# Scientific Programming in Python crash course

Aprendizaje Automático para la Robótica Máster Universitario en Ingeniería Industrial

Departamento de Automática





## Objectives

- 1. Introduce scientific programming problems
- 2. Efficient matrix computations in Python
- 3. Visualize data in Python

## Bibliography

Jake VanderPlas. Python Data Science Handbook. Chapter 1. O'Reilly. (Link).

## Table of Contents

- I. The data scientist toolkit
  - Motivation
  - Overview
  - Anaconda
  - Conda
- 2. NumPy
  - Understanding data types in Python
  - Introduction
  - NumPy array attributes
  - NumPy data types
  - NumPy notebook

- 3. Pandas
  - Introduction
  - The Pandas Series object
  - The Pandas DataFrame object
  - Pandas notebook
- 4. Data visualization
  - Introduction
  - Matplotlib
  - Matplotlib notebook
  - Searborn
  - Seaborn notebook



#### Motivation

#### Data science is about manipulating data

- Need of specialized tools
- Two main languajes: R and Python

#### Python is a general purpose programming language

- Easy integration
- Huge ecosystem of packages and tools

#### Need of data-oriented tools

• Features provided by third-party tools



#### Overview

Tool	Туре	Description
conda	Software	Python environments and package management
iPython	Software	Advaced Python interpreter
Jupiter	Software	Python notebooks (Python interpreter)
Numpy	Package	Efficient array operations
Pandas	Package	Dataframe support
Matplotlib	Package	Data visualization
Seaborn	Package	Data visualization with dataframes
Scikit-learn	Package	AI/ML package for Python



#### Anaconda

## Most of those tools are packaged in Anaconda

- Python distribution for Data Science
- Environment management for Python
- Package management system

#### Anaconda provides conda

- Packages management tool
- Environment management for Python

#### In addition, Anaconda provides Spyder

• Python IDE designed for Data Science







#### Conda crush introduction

## Conda environment for Data Science

- i. conda create --name ml
  seaborn=0.9.0
- 2. source activate ml
- 3. conda install ipython
- 4. conda install nb\_conda
- 5. conda install scikit-learn

List environments:
 conda info --envs

Activate environment:
 source activate <env>
Install package:
 conda install <package>
List packages:
 conda list

Exit environment:
 conda deactivate



## Understanding data types in Python (I)

```
Static typing
/* C code */
int result = 0;
for(int i=0; i<100; i++){
    result += i;
}</pre>
```

- Data types must be declared
- Data types cannot change
- Error detection in compilation
- Variables names are, basicly, labels

#### Dynamic typing

```
# Python code
result = 0
for i in range(100):
    result += i
```

- Data types are not declared
- Data types can change
- Error detection in run-time
- Variables are complex data structures (even for simple types)

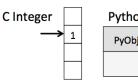


## NumPy

## Understanding data types in Python (II)

Dynamic typing must be implemented somewhere ...

```
Python 3.4 source code
struct _longobject {
    long ob_refcnt;
    PyTypeObject *ob_type;
    size t ob size;
    long ob_digit[1];
};
```

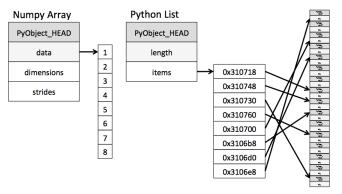




## Understanding data types in Python (III)

### A Python list may contain different types

```
In [1]: L3 = [True, "2", 3.0, 4]
   ...: [type(item) for item in L3]
Out[1]: [bool, str, float, int]
```





## Understanding data types in Python (IV)

Standard Python data types are powerful and flexible

- Flexibility has a price: Reduced performance
- Not an big issue in generic programming
- A big issue in scientific programming
- We require efficient data manipulation mechanisms: NumPy

NumPy: Python package for numeric computation

- Efficient array implementation
- Fast mathematical functions
- Random numbers generation
- Static data types: Less flexibility

Most Python modules for AI/ML depend on NumPy, in particular

Pandas (dataframes), Scikit-learn (ML), Seaborn (data visualization)



## NumPy must be imported in order to be available

• You can use np? or np. <TAB>

The main component of NumPy is ndarray

- Python object
- Efficient matrix representation
- Homogeneus elements

#### Convention

import numpy as np

```
In [1]: array = np.array
          ([1,2,3])
In [2]: array
Out[1]: array([1,2,3])
In [3]: array = np.array
     ([[1,2],[3,4]])
```



## NumPy array attributes

### Ndarray objects expose several attributes

- ndim: Dimensions
- shape: Size of each dimension
- size: Number of elements
- dtype: Data type
- itemsize: Size of each element (in bytes)
- nbytes: Size of the array (in bytes)

```
x = np.random.randint(10, size
    =(3,4)
print("x ndim: ", x.ndim)
print ("x shape:", x.shape)
print("x size: ", x.size)
print("dtype:", x.dtype)
print("itemsize:", x.itemsize)
print("nbytes:", x.nbytes)
```



# NumPy data types

## Python is implemented in C

• Data types in NumPy are based on those in C

## Two styles to declare types

- String: np.zeros(10, dtype='int16')
- NumPy object: np.zeros(10, dtype=np.int16)

<b>Дата түре</b>	Description
bool_	Boolean (True or False) stored as a byte
int_	Default integer type
intc	Identical to C
intp	Integer used for indexing
int8	Byte
int16	Integer
int32	Integer
int64	Integer
uint8	Unsigned integer
uint16	Unsigned integer
uint32	Unsigned integer
uint64	Unsigned integer
float_	Shorthand for float64
float16	Half precision float
float32	Single precision float
float64	Double precision float
complex_	Shorthand for complex128
complex64	Complex number
complex128	Complex number

# NumPy notebook

## NumPy notebook

(Link to notebook)



#### Introduction

#### A DS/ML workflow needs more features

- Missing data
- Data input
- Operations on groups
- Label columns and rows

Pandas provides all those features, and more

- Pandas = PANel DAta System
- Built on NumPy's ndarray
- Provides dataframes

Pandas provides two main objects

Series and DataFrame









#### Convention

import numpy as np import pandas as pd



### Pandas

## The Pandas Series object (I)

A Series is a one-dimensional array of indexed data

- NumPy arrays indices are implicit (i.e. its position)
- Series indices are explicit, and can be any type

#### Two attributes

- values: ndarray
- index: pd. Index object

#### Two indices

- Implicit: Regular index
- Explicit: Custom index

Index	VALUES
'a'	0.25
'b'	0.5
'c'	0.75
'd'	0.99

### Pandas

## The Pandas Series object (II)

```
In[1]: data = pd. Series ([0.25, 0.5, 0.75, 1.0],
   index = ['a', 'b', 'c', 'd'])
In [2]: data
Out [1]:
     0.25
   b 0.50
   c 0.75
    d 1.00
    dtype: float64
    In [3]: data['a']
    Out [2]: 0.25
    In [4]: data[0]
    Out [3]: 0.25
```

#### Pandas

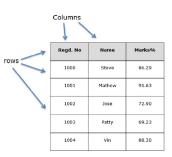
## The Pandas DataFrame object (I)

#### A DataFrame is a 2-D tabular data structure

- Similar to a spreadsheet
- Homogeneous columns
- Heterogeneous rows

Two read-only attributes, both pd. Index

- index: Rows
- columns: Columns



(Source)

## The Pandas DataFrame object (II)

```
In [1]: import seaborn as sns
In [2]: iris = sns.load_dataset('iris')
In [3]: iris.head()
Out [1]:
sepal_length sepal_width petal_length petal_width species
0
           5.I
                        3 - 5
                                     I.4
                                                   o.2 setosa
                                                   o.2 setosa
           4.9
                        3.0
                                     I.4
                                                  o.2 setosa
           4 - 7
                      3.2
                                     I.3
                                                o.2 setosa
           4.6
3
                       3.I
                                     I.5
           5.0
                       3.6
                                     I.4
                                              o.2 setosa
In [246]: iris.columns
Out [246]:
Index(['sepal_length', 'sepal_width', 'petal_length',
        'petal_width', 'species'], dtype='object')
```

## The Pandas DataFrame object (III)

#### Read from a file

- Excel: pd.read excel('filename.xlsx', sheetname='mysheet')
- CSV (very common!!!): pd.read csv('filename.csv')

```
# This CSV file contains data about weights and heights
"id", "weight", "height", "sex", "race"
1, 143.5, 81.6, "Female", "White"
2, 109.1, 83.7, "Female", "Black"
4, 104.8, 54.6, "Female", "Hisp"
7, 130.2, 81.7, "Male", "White"
```

CVS can be exported from MS Excel or programatically



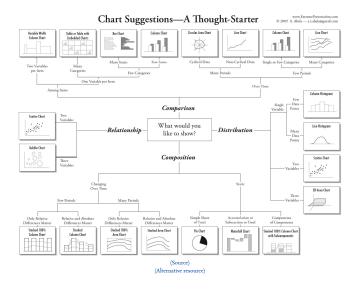
## Pandas notebook

## NumPy notebook

(Link to notebook)



#### Motivation



## Matplotlib (I)

#### Matplotlib is a Python package

- Based on NumPy
- Imitates Matlab

#### Three operation modes

- Scripts.
   Must use plt.show() to enter event loop. Use it once!
- IPython shell.
   Must use %matplotlib
- IPython notebook. Two modes
  - %matplotlib inline
  - %matplotlib notebook

#### Convention

```
import matplotlib as mpl
import matplotlib.pyplot as plt
```

#### myplot.py

```
\begin{array}{ll} \textbf{import} & \textbf{matplotlib} \ . \ \textbf{pyplot} & \textbf{as} & \textbf{plt} \\ \textbf{import} & \textbf{numpy} & \textbf{as} & \textbf{np} \end{array}
```

```
x = np. linspace(o, 10, 100)
```

```
plt.plot(x, np.sin(x))
plt.plot(x, np.cos(x))
```

```
plt.show()
```

## Matplotlib (II)

#### Matplotlib comes with two interfaces

- Matlab-like, Old-fashioned function-oriented API.
- Object-oriented. Object-oriented and more powerfull API.

#### Matlab API

#### OO API



## Pandas

## Matplotlib notebook

## Matplotlib notebook

(Link to notebook)



Seaborn (I)

#### Seaborn is a modern data-visualization Python package

- Based on matplotlib
- ... it uses matplotlib indeed
- Pandas-aware
- High level
- Advanced visualizations.
- Easy to use

Still under development! (v. o.9)



import seaborn as sns

This documentation is for Seaborn

o.9 or newer



Seaborn (II)

#### Display initialization

- plt.show()
- %matplotlib

#### Style initialization

- Default Seaborn style sns.set()
- By default, same style than matplotlib

#### Several functions ...

• ... similar parameters

#### Parameters

- x: Data axis x
- y: Data axis Y
- data: Dataframe name
- hue: Color
- style: Style
- sizes: Size
- kind: Alternate representation



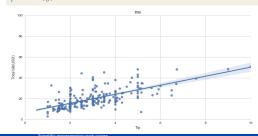
ee data scientist toolkit NumPy Pandas **Data visualization** 

#### Data visualization

## Seaborn (III)

## Typical Seaborn usage

- 1. Prepare data
- 2. Set up aesthetics
- 3. Plot
- 4. Customize the plot



## Seaborn

### Seaborn notebook

## Seaborn notebook

(Link to notebook)

