

# Lesson 2: Second steps with Python and the NumPy module

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Alessio Cardillo

Department of computer science and mathematics (DSIM),  
Universitat Rovira i Virgili, Tarragona, Spain

Python Days

Dept. of physics and astronomy – University of Catania, Catania, Italy



UNIVERSITAT ROVIRA I VIRGILI

# Foreword

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## Course

1. ~~Introduction to the basics of the Python language.~~

# Outline of the course/lesson

## Course

1. ~~Introduction to the basics of the Python language.~~
2. More “*basics*” of Python & basics of NumPy.
3. IPython notebook & Visualization of data (Matplotlib).
4. Data analysis (Pandas).

## Lesson 2

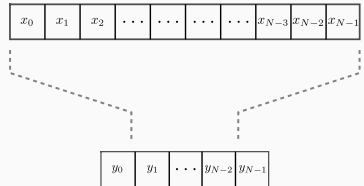
- Manipulating lists
- File input/output, JSON, functions, & handling errors
- Running third party software in Python
- The NumPy module
- Hands-on session

# Manipulating lists



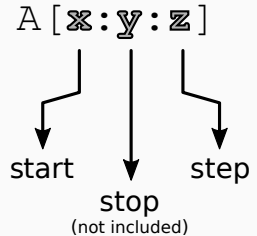
# List slicing/comprehension

- By *list slicing* we refer to a way of extracting (and eventually copying) subsets (slices) of list objects (and other ordered iterables).



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- Slicing is based on three *stride* indicators (even not all together).



## Hands-on

```
>>> # defining a list
>>> a = list(range(10))
>>> a
[0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
>>> # setting a new list (slicing)
>>> b = a[2:6:2]
>>> b
[2, 4]
```

# List slicing/comprehension

- By *list slicing* we refer to a way of extracting (and eventually copying) subsets (slices) of list objects (and other ordered iterables).
- Slicing is based on three *stride* indicators (even not all together).
- Slicing can be done both forward and backward.

## Hands-on

```
>>> # defining a list
>>> a = list(range(10))
>>> a
[0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
>>> # striding forward
>>> b = a[2:7:2]
>>> b
[2, 4, 6]
>>> # striding backward
>>> c = a[7:2:-2]
>>> c
[7, 5, 3]
```



# List slicing/comprehension

- By *list slicing* we refer to a way of extracting (and eventually copying) subsets (slices) of list objects (and other ordered iterables).
- Slicing is based on three **stride** indicators (even not all together).
- Slicing can be done both forward and backward.
- Note: When we do assignment through slicing the size of the two objects must be **THE SAME**.

## Hands-on

```
>>> a = list(range(10))
>>> b = list(range(5))
>>> a[2:6:2] = b[1:3:2]
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
ValueError: attempt to assign sequence
of size 1 to extended slice of size 2
```

# List slicing/comprehension

By **list comprehension** we indicate any “*compact way*” to perform complex operations on lists.

## Hands-on

```
>>> l = range(10)
>>> # list comprehension
>>> a = [x for x in l]
>>> a
[0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
>>> # a more sophisticated one
>>> b = [x**2 for x in l]
>>> b
[0, 1, 4, 9, 16, 25, 36, 49, 64, 81]
>>> # you can nest multiple instances
>>> c = [[x**i for x in l[2:5]] for i in range(3)]
>>> c
[[1, 1, 1], [2, 3, 4], [4, 9, 16]]
```

## File input/output, JSON, Functions, & Handling errors

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# Input: command line arguments

## Getting input from stdin

- The simplest way to pass variables to a Python script via the standard input (stdin) involves the `sys.argv` variable. More sophisticated ways involve parsing command line options (for instance via the `getopt` or the `argparse` modules).
- The `sys.argv` variable (available through the `sys` module) works exactly as its homologous in C (e.g., `sys.argv[0]` is the name of the program).

## Hands-on

```
# basic usage of the sys.argv method  
import sys  
  
print('Nr. of args:', len(sys.argv))  
print('Args List:', str(sys.argv))
```

Then, run the script as:

```
$ python3 myscript.py var1 3.5 True
```

## Note

REMEMBER: Python parses arguments as **strings**. Hence, you need to convert them to the appropriate type!

# Read/Write from/to file

## Input file (example-read-io.txt)

```
This file has 3 lines  
This is the second line  
This is the third line
```

## Reading from file

```
>>> # opening the file object  
>>> f = open('hands-on/example-read-io.txt', 'r')  
>>> # reading the WHOLE file  
>>> f.read()  
'This file has 3 lines\nThis is the second line\nThis is the third line\n'  
>>> # closing the file object  
>>> f.close()
```

## Opening modes

| Code | Mode           |
|------|----------------|
| 'r'  | read only      |
| 'w'  | write (over)   |
| 'a'  | append         |
| 'r+' | read and write |

## Tip

Be **very careful** with the `.read()` method because it reads the WHOLE file at once!

# Read/Write from/to file

## Python

```
>>> # opening the file inside a with statement
>>> with open('hands-on/example-read-io.txt', 'r') as f:
...     f.read()
```

```
>>> # reading one row at a time
>>> with open('hands-on/example-read-io.txt', 'r') as f:
...     for line in f:
...         print(line, end='')
...
This file has 3 lines
This is the second line
This is the third line
```

```
>>> # reading skipping the first two rows
>>> with open('hands-on/example-read-io.txt', 'r') as f:
...     for _ in range(2):
...         a = next(f)
...     for line in f:
...         print(line, end='')
...
This is the third line
```

## C

```
#include <stdio.h>

int main() {

    FILE *f;
    char buff[255];

    f = fopen("example-read-io.txt", "r");
    fscanf(f, "%s", buff);
    printf("%s\n", buff);

    fgets(buff, 255, (FILE*)f);
    printf("%s\n", buff );

    fclose(f);

    return 0;
}
```

- <https://docs.python.org/3/tutorial/inputoutput.html#reading-and-writing-files>
- [https://docs.python.org/3/reference/compound\\_stmts.html#with](https://docs.python.org/3/reference/compound_stmts.html#with)

## Read/Write from/to file

## Writing to file

```
>>> # declaring a matrix
>>> mydata = [[1,2,3],[4,5,6],[7,8,9]]
>>> # opening a file and writing on it
>>> with open('hands-on/example-write-io.txt', 'w') as f:
...     for row in mydata:
...         for col in row:
...             f.write("%d\t" % col)
...         f.write("\n")
... 
```

# Read/Write from/to file

## Good to know

Two very handy modules are the `tarfile` and the `zipfile`. The former makes it possible to read and write tar archives, including those using *gzip*, *bz2*, and *lzma* compression. The latter, instead, can be used to read/write `.zip` files.

## Hands-on

```
>>> import tarfile
>>>
>>> # loading the tarfile
>>> fname = 'tarfile-example.tar.gz'
>>> mytar = tarfile.open('hands-on/'+fname, 'r:gz')
>>>
>>> print('Printing the content of the tarfile %s \n' % fname)
>>>
>>> # iterating over the files in the tarfile
>>> for members in mytar.getmembers():
...     print(members)
...
Printing the content of the tarfile tarfile-example.tar.gz

<TarInfo 'archive-1.dat' at 0x7ff6a40ee818>
<TarInfo 'archive-2.dat' at 0x7ff6a40ee688>
<TarInfo 'archive-3.dat' at 0x7ff6a40ee750>
```

- <https://docs.python.org/3/library/tarfile.html>
- <https://docs.python.org/3/library/zipfile.html>



# The JSON format

## What is a JSON?

JSON (*JavaScript Object Notation*) is a data-interchange format easy for humans to read/write, and for machines to parse/generate. It is completely language independent (ideal for data-interchange), but uses conventions that are familiar to programmers of the C-family of languages (e.g., C, C++, C#, Java, JavaScript, Perl), and **OF COURSE** Python.

- <https://www.json.org/json-en.html>

# The JSON format

## A JSON is based on two structures

**object** A collection of name/value pairs  
(*i.e.*, an associative array).

**array** An ordered list of values.

# The JSON format

A JSON is based on two structures

**object** A collection of name/value pairs  
(*i.e.*, an associative array).

**array** An ordered list of values.

## Question

What are the Python equivalents of the **object** and the **array**?

# The JSON format

## Example of a JSON

```
myjson = {  
    'key1': 0.358,  
    'key2': [1, 3, 5],  
    'key3': True,  
    'key4': {  
        'key41': 'mystring',  
        'key42': [[3.5, 'string'], [-7.5, 3]]  
    }  
}
```

## Note

**JSON does not support** comments!

# The JSON format

## The json module

```
import json

# loading a json from file
with open('data.txt', 'r') as infile:
    myjson = json.load(infile)

print(myjson)

# printing a json to file
myjson = {'a': True, 'b': False}

with open('test_file-dump.json', 'w') as outfile:
    json.dump(myjson, outfile)

# pretty printing
with open('test_file-dump.json', 'w') as outfile:
    json.dump(myjson, outfile, indent=4)
```

# Functions/modules

Functions are declared using the keyword `def`. There is no need to declare the type of the function or of its arguments (*duck typing*). All functions **must** return something. By default, the return of a function is `None`.

## Hands-on

```
>>> # define universal sum
>>> def mysum(v1, v2):
...     return v1+v2
...
>>> mysum(3,5)
8
>>> mysum('Hello', ' World!')
'Hello World!'
>>> mysum([1,3,5], ['a', 'b', 'c'])
[1, 3, 5, 'a', 'b', 'c']
```

- <https://docs.python.org/3/glossary.html#term-function>

# Functions/modules

Functions are declared using the keyword `def`. There is no need to declare the type of the function or of its arguments (*duck typing*). All functions **must** return something. By default, the return of a function is `None`.

## Hands-on

```
>>> # factorial
>>> def factorial(n):
...     if n < 2:
...         return 1
...     else:
...         return n*factorial(n-1)
...
>>> for i in [1, 3, 5, 10, 40]:
...     print('fact(%d) = %d' %(i, factorial(i)))
...
fact(1) = 1
fact(3) = 6
fact(5) = 120
fact(10) = 3628800
fact(40) = 81591528324789773434561126959611589427200
```

- <https://docs.python.org/3/glossary.html#term-function>

# Functions/modules

Parameters can be passed either by **position** or by **name** (reference).

Parameters might have **default values**. We can also write functions with an **arbitrary number of parameters**.

## Hands-on

```
>>> # define a function
>>> def mysum(v1, v2):
...     return v1+v2
...
>>> # pass by position & name
>>> mysum(1, v2=6)
7
>>> # pass only by name (notice the order)
>>> mysum(v2=6, v1=2)
8
```

- <https://stackoverflow.com/questions/986006/how-do-i-pass-a-variable-by-reference>



# Functions/modules

Parameters can be passed either by **position** or by **name** (reference).

Parameters might have **default values**. We can also write functions with an **arbitrary number of parameters**.

## Hands-on

```
>>> # setting default value
>>> def mysum(v1, v2=5):
...     return v1+v2
...
>>> # passing only v1
>>> mysum(2)
7
>>> # passing v1 and v2
>>> mysum(2, 7)
9
>>> # passing v1 by name
>>> mysum(v1=2, 7)
File "<stdin>", line 1
SyntaxError: positional argument follows
keyword argument
```

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# Functions/modules

Parameters can be passed either by **position** or by **name** (reference).

Parameters might have **default values**. We can also write functions with an **arbitrary number of parameters**.

## Hands-on

```
>>> # defining a function
>>> def mysum(v1, *args):
...     for elem in args:
...         v1 += elem
...     return v1
...
>>> mysum(1,3)
4
>>> # declaring a list l
>>> l = [1,2,3]
>>> # passing the content of l to the function
>>> mysum(1,*l)
7
```

• <https://stackoverflow.com/questions/986006/how-do-i-pass-a-variable-by-reference>

Parameters can be passed either by **position** or by **name** (reference).

Parameters might have **default values**. We can also write functions with an **arbitrary number of parameters**.

## Hands-on

```
>>> # function with arbitrary nr. of parameters
>>> def myfunc(a, **kwargs):
...     print(a)
...     print(kwargs)
...     return None # optional
...
>>> # passing parameters by name
>>> myfunc(5, x=1, y=2)
5
{'x': 1, 'y': 2}
>>> # passing parameters as dictionary
>>> c = {'x':1, 'y': 'Hello'}
>>> myfunc(5, **c)
5
{'x': 1, 'y': 'Hello'}
>>> # ERROR multiple definition
>>> c = {'x':1, 'a': 'Hello'}
>>> myfunc(5, **c)
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
TypeError: myfunc() got multiple values for argument 'a'
```

# Functions/modules

We can associate to each function a *doc string* explaining what the function does. We can access the doc string via the command `myfunc.__doc__` or by `help(myfunc)`. It is usually delimited with `""" """`.

## Hands-on

```
>>> def mysum(v1, v2):  
...     """This function computes the sum between  
...         two elements of arbitrary type."""  
...     return v1 + v2  
...  
>>> print(mysum.__doc__)  
This function computes the sum between  
two elements of arbitrary type.  
>>> help(mysum)
```

- <https://www.python.org/dev/peps/pep-0257/>

# Handling errors

- Error handling, despite being a complex task, ensures the functionality of code.
- Older and modern languages handle errors in completely different ways.
- A modern error handling system should have these characteristics:
  1. Low impact on performances (if errors do not occur).
  2. Small overhead on the code.
  3. Possibility to handle the error in an automated way.

# Handling errors

Let's look at “typical” error occurring  
in scientific computing:  
the **zero division** error!

$$\frac{x}{0} = \infty$$

# Handling errors

C

```
#include <stdio.h>
#include <stdlib.h>

main() {

    int dividend = 20;
    int divisor = 0;
    int quotient;

    if( divisor == 0){
        fprintf(stderr, "Division by zero! Exiting...\n");
        exit(-1);
    }

    quotient = dividend / divisor;
    fprintf(stderr, "Value of quotient : %d\n", quotient );

    exit(0);
}
```

# Handling errors

## Python

```
>>> num = 10.  
>>> denom = 0.  
>>>  
>>> try:  
...     print('division = %.2f' %(num/denom))  
... except ZeroDivisionError:  
...     print('You cannot divide by zero!')  
...  
...
```

- We can handle the error by placing the code inside a so-called **try/except** block.
- We can catch either specific **classes** of errors or all of them.



# Handling errors

## Hands-on

```
try:
... # tries to execute the code here
...
except Exc0: # captures exceptions of type Exc0
... # do something in case Exc0 occurs
...
except (Exc0, Exc1): # captures exceptions of type Exc0 or Exc1
... # do something in case Exc0 or Exc1 occurs
...
except: # captures EVERY type of exceptions
... # do something in case an exception occurs
...
else: # what should be done in case no exception occurs
... # do something in case of no exception
...
finally: # what must be done regardless of what happens
... # execute this code ALWAYS
...
```

## Note

We can generate an error/exception using the command `raise`.

## Running third party software

---

# Running external software in Python

It is possible to launch/run “external code” from a Python script.  
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It is possible to launch/run “external code” from a Python script. There are two distinct way of doing it:

1. Calling **directly** functions (e.g., of a shared C library) in the code.
2. Running **externally** the code (e.g., a bash script).

# Running external software in Python

## Extension modules

Such modules can do two things that can't be done directly in Python: they can implement new built-in object types, and they can call C library functions and system calls.

## the ctypes module

It provides C compatible data types, and allows calling functions in DLLs or shared libraries. It can be used to wrap these libraries in pure Python.

- <https://docs.python.org/3/extending/extending.html>
- <https://docs.python.org/3/library/ctypes.html#module-ctypes>

# Running external software in Python

## Running external code through Python

Let's suppose that we need to interact directly with the operative system, or that we have an old, but well-working, `bash` script that automatically backups all our important files/directories. How can we do such operations through Python?

## Answer

We can use either the `os` module, or the `subprocess` one!

- <https://stackabuse.com/executing-shell-commands-with-python/>

# Running external software in Python

## Our bash script: myscript-bash.sh

```
#!/bin/bash

echo -e "HELLO! I AM A BASH SCRIPT\n"
echo -e "THE VALUE INSERTED IS: $1\n\n"

echo -e "IN THIS DIRECTORY WE HAVE THE FOLLOWING FILES:\n"
ls -lght

# end of script
```

```
alessio@aleurv:~/corso-python-dfa/lectures/lect2/hands-on$ bash myscript-bash.sh 7
HELLO! I AM A BASH SCRIPT

THE VALUE INSERTED IS: 7

IN THIS DIRECTORY WE HAVE THE FOLLOWING FILES:

total 4,0K
-rw-rw-r-- 1 alessio 176 nov 12 15:47 myscript-bash.sh
```



# Running external software in Python

## Using the os module

```
import os
```

```
os.system("bash hands-on/myscript-bash.sh 7")
```

```
HELLO! I AM A BASH SCRIPT
```

```
THE VALUE INSERTED IS: 7
```

```
IN THIS DIRECTORY WE HAVE THE FOLLOWING FILES:
```

```
total 572K
-rw-r--r-- 1 alessio 12K nov 12 16:26 lecture-2.tex
-rw-r--r-- 1 alessio 65K nov 12 16:24 lecture-2.log
-rw-r--r-- 1 alessio 439K nov 12 16:24 lecture-2.pdf
-rw-r--r-- 1 alessio 3,7K nov 12 16:24 lecture-2.nav
-rw-r--r-- 1 alessio 463 nov 12 16:24 lecture-2.toc
-rw-r--r-- 1 alessio 426 nov 12 16:24 lecture-2.out
-rw-r--r-- 1 alessio 9,6K nov 12 16:24 lecture-2.aux
drwxr-xr-x 2 alessio 20K nov 12 16:24 _minted-lecture-2
-rw-r--r-- 1 alessio 0 nov 12 16:24 lecture-2.snm
-rw-r--r-- 1 alessio 40 nov 12 16:24 lecture-2.vrb
drwxrwxr-x 2 alessio 4,0K nov 12 16:15 hands-on
0
```

# Running external software in Python

## Using the os module

```
import os
```

```
os.system("bash hands-on/myscript-bash.sh 7")
```

HELLO! I AM A BASH SCRIPT

THE VALUE INSERTED IS: 7

IN THIS DIRECTORY WE HAVE THE FOLLOWING FILES:

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total 572K
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-rw-r--r-- 1 alessio 40 nov 12 16:24 lecture-2.vrb
drwxrwxr-x 2 alessio 4,0K nov 12 16:15 hands-on
0
```

## Using the subprocess module

```
import subprocess as subproc
```

```
out_stat = subproc.run(["bash", \
                        "hands-on/myscript-bash.sh", \
                        "7"])
```

```
print("\nThe exit code was %d" % out_stat.returncode)
```

SAME OUTPUT AS WITH `os` MODULE, PLUS:

The `exit` code was 0

## Redirect of the output

```
import subprocess as subproc
```

```
out_stat = subproc.run(["bash", \
                        "hands-on/myscript-bash.sh", \
                        "7"], \
                        stdout=subprocess.DEVNULL)
```

```
print("\nThe exit code was %d" % out_stat.returncode)
```

THE OUTPUT BECOMES:

The `exit` code was 0

# The NumPy module

---



## What is NumPy?

- Is a Python library used for **working** (mainly) **with arrays**.



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- In Python lists serve as arrays, but they are slow to process. A NumPy array is up to **50x faster** than lists.



## What is NumPy?

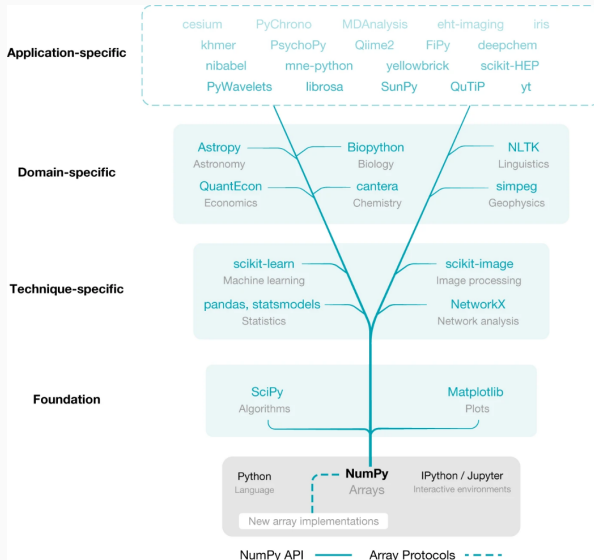
- Is a Python library used for **working** (mainly) **with arrays**.
- In Python lists serve as arrays, but they are slow to process. A NumPy array is up to **50x faster** than lists.
- NumPy arrays (unlike lists) are stored at one continuous place in memory (locality of reference)  $\Rightarrow$  very efficient access & manipulation.



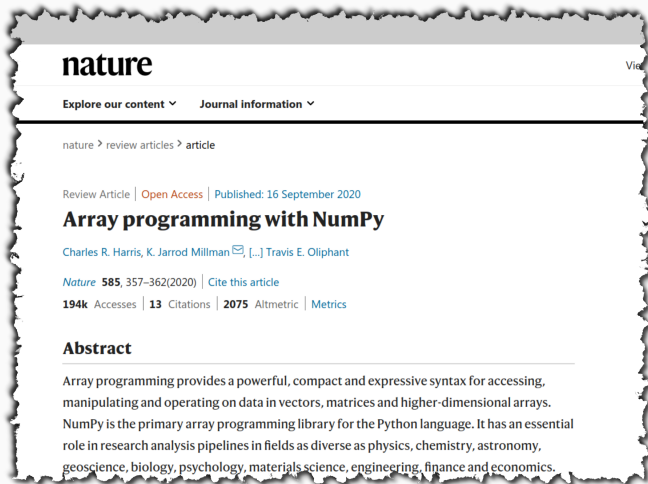
## What is NumPy?

- Is a Python library used for **working** (mainly) **with arrays**.
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- NumPy arrays (unlike lists) are stored at one continuous place in memory (locality of reference)  $\Rightarrow$  very efficient access & manipulation.
- NumPy is written **partially** in Python, but **most of the parts requiring fast computation are written in C or C++**.

# Numerical Python (NumPy)







# Arrays & Main operations

## Hands-on

```
>>> import numpy as np
```

3 columns

4 rows

|   |    |    |
|---|----|----|
| 0 | 1  | 2  |
| 3 | 4  | -1 |
| 6 | 7  | 10 |
| 9 | 10 | 0  |

### DATA

|   |   |   |   |   |    |   |   |    |   |    |   |
|---|---|---|---|---|----|---|---|----|---|----|---|
| 0 | 1 | 2 | 3 | 4 | -1 | 6 | 7 | 10 | 9 | 10 | 0 |
|---|---|---|---|---|----|---|---|----|---|----|---|

24 bytes

### DATA TYPE

8 byte int

### SHAPE

(4,3) (rows,columns)

### STRIDES

(24,8) (jump row,jump col)

# Arrays & Main operations

## Hands-on

```
>>> import numpy as np
>>> # arrays must contain elements of the same type
>>> a = np.array([1, 3.4, 'test'])
>>> a
array(['1', '3.4', 'test'], dtype='<U32')
>>> a = np.array([1, 3.4, -5.+3j])
>>> a
array([ 1. +0.j,  3.4+0.j, -5. +3.j])
```

## Note

Contrary to lists, NumPy arrays must contain objects of the **same type!**

# Arrays & Main operations

## Hands-on

```
>>> import numpy as np
>>> # setting up a list
>>> a = [1, 6, 'hello', -6.5, True]
>>> # converting to numpy array
>>> b = np.asarray(a)
>>> # printing the result
>>> a
[1, 6, 'hello', -6.5, True]
>>> type(a)
<class 'list'>
>>> b
array(['1', '6', 'hello', '-6.5', 'True'], dtype='<U21')
>>> type(b)
<class 'numpy.ndarray'>
```

## Good to know

It is possible to convert a list into a NumPy array using the method `.asarray()`. It is also possible to convert a NumPy array to a list using the method `.tolist()` (or, equivalently, `list()`).

# Arrays & Main operations

## Slicing & masking

Slicing works like for lists, but we can apply it to **multiple dimensions**.

`a[:2, ::2]`

`a[:, 1]`

`a[2:, 2:]`

|   |    |    |   |
|---|----|----|---|
| 0 | 1  | 2  | 3 |
| 3 | 4  | -1 | 0 |
| 6 | 7  | 10 | 9 |
| 9 | 10 | 0  | 4 |

# Arrays & Main operations

## Slicing & masking

```
>>> # creating a 4x4 matrix
>>> a = np.array([[0,1,2,3],[3,4,-1,0],\
...               [6,7,10,9],[9,10,0,4]])
>>> a
array([[ 0,  1,  2,  3],
       [ 3,  4, -1,  0],
       [ 6,  7, 10,  9],
       [ 9, 10,  0,  4]])
>>> # returning only elements > 5
>>> a[a>5]
array([ 6,  7, 10,  9,  9, 10])
```

$a[a > 5]$

|   |    |    |   |
|---|----|----|---|
| 0 | 1  | 2  | 3 |
| 3 | 4  | -1 | 0 |
| 6 | 7  | 10 | 9 |
| 9 | 10 | 0  | 4 |

# Arrays & Main operations

## Vectorization

```
>>> a = np.array([[0,1],[3,4],[6,7],[9,10]])
>>> b = np.array([[1,1],[1,1],[1,1],[1,1]])
>>> c = a + b
>>> c
array([[ 1,  2],
       [ 4,  5],
       [ 7,  8],
       [10, 11]])
```

|   |    |   |   |   |   |    |    |
|---|----|---|---|---|---|----|----|
| 0 | 1  | + | 1 | 1 | = | 1  | 2  |
| 3 | 4  |   | 1 | 1 |   | 4  | 5  |
| 6 | 7  |   | 1 | 1 |   | 7  | 8  |
| 9 | 10 |   | 1 | 1 |   | 10 | 11 |

# Arrays & Main operations

## Broadcasting

```
>>> # to have a column vector
>>> # I have to set it in this way
>>> a = np.array([[0],[3],[6],[9]])
>>> b = np.array([0,1])
>>> c = a * b
>>> c
array([[0, 0],
       [0, 3],
       [0, 6],
       [0, 9]])
```

## Note

There are **more pythonic** ways to *transpose* a row array into a column one. For instance:

```
>>> a = np.array([0,3,6,9])
>>> a = a[:, np.newaxis]
array([[0],
       [3],
       [6],
       [9]])
```

|   |   |
|---|---|
| 0 | 0 |
| 3 | 3 |
| 6 | 6 |
| 9 | 9 |

 X 

|   |   |
|---|---|
| 0 | 1 |
| 0 | 1 |
| 0 | 1 |
| 0 | 1 |

 = 

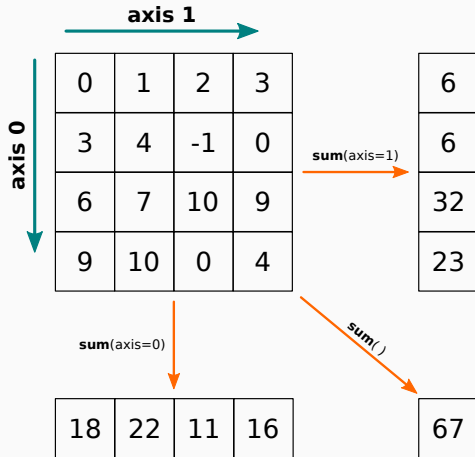
|   |   |
|---|---|
| 0 | 0 |
| 0 | 3 |
| 0 | 6 |
| 0 | 9 |



# Arrays & Main operations

## Reduction

```
>>> # define a matrix
>>> a = np.array([[0,1,2,3],[3,4,-1,0],\
...               [6,7,10,9],[9,10,0,4]])
>>> a
array([[ 0,  1,  2,  3],
       [ 3,  4, -1,  0],
       [ 6,  7, 10,  9],
       [ 9, 10,  0,  4]])
>>> # sum over axis 0
>>> s1 = a.sum(axis=0)
>>> # sum over axis 1
>>> s2 = a.sum(axis=1)
>>> # sum over axis 0 first and 1 then
>>> s3 = a.sum()
>>> # displaying results
>>> s1
array([18, 22, 11, 16])
>>> s2
array([ 6,  6, 32, 23])
>>> s3
67
```



# Arrays & Main operations

Evaluating mathematical expressions with NumPy is quite easy!

## Mean square error

$$\langle \varepsilon^2 \rangle = \frac{1}{N} \sum_{i=1}^N (X_i^* - X_i)^2$$

$N \rightarrow$  Number of measures.

$X^* \rightarrow$  Expected value.

$X \rightarrow$  Measure.

# Arrays & Main operations

C

```
#include <stdlib.h>
#include <stdio.h>

float myexpt[4] = {1., 3.5, -6., 0.4};
float myvals[4] = {1.05, 2.8, -4., -0.3};
float dummy, v1, v2, avg_sq_err;

int N = sizeof(myexpt)/sizeof(float); // 4
int i;

int main(){

    dummy = 0.;

    for(i=0; i<N; i++)
    {
        v1 = myexpt[i];
        v2 = myvals[i];
        dummy += (v1-v2)*(v1-v2);
    }
```

```
avg_sq_err = (1./((float) N))*dummy;

printf("expect \t measures\n");

for(i=0; i<N; i++)
{
    printf("%.2f\t%.2f\n", myexpt[i], myvals[i]);
}

printf("mean quadratic error = %.4f\n", avg_sq_err);

return 0;
}
```

# Arrays & Main operations

## Python

```
1  myexpt = [1., 3.5, -6., 0.4]
2  myvals = [1.05, 2.8, -4., -0.3]
3
4  N = len(myexpt)
5
6  dummy = 0.
7
8  for i in range(N):
9      v1 = myexpt[i]
10     v2 = myvals[i]
11     dummy += (v1-v2)*(v1-v2)
12
13  avg_sq_err = (1./N)*dummy
14
15  print('expectations = ', myexpt)
16  print('measures = ', myvals)
17  print('mean quadratic error = %.4f'\
18        % avg_sq_err)
```

## NumPy

```
1  import numpy as np
2
3  myexpt = np.array([1., 3.5, -6., 0.4])
4  myvals = np.array([1.05, 2.8, -4., -0.3])
5
6  N = np.shape(myexpt)[0]
7
8  avg_sq_err = ((1./N)
9               * np.sum(np.square(myexpt-myvals)))
10
11  print('expectations = ', myexpt)
12  print('measures = ', myvals)
13  print('mean quadratic error = %.4f'\
14        % avg_sq_err)
```

## Cheat Sheets

Together with the course materials, there are also the cheat sheets for NumPy and SciPy.

# Useful NumPy's functions

## shape & reshape

```
>>> import numpy as np
>>> # find the shape of an array
>>> a = np.array([[1,2,3],[4,5,6]])
>>> a.shape
(2, 3)
>>> np.shape(a)
(2, 3)
>>> # reshape of an array
>>> a = np.array([1,2,3,4,5,6])
>>> a.reshape(2,3)
array([[1, 2, 3],
       [4, 5, 6]])
```

# Useful NumPy's functions

## zeros & ones

```
>>> import numpy as np
>>> # create an array full of zeros
>>> a = np.zeros(5)
>>> a
array([0., 0., 0., 0., 0.])
>>> # create an array full of ones
>>> a = np.ones(5)
>>> a
array([1., 1., 1., 1., 1.])
```

## Note

There are two functions called `zeros_like()` and `ones_like()` which generate arrays of zeros and ones with a shape **identical** to another array (e.g., (3,4)).

# Useful NumPy's functions

## arange & linspace

Both return sequences of evenly spaced numbers, but they are not the same thing.

```
>>> import numpy as np
>>> # arange(start, stop, step)
>>> # (stop not included)
>>> a = np.arange(0,10,2)
>>> a
array([0, 2, 4, 6, 8])
>>> # linspace(start, stop, nrpoints)
>>> # (stop is included)
>>> a = np.linspace(0,10,2)
>>> a
array([ 0., 10.] )
```

## Note

There exist two logarithmic counterparts of `linspace()` called `logspace()` and `geomspace()`.



# Useful NumPy's functions

## mean, std, & var

```
>>> import numpy as np
>>> # generate an array of size 10 filled of random numbers in [0,1)
>>> a = np.random.random(10)
>>> a
array([0.82191943, 0.8119978 , 0.93116978, 0.74907317, 0.35424478,
        0.41143222, 0.1770835 , 0.98988076, 0.30800019, 0.37610524])
>>> # arithmetic mean
>>> np.mean(a)
0.593090686506173
>>> # standard deviation
>>> np.std(a)
0.28068161984733503
>>> # variance
>>> np.var(a)
0.0787821717201239
```

## Note

There exist similar functions to account for the presence of NaN values called `nanmean()`, `nanstd()`, and `nanvar()`.

# Useful NumPy's functions

## histogram

```
>>> import numpy as np
>>> # generate a histogram with three bins
>>> np.histogram([1, 2, 1], bins=[0, 1, 2, 3])
(array([0, 2, 1]), array([0, 1, 2, 3]))
>>> # you can feed multiple arrays to the histogram
>>> np.histogram([[1, 2, 1], [1, 0, 1]], bins=[0,1,2,3])
(array([1, 4, 1]), array([0, 1, 2, 3]))
>>> # compute a "kernel density estimator" (KDE)
>>> hist, bin_edges = np.histogram(np.random.randint(0,4,size=100), bins=np.linspace(0,4,10),\
...                               density=True)
>>> hist
array([0.585 , 0.      , 0.4725, 0.      , 0.63  , 0.      , 0.5625, 0.      , 0.      ])
>>> bin_edges
array([0.      , 0.44444444, 0.88888889, 1.33333333, 1.77777778,
       2.22222222, 2.66666667, 3.11111111, 3.55555556, 4.      ])
>>> hist.sum()
2.25
>>> # computing the "area under the curve"
>>> np.sum(hist * np.diff(bin_edges))
1.0
```

## Warning

In previous versions of the NumPy (< 1.6), the normed histogram did not compute well the probability density. Remember: **ALWAYS BENCHMARK THE CODE!**

# Useful NumPy's functions

## savetxt

```
>>> a = np.array([[1,2,3,5],[2,4,10,1],\
...               [0,30,5,-1],[16,8,20,20]])
>>> np.savetxt('hands-on/ex_np_savetxt.dat',
...            a, fmt='%.1f', delimiter=';')
```

```
alessio@aleurv:~$ less ex_np_savetxt.dat
```

```
1.0;2.0;3.0;5.0
2.0;4.0;10.0;1.0
0.0;30.0;5.0;-1.0
16.0;8.0;20.0;20.0
```

## loadtxt

```
>>> b = np.loadtxt('hands-on/ex_np_savetxt.dat',
...                delimiter=';', usecols=(1,3))
>>> b
array([[ 2.,  5.],
       [ 4.,  1.],
       [30., -1.],
       [ 8., 20.]])
```

## Good to know

The `loadtxt()` function is quite versatile as it allows to define **comments**, **delimiters**, **rows to skip**, and **columns to use**. Similarly, the `savetxt()` function allows to define also **formats**, **headers**, and **footers**.

# Hands-on Session



# Exercise 1

## Task

1. Compute  $\int_a^b \alpha \sin(x) e^{-\beta x} dx$  using the **Monte Carlo method**.

$$I = \int_a^b g(x) dx \simeq \langle I_N \rangle.$$

Being

$$\langle I_N \rangle \equiv \frac{1}{N} \sum_{i=1}^N I_i = \frac{1}{N} \left[ (b-a) \sum_{i=1}^N g(x_i) \right],$$

with  $x_i$  a random number in the  $[a, b]$  range.

2. Print the solution and its accuracy (in absolute value and in %) as a function of the size of the sample,  $N$ .
3. Parameters' values:  $a = 0$ ,  $b = 2\pi$ ,  $\alpha = 2.5$ ,  $\beta = 1$ , and  $N \in \{10, 100, 1000, 10000, 100000, 1000000, 10000000\}$ .

# Exercise 1

## Random numbers in $[a, b]$ range

Use the function

```
numpy.random.uniform().
```

## Compute the exact solution

```
from scipy import integrate
```

```
val, err = integrate.quad(myfunc, val_a, val_b)
```

## Tips

- Leverage the *power* of NumPy arrays!
- Use a *function* to define the integrand.
- NumPy has built-in  $\sin(x)$  (`numpy.sin()`) and  $\exp(x)$  (`numpy.exp()`) functions as well as the  $\pi$  constant (`numpy.pi`).

• <https://towardsdatascience.com/monte-carlo-integration-in-python-a71a209d277e>

• [http://people.duke.edu/~ccc14/sta-663-2016/15C\\_MonteCarloIntegration.html](http://people.duke.edu/~ccc14/sta-663-2016/15C_MonteCarloIntegration.html)

## Exercise 2

### Task

1. Compute the **Shannon entropy**,  $S$ , of a text using the following relation:

$$S = -\frac{1}{\log_2 N_W} \sum_{i=1}^{N_W} p_i \log_2 p_i .$$

Where  $N_W$  is the number of distinct words in the text. The probability of extracting word  $i$  uniformly at random from the text,  $p_i$  is equal to:

$$p_i = \frac{N_i}{L_{TOT}} ,$$

with  $N_i$  the number of occurrences of word  $i$  and  $L_{TOT}$  the total length of the text.

- [https://en.wikipedia.org/wiki/Entropy\\_\(information\\_theory\)](https://en.wikipedia.org/wiki/Entropy_(information_theory))

## Exercise 2

### Task (continuing)

2. For a given text, compute the following quantities:
  - The number of distinct words,  $N_W$ .
  - The total length of the text,  $L_{TOT}$ .
  - The entropy of the text,  $S$ .
  - The number of times each word appears (stored as a dictionary).
  - The lists of the top 5% most frequent and least frequent words (two distinct lists).
3. Print the results on screen and store them in a JSON file.
4. Repeat points 1-3 for all the text files available.



## Exercise 2

### Tips

- Use **list slicing** to remove the file extension from file names.
- Not all the lines in the text are useful. Find a way to understand where the “interesting” part of the text is, and tell Python which rows to parse.
- Try to get rid of the punctuation and other “undesired” characters (e.g., ‘-’ or ‘\n’) from words.
- NumPy has a way to compute the **percentile** of a set of values.
- Print the JSON using the “pretty print” style.

### Split string into “words”

```
fin = open(filename, 'r')
for line in fin:
    # splitting using space as separator
    words = line.split(" ")
```






### Remove punctuation

```
import string

str_translator = str.maketrans('', '',
                                string.punctuation)

line = line.translate(str_translator)
```

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