```
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]:
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]:
а
Out[13]:
array([1, 2, 3])
In [14]:
```

a*10

O11+ [1 / 1 .

```
Out[14]:
array([10, 20, 30])
In [15]:
Out[15]:
array([1, 2, 3])
In [16]:
#universal functions are also called ufuncs
In [17]:
a.fill(10.1)
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]])
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]:
Out[26]:
array([[ 1, 2, 3, 100],
             2,
                  3,
      [ 1,
                       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]:
а
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 \mus \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)
```

```
[0 \ 3 \ -5]
In [34]:
a = np.arange(55).reshape(11,5)
In [35]:
а
Out[35]:
array([[ 0, 1, 2,
                     3,
       [5, 6, 7, 8,
                          9],
       [10, 11, 12, 13, 14],
       [15, 16, 17, 18, 19],
       [20, 21, 22, 23, 24],
       [25, 26, 27, 28, 29],
       [30, 31, 32, 33, 34],
       [35, 36, 37, 38, 39],
       [40, 41, 42, 43, 44],
       [45, 46, 47, 48, 49],
       [50, 51, 52, 53, 54]])
In [36]:
a[[0,1,2,3],[0,2,3,4]]
Out[36]:
array([ 0, 7, 13, 19])
In [37]:
a[3:,[1,3,4]]
Out[37]:
array([[16, 18, 19],
       [21, 23, 24],
[26, 28, 29],
       [31, 33, 34],
       [36, 38, 39],
       [41, 43, 44],
       [46, 48, 49],
       [51, 53, 54]])
In [39]:
mask=np.array([1,0,1,0,1,0,1,1,1,1],dtype=bool)
a[mask,2]
Out[39]:
array([ 2, 12, 22, 32, 37, 42, 47, 52])
In [43]:
a*np.nan #not a number so we use if there is a value which is not a number and want get r
id of it.
Out[43]:
array([[nan, nan, nan, nan, nan],
       [nan, nan, nan, nan, nan],
```

```
[nan, nan, nan, nan]])
In [69]:
a= np.ones((5,6))
In [72]:
timeit a.flatten()
1.77 \mu s \pm 90.2 ns per loop (mean \pm std. dev. of 7 runs, 1000000 loops each)
In [73]:
timeit a.ravel()
404 ns \pm 84.1 ns per loop (mean \pm 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
```