
print('-----')
print(data[1:6:2])
print('-----')
print(data[5])
print('-----')
```

```
1      6
2      9
dtype: int64
```

```
-----
```

```
1      6
3      8
5      4
dtype: int64
```

```
-----
```

```
4
```

create index “labels”

- we can set indexes as we want not just use default indices

```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8]
indexs = ['a', 'b', 'c', 'd']
new_data_list = dict(zip(indexs, data_list))

data1 = pd.Series(data_list, index={'a':3, 'b':6, 'c':9, 'd':8})
data2 = pd.Series(data_list, index=indexs)
data3 = pd.Series(new_data_list)
data4 = pd.Series({'a':3, 'b':6, 'c':9, 'd':8})

print(data1)
print('-----')
print(data2)
print('-----')
print(data3)
print('-----')
print(data4)
print('-----')
```

```
a    3
b    6
c    9
d    8
dtype: int64
```

```
a    3
b    6
c    9
d    8
dtype: int64
```

```
a    3
b    6
c    9
d    8
dtype: int64
```

```
a    3
b    6
c    9
d    8
dtype: int64
```

Operators (&, |, ^)

- we use them to filter data
 - &: to get intersection between 2 lists
 - |: union
 - ^: values not repeated in just one of lists and not in another

```
[ ]: import pandas as pd

a = pd.Index([0, 1, 3, 5, 7, 9])
b = pd.Index([2, 3, 5, 6, 9])

print(a)
print('-----')
print(b)
print('-----')
print(a&b)           #intersection
print('-----')
print(a|b)           #union
print('-----')
print(a^b)           #means get numbers not repeated ('just in a or b not in both')
print('-----')
```

```
Int64Index([0, 1, 3, 5, 7, 9], dtype='int64')
```

```
Int64Index([2, 3, 5, 6, 9], dtype='int64')
```

```
Int64Index([3, 5, 9], dtype='int64')
```

```
Int64Index([0, 1, 2, 3, 5, 6, 7, 9], dtype='int64')
```

```
Int64Index([0, 1, 2, 6, 7], dtype='int64')
```

```
/tmp/ipykernel_48679/734037960.py:10: FutureWarning: Index.__and__ operating as
a set operation is deprecated, in the future this will be a logical operation
matching Series.__and__. Use index.intersection(other) instead.
```

```
print(a&b)           #intersection
```

```
/tmp/ipykernel_48679/734037960.py:12: FutureWarning: Index.__or__ operating as a
set operation is deprecated, in the future this will be a logical operation
matching Series.__or__. Use index.union(other) instead.
```

```
print(a|b)           #union
```

```
/tmp/ipykernel_48679/734037960.py:14: FutureWarning: Index.__xor__ operating as
a set operation is deprecated, in the future this will be a logical operation
matching Series.__xor__. Use index.symmetric_difference(other) instead.
```

```
print(a^b)           #means get numbers not repeated ('just in a or b not in both')
```

1.2 Plotting

plot()

- uses to create diagrams.
- We can use Pyplot, a submodule of the Matplotlib library to visualize the diagram on the screen.

kind - line : default - pie - bar, barh - hist - box - kde - area

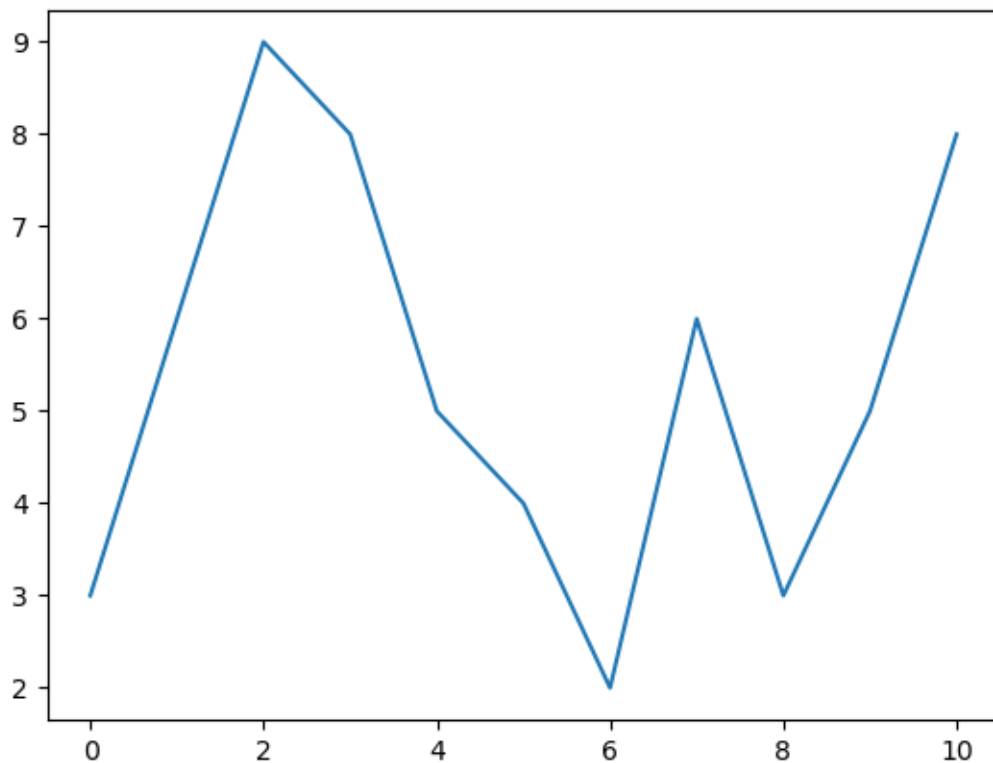
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot()
#data.plot(kind='line')
```

```
[ ]: <AxesSubplot:>
```



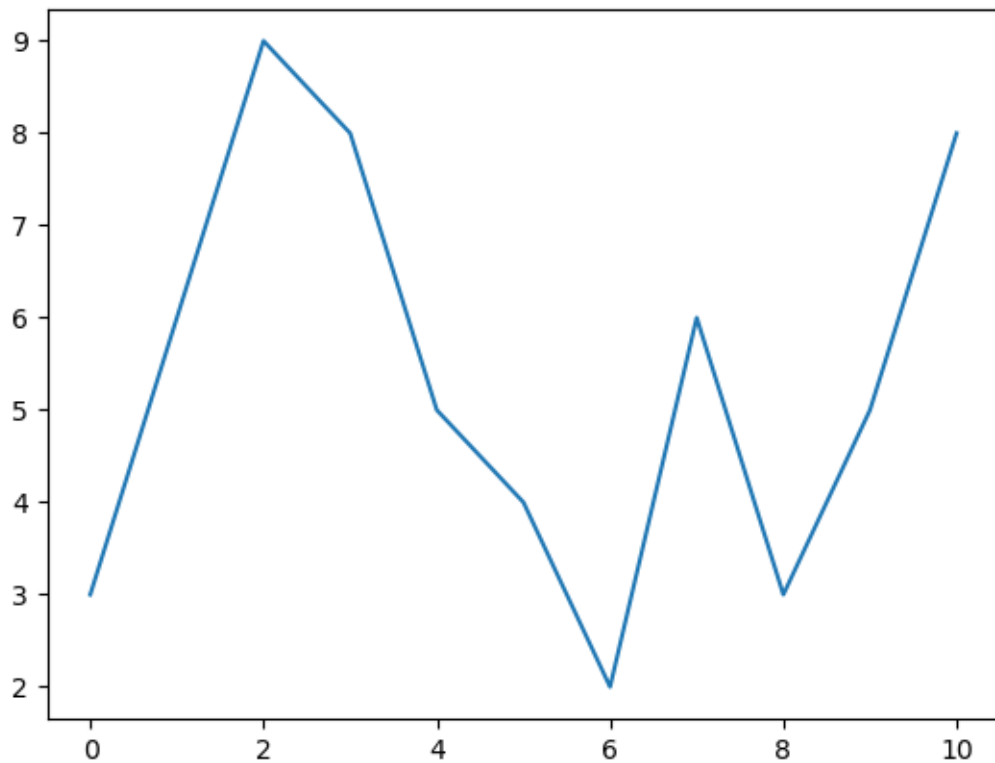
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]
```

```
data = pd.Series(data_list)

data.plot(kind='line')
```

```
[ ]: <AxesSubplot:>
```



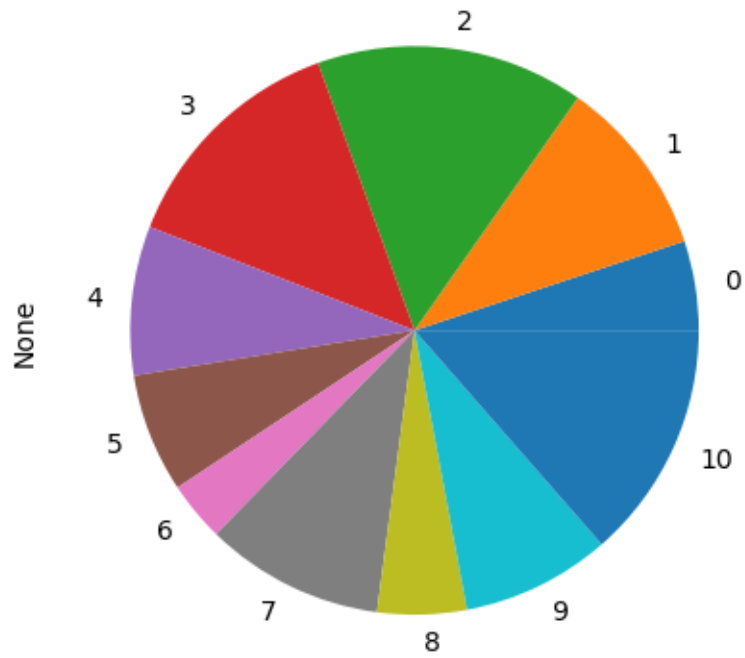
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

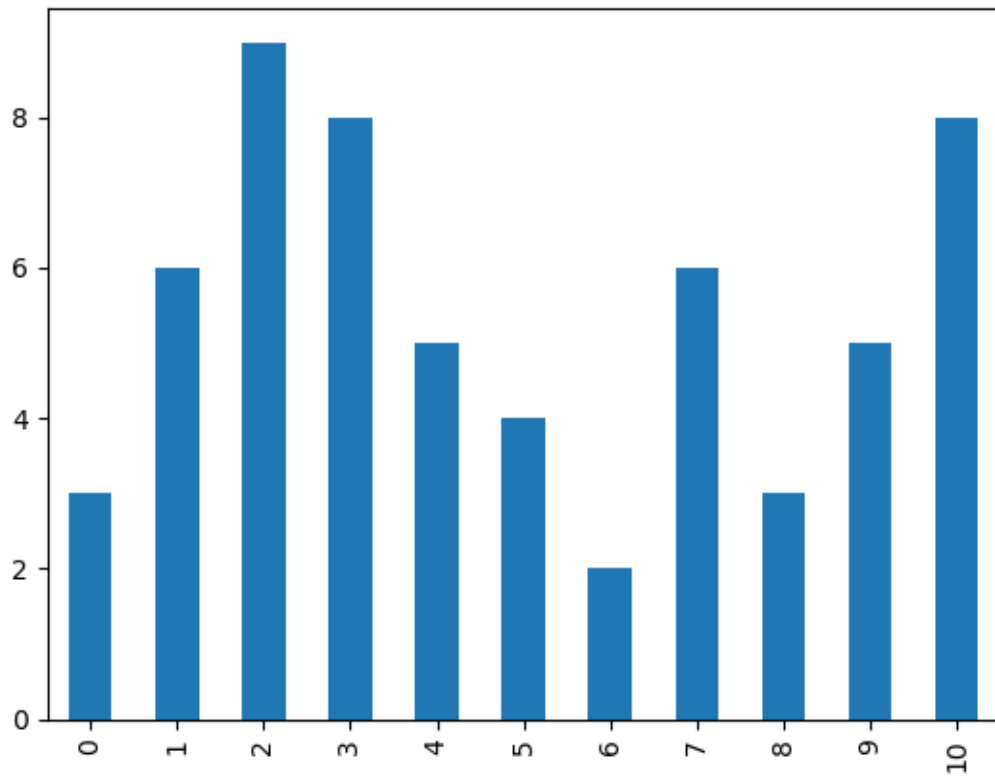
data.plot(kind='pie')
```

```
[ ]: <AxesSubplot:ylabel='None'>
```



```
[ ]: import pandas as pd  
  
data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]  
  
data = pd.Series(data_list)  
  
data.plot(kind='bar')
```

```
[ ]: <AxesSubplot:>
```

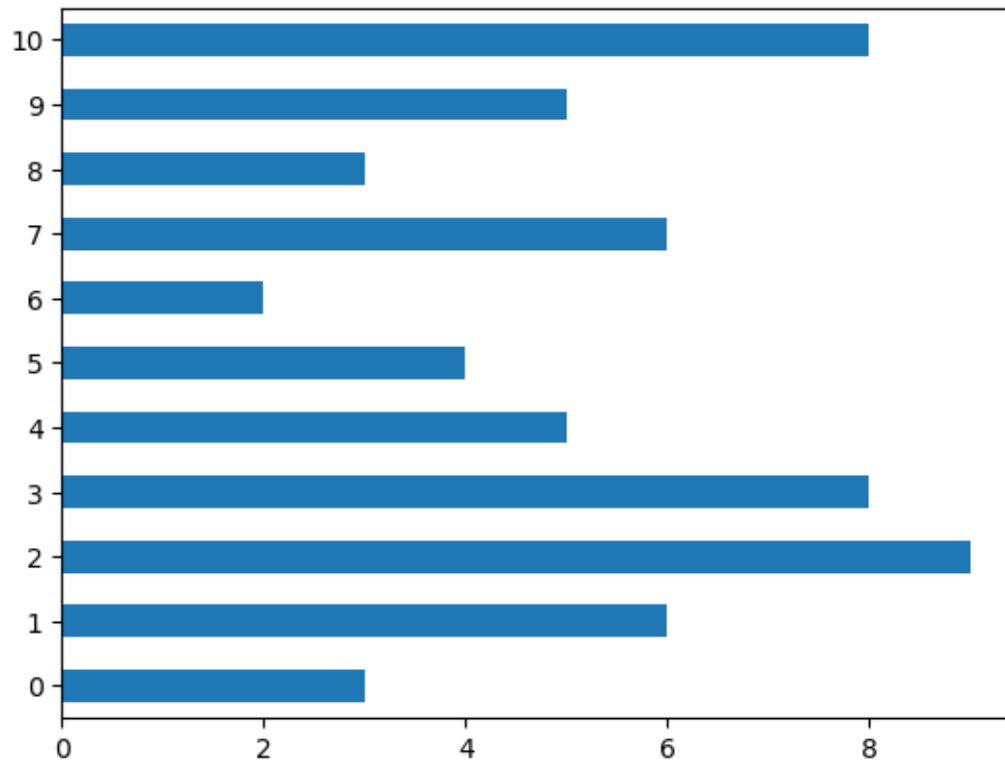
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot(kind='barh')
```

```
[ ]: <AxesSubplot:>
```



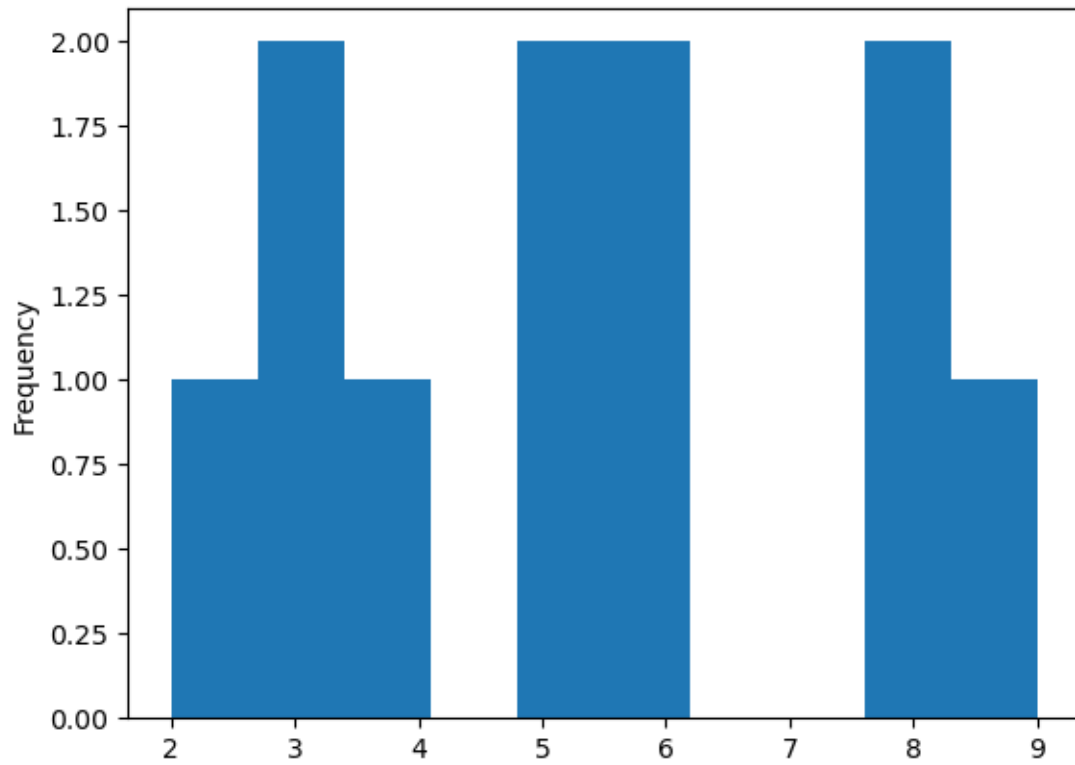
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot(kind='hist')
```

```
[ ]: <AxesSubplot:ylabel='Frequency'>
```



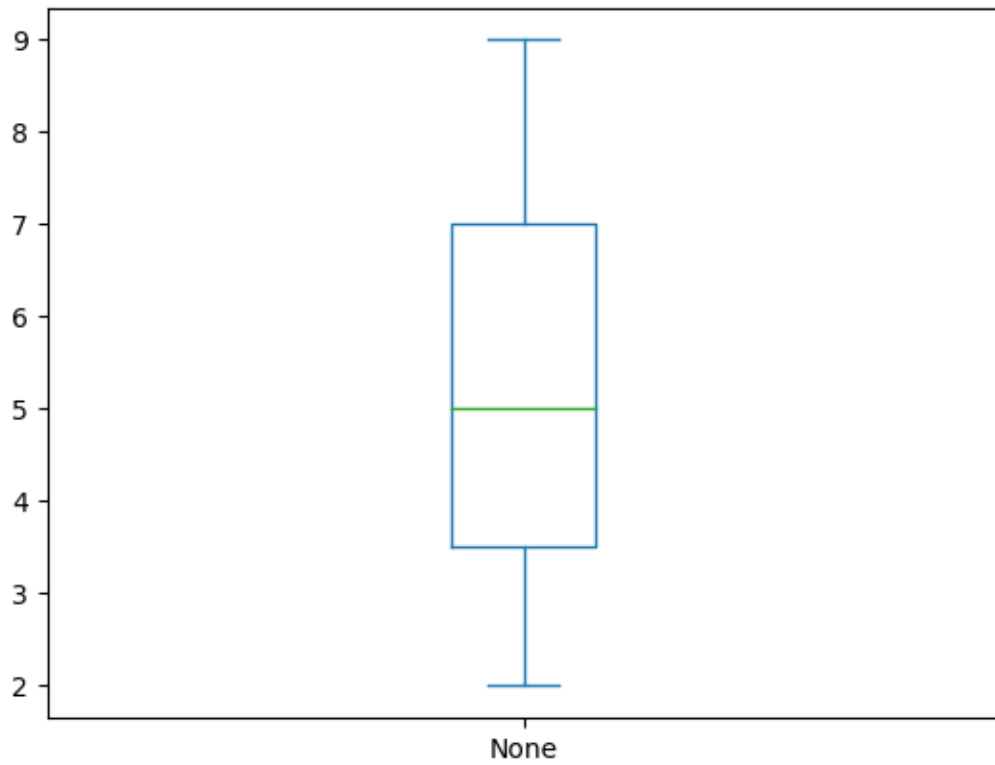
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot(kind='box')
```

```
[ ]: <AxesSubplot:>
```



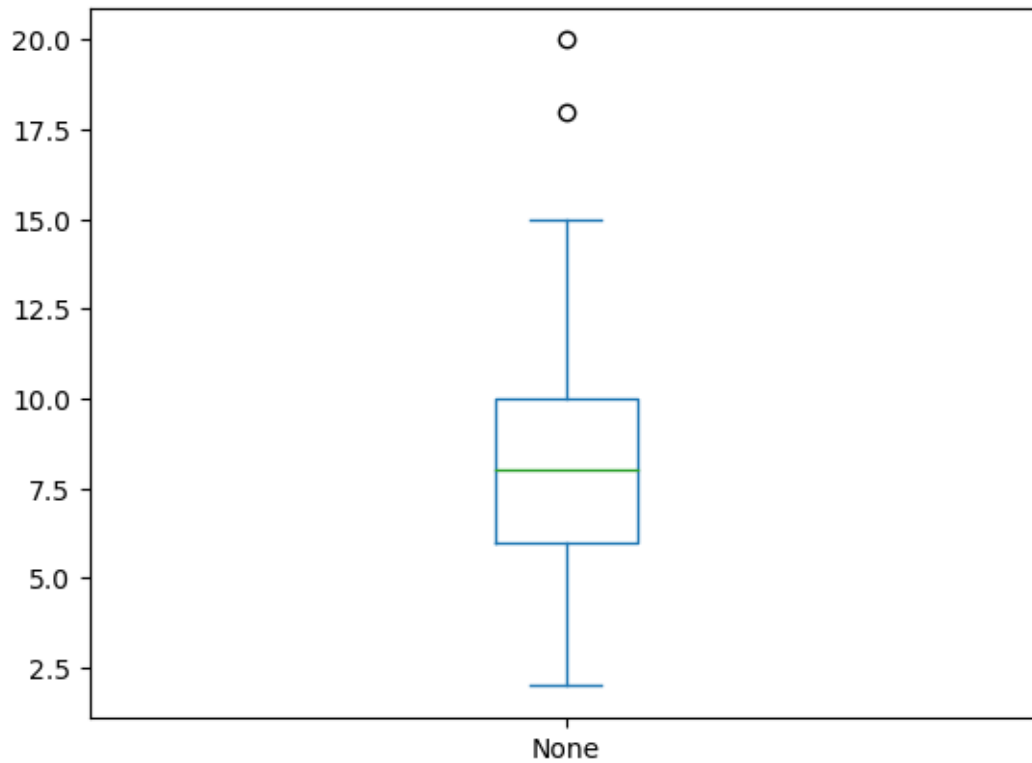
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 10, 4, 2, 6, 10, 20, 15, 18, 8]

data = pd.Series(data_list)

data.plot(kind='box')
```

```
[ ]: <AxesSubplot:>
```



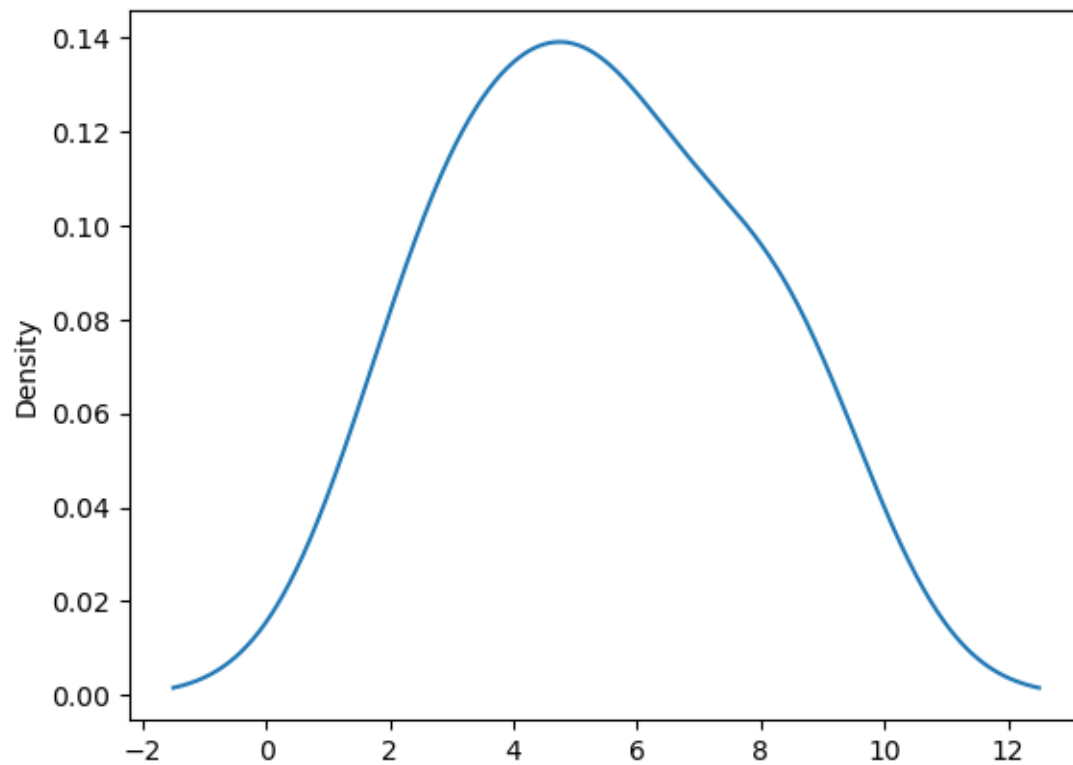
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot(kind="kde")
```

```
[ ]: <AxesSubplot:ylabel='Density'>
```



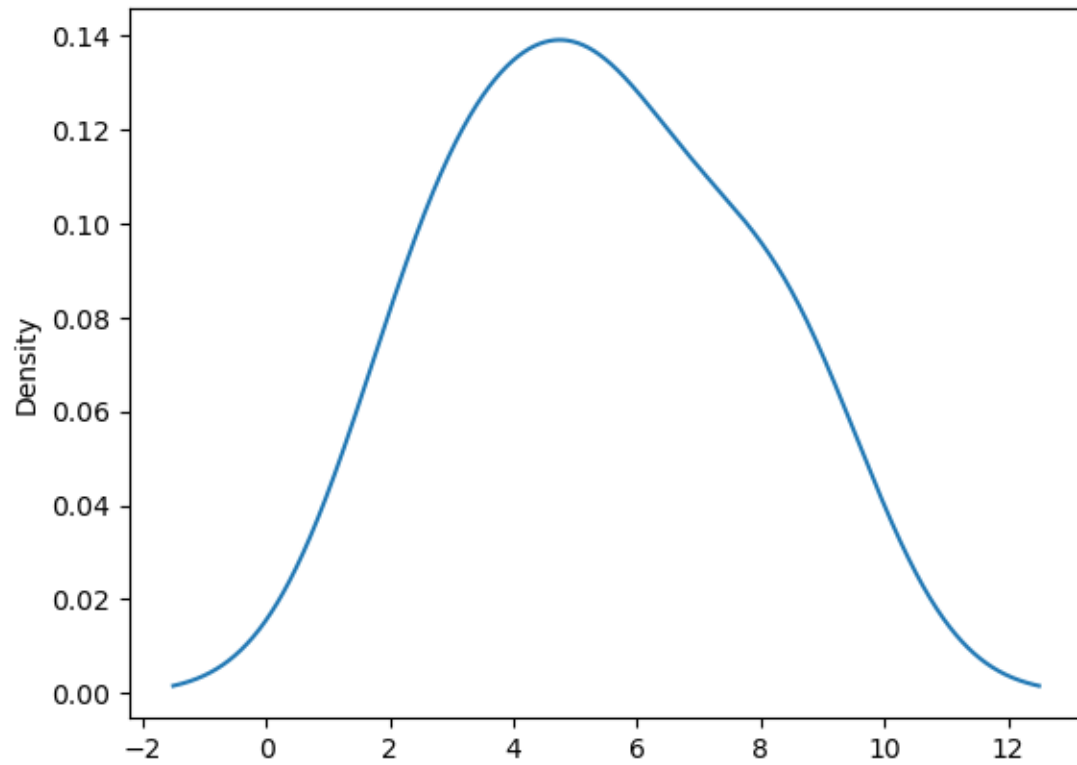
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot(kind='density')
```

```
[ ]: <AxesSubplot:ylabel='Density'>
```



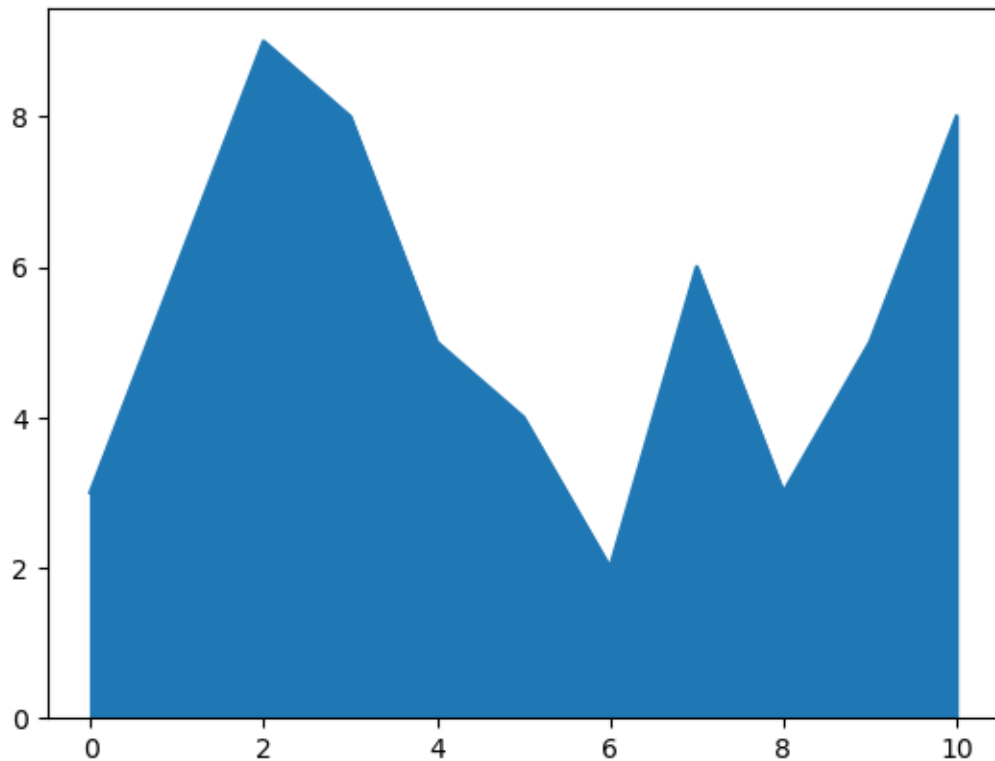
```
[ ]: import pandas as pd

data_list = [3, 6, 9, 8, 5, 4, 2, 6, 3, 5, 8]

data = pd.Series(data_list)

data.plot(kind='area')
```

```
[ ]: <AxesSubplot:>
```



1.3 DataFrame

- is a 2 dimensional data structure, like a 2 dimensional array, or a table with rows and columns.

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=15).reshape(3, 5)

rows_names = ['a', 'b', 'c']
columns_names = ['I', 'II', 'III', 'IV', 'V']

data_frame = pd.DataFrame(data=data, index=rows_names, columns=columns_names)

print(data_frame)
print('-----')
```

	I	II	III	IV	V
a	6	51	87	13	52
b	11	58	36	15	67
c	9	79	14	94	15

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w , x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print('-----data-----')
print(data)
print('-----w-----')
print(w)
print('-----x-----')
print(x)
print('-----y-----')
print(y)
print('-----z-----')
print(z)
print('-----df-----')
print(df)
print('-----')
```

```
-----data-----
[[24 35 98 92 73]
 [19 95 22  4 22]
 [55 53 20 90 27]
 [88  8 29 21 76]]

-----w-----
{'a': 24, 'b': 35, 'c': 98, 'd': 92, 'e': 73}

-----x-----
{'a': 19, 'b': 95, 'c': 22, 'd': 4, 'e': 22}

-----y-----
{'a': 55, 'b': 53, 'c': 20, 'd': 90, 'e': 27}

-----z-----
{'a': 88, 'b': 8, 'c': 29, 'd': 21, 'e': 76}

-----df-----
   Math  Physics  French  Chemistry
a     24      19     55         88
b     35      95     53          8
c     98      22     20         29
d     92       4     90         21
e     73      22     27         76

-----
```

Accessing Columns

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w , x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print(df['Math'])
print('-----')
```

```
a    53
b    22
c     2
d     3
e    22
Name: Math, dtype: int64
-----
```

Transpose

- T property is used to transpose index and columns of the data frame. The property T is somehow related to method transpose().
- The main function of this property is to create a reflection of the data frame over the main diagonal by making rows as columns and vice versa.

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w , x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print(df)
print('-----')
print(df.T)
print('-----')
```

```
      Math  Physics  French  Chemistry
a      88      78      24         96
b      99      54      28          6
c      87      19      74         90
d      19      16      36         13
e      37      94      57         41
-----
```

```
      a      b      c      d      e
```

Math	88	99	87	19	37
Physics	78	54	19	16	94
French	24	28	74	36	57
Chemistry	96	6	90	13	41

keys() & values

- keys(): use to get column names
- values: return row values
- index: return row names

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w, x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print(df.keys())
print('-----')
print(df.values)
print('-----')
print(df.index)
print('-----')
```

```
Index(['Math', 'Physics', 'French', 'Chemistry'], dtype='object')
```

```
[[15 25 37 87]
 [ 6 95 10 65]
 [85 22 14 98]
 [90 59 17 42]
 [98 33 51 94]]
```

```
Index(['a', 'b', 'c', 'd', 'e'], dtype='object')
```

check

- we can check if an element is in DataFrame or not with using in
 - check in keys()
 - check in values

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
```

```
w , x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print('Math' in df.keys())
print('-----')
print('math' in df.keys())
print('-----')
print(12 in df.values)
print('-----')
print(55 in df.values)
print('-----')
```

True

False

True

False

stack & stack

- **stack**: used to stack the prescribed level(s) from columns to index. Return a reshaped DataFrame or Series having a multi-level index with one or more new inner-most levels compared to the current DataFrame.
- **unstack**: unstack the prescribed level(s) from row to column.

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w , x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print(df)
print('-----')
new_df = df.stack()
print(new_df)
print('-----')
print(new_df.unstack())
print('-----')
```

	Math	Physics	French	Chemistry
a	25	56	3	47
b	43	80	37	20

c	27	57	56	77
d	96	7	80	74
e	87	80	48	44

```

-----
a  Math      25
   Physics   56
   French     3
   Chemistry 47
b  Math      43
   Physics   80
   French    37
   Chemistry 20
c  Math      27
   Physics   57
   French    56
   Chemistry 77
d  Math      96
   Physics    7
   French    80
   Chemistry 74
e  Math      87
   Physics   80
   French    48
   Chemistry 44
dtype: int64

```

```

-----
      Math  Physics  French  Chemistry
a      25     56      3       47
b      43     80     37       20
c      27     57     56       77
d      96      7     80       74
e      87     80     48       44
-----

```

locate

As you can see from the result above, the DataFrame is like a table with rows and columns.

- Pandas use the `loc` attribute to return one or more specified row(s)
- we can use `loci` to locate with indices

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w, x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

```

```
print(df.iloc[:3, :2])
print('-----')
print(df.loc['b':'c', 'Math':])
print('-----')
```

	Math	Physics
a	72	14
b	45	54
c	15	19

	Math	Physics	French	Chemistry
b	45	54	34	63
c	15	19	35	49

Filtering

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(4, 5)
rows_names = ['a', 'b', 'c', 'd', 'e']
w, x, y, z = [dict(zip(rows_names, row)) for row in data]

df = pd.DataFrame({'Math':w, 'Physics':x, 'French':y, 'Chemistry':z})

print(df.loc[df.Math > 50])
print('-----')
print(df.loc[df.Math > 50, ['Math', 'French']])
print('-----')
print(df.loc[(df.Math > 50) & (df.French > 65)])
print('-----')
```

	Math	Physics	French	Chemistry
c	56	62	88	32
d	89	70	44	32
e	57	79	61	17

	Math	French
c	56	88
d	89	44
e	57	61

	Math	Physics	French	Chemistry
c	56	62	88	32

Retrieve with Sorting Values

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(5, 4)
rows_names = ['a', 'b', 'c', 'd', 'e']
columns_names = ['Math', 'Phyiscis', 'French', 'Chemistry']
cols = [dict(zip(columns_names, row)) for row in data]
table = dict(zip(rows_names, cols))

grades = pd.DataFrame(table).T
print(data)
print('-----')
print(grades)
print('-----')
print(grades.sort_values(['Math'], ascending=True))
print('-----')
print(grades.sort_values(['French'], ascending=False))
print('-----')
```

```
[[33 23 45 62]
 [66 58 23 78]
 [92 58 83  9]
 [37 33 90 97]
 [41 75  8 75]]
```

```
-----
      Math  Phyiscis  French  Chemistry
a      33         23      45         62
b      66         58      23         78
c      92         58      83          9
d      37         33      90         97
e      41         75       8         75
```

```
-----
      Math  Phyiscis  French  Chemistry
a      33         23      45         62
d      37         33      90         97
e      41         75       8         75
b      66         58      23         78
c      92         58      83          9
```

```
-----
      Math  Phyiscis  French  Chemistry
d      37         33      90         97
c      92         58      83          9
a      33         23      45         62
b      66         58      23         78
e      41         75       8         75
-----
```

1.3.1 Statistics

- `max()`
- `min()`
- `mean()`
- `std()`

```
[ ]: import numpy as np
import pandas as pd

arr = np.arange(1, 21).reshape(5, 4)
rows_names = ['a', 'b', 'c', 'd', 'e']
columns_names = ['Math', 'Phyiscis', 'French', 'Chemistry']
cols = [dict(zip(columns_names, row)) for row in arr]
table = dict(zip(rows_names, cols))

grades = pd.DataFrame(table).T

print(grades)
print('-----')
print(grades.max())
print('-----')
print(grades.min())
print('-----')
print(grades.mean())
print('-----')
print(grades.std())
print('-----')
```

	Math	Phyiscis	French	Chemistry
a	1	2	3	4
b	5	6	7	8
c	9	10	11	12
d	13	14	15	16
e	17	18	19	20

Math	17
Phyiscis	18
French	19
Chemistry	20

dtype: int64

Math	1
Phyiscis	2
French	3
Chemistry	4

dtype: int64

```
Math          9.0
Physcis       10.0
French        11.0
Chemistry     12.0
dtype: float64
```

```
-----
Math          6.324555
Physcis       6.324555
French        6.324555
Chemistry     6.324555
dtype: float64
-----
```

```
[ ]: import numpy as np
import pandas as pd

arr = np.arange(1, 21).reshape(5, 4)
rows_names = ['a', 'b', 'c', 'd', 'e']
columns_names = ['Math', 'Physcis', 'French', 'Chemistry']

grades = pd.DataFrame(arr, index=rows_names, columns=columns_names)

print(grades)
print('-----')
print(grades.Math.max())
print('-----')
print(grades.Physcis.std())
print('-----')
print(grades.French.mean())
print('-----')
```

```
      Math  Physcis  French  Chemistry
a         1         2         3         4
b         5         6         7         8
c         9        10        11        12
d        13        14        15        16
e        17        18        19        20
```

```
-----
17
-----
```

```
6.324555320336759
-----
```

```
11.0
-----
```

```
[ ]: import numpy as np
import pandas as pd
```

```

data = np.random.randint(1, 100, size=20).reshape(10, 2)

df = pd.DataFrame(data, columns=['A', 'B'])

print(df)
print('-----')
print(df.sum())
print('-----')
print(df.prod())
print('-----')
print(df.mean())
print('-----')
print(df['A'].sum())
print('-----')
print(df['B'].mean())
print('-----')
print(df.mean(axis='columns'))
print('-----')
print(df.sum(axis='index'))
print('-----')

```

```

   A  B
0  98 34
1  97 44
2  43 22
3  98 30
4  39 26
5  83 28
6  51 49
7  94 14
8  94 99
9   4 82

```

```

-----
A      701
B      428
dtype: int64

```

```

-----
A      233733474638943552
B      4002949130019840
dtype: int64

```

```

-----
A      70.1
B      42.8
dtype: float64

```

```

-----
701

```

42.8

0 66.0
1 70.5
2 32.5
3 64.0
4 32.5
5 55.5
6 50.0
7 54.0
8 96.5
9 43.0
dtype: float64

A 701
B 428
dtype: int64

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(10, 2)

df = pd.DataFrame(data, columns=['A', 'B'])

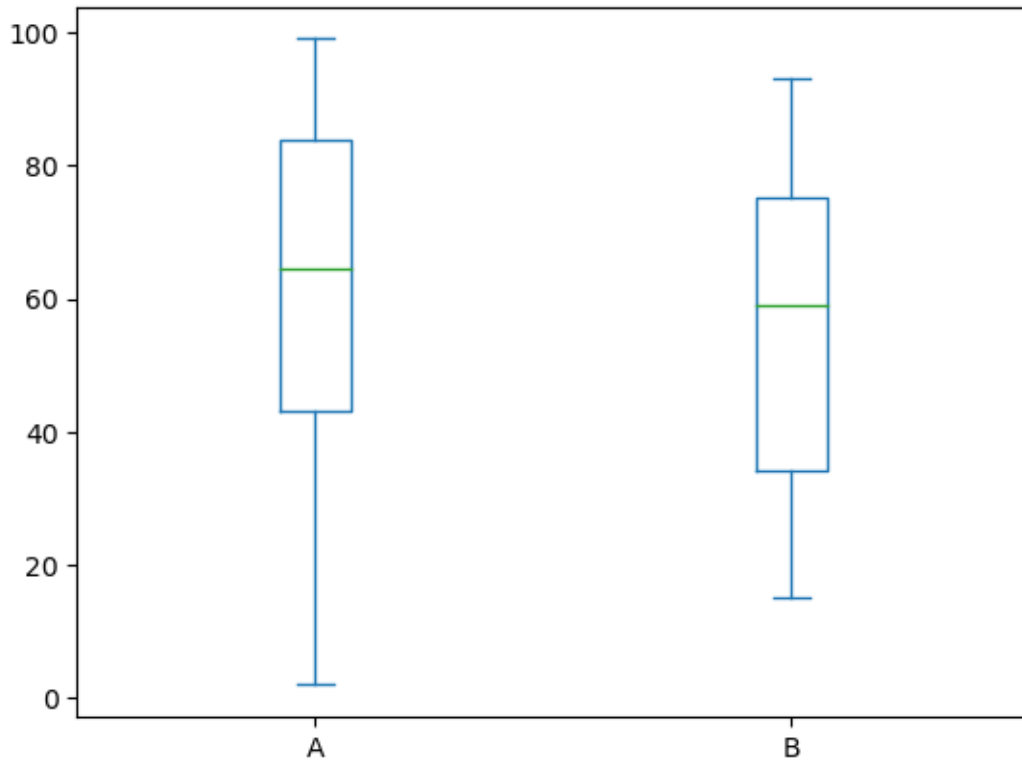
print(df)
print('-----')
print(df.describe())
print('-----')
df.plot(kind='box')
```

 A B
0 92 93
1 2 18
2 11 78
3 80 30
4 56 89
5 73 67
6 85 65
7 99 15
8 47 53
9 42 47

	A	B
count	10.000000	10.000000
mean	58.700000	55.500000

std	33.326833	27.953334
min	2.000000	15.000000
25%	43.250000	34.250000
50%	64.500000	59.000000
75%	83.750000	75.250000
max	99.000000	93.000000

```
[ ]: <AxesSubplot:>
```



Correlation

used to find the pairwise correlation of all columns in the Pandas Dataframe in Python.

```
[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame(np.random.rand(5, 3), columns=['A', 'B', 'C'])

print(df)
print('-----')
print(df.corr())
print('-----')
```

	A	B	C
0	0.413500	0.874991	0.762693
1	0.776048	0.690615	0.224443
2	0.126178	0.118267	0.159303
3	0.023866	0.975316	0.073367
4	0.250222	0.891413	0.992557

	A	B	C
A	1.000000	0.073569	0.116899
B	0.073569	1.000000	0.399207
C	0.116899	0.399207	1.000000

Skew

The `skew()` method calculates the skew for each column.

```
[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame(np.random.rand(5, 3), columns=['A', 'B', 'C'])

print(df)
print('-----')
print(df.skew())
print('-----')
```

	A	B	C
0	0.404252	0.314501	0.282954
1	0.662645	0.553560	0.783029
2	0.193921	0.003088	0.039726
3	0.677650	0.766873	0.721927
4	0.578256	0.734138	0.786718

A	-1.018482
B	-0.844611
C	-0.861548

dtype: float64

1.3.2 Operations on DataFrames

dataframe with list comprehension

```
[ ]: import pandas as pd

data = [{'square': i**2} for i in range(10)]
df = pd.DataFrame(data)
```

```
print(df)
print('-----')
```

	square
0	0
1	1
2	4
3	9
4	16
5	25
6	36
7	49
8	64
9	81

```
[ ]: import pandas as pd

data = [{'square': i**2, 'cube': i**3, 'root': i**0.5} for i in range(10)]

df = pd.DataFrame(data)

print(df)
print('-----')
```

	square	cube	root
0	0	0	0.000000
1	1	1	1.000000
2	4	8	1.414214
3	9	27	1.732051
4	16	64	2.000000
5	25	125	2.236068
6	36	216	2.449490
7	49	343	2.645751
8	64	512	2.828427
9	81	729	3.000000

NaN

NaN stands for Not A Number and is one of the common ways to represent the missing value in the data

```
[ ]: import pandas as pd

df = pd.DataFrame([{'a': 1, 'b': 2}, {'b': 3, 'c': 4}, {'c': 5, 'd': 6}])

print(df)
```

	a	b	c	d
0	1.0	2.0	NaN	NaN
1	NaN	3.0	4.0	NaN
2	NaN	NaN	5.0	6.0

Create New Column

```
[ ]: import numpy as np
import pandas as pd

data = np.random.randint(1, 100, size=20).reshape(5, 4)

grades = pd.DataFrame(data, index=['a', 'b', 'c', 'd', 'e'], columns=['Math', 'Physics', 'French', 'Chemistry'])

print(grades)
print('-----')

grades['Total'] = grades.sum(axis=1) / 4
print(grades)
print('-----')
print(grades.loc[grades.Total > 80])
print('-----')
```

	Math	Physics	French	Chemistry
a	52	12	6	72
b	51	1	33	1
c	44	98	18	39
d	52	84	94	84
e	47	66	80	94

	Math	Physics	French	Chemistry	Total
a	52	12	6	72	35.50
b	51	1	33	1	21.50
c	44	98	18	39	49.75
d	52	84	94	84	78.50
e	47	66	80	94	71.75

```
Empty DataFrame
Columns: [Math, Physics, French, Chemistry, Total]
Index: []
```

```
[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame(np.random.randint(1, 100, size=15).reshape(5, 3),
                  columns=['A', 'B', 'C'])
```

```

result = (df['A'] + df['B']) / (df['C'] - 1)

print(df)
print('-----')
print(result)
print('-----')

```

```

      A   B   C
0     9  67  24
1    10  56  74
2    48  97  76
3    74  97  40
4    93  82  28

```

```

-----
0     3.304348
1     0.904110
2     1.933333
3     4.384615
4     6.481481
dtype: float64
-----

```

eval()

function evaluates a string describing operations on DataFrame columns. Operates on columns only, not specific rows or elements. This allows eval to run arbitrary code, which can make you vulnerable to code injection if you pass user input to this function.

```

[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame(np.random.randint(1, 100, size=15).reshape(5, 3),
                  columns=['A', 'B', 'C'])
result = pd.eval('(df.A + df.B) / (df.C - 1)')

print(df)
print('-----')
print(result)
print('-----')

```

```

      A   B   C
0    82  65  62
1     8  91  64
2     2  99  62
3    46  11  47
4    83  17  16

```

```

-----
0     2.409836

```



```

1    1.571429
2    1.655738
3    1.239130
4    6.666667
dtype: float64
-----

```

query()

method is used to query the rows based on the expression (single or multiple column conditions) provided and returns a new DataFrame.

```

[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame(np.random.rand(5, 3), columns=['A', 'B', 'C'])

print(df)
print('-----')
print(df.query('A > 0.5 and B < 0.5'))
print('-----')

```

	A	B	C
0	0.071751	0.813642	0.710513
1	0.289516	0.714604	0.396607
2	0.247046	0.033635	0.295897
3	0.798599	0.565521	0.812307
4	0.631097	0.136117	0.962577

	A	B	C
4	0.631097	0.136117	0.962577

```

[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame(np.random.rand(5, 3), columns=['A', 'B', 'C'])

tmp1 = df.A < 0.5
tmp2 = df.C > 0.5
tmp3 = tmp1 & tmp2

print(df)
print('-----')
print(tmp3)
print('-----')
print(df[tmp3])
print('-----')

```

	A	B	C
0	0.055417	0.545461	0.437569
1	0.608685	0.978338	0.701264
2	0.178670	0.764195	0.635076
3	0.338034	0.344962	0.094271
4	0.091445	0.367199	0.223734

```

0    False
1    False
2     True
3    False
4    False
dtype: bool

```

	A	B	C
2	0.17867	0.764195	0.635076

```
[ ]: import numpy as np
import pandas as pd

def make_df(rows, cols):
    data = {col: [str(col)+str(row) for row in rows] for col in cols}
    return pd.DataFrame(data, rows)

rows = range(3)
cols = 'ABC'
df = make_df(rows, cols)
print(df)
```

	A	B	C
0	A0	B0	C0
1	A1	B1	C1
2	A2	B2	C2

merge()

- function is used to merge two DataFrame objects with a database-style join operation.
- The joining is performed on columns or indexes.
- If the joining is done on columns, indexes are ignored.
- This function returns a new DataFrame and the source DataFrame objects are unchanged.

```
[ ]: import pandas as pd

df1 = pd.DataFrame({'employee': ['Bob', 'Jake', 'Lisa', 'Sue'],
                    'group': ['Accounting', 'Engineering', 'Engineering',
                              'HR']})
```

```

df2 = pd.DataFrame({'employee': ['Lisa', 'Bob', 'Jake', 'Sue'],
                    'hire_date': [2004, 2008, 2012, 2014]})

df3 = pd.merge(df1, df2)
print(df3)
print('-----')

df4 = pd.DataFrame({'group': ['Accountin', 'Engineering', 'HR'],
                    'supervisor': ['Carly', 'Guido', 'Steve']})

df5 = pd.merge(df3, df4)
print(df5)
print('-----')

```

	employee	group	hire_date
0	Bob	Accounting	2008
1	Jake	Engineering	2012
2	Lisa	Engineering	2004
3	Sue	HR	2014

	employee	group	hire_date	supervisor
0	Jake	Engineering	2012	Guido
1	Lisa	Engineering	2004	Guido
2	Sue	HR	2014	Steve

Merge Options - merge has some options by passing value to `how` arg - `'inner'`: just merge intersection between them *default option* - `'outer'`: merge 2 dataframes and create new one that contains all itmes - `'right'`: merge 2 dataframes and create new one that contains all itmes of 1st dataframe and only intersected items from second - `'left'`: merge 2 dataframes and create new one that contains all itmes of 2nd dataframe and only intersected items from first

```

[ ]: import pandas as pd

food_orders = pd.DataFrame({
    'name': ['Peter', 'Paul', 'Mary'],
    'food': ['fish', 'beans', 'bread']
})

drink_orders = pd.DataFrame({
    'name': ['Mary', 'Joseph'],
    'drink': ['cola', 'orange juice']
})

print(food_orders)
print('-----')
print(drink_orders)
print('-----')

```

```

print(pd.merge(food_orders, drink_orders))
print('-----')
print(pd.merge(food_orders, drink_orders, how='inner'))
print('-----')
print(pd.merge(food_orders, drink_orders, how='outer'))
print('-----')
print(pd.merge(food_orders, drink_orders, how='right'))
print('-----')
print(pd.merge(food_orders, drink_orders, how='left'))
print('-----')

```

```

      name  food
0  Peter  fish
1   Paul  beans
2   Mary  bread

```

```

-----
      name      drink
0   Mary      cola
1  Joseph  orange juice

```

```

-----
      name  food drink
0  Mary  bread  cola

```

```

-----
      name  food drink
0  Mary  bread  cola

```

```

-----
      name  food      drink
0  Peter  fish      NaN
1   Paul  beans      NaN
2   Mary  bread     cola
3  Joseph   NaN  orange juice

```

```

-----
      name  food      drink
0   Mary  bread     cola
1  Joseph   NaN  orange juice

```

```

-----
      name  food drink
0  Peter  fish  NaN
1   Paul  beans  NaN
2   Mary  bread  cola

```

drop()

method removes the specified row or column. By specifying the column axis (`axis='columns'`), the `drop()` method removes the specified column. By specifying the row axis (`axis='index'`), the `drop()` method removes the specified row.

```
[ ]: import pandas as pd

df1 = pd.DataFrame({'employee': ['Bob', 'Jake', 'Lisa', 'Sue'],
                    'group': ['Accounting', 'Engineering', 'Engineering', 'HR']})

df2 = pd.DataFrame({'name': ['Bob', 'Jake', 'Lisa', 'Sue'],
                    'salary': [70000, 80000, 120000, 90000]})

print(df1)
print('-----')
print(df2)
print('-----')

df3 = pd.merge(df1, df2, left_on='employee', right_on='name')
print(df3)
print('-----')
print(df3.drop('name', axis=1))
print('-----')
```

	employee	group
0	Bob	Accounting
1	Jake	Engineering
2	Lisa	Engineering
3	Sue	HR

	name	salary
0	Bob	70000
1	Jake	80000
2	Lisa	120000
3	Sue	90000

	employee	group	name	salary
0	Bob	Accounting	Bob	70000
1	Jake	Engineering	Jake	80000
2	Lisa	Engineering	Lisa	120000
3	Sue	HR	Sue	90000

	employee	group	salary
0	Bob	Accounting	70000
1	Jake	Engineering	80000
2	Lisa	Engineering	120000
3	Sue	HR	90000

`set_index()`

method allows one or more column values become the row index.

```
[ ]: import pandas as pd

data = {
    'name': ['Mark', 'John', 'Lisa', 'Monica'],
    'age': [50, 40, 30, 40],
    'qualified': [True, False, False, True]
}

df = pd.DataFrame(data)

print(df)
print('-----')
new_df = df.set_index('name')
print(new_df)
print('-----')
```

	name	age	qualified
0	Mark	50	True
1	John	40	False
2	Lisa	30	False
3	Monica	40	True

	age	qualified
name		
Mark	50	True
John	40	False
Lisa	30	False
Monica	40	True

groupby()

- is used for grouping the data according to the categories and apply a function to the categories.
- It also helps to aggregate data efficiently. Pandas `dataframe.groupby()` function is used to split the data into groups based on some criteria.

```
[ ]: import pandas as pd

data = {
    'key': ['A', 'B', 'C', 'A', 'B', 'C'],
    'data': range(6)
}

df = pd.DataFrame(data)

print(df)
print('-----')
gdf = df.groupby('key')
```

```

print(gdf.first())           #get first value of each group
print('-----')
print(gdf.sum())             #sum of each group
print('-----')
print(gdf.describe())
print('-----')
print(gdf.describe().unstack())
print('-----')

```

```

key  data
0   A    0
1   B    1
2   C    2
3   A    3
4   B    4
5   C    5

```

```

-----
      data
key
A      0
B      1
C      2

```

```

-----
      data
key
A      3
B      5
C      7

```

```

-----
      data
count mean      std  min   25%  50%   75%  max
key
A      2.0  1.5  2.12132  0.0  0.75  1.5  2.25  3.0
B      2.0  2.5  2.12132  1.0  1.75  2.5  3.25  4.0
C      2.0  3.5  2.12132  2.0  2.75  3.5  4.25  5.0

```

```

-----
      data      key
count  A      2.00000
      B      2.00000
      C      2.00000
mean   A      1.50000
      B      2.50000
      C      3.50000
std    A      2.12132
      B      2.12132
      C      2.12132
min    A      0.00000

```

	B	1.00000
	C	2.00000
25%	A	0.75000
	B	1.75000
	C	2.75000
50%	A	1.50000
	B	2.50000
	C	3.50000
75%	A	2.25000
	B	3.25000
	C	4.25000
max	A	3.00000
	B	4.00000
	C	5.00000

dtype: float64

aggregate()

method allows you to apply a function or a list of function names to be executed along one of the axis of the DataFrame

```
[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame({
    'key': ['A', 'B', 'C', 'A', 'B', 'C'],
    'data1': range(6),
    'data2': np.random.randint(0, 10, size=6)
})

print(df)
print('-----')
new_df = df.groupby('key').aggregate({'data1': 'min', 'data2': 'max'})
print(new_df)
```

	key	data1	data2
0	A	0	0
1	B	1	9
2	C	2	2
3	A	3	9
4	B	4	4
5	C	5	2

	key	data1	data2
A		0	9
B		1	9
C		2	2

filter()

- function filters the DataFrame for rows and columns.
- The returned DataFrame contains only rows and columns that are specified with the function.
- It doesn't update the existing DataFrame instead it always returns a new one.

```
[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame({
    'key': ['A', 'B', 'C', 'A', 'B', 'C'],
    'data1': range(6),
    'data2': np.random.randint(0, 10, size=6)
})

print(df)
print('-----')

def filter_func(x):
    return x['data2'].std() > 4

print(df.groupby('key').filter(filter_func))
print('-----')
```

	key	data1	data2
0	A	0	5
1	B	1	9
2	C	2	7
3	A	3	5
4	B	4	1
5	C	5	8

	key	data1	data2
1	B	1	9
4	B	4	1

transform()

function call func on self producing a DataFrame with transformed values and that has the same axis length as self.

```
[ ]: import numpy as np
import pandas as pd

df = pd.DataFrame({
    'key': ['A', 'B', 'C', 'A', 'B', 'C'],
    'data1': range(6),
    'data2': np.random.randint(0, 10, size=6)
})
```

```

})

print(df)
print('-----')
new_df = df.groupby('key').transform(lambda x: x**2)
print(new_df)
print('-----')

```

	key	data1	data2
0	A	0	3
1	B	1	8
2	C	2	5
3	A	3	4
4	B	4	4
5	C	5	8

```

-----

```

	data1	data2
0	0	9
1	1	64
2	4	25
3	9	16
4	16	16
5	25	64

```

-----

```

contact me [Mahmoud Gadallah](#)