An Introduction to Numpy

Data Science 2 / Data & AI 3



Agenda







- 1. Introduction
- 2. Indexing, slicing and reshaping
- 3. Computation on NumPy Arrays
 - Vectorized operations
 - Agregations
 - Broadcasting
- 4. Boolean Arrays and Masks
- 5. Fancy indexing



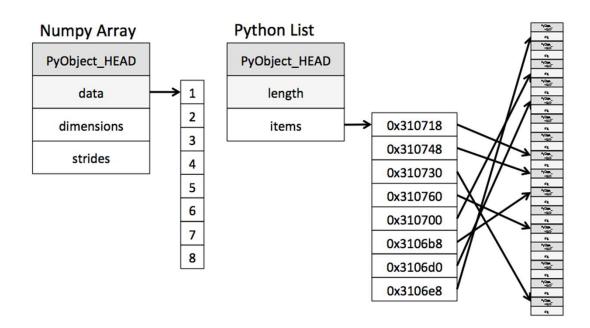




Understanding data types

What is Numpy

- NumPy: package for scientific computing in Python
- Python library that provides multidimensional array
- Routines for fast operations on arrays,
 - Mathematical
 - Logical
 - Shape manipulation
 - Selecting
 - Basic linear algebra
 - Basic statistical operations



Array creation

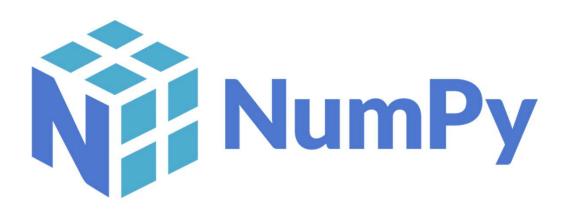
import numpy package

import numpy as np

Creating Arrays from Python Lists

```
a = np.array([1, 4, 2, 5, 3])
# Mind recent version of Python have a built-in package
"array" that also define a data type "array". That is way
we commonly import Numpy with alias "np" to make clear we
use Numpy arrays.
```

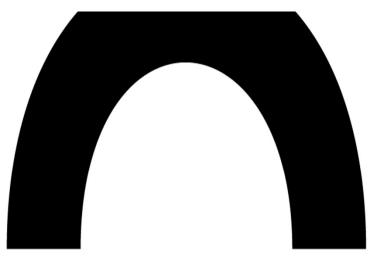
Creating Arrays from Scratch

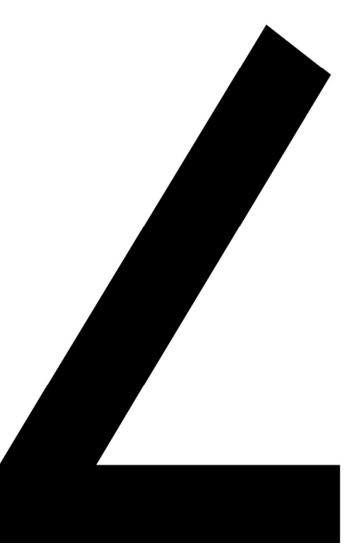




Indexing, slicing and reshaping

array[1:]				array[:,0]					array[1:,1:-1]				
2	3	4		1	2	3	4		1	2	3	4	
6	7	8		5	6	7	8		5	6	7	8	
10	11	12		9	10	11	12		9	10	11	12	
array[:-1,::2]				array[::2,1::2]					array[2,::-1]				
2	3	4		1	2	3	4		1	2	3	4	
6	7	8		5	6	7	8		5	6	7	8	
10	11	12		9	10	11	12		9	10	11	12	
	2 6 10 rray[: 2 6	2 3 6 7 10 11 rray[:-1,:: 2 3 6 7	2 3 4 6 7 8 10 11 12 rray[:-1,::2] 2 3 4 6 7 8	2 3 4 6 7 8 10 11 12 rray[:-1,::2] 2 3 4 6 7 8	2 3 4 1 6 7 8 5 10 11 12 9 rray[:-1,::2] arr 2 3 4 1 6 7 8 5	2 3 4 1 2 6 7 8 5 6 10 11 12 9 10 rray[:-1,::2] array[:: 2 3 4 1 2 6 7 8 5 6	2 3 4 1 2 3 6 7 8 5 6 7 10 11 12 9 10 11 rray[:-1,::2] array[::2,1: 2 3 4 1 2 3 6 7 8 5 6 7	2 3 4 1 2 3 4 6 7 8 5 6 7 8 10 11 12 9 10 11 12 ray[:-1,::2]	2 3 4 1 2 3 4 6 7 8 5 6 7 8 10 11 12 9 10 11 12 rray[:-1,::2] array[::2,1::2] 2 3 4 1 2 3 4 6 7 8 5 6 7 8	2 3 4 1 2 3 4 1 6 7 8 5 6 7 8 5 10 11 12 9 10 11 12 9 ray[:-1,::2] array[::2,1::2] ar 2 3 4 1 2 3 4 1 6 7 8 5 6 7 8 5	2 3 4 1 2 3 4 1 2 6 7 8 5 6 7 8 5 6 10 11 12 9 10 11 12 9 10 rray[:-1,::2] array[::2,1::2] array[2 3 4 1 2 3 4 1 2 6 7 8 5 6 7 8 5 6	2 3 4 1 2 3 4 1 2 3 6 7 8 5 6 7 8 5 6 7 10 11 12 9 10 11 12 9 10 11 ray[:-1,::2] array[::2,1::2] array[2,::-2] 2 3 4 1 2 3 6 7 8 5 6 7	



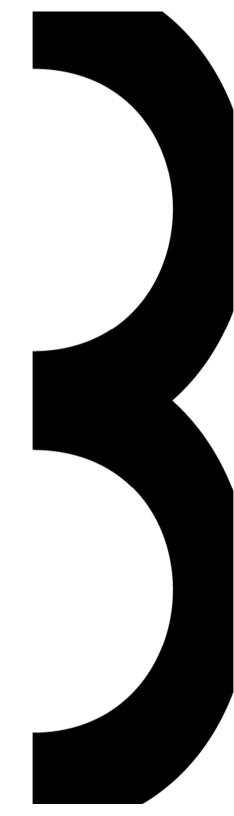


Indexing, slicing and reshaping

```
# Array Attributes
x = np.array([[1, 2, 3],
                [4, 5, 6]
print("x ndim: ", x.ndim) # 2
print("x shape:", x.shape) # (2, 3), 2 rows and 3 columns
print("x size: ", x.size) # 6
# Array Indexing
                                 # 6
x[1, 2]
                                 # [1, 2, 3]
x[0]
                                 # [4, 5, 6]
\times [-1]
                                                   Make sure you are fluent in
                                                   indexing and slicing arrays!
# Array Slicing
x[0, :2]
                                 # [1, 2]
                                 # [2, 5]
x[:, 1]
                                 # [[1, 2], [4, 5]]
x[:, :2]
# Reshaping
np.arange(1, 10).reshape((3, 3))
```



Computation on NumPy Arrays



Vectorized operations

- Python for loops are slow
- Solution: Numpy's vectorized operations and functions

When using Numpy & Pandas (later on) using loops is suboptimal and will affect your evaluation scores negatively!

Array arithmetic

```
x = np.arange(3)  # [0, 1, 2]
2 ** x  # [1, 2, 4]

x / np.arange(1, 4)  # [0, 0.5, 0.66666667]
-(0.5*x + 1)  # [-1, -1.5, -2]

np.power(3, x)  # [1, 3, 9]
```

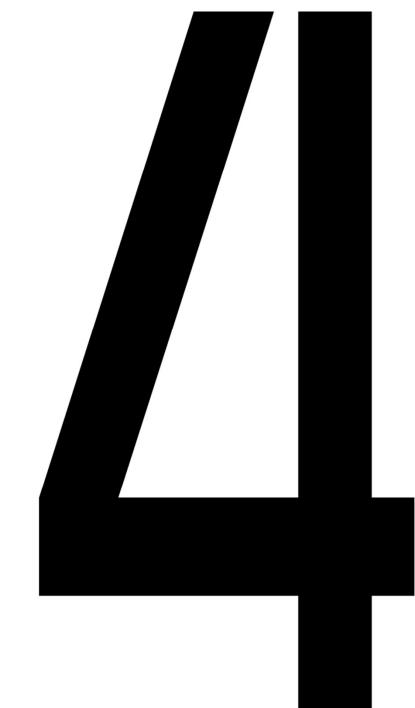
Aggregates

```
np.add.reduce(x) # 3
np.add.accumulate(x) # [0, 1, 3]
np.multiply.reduce(x) #0
np.multiply.accumulate(x) # [0, 0, 0]
```



Computation on NumPy Arrays





Agregations

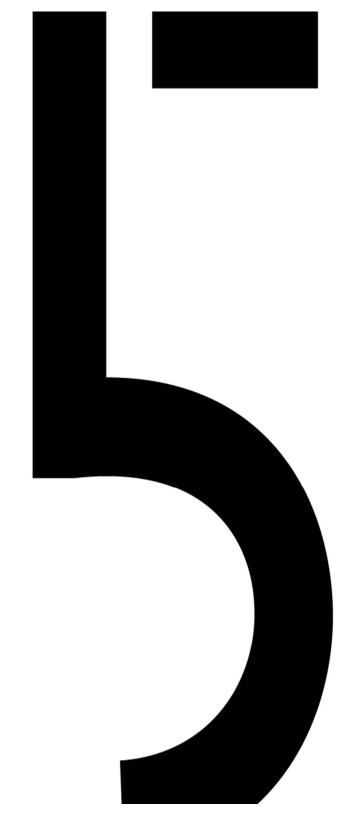
```
x = np.array([[1, 5, 3],
          [4, 2, 6]])
                      # 21
x.sum
                      # 1
x.min
                      # [1, 2, 3],
x.max(axis=0)
                      # axis is dimension that is collapsed
x.max(axis=1)
                      # [1, 2]
# Statistics
                  # [2.5, 3.5, 4.5]
x.mean(axis=0)
x.var()
x.std(axis=1)
x.var()
x.median()
                     # as method
                 # as function
np.median(x)
np.percentile(x, 25)
```

- p. 11



Computation on NumPy Arrays

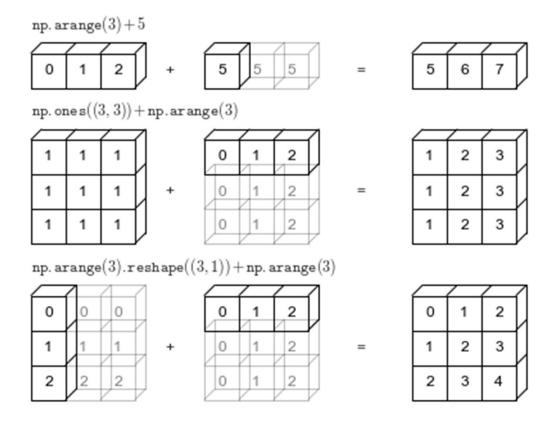
Broadcasting



Broadcasting

Rules

- 1. If the two arrays differ in their number of dimensions, the shape of the one with fewer dimensions is padded with ones on its leading (left) side
- 2. If the shape of the two arrays does not match in any dimension, the array with shape equal to 1 in that dimension is stretched to match the other shape
- 3. If in any dimension the sizes disagree and neither is equal to 1 -> error



Take some time to understand the broadcasting principles!



Boolean Arrays and Masks



Boolean Arrays and Masks

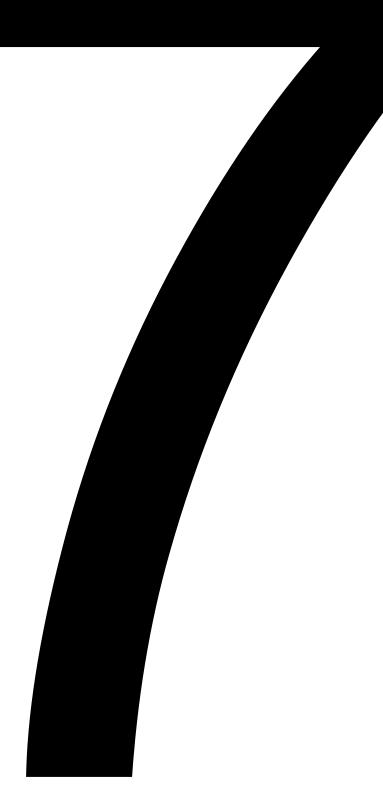
x[x < 5]

```
x = np.array([[1, 5, 3],
          [4, 2, 6]])
x < 5
                               # [[True, False, True], [True,
True, False]]
np.sum(x < 5)
                               # 4, False -> 0, True -> 1
np.sum(x < 5, axis=1)
                              # [2, 2]
                               # True
np.any(x < 5)
                               # False
np.all(x < 5)
np.sum((x > 2) & (x < 5))
                                      # 2
# Boolean Arrays as Masks
```

[1, 3, 4, 2], returns 1-dim array



Fancy indexing



Fancy indexing

```
x = np.arange(1,10)

ind = [3, 7, 4, 0]

x[ind] # [4, 8, 5, 1]
```