

NUMPY

NumPy's main object is the homogeneous multidimensional array. It is a table of elements (usually numbers), all of the same type, indexed by a tuple of non-negative integers. In NumPy dimensions are called axes.

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

```
In [2]: a = np.arange(15).reshape(3, 5)
a
```

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

```
In [3]: # ndarray.ndim
# the number of axes (dimensions) of the array.
a.ndim
```

```
Out[3]: 2
```

```
In [4]: # This is a tuple of integers indicating the size of the array in each dimension.
a.shape
```



```
Out[4]: (3, 5)
```

```
In [5]: # the total number of elements of the array.
a.size
```

```
Out[5]: 15
```

```
In [6]: # an object describing the type of the elements in the array.
a.dtype
```

```
Out[6]: dtype('int32')
```

```
In [7]: # the size in bytes of each element of the array.
a.itemsize
```

```
Out[7]: 4
```

```
In [8]: # the buffer containing the actual elements of the array
a.data
```

```
Out[8]: <memory at 0x00000150B0EA4E10>
```

```
In [9]: type(a)
```

```
Out[9]: numpy.ndarray
```

```
In [10]: x=np.array([[1,3],[3,4]],dtype=float)
x
```

```
Out[10]: array([[1., 3.],
               [3., 4.]])
```

```
In [11]: x=np.array([[1,3],[3,4]],dtype=complex)
x
```

```
Out[11]: array([[1.+0.j, 3.+0.j],
               [3.+0.j, 4.+0.j]])
```

The function `zeros` creates an array full of zeros, the function `ones` creates an array full of ones, and the function `empty` creates an array whose initial content is random and depends on the state of the memory. By default, the `dtype` of the created array is `float64`, but it can be specified via the key word argument `dtype`.

```
In [12]: np.zeros((3, 4))
```

```
Out[12]: array([[0., 0., 0., 0.],
               [0., 0., 0., 0.],
               [0., 0., 0., 0.]])
```

```
In [13]: np.ones((2,5))
```

```
Out[13]: array([[1., 1., 1., 1., 1.],
               [1., 1., 1., 1., 1.]])
```

```
In [14]: np.empty((2,4))
```

```
Out[14]: array([[4.67296746e-307, 1.69121096e-306, 1.29061142e-306,
                1.89146896e-307],
               [7.56571288e-307, 3.11525958e-307, 1.24610723e-306,
                1.29061142e-306]])
```

To create sequences of numbers, NumPy provides the `arange` function which is analogous to the Python built-in `range`, but returns an array.

```
In [15]: np.arange(10,50,2)
```

```
Out[15]: array([10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38, 40, 42,
                44, 46, 48])
```

Basic Operations

```
In [16]: a=np.array([2,3,4,5,7])  
b=np.ones((1,5))  
print(a-b)
```

```
[[1. 2. 3. 4. 6.]]
```

```
In [17]: a*5
```

```
Out[17]: array([10, 15, 20, 25, 35])
```

```
In [18]: a**2
```

```
Out[18]: array([ 4,  9, 16, 25, 49])
```

```
In [19]: a>3
```

```
Out[19]: array([False, False,  True,  True,  True])
```

```
In [20]: A = np.array([[1, 1],  
                        [0, 1]])  
B = np.array([[2, 0],  
              [3, 4]])
```

```
In [21]: # elementwise product  
A*B
```

```
Out[21]: array([[2, 0],  
               [0, 4]])
```

```
In [22]: # matrix product  
A@B
```

```
Out[22]: array([[5, 4],  
               [3, 4]])
```

```
In [23]: A.dot(B)
```

```
Out[23]: array([[5, 4],  
               [3, 4]])
```

```
In [24]: A.sum()
```

```
Out[24]: 3
```

```
In [25]: A.min()
```

```
Out[25]: 0
```

```
In [26]: A.max()
```

```
Out[26]: 1
```

```
In [27]: A.mean()
```

```
Out[27]: 0.75
```

```
In [28]: A.transpose()
```

```
Out[28]: array([[1, 0],
               [1, 1]])
```

```
In [64]: def f(x,y):
          return x+y;
          x=np.fromfunction(f,(3,3),dtype="int64")
          x
```

```
Out[64]: array([[0, 1, 2],
               [1, 2, 3],
               [2, 3, 4]], dtype=int64)
```

```
In [29]: A.sort()
          A
```

```
Out[29]: array([[1, 1],
               [0, 1]])
```

```
In [54]: A.cumsum() # cumulative sum of elements
```

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

```
In [31]: A.sum(axis=1)
```

```
Out[31]: array([2, 1])
```

```
In [55]: a.cumprod() # cumulative product of elements
```

```
Out[55]: array([ 2,  6, 24, 120, 840])
```

```
In [56]: #np.compress(condition, a, axis=None, out=None)
          np.compress(a>3,a) # a slice along that axis is returned in output for each
```

```
Out[56]: array([4, 5, 7])
```

```
In [57]: # np.extract(condition, arr)
          np.extract(a>2,a) # compress is equivalent to extract.
```

```
Out[57]: array([3, 4, 5, 7])
```

Universal Functions

```
In [32]: s=np.arange(4)
s
```

```
Out[32]: array([0, 1, 2, 3])
```

```
In [33]: np.exp(A)
```

```
Out[33]: array([[2.71828183, 2.71828183],
               [1.          , 2.71828183]])
```

```
In [35]: np.log(B)
```

```
C:\Users\heman\AppData\Local\Temp\ipykernel_27544\4061543246.py:1: RuntimeWarning: divide by zero encountered in log
  np.log(B)
```

```
Out[35]: array([[0.69314718, -inf],
               [1.09861229, 1.38629436]])
```

```
In [37]: np.sqrt(49)
```

```
Out[37]: 7.0
```

Statistics

```
In [38]: A.mean()
```

```
Out[38]: 0.75
```

```
In [39]: A.std() #standard deviation
```

```
Out[39]: 0.4330127018922193
```

```
In [40]: A.var() #variance
```

```
Out[40]: 0.1875
```

```
In [44]: np.cov(A) #covariance
```

```
Out[44]: array([[0. , 0. ],
               [0. , 0.5]])
```

Questions

```
In [47]: A.all()
```

```
Out[47]: False
```

```
In [48]: A.any()
```

```
Out[48]: True
```

```
In [49]: A.nonzero()
```

```
Out[49]: (array([0, 0, 1], dtype=int64), array([0, 1, 1], dtype=int64))
```

```
In [51]: np.where(a>3)
```

```
Out[51]: (array([2, 3, 4], dtype=int64),)
```

Indexing, Slicing

```
In [58]: a[:3]
```

```
Out[58]: array([2, 3, 4])
```

```
In [59]: a[::-1]
```

```
Out[59]: array([7, 5, 4, 3, 2])
```

```
In [60]: w=np.array([[1,2,3],[5,6,7],[9,7,6]])  
w[:,2] #all row and third col
```

```
Out[60]: array([3, 7, 6])
```

Shape

```
In [67]: w.shape
```

```
Out[67]: (3, 3)
```

```
In [69]: w.ravel() # returns the array, flattened
```

```
Out[69]: array([1, 2, 3, 5, 6, 7, 9, 7, 6])
```

```
In [72]: w.reshape(9,1)
```

```
Out[72]: array([[1],  
                [2],  
                [3],  
                [5],  
                [6],  
                [7],  
                [9],  
                [7],  
                [6]])
```

The reshape function returns its argument with a modified shape, whereas the ndarray.resize method modifies the array itself

```
In [75]: w.resize(1,9)
w
```

```
Out[75]: array([[1, 2, 3, 5, 6, 7, 9, 7, 6]])
```