

### 3. NumPy

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```
In [1]: 1 import numpy as np
        2 import matplotlib.pyplot as plt
        3 %matplotlib inline
```

```
In [2]: 1 x = np.array([6, 7, 8])      # create an np array
        2 x
```

Out[2]: array([6, 7, 8])

```
In [3]: 1 # printing an NumPy object
        2 print (x)
```

[6 7 8]

```
In [4]: 1 # multi-dimensional array
        2 y = np.array([[0,1,2],[3,4,5],[6,7,8]])
        3 print ('y =', y)
        4 print ('size: ', y.size)           # number of elements
        5 print ('type: ', y.dtype)         # data type of elements
        6 print ('ndim: ', y.ndim)         # dimension of the array
        7 print ('shape:', y.shape)        # shape of the array
```

```
y = [[0 1 2]
      [3 4 5]
      [6 7 8]]
size: 9
type: int32
ndim: 2
shape: (3, 3)
```

```
In [5]: 1 # various types
        2 print (np.array([ [1,2], [3,4] ], dtype=int))
        3 print (np.array([ [1,2], [3,4] ], dtype=float))
        4 print (np.array([ [1,2], [3,4] ], dtype=complex))
        5 x = 1+5j
        6 print (x)
```

```
[[1 2]
 [3 4]]
[[1. 2.]
 [3. 4.]]
[[1.+0.j 2.+0.j]
 [3.+0.j 4.+0.j]]
(1+5j)
```

```
In [6]: 1 # numerous array functions
        2 print ('identity:\n', np.identity(3))
```

```
identity:
[[1. 0. 0.]
 [0. 1. 0.]
 [0. 0. 1.]]
```

```
In [7]: 1 print ('zeros:\n', np.zeros((3,4)))
```

```
zeros:
[[0. 0. 0. 0.]
 [0. 0. 0. 0.]
 [0. 0. 0. 0.]]
```

```
In [8]: 1 print ('ones:\n', np.ones( (2,3,4), dtype=np.int16 ) )
```

```
ones:
[[[1 1 1 1]
  [1 1 1 1]
  [1 1 1 1]]

 [[1 1 1 1]
  [1 1 1 1]
  [1 1 1 1]]]
```

## vectorized computing

```
In [9]: 1 # simple array arithmetic
        2 a = np.array([ [1,2], [3,4] ])
        3 b = np.array([ [5,6], [7,8] ], dtype=complex)
        4
        5 print ('a + b\n', a + b)
        6 print ('a - b\n', a - b)
        7 print ('a * b\n', a * b)
        8 print ('a / b\n', a / b)
        9
```

```
a + b
[[ 6.+0.j  8.+0.j]
 [10.+0.j 12.+0.j]]
a - b
[[-4.+0.j -4.+0.j]
 [-4.+0.j -4.+0.j]]
a * b
[[ 5.+0.j 12.+0.j]
 [21.+0.j 32.+0.j]]
a / b
[[0.2      +0.j 0.33333333+0.j]
 [0.42857143+0.j 0.5      +0.j]]
```

```
In [10]: 1 print ('a ** 2\n', a ** 2)
2 print ('10 * sin(a)\n', 10 * np.sin(a))

a ** 2
[[ 1  4]
 [ 9 16]]
10 * sin(a)
[[ 8.41470985  9.09297427]
 [ 1.41120008 -7.56802495]]
```

```
In [11]: 1 # numerous aux functions
2 a = np.random.random((2,5))
3 print ('a =\n', a)
4 print ('a.sum =', a.sum())
5 print ('a.min =', a.min())
6 print ('a.max =', a.max())

a =
[[0.64458286 0.15065244 0.25598586 0.02459517 0.2319529 ]
 [0.82608774 0.43347058 0.67035519 0.92331687 0.15584371]]
a.sum = 4.316843342203911
a.min = 0.02459517452385529
a.max = 0.92331687346472
```

## powerful indexing

```
In [12]: 1 # sub-array in row
2 a = np.random.random((5,3))
3 print ('a =\n', a)
4 print ('a[1:4] =\n', a[1:4])
5

a =
[[0.69963029 0.60533353 0.45818071]
 [0.85565126 0.18453609 0.9522594 ]
 [0.16555263 0.84277372 0.94436354]
 [0.88798277 0.29993386 0.95844613]
 [0.70249742 0.16514484 0.42562092]]
a[1:4] =
[[0.85565126 0.18453609 0.9522594 ]
 [0.16555263 0.84277372 0.94436354]
 [0.88798277 0.29993386 0.95844613]]
```

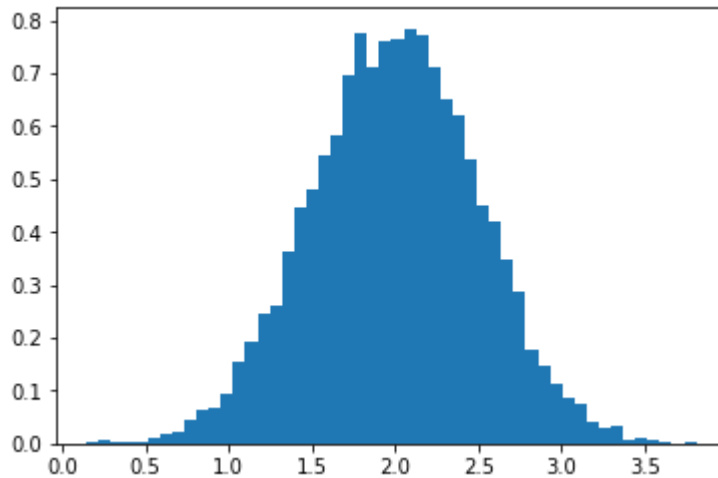
```
In [13]: 1 # sub-array in col
2 print ('a[:,1:] =\n', a[:,1:])

a[:,1:] =
[[0.60533353 0.45818071]
 [0.18453609 0.9522594 ]
 [0.84277372 0.94436354]
 [0.29993386 0.95844613]
 [0.16514484 0.42562092]]
```

## numpy has its own powerful random library

```
In [14]: 1 # work well with graphic library
          2 mu, sigma = 2, 0.5
          3 v = np.random.normal(mu,sigma,10000)
```

```
In [15]: 1 # Plot a normalized histogram with 50 bins
          2 plt.hist(v, bins=50, density=1) # Plot a normalized histogram with 50 bi
          3 plt.show()
```



## Numpy vs. MatLab

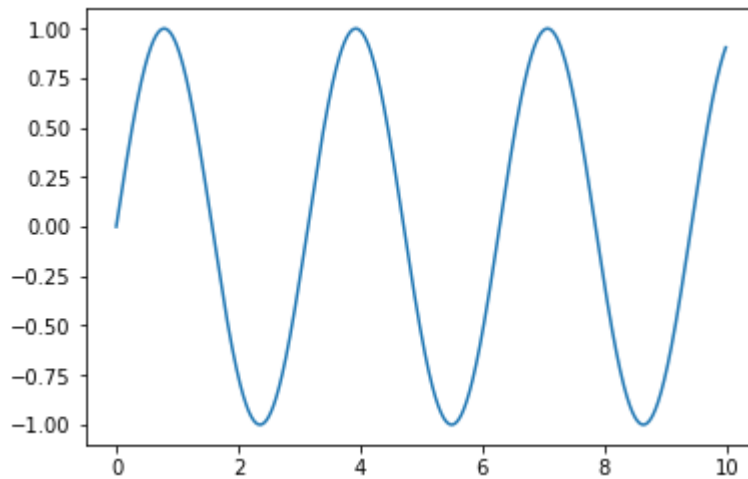
I found the following from a MatLab tutorial

```
x = [0:.01:10];
y = sin(2 * x);
plot(x,y)
```

Its equivalent NumPy code will be as below:

```
In [16]: 1 x = np.arange(0,10, 0.01)
          2 y = np.sin(2 * x)
          3 plt.plot(x,y)
```

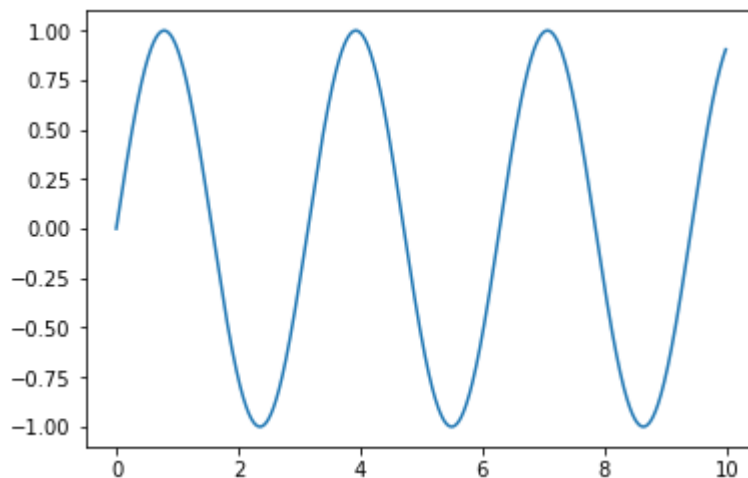
Out[16]: [`<matplotlib.lines.Line2D at 0x22d292c0ef0>`]



even without the namespace ...

```
In [17]: 1 from numpy import *
          2 from matplotlib.pyplot import *
          3
          4 x = arange(0,10, 0.01)
          5 y = sin(2 * x)
          6 plot(x,y)
```

Out[17]: [`<matplotlib.lines.Line2D at 0x22d293302e8>`]



## Linear algebra (numpy.linalg)

<https://docs.scipy.org/doc/numpy/reference/routines.linalg.html>

(<https://docs.scipy.org/doc/numpy/reference/routines.linalg.html>)

In [18]:

```
1 import numpy as np
2 r = np.array([
3     [0, 1, 1, 1],
4     [1, 0, 0, 1],
5     [1, 0, 0, 1],
6     [1, 1, 1, 0]])
7 w,v = np.linalg.eig(r)
8 print ("w =", w)
9 print ("v =", v)
```

```
w = [ 2.56155281e+00 -1.00000000e+00 -1.56155281e+00 -5.49681202e-18]
v = [[-5.57345410e-01  7.07106781e-01 -4.35162146e-01  1.98492213e-17]
     [-4.35162146e-01 -3.97111101e-16  5.57345410e-01 -7.07106781e-01]
     [-4.35162146e-01 -7.79219809e-17  5.57345410e-01  7.07106781e-01]
     [-5.57345410e-01 -7.07106781e-01 -4.35162146e-01 -6.87807092e-17]]
```

In [ ]:

1