

# An Introduction to Numpy

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# NumPy Fundamentals

- Numpy is a module for faster vector processing with numerous other routines
- Scipy is a more extensive module that also includes many other functionalities such as machine learning and statistics
  -

# NumPy Fundamentals

- Why Numpy?
  - Remember that Python does not limit lists to just elements of a single class
  - If we have a large list  $[a_1, a_2, a_3, \dots, a_n]$  and we want to add a number to all of the elements, then Python will ask for each element:
    - What is the type of the element
    - Does the type support the + operation
    - Look up the code for the + and execute
  - This is slow

# NumPy Fundamentals

- Why Numpy?
  - Primary feature of Numpy are arrays:
    - List like structure where all the elements have the same type
      - Usually a floating point type
    - Can calculate with arrays much faster than with list
    - Implemented in C / Java for Cython or Jython

# NumPy Fundamentals

- How to get Numpy?
  - Get the Anaconda distribution
    - Comes these days with all sorts of goodies
      - No need to install numpy, but could with
        - `conda numpy`
  - I still want to use Idle, so I install components individually
    - Use `pip3` install numpy
      - Be careful, some OS come with a Python 2.7 version
        - Do not update those

# NumPy Arrays

- Numpy arrays have dimensions
  - Vectors: one-dimensional
  - Matrices: two-dimensional
  - Tensors: more dimensions, but much more rarely used
- Nota bene: A matrix can have a single row and a single column, but has still two dimensions

# NumPy Arrays

- After installing, try out `import numpy as np`
- Making arrays:
  - Can use lists, though they better be of the same type

```
import numpy as np
my_list = [1,5,4,2]
my_vec = np.array(my_list)
my_list = [[1,2],[4,3]]
my_mat = np.array(my_list)
```

# NumPy Arrays

- Use `np.arange`
  - Similar to the `range` function

```
np.arange(start, stop, step)
```

- Example:

```
print(np.arange(0,10))
```

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



# NumPy Arrays

- Other generation methods:
  - `np.zeros`
    - Takes a number or a tuple of numbers
    - Fills in a tensor with zeroes
    - default datatype is a 'float'

```
>>> np.zeros((3,3), dtype='int')
```

```
array([[0, 0, 0],  
       [0, 0, 0],  
       [0, 0, 0]])
```

# NumPy Arrays

- Similarly `np.ones`

```
>>> np.ones((3,4))  
array([[1., 1., 1., 1.],  
       [1., 1., 1., 1.],  
       [1., 1., 1., 1.]])
```

# NumPy Arrays

- Creating arrays:
  - `np.full` to fill in with a given value

```
np.full(5, 3.141)
```

```
array([3.141, 3.141, 3.141, 3.141, 3.141])
```

# NumPy Arrays

- Creating arrays:
  - Use linspace to evenly space between two values:
    - `np.linspace(start, end, number)`

```
>>> np.linspace(0, 2, 5)
```

```
array([0. , 0.5, 1. , 1.5, 2. ])
```

# NumPy Arrays

- Can also use random values.
  - Uniform distribution between 0 and 1

```
>>> np.random.random( (3,2) )
array([[0.39211415, 0.50264835],
       [0.95824337, 0.58949256],
       [0.59318281, 0.05752833]])
```

# NumPy Arrays

- Or random integers

```
>>> np.random.randint(0,20,(2,4))
```

```
array([[ 5,  7,  2, 10],  
       [19,  7,  1, 10]])
```

# NumPy Arrays

- Or other distributions, e.g. normal distribution with mean 2 and standard deviation 0.5

```
>>> np.random.normal(2, 0.5, (2, 3))  
array([[1.34857621, 1.34419178, 1.977698   ],  
       [1.31054068, 2.35126538, 3.25903903]])
```

# NumPy Arrays

- There is a special notation for the identity matrix /

```
>>> np.eye(4)
array([[1., 0., 0., 0.],
       [0., 1., 0., 0.],
       [0., 0., 1., 0.],
       [0., 0., 0., 1.]])
```



# NumPy Array Attributes

- The number of dimensions: `ndim`
- The values of the dimensions as a tuple: `shape`
- The size (number of elements)

```
>>> tensor
array([[ [2.11208424, 2.01510638, 2.03126777, 1.89670846],
        [1.94359036, 2.02299445, 2.08515919, 2.05402626],
        [1.8853457 , 2.01236192, 2.07019962, 1.93713157]]],

       [[ [1.84275427, 1.99537922, 1.96060154, 1.90020305],
        [2.00270166, 2.11286224, 2.03144254, 2.06924855],
        [1.95375653, 2.0612986 , 1.82571628, 1.86067971]]]])
>>> tensor.ndim
3
>>> tensor.shape
(2, 3, 4)
>>> tensor.size
24
```

# NumPy Array Attributes

- The data type: dtype
  - can be bool, int, int64, uint, uint64, float, float64, complex ...
- The size of a single element in bytes: itemsize
- The size of the total array: nbytes

# NumPy Array Indexing

- Single elements
  - Use the bracket notation [ ]
    - Single array: Same as in standard python

```
>>> vector = np.random.normal(10,1,(5))
>>> print(vector)
[10.25056641 11.37079651 10.44719557 10.54447875 10.43634562]
>>> vector[4]
10.436345621654919
>>> vector[-2]
10.544478746079845
```

# NumPy Arrays Indexing

- Matrix and tensor elements: Use a single bracket and a comma separated tuple

```
>>> tensor
array([[[[2.11208424, 2.01510638, 2.03126777, 1.89670846],
         [1.94359036, 2.02299445, 2.08515919, 2.05402626],
         [1.8853457 , 2.01236192, 2.07019962, 1.93713157]]],

       [[1.84275427, 1.99537922, 1.96060154, 1.90020305],
        [2.00270166, 2.11286224, 2.03144254, 2.06924855],
        [1.95375653, 2.0612986 , 1.82571628, 1.86067971]]]])
>>> tensor[0,0,1]
2.015106376191313
```

# NumPy Arrays Indexing

- Multiple bracket notation
  - We can also use the Python indexing of multi-dimensional lists using several brackets

```
>>> tensor[0][1][2]  
2.085159191502853
```

- It is more writing and more error prone than the single bracket version

# NumPy Arrays Indexing

- We can also define slices

```
>>> vector = np.random.normal(10,1,(3))
>>> vector
array([10.61948855,  7.99635252,  9.05538706])
>>> vector[1:3]
array([7.99635252,  9.05538706])
```

# NumPy Arrays Indexing

- In Python, slices are new lists
- In NumPy, slices are **not** copies
  - Changing a slice changes the original

# NumPy Arrays Indexing

- Example:

- Create an array

```
>>> vector = np.random.normal(10, 1, (3))
```

```
>>> vector
```

```
array([10.61948855,  7.99635252,  9.05538706])
```

- Define a slice

```
>>> x = vector[1:3]
```



# NumPy Arrays Indexing

- Example (cont.)
  - Change the first element in the slice

```
>>> x[0] = 5.0
```

- Verify that the change has happened

```
>>> x  
array([5.          , 9.05538706])
```

- But the original has also changed:

```
>>> vector  
array([10.61948855, 5.          , 9.05538706])
```

# NumPy Arrays Indexing

- Slicing does **not** makes copies
  - This is done in order to be efficient
    - Numerical calculations with a large amount of data get slowed down by unnecessary copies
    -

# NumPy Arrays Indexing

- If we want a copy, we need to make one with the copy method
- Example:

- Make an array

```
>>> vector = np.random.randint(0,10,5)
>>> vector
array([0, 9, 5, 7, 8])
```

- Make a copy of the array

```
>>> my_vector_copy = vector.copy()
```

# NumPy Arrays Indexing

- Example (continued)

- Change the middle elements in the copy

```
>>> my_vector_copy[1:-2]=100
```

- Check the change

```
>>> my_vector_copy  
array([  0, 100, 100,   7,   8])
```

- Check the original

```
>>> vector  
array([0, 9, 5, 7, 8])
```

- No change!

# NumPy Arrays Indexing

- Multi-dimensional slicing
  - Combines the slicing operation for each dimension

```
>>> slice = tensor[1:, :2, :1]
>>> slice
array([[ [1.84275427],
        [2.00270166]]])
```

# NumPy Arrays

## Conditional Selection

- We can create an array of Boolean values using comparisons on the array

```
>>> array = np.random.randint(0,10,8)
>>> array
array([2, 4, 4, 0, 0, 4, 8, 4])
>>> bool_array = array > 5
>>> bool_array
array([False, False, False, False, False,
       False,  True, False])
```

# NumPy Arrays

## Conditional Selection

- We can then use the Boolean array to create a selection from the original array

```
>>> selection=array[bool_array]
>>> selection
array([8])
```

- The new array only has one element!

# Selftest

- Can you do this in one step?
  - Create a random array of 10 elements between 0 and 10
  - Then select the ones larger than 5



# Selftest Solution

- Solution:
  - Look a bit cryptic
    - First, we create an array

```
>>> arr = np.random.randint(0,10,10)
>>> arr
array([3, 2, 7, 8, 7, 2, 1, 0, 4, 8])
```

- Then we select in a single step

```
>>> sel = arr[arr>5]
>>> sel
array([7, 8, 7, 8])
```

# NumPy Arrays

## Conditional Selection

- Let's try this out with a matrix
  - We create a vector, then use **reshape** to make the array into a vector
  - Recall: the number of elements needs to be the same

```
>>> mat = np.arange(1,13).reshape(3,4)
>>> mat
array([[ 1,  2,  3,  4],
       [ 5,  6,  7,  8],
       [ 9, 10, 11, 12]])
```

# NumPy Arrays

## Conditional Selection

- Now let's select:

```
>>> mat1 = mat[mat>6]
>>> mat1
array([ 7,  8,  9, 10, 11, 12])
```

- This is no longer a matrix, which makes sense