

NumPy Cheat Sheet

EPFL CS 233

Introduction to Machine Learning

(Version 1)

Array initialization

Create arrays from (potentially nested) Python lists:

Rank 1	<code>a = np.array([1, 2, 3])</code>
Rank 2	<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((1,2))</code>
Random array	<code>d = np.random.rand((2,2))</code>
Identity matrix	<code>c = np.eye(2)</code>
Increasing sequence	<code>e = np.linspace(2.0, 8.0, 10)</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 a list of indices:

Select elements (1,1) and (2,1)	<code>b[[1,2],[1,1]]</code>
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Indexing using masking:

Select elements less than 4	<code>b[b<4]</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> <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>
Cast an array to a different type	<code>a.astype(np.float32)</code>

Reshaping, Copying

Reshape an array	<code>d = a.reshape(6, 1)</code>
Extend the dimensionality	<code># shape (2, 3)</code> <code>a = np.array([[1, 2, 3], [4, 5, 6]])</code> <code># shape (2, 1, 3)</code> <code>a[:, None, :]</code>
Flatten array to 1D	<code>a.flatten()</code> <code>a.reshape((-1,))</code>
Create a deep copy of b	<code>c = np.copy(b)</code>

Broadcasting

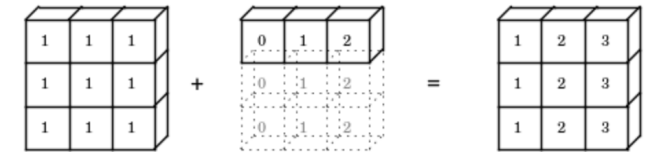
Broadcasting enables operations that combine arrays of different shapes.

`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.

`np.ones((3, 3)) + np.arange(3)`



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

`np.arange(3).reshape((3, 1)) + np.arange(3)`

