

In [14]:

```
import numpy as np
```

Elementwise Operations

1. Basic Operations

with scalars

In [5]:

```
a = np.array([1, 2, 3, 4]) #create an array
```

```
a + 1
```

Out[5]:

```
array([2, 3, 4, 5])
```

In [6]:

```
a ** 2
```

Out[6]:

```
array([ 1,  4,  9, 16], dtype=int32)
```

All arithmetic operates elementwise

In [8]:

```
b = np.ones(4) + 1 #1,1,1,1
print(b)
print(a)
a - b
```

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

Out[8]:

```
array([-1.,  0.,  1.,  2.])
```

In []:

```
a * b #[1,2,3,4] [2,2,2,2]
```

Out[5]:

```
array([ 2.,  4.,  6.,  8.])
```

In [11]: *# Matrix multiplication*

```
c = np.diag([1, 2, 3, 4])
```

```
print(c * c) #element by element
print("*****")
print(c.dot(c)) #matrix mul
```

```
[[ 1  0  0  0]
 [ 0  4  0  0]
 [ 0  0  9  0]
 [ 0  0  0 16]]
*****
[[ 1  0  0  0]
 [ 0  4  0  0]
 [ 0  0  9  0]
 [ 0  0  0 16]]
```

In [12]: `import numpy as np`
`np.matmul(c,c) #matrix mutiplication`

Out[12]: `array([[1, 0, 0, 0],
 [0, 4, 0, 0],
 [0, 0, 9, 0],
 [0, 0, 0, 16]])`

comparisions

In [13]: `a = np.array([1, 2, 3, 4])`
`b = np.array([5, 2, 2, 4])`
`a == b`

Out[13]: `array([False, True, False, True])`

In []: `a > b`

Out[11]: `array([False, False, True, False], dtype=bool)`

In []: *#array-wise comparisions*
`a = np.array([1, 2, 3, 4])`
`b = np.array([5, 2, 2, 4])`
`c = np.array([1, 2, 3, 4])`

`np.array_equal(a, b)`

Out[12]: `False`

In []: `np.array_equal(a, c)`

Out[13]: `True`

Logical Operations

```
In [ ]: a = np.array([1, 1, 0, 0], dtype=bool) #1=True, 0=False
b = np.array([1, 0, 1, 0], dtype=bool)

np.logical_or(a, b)
```

Out[14]: array([True, True, True, False], dtype=bool)

```
In [ ]: np.logical_and(a, b)
```

Out[15]: array([True, False, False, False], dtype=bool)

Transcendental functions:

```
In [7]: a = np.arange(5)

np.sin(a)
```

Out[7]: array([0. , 0.84147098, 0.90929743, 0.14112001, -0.7568025])

```
In [8]: np.log(a)
```

C:\Users\91920\anaconda3\lib\site-packages\ipykernel_launcher.py:1: RuntimeWarning: divide by zero encountered in log
 """Entry point for launching an IPython kernel.

Out[8]: array([-inf, 0. , 0.69314718, 1.09861229, 1.38629436])

```
In [ ]: np.exp(a) #evaluates e^x for each element in a given input
```

Out[18]: array([1. , 2.71828183, 7.3890561 , 20.08553692, 54.59815003])

Shape Mismatch

```
In [9]: a = np.arange(4)
print(a.shape)

a + np.array([1, 2])

(4,)
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-9-cfac243e1786> in <module>
      2 print(a.shape)
      3
----> 4 a + np.array([1, 2])
```

ValueError: operands could not be broadcast together with shapes (4,) (2,)

Basic Reductions

computing sums

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

Out[4]: 10

```
In [9]: #sum by rows and by columns  
  
x = np.array([[1, 1], [2, 2]])  
x
```

Out[9]: array([[1, 1],
[2, 2]])

```
In [10]: x.sum()
```

Out[10]: 6

```
In [21]: x.sum(axis=0) #columns first dimension
```

Out[21]: array([3, 3])

```
In [ ]: x.sum(axis=1) #rows (second dimension)
```

Out[7]: array([2, 4])

Other reductions

```
In [3]: x = np.array([1, 3, 2])  
x.min()
```

Out[3]: 1

```
In [4]: x.max()
```

Out[4]: 3

```
In [5]: x.argmin() # index of minimum element
```

Out[5]: 0

```
In [13]: y = np.array([[5, 2], [1, 4]])  
print(y.argmin())
```

2

```
In [6]: x.argmax() # index of maximum element
```

Out[6]: 1

Logical Operations

```
In [4]: np.all([True, True, False]) #0, False
```

Out[4]: False

```
In [5]: np.any([True, False, False]) #True
```

Out[5]: True

```
In [23]: #Note: can be used for array comparisions
a = np.zeros((10, 10))#0 a==0
print(a)
np.any(a != 0)
```

```
[[0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]]
```

Out[23]: False

```
In [24]: np.all(a == a)
```

Out[24]: True

```
In [26]: a = np.array([1, 2, 3, 2])
b = np.array([2, 2, 3, 2])
c = np.array([6, 4, 4, 5])
((a <= b) & (b <= c)).all()
```

Out[26]: True

Statistics

```
In [27]: x = np.array([1, 2, 3, 1]) #1,1,2,3
y = np.array([[1, 2, 3], [5, 6, 1]])
x.mean()
y.mean()
```

Out[27]: 3.0

```
In [29]: np.median(x) #middle element
```

Out[29]: 1.5

```
In [ ]: np.median(y, axis= 1) # last axis
```

```
Out[25]: array([ 2.,  5.])
```

```
In [ ]: x.std() # full population standard dev.
```

```
Out[26]: 0.82915619758884995
```

Example:

Data in populations.txt describes the populations of hares and lynxes (and carrots) in northern Canada during 20 years.

```
In [15]: #Load data into numpy array object
data = np.loadtxt('population.txt') #C:/Users/91920/Python/population.txt
```

```
In [5]: '''file=open("population.txt","r")
print(file.read())
file.close()'''
```

```
Out[5]: 'file=open("population.txt","r") \nprint(file.read())\nfile.close()'
```

```
In [6]: data
```

```
Out[6]: array([[ 1900., 30000.,  4000., 48300.],
 [ 1901., 47200.,  6100., 48200.],
 [ 1902., 70200.,  9800., 41500.],
 [ 1903., 77400., 35200., 38200.],
 [ 1904., 36300., 59400., 40600.],
 [ 1905., 20600., 41700., 39800.],
 [ 1906., 18100., 19000., 38600.],
 [ 1907., 21400., 13000., 42300.],
 [ 1908., 22000.,  8300., 44500.],
 [ 1909., 25400.,  9100., 42100.],
 [ 1910., 27100.,  7400., 46000.],
 [ 1911., 40300.,  8000., 46800.],
 [ 1912., 57000., 12300., 43800.],
 [ 1913., 76600., 19500., 40900.],
 [ 1914., 52300., 45700., 39400.],
 [ 1915., 19500., 51100., 39000.],
 [ 1916., 11200., 29700., 36700.],
 [ 1917.,  7600., 15800., 41800.],
 [ 1918., 14600.,  9700., 43300.],
 [ 1919., 16200., 10100., 41300.],
 [ 1920., 24700.,  8600., 47300.]])
```

```
In [20]: x = data[ : ,0:3]
y = data[ : , -1:]
y
```

```
Out[20]: array([[48300.],
               [48200.],
               [41500.],
               [38200.],
               [40600.],
               [39800.],
               [38600.],
               [42300.],
               [44500.],
               [42100.],
               [46000.],
               [46800.],
               [43800.],
               [40900.],
               [39400.],
               [39000.],
               [36700.],
               [41800.],
               [43300.],
               [41300.],
               [47300.]])
```

```
In [24]: y.max()
b=y.argmax()
```

```
In [25]: x[b]
```

```
Out[25]: array([ 1900., 30000.,  4000.])
```

```
In [9]: year, hares, lynxes, carrots = data.T #columns to variables
data.T
```

```
Out[9]: array([[ 1900.,  1901.,  1902.,  1903.,  1904.,  1905.,  1906.,  1907.,
                1908.,  1909.,  1910.,  1911.,  1912.,  1913.,  1914.,  1915.,
                1916.,  1917.,  1918.,  1919.,  1920.],
               [30000., 47200., 70200., 77400., 36300., 20600., 18100., 21400.,
                22000., 25400., 27100., 40300., 57000., 76600., 52300., 19500.,
                11200.,  7600., 14600., 16200., 24700.],
               [ 4000.,  6100.,  9800., 35200., 59400., 41700., 19000., 13000.,
                8300.,  9100.,  7400.,  8000., 12300., 19500., 45700., 51100.,
                29700., 15800.,  9700., 10100.,  8600.],
               [48300., 48200., 41500., 38200., 40600., 39800., 38600., 42300.,
                44500., 42100., 46000., 46800., 43800., 40900., 39400., 39000.,
                36700., 41800., 43300., 41300., 47300.]])
```

```
In [ ]: year, hares, lynxes, carrots = data.T #columns to variables  
print(year)
```

```
[ 1900.  1901.  1902.  1903.  1904.  1905.  1906.  1907.  1908.  1909.  
 1910.  1911.  1912.  1913.  1914.  1915.  1916.  1917.  1918.  1919.  
 1920.]
```

```
In [ ]: #The mean population over time  
populations = data[:, 1:]  
populations
```

```
Out[21]: array([[ 30000.,   4000.,  48300.],  
               [ 47200.,   6100.,  48200.],  
               [ 70200.,   9800.,  41500.],  
               [ 77400.,  35200.,  38200.],  
               [ 36300.,  59400.,  40600.],  
               [ 20600.,  41700.,  39800.],  
               [ 18100.,  19000.,  38600.],  
               [ 21400.,  13000.,  42300.],  
               [ 22000.,   8300.,  44500.],  
               [ 25400.,   9100.,  42100.],  
               [ 27100.,   7400.,  46000.],  
               [ 40300.,   8000.,  46800.],  
               [ 57000.,  12300.,  43800.],  
               [ 76600.,  19500.,  40900.],  
               [ 52300.,  45700.,  39400.],  
               [ 19500.,  51100.,  39000.],  
               [ 11200.,  29700.,  36700.],  
               [   7600.,  15800.,  41800.],  
               [ 14600.,   9700.,  43300.],  
               [ 16200.,  10100.,  41300.],  
               [ 24700.,   8600.,  47300.]])
```

```
In [ ]: #sample standard deviations  
populations.std(axis=0)
```

```
Out[22]: array([ 20897.90645809,  16254.59153691,   3322.50622558])
```

```
In [ ]: #which species has the highest population each year?  
np.argmax(populations, axis=1)
```

```
Out[23]: array([2, 2, 0, 0, 1, 1, 2, 2, 2, 2, 2, 2, 0, 0, 0, 1, 2, 2, 2, 2, 2])
```

Broadcasting

Basic operations on numpy arrays (addition, etc.) are elementwise

This works on arrays of the same size. Nevertheless, It's also possible to do operations on arrays of different sizes if NumPy can transform these arrays so that they all have the same size: this conversion is called broadcasting.

The image below gives an example of broadcasting:



```
In [7]: np.arange(0, 40, 10)
```

```
Out[7]: array([ 0, 10, 20, 30])
```

```
In [8]: np.tile(np.arange(0, 40, 10), (3,1))
```

```
Out[8]: array([[ 0, 10, 20, 30],
               [ 0, 10, 20, 30],
               [ 0, 10, 20, 30]])
```

```
In [ ]: a = np.tile(np.arange(0, 40, 10), (3,1))
        print(a)
```

```
print("*****")
a=a.T
print(a)
```

```
[[ 0 10 20 30]
 [ 0 10 20 30]
 [ 0 10 20 30]]
*****
[[ 0  0  0]
 [10 10 10]
 [20 20 20]
 [30 30 30]]
```

```
In [ ]: b = np.array([0, 1, 2])
        b
```

```
Out[35]: array([0, 1, 2])
```

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

```
Out[36]: array([[ 0,  1,  2],
               [10, 11, 12],
               [20, 21, 22],
               [30, 31, 32]])
```

```
In [ ]: a = np.arange(0, 40, 10)
        a.shape
```

```
Out[45]: (4,)
```

```
In [ ]: a = a[:, np.newaxis] # adds a new axis -> 2D array
a.shape
```

```
Out[42]: (4, 1)
```

```
In [ ]: a
```

```
Out[43]: array([[ 0],
               [10],
               [20],
               [30]])
```


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

```
Out[58]: array([[ 0,  1,  2],
               [10, 11, 12],
               [20, 21, 22],
               [30, 31, 32]])
```

Array Shape Manipulation

Flattening

```
In [10]: a = np.array([[1, 2, 3], [4, 5, 6]])
print(a)
a.ravel() #Return a contiguous flattened array. A 1-D array, containing the elements of the input array.
```



```
[[1 2 3]
 [4 5 6]]
```

```
Out[10]: array([1, 2, 3, 4, 5, 6])
```

```
In [11]: a.T #Transpose
```

```
Out[11]: array([[1, 4],
               [2, 5],
               [3, 6]])
```

```
In [12]: a.T.ravel()
```

```
Out[12]: array([1, 4, 2, 5, 3, 6])
```

Reshaping

The inverse operation to flattening:

```
In [17]: print(a.shape)
print(a)    #4,3 =12 ,,,, 1*12,,,,2*6 ,,3*4
b=a.reshape((6,1))
print(b)
c=a.reshape((3,2))
print(c)
d=a.reshape((4,4))
print(d)
```

```
(2, 3)
[[1 2 3]
 [4 5 6]]
[[1]
 [2]
 [3]
 [4]
 [5]
 [6]]
[[1 2]
 [3 4]
 [5 6]]
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-17-33c50e170589> in <module>
      5 c=a.reshape((3,2))
      6 print(c)
----> 7 d=a.reshape((4,4))
      8 print(d)
```

ValueError: cannot reshape array of size 6 into shape (4,4)

```
In [ ]: b = a.ravel()
print(b)
```

```
[1 2 3 4 5 6]
```

```
In [ ]: b = b.reshape((2, 3))
b
```

```
Out[57]: array([[1, 2, 3],
               [4, 5, 6]])
```

```
In [ ]: b[0, 0] = 100
a
```

```
Out[59]: array([[100,  2,  3],
               [ 4,  5,  6]])
```

Note and Beware: reshape may also return a copy!

```
In [ ]: a = np.zeros((3, 2))
        b = a.T.reshape(3*2)
        b[0] = 50
        a
```

```
Out[60]: array([[ 0.,  0.],
                [ 0.,  0.],
                [ 0.,  0.]])
```

Adding a Dimension

Indexing with the `np.newaxis` object allows us to add an axis to an array

`newaxis` is used to increase the dimension of the existing array by one more dimension, when used once. Thus,

1D array will become 2D array

2D array will become 3D array

3D array will become 4D array and so on

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

```
Out[61]: array([1, 2, 3])
```

```
In [ ]: z[:, np.newaxis]
```

```
Out[62]: array([[1],
                [2],
                [3]])
```

Dimension Shuffling

```
In [ ]: a = np.arange(4*3*2).reshape(4, 3, 2)
        a.shape
```

```
Out[63]: (4, 3, 2)
```

In []: a

```
Out[77]: array([[ 0,  1],
               [ 2,  3],
               [ 4,  5]],

              [[ 6,  7],
               [ 8,  9],
               [10, 11]],

              [[12, 13],
               [14, 15],
               [16, 17]],

              [[18, 19],
               [20, 21],
               [22, 23]])
```

In []: a[0, 2, 1]

Out[64]: 5

Resizing

```
In [3]: a = np.arange(2)
        a.resize((8,))
        a
```

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

However, it must not be referred to somewhere else:

```
In [ ]: b = a
        a.resize((4,))
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-68-702766c88583> in <module>()
      1 b = a
----> 2 a.resize((4,))
```

ValueError: cannot resize an array that references or is referenced by another array in this way. Use the `resize` function

Sorting Data

```
In [ ]: #Sorting along an axis:  
a = np.array([[5, 4, 6], [2, 3, 2]])  
b = np.sort(a, axis=1)  
b
```

```
Out[69]: array([[4, 5, 6],  
               [2, 2, 3]])
```

```
In [ ]: #in-place sort  
a.sort(axis=1)  
a
```

```
Out[84]: array([[4, 5, 6],  
               [2, 2, 3]])
```

```
In [ ]: #sorting with fancy indexing  
a = np.array([4, 3, 1, 2])  
j = np.argsort(a)  
j
```

```
Out[85]: array([2, 3, 1, 0])
```

```
In [ ]: a[j]
```

```
Out[86]: array([1, 2, 3, 4])
```