



iscte

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UNIVERSITÁRIO
DE LISBOA



emprego
digital



```
class BigFile:
```

```
    def __init__(self, datadir, ndims):  
        idfile = os.path.join(datadir, "id.txt")  
        self.names = [x.strip() for x in str.split(open(idfile).read()) if x.strip()]  
        self.name2index = dict(zip(self.names, range(len(self.names))))  
        self.ndims = ndims  
        self.featurefile = os.path.join(datadir, "feature.bin")  
        print "[BigFile] %d features, %d dimensions" % (len(self.names), self.ndims)  
        print "        binary: %s" % self.featurefile  
        print "        txt: %s" % idfile
```

```
    def read(self, requested, isname=True):  
        if isname:  
            index_name_array = [(self.name2index[x], x) for x in requested if x in self.names]  
        else:  
            assert len(requested) > 0  
            assert all((requested[i] in self.names) for i in range(len(requested)))  
            index_name_array = [(x, self.names[x]) for x in requested]  
            index_name_array.sort()  
            vecs = seq_read(self.featurefile, self.ndims, [x[0] for x in index_name_array])  
            return [x[1] for x in index_name_array], vecs  
  
    def shape(self):  
        return [len(self.names), self.ndims]
```



pythonTM

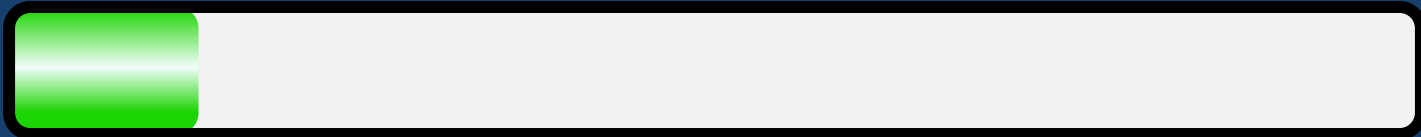
1.

Overall Program Content

Web development with Python	Hours
Work skills development	50
Python Programming Introduction	150
Web Programming Introduction (html/css)	100
Databases Concepts and Structures	50
Web Servers Programming	150
Web services development	150
Total	650

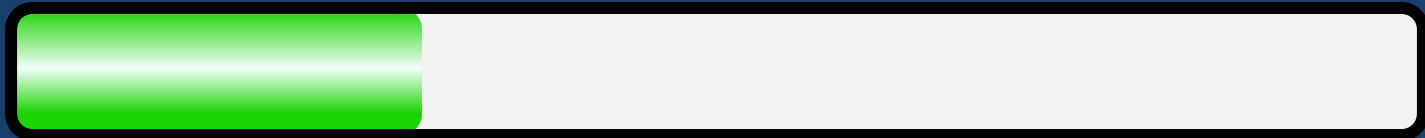
Python programming Introduction Content

1. Course Introduction
 - Why Python?
 - Python Applications
 - Installation Tools
 - Building your code catalog
 - Useful websites



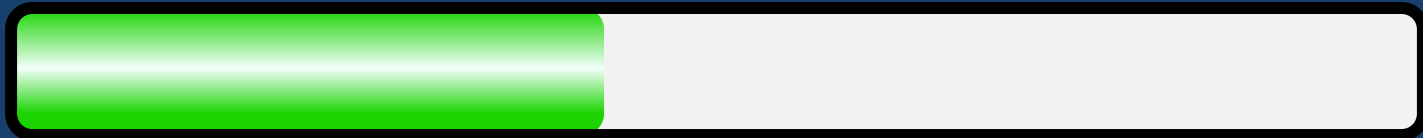
Python programming Introduction Content

2. Data types/outputs/inputs
3. Operators
4. Functions and Modules



Python programming Introduction Content

- 5. Conditional statements and expression
- 6. Loops
- 7. Work with standard Library and Modules



Python programming Introduction Content

- 8. Data structure in python
- 9. List,
- 10. Tuple,
- 11. Dictionaries,
- 12. Set



Python programming Introduction Content

- 13. Files
- 14. Functions and Modules
- 15. Classes
- 16. Introduction to Numpy
- 17. Introduction to Pandas



Python programming Introduction Content

- 18. Introduction to matplotlib for data visualization
- 19. Data Preprocessing

100% Loaded

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```
class BigFile:
```

```
    def __init__(self, datadir, ndims):
        idfile = os.path.join(datadir, "id.txt")
        self.names = [x.strip() for x in str.split(open(idfile).read()) if x.strip()]
        self.name2index = dict(zip(self.names, range(len(self.names))))
        self.ndims = ndims
        self.featurefile = os.path.join(datadir, "feature.bin")
        print "[BigFile] %d features, %d dimensions" % (len(self.names), self.ndims)
        print "        binary: %s" % self.featurefile
        print "        txt: %s" % idfile
```

```
    def read(self, requested, isname=True):
        if isname:
            index_name_array = [self.name2index[x], x] for x in requested if x in self.names
        else:
            assert(min(requested) >= 0)
            assert(max(requested) < len(self.names))
            index_name_array = [(x, self.names[x]) for x in requested]
            index_name_array.sort()
            vecs = seq_read(self.featurefile, self.ndims, [x[0] for x in index_name_array])
            return [x[1] for x in index_name_array], vecs

    def shape(self):
        return [len(self.names), self.ndims]
```

<Let's get started >

Contents

1. Statistics

2. Data Visualization

Statistics

Mean / Average

Arithmetic Mean (AM)

Geometric Mean (GM)

Harmonic Mean (HM)

$$AM(x_1, \dots, x_n) = \frac{1}{n} (x_1 + \dots + x_n)$$

$$GM(x_1, \dots, x_n) = \sqrt[n]{|x_1 \times \dots \times x_n|}$$

$$HM(x_1, \dots, x_n) = \frac{n}{\frac{1}{x_1} + \dots + \frac{1}{x_n}}$$

$$HM\left(\frac{1}{x_1}, \dots, \frac{1}{x_n}\right) = \frac{1}{AM(x_1, \dots, x_n)}$$

$$GM\left(\frac{1}{x_1}, \dots, \frac{1}{x_n}\right) = \frac{1}{GM(x_1, \dots, x_n)}$$

Weighted arithmetic mean

$$\bar{x} = \frac{\sum_{i=1}^n w_i x_i}{\sum_{i=1}^n w_i}$$

$$\min \leq HM \leq GM \leq AM \leq \max$$

Mean Example

For these values 4, 36, 45, 50, 75 calculate mean?

Arithmetic Mean (AM)

$$\bar{x} = \frac{1}{n} \left(\sum_{i=1}^n x_i \right) = \frac{x_1 + x_2 + \cdots + x_n}{n}$$

$$\frac{4 + 36 + 45 + 50 + 75}{5} = \frac{210}{5} = 42$$

Geometric Mean (GM)

$$\bar{x} = \left(\prod_{i=1}^n x_i \right)^{\frac{1}{n}} = (x_1 x_2 \cdots x_n)^{\frac{1}{n}}$$

$$(4 \times 36 \times 45 \times 50 \times 75)^{\frac{1}{5}} = \sqrt[5]{24\,300\,000} = 30.$$

Harmonic Mean (HM)

$$\bar{x} = n \left(\sum_{i=1}^n \frac{1}{x_i} \right)^{-1}$$

$$\frac{5}{\frac{1}{4} + \frac{1}{36} + \frac{1}{45} + \frac{1}{50} + \frac{1}{75}} = \frac{5}{\frac{1}{3}} = 15.$$

Mean Median Mode

*First Sort the values
Then find median!*

Comparison of common **averages** of values { 1, 2, 2, 3, 4, 7, 9 }

Type	Description	Example	Result
Arithmetic mean	Sum of values of a data set divided by number of values	$(1+2+2+3+4+7+9) / 7$	4
Median	Middle value separating the greater and lesser halves of a data set	1, 2, 2, 3, 4, 7, 9	3
Mode	Most frequent value in a data set	1, 2, 2, 3, 4, 7, 9	2

Mean

Median

Mode

Example

```
import statistics
from scipy import stats
import numpy as np

a = [4, 36, 45, 50, 75]
b = [1, 2, 2, 3, 4, 7, 9]
c = [6, 3, 9, 6, 6, 5, 9, 9, 3, 1]
```

```
print(statistics.mean(a))
print(np.mean(b))
print(np.mean(c))
```

42
4.0
5.7

```
print(statistics.mode(a))
print(stats.mode(b))
print(stats.mode(c))
```

4
ModeResult(mode=array([2]), count=array([2]))
ModeResult(mode=array([6]), count=array([3]))

```
print(statistics.median(a))
print(statistics.median(b))
print(np.median(c))
```

45
3
6.0

Variance

Standard deviation

Standard deviation (SD)

The measure of the amount of variation or dispersion of a set of values. A low standard deviation indicates that the values tend to be close to the mean of the set, while a high standard deviation indicates that the values are spread out over a wider range.

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}$$

σ , sigma for the **population** standard deviation

μ , the population mean

$$\mu = \frac{1}{N} \sum_{i=1}^N x_i$$

$$s = \sqrt{\frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2}$$

s , the **sample** standard deviation

\bar{x} , a sample mean

Variance (Var)

it measures how far a set of numbers is spread out from their average value.
The average of the **squared** differences from the Mean

$$\sigma^2, s^2, \text{Var}(X)$$

$$\text{StandardDeviation} = \sqrt{\text{variance}}$$

Why Take a Sample?

Mostly because it is easier and cheaper.

Imagine you want to know what the whole country thinks ... you can't ask millions of people, so instead you ask maybe 1,000 people.

To find out information about the population (such as mean and standard deviation), we do not need to look at all members of the population; we only need a sample.

But when we take a sample, we lose some accuracy.

Variance SD Population Example

Population standard deviation example



Example: Sam has 20 Rose Bushes.

The number of flowers on each bush is

9, 2, 5, 4, 12, 7, 8, 11, 9, 3, 7, 4, 12, 5, 4, 10, 9, 6, 9, 4

Work out the Standard Deviation.

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}$$

The handy [Sigma Notation](#) says to sum up as many terms as we want:

start at this value
go to this value

$$\sum_{n=1}^4 n = 1+2+3+4 = 10$$

what to sum

Sigma Notation

$$\mu = \frac{1}{N} \sum_{i=1}^N x_i$$

Example: 9, 2, 5, 4, 12, 7, 8, 11, 9, 3, 7, 4, 12, 5, 4, 10, 9, 6, 9, 4

The mean is:

$$\frac{9+2+5+4+12+7+8+11+9+3+7+4+12+5+4+10+9+6+9+4}{20} = \frac{140}{20} = 7$$

So:

$$\mu = 7$$

Variance SD Population Example

Example (continued):

$$(9 - 7)^2 = (2)^2 = 4$$

$$(2 - 7)^2 = (-5)^2 = 25$$

$$(5 - 7)^2 = (-2)^2 = 4$$

$$(4 - 7)^2 = (-3)^2 = 9$$

$$(12 - 7)^2 = (5)^2 = 25$$

$$(7 - 7)^2 = (0)^2 = 0$$

$$(8 - 7)^2 = (1)^2 = 1$$

... etc ...

And we get these results:

4, 25, 4, 9, 25, 0, 1, 16, 4, 16, 0, 9, 25, 4, 9, 4, 1, 4, 9

Example (continued):

$$\sum_{i=1}^N (x_i - \mu)^2$$

Which means: Sum all values from $(x_1 - 7)^2$ to $(x_N - 7)^2$

We already calculated $(x_1 - 7)^2 = 4$ etc. in the previous step, so just sum them up:

$$= 4 + 25 + 4 + 9 + 25 + 0 + 1 + 16 + 4 + 16 + 0 + 9 + 25 + 4 + 9 + 4 + 1 + 4 + 9 = \mathbf{178}$$

Example (concluded):

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}$$

$$\sigma = \sqrt{(8.9)} = \mathbf{2.983...}$$

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}$$

Variance SD Sample Example



Example: Sam has **20** rose bushes, but only counted the flowers on **6 of them!**

The "population" is all 20 rose bushes,
and the "sample" is the 6 bushes that Sam counted the flowers of.

Let us say Sam's flower counts are:

9, 2, 5, 4, 12, 7

$$s = \sqrt{\frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2}$$

The important change is "N-1" instead of "N"

The symbols also change to reflect that we are working on a sample instead of the whole population:

- The mean is now \bar{x} (for sample mean) instead of μ (the population mean),
- And the answer is **s** (for Sample Standard Deviation) instead of σ .

But that does not affect the calculations. **Only N-1 instead of N changes the calculations.**

Variance SD Sample Example

Sample Standard Deviation example

Example 2: Using sampled values 9, 2, 5, 4, 12, 7

The mean is $(9+2+5+4+12+7) / 6 = 39/6 = 6.5$

So:

$$\bar{x} = 6.5$$

$$s = \sqrt{\frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2}$$

Example 2 (continued):

$$(9 - 6.5)^2 = (2.5)^2 = 6.25$$

$$(2 - 6.5)^2 = (-4.5)^2 = 20.25$$

$$(5 - 6.5)^2 = (-1.5)^2 = 2.25$$

$$(4 - 6.5)^2 = (-2.5)^2 = 6.25$$

$$(12 - 6.5)^2 = (5.5)^2 = 30.25$$

$$(7 - 6.5)^2 = (0.5)^2 = 0.25$$

Example 2 (concluded):

$$s = \sqrt{\frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2}$$

$$s = \sqrt{(13.1)} = 3.619...$$

Variance

SD

Statistics vs NumPy

```
import statistics
import numpy as np
```

```
a = [9, 2, 5, 4, 12, 7, 8, 11, 9, 3, 7, 4, 12, 5, 4, 10, 9, 6, 9, 4]
b = [9, 2, 5, 4, 12, 7]
```

Entire Population

a Sample

```
print(np.mean(a))
print(np.mean(b))
```

```
7.0
6.5
```

```
print(np.std(a))           # standard deviation of an entire population
print(statistics.pstdev(a)) # standard deviation of an entire population
print(statistics.stdev(a))
```

```
2.9832867780352594
2.9832867780352594
3.0607876523260447
```

Wrong answers: stdev(a) , std(b)

```
print(np.std(b))
print(statistics.stdev(b)) # standard deviation of a Sample
```

```
3.304037933599835
3.6193922141707713
```

```
print(np.var(a))           # Variance of an entire population
print(statistics.pvariance(a)) # Variance of an entire population
print(statistics.variance(a))
```

```
8.9
8.9
9.368421052631579
```

Wrong answers: variance(a) , var(b)

```
print(np.var(b))
print(statistics.variance(b)) # Variance of a Sample
```

```
10.916666666666666
13.1
```

for Entire Population
`numpy.std(a)`
`statistics.pstdev(a)`
`numpy.var(a)`
`statistics.pvariance(a)`

for a Sample
`statistics.stdev(b)`
`statistics.variance(b)`

Quartile

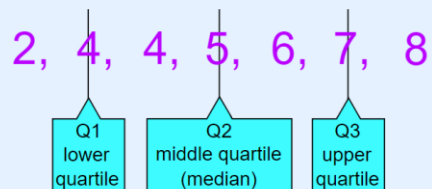
Quartiles are the values that divide a list of numbers into quarters

- ✓ *Put the list of numbers in order*
- ✓ *Then cut the list into four equal parts*
- ✓ *The Quartiles are at the "cuts"*

Example: 5, 7, 4, 4, 6, 2, 8

Put them in order: 2, 4, 4, 5, 6, 7, 8

Cut the list into quarters:



And the result is:

- Quartile 1 (Q1) = 4
- Quartile 2 (Q2), which is also the Median, = 5
- Quartile 3 (Q3) = 7

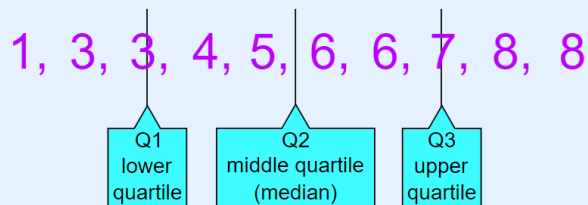
Quartile

*Sometimes a "cut" is between two numbers ...
the Quartile is the average of the two numbers*

Example: 1, 3, 3, 4, 5, 6, 6, 7, 8, 8

The numbers are already in order

Cut the list into quarters:



In this case Quartile 2 is half way between 5 and 6:

$$Q2 = (5+6)/2 = 5.5$$

And the result is:

- Quartile 1 (Q1) = **3**
- Quartile 2 (Q2) = **5.5**
- Quartile 3 (Q3) = **7**

Quartile

5 Methods to calculate Quartiles

when the desired quantile lies between two data points $i < j$

Linear

$$i + (j - i) * \text{fraction}$$

where `fraction` is the fractional part of the index surrounded by i and j

Lower

i

Higher

j

Nearest

i or j

whichever is nearest

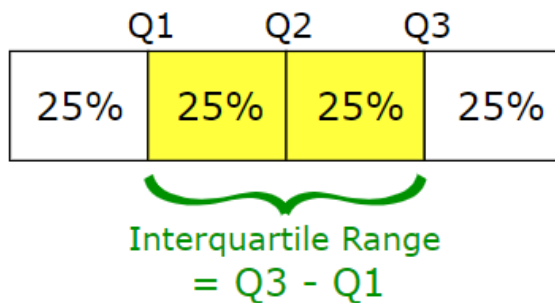
Midpoint

$$(i + j) / 2$$

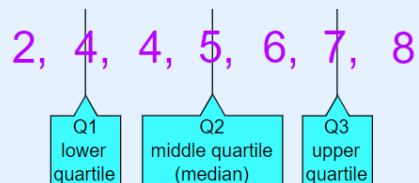
IQR

Interquartile Range

The "Interquartile Range" is from $Q1$ to $Q3$



Example:



The **Interquartile Range** is:

$$Q3 - Q1 = 7 - 4 = 3$$

Data Visualization

Matplotlib

Plot

Simple Example

```
import matplotlib.pyplot as plt
```

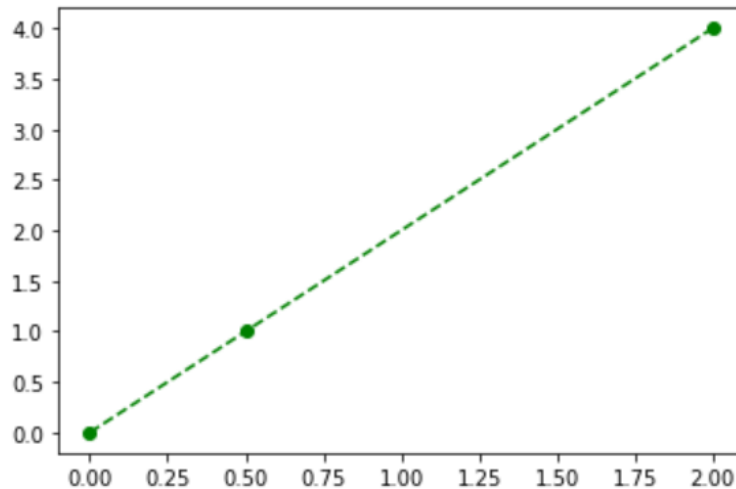
`plt.plot?` *Plot y versus x as lines and/or markers.*

for points A(0, 0) , B(0.5, 1) , C(2, 4)

```
x = [0, 0.5, 2]  
y = [0, 1, 4]
```

```
plt.plot(x, y, 'go--');
```

```
# plt.plot(x, y, color='green', marker='o', linestyle='dashed');
```



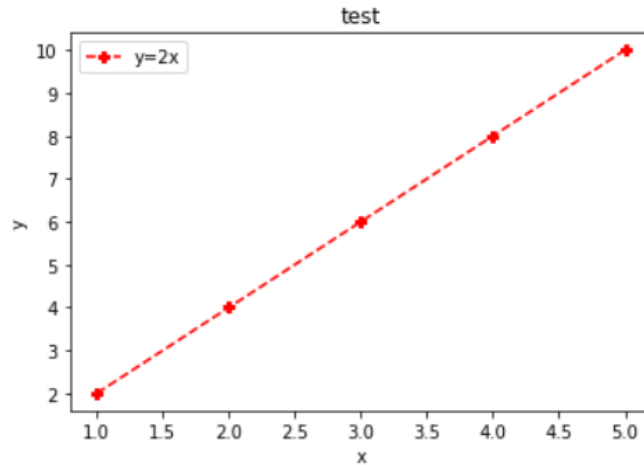
Plot Legend Title xlabel ylabel

```
import matplotlib.pyplot as plt
```

```
x=[1,2,3,4,5]  
y=[2,4,6,8,10]
```

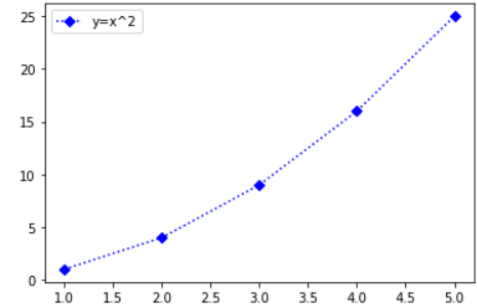
```
plt.plot(x,y,'r--P',label='y=2x')  
plt.legend()  
plt.title('test')  
plt.xlabel('x')  
plt.ylabel('y')
```

```
Text(0, 0.5, 'y')
```

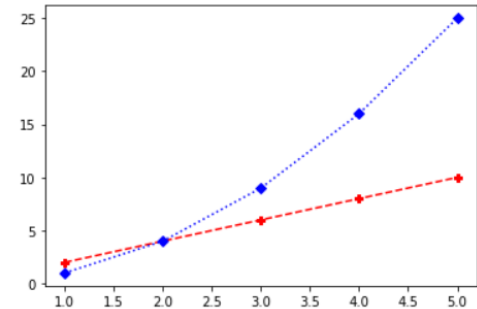


```
x1=[1,2,3,4,5]  
y1=[1,4,9,16,25]
```

```
plt.plot(x1,y1,'b:D' , label='y=x^2');  
plt.legend();
```



```
plt.plot(x,y,'r--P',label='y=2x');  
plt.plot(x1,y1,'b:D' , label='y=x^2');
```



Plot Subplot Legend Savefig

```
import matplotlib.pyplot as plt
```

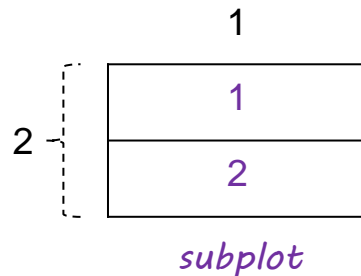
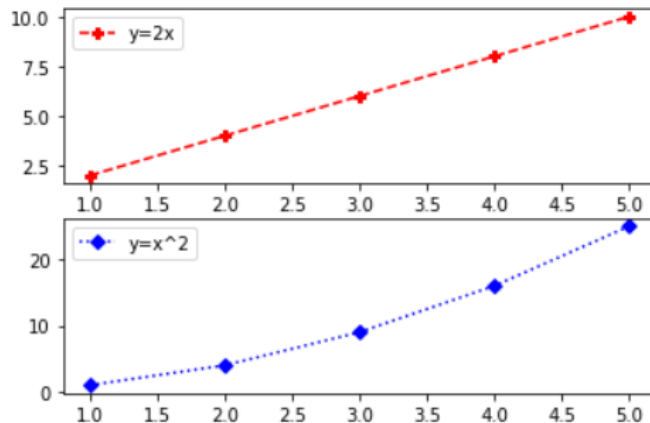
```
x=[1,2,3,4,5]  
y=[2,4,6,8,10]
```

```
x1=[1,2,3,4,5]  
y1=[1,4,9,16,25]
```

```
plt.subplot(211)  
plt.plot(x,y,'r--P',label='y=2x')  
plt.legend()
```

```
plt.subplot(212)  
plt.plot(x1,y1,'b:D' , label='y=x^2')  
plt.legend()
```

```
plt.savefig('D:/plot.png')
```



Plot Step Grid Legend Title

```
import numpy as np
import matplotlib.pyplot as plt
```

```
x = np.arange(14)
y = np.sin(x / 2)
```



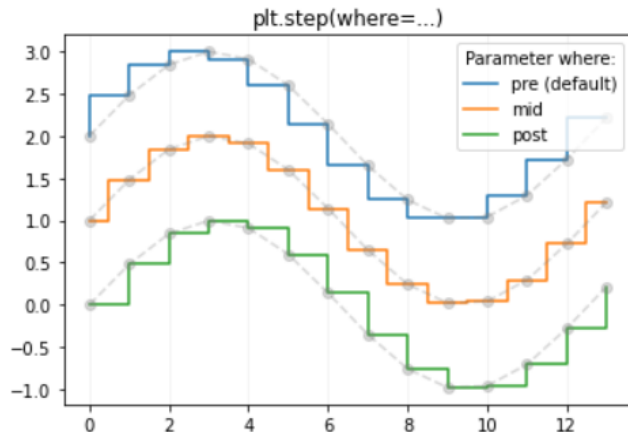
Function

```
plt.step(x, y + 2, label='pre (default)')
plt.plot(x, y + 2, 'o--', color='grey', alpha=0.3)
```

```
plt.step(x, y + 1, where='mid', label='mid')
plt.plot(x, y + 1, 'o--', color='grey', alpha=0.3)
```

```
plt.step(x, y, where='post', label='post')
plt.plot(x, y, 'o--', color='grey', alpha=0.3)
```

```
plt.grid(axis='x', color='0.95')
plt.legend(title='Parameter where:')
plt.title('plt.step(where=...)')
plt.show()
```



Plot

Subplots

Suptitle

Set_xlabel

Set_ylabel

```
import numpy as np
import matplotlib.pyplot as plt

x1 = np.linspace(0.0, 5.0)
x2 = np.linspace(0.0, 2.0)

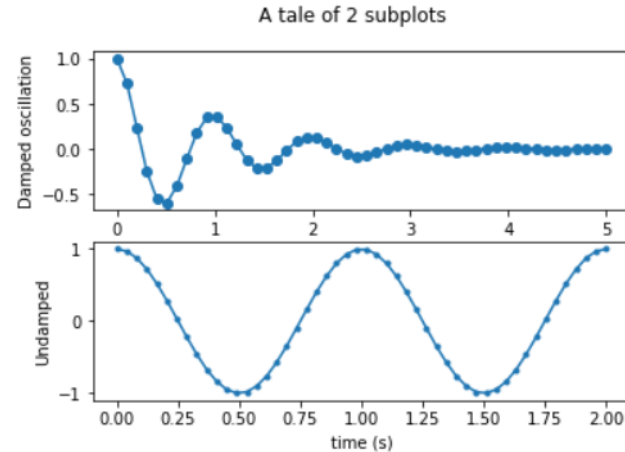
y1 = np.cos(2 * np.pi * x1) * np.exp(-x1)
y2 = np.cos(2 * np.pi * x2)

fig, (ax1, ax2) = plt.subplots(2, 1)
fig.suptitle('A tale of 2 subplots')

ax1.plot(x1, y1, 'o-')
ax1.set_ylabel('Damped oscillation')

ax2.plot(x2, y2, '-.')
ax2.set_xlabel('time (s)')
ax2.set_ylabel('Undamped')

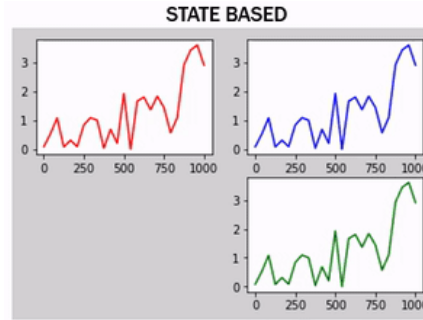
plt.show()
```



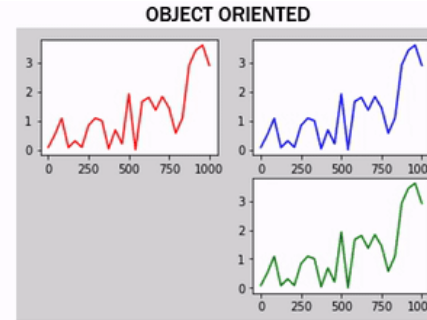
State-based VS Object-oriented

Subplot VS Subplots

State-based vs Object-oriented



```
plt.figure(facecolor='lightgrey')
plt.subplot(2,2,1)
plt.plot(data_x, data_y, 'r-')
plt.subplot(2,2,2)
plt.plot(data_x, data_y, 'b-')
plt.subplot(2,2,4)
plt.plot(data_x, data_y, 'g-')
```



```
fig, ax = plt.subplots(2,2)
fig.set_facecolor('lightgrey')
ax[0,0].plot(data_x, data_y, 'r-')
ax[1,0].plot(data_x, data_y, 'b-')
fig.delaxes(ax[1,0])
ax[1,1].plot(data_x, data_y, 'g-')
# ... complete!
```

1	2
3	4

subplot()

vs

subplots()

STATE BASED

OBJECT ORIENTED

Plot

Subplots

Set(title, xlabel, ylabel)

Grid

Legend

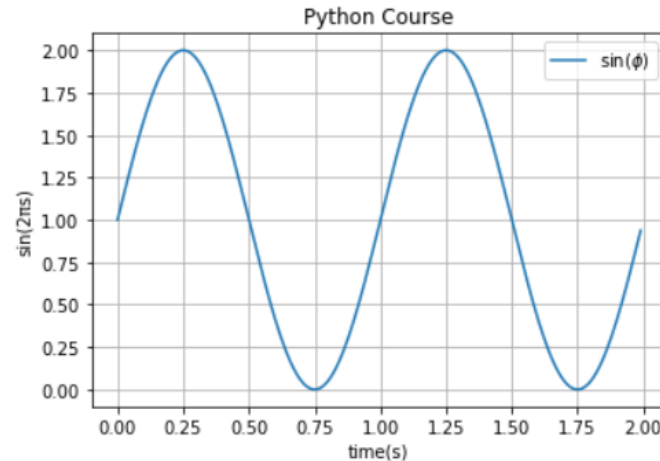
Savefig

```
import matplotlib.pyplot as plt
import numpy as np

# Data for plotting
x = np.arange(0.0, 2.0, 0.01)
y = 1 + np.sin(2 * np.pi * x)

fig, ax = plt.subplots()
ax.plot(x, y, label='sin( $\phi$ )')

ax.set(xlabel='time(s)', ylabel='sin(2\pi s)', title='Python Course')
ax.grid()
ax.legend()
fig.savefig("test.png")
plt.show()
```



https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

<https://pythonforundergradengineers.com/unicode-characters-in-python.html>

Boxplot

Boxplots are a standardized way of displaying the distribution of data based on a five number summary

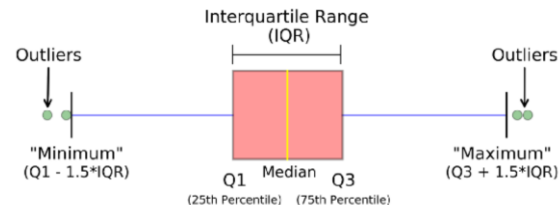
*Minimum $\geq (Q1 - 1.5 * IQR)$*

first quartile (Q1)

Median (Q2)

third quartile (Q3)

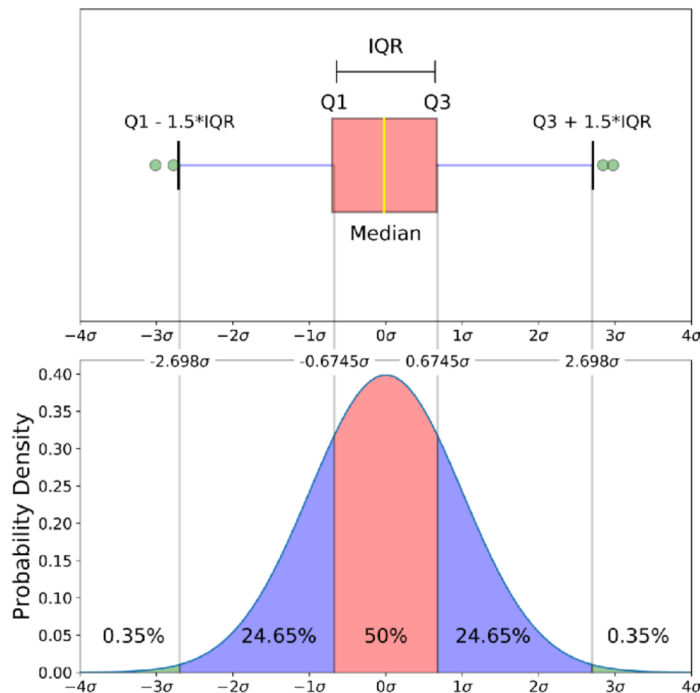
*Maximum $\leq (Q3 + 1.5 * IQR)$*



1. Tell you the values of your outliers
2. Identify if data is symmetrical
3. Determine how tightly data is grouped
4. See if your data is skewed

Boxplot

Boxplot on a Normal Distribution



$$f(x | \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

PDF for a Normal Distribution

mean (μ) of 0
standard deviation (σ) of 1

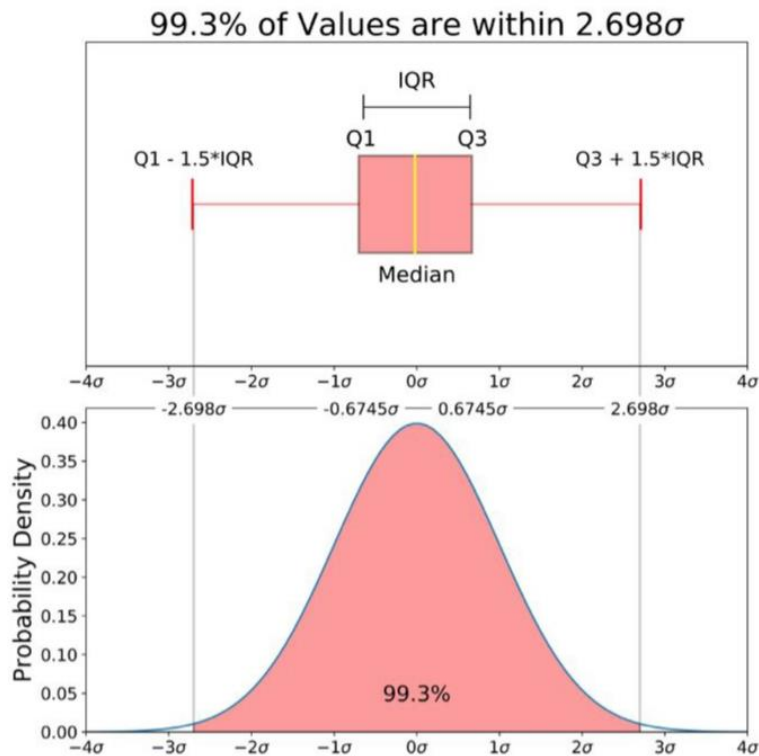
$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

PDF for a Normal Distribution

Boxplot

Boxplot on a Normal Distribution

By removing outliers, we have access to 99.3% of data on a normal distribution



Boxplot Example

```
import numpy as np
```

```
import matplotlib.pyplot as plt  
# from matplotlib import pyplot as plt
```

```
data=np.array([-10 , -5 , -2 , -1 , 0 , 1 , 2 , 3 , 4])
```

```
q1 = np.quantile(data, .25)  
q1
```

-2.0

```
q2 = np.quantile(data, .50)  
q2
```

0.0

```
q3 = np.quantile(data, .75)  
q3
```

2.0

```
iqr = q3 - q1  
iqr
```

4.0

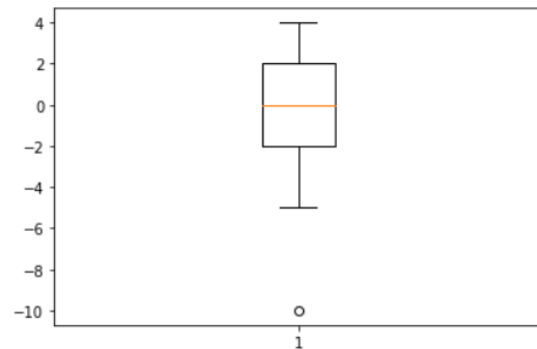
```
lv = q1 - 1.5 * iqr  
lv
```

-8.0

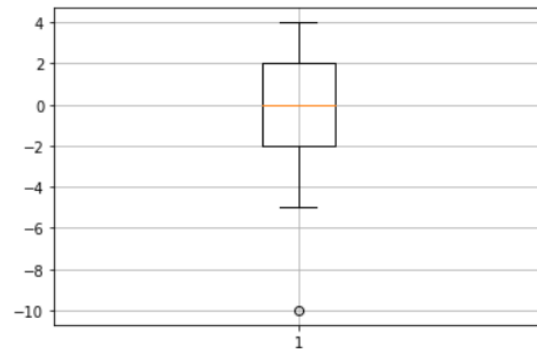
```
hv = q3 + 1.5 * iqr  
hv
```

8.0

```
plt.boxplot(data);
```



```
plt.boxplot(data);  
plt.grid()
```



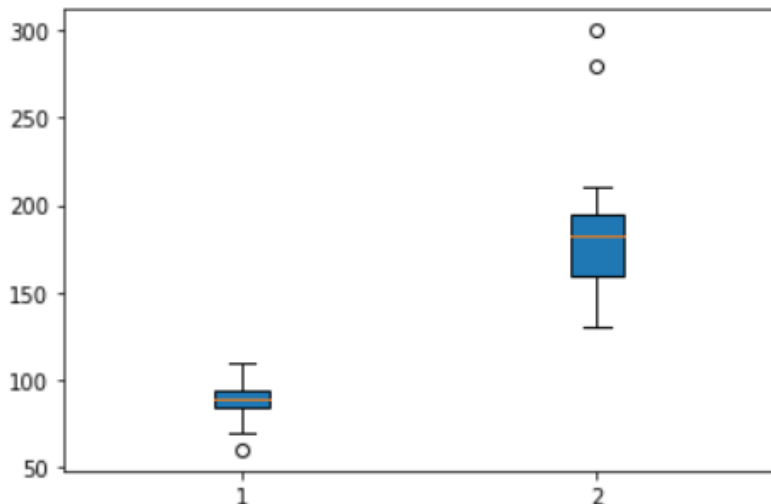
Boxplot Example

```
import numpy as np
import matplotlib.pyplot as plt
```

```
class1= np.array([60,70,80,83,85,87,88,89,90,92,94,95,97,100,110])
```

```
class2 = np.array([130,143,150,158,160,170,175,182,185,188,190,200,210,280,300])
```

```
plt.boxplot([class1,class2],patch_artist=True);
```



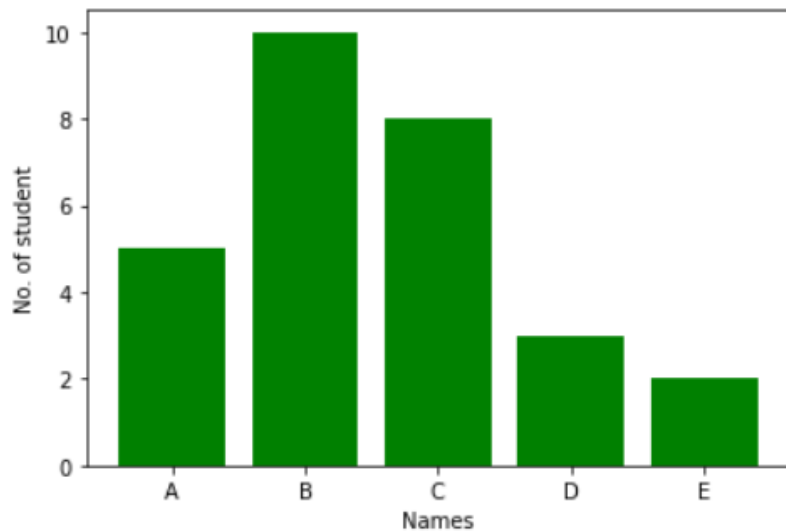
Bar

```
import matplotlib.pyplot as plt  
import numpy as np
```

```
names = ['A', 'B', 'C', 'D', 'E']
```

```
values = [ 5, 10, 8, 3, 2]
```

```
plt.bar(names, values, color='green');  
plt.xlabel('Names');  
plt.ylabel('No. of student');
```

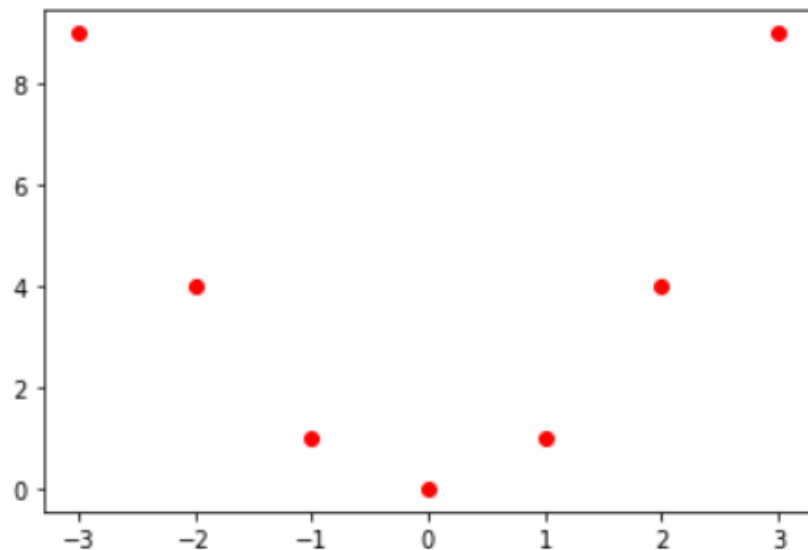


Scatter

```
import numpy as np
import matplotlib.pyplot as plt
```

```
x=np.array([-3,-2,-1,0,1,2,3])
y=np.array([9,4,1,0,1,4,9])
```

```
plt.scatter(x, y, c='r');
```

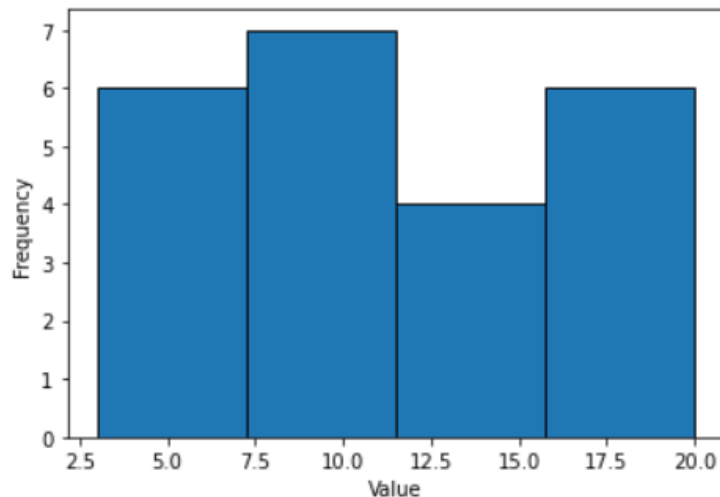


Histogram

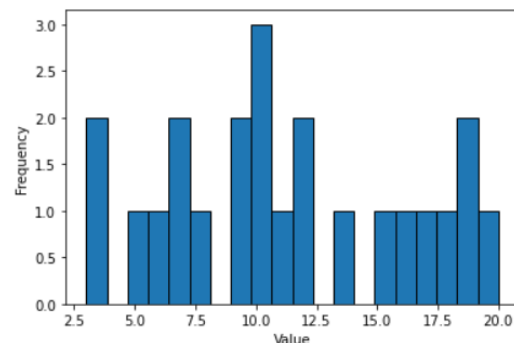
```
import numpy as np
import matplotlib.pyplot as plt
```

```
data = np.array([3,3,5,6,7,7,8,9,9,10,10,10,11,12,12,14,15,16,17,18,19,19,20])
```

```
plt.hist(data, bins=4, edgecolor='black');
plt.xlabel('Value');
plt.ylabel('Frequency');
```



```
plt.hist(data, bins=20, edgecolor='black');
plt.xlabel('Value');
plt.ylabel('Frequency');
```

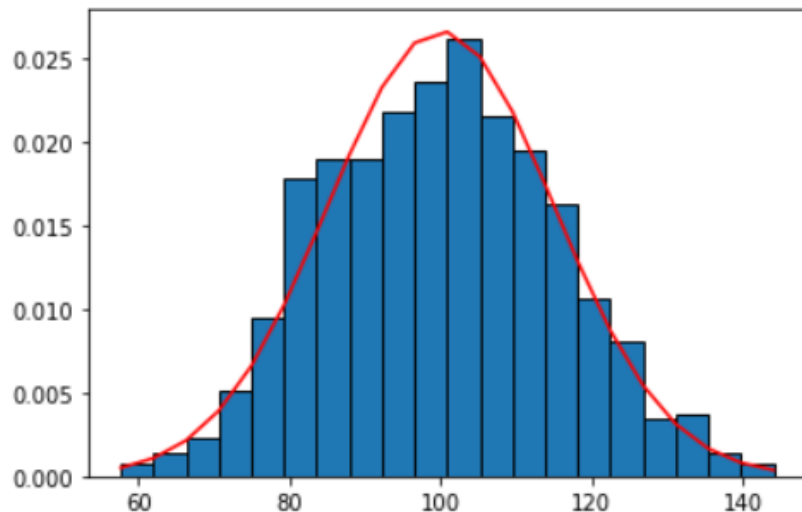


Histogram

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import norm
```

```
x = 100 + 15 * np.random.randn(1000)
```

```
n, bins, _ = plt.hist(x, bins=20, edgecolor='black', density=1)
y = norm.pdf(bins, 100, 15)
plt.plot(bins, y, 'r');
```

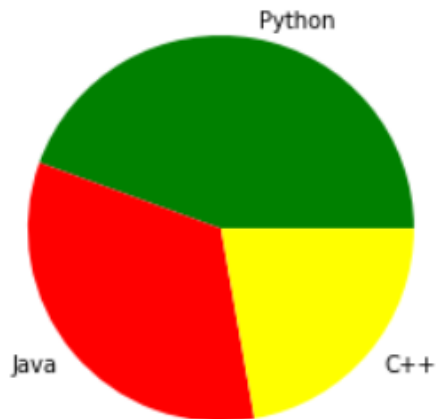


Pie

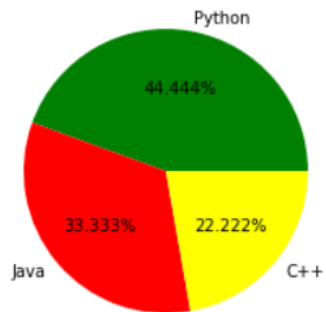
```
import matplotlib.pyplot as plt
```

```
l = ['Python', 'Java', 'C++']  
s = [ 200, 150, 100]  
c = [ 'green', 'red', 'yellow']
```

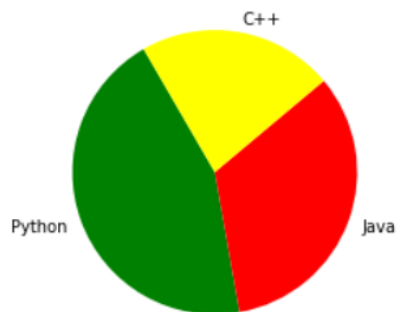
```
plt.pie(s, labels=l, colors=c);
```



```
plt.pie(s, labels=l, colors=c, autopct='%1.3f%%');
```



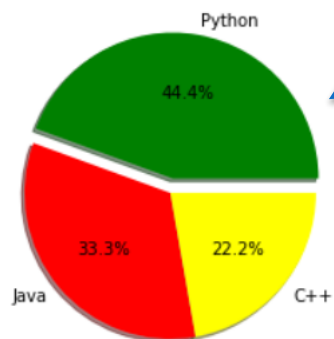
```
plt.pie(s, labels=l, colors=c, startangle=120);
```



Rotate

Pie

```
plt.pie(s, labels=l, colors=c, explode=(0.1, 0, 0), shadow=True, autopct='%1.1f%%');
```



Data Visualization with Pandas

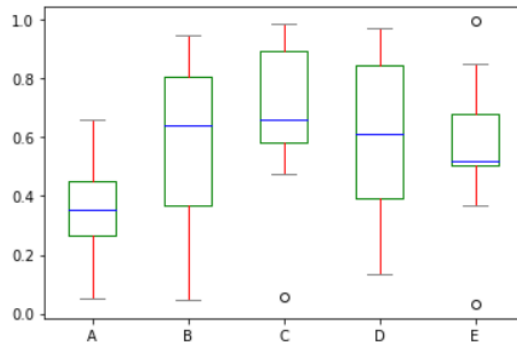
Pandas Box

```
import pandas as pd
```

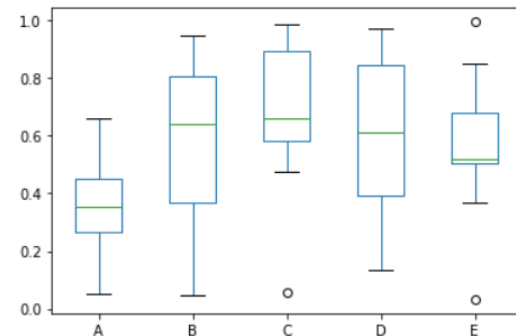
```
df = pd.DataFrame(np.random.rand(10, 5),  
                  columns=['A', 'B', 'C', 'D', 'E'])  
df
```

	A	B	C	D	E
0	0.049709	0.045436	0.055949	0.843779	0.512288
1	0.660786	0.840155	0.724785	0.132184	0.850772
2	0.050459	0.669042	0.627813	0.286994	0.032594
3	0.248601	0.601896	0.568547	0.573618	0.524015
4	0.356679	0.829488	0.474301	0.548752	0.503438
5	0.548138	0.290860	0.984472	0.343178	0.369197
6	0.323869	0.739252	0.952468	0.651361	0.717576
7	0.385569	0.162749	0.972111	0.906281	0.573732
8	0.475793	0.608820	0.683542	0.841382	0.502442
9	0.347030	0.949040	0.636254	0.970489	0.995102

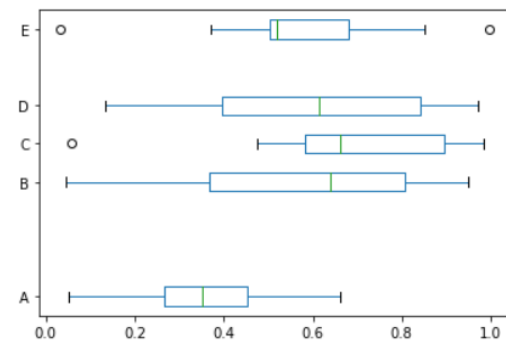
```
df.plot.box(color={'boxes': 'Green', 'whiskers': 'red',  
                  'medians': 'Blue', 'caps': 'Gray'});
```



```
df.plot.box();
```



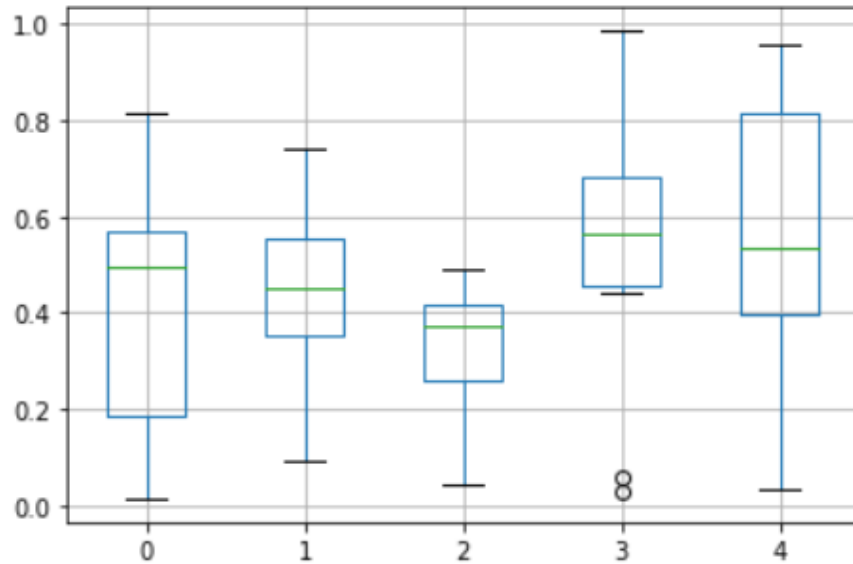
```
df.plot.box(vert=False, positions=[1, 4, 5, 6, 8]);
```



Pandas Boxplot

```
import pandas as pd
```

```
df = pd.DataFrame(np.random.rand(10, 5))  
df.boxplot();
```



Pandas

Bar

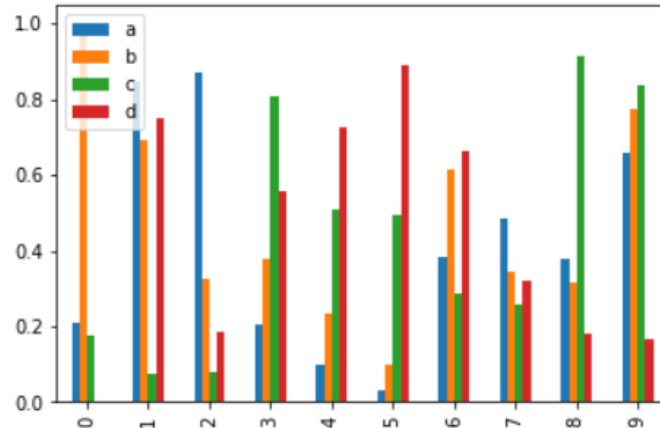
Barh

```
import pandas as pd
```

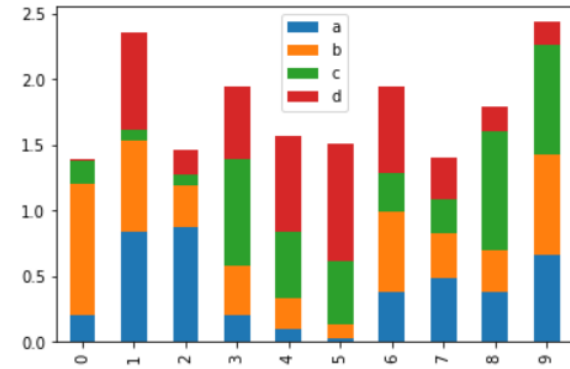
```
df = pd.DataFrame(np.random.rand(10, 4), columns=['a', 'b', 'c', 'd'])  
df
```

	a	b	c	d
0	0.208853	0.999368	0.177955	0.004178
1	0.844905	0.692018	0.074666	0.749286
2	0.871019	0.324711	0.080421	0.184144
3	0.204348	0.379936	0.809383	0.554804
4	0.097902	0.235168	0.507640	0.725698
5	0.032330	0.096933	0.492393	0.891947
6	0.380896	0.613993	0.288315	0.664387
7	0.485372	0.342866	0.256060	0.318199
8	0.380383	0.317800	0.911732	0.178845
9	0.658787	0.773167	0.835882	0.168122

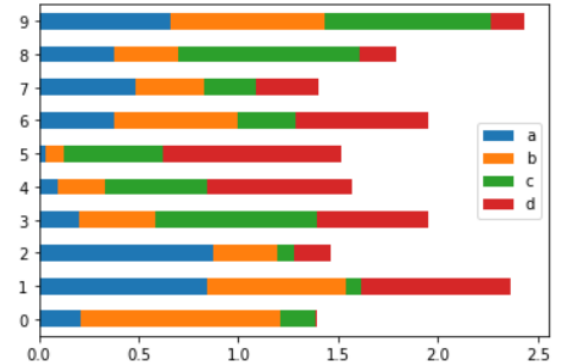
```
df.plot.bar();
```



```
df.plot.bar(stacked=True);
```



```
df.plot.barh(stacked=True);
```



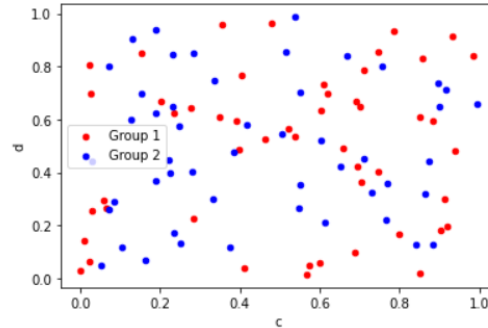
Pandas Scatter

```
import pandas as pd
import numpy as np
```

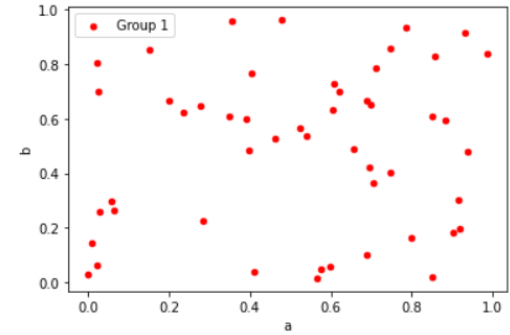
```
df = pd.DataFrame(np.random.rand(50, 4), columns=['a', 'b', 'c', 'd'])
df.head()
```

	a	b	c	d
0	0.390322	0.597285	0.232865	0.173990
1	0.913780	0.302225	0.051365	0.048679
2	0.882417	0.593372	0.332704	0.300124
3	0.057654	0.296669	0.728719	0.323799
4	0.745395	0.857029	0.188868	0.366672

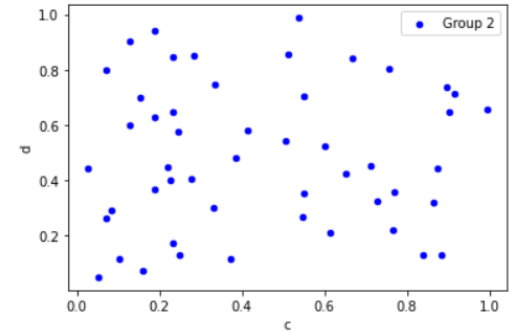
```
k=df.plot.scatter(x='a', y='b', color='red', label='Group 1')
df.plot.scatter(x='c', y='d', color='blue', label='Group 2',ax=k);
```



```
df.plot.scatter(x='a', y='b', color='red', label='Group 1');
```



```
df.plot.scatter(x='c', y='d', color='blue', label='Group 2');
```



Pandas

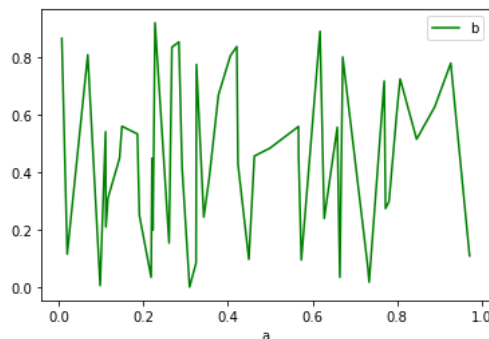
Plot (kind = ...)

```
import numpy as np
import pandas as pd
```

```
df = pd.DataFrame(np.random.rand(50, 3), columns=['a', 'b', 'c'])
df['c']*=200
df.sort_values(by = 'a', inplace=True)
df.head(7)
```

	a	b	c
10	0.007162	0.867517	195.458007
9	0.019932	0.116770	39.280149
23	0.068505	0.810289	91.971103
15	0.097792	0.006978	145.376167
36	0.110587	0.542228	106.221738
18	0.111375	0.211138	161.600937
31	0.116303	0.311410	13.561107

```
df.plot(kind='line', x= 'a', y='b' , color='green');
```



kind : str

The kind of plot to produce:

- 'line' : line plot (default)
- 'bar' : vertical bar plot
- 'barh' : horizontal bar plot
- 'hist' : histogram
- 'box' : boxplot
- 'kde' : Kernel Density Estimation plot
- 'density' : same as 'kde'
- 'area' : area plot
- 'pie' : pie plot
- 'scatter' : scatter plot
- 'hexbin' : hexbin plot.

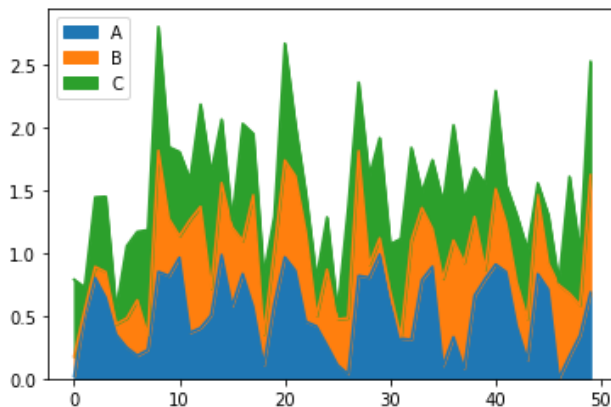
Pandas Area

```
import numpy as np
import pandas as pd
```

```
df = pd.DataFrame(np.random.rand(50, 3), columns=['A', 'B', 'C'])
df.head(7)
```

	A	B	C
0	0.393983	0.831335	0.934014
1	0.891985	0.953771	0.052773
2	0.081741	0.386546	0.766863
3	0.739023	0.967765	0.043960
4	0.072500	0.872009	0.205749
5	0.713897	0.390277	0.075670
6	0.849453	0.796737	0.862341

```
df.plot(kind='area');
```



Pandas Scatter style

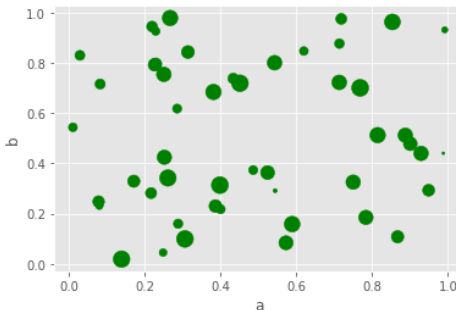
```
import numpy as np
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
plt.style.use('ggplot') ←
df = pd.DataFrame(np.random.rand(50, 3), columns=['a', 'b', 'c'])
df['c'] *= 200
df.head()
```

	a	b	c
0	0.750055	0.325310	141.280214
1	0.125275	0.028598	6.777477
2	0.713286	0.722529	147.216246
3	0.768355	0.700825	192.918051
4	0.783541	0.184801	139.875161

```
df.plot(kind='scatter', x='a', y='b', s='c', color='green');
```



```
plt.style.available
```

```
['Solarize_Light2',
 '_classic_test_patch',
 'bmh',
 'classic',
 'dark_background',
 'fast',
 'fivethirtyeight',
 'ggplot',
 'grayscale',
 'seaborn',
 'seaborn-bright',
 'seaborn-colorblind',
 'seaborn-dark',
 'seaborn-dark-palette',
 'seaborn-darkgrid',
 'seaborn-deep',
 'seaborn-muted',
 'seaborn-notebook',
 'seaborn-paper',
 'seaborn-pastel',
 'seaborn-poster',
 'seaborn-talk',
 'seaborn-ticks',
 'seaborn-white',
 'seaborn-whitegrid',
 'tableau-colorblind10']
```

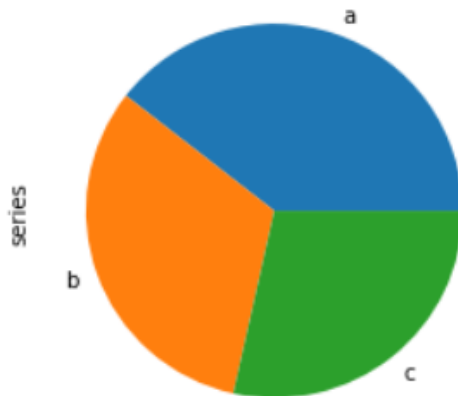
Pandas Series Pie

```
import pandas as pd  
import numpy as np
```

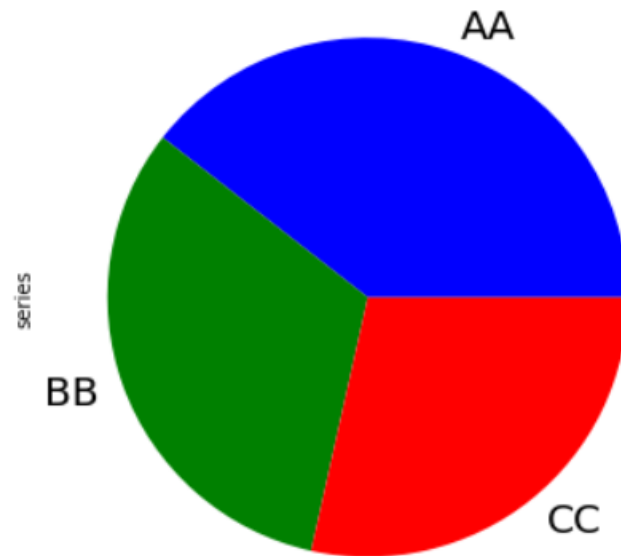
```
s = pd.Series(3 * np.random.rand(3),  
              index=['a', 'b', 'c'],  
              name='series')
```

```
s  
  
a    1.635387  
b    1.329396  
c    1.181540  
Name: series, dtype: float64
```

```
s.plot.pie();
```



```
s.plot.pie(labels=['AA', 'BB', 'CC'],  
           colors=['b', 'g', 'r'],  
           fontsize=20,  
           figsize=(6, 6));
```



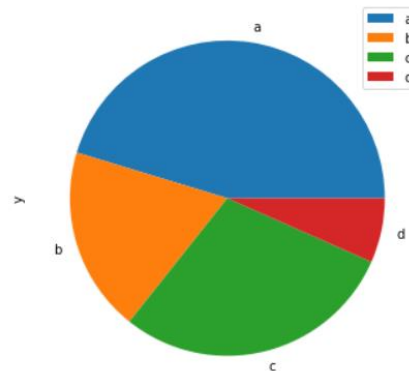
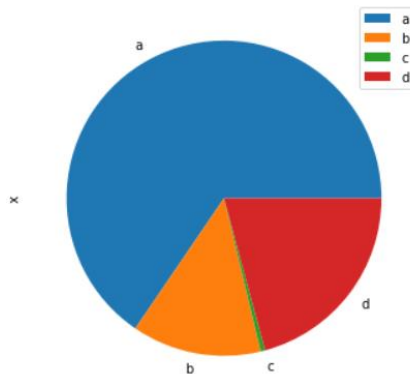
Pandas DataFrames

Pie

```
df = pd.DataFrame(3 * np.random.rand(4, 2),  
                  index=['a', 'b', 'c', 'd'],  
                  columns=['x', 'y'])  
df
```

	x	y
a	2.280157	2.398253
b	0.462583	1.005063
c	0.015842	1.538840
d	0.724033	0.350281

```
df.plot.pie(subplots=True, figsize=(12, 8));
```



Meshgrid

The `numpy.meshgrid` function is used to create a rectangular grid out of two given one-dimensional arrays representing the Cartesian indexing or Matrix indexing.

```
import numpy as np
```

```
x = np.array([1,2,3,4])  
y = np.array([7,8])
```

```
a,b = np.meshgrid(x, y)
```

a

```
array([[1, 2, 3, 4],  
       [1, 2, 3, 4]])
```

b

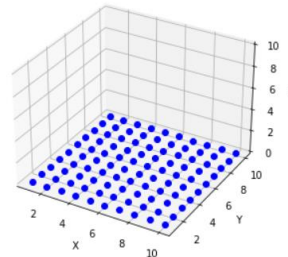
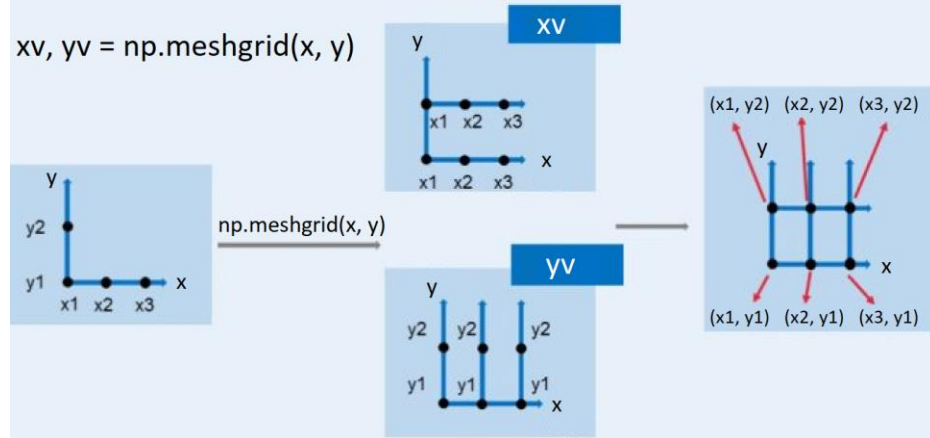
```
array([[7, 7, 7, 7],  
       [8, 8, 8, 8]])
```

a.shape

```
(2, 4)
```

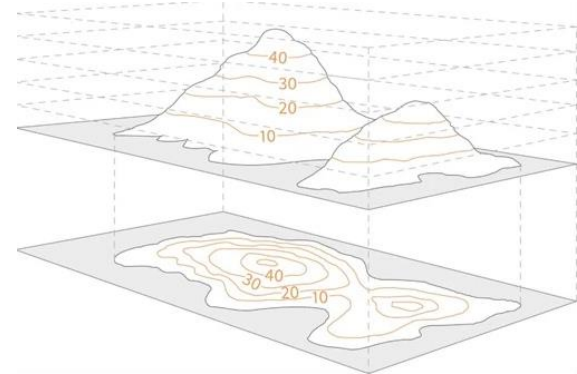
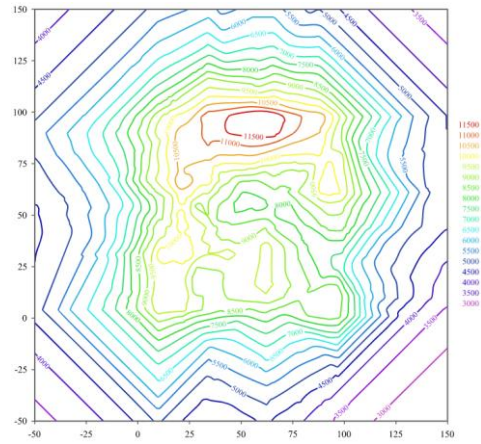
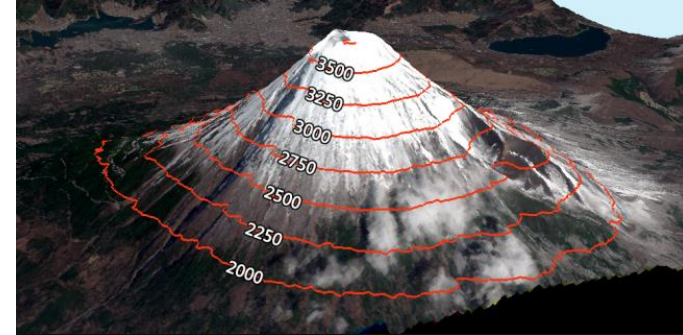
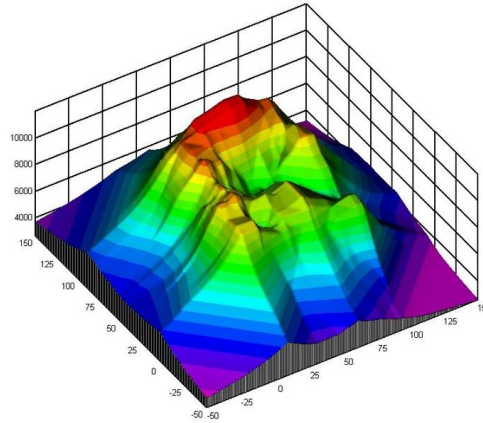
Numpy.meshgrid

xv, yv = np.meshgrid(x, y)



Contour?

What is the Contour? (TOPOGRAPHIC MAPS)



3D Plot

`mpl_toolkits.mplot3d` Plot_surface Contour

Create an empty 3D figure →

Plot 3D surface →

Contour →

Color bar →

Contour plots (sometimes called Level Plots) are a way to show a three-dimensional surface on a two-dimensional plane.

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
```

```
x = np.arange(-10, 10, 0.5)
y = np.arange(-10, 10, 0.5)
X, Y = np.meshgrid(x, y)
Z = X ** 2 + Y ** 2
```

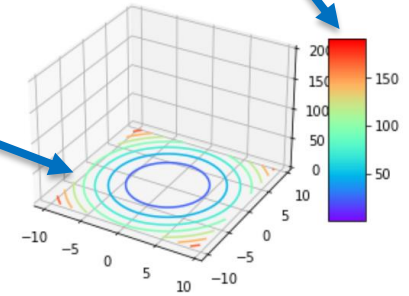
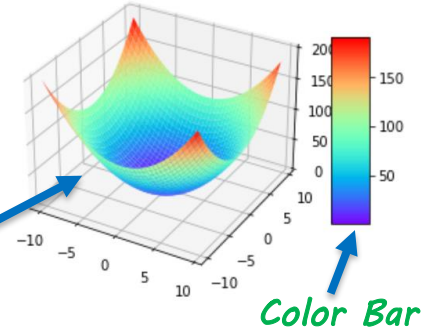
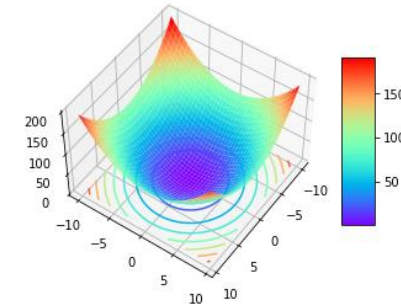
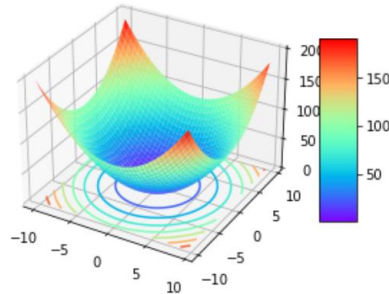
```
fig = plt.figure(figsize=(5, 5))
ax = fig.gca(projection='3d')

s = ax.plot_surface(X, Y, Z, cmap=plt.cm.rainbow)

cset = ax.contour(X, Y, Z, zdir='z', offset=0, cmap=plt.cm.rainbow)

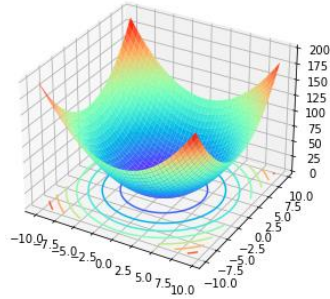
fig.colorbar(s, shrink=0.5, aspect=5);
```

```
ax.view_init(50, 35)
fig
```

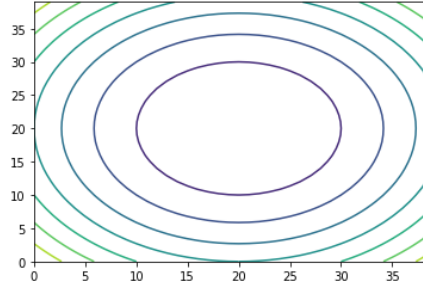


Contour Contourf

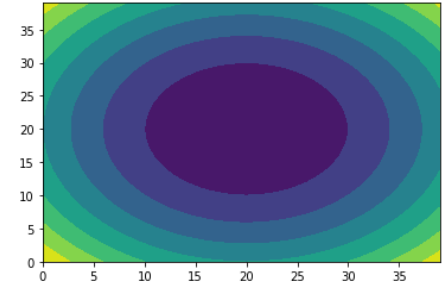
contour and contourf draw contour lines and filled contours, respectively. Except as noted, function signatures and return values are the same for both versions.



```
plt.contour(Z);
```



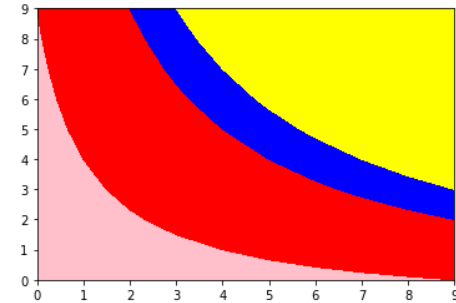
```
plt.contourf(Z);
```



```
import numpy as np
import matplotlib.pyplot as plt
```

```
x = np.arange(1, 11)
y = x.reshape(-1, 1)
h = x*y
```

```
cs = plt.contourf(h, levels=[10, 30, 40], colors=['r', 'b'], extend='both')
cs.cmap.set_over('yellow')
cs.cmap.set_under('pink')
cs.changed()
```



“

- *Make it work*
- *Make it Right*
- *Make it Fast*

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