

# NumPy Cheat Sheet

EPFL CS 328

Numerical Methods for Visual Computing

(Version 2)

## Array initialization

Create arrays from (potentially nested) Python lists:

1-D Array	<code>a = np.array([1, 2, 3])</code>
2-D Array	<code>b = np.array([[1, 2, 3], [4, 5, 6]])</code>
Force a datatype	<code>c = np.array([0, 1, 2], dtype=np.int32)</code>

Functions to create standard arrays (e.g. all zeros):

Zero-filled array	<code>a = np.zeros((2,2))</code>
One-filled array	<code>b = np.ones((3,2))</code>
Random array	<code>d = np.random.random((2,2))</code>
Identity matrix	<code>c = np.eye(2)</code>
Increasing sequence	<code>e = np.linspace(2.0, 8.0, 10)</code>
Evenly spaced values	<code>f = np.arange(3)</code>

Note that the first three functions require a shape *tuple* argument, hence the double parentheses.

## Data Types

Basic data types:

Unsigned 32 bit integer	<code>np.uint32</code>
Signed 64 bit integer	<code>np.int64</code>
Single precision floating point	<code>np.float32</code>
Double precision floating point	<code>np.float64</code>
Boolean	<code>np.bool_</code>

## Array indexing, slicing

Basic indexing notation:

Select the element at the 3rd index	<code>a[3]</code>
Select the element at row 2, column 0	<code>b[2, 0]</code>

Slicing:

Select elements at index 0 and 1	<code>a[0:2]</code>
Select all elements in column 1	<code>b[:, 1:2]</code>
Select first two rows and last two columns	<code>b[:2, -2:]</code>

Indexing using conditional expressions:

Select elements less than 4	<code>b[b&lt;4]</code>
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Indexing using a list of indices:

Select elements (1,1) and (2,1)	<code>b[[1,2], [1,1]]</code>
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## Standard arithmetic operations

Elementwise arithmetic operations:

Addition	<code>d = e + f</code>
Subtraction	<code>d = e - f</code>
Multiplication	<code>d = e * f</code>
Division	<code>d = e / f</code>
Square root	<code>d = np.sqrt(e)</code>
Exponentiation	<code>d = np.exp(e)</code>
Natural logarithm	<code>d = np.log(e)</code>
Cosine	<code>d = np.cos(e)</code>

Other functions:

Dot/Matrix product	<code>d = np.dot(e, f)</code> # Shorter: <code>d = e @ f</code>
Compute sum of each column	<code>d = np.sum(b, axis=0)</code>
Compute max value of each row	<code>d = np.max(b, axis=1)</code>
Compute min value of array	<code>d = np.min(b)</code>
Transpose matrix	<code>d = e.T</code>

Note that `*` and `@` are different. The former does elementwise multiplication, while the latter is a dot or matrix-matrix/vector product depending on the input shapes.

## Inspecting arrays

Basic definitions:


Array dimensions	<code>a.shape</code>
Number of array dimensions	<code>a.ndim</code>
Number of array elements	<code>a.size</code>
Number of array elements in a row	<code>a.shape[0]</code>
Data type of array elements	<code>a.dtype</code>
Convert an array to a different type	<code>a.astype(np.float32)</code>

## Copying, sorting, reshaping

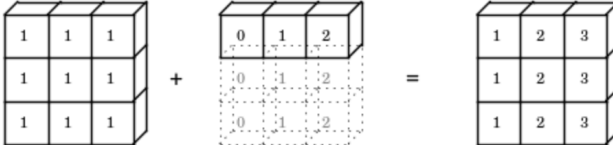
Create an array copy of b	<code>c = np.copy(b)</code>
Create view of array with <b>dtype</b>	<code>c = a.view(np.int32)</code>
Return a sorted copy of array	<code>np.sort(a)</code>
Return a copy of flattened 1D array	<code>a.flatten()</code>
Reshape array; the new shape must have the same number of entries.	<code>d = a.reshape(6, 1)</code>

## Broadcasting

Broadcasting enables operations that combine arrays of different shapes.

$$\text{np.arange}(3) + 5$$


A two dimensional array multiplied by a one dimensional array results in broadcasting if number of 1D array elements matches the number of 2D array columns.

$$\text{np.ones}((3, 3)) + \text{np.arange}(3)$$


Broadcasting can stretch both arrays to form an output array larger than either of the initial arrays.

$$\text{np.arange}(3).reshape((3, 1)) + \text{np.arange}(3)$$
