Basics of NumPy Arrays - 2

```
In [2]:
import numpy as np
Array Indexing
Similar to list indexing, arrays can be accessed using the indices.
In [3]:
nparray = np.array([1, 2, 3, 4, 5])
In [5]:
#Access first element, indexing starts from 0
nparray[0]
Out[5]:
In [6]:
#Access second element, indexing starts from 0
nparray[1]
Out[6]:
2
In [7]:
#Access last element, negative indices works
nparray[-1]
Out[7]:
5
In [8]:
```

In multi-dimensional arrays, comma separated tupe of indices can be used to access elements.

nparray[-2]

Out[8]:

4

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In [12]:
x = np.array([[1,2,3], [4, 5, 6], [7, 8, 9]])
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Out[12]:
array([[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
In [10]:
#Access element at location (0, 0), zeroth row and zeroth column
x[0,0]
Out[10]:
1
In [11]:
#Access element at location (1, 2), first row and second column
x[1,2]
Out[11]:
Values can be modified using indices,
In [13]:
#Modify the (0, 0)th entry to 555
x[0, 0] = 555
Out[13]:
array([[555, 2, 3],
       [ 4, 5, 6],
[ 7, 8, 9]])
Array Slicing
[] can be used access the sub arrays i.e. array sicing is supported similar to list. To access a slice of array x,
x[start:stop:step]
In [40]:
nparray
Out[40]:
array([1, 2, 3, 4, 5])
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In [43]:
nparray[1:] # all values after index 1
Out[43]:
array([2, 3, 4, 5])
In [48]:
nparray[ : 3] # all values till index 3
Out[48]:
array([1, 2, 3])
In [49]:
nparray[2:4] # all value between index 2 and 4
Out[49]:
array([3, 4])
In [47]:
nparray[::2] #every other element
Out[47]:
array([1, 3, 5])
In [51]:
nparray[1::2] #every other element starting from element at index 1
Out[51]:
array([2, 4])
In [52]:
nparray[::-1] # all elements in reversed direction
Out[52]:
array([5, 4, 3, 2, 1])
In [53]:
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Out[53]:
array([[555,
               2,
                    3],
       [ 4, 5,
                    6],
                    9]])
       [ 7,
```

```
In [55]:
x[:2,:3] # first two rows, first three columns
Out[55]:
array([[555, 2, 3],
       [ 4, 5, 6]])
In [57]:
#Accessing array rows
x[0, :] # zeroth row , all columns
Out[57]:
array([555, 2, 3])
In [59]:
#Accessing array columns
x[:, 1] # all row , second columns
Out[59]:
array([2, 5, 8])
Array copying
Array slices are views, hence if they are modified then original array is changed.
In [60]:
nparray
Out[60]:
array([1, 2, 3, 4, 5])
In [63]:
nparray_sub = nparray[2:4] #Slice the array
nparray_sub
Out[63]:
array([3, 4])
In [64]:
nparray_sub[0] = 5 # modify the element of slice
nparray_sub
Out[64]:
array([5, 4])
```

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In [66]:
nparray #original array is changed
Out[66]:
array([1, 2, 5, 4, 5])
In order to prevent this behaviour copy needs to be done.
In [70]:
nparray_copy = nparray[2:4].copy() #create copy of array slice
In [74]:
nparray_copy[0] = 90 # modify the slice
nparray_copy
Out[74]:
array([90, 4])
In [76]:
nparray #original array is still intact
Out[76]:
array([1, 2, 5, 4, 5])
Reshaping of Arrays
Reshape can be used to change the structure of array. Size of original and reshaped arrays should be same.
In [80]:
a = [1, 2, 3, 4, 5, 6, 7, 8, 9]
array = np.array(a)
array
Out[80]:
array([1, 2, 3, 4, 5, 6, 7, 8, 9])
In [81]:
array.reshape(3, 3)
Out[81]:
array([[1, 2, 3],
       [4, 5, 6],
```

[7, 8, 9]])

Reshaping can be used to convert an array into row matrix or column matrix.

array([1, 2, 3, 10, 20, 30, 99, 99, 99])

```
In [77]:
Х
Out[77]:
array([[555, 2,
                    3],
               5,
                    6],
       [ 4,
               8,
       [ 7,
                    9]])
In [78]:
x.reshape(9, 1)
Out[78]:
array([[555],
       [ 2],
          3],
       [
         4],
       [
         5],
       [
         6],
       [
         7],
       [
         8],
       [ 9]])
In [79]:
x.reshape(1, 9)
Out[79]:
array([[555, 2,
                    3, 4, 5, 6, 7, 8, 9]])
Array Concatenation
Joining of two arrays is possible through np.concatenate, np.vstack and np.hstack.
In [84]:
x = np.array([1, 2, 3])
y = np.array([10, 20, 30])
z = np.array([99, 99, 99])
In [86]:
np.concatenate([x, y, z])
Out[86]:
```

```
In [89]:
two_d_array = np.array([[1, 2, 3], [4, 5,6]])
two_d_array
Out[89]:
array([[1, 2, 3],
       [4, 5, 6]]
In [90]:
np.concatenate([two_d_array, two_d_array])
Out[90]:
array([[1, 2, 3],
       [4, 5, 6],
       [1, 2, 3],
       [4, 5, 6]])
In [92]:
np.vstack([two_d_array, y])
Out[92]:
array([[ 1, 2, 3],
       [4, 5, 6],
       [10, 20, 30]])
In [95]:
v_{array} = np.array([[99], [99]])
In [97]:
np.hstack([two_d_array, v_array])
Out[97]:
array([[ 1, 2, 3, 99],
       [4, 5, 6, 99]])
Array Splitting
Splititng is done by split, vsplit and hsplit.
In [98]:
x = np.array([1, 2, 3, 4, 5, 6, 7, 8, 9, 10])
In [101]:
n1, n2, n3, n4 = np.split(x, [2, 5, 8]) # splitting position 2, 5, and 8
```

```
In [102]:
print(n1, n2, n3, n4)
[1 2] [3 4 5] [6 7 8] [ 9 10]
In [104]:
grid = np.arange(5, 20).reshape(5,3)
grid
Out[104]:
array([[ 5, 6, 7],
       [ 8, 9, 10],
       [11, 12, 13],
       [14, 15, 16],
       [17, 18, 19]])
In [110]:
n1, n2 = np.split(grid, [3]) #splitting postion third row
In [109]:
print(n1, "\n\n", n2)
[[5 6 7]
[8 9 10]
 [11 12 13]]
 [[14 15 16]
 [17 18 19]]
In [112]:
n1, n2 = np.vsplit(grid, [1])
In [113]:
print(n1, "\n\n", n2)
[[5 6 7]]
 [[ 8 9 10]
 [11 12 13]
 [14 15 16]
 [17 18 19]]
In [114]:
n1, n2 = np.hsplit(grid, [1])
```

```
print(n1, "\n\n", n2)
[[ 5]
 [ 8]
 [11]
 [14]
 [17]]
 [[67]
 [ 9 10]
 [12 13]
 [15 16]
 [18 19]]
Fancy Indexing
Here arrays of indices is passed to access the elements.
In [16]:
nparray
Out[16]:
array([1, 2, 3, 4, 5])
In [21]:
indices = [1, 3, 4] #indices first, third and fourth
nparray[indices] #elements at first , third and forth elements are returned
Out[21]:
array([2, 4, 5])
In [38]:
#Create a 2-d array using the indices
indices = np.array([[0,1], [3,4]])
nparray[indices]
Out[38]:
array([[1, 2],
       [4, 5]])
In [20]:
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Out[20]:
array([[555,
                    3],
               2,
               5,
                    6],
       [ 4,
```

In [115]:

7,

9]])

```
In [36]:

row = np.array([0, 1])

column = np.array([0,1])

In [39]:

x[row, column] #(0,0), (1, 1) are returned

Out[39]:
array([555, 5])
```