# **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



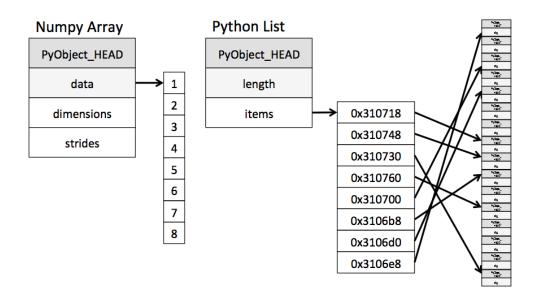






#### What is Numpy

- 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])
# Creating Arrays from Scratch
np.zeros(3, dtype=int)
                                         # array([0, 0, 0])
np.ones((2, 1), dtype=float)
                                         # array([[ 1.],
                                                   [1.]])
np.random.randint(0, 10, (3, 3))
                                         # random integers in [0, 10 )
np.arange(1, 5)
                                         #[1, 2, 3, 4]
```

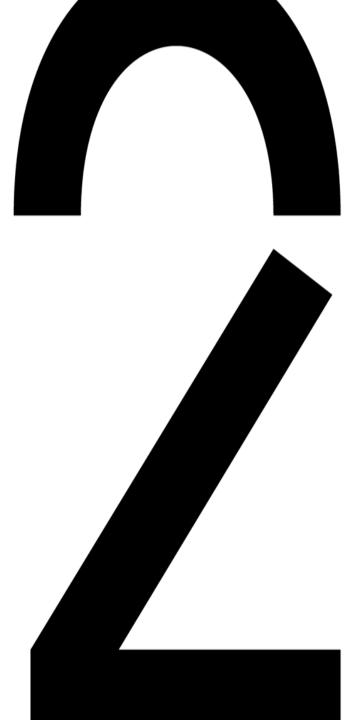
#### Notebook

See ...

Exercise time!



Indexing, slicing and reshaping



# 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]
x[0]
                                          # [1, 2, 3]
x[-1]
                                          # [4, 5, 6]
# Array Slicing
x[0, :2]
                                          #[1, 2]
x[:, 1]
                                          #[2, 5]
x[:, :2]
                                          # [[1, 2], [4, 5]]
# Reshaping
np.arange(1, 10).reshape((3, 3))
```

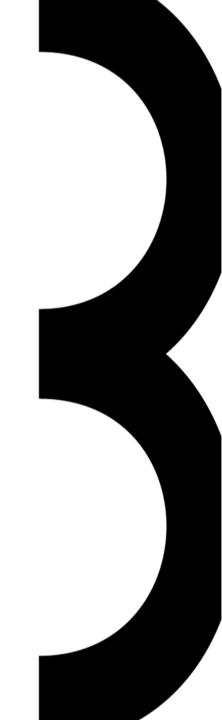
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# Computation on NumPy Arrays



#### **Vectorized operations**

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

```
# Array arithmetic
                                                              # [0, 1, 2]
x = np.arange(3)
2 ** x
                                                                                   #
[1, 2, 4]
x / np.array(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
                                                              # [0, 1, 3]
np.add.accumulate(x)
np.multiply.reduce(x)
                                                              # 0
                                                            - #110, 0, 01
np.multiply.accumulate(x)
```

#### Notebook

See ...

Exercise time!

# **Agregations**

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

- p.13

#### Notebook

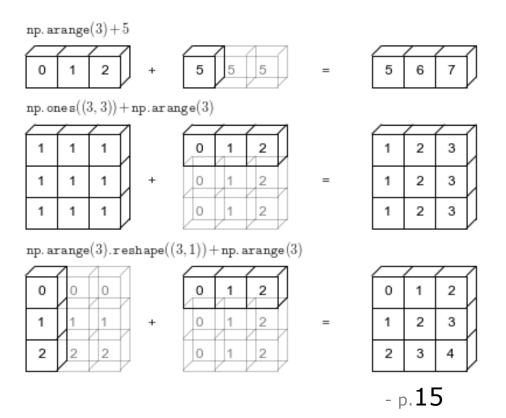
See ...

Exercise time!

#### **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
- 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



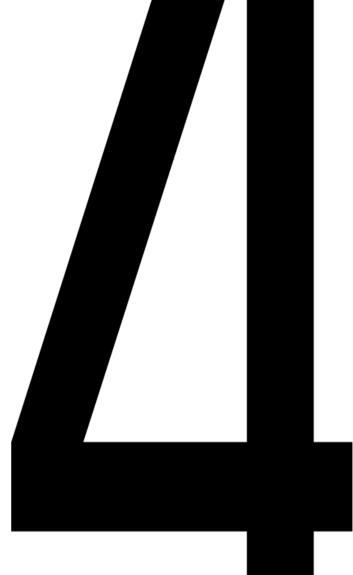
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See ...

Exercise time!



**Boolean Arrays and Masks** 



### **Boolean Arrays and Masks**

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

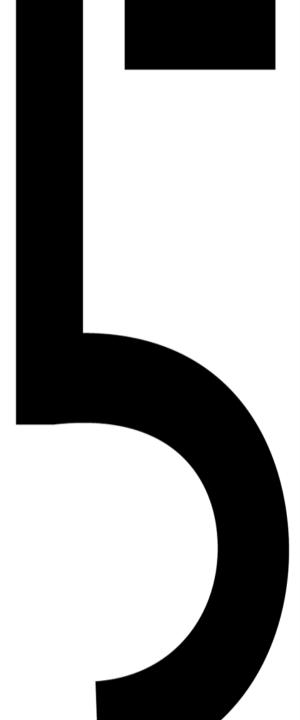
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Exercise time!



Fancy indexing



# **Fancy indexing**

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

ind = [3, 7, 4, 0]

x[ind]
```

# [4, 8, 5, 1]

#### Notebook

See ...

Exercise time!