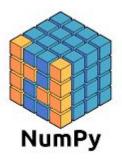
Numpy



Numpy

It is a module with a very convenient data type - array

Array characteristics:

- 1. Fixed size size predetermined in the creation time
- 2. Homogeneity contains values with one type
- 3. Memory efficiency
- 4. Speed efficiency
- 5. Vectorization same behaviour as R vectors

Creation

We will encounter several shared arguments for many numpy functions

- 1. shape dimensionality of desired array, e.g. 1 x 10, 5 x 5, usually passed as tuple
- 2. dtype type of data in an array, typically includes specific allocated number of bits per value, e.g. int, float, np.int8, and so on

Creation from python sequences

import numpy as np

From list

xs = np.array([1, 2, 3])

xs

dtype parameter

dtype can be passed with several ways:

- 1. python type int, float
- 2. string name 'int', 'float'
- numpy types np.int, np.int8, np.int16, np.float64

```
# Now floats
fractions = np.array([1, 2, 3, 4, 5], dtype='float')
fractions
array([1., 2., 3., 4., 5.])
# Another variant
```

np.array([1, 2, 3, 4, 5], dtype=float)

array([1., 2., 3., 4., 5.])

```
# Same
np.array([1, 2, 3, 4, 5], dtype=np.float)
array([1., 2., 3., 4., 5.])
# With specified number of bits per value - 16
floats16 = np.array([1, 2, 3, 4, 5],
dtype=np.float16)
floats16
array([1., 2., 3., 4., 5.], dtype=float16)
```

import sys

```
# In bytes
sys.getsizeof(fractions)
136
sys.getsizeof(floats16)
106
```

Other types of initialization

```
# Shape is equal to 3 here
# With specific value - 5
np.full(3, 5)
array([5, 5, 5])
# Another shape - 2 x 2
np.full((2, 2), 5)
array([5, 5],
       [5, 5]]
```

Some edge cases

```
# A11 0
np.zeros(3)
array([0., 0., 0.])
# All 0
np.zeros((3, 3))
array([[0., 0., 0.],
       [0., 0., 0.],
       [0., 0., 0.]
```

```
# All ones
np.ones((2, 3))
array([[1., 1., 1.],
       [1., 1., 1.]
# Identity matrix
# Only for 2-dimensional arrays
np.eye(3)
array([[1., 0., 0.],
       [0., 1., 0.],
       [0., 0., 1.]
```

```
# Ranges
np.arange(5)
array([0, 1, 2, 3, 4])
# From 3 to 12
np.arange(3, 12)
array([ 3, 4, 5, 6, 7, 8, 9, 10, 11])
# With step equal to 2
np.arange(3, 12, 2)
array([3, 5, 7, 9, 11])
```

Vectorization

```
a = np.arange(3)
b = np.ones(3)
a, b
(array([0, 1, 2]), array([1., 1., 1.]))
a + 3
array([3, 4, 5])
```

```
a * 3
array([0, 3, 6])
a + b
array([1., 2., 3.])
a < b
array([ True, False, False])
```

```
a * b
array([0., 1., 2.])

# Vector multiplication
a.dot(b)
3.0
```

```
a = np.array(((1, 2),
             (3, 4)))
b = np.arange(2)
a + b
array([[1, 3],
       [3, 5]])
a * b
array([[0, 2],
       [0, 4]])
```

a.dot(b) array([2, 4]) b.dot(a) array([3, 4]) a @ b array([2, 4])

Useful attributes

- 1. shape lengths of dimensions, tuple, first rows
- 2. size number of elements in the array, same as product of shape

```
a.shape
(2, 2)
a.size
4
# Don't use it
len(a)
```

2

Indexing

Vast topic. A couple of methods to start with:

- a[start:stop:step] same as with list, just extended to multiple dimensions
- 2. a[[1, 2, 3]] "fancy" indexing, will get elements with indices 1, 2 and 3

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

matrix[0]

[11, 12, 13, 14]

```
matrix[-1]
array([11, 12, 13, 14])
matrix[::2]
array([[ 3, 4, 5, 6],
       [11, 12, 13, 14]])
matrix[0, 0]
3
```

```
# 1st column
matrix[:, 0]
array([ 3, 7, 11])
matrix[:2, 1:3]
array([[4, 5],
       [8, 9]])
matrix[::2, ::2]
array([[ 3, 5],
       [11, 13]
```