

In [5]:

```
import numpy as np
```

In [6]:

```
a= np.array([1,2,3])  
b= np.array([10,20,30])
```

In [7]:

```
print(a+b)
```

```
[11 22 33]
```

In [8]:

```
print(b**a)
```

```
[  10   400 27000]
```

In [9]:

```
a.shape
```

Out[9]:

```
(3,)
```

In [10]:

```
a.size
```

Out[10]:

```
3
```

In [11]:

```
a.ndim
```

Out[11]:

```
1
```

In [12]:

```
a+b
```

Out[12]:

```
array([11, 22, 33])
```

here in the background some vectorized operations are going on. `a[0]` is also replaced by `getitem(0)`. so this square brackets are also called **syntactic sugar** because it gives us simple and sweet way to write down the code.

In [13]:

```
a
```

Out[13]:

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

In [14]:

```
a*10
```

Out[14]:

```
Out[14]:  
array([10, 20, 30])
```

```
In [15]:
```

```
a
```

```
Out[15]:  
array([1, 2, 3])
```

```
In [16]:
```

```
#universal functions are also called ufuncs
```

```
In [17]:
```

```
a.fill(10.1)  
a
```

```
Out[17]:  
array([10, 10, 10])
```

```
In [18]:
```

```
type(a)
```

```
Out[18]:  
numpy.ndarray
```

```
In [19]:
```

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

```
Out[19]:  
array([[1, 2, 3, 4],  
       [1, 2, 3, 4]])
```

```
In [20]:
```

```
a.size
```

```
Out[20]:  
8
```

```
In [21]:
```

```
a.shape
```

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

```
In [22]:
```

```
a.ndim
```

```
Out[22]:  
2
```

```
In [23]:
```

```
a[1]
```

```
Out[23]:  
array([1, 2, 3, 4])
```

In [24]:

```
a[0]
```

Out[24]:

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

In [25]:

```
a[0,3]=100
```

In [26]:

```
a
```

Out[26]:

```
array([[ 1,  2,  3, 100],
       [ 1,  2,  3,  4]])
```

In [27]:

```
a[1, 2:4]
```

Out[27]:

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

In [28]:

```
a[:,3]
```

Out[28]:

```
array([100,  4])
```

In [29]:

```
a= np.arange(25).reshape(5,5)
```

In [30]:

```
a
```

Out[30]:

```
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, 24]])
```

In [31]:

```
timeit(a[1:4:2,:3:2])
```

1.1 μ s \pm 77.2 ns per loop (mean \pm std. dev. of 7 runs, 1000000 loops each)

In [32]:

```
a=np.array([12,0,-1,3,5,-5,-6])
a<0
```

Out[32]:

```
array([False, False,  True, False, False,  True,  True])
```

In [33]:

```
mask=np.array([0,1,0,1,0,1,0], dtype=bool)
y= a[mask]
print(y)
```



```
[nan, nan, nan, nan, nan]])
```

```
In [69]:
```

```
a= np.ones((5,6))
```

```
In [72]:
```

```
timeit a.flatten()
```

```
1.77 µs ± 90.2 ns per loop (mean ± std. dev. of 7 runs, 1000000 loops each)
```

```
In [73]:
```

```
timeit a.ravel()
```

```
404 ns ± 84.1 ns per loop (mean ± std. dev. of 7 runs, 1000000 loops each)
```

```
In [61]:
```

```
a.sum(axis=0)
```

```
Out[61]:
```

```
array([5., 5., 5., 5., 5., 5.])
```

```
In [ ]:
```

how to find location of values using where functions, flatten(safe method) , ravel (efficient)

```
In [66]:
```

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

```
In [68]:
```

```
a.argmin()
```

```
Out[68]:
```

```
0
```