

1.3-NumPy Indexing and Selection

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💡💡💡💡 NumPy Indexing and Selection 💡💡💡💡

In this notebook we will discuss how to select elements or groups of elements from an array.

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

```
In [2]: #Creating sample array
arr = np.arange(0,11)
```

```
In [3]: #Show
arr
```

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

1 Bracket Indexing and Selection

The simplest way to pick one or some elements of an array looks very similar to python lists:

```
In [4]: #Get a value at an index
arr[8]
```

```
Out[4]: 8
```

```
In [5]: #Get values in a range
arr[1:5]
```

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

```
In [6]: #Get values in a range  
arr[0:5]
```

```
Out[6]: array([0, 1, 2, 3, 4])
```

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2 Broadcasting

Numpy arrays differ from a normal Python list because of their ability to broadcast:

```
In [7]: #Setting a value with index range (Broadcasting)  
arr[0:5]=100  
  
#Show  
arr
```

```
Out[7]: array([100, 100, 100, 100, 100,  5,  6,  7,  8,  9, 10])
```

```
In [8]: # Reset array, we'll see why I had to reset in a moment  
arr = np.arange(0,11)  
  
#Show  
arr
```

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

```
In [9]: #Important notes on Slices  
slice_of_arr = arr[0:6]  
  
#Show slice  
slice_of_arr
```

```
Out[9]: array([0, 1, 2, 3, 4, 5])
```

```
In [10]: #Change Slice  
slice_of_arr[:]=99  
  
#Show Slice again  
slice_of_arr
```

```
Out[10]: array([99, 99, 99, 99, 99, 99])
```

Now note the changes also occur in our original array!

```
In [11]: arr
```

```
Out[11]: array([99, 99, 99, 99, 99, 99,  6,  7,  8,  9, 10])
```

Data is not copied, it's a view of the original array! This avoids memory problems!

```
In [12]: #To get a copy, need to be explicit  
arr_copy = arr.copy()  
  
arr_copy
```

```
Out[12]: array([99, 99, 99, 99, 99, 99,  6,  7,  8,  9, 10])
```

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3 Indexing a 2D array (matrices)

The general format is **arr_2d[row][col]** or **arr_2d[row,col]**. I recommend usually using the comma notation for clarity.

```
In [13]: arr_2d = np.array([[5,10,15],[20,25,30],[35,40,45]])  
  
#Show  
arr_2d
```

```
Out[13]: array([[ 5, 10, 15],  
               [20, 25, 30],  
               [35, 40, 45]])
```

```
In [14]: #Indexing row  
arr_2d[1]
```

```
Out[14]: array([20, 25, 30])
```

```
In [15]: # Format is arr_2d[row][col] or arr_2d[row,col]  
  
# Getting individual element value  
arr_2d[1][0]
```

```
Out[15]: 20
```

```
In [16]: # Getting individual element value  
arr_2d[1,0]
```

```
Out[16]: 20
```

```
In [17]: # 2D array slicing  
  
#Shape (2,2) from top right corner  
arr_2d[:2,1:]
```

```
Out[17]: array([[10, 15],  
               [25, 30]])
```

```
In [18]:
```

```
#Shape bottom row  
arr_2d[2]
```

Out[18]: array([35, 40, 45])

```
In [19]: #Shape bottom row  
arr_2d[2,:]
```

Out[19]: array([35, 40, 45])

3.1 Fancy Indexing

Fancy indexing allows you to select entire rows or columns out of order, to show this, let's quickly build out a numpy array:

```
In [20]: #Set up matrix  
arr2d = np.zeros((10,10))
```

```
In [21]: #Length of array  
arr_length = arr2d.shape[1]
```

```
In [22]: #Set up array  
  
for i in range(arr_length):  
    arr2d[i] = i  
  
arr2d
```

```
Out[22]: array([[0., 0., 0., 0., 0., 0., 0., 0., 0., 0.],  
                [1., 1., 1., 1., 1., 1., 1., 1., 1., 1.],  
                [2., 2., 2., 2., 2., 2., 2., 2., 2., 2.],  
                [3., 3., 3., 3., 3., 3., 3., 3., 3., 3.],  
                [4., 4., 4., 4., 4., 4., 4., 4., 4., 4.],  
                [5., 5., 5., 5., 5., 5., 5., 5., 5., 5.],  
                [6., 6., 6., 6., 6., 6., 6., 6., 6., 6.],  
                [7., 7., 7., 7., 7., 7., 7., 7., 7., 7.],  
                [8., 8., 8., 8., 8., 8., 8., 8., 8., 8.],  
                [9., 9., 9., 9., 9., 9., 9., 9., 9., 9.]])
```

Fancy indexing allows the following

```
In [23]: arr2d[[2,4,6,8]]
```

```
Out[23]: array([[2., 2., 2., 2., 2., 2., 2., 2., 2., 2.],  
                [4., 4., 4., 4., 4., 4., 4., 4., 4., 4.],  
                [6., 6., 6., 6., 6., 6., 6., 6., 6., 6.],  
                [8., 8., 8., 8., 8., 8., 8., 8., 8., 8.]])
```

```
In [24]: #Allows in any order  
arr2d[[6,4,2,7]]
```

```
array([[6., 6., 6., 6., 6., 6., 6., 6., 6., 6.],
```

```
Out[24]:      [4., 4., 4., 4., 4., 4., 4., 4., 4., 4.],  
           [2., 2., 2., 2., 2., 2., 2., 2., 2., 2.],  
           [7., 7., 7., 7., 7., 7., 7., 7., 7., 7.]])
```

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4 Selection

Let's briefly go over how to use brackets for selection based off of comparison operators.

```
In [25]: arr = np.arange(1,11)  
arr
```

```
Out[25]: array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10])
```

```
In [26]: arr > 4
```

```
Out[26]: array([False, False, False, False,  True,  True,  True,  True,  True,  
                True])
```

```
In [27]: bool_arr = arr>4
```

```
In [28]: bool_arr
```

```
Out[28]: array([False, False, False, False,  True,  True,  True,  True,  True,  
                True])
```

```
In [29]: arr[bool_arr]
```

```
Out[29]: array([ 5,  6,  7,  8,  9, 10])
```

```
In [30]: arr[arr>2]
```

```
Out[30]: array([ 3,  4,  5,  6,  7,  8,  9, 10])
```

```
In [31]: x = 2  
arr[arr>x]
```

```
Out[31]: array([ 3,  4,  5,  6,  7,  8,  9, 10])
```

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Great Job!
