Chapter 6, Data Analytic Tools

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Contents

0	Introduction to the Toolkit	1
1	Creating Numpy Arrays	1
2	Arithmetic on Numerical Arrays	4

0 Introduction to the Toolkit

To install, we recommend you use the Anaconda package manager for Python. Complete instructions can be found at https://docs.anaconda.com/anaconda/ install/. You will want to make sure you have these things.

- 1. ipython This is an enhanced Python interactive shell.
- 2. numpy This is a high-performance library for matrix calculations
- 3. matplotlib This is a library that can produce mathematical plots. They can be viewed locally on your machine, or it can generate graphical files on a server, which can be viewed off your account's web presence (public_html).
- 4. pandas This is a powerful tool that supports R-style data frames.

1 Creating Numpy Arrays

To see if numpy is working correctly, open ipython and do this import.

Python 3.6.8 |Anaconda, Inc.| (default, Dec 29 2018, 19:04:46) Type 'copyright', 'credits' or 'license' for more information IPython 7.2.0 -- An enhanced Interactive Python. Type '?' for help.

```
In [1]: import numpy as np
In [2]:
```

If there is no error, you are good to go. Numpy supports the ndarray, or n-dimensional array object. These arrays are fixed-size mutable objects. They are homogeous; to wit, all objects must be of the same type. The type of objects present in an ndarray is called its dtype; this is short for "data type." Let us create an array and explore.

```
In [2]: x = np.array([0,0,0,0])
```

In the first instance, Numpy saw the 0s and said, "Oh, integers." Note that these integers are 64 bit integers in two's complement notation. It auto-detected and assigned a default dtype.

In the second instance, Numpy was told the dtype was float64 and cast the zeroes appropriately. It also saw list with two lists inside and made a 2×2 array. You have a choice when indexing into a two dimensional array.

In [6]: y[0,0] Out[6]: 0.0 In [7]: y[0][0] Out[7]: 0.0 The shape of an array is maintained as a tuple In [8]: y.shape Out[8]: (2, 2) \section{Arithmetic, Functions, and Numpy Arrays} In [9]: x.shape

Out[9]: (4,)

To fill make an array with nothing but zeroes you can do this.

You can also specify the data type.

Watch the array get reshaped two ways.

```
In [13]: y.reshape(6,6)
Out[13]:
array([[0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0]])
In [14]: y.reshape(9,4)
Out[14]:
array([[0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0],
       [0, 0, 0, 0]])
```

You can make arrays with any number of dimensions; just specify a tuple. We also demonstrate the fill method.

```
In [15]: z
Out[15]:
array([[[ 0., 0., 0.],
        [ 0., 0., 0.],
        [ 0., 0., 0.],
```

```
[0., 0., 0.]],
      [[ 0., 0., 0.],
       [0., 0., 0.],
       [0., 0., 0.],
       [0., 0., 0.]]])
In [16]: z.fill(5)
In [16]: z
Out [25]:
array([[[ 5., 5., 5.],
       [5., 5., 5.],
       [5., 5., 5.],
       [5., 5., 5.]],
      [[5., 5., 5.],
       [5., 5., 5.],
       [5., 5., 5.],
       [5., 5., 5.]])
```

Exercises You should create some array so you can test the functions speced below.

- 1. If x is an array, what is x.T? Make sure you try this on a two-dimensional array.
- 2. Write a function isSquare(x) which tells if an two-dimensional array passed to it is a square array.
- 3. Write a function swapRows(x, j, k) that will swap the jth and kth rows in an array, and which throws a ValueError if the rows specified are out of bounds.
- Write a function swapCols(x, j, k) that will swap the jth and kth columns in an array, and which throws a ValueError if the columns specified are out of bounds.
- 5. What does np.eye(3) produce?

2 Arithmetic on Numerical Arrays

Let us look at the behavior of numerical arrays when presented with arithmetic operators. We begin by making two 3×3 arrays.

In [1]: import numpy as np

```
In [2]: x = np.array([[1,2,3],[4,5,6],[7,8,9]])
In [3]: x
Out[3]:
array([[1, 2, 3],
       [4, 5, 6],
       [7, 8, 9]])
In [4]: y = np.eye(3)
In [5]: y
Out[5]:
array([[ 1., 0., 0.],
       [ 0., 1., 0.],
       [ 0., 0., 1.]])
```

Now let us play with arithmetic.

In [6]: x + y
Out[6]:
array([[2., 2., 3.],
 [4., 6., 6.],
 [7., 8., 10.]])

The arrays were added element by element.

Programming Exercises Do these NOW!

- 1. What happens if you attempt to add arrays of different sizes?
- 2. What happens if you try to add an array to a number?

Apparently, multiplying an array by a number just multiplies every entry by that number.

```
In [7]: x + 5*y
Out[7]:
array([[ 6., 2., 3.],
        [ 4., 10., 6.],
        [ 7., 8., 14.]])
```

If you are a linear algebra fan, start cringing now. However, you will see later that all will end well

In [8]: x*y
Out[8]:

array([[1., 0., 0.], [0., 5., 0.], [0., 0., 9.]])

If you multiply arrays, they multiply entrywise. Let's inject some power.

```
In [9]: x**3
Out[9]:
array([[ 1, 8, 27],
       [ 64, 125, 216],
       [343, 512, 729]])
```

Now we will apply a function. All of the math functions have **np**. analogs that apply the entry-by-entry.

```
In [10]: np.log10(x)
Out[10]:
array([[ 0. , 0.30103 , 0.47712125],
       [ 0.60205999, 0.69897 , 0.77815125],
       [ 0.84509804, 0.90308999, 0.95424251]])
In [11]: np.sin(x)
Out[11]:
array([[ 0.84147098, 0.90929743, 0.14112001],
       [-0.7568025 , -0.95892427, -0.2794155 ],
       [ 0.6569866 , 0.98935825, 0.41211849]])
```