

Python

Handling vectors with NumPy

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Numpy ?

- NumPy (numerical python) is a package for scientific computing. It provides tools for handling n-dimensional arrays (especially vectors and matrices).
- The objects are all the same type into a NumPy arrays structure
- The package offers a large number of routines for fast access to data (e.g. search, extraction), for various manipulations (e.g. sorting), for calculations (e.g. statistical computing)
- Numpy arrays are more efficient (speed, volume management) than the usual Python collections (list, tuple).
- Numpy arrays are underlying to many packages dedicated to scientific computing in Python.
- Note that a vector is actually a 1 single dimension array

To go further, see the reference manual (used to prepare this slideshow).

<http://docs.scipy.org/doc/numpy/reference/index.html>

Creation on the fly, generation of a sequence, loading from a file

CREATING A NUMPY VECTOR

First, we must import the module “numpy”

```
import numpy as np
```

np is the alias used for accessing to the routines of the package 'numpy'.

Converting Python array_like objects (e.g. list)

```
a = np.array([1.2,2.5,3.2,1.8])
```

[] is a list of values (float)

Information about the structure

```
#object type  
print(type(a)) #<class 'numpy.ndarray'>  
  
#data type  
print(a.dtype) #float64  
  
#number of dimensions  
print(a.ndim) #1 (we have 2 if it is a matrix, etc.)  
  
#number of rows and columns  
print(a.shape) #(4,) → tuple! 4 elements for the 1st dim (n°0)  
  
#total number of elements  
print(a.size) #4, nb.rows x nb.columns if a matrix
```

Setting the data type

Specifying the data type
can be implicit or explicit

```
#creating a vector – implicit typing
a = np.array([1,2,4])
print(a.dtype) #int32
```

Creating an array with
objects of non-standard
type is possible

```
#creating a vector – explicit typing – preferable !
a = np.array([1,2,4],dtype=float)
print(a.dtype) #float64
print(a) #[1. 2. 4.]
```

```
#a vector of Boolean values is possible
b = np.array([True,False,True,True], dtype=bool)
print(b) #[True False True True]
```

```
# the array value may be an object
a = np.array([{"Toto":(45,2000)},{"Tata":(34,1500)}])
print(a.dtype) #object
```

Creating sequence of numbers

#evenly spaced values within a given interval (step = 1 here)

```
a = np.arange(start=0,stop=10)
```

```
print(a) #[0 1 2 3 4 5 6 7 8 9], the last value is excluded
```

#specifying the step property

```
a = np.arange(start=0,stop=10,step=2)
```

```
print(a) #[0 2 4 6 8]
```

#evenly spaced value, specify the number of elements

```
a = np.linspace(start=0,stop=10,num=5)
```

```
print(a) #[0. 2.5 5. 7.5 10.], the last value is included here
```

#repeating 5 times the value 1 – number of values = 5 (1 dimension)

```
a = np.ones(shape=5)
```

```
print(a) # [1. 1. 1. 1. 1.]
```

#repeating 5 times (1 dimension) the value 3.2

```
a = np.full(shape=5,fill_value=3.2)
```

```
print(a) #[3.2 3.2 3.2 3.2 3.2]
```

Loading a vector from a data file

The values can be stored in a text file (`loadtxt` for reading, `savetxt` for writing)

```
#loading from a text file  
#we can set the type of the data  
a = np.loadtxt("vecteur.txt",dtype=float)  
print(a) #[4. 5. 8. 16. 68. 14. 35.]
```

Only 1 column here

1	4
2	5
3	8
4	16
5	68
6	14
7	35

Note: If necessary, we change the default directory with the function `chdir()` from the `os` module (that must be imported)

We can convert a Python sequence type in a “numpy” array

```
#lst is a list of values (float)  
lst = [1.2,3.1,4.5]  
print(type(lst)) #<class 'list'>  
#converting the list  
a = np.asarray(lst,dtype=float)  
print(type(a)) #<class 'numpy.ndarray'>  
print(a) #[1.2 3.1 4.5]
```

Adding and removing elements

Add a value in last position

```
#a is a vector  
a = np.array([1.2,2.5,3.2,1.8])  
#append the value 10 into the vector a  
a = np.append(a,10)  
print(a) #[1.2 2.5 3.2 1.8 10.]
```

Remove a value from its index

```
#remove the value n°2  
b = np.delete(a,2) #a range of indices can be used  
print(b) #[1.2 2.5 1.8 10.]
```

Modify the size of a vector

```
a = np.array([1,2,3])  
#adding two cells  
#fills zero for the new cell  
a.resize(new_shape=5)  
print(a) #[1 2 3 0 0]
```

Concatenation of vectors

```
#concatenate 2 vectors  
x = np.array([1,2,5,6])  
y = np.array([2,1,7,4])  
z = np.append(x,y)  
print(z) #[1 2 5 6 2 1 7 4]
```

Indexing with indices or Boolean array

EXTRACTING VALUES

Indexed access – `v = np.array([1.2,7.4,4.2,8.5,6.3])`

#printing all the values

```
print(v)
```

#or

```
print(v[:]) # note the role of : ; here, from start to end
```

#indexed access - first value

```
print(v[0]) # 1.2 – the first index is 0 (zero)
```

#last value

```
print(v[v.size-1]) #6.3, v.size is okay because v is a vector
```

#contiguous indices

```
print(v[1:3]) # [7.4 4.2]
```

#extreme values, start to 3 (not included)

```
print(v[:3]) # [1.2 7.4 4.2]
```

#extreme values, 2 to end

```
print(v[2:]) # [4.2 8.5 6.3]
```

#negative indices

```
print(v[-1]) # 6.3, last value
```

#negative indices

```
print(v[-3:]) # [4.2 8.5 6.3], 3 last values
```

Note : Apart from
singletons, the
generated vectors are of
type `numpy.ndarray`

Generic writing of indices is : `first:last:step`
`last` is not included

```
#value n°1 to n°3 with a step = 1  
print(v[1:4:1]) # [7.4, 4.2, 8.5]
```

```
#step = 1 is implicit  
print(v[1:4]) # [7.4, 4.2, 8.5]
```

```
#n°0 to n°2 with a step = 2  
print(v[0:3:2]) # [1.2, 4.2]
```

```
#the step can be negative, n°3 to n°1 with a step = -1  
print (v[3:0:-1]) # [8.5, 4.2, 7.4]
```

```
#we can use this idea (negative step) to reverse a vector  
print(v[::-1]) # [6.3, 8.5, 4.2, 7.4, 1.2]
```

Boolean indexing – `v = np.array([1.2,7.4,4.2,8.5,6.3])`

```
#extraction with a vector of Booleans  
#if b too short, the remainder is considered False  
b = np.array([False,True,False,True,False],dtype=bool)  
print(v[b]) # [7.4 8.5]
```

```
#one can use a condition for extraction  
print(v[v < 7]) # [1.2 4.2 6.3]
```

```
#because a condition generates a vector of Booleans  
b = v < 7  
print(b) # [True False True False True]  
print(type(b)) # <class 'numpy.ndarray'>
```

```
#one can use also the extract() function  
print(np.extract(v < 7, v)) # [1.2 4.2 6.3]
```

```
#get the max value  
print(np.max(v)) # 8.5
```

```
#find the index of the max value  
print(np.argmax(v)) # 3
```

```
#sort the values  
print(np.sort(v)) # [1.2 4.2 6.3 7.4 8.5]
```

```
#get the indices that would sort the values  
print(np.argsort(v)) # [0 2 4 1 3]
```

```
#unique elements of the vector  
a = np.array([1,2,2,1,1,2])  
print(np.unique(a)) # [1 2]
```

Note : The equivalent exists for `min()`

STATISTICAL ROUTINES

Statistical functions – `v = np.array([1.2,7.4,4.2,8.5,6.3])`

```
#mean  
print(np.mean(v)) # 5.52
```

```
#median  
print(np.median(v)) # 6.3
```

```
#variance  
print(np.var(v)) # 6.6856
```

```
#percentile  
print(np.percentile(v,50)) #6.3 (50% = médiane)
```

```
#sum  
print(np.sum(v)) # 27.6
```

```
#cumulative sum  
print(np.cumsum(v)) # [1.2 8.6 12.8 21.3 27.6]
```



The statistical functions are not numerous, we will need SciPy (and other)

Calculations between vectors – “Elementwise” operations

```
#two vectors : x and y  
x = np.array([1.2,1.3,1.0])  
y = np.array([2.1,0.8,1.3])  
#multiplication  
print(x*y) # [2.52 1.04 1.3]  
#addition  
print(x+y) # [3.3 2.1 2.3]  
#multiplication by a scalar  
print(2*x) # [2.4 2.6 2.]
```

#comparison of vectors

```
x = np.array([1,2,5,6])  
y = np.array([2,1,7,4])  
b = x > y  
print(b) # [False True False True]
```

#logical operations

```
a = np.array([True,True,False,True],dtype=bool)  
b = np.array([True,False,True,False],dtype=bool)
```

#AND operator

```
np.logical_and(a,b) # [True False False False]
```

#XOR operator (exclusive or)

```
np.logical_xor(a,b) # [False True True True]
```

The calculations are made in the element wise fashion - We have the same principle under R.

The list of functions is long.
See :

<http://docs.scipy.org/doc/numpy/reference/routines.logic.html>

```
x = np.array([1.2,1.3,1.0])  
y = np.array([2.1,0.8,1.3])
```

```
#dot product of two vectors  
z = np.vdot(x,y)  
print(z) # 4.86
```

```
#or, equivalently  
print(np.sum(x*y)) # 4.86
```

```
#vector norm  
n = np.linalg.norm(x)  
print(n) # 2.03
```

```
#or, equivalently  
import math  
print(math.sqrt(np.sum(x**2))) # 2.03
```

The functions for matrix operations exist, some of them can be applied to vectors

```
#set routines
```

```
x = np.array([1,2,5,6])  
y = np.array([2,1,7,4])
```

```
#intersection
```

```
print(np.intersect1d(x,y)) # [1 2]
```

```
#union – this is not a concatenation
```

```
print(np.union1d(x,y)) # [1 2 4 5 6 7]
```

```
#difference i.e. values in x but not in y
```

```
print(np.setdiff1d(x,y)) # [5 6]
```

A vector of values (especially integer) can be considered as a set of values.

Course materials (in French)

http://eric.univ-lyon2.fr/~ricco/cours/cours_programmation_python.html

Python website

Welcome to Python - <https://www.python.org/>

Python 3.4.3 documentation - <https://docs.python.org/3/index.html>

NumPy Manual

[Numpy User Guide](#) and [Numpy Reference](#)

POLLS (KDnuggets)

Data Mining / Analytics Tools Used

Python, 4th in 2015

Primary programming language for Analytics, Data Mining, Data Science tasks

Python, 2nd in 2015 (next R)