

The numpy module

Working with efficient arrays
of values

Arrays of Numbers

- **Arrays** - not built into Python
 - *fundamental data structure in computing*
- All elements same type
- Elements known to be adjacent to each other in memory
- Regular accesses because of uniformity
 - *leads to faster access*
- Require less memory than lists

The Numpy Module

- Widely used module "numpy" provides an implementation of arrays
 - **np.ndarray** type represents numeric arrays with arbitrary dimensions
 - empty arrays
 - zero-filled arrays
- np.array() method converts Python sequences into arrays
 - converts cube list-of-lists-of-lists to array
 - access with three indexes
 - » like the list-of-lists-of-lists approach

Numpy Help

```
import numpy as np
help(np.ndarray)
```

Use numpy with a convenient name...

Help on class ndarray in module numpy:

```
class ndarray(builtins.object)
| ndarray(shape, dtype=float, buffer=None, offset=0,
|          strides=None, order=None)
```

An array object represents a multidimensional, homogeneous array of fixed-size items. An associated data-type object describes the format of each element in the array (its byte-order, how many bytes it occupies in memory, whether it is an integer, a floating point number, or something else, etc.)

Arrays should be constructed using ``array``, ``zeros`` or ``empty`` (refer to the See Also section below). The parameters given here refer to a low-level method (``ndarray(...)``) for instantiating an array.

For more information, refer to the ``numpy`` module and examine the methods and attributes of an array.

Parameters

np.ndarray Arrays

```
triple = np.zeros( [3] )
square = np.zeros( [2, 2] )
line = np.empty( [8] )

print(triple)
print()
print(square)
print()
print(line)
```

```
[ 0.  0.  0.]
```

```
[[ 0.  0.]
 [ 0.  0.]
```

```
[ 6.90865611e-310  4.68012469e-310  0.00000000e+000  0.00000000e+000
 0.00000000e+000  0.00000000e+000  0.00000000e+000  nan]
```

- 1-dimensional, 3-element array of floating-point zeros
- 2x2-element square array of floating-point zeros
- 8-element array of floating-points, uninitialized

Generate Numpy Arrays

- numpy alternatives to "range()" function:
 - `vec1 = np.arange(start, stop, step)`
 - » array is half-open interval, excludes "stop" value
 - » Similar to python "range()" - not good for real #'s
 - `vec2 = np.linspace(start, end, nsamples)`
 - » array is closed interval, includes both "start" and "end"
 - » Best for non-integer intervals
- Change an array's shape:
 - `a1 = vec1.reshape(shape-tuple)`
 - `shape-tuple` argument defines new array's shape
 - » number of elements gives array's dimensions
 - » element values give dimension extents
 - example: (6, 6, 6) - three dimensions, 6 elements in each

Convert Python Lists to Numpy Arrays

- 1-dimensional list to array:
 - `mylist1 = [x**2 for x in range(100)]`
 - `myarray1 = np.array(mylist1)`
 - `print(myarray1.shape, myarray1.ndim)`
- 2-dimensional list-of-lists to array:
 - `mylist2 = [\`
`[x**2/y for x in range(50)] \`
`for y in range(1,10)]`
 - `myarray2 = np.array(mylist2)`
 - `print(myarray2.shape, myarray2.ndim)`

Array-Making Functions

- `np.zeros(shape-tuple)`
`np.ones(shape-tuple)`
 - create arrays filled with 0s or with 1s
- `np.empty(shape-tuple)`
 - create an array with no particular values
- `shape-tuple` argument
 - defines shape of the array
 - number of elements gives array's dimensions
 - element values give dimension extents
 - » ex. (4, 4, 4) – three dimensions, 4 elements in each

More Functions to Make Arrays Directly

- `np.full(shape-tuple , fill-value)`
 - create an array filled with any desired value

Create an array shaped like a given pre-existing array:

- `np.zeros_like(array)`
- `np.ones_like(array)`
- `np.empty_like(array)`
- `np.full_like(array, fill-value)`

numpy Versus math

- "math" module provides advanced, floating-point functions, for example:
 - `math.log`, `math.exp`
 - `math.pi`, `math.e`
 - `math.sin`, `math.cos`, `math.tan`
- "numpy" module provides comparable functions, extended to apply to ndarrays
- Also, related submodules/functions:
 - `np.random.random()`, `np.random.random_integer()`
 - (More functions in `np.random`)

Random Arrays; Array-Oriented Actions

- Create arrays of random values
- Math operators apply to arrays, and *scalars*

```

sq1 = np.random.randint(0,11, (2,2))
sq2 = 10 * np.random.random((2,2,))

print('sq1:\n', sq1, '\nsq2:\n', sq2)

print('\nproduct:\n', np.round(sq1 * sq2, 2))

```

scalar w/
array

```

sq1:
 [[2 6]
 [2 0]]
sq2:
 [[8.39034424 8.10763239]
 [2.62528707 9.84959535]]

product:
 [[16.78 48.65]
 [ 5.25  0.  ]]

```

array - array product;
rounding the results

Multiplying Arrays

- Two arrays can be multiplied element-by-element
 - Also: add, subtract, divide, other operations
- Linear/matrix algebra
 - Numpy provides `.dot()`, `.cross()`, `.diagonal()` functions
 - Array shapes must be compatible
 - Usage is beyond the scope of this introduction

Reduction Operations

- "Collapse arrays to fewer dimensions or to scalars

```
cube = np.array([ [1,2,3], [0.5,-7.3,4.44] ])
print(cube.shape, '\n', cube, '\n#-----')
print('Column-wise sums:', np.sum(cube, axis=0), '\n#-----')
print('  Row-wise sums:', np.sum(cube, axis=1), '\n#-----')
print('  Overall sum:', np.sum(cube), '\n#-----')
```

```
(2, 3)
[[ 1.  2.  3. ]
 [ 0.5 -7.3  4.44]]
#-----
Column-wise sums: [ 1.5 -5.3  7.44]
#-----
  Row-wise sums: [ 6.  -2.36]
#-----
  Overall sum: 3.6400000000000006
#-----
```

sum() function
reduces array's
dimensionality

Example: powers of 2

```
# Naive, pythonic approach:
exponents = np.empty( (8,) )
for i in range(8):
    exponents[i] = i

exponents # floating-point!

array([0., 1., 2., 3., 4., 5., 6., 7.]
```

```
# Numpy approach:
exponents = np.arange(8)

exponents # integer

array([0, 1, 2, 3, 4, 5, 6, 7])
```

```
# Numpy approach 2:
exponents = np.linspace(0, 7, 8)

exponents # floating-point again

array([0., 1., 2., 3., 4., 5., 6., 7.]
```

```
powers = 2 ** exponents
powers

array([ 1.,  2.,  4.,  8., 16., 32., 64., 128.])
```

*Example:
trig
function*

```
x = np.linspace(0, 6*np.pi, 100)  
y = np.sin(x) * np.cos(3*x) * np.exp(-.1 * x)
```

```
import matplotlib.pyplot as plt  
plt.plot(x,y)  
plt.show()
```

