NP01_Introduction_to_Numpy

December 10, 2022

1 Introduction to NumPy

```
[1]: import numpy
     numpy.__version__
[1]: '1.23.5'
     import numpy as np
[3]: np?
                  module
    String form: <module 'numpy' from 'C:</pre>
     →\\Users\\onkar\\AppData\\Local\\Packages\\PythonSoftwareFoundation.Python.3.
     411_qbz5n2kfra8p0\\LocalCache\\local-packages\\Python311\\site-packages\\numpy\\__init__.
     y'>
    File:
                  c:\users\onkar\appdata\local\packages\pythonsoftwarefoundation.
     411_qbz5n2kfra8p0\localcache\local-packages\python311\site-packages\numpy\__init__.
     ∽ру
    Docstring:
    NumPy
    =====
    Provides
      1. An array object of arbitrary homogeneous items
      2. Fast mathematical operations over arrays
      3. Linear Algebra, Fourier Transforms, Random Number Generation
    How to use the documentation
    Documentation is available in two forms: docstrings provided
    with the code, and a loose standing reference guide, available from
    `the NumPy homepage <a href="https://numpy.org">https://numpy.org">...</a>
    We recommend exploring the docstrings using
    `IPython <a href="https://ipython.org">> , an advanced Python shell with
    TAB-completion and introspection capabilities. See below for further
```

instructions.

```
The docstring examples assume that `numpy` has been imported as `np`::
```

```
>>> import numpy as np
```

Code snippets are indicated by three greater-than signs::

```
>>> x = 42
>>> x = x + 1
```

Use the built-in ``help`` function to view a function's docstring::

```
>>> help(np.sort)
... # doctest: +SKIP
```

For some objects, ``np.info(obj)`` may provide additional help. This is particularly true if you see the line "Help on ufunc object:" at the top of the help() page. Ufuncs are implemented in C, not Python, for speed. The native Python help() does not know how to view their help, but our np.info() function does.

To search for documents containing a keyword, do::

```
>>> np.lookfor('keyword')
... # doctest: +SKIP
```

General-purpose documents like a glossary and help on the basic concepts of numpy are available under the ``doc`` sub-module::

```
>>> from numpy import doc
>>> help(doc)
... # doctest: +SKIP
```

Available subpackages

lib

Basic functions used by several sub-packages.

random

Core Random Tools

linalg

Core Linear Algebra Tools

fft

Core FFT routines

polynomial

Polynomial tools

testing

NumPy testing tools

distutils

Enhancements to distutils with support for Fortran compilers support and more.

Utilities

test

Run numpy unittests

show_config

Show numpy build configuration

dual

Overwrite certain functions with high-performance SciPy tools.

Note: `numpy.dual` is deprecated. Use the functions from NumPy or Scipy directly instead of importing them from `numpy.dual`.

matlib

Make everything matrices.

__version__

NumPy version string

Viewing documentation using IPython

Start IPython with the NumPy profile (``ipython -p numpy``), which will import `numpy` under the alias `np`. Then, use the ``cpaste`` command to paste examples into the shell. To see which functions are available in `numpy`, type ``np.<TAB>`` (where ``<TAB>`` refers to the TAB key), or use ``np.*cos*?<ENTER>`` (where ``<ENTER>`` refers to the ENTER key) to narrow down the list. To view the docstring for a function, use ``np.cos?<ENTER>`` (to view the docstring) and ``np.cos??<ENTER>`` (to view the source code).

Copies vs. in-place operation

Most of the functions in `numpy` return a copy of the array argument (e.g., `np.sort`). In-place versions of these functions are often available as array methods, i.e. ``x = np.array([1,2,3]); x.sort()`. Exceptions to this rule are documented.

1.1 Fixed Type Arrays in Python

```
[4]: import array
L = list(range(10))
A = array.array('i', L) # 'i' indicates that contents are integers
A
```

[4]: array('i', [0, 1, 2, 3, 4, 5, 6, 7, 8, 9])

1.2 Creating Arrays from Python Lists

```
[5]: np.array([1,2,3,4,5])
 [5]: array([1, 2, 3, 4, 5])
 [6]: np.array([1,2,3.14,4,5]) # NumPy will upcast if types do not match
 [6]: array([1. , 2. , 3.14, 4. , 5. ])
 [7]: np.array([1,2,3.14,4,5],dtype='float32') # Data type can be explicitly set_{\square}
       →using keyword 'dtype'
 [7]: array([1. , 2. , 3.14, 4. , 5. ], dtype=float32)
 [8]: np.array([range(i,i+3) for i in range(2,7,2)]) # nested lists are converted to
       ⇔multidimensional arrays
 [8]: array([[2, 3, 4],
             [4, 5, 6],
             [6, 7, 8]])
     1.3 Creating Arrays from Scratch
 [9]: np.zeros(10,dtype="int") # length 10 int array with zeroes
 [9]: array([0, 0, 0, 0, 0, 0, 0, 0, 0])
[10]: np.ones((3,5),dtype="float") # 3*5 floating-point array with ones
[10]: array([[1., 1., 1., 1., 1.],
             [1., 1., 1., 1., 1.],
             [1., 1., 1., 1., 1.]
[11]: np.full((3,5),3.14) # 3*5 floating-point array with 3.14
[11]: array([[3.14, 3.14, 3.14, 3.14, 3.14],
             [3.14, 3.14, 3.14, 3.14, 3.14],
             [3.14, 3.14, 3.14, 3.14, 3.14])
[12]: np.arange(0,20,2) # starting at 0, ending at 20, sstepping size 2
[12]: array([0, 2, 4, 6, 8, 10, 12, 14, 16, 18])
[13]: np.linspace(0,1,5) # 5 values evenly spaced between 0 and 1
[13]: array([0. , 0.25, 0.5 , 0.75, 1. ])
```

```
[14]: np.random.random((3,3)) # 3*3 array (uniformly distributed) with random values
       \hookrightarrowbetween 0 and 1
[14]: array([[0.9174918 , 0.4896383 , 0.95786394],
             [0.2186783, 0.84429385, 0.10723626],
             [0.96688974, 0.2739895, 0.31981254]])
[15]: np.random.normal((3,3)) # 3*3 array (normally distributed) with random values
       \hookrightarrowbetween 0 and 1
[15]: array([3.40408317, 3.69649029])
[16]: np.random.randint(0,10,(3,3)) # 3*3 array with random int in interval [0,10)
[16]: array([[9, 3, 3],
             [3, 3, 5],
             [0, 5, 6]])
[17]: np.eye(3) # 3*3 identity matrix
[17]: array([[1., 0., 0.],
             [0., 1., 0.],
             [0., 0., 1.]])
[18]: np.empty(3) # uninitialized array of 3 int with value from previous allocation_
       ⇔in that memory location
[18]: array([1., 1., 1.])
```

1.4 NumPy Standard Data Types

Data Type	Description
bool_	Boolean (True or False) stored as a byte
int	Default integer type (same as C long; normally either int64 or
	int32)
intc	Identical to C int (normally int32 or int64)
intp	Integer used for indexing (same as C ssize_t; normally either
	int32 or int64)
int8	Integer (-127 to 128)
int16	Integer $(-32768 \text{ to } 32767)$
int32	Integer (-2147483648 to 2147483647)
int64	Integer $(-9223372036854775808 \text{ to } 9223372036854775807)$
uint8	Unsigned integer (0 to 255)
uint16	Unsigned integer (0 to 65535)
uint32	Unsigned integer $(0 \text{ to } 4294967295)$
uint64	Unsigned integer (0 to 18446744073709551615)
${ m float}$	Shorthand for float64

Data Type	Description
	Half-precision foat: sign bit, 5 bits exponent, 10 bits mantissa Single-precision foat: sign bit, 8 bits exponent, 23 bits mantissa Double-precision foat: sign bit, 11 bits exponent, 52 bits mantissa
complex_ complex64 complex128	Shorthand for complex128 Complex number, represented by two 32-bit foats Complex number, represented by two 64-bit foats

```
[19]: np.zeros(10,dtype="int16") # length 10 int16 array with zeroes
```

[19]: array([0, 0, 0, 0, 0, 0, 0, 0, 0], dtype=int16)

[20]: np.zeros(10,dtype=np.int16) # np.int16 is associated NumPy object

[20]: array([0, 0, 0, 0, 0, 0, 0, 0, 0], dtype=int16)