

# Numerical Python - NumPy Package

## Computer Programming for Data Science 7CCSMCMP

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## NumPy

### ▶ **Part 1:**

- ▶ **Overview of NumPy**
- ▶ **Creating an array**
- ▶ **Indexing and slicing arrays**

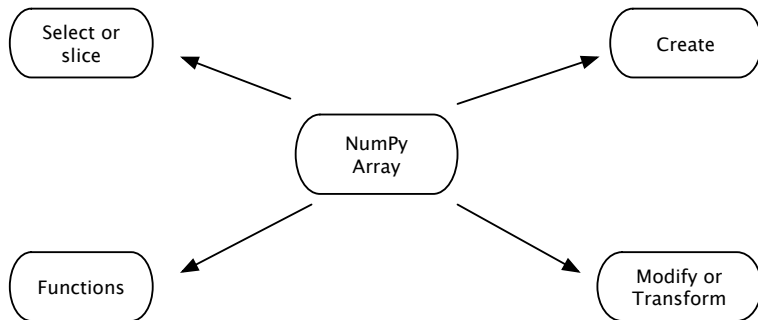
### ▶ **Part 2:**

- ▶ Operations on arrays
- ▶ Sorting arrays
- ▶ Writing and reading arrays from and to disk
- ▶ Generating random number arrays
- ▶ Histogram in NumPy

# NumPy Overview

- ▶ Fast multidimensional arrays
- ▶ Speed of C and developer friendliness of Python
- ▶ Why are arrays needed?
  - ▶ Handle large sequences of data
  - ▶ To process many data points at the same time - matrix maths and regression for example
  - ▶ Numpy handled arrays faster than standard python lists and tuples
- ▶ in order to use the library: `import numpy as np`

# NumPy Highlights



# About arrays

An array is the central structure of NumPy

- ▶ Each element is of the same type
- ▶ The rank of an array is the number of dimensions
- ▶ The length of each dimension do not have to be the same

```
1 array([[1, 2, 3, 4],  
2        [5, 6, 7, 8]])
```



## Creating an array

## Creating an array with lists:

```
1 data1=[6,7.5,8,0,1]
2 arr1=np.array(data1)
3 arr1
```

Results in this array:

```
1 array([ 6. ,  7.5,  8. ,  0. ,  1. ])
```

An array of zeros can be created using:

```
1 np.zeros(8)
```

This results in:

```
1 array([ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.]
```

# Creating an array (contd...)

There are more ways of creating an array:

```
1 np.ones(4)
2 np.arange(15)
```

What will these generate?



# Creating an array (contd...)

There are more ways of creating an array:

```
1 np.ones(4)
2 np.arange(15)
```

These will generate the following arrays:

```
1 array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11,
        12, 13, 14])
2 array([ 1.,  1.,  1.,  1.] )
```

# Using recursion to populate an array

Populating an array using recursion:

```
In [18]: arr2=np.empty((10,4))  
         for i in range(10):  
             arr2[i]=i
```

```
In [19]: arr2
```

```
Out[19]: array([[ 0.,  0.,  0.,  0.],  
                [ 1.,  1.,  1.,  1.],  
                [ 2.,  2.,  2.,  2.],  
                [ 3.,  3.,  3.,  3.],  
                [ 4.,  4.,  4.,  4.],  
                [ 5.,  5.,  5.,  5.],  
                [ 6.,  6.,  6.,  6.],  
                [ 7.,  7.,  7.,  7.],  
                [ 8.,  8.,  8.,  8.],  
                [ 9.,  9.,  9.,  9.]])
```

# Multidimensional Arrays

Arrays can have more than just one dimension, for example:

```
1 data2 = [[1, 2, 3, 4], [5, 6, 7, 8]]
2 arr2=np.array(data2)
3 arr2
```

arr2 will look like:

```
1 array([[1, 2, 3, 4],
2        [5, 6, 7, 8]])
```

Using size and shape:

```
1 arr2.shape
2 (2, 4)
3
4 arr2.size
5 8
```

# Array Data Type

The data type of the array:

```
1 arr=np.arange(8)
2 arr.dtype
```

Will return in this case:

```
1 dtype('int64')
```

Specifying the type when generating the array

```
1 arr2 = np.array([1, 2, 3], dtype=np.int32)
2 arr2.dtype
```

will result in this:

```
1 dtype('int32')
```

# Array Indexing and Slicing

Select a subset or individual elements from a one dimensional array:

```
1 arr = np.arange(10)
2 arr
3 array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
4 arr[5]
5 arr[5:8]
```

What do the selections above return?

# Array Indexing and Slicing contd.

The results of the indexing from the previous slide:

```
In [4]: arr = np.arange(10)
        arr
```

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

```
In [5]: arr[5]
```

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

```
In [6]: arr[5:8]
```

```
Out[6]: array([5, 6, 7])
```

# Array Indexing and Slicing

Assign a value to a slice:

```
1 arr[5:8] = 12
2 arr
3 array([ 0, 1, 2, 3, 4, 12, 12, 12, 8, 9])
```

A slice is a view of the array, so any changes made to a slice are propagated to the source array.

# Indexing and Slicing Multidimensional arrays

Lets look at an array created as follows:

```
1 arr2d = np.array([[1, 2, 3,4], [5, 6, 7,8], [9, 10,  
    11,12]])
```

Axis 1 Columns					
		1	2	3	4
Axis 0 rows		5	6	7	8
		9	10	11	12



# A two dimensional array

The positions of each element in the two dimensional array are (row,column):

	Axis 1 Columns			
	1 (0,0)	2 (0,1)	3 (0,2)	4 (0,3)
Axis 0 rows	5 (1,0)	6 (1,1)	7 (1,2)	8 (1,3)
	9 (2,0)	10 (2,1)	11 (2,2)	12 (2,3)

Individual elements can be accessed:

```
1 arr2d [0 , 2]  
2 3
```

# Two dimensional array indexing and slicing example

1	2	3	4
5	6	7	8
9	10	11	12

```
1 arr2d [2 , :]
```

1	2	3	4
5	6	7	8
9	10	11	12

This results in this shaded area of the array:

# Two dimensional array indexing and slicing example

1	2	3	4
5	6	7	8
9	10	11	12

```
1 arr2d[:, :2]
```

What will this return?

1. First 2 columns, all rows
2. Last 2 rows, all columns
3. Last 2 columns, all rows

# Two dimensional array indexing and slicing example

1	2	3	4
5	6	7	8
9	10	11	12

```
1 arr2d[:, :2]
```

1	2	3	4
5	6	7	8
9	10	11	12

Answer: First two columns, all rows.

## 3D arrays

Assuming a 3D array is generated:

```
1 arr3d = np.array([[[1, 2, 3], [4, 5, 6]], [[7, 8, 9],  
            [10, 11, 12]]])
```

[1,2,3]	[4,5,6]
[7,8,9]	[10,11,12]

The shape of this array is  $(2, 2, 3)$  i.e.  $2 * 2 * 3$ .

## 3D arrays

In order to select the following from the 3D array:

[1,2,3]	[4,5,6]
[7,8,9]	[10,11,12]

Then refer to it:

```
1 arr3d [1 , 1 , ::]  
2 arr3d [1 , 1 , ]
```

## 3D arrays

If later indices are omitted, the returned object will be a lower-dimensional array consisting of all the data along the higher dimensions.

```
1 arr3d [0]
```

Will result in a 2\*3 array:

```
1 array([[1, 2, 3],  
2        [4, 5, 6]])
```

# NumPy - Summary of Part 1

- ▶ Overview of Numpy
- ▶ Methods to create an array: manual inputs, using lists, using zeros, ones, random and recursion.
- ▶ Attributes of arrays: shape, number of dimensions and data types
- ▶ Indexing and slicing: selections, 2d and 3d