

NumPy Indexing and Selection

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

```
In [2]: 1 import numpy as np
```

```
In [3]: 1 #Creating sample array
        2 arr = np.arange(0,11)
```

```
In [4]: 1 #Show
        2 arr
```

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

Bracket Indexing and Selection

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

```
In [5]: 1 #Get a value at an index
        2 arr[8]
```

```
Out[5]: 8
```

```
In [6]: 1 #Get values in a range
        2 arr[1:5]
```

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

```
In [7]: 1 #Get values in a range
        2 arr[0:5]
```

```
Out[7]: array([0, 1, 2, 3, 4])
```

Broadcasting

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

```
In [8]: 1 #Setting a value with index range (Broadcasting)
        2 arr[0:5]=100
        3
        4 #Show
        5 arr
```

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

```
In [9]: 1 # Reset array, we'll see why I had to reset in a moment
        2 arr = np.arange(0,11)
        3
        4 #Show
        5 arr
```

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

```
In [10]: 1 #Important notes on Slices
        2 slice_of_arr = arr[0:6]
        3
        4 #Show slice
        5 slice_of_arr
```

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

```
In [11]: 1 #Change Slice
        2 slice_of_arr[:]=99
        3
        4 #Show Slice again
        5 slice_of_arr
```

```
Out[11]: array([99, 99, 99, 99, 99, 99])
```

Now note the changes also occur in our original array!

```
In [12]: 1 arr
```

```
Out[12]: 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 [13]: 1 #To get a copy, need to be explicit
        2 arr_copy = arr.copy()
        3
        4 arr_copy
```

```
Out[13]: array([99, 99, 99, 99, 99, 99,  6,  7,  8,  9, 10])
```

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 [14]: 1 arr_2d = np.array([[5,10,15],[20,25,30],[35,40,45]])
          2
          3 #Show
          4 arr_2d
```

```
Out[14]: array([[ 5, 10, 15],
                [20, 25, 30],
                [35, 40, 45]])
```

```
In [15]: 1 #Indexing row
          2 arr_2d[1]
          3
```

```
Out[15]: array([20, 25, 30])
```

```
In [16]: 1 # Format is arr_2d[row][col] or arr_2d[row,col]
          2
          3 # Getting individual element value
          4 arr_2d[1][0]
```

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

```
In [17]: 1 # Getting individual element value
          2 arr_2d[1,0]
```

```
Out[17]: 20
```

```
In [18]: 1 # 2D array slicing
          2
          3 #Shape (2,2) from top right corner
          4 arr_2d[:2,1:]
```

```
Out[18]: array([[10, 15],
                [25, 30]])
```

```
In [19]: 1 #Shape bottom row
          2 arr_2d[2]
```

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

```
In [20]: 1 #Shape bottom row
          2 arr_2d[2,:]
```

```
Out[20]: array([35, 40, 45])
```

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 [21]: 1 #Set up matrix
         2 arr2d = np.zeros((10,10))
```

```
In [22]: 1 #Length of array
         2 arr_length = arr2d.shape[1]
```

```
In [23]: 1 #Set up array
         2
         3 for i in range(arr_length):
         4     arr2d[i] = i
         5
         6 arr2d
```

```
Out[23]: 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 [24]: 1 arr2d[[2,4,6,8]]
```

```
Out[24]: 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 [25]: 1 #Allows in any order
         2 arr2d[[6,4,2,7]]
```

```
Out[25]: array([[ 6.,  6.,  6.,  6.,  6.,  6.,  6.,  6.,  6.,  6.],
                [ 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.]])
```

More Indexing Help

Indexing a 2d matrix can be a bit confusing at first, especially when you start to add in step size. Try google image searching NumPy indexing to find useful images, like this one:

Selection

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

```
In [28]: 1 arr = np.arange(1,11)
        2 arr
```

```
Out[28]: array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10])
```

```
In [30]: 1 arr > 4
```

```
Out[30]: array([False, False, False, False,  True,  True,  True,  True,  True,  True], dtype=bool)
```

```
In [31]: 1 bool_arr = arr>4
```

```
In [32]: 1 bool_arr
```

```
Out[32]: array([False, False, False, False,  True,  True,  True,  True,  True,  True], dtype=bool)
```

```
In [33]: 1 arr[bool_arr]
```

```
Out[33]: array([ 5,  6,  7,  8,  9, 10])
```

```
In [34]: 1 arr[arr>2]
```

```
Out[34]: array([ 3,  4,  5,  6,  7,  8,  9, 10])
```

```
In [37]: 1 x = 2
        2 arr[arr>x]
```

```
Out[37]: array([ 3,  4,  5,  6,  7,  8,  9, 10])
```