

Python 6: NumPy

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Intro to NumPy

❖ What is NumPy?

- Fast array processing
- Iteration via loops in interpreted languages (including Python) is relatively slow
- Operations are sent in batches to optimized C and Fortran code

❖ NumPy Arrays

```
In [1]: import numpy as np
```

```
In [2]: a = np.zeros(3)
```

```
In [3]: a
```

```
Out [3]: array([ 0.,  0.,  0.])
```

```
In [4]: type(a)
```

```
Out [4]: numpy.ndarray
```

NumPy dtype

❖ NumPy arrays vs. Python lists

- Data must be homogeneous (all elements of the same type)
- These types (dtypes) are provided by NumPy

❖ dtypes

- float64 (default type)
- float32
- int64
- int32
- bool

```
In [7]: a = np.zeros(3)
```

```
In [8]: type(a[0])
```

```
Out[8]: numpy.float64
```

```
In [9]: a = np.zeros(3, dtype=int)
```

```
In [10]: type(a[0])
```

```
Out[10]: numpy.int32
```

Shape and Dimension

❖ “flat” array

```
In [11]: z = np.zeros(10)
```

```
In [12]: z.shape
```

```
Out[12]: (10,) # Note syntax for tuple with one element
```

❖ Changing shape

```
In [13]: z.shape = (10, 1)
```

```
In [14]: z
```

```
Out[14]:
```

```
array([[ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.],  
       [ 0.]])
```

```
In [15]: z = np.zeros(4)
```

```
In [16]: z.shape = (2, 2)
```

```
In [17]: z
```

```
Out[17]:
```

```
array([[ 0.,  0.],  
       [ 0.,  0.]])
```

Creating Arrays

❖ Empty array

```
In [18]: z = np.empty(3)
```

```
In [19]: z
```

```
Out[19]: array([ 8.90030222e-307,  4.94944794e+173,  4.04144187e-262])
```

❖ Grid of evenly spaced numbers

```
In [20]: z = np.linspace(2, 4, 5) # From 2 to 4, with 5 elements
```

❖ Identity

```
In [21]: z = np.identity(2)
```

```
In [22]: z
```

```
Out[22]:
```

```
array([[ 1.,  0.],  
       [ 0.,  1.]])
```

Creating Array

❖ From Python lists, tuples, etc.

```
In [23]: z = np.array([10, 20])
```

```
In [24]: z
```

```
Out[24]: array([10, 20])
```

```
In [25]: type(z)
```

```
Out[25]: numpy.ndarray
```

```
In [26]: z = np.array((10, 20), dtype=float)
```

```
In [27]: z
```

```
Out[27]: array([ 10.,  20.])
```

```
In [28]: z = np.array([[1, 2], [3, 4]])
```

```
In [29]: z
```

```
Out[29]:
```

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

Indexing

```
In [30]: z = np.linspace(1, 2, 5)
```

```
In [31]: z
```

```
Out[31]: array([ 1.   ,  1.25,  1.5   ,  1.75,  2.   ])
```

```
In [32]: z[0]
```

```
Out[32]: 1.0
```

```
In [33]: z[0:2] # Slice numbering is left closed, right open
```

```
Out[33]: array([ 1.   ,  1.25])
```

```
In [34]: z[-1]
```

```
Out[34]: 2.0
```

2D

```
In [35]: z = np.array([[1, 2], [3, 4]])
```

```
In [36]: z
```

```
Out[36]:
```

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

```
In [37]: z[0, 0]
```

```
Out[37]: 1
```

```
In [39]: z[0,:]
```

```
Out[39]: array([1, 2])
```

```
In [38]: z[0, 1]
```

```
Out[38]: 2
```

```
In [40]: z[:,1]
```

```
Out[40]: array([2, 4])
```

```
In [42]: z
```

```
Out[42]: array([ 2. ,  2.5,  3. ,  3.5,  4. ])
```

```
In [43]: indices = np.array((0, 2, 3))
```

```
In [44]: z[indices]
```

```
Out[44]: array([ 2. ,  3. ,  3.5])
```

```
In [46]: d = np.array([0, 1, 1, 0, 0], dtype=bool)
```

```
In [47]: d
```

```
Out[47]: array([False,  True,  True, False, False], dtype=bool)
```

```
In [48]: z[d]
```

```
Out[48]: array([ 2.5,  3. ])
```

```
In [49]: z = np.empty(3)
```

```
In [50]: z
```

```
Out[50]: array([-1.25236750e-041,  0.00000000e+000,  5.45693855e-313])
```

```
In [51]: z[:] = 42
```

```
In [52]: z
```

```
Out[52]: array([ 42.,  42.,  42.])
```


Methods

Attribute		
	dtype	ndim
	shape	size
	T	
Method		
	all	any
	argmax	argmin
	max	min
	mean	
	std	var
	prod	sum
	cumprod	cumsum
	dot	transpose
	diagonal	trace
	sort	copy

```
In [53]: A = np.array((4, 3, 2, 1))
```

```
In [54]: A
```

```
Out[54]: array([4, 3, 2, 1])
```

```
In [55]: A.sort()                                # Sorts A in place
```

```
In [56]: A
```

```
Out[56]: array([1, 2, 3, 4])
```

```
In [57]: A.sum()                                # Sum
```

```
Out[57]: 10
```

```
In [58]: A.mean()                               # Mean
```

```
Out[58]: 2.5
```

```
In [64]: A.std()
```

```
Out[64]: 1.1180339887498949
```

```
In [65]: A.shape = (2, 2)
```

```
In [66]: A.T
```

```
Out[66]:
array([[1, 3],
       [2, 4]])
```

Algebraic Operation

```
In [75]: a = np.array([1, 2, 3, 4])
```

```
In [76]: b = np.array([5, 6, 7, 8])
```

```
In [77]: a + b
```

```
Out[77]: array([ 6,  8, 10, 12])
```

```
In [78]: a * b
```

```
Out[78]: array([ 5, 12, 21, 32])
```

```
In [79]: a + 10
```

```
Out[79]: array([11, 12, 13, 14])
```

```
In [82]: a * 10
```

```
Out[82]: array([10, 20, 30, 40])
```

```
In [86]: A = np.ones((2, 2))
```

```
In [87]: B = np.ones((2, 2))
```

```
In [88]: A + B
```

```
Out[88]:  
array([[ 2.,  2.],  
       [ 2.,  2.]])
```

```
In [89]: A + 10
```

```
Out[89]:  
array([[ 11.,  11.],  
       [ 11.,  11.]])
```

```
In [90]: A * B
```

```
Out[90]:  
array([[ 1.,  1.],  
       [ 1.,  1.]])
```

In particular, $A * B$ is *not* the matrix product, it is an elementwise product

Matrix Multiplication

```
In [137]: A = np.ones((2, 2))
```

```
In [138]: B = np.ones((2, 2))
```

```
In [139]: np.dot(A, B)
```

```
Out[139]:
```

```
array([[ 2.,  2.],  
       [ 2.,  2.]])
```

```
In [91]: A = np.array([1, 2])
```

```
In [92]: B = np.array([10, 20])
```

```
In [93]: np.dot(A, B)    # Returns a scalar in this case
```

```
Out[93]: 50
```

Comparisons

❖ Elementwise comparisons

```
In [97]: z = np.array([2, 3])
```

```
In [98]: y = np.array([2, 3])
```

```
In [99]: z == y
```

```
Out[99]: array([ True,  True], dtype=bool)
```

```
In [100]: y[0] = 5
```

```
In [101]: z == y
```

```
Out[101]: array([False,  True], dtype=bool)
```

```
In [102]: z != y
```

```
Out[102]: array([ True, False], dtype=bool)
```

```
In [103]: z = np.linspace(0, 10, 5)
```

```
In [109]: z[z > 3]
```

```
Out[109]: array([ 5. ,  7.5, 10. ])
```

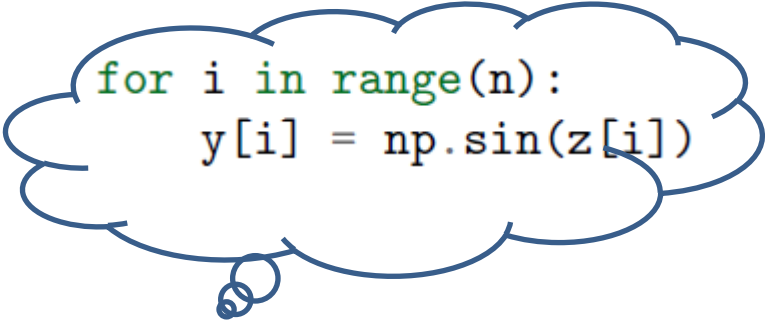
```
In [104]: z
```

```
Out[104]: array([ 0. ,  2.5,  5. ,  7.5, 10. ])
```

```
In [105]: z > 3
```

```
Out[105]: array([False, False,  True,  True,  True], dtype=bool)
```

Vectorized Functions



```
for i in range(n):  
    y[i] = np.sin(z[i])
```

```
In [110]: z = np.array([1, 2, 3])
```

```
In [111]: np.sin(z)
```

```
Out[111]: array([ 0.84147098,  0.90929743,  0.14112001])
```

```
In [113]: (1 / np.sqrt(2 * np.pi)) * np.exp(- 0.5 * z**2)
```

```
Out[113]: array([ 0.24197072,  0.05399097,  0.00443185])
```

np.where

```
In [114]: import numpy as np
```

```
In [115]: x = np.random.randn(4)
```

```
In [116]: x
```

```
Out[116]: array([-0.25521782,  0.38285891, -0.98037787, -0.083662  ])
```

```
In [117]: np.where(x > 0, 1, 0)  # Insert 1 if x > 0 true, otherwise 0
```

```
Out[117]: array([0, 1, 0, 0])
```

❖ Vectorized function

```
In [118]: def f(x): return 1 if x > 0 else 0
```

```
In [119]: f = np.vectorize(f)
```

```
In [120]: f(x)  # Passing same vector x as previous example
```

```
Out[120]: array([0, 1, 0, 0])
```

Other NumPy Functions

```
In [131]: A = np.array([[1, 2], [3, 4]])
```

```
In [132]: np.linalg.det(A)                # Compute the determinant
```

```
Out[132]: -2.0000000000000004
```

```
In [133]: np.linalg.inv(A)                # Compute the inverse
```

```
Out[133]:
```

```
array([[ -2. ,  1. ],  
       [ 1.5, -0.5]])
```

```
In [134]: Z = np.random.randn(10000)    # Generate standard normals
```

```
In [135]: y = np.random.binomial(10, 0.5, size=1000)
```

Q & A

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