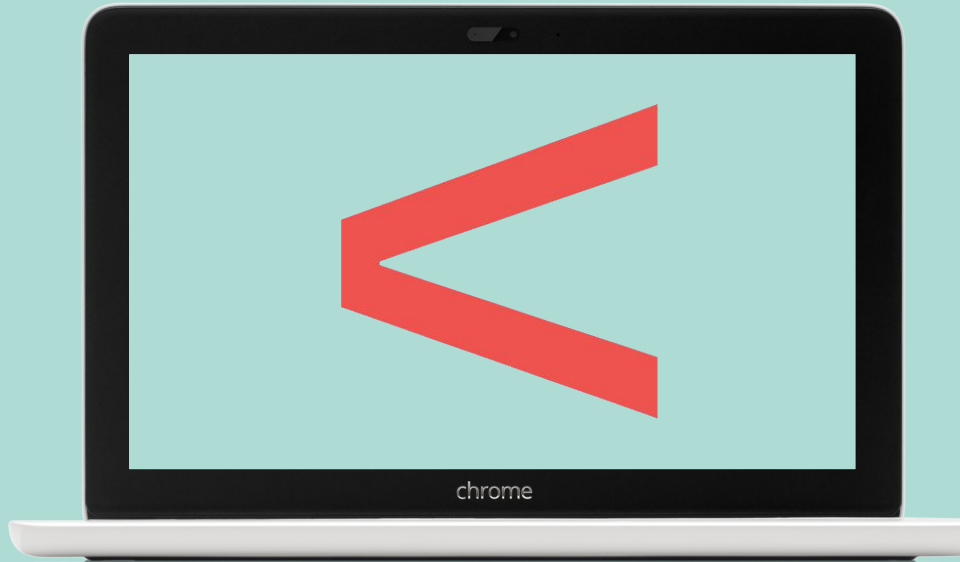




Python libraries: NumPy



What is NumPy?

NumPy is the fundamental library of python, used to perform scientific computing on **numerical data**.

NumPy provides **NumPy arrays**, a data structure that stores values of the same type that are indexed and allow to **perform calculations** across them. Think of a NumPy array as rows and columns of table.

What can NumPy do?



Inspecting and describing your array



Performing mathematical operations: statistics, arithmetics, comparison and aggregate functions



Generating random values



Copying and sorting arrays



Subsetting, slicing and indexing



Manipulating arrays



Visualizing (matplotlib and seaborn)



Among other things

Subsetting and slicing with Python

String:

“h	e	l	l	o		w	o	r	l	d	!”
0	1	2	3	4	5	6	7	8	9	10	11
-12	-11	-10	-9	-8	-7	-6	-5	-4	-3	-2	-1

Subsetting: selecting a single element

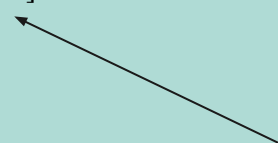
```
first_letter = hello[ 0 ]
```

“h”

Slicing: selecting or several elements

```
first_word = hello[ 0 : 5 ]
```

“hello”



Notice how we select one element up to the one we want

Subsetting and slicing with Python

List:

["apple",	"orange",	"melon",	"pear",	"apple",	"lemon",	"grape"]
0	1	2	3	4	5	6

Subsetting: selecting a single element

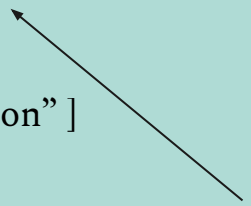
```
third_fruit = fruit[ 2 ]
```

"melon"

Slicing: selecting several elements

```
several_fruit = hello[ 0 : 3 ]
```

["apple", "orange", "melon"]



Notice how we select one element up to the one we want

What are the differences between Python lists and NumPy arrays?

Python list	NumPy array
Lists cannot directly handle math operations	You can perform calculations over entire arrays
List don't need to be declared	Arrays need to be declared with <code>np.array()</code> to be created
Takes up a lot of memory space and execution time is slower	Consumes less memory and is faster at performing operations
Only have one dimension	Can have N-dimensions
Allows several data types	Not optimal for different data types

Dimensions in arrays - n dimensions

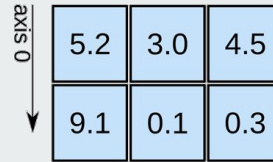
1D array



axis 0 →

shape: (4,)

2D array

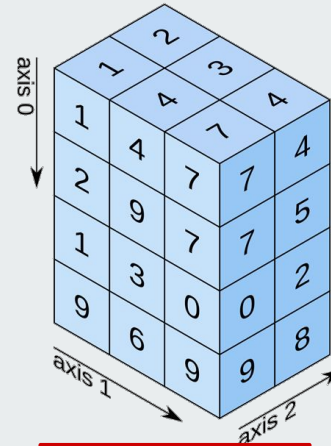


axis 0 ↓

axis 1 →

shape: (2, 3)

3D array



axis 0 ↓

axis 1 →

axis 2 →

shape: (4, 3, 2)

How to create an array from scratch?

1D array:

```
onedim_arr = np.array([ 1, 2, 3])
```

Result: array([1, 2, 3])

2D array:

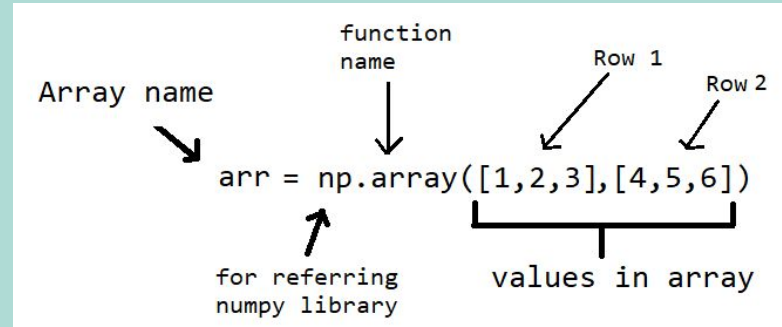
```
twodim_arr = np.array([[ 1, 2, 3],[ 4, 5, 6]])
```

Result: array([[1, 2, 3],
[4, 5, 6]])

3D array:

```
threedim_arr = np.array([[[ 1, 2, 3],[ 4, 5, 6]],[[ 7, 8, 9],[ 10, 11, 12]]])
```

Result: array([[[1, 2, 3],
[4, 5, 6]],
[[[7, 8, 9],
[10, 11, 12]]]])



How to create an array from a table?

Fruit	Apple	Grape	Banana	Melon	Lemon	Orange
Weight	10	2	3.5	4	1	6
Price	1.25	2.30	4.0	1.80	2.25	3.45
Discount	10	5	15	25	20	30

```
array([[ 10,      2,  3.5,   4,    1,   6 ],  
       [ 1.25,  2.3,   4,  1.8,  2.25, 3.45 ],  
       [ 10,    5,  15,  25,   20,  30 ] ])
```

← Weight row

← Price row

← Discount row

How to know the size, shape and number of dimensions?

- **arr.ndim:** number of axes, or dimensions, of the array.
- **arr.size:** total number of elements of the array. This is the product of the elements of the array's shape.
- **arr.shape:** tuple of integers indicating the number of elements stored along each dimension of the array
 - This function is particularly useful when cleaning the data, as it will help you understand how many rows or columns have been dropped

Subsetting arrays

```
array([[ 10,      2,  3.5,   4,    1,    6 ],
       [ 1.25,   2.3,   4,   1.8,  2.25, 3.45 ],
       [ 10,     5,  15,  25,   20,  30 ] ])
```

← Weight row

← Price row

← Discount row

Subsetting: selecting a single element

```
weight_list = arr[ 0 ]
```

```
[ 10, 2, 3.5, 4, 1, 6 ]
```

```
third_weight = arr[ 0, 2 ]
```

```
[ 3.5 ]
```

Slicing arrays: rows

```
array([[ 10,      2,  3.5,   4,    1,    6 ],  
       [ 1.25,   2.3,   4,   1.8,  2.25, 3.45 ],  
       [ 10,     5,  15,  25,   20,   30 ] ])
```

← Weight row

← Price row

← Discount row

Slicing: selecting or several elements

```
price_discount = arr[ 1 : ]
```

```
[[ 1.25, 2.3,  4, 1.8, 2.25, 3.45 ],  
 [ 10,  5, 15, 25,  20,  30 ] ]
```

Slicing arrays: columns

```
array([ [ 10,      2,  3.5,   4,      1,   6 ],
       [ 1.25,   2.3,   4,   1.8,  2.25, 3.45 ],
       [ 10,      5,  15,   25,   20,   30 ] ])
```

If columns are not next to each other:

look at the brackets and comma in the syntax

```
first_third_columns= arr[:, [ 0,2 ]]
```

```
[ [ 10.  3.5 ]
  [ 1.25  4. ]
  [10.  15. ]]
```

If columns are contiguous:

there no brackets and you separate with column

```
fourth_fifth_columns=arr[:, 3:5 ]
```

```
[ [ 4.    1. ]
  [ 1.8  2.25]
  [25.  20. ]]
```

Algebraic operations with operators on single array

```
twice_arr = arr * 2
```

```
array([ [ 20.,  4.,  7.,  8.,  2., 12. ],  
       [ 2.5, 4.6,  8.,  3.6, 4.5, 6.9 ],  
       [20., 10., 30., 50., 40., 60. ]])
```

Algebraic operations with operators between two arrays

```
sum_of_arrs = arr + twice_arr
```

```
array([ [ 30. ,  6. , 10.5, 12. ,  3. , 18. ],  
       [ 3.75,  6.9, 12. ,  5.4,  6.75, 10.35],  
       [ 30. , 15. , 45. , 75. , 60. , 90. ]])
```

Aggregation functions

Find max value:

`arr.max()`

30.0

Find min value:

`arr.min()`

1.0

Find average value:

`data.mean()`

8.141666666666667

Check the documentation for more examples

Sorting and concatenating

```
a = np.array( [4, 2, 1, 3] )
```

```
b = np.array( [5, 6, 7, 8] )
```

Sorting:

```
a = np.sort( a )
```

```
array( [1, 2, 3, 4] )
```

Concatenate:

```
new_arr = np.concatenate((a, b))
```

```
array( [4, 2, 1, 3, 5, 6, 7, 8] )
```

Always read the documentation!