	ython and scientific computation: navigating in a NumPy array
M	earn to navigate in a NumPy array which is a powerful data structure and particularly efficient for scientific computing. lain concepts covered: 1. Indexing: navigating an array from indexes. 2. Slicing/Subsetting: extraction of a portion of an array. 3. Broadcasting: broadcasting operation between arrays.
į	4. Selection/Masking: verification of a condition for each item of an array. 5. Boolean indexing: conditional navigation in an array (selection of items that satisfy a condition). stline]
# 1 img	exing Slicing Broadcasting Masking et Boolean indexing importation de NumPy port numpy as np dexing
S	his is an array browsing mechanism, based on the use of indexes to search for elements (in an array). he general principle is to specify in parameters the respective values of the dimensions corresponding to the required element. yntax: • For a 1 dimension array (column vector): my_arr[index] • For other dimensions: my_arr[index_dim1, index_dim2,]
H	int: egative indexes can be used to specify navigation from the end: index -1 corresponds to the last element of the corresponding dimension (i.e. the last row or last column in a 2-dimension array), index -2 corresponds to the second last one,
-1	2 -1 2 -1 2 -1 2 my_arr[axis1] my_arr[0] my_arr[1] my_arr[2] my_arr[-2] my_arr[-1] 0 1 2 3 4 0 1 2 3 4 0 1 2 3 4 0 1 2 3
-1	0
# (c # (my_	dexing on a 1-dimension array creation of a 1-dimension array (with random numbers between 0 and 1) arr_1d = np.random.rand(3) check the shape of the created array
# control # cont	which corresponds to a column vector int(my_arr_ld.shape) display the content of the array int(my_arr_ld) retrieve the first item (index 0 !!!) int(my_arr_ld[0]) the second one (index 1 !!!)
# # pri	int(my_arr_ld[1]) out the item is also the second two last (index -2) int(my_arr_ld[-2]) the 3rd item (index 2 !!!) int(my_arr_ld[2]) the 3rd item is the last one (index -1)
# 0 # v my_	dexing on a 2-dimension array creation of a 2-d array with random numbers between 0 and 1 arr_2d = np.random.rand(3,5) check the dimension of the array
# 0 pri	int(my_arr_2d.shape) display its content int(my_arr_2d) retrieve the items of the 1st row (no need to specify the indexes of the column) int(my_arr_2d[0])
# i pri	<pre>2nd row int(my_arr_2d[1]) 2st row and 2nd column int(my_arr_2d[0,1]) 2ast column of the the 1st row int(my_arr_2d[0,-1])</pre>
pri [Ou	int (my_arr_2d[-1,-1]) icing/Subsetting his is the mechanism for extracting a sub-part of an array.
T B S	he principle consists in recovering the items from a starting index to an ending one. y default, the path is made by steps of 1, otherwise the step must be explicitly specified as the last parameter. yntax: For a 1 dimension array (column vector): my_arr[start:end:step] For other dimensions: my_arr[start:end, start:end,]
	my_arr[deb:fin, deb:fin]
# 0 # 0 # 8 my_	ray of the illustration creation of a 1-d array composed of composed of the first 15 consecutive integers (0 excluded) and reformatting into a 2-d array arr = np.arange(1,16).reshape(3,5)
# 1 pri	<pre>arr = np.arange(1,16).resnape(3,5) int(my_arr) camples of slicing retrieve the first row int(my_arr[0,:]) the 1st column int(my_arr[:,0])</pre>
# 1 pri	<pre>int(my_arr[:,0]) from the 2nd to the 4th column, for the 2nd to the 3rd row int(my_arr[1:3,1:4]) same result int(my_arr[1:,1:4]) same result int(my_arr[-2:,1:4])</pre>
# s pri	<pre>int(my_arr[-2:,1:4]) same result int(my_arr[-2:,-4:-1]) utline] roadcasting</pre>
In m	Vithout the vectorization mechanism, to modify a sub-part of an array, a loop would have been used to navigate through the array. In NumPy, one can, in a very simplified way, manipulate the values of a sub-part of an array by broadcasting the desired manipulation (for example, for a valued including the desired manipulation). Note: Slicing can then be used to specify the broadcasting area for complex usage.
# pr # # my	recall the previously 2-d created array int(my_arr) let's create a new one by copy of the original one (which means modification on the copy will not affect the original)
# my pr	and another array, by creating a view of the original (modifications from one array will create cchanges on the other) carr_view = my_arr int(my_arr_view) roadcasting of an assignment let's modfiy the 1st row, replacing all the values by 9999
# my pr # # my	<pre>(no need to loop, NumPy will broadcast the assignment) c_arr[0,:] = 999 int(my_arr) modification of the last column of the original array replacing the corresponding values by the number 555 c_arr[:,-1] = 555 int(my_arr)</pre>
# pr # my	the two modifications on the original affected the view int(my_arr_view) but not the copy int(my_arr_copy) modification of the last column of the copy of the original array with the number 777 carr_copy[:, -1] = 777 int(my_arr_copy)
# pr # # my	<pre>modification on the copy did not affect the original array int(my_arr) modification of the last column of the view of the original array with the number 333 c_arr_view[:, -1] = 333 int(my_arr_view)</pre>
BI	modification on the view has changed the original array rint(my_arr) roadcasting operations between 2 arrays An operation, between arrays, e.g. addition, can be made by broadcasting without the need for loops. However, one fundamental rule must be respected:
	 either the dimensions (axes) of the two arrays are of same size, either one of the dimensions is of size 1, and the size of the other dimension can be extended to the other array. (3,4) incompatible dimensions
	+ (3, 4) impossible to extend the dimension
Br	roadcasting addition of 2 arrays of same dimension
	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$
# #	1
# # ar pr	<pre>cint(arr_1) creation of another (3 x 4) array lst column composed of 0, 2nd with 10,, 4-th with 30 cr_2 = np.array([0,10,20,30,0,10,20,30]).reshape(3,4) cint(arr_2) check the shape (dimension) of the 2 array cint(f'Shape of array 1: {arr_1.shape}') cint(f'Shape of array 2: {arr_2.shape}')</pre>
Br Or	broadcasting addition on 2 arrays of same dimension (3, 4) int(arr_1 + arr_2) roadcasting on 2 arrays of different dimensions ne of the arrays has one dimension of size 1 and the other one can be extended on the second array. stension of 2 dimensions (axis 1 and axis 2)
.∧ •	+ =
	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$
# pr pr	2 2 2 2 10 10 10 10 Dimensions of the 2 arrays int(f'Shape of array 1: {arr_1.shape}') int(f'Shape of the scalar (array 2): {np.shape(10)}') broadcasting on arrays of different dimensions : (3 , 4) et (1 , 1) int(arr_1 + 10)
	extension of only one dimension (axis 1)
	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$
# ar pr	2 2 2 2 0 10 20 30 creation of an array of dimension (3, 4) r_3 = np.array([0,10,20,30]).reshape(1,4) int(arr_3) Check the imensions of the 2 arrays int(f'Shape of array 1: {arr_1.shape}')
# pr	<pre>cint(f'Shape of array 1: {arr_1.shape}') cint(f'Shape of array 3: {arr_3.shape}') broadcasting on arrays of different dimensions: (3 , 4) et (1 , 4) cint(arr_1 + arr_3) ctension of only one dimension (axis 2 on the first array / axis 1 on the second array)</pre>
√	3 x 1
	0 10 20 30
ar pr # pr pr # #	<pre>creation of an array of dimension (3 , 1) r_4 = np.array([0,1,2]).reshape(3,1) int(arr_4) Check the shapes of the 2 arrays int(f'Shape of array 4: {arr_4.shape}') int(f'Shape of array 3: {arr_3.shape}') broadcasting on arrays of different dimensions (3 , 1) et (1 , 4)</pre>
[O	election/Masking et Boolean indexing The property of the content
	7
# my pr E)	creation of an array from a nested list _arr = np.array([[2,13,2,4,7], [7,4,21,9,31], [23,1,8,14,3]]) int(my_arr) kamples of indexing and boolean indexing indexing
# # # # my	<pre>indexing int(my_arr > 5) boolean indexing int(my_arr[my_arr > 5]) practical example: using boolean indexing to modify the values of an array -> broadcasting carr[my_arr > 5] = 99 int(my_arr)</pre>
gr SI # fr im %m	licing on unstructured data (e.g. images) Let's import useful librairies for images processing community import misc prot matplotlib.pyplot as plt matplotlib inline
# # fa pr # pl	retrieve the image in the form of a NumPy array visulaisation of its shape (3-dimensional arrray): dim 1 = gray scale on axis 1, dim 2 = gray scale on axis 2 dim 3 = colour on axis 3 (Red, Green, or Blue) ce = misc.face() int(face.shape) display the image -> resolution of 1024 x 768 t.imshow(face);
# # pl # fa pr	<pre>image compression, skipping 2 pixels in width and height resolution divided by 2: 512 x 384 => memory save Note: by increasing the skip step -> memory save but loss of quality t.imshow(face[::2,::2]); Deletion of the 3rd dimension = colour ce_gray = misc.face(gray=True) int(face_gray.shape) image en noir et blanc</pre>
# pl	<pre>image en noir et blanc t.imshow(face_gray, cmap=plt.cm.gray); compression of the black and white image t.imshow(face_gray[::2,::2], cmap=plt.cm.gray); zoom on the animal t.imshow(face_gray[50:490,370:945], cmap=plt.cm.gray); utline]</pre>
С	ongratulations !!!