# Numpy multidimensional arrays

By examples

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#### **Creation:**

#### **Dimension and shape:**

print (a.ndim)	2
print (a.shape)	(3, 4) # 3 rows, 4 columns
print (a.size)	12 # 12 elements

First dimension: rows, second dimension: columns

#### **Reshape:**

· · · · · · · · · · · · · · · · · · ·	
b = a.reshape(4,3)	[[ 1 2 3]
print(b)	
	[10 11 12]]

This might not be what you wanted

#### The reference trap:

c = a	[[314	2	3	4]	
c[0,0] = 314	[ 5			8]	
print(a)	[ 9			12]]	
print(c)	[[314	2	3	4]	
	[ 5	6		8]	
	[ 9	10	11	12]]	

Setting one or more elements of the copy c to new values, changes not only the copied array c, but also the original array a.

Take care:

The arrays a and c reference the same array, after c = a.

So if an element of c is changed, the same element of a is also changed.

(This is true for other kinds of lists also, in general for mutable objects)

If you do not want this, use the copy function:

c = a.copy()	[[31	4	2	3	4]
c[0,0] = 512	E	5	6	7	8]
print(a)	]	9	10	11	12]]
<pre>print(c)</pre>	[[51	2	2	3	4]
	]				8]
	Ε	9	10	11	12]]

(Remember that a[0,0] was changed in the previous example)

### Indexing and slicing

Remember that all indexing starts with 0! So the 2<sup>nd</sup> row for example has index 1.

#### Let's begin with the same matrix a:

import numpy as np
a = np.array([[ 1, 2, 3, 4],
[5, 6, 7, 8],
[9, 10, 11, 12]])

#### Get the second row:

<pre>b = a[1] print(b)</pre>	[5 6 7 8]

#### Get the last row:

c = a[-1]	[ 9 10 11 12]
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#### Get the element in the 3rd row and 2nd column:

d = a[2, 1]	10	
1 50 13		

Indexing goes by [row, column]

By the way, another valid syntax would be d = a[2][1]

#### Get the 3rd column:

e = a[:, 2]	[ 3 7 11]

Indexing goes by [row, column] and ":" means: all in this, so all rows of column 2

### Get a submatrix consisting of 2<sup>nd</sup> and 3<sup>rd</sup> columns:

f = a[:, 1:3]	[[ 2 3]	
	[ 6 7] [10 11]]	

This gives all rows of the columns with index 1 and 2 (remember that "1:3" means "from 1 to 3-1=2" !

( as 1:3 includes 1, but excludes 3, one of the weirdnesses of Python)

### Floating point arrays

#### The integer trap:

All the above examples were done with integers. Numpy has looked at the defined array a and found it all integers, so the resulting arrays also were integer arrays.

Even an assignment like

a[0,0] = 3.14

would not change the type of the array, the result would be a cast of 3.14 to the ineger value 3.

Defining a floating point array:

import numpy as np	]]	3.14	2.	3.	4.	]	
a = np.array([[ 3.14, 2, 3, 4],	[	5.	6.	7.	8.	]	
[5, 6, 7, 8],	[	9.	10.	11.	12.	]]	
[9, 10, 11, 12]])							
print(a)							

Even if only one element is a floating point number, Numpy sets all elements to floating point, as can be seen in the result.

## **Other Numpy functions**

Make an array from lists and / or arrays:

<pre>import numpy as np l1 = (3.14, 2, 5) l2 = [2, 3, 4]</pre>	((3.14, 2, 5), [2, 3, 4])
l = l1, l2	[[3.14 2. 5.] [2. 3. 4.]]

Make zeros array:

<pre># one dimensional: z = np.zeros(5) print(z)</pre>	[0. 0. 0. 0. 0.]
<pre># 2 dimensional z2 = np.zeros((2,3)) print(z2)</pre>	[[0. 0. 0.] [0. 0. 0.]]

#### Make an array of evenly spaced numbers:

Example: 5 values between 2 and 3

## **Operations on arrays**

Operations are done element wise.

This is also true for multiplication and division.

Functions can directly operate on arrays:

d = np.sin(a)	[[ 0.84147098	0.90929743	0.14112001	-0.7568025 ]
print(d)	[-0.95892427	-0.2794155	0.6569866	0.98935825]
	[ 0.41211849	-0.54402111	-0.99999021	-0.53657292]]

Calculations with constants are also done elementwise:

e = a + 5	[[ 6 7 8 9]
print(e)	[10 11 12 13]
	[14 15 16 17]]

## **Mathematical Matrix operations**

Matrix multiplication  $\rightarrow$  dot function

<pre>import numpy as np a = np.array([[ 1, 2 ],</pre>	
<pre>b = np.array([[5, 6],</pre>	[[19 22] [43 50]]
print(c)	[43 30]]