

Scientific Python bootcamp  
2015 Berkeley Statistics Masters program

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# Chapter 1

## Python review

*Brief review of basic Python. To begin, sign onto your GitHub account and fork the bootcamp repository: <https://github.com/jarrodmillman/python-bootcamp-2015>.*

*Now from your BASH terminal clone your fork(ed) repository. If you are using HTTP to authenticate, you will do something like this:*

```
$ cd <to where you keep your source repositories>
$ git clone https://github.com/<you>/python-bootcamp-2015.git
$ cd python-bootcamp-2015
```

*For SSH authentication, do something like this:*

```
$ cd <to where you keep your source repositories>
$ git clone git@github.com:<you>/python-bootcamp-2015.git
$ cd python-bootcamp-2015
```

### 1.1 Introduction

- <https://docs.python.org/2/index.html>
- <https://docs.python.org/2/library/index.html>
- <https://scipy-lectures.github.io/>
- <http://software-carpentry.org/v4/python/>
- <https://docs.python.org/3/howto/pyporting.html>

#### 1.1.1 Interpreter

For much of the bootcamp, you should type the example code snippets at an interactive Python terminal. I recommend using either the IPython shell or the IPython notebook. To start an IPython shell, type the following at a BASH prompt:

```
$ ipython
```

To start an IPython notebook, type

```
$ ipython notebook
```

### 1.1.2 Objects

Everything is an object in Python. Roughly, this means that it can be tagged with a variable and passed as an argument to a function. Often it means that everything has *attributes* and *methods*.

Certain objects in Python are mutable (e.g., lists, dictionaries), while other objects are immutable (e.g., tuples, strings, sets). Many objects can be composite (e.g., a list of dictionaries or a dictionary of lists, tuples, and strings).

### 1.1.3 Variables

Recall from last semester that variables are not their values in Python (think "I am not my name, I am the person named XXX"). You can think of variables as tags on objects. In particular, variables can be bound to an object of one type and then reassigned to an object of another type without error.

### 1.1.4 Modules, files, packages, import

While you will often explore things from an interactive Python prompt, you will save your code in files for reuse as well as to document what you've done. You can use Python code saved in a plain text file from a Python prompt or other files by importing it. Typically, this is done at the top of a file (if you are working at a prompt, you just need to import it before you want to use the functionality).

Here are some examples of importing:

```
import math
from math import cos
import numpy as np
import scipy as sp
import matplotlib.pyplot as plt
```

### 1.1.5 Style

Adopting standard coding conventions is good practice.

- <https://www.python.org/dev/peps/pep-0008/>
- <https://docs.python.org/2/tutorial/controlflow.html#intermezzo-coding-style>
- [https://github.com/numpy/numpy/blob/master/doc/HOWTO\\_DOCUMENT.rst.txt](https://github.com/numpy/numpy/blob/master/doc/HOWTO_DOCUMENT.rst.txt)
- [http://matplotlib.org/devel/coding\\_guide.html](http://matplotlib.org/devel/coding_guide.html)

The first link above is the official "Style Guide for Python Code", usually referred to as PEP8 (PEP is an acronym for Python Enhancement Proposal). There are a couple of potentially helpful tools for helping you conform to the standard. The [pep8](#) package that provides a commandline tool to check your code against some of the PEP8 standard conventions. Similarly, [autopep8](#) provides a tool to automatically format your code so that it conforms to the PEP8 standards. I have used both a little and they seem to work fairly well.

The last two links discuss the NumPy docstring<sup>1</sup> standard. Let's briefly see how you might benefit from NumPy docstrings in practice.

---

<sup>1</sup>Docstrings are an important part of a Python program:

A docstring is a string literal that occurs as the first statement in a module, function, class, or method definition. Such a docstring becomes the `__doc__` special attribute of that object. All modules should normally have docstrings, and all functions and classes exported by a module should also have docstrings. Public methods (including the `__init__` constructor) should also have docstrings.

— <https://www.python.org/dev/peps/pep-0257/>

Docstrings also allow for the use of doctests.

The doctest module searches for pieces of text that look like interactive Python sessions, and then executes those sessions to verify that they work exactly as shown.

— <http://docs.python.org/2/library/doctest.html>

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

```
In [2]: np.ndim?
```

```
Type:      function
String form: <function ndim at 0x7fcabd864938>
File:      /usr/lib64/python2.7/site-packages/numpy/core/fromnumeric.py
Definition: np.ndim(a)
Docstring:
Return the number of dimensions of an array.
```

Parameters

-----

```
a : array_like
    Input array. If it is not already an ndarray, a conversion is
    attempted.
```

Returns

-----

```
number_of_dimensions : int
    The number of dimensions in `a`. Scalars are zero-dimensional.
```

See Also

-----

```
ndarray.ndim : equivalent method
shape : dimensions of array
ndarray.shape : dimensions of array
```

Examples

-----

```
>>> np.ndim([[1,2,3],[4,5,6]])
2
>>> np.ndim(np.array([[1,2,3],[4,5,6]]))
2
>>> np.ndim(1)
0
```

## Exercises

- What happens if you type `np.ndim??` (i.e., use two question marks)?
- Type `np.tril?` at an IPython prompt. What does `np.tril` do?
- Type `np.ndarray?` at an IPython prompt. Briefly skim the docstring. `ndarray` is the basic datastructure provided by NumPy. We will examine it in much more detail in the next chapter.
- Type `np.` followed by the <Tab> key at an IPython prompt. Choose two or three of the completions and use `?` to view their docstrings. In particular, pay attention to the examples provided near the end of the docstring and see whether you can figure out how you might use this functionality. Use on them as well.

### 1.1.6 Python 2 vs. 3

Python 3 is a new version of Python, which is incompatible with Python 2. We will use Python 2 in this bootcamp, but Python 3 is the future. Due to the large installed codebase of Python 2, the transition will take years.





[illegible]

```
In [9]: cos(0)
```

```
NameError                                Traceback (most recent call last)
<ipython-input-6-edaadd132e03> in <module>()
----> 1 cos(1)
```

**NameError:** name 'cos' is not defined

```
In [10]: import math
```

```
In [11]: math.cos(0)
Out[11]: 1.0
```

```
In [12]: math.cos(math.pi)
Out[12]: -1.0
```

```
In [13]: (type(1), type(1.1), type(1+2j))
Out[13]: (int, float, complex)
```

The above line is an example of a composite object called a tuple, which we will discuss more in § 1.2.3 below. At an IPython prompt, use `type?` to see what `type` does.

The `math` package in the standard library includes many additional numerical operations.

```
In [14]: math.
```

<code>math.acos</code>	<code>math.degrees</code>	<code>math.fsum</code>	<code>math.pi</code>
<code>math.acosh</code>	<code>math.e</code>	<code>math.gamma</code>	<code>math.pow</code>
<code>math.asin</code>	<code>math.erf</code>	<code>math.hypot</code>	<code>math.radians</code>
<code>math.asinh</code>	<code>math.erfc</code>	<code>math.isinf</code>	<code>math.sin</code>
<code>math.atan</code>	<code>math.exp</code>	<code>math.isnan</code>	<code>math.sinh</code>
<code>math.atan2</code>	<code>math.expm1</code>	<code>math.ldexp</code>	<code>math.sqrt</code>
<code>math.atanh</code>	<code>math.fabs</code>	<code>math.lgamma</code>	<code>math.tan</code>
<code>math.ceil</code>	<code>math.factorial</code>	<code>math.log</code>	<code>math.tanh</code>
<code>math.copysign</code>	<code>math.floor</code>	<code>math.log10</code>	<code>math.trunc</code>
<code>math.cos</code>	<code>math.fmod</code>	<code>math.log1p</code>	
<code>math.cosh</code>	<code>math.frexp</code>	<code>math.modf</code>	

## Exercises

Using the section on "Built-in Types" from the official "The Python Standard Library" reference (follow the first link at the top of § 1.2), figure out how to compute:

1.  $3 \leq 4$ ,
2.  $3 \bmod 4$ ,
3.  $|-4|$ ,
4.  $(\lceil \frac{3}{4} \rceil \times 4)^3 \bmod 2$ , and
5.  $\sqrt{-1}$ .

## Questions

1. How do you get the list of completions for `x.`?
2. What is the difference in the old and new behavior of division?
3. Read the "Truth Value Testing" and "Boolean Operations" subsections at the top of the "Built-in Types" section of the Library reference. How does this compare to how R handles things?

## 1.2.2 Strings

Strings are immutable sequences of (zero or more) characters.

### Sequences

Unlike numbers, Python strings are container objects. Specifically, it is a sequence. Python has several sequence types including strings, tuples, and lists. Sequence types share some common functionality, which we can demonstrate with strings.

- **Indexing** To see how indexing works in Python let's use the string containing the digits 0 through 9.

```
In [1]: import string
```

```
In [2]: string.digits
```

```
Out[2]: '0123456789'
```

```
In [3]: string.digits[1]
```

```
Out[3]: '1'
```

```
In [4]: string.digits[-1]
```

```
Out[4]: '9'
```

Note that indexing starts at 0 (unlike R and Fortran, but like C). Also negative integers index starting from the end of the sequence. You can find the length of a sequence using the `len()` function.

- **Slicing** Slicing allows you to select a subset of a string (or any sequence) by specifying start and stop indices as well as a step, which you specify using the `start:stop:step` notation inside of square braces.

```
In [5]: string.digits[1::2]
```

```
Out[5]: '13579'
```

```
In [6]: string.digits[9::-1]
```

```
Out[6]: '9876543210'
```

- **Subsequence testing**

```
In [7]: '23' in string.digits
```

```
Out[7]: True
```

```
In [16]: '25' not in string.digits
```

```
Out[16]: True
```

## String methods

```
In [1]: string1 = "my string"
```

```
In [2]: string1.
string1.capitalize  string1.islower      string1.rpartition
string1.center       string1.isspace      string1.rsplitt
string1.count         string1.istitle      string1.rstrip
string1.decode        string1.isupper      string1.split
string1.encode        string1.join          string1.splitlines
string1.endswith     string1.ljust        string1.startswith
string1.expandtabs    string1.lower        string1.strip
string1.find          string1.lstrip       string1.swapcase
string1.format        string1.partition    string1.title
string1.index         string1.replace      string1.translate
string1.isalnum       string1.rfind        string1.upper
string1.isalpha       string1.rindex       string1.zfill
string1.isdigit       string1.rjust
```

```
In [2]: string1.upper()
```

```
Out[2]: 'MY STRING'
```

```
In [3]: string1.upper?
```

```
Type:          builtin_function_or_method
```

```
String form: <built-in method upper of str object at 0x7fa136f8ced0>
```

```
Docstring:
```

```
S.upper() -> string
```

Return a copy of the string S converted to uppercase.

```
In [4]: string1 + " is your string."
```

```
Out[4]: 'my string is your string.'
```

```
In [5]: "*" * 10
```

```
Out[5]: '*****'
```

```
In [6]: string1[3:]
```

```
Out[6]: 'string'
```

```
In [7]: string1[3:4]
```

```
Out[7]: 's'
```

```
In [8]: string1[4::2]
```

```
Out[8]: 'tig'
```

```
In [9]: string1[3:5] = 'ts'
```

```
-----
TypeError                                Traceback (most recent call last)
<ipython-input-12-d7a58dc91703> in <module>()
----> 1 string1[3:5] = 'ts'
```

```
TypeError: 'str' object does not support item assignment
```

```
In [10]: string1.__
```

<code>string1.__add__</code>	<code>string1.__len__</code>
<code>string1.__class__</code>	<code>string1.__lt__</code>
<code>string1.__contains__</code>	<code>string1.__mod__</code>
<code>string1.__delattr__</code>	<code>string1.__mul__</code>
<code>string1.__doc__</code>	<code>string1.__ne__</code>
<code>string1.__eq__</code>	<code>string1.__new__</code>
<code>string1.__format__</code>	<code>string1.__reduce__</code>
<code>string1.__ge__</code>	<code>string1.__reduce_ex__</code>
<code>string1.__getattr__</code>	<code>string1.__repr__</code>
<code>string1.__getitem__</code>	<code>string1.__rmod__</code>
<code>string1.__getnewargs__</code>	<code>string1.__rmul__</code>
<code>string1.__getslice__</code>	<code>string1.__setattr__</code>
<code>string1.__gt__</code>	<code>string1.__sizeof__</code>
<code>string1.__hash__</code>	<code>string1.__str__</code>
<code>string1.__init__</code>	<code>string1.__subclasshook__</code>

### Exercises

At an interactive Python prompt, type `x = "The ant wants what all ants want."`. Using string indexing, slicing, subsequence testing, and methods, solve the following:

1. Convert the string to all lower case letters (don't change `x`).
2. Count the number of occurrences of the substring `"ant"`.
3. Create a list of the words occurring in `x`. Make sure to remove punctuation and convert all words to lowercase.
4. Using only string methods on `x`, create the following string: `"The chicken wants what all chickens want."`
5. Using indexing and the `+` operator, create the following string: `"The tna wants what all ants want."`
6. Do the same thing except using a string method instead.

### Questions

1. How do the string method's `split` and `rsplit` differ? [Hint: use `?` to view the method's docstrings.]
2. What happens when you multiple a string by a number? How does this relate to the string method `__mul__`? [Hint: look at the docstring.]
3. How does the `len()` function know how to find the length of a sequence?
4. How do the `in` and `not in` operators work?

### 1.2.3 Tuples

Tuples are immutable sequences of (zero or more) objects. Functions in Python often return tuples.

```
In [1]: x = 1; y = 2
```

```
In [2]: xy = (x, y)
```

```
In [3]: xy
```

```
Out[3]: (1, 2)
```

```
In [4]: xy[1]
Out[4]: 2
```

```
In [5]: xy[1] = 3
```

```
-----
TypeError                                Traceback (most recent call last)
<ipython-input-7-b22951f8a33e> in <module>()
----> 1 xy[1] = 3
```

```
TypeError: 'tuple' object does not support item assignment
```

```
In [6]: (x, y)
Out[6]: (1, 2)
```

```
In [7]: x, y
Out[7]: (1, 2)
```

### Exercises

1. Note that `x`, `y` and `(x, y)` both print the same string. To see why that is assign them to variables and check their type.
2. Create the following `x=5` and `y=6`. Now swap their values. (How would you do this in R?)

#### 1.2.4 List

Lists are mutable sequences of (zero or more) objects.

```
In [1]: dice = [1, 2, 3, 4, 5, 6]
```

```
In [2]: dice[1::2]
Out[2]: [2, 4, 6]
```

```
In [3]: dice[1::2] = dice[::2]
```

```
In [4]: dice
Out[4]: [1, 1, 3, 3, 5, 5]
```

```
In [5]: dice*2
Out[5]: [1, 1, 3, 3, 5, 5, 1, 1, 3, 3, 5, 5]
```

```
In [6]: dice+dice[::-1]
Out[6]: [1, 1, 3, 3, 5, 5, 5, 5, 3, 3, 1, 1]
```

```
In [7]: 1 in dice
Out[7]: True
```

### Exercises

1. Create a list of numbers. Reverse the order of the items in the list using slicing. Now reverse the order of the items using a list method. How does using the method differ from slicing? Do you think you think tuples have a method to reverse the order of its items? Why or why not? Check to see if you are correct or not.

2. Using a list method sort your numbers. Create a list of strings and sort it. Put your list of numbers and strings together in one list and sort it. What happened?

### 1.2.5 Dictionaries

Dictionaries are mutable, unordered collections of key-value pairs.

```
In [99]: students = {"Jarrod Millman": [10, 11, 9],
....:               "Thomas Kluyver": [11, 9, 10],
....:               "Stefan van der Walt": [12, 9, 9]}

In [100]: students
Out[100]:
{'Jarrod Millman': [10, 11, 9],
 'Stefan van der Walt': [12, 9, 9],
 'Thomas Kluyver': [11, 9, 10]}

In [102]: students.keys()
Out[102]: ['Thomas Kluyver', 'Stefan van der Walt', 'Jarrod Millman']

In [103]: students["Jarrod Millman"]
Out[103]: [10, 11, 9]

In [104]: students["Jarrod Millman"][1]
Out[104]: 11
```

### 1.2.6 Sets

Sets are immutable, unordered collections of unique elements.

```
In [1]: x = {1, 2, 4, 1, 4}

In [2]: x
Out[2]: {1, 2, 4}

In [3]: x.
x.add                x.issubset
x.clear              x.issuperset
x.copy               x.pop
x.difference          x.remove
x.difference_update  x.symmetric_difference
x.discard             x.symmetric_difference_update
x.intersection        x.union
x.intersection_update x.update
x.isdisjoint
```

### 1.2.7 And more

```
In [1]: import collections

In [2]: collections.
collections.Callable      collections.MutableSequence
collections.Container     collections.MutableSet
collections.Counter       collections.OrderedDict
collections.Hashable      collections.Sequence
```

<code>collections.ItemsView</code>	<code>collections.Set</code>
<code>collections.Iterable</code>	<code>collections.Sized</code>
<code>collections.Iterator</code>	<code>collections.ValuesView</code>
<code>collections.KeysView</code>	<code>collections.defaultdict</code>
<code>collections.Mapping</code>	<code>collections.deque</code>
<code>collections.MappingView</code>	<code>collections.namedtuple</code>
<code>collections.MutableMapping</code>	

## 1.3 Built-in functions

- <https://docs.python.org/2/library/functions.html>

Python has several built-in functions (you can find a full list using the link above). We've already used a few (e.g., `len()`, `type()`, `print()`). Here are a few more that you will find useful.

### 1.3.1 zip

```
In [108]: zip([1, 2], ["a", "b"])
Out[108]: [(1, 'a'), (2, 'b')]
```

### 1.3.2 enumerate

```
In [109]: enumerate(["a", "b"])
Out[109]: <enumerate at 0x7f5e3e018640>
```

```
In [110]: list(enumerate(["a", "b"]))
Out[110]: [(0, 'a'), (1, 'b')]
```

#### Question

- What do the built-in functions `abs()`, `all()`, `any()`, `dict()`, `dir()`, `id()`, `list()`, and `set()` do? Make sure to use `?` from the IPython prompt as well as looking at the documentation in the official Python Standard Library reference (use the above link).

## 1.4 Control flow

- <https://docs.python.org/2/tutorial/controlflow.html>

### 1.4.1 If-then-else

- <https://docs.python.org/2/tutorial/controlflow.html#if-statements>

```
In [44]: x = 2
```

```
In [45]: if x < 2:
....:     print("Yes")
....: else:
....:     print("No")
....:
```

No

### 1.4.2 For-loops (and list comprehension)

- <https://docs.python.org/2/tutorial/controlflow.html#for-statements>
- <https://docs.python.org/2/whatsnew/2.0.html#list-comprehensions>

```
In [49]: for x in [1,2,3,4]:
...:     print(x)
...:
1
2
3
4
```

```
In [50]: for x in [1,2,3,4]:
...:     print(x, end="")
...:
1234
```

Building up a list piece-by-piece is a common task, which can easily be done in a for-loop. List comprehension provide a compact syntax to handle this task.

```
In [64]: x = [1, 2, 3, 4]

In [65]: zip(x, x[::-1])
Out[65]: [(1, 4), (2, 3), (3, 2), (4, 1)]

In [66]: [y for y in zip(x, x[::-1]) if y[0] > y[1]]
Out[66]: [(3, 2), (4, 1)]
```

#### Exercises

- Write a for-loop that produces [(3, 2), (4, 1)] from x. How does it compare to the list comprehension above?
- Use `print?` to see what the `end` argument to the print function does. Are there any additional arguments to `print()`? If so, try using the additional arguments.
- Find the section on the `range()` function in Python tutorial. Rewrite the two for-loops above using it rather than explicitly constructing the list of numbers.
- See what `[1, 2, 3] + 3` returns. Try to explain what happened and why. In R, when you add a scalar to a vector the result is the element-wise addition.

```
> 3 + c(1,2,3)
[1] 4 5 6
```

Use list comprehension to perform element-wise addition of a scalar to a list of scalars.

### 1.4.3 Functions

- <https://docs.python.org/2/tutorial/controlflow.html#defining-functions>

```
In [105]: def add(x, y):
...:     return x+y
...:
```

```
In [106]: add(2, 3)
```



```
Out[106]: 5
```

```
In [105]: def add(x, y=1):
.....:     return x+y
.....:
```

```
In [106]: add(3)
```

```
Out[106]: 4
```

## 1.5 Exercise: DNA

For this exercise, please see `ex1/dna.py`.

## 1.6 Classes

- <https://docs.python.org/2/tutorial/classes.html>

```
In [224]: class Rectangle(object):
.....:     def __init__(self, height, width):
.....:         self.height = height
.....:         self.width = width
.....:     def __repr__(self):
.....:         return "{0} by {1}".format(self.height, self.width)
.....:     def area(self):
.....:         return self.height*self.width
.....:
```

```
In [225]: x = Rectangle(10,5)
```

```
In [228]: x
```

```
Out[228]: 10 by 5
```

```
In [229]: x.area()
```

```
Out[229]: 50
```

## 1.7 Exercise: Cipher

For this exercise, please see `ex2/cipher.py`.

## 1.8 Data formats

### 1.8.1 CSV

- <https://docs.python.org/2/library/csv.html>

The Python standard library provides a package for reading and writing CSV files. This is a somewhat low-level library, so in practice you will often use NumPy, SciPy, or Pandas CSV functionality.

### 1.8.2 JSON

- <https://docs.python.org/2/library/json.html>

However the JSON package in the standard library is much more useful.

```

In [182]: import json

In [183]: x = {"name": "Jarrod", "department": "Biostatistics"}

In [186]: with open("tmp.json", "w") as outfile:
.....:     json.dump(x, outfile)
.....:

In [187]: cat tmp.json
{"department": "Biostatistics", "name": "Jarrod"}

In [192]: with open("tmp.json") as infile:
.....:     y = json.load(infile)
.....:

In [193]: y
Out[193]: {'department': u'Biostatistics', u'name': u'Jarrod'}

```

Note that `cat` is not a Python statement. IPython is clever enough to guess that you want it to call out to the underlying operating system.

### Exercise

- One of the nice things about the JSON format is that it is so well structured that it is easy for a machine to parse, but simple enough that it is easy for humans to read. By default `json.dump` writes everything out to disk without line breaks. For readability purposes, use `json.dump?` to figure out how to pretty-print the text as well as sort it alphabetically by key.

### 1.8.3 HTML

We will use Thomas Kluyver's web scraping example notebook for this section. You can view a rendered version of it [here](#). To get an interactive version of it, you can do the following from your BASH prompt:

```

$ git clone https://github.com/dlab-berkeley/python-fundamentals.git
$ cd python-fundamentals/cheat-sheets/
$ ipython notebook Web-Scraping.ipynb

```

## 1.9 Standard library

- <https://docs.python.org/2/tutorial/stdlib.html>

### 1.9.1 Batteries included

Python provides a wealth of functionality in its huge standard library. We've already seen several (e.g., `math`, `csv`, `json`). If you need some functionality the standard library is one of the first places to look.

Here are a couple packages that you may find useful.

### 1.9.2 os

- <https://docs.python.org/2/tutorial/stdlib.html#operating-system-interface>

```
In [147]: import os
```

```
In [148]: os.getcwd()
```

```
Out[148]: '/home/jarrodd'
```

```
In [149]: pwd
```

```
Out[149]: u'/home/jarrodd'
```

### Exercise

- Use `os?` and `dir(os)` to explore the `os` package.

### 1.9.3 re

- <https://docs.python.org/2/howto/regex.html>

The `re` package provides support for regular expressions.

## 1.10 Exercise: State of the Union addresses

For this exercise, you will revisit the State of the Union Addresses, which you worked with in Statistics 243 (problem set 3).

I've provided a Python script (`ex3/munge.py`) that scraps the web, processes the speeches, and saves the processed information to a JSON file (`ex3/speeches.json`).

You should write a Python script<sup>3</sup> to:

1. Load the data from the JSON file.
2. Count the number of speeches (you may need to do some data cleaning).
3. Create a list with just the names of the presidents.
4. Create a list with just the dates of the speeches.
5. Write a function that takes a speech object and returns the tuple (`<president>`, `<number of words>`, `<date>`).
6. Create a list of tuples by calling your function on all the speech objects.
7. Create a list of the speech dates sorted based on the length of speech.

For the last task you may want to take a look at the Python Sorting How To:

- <https://wiki.python.org/moin/HowTo/Sorting>

---

<sup>3</sup>You will probably need to explore the data interactively from an IPython prompt and in tandem write your script



## Chapter 2

# NumPy, SciPy, and matplotlib

*Introduce Python's core numerical, scientific, and plotting packages.*

- *Fernando Pérez, Brian E. Granger, and John D. Hunter. "Python: an ecosystem for scientific computing." Computing in Science & Engineering 13, no. 2 (2011): 13-21.*
- *Stéfan van der Walt, S. Chris Colbert, and Gael Varoquaux. "The NumPy array: a structure for efficient numerical computation." Computing in Science & Engineering 13, no. 2 (2011): 22-30.*
- *John D. Hunter. "Matplotlib: A 2D graphics environment." Computing in Science & Engineering 9, no. 3 (2007): 0090-95.*

## 2.1 Introduction

- <http://docs.scipy.org/doc/>
- <http://matplotlib.org/gallery.html>
- <https://scipy-lectures.github.io/>
- <https://github.com/ipython/ipython/wiki/A-gallery-of-interesting-IPython-Notebooks>

## 2.2 NumPy and matplotlib

### 2.2.1 Exercise: lock 'n load

For this exercise please work through Stéfan van der Walt's NumPy lock 'n load.

- [https://github.com/stefanv/teaching/tree/master/2008\\_numpy\\_load\\_n\\_load](https://github.com/stefanv/teaching/tree/master/2008_numpy_load_n_load)

### 2.2.2 ndarray

- [http://scipy-lectures.github.io/intro/numpy/array\\_object.html](http://scipy-lectures.github.io/intro/numpy/array_object.html)
- <http://scipy-lectures.github.io/intro/numpy/operations.html>

### 2.2.3 2D plotting

- <http://scipy-lectures.github.io/intro/matplotlib/matplotlib.html>

## 2.3 Example: random walk redux

Recall the random walk simulation from 243.

```
import numpy as np
import matplotlib.pyplot as plt

np.random.seed(2)

code = {"up": (0,1),
        "down": (0,-1),
        "left": (-1,0),
        "right": (1,0)}

def random_2d_walk(nsteps=100):
    steps = np.random.choice(code.keys(), nsteps)
    walk = np.array([code[step] for step in steps])
    xy = walk.cumsum(axis=0)
    return xy

xy = random_2d_walk()
plt.plot(xy[:,0], xy[:,1])
plt.savefig("test.png")
```

Here is a version implemented as a class.

```
import numpy as np
import matplotlib.pyplot as plt

class Random2DWalk(object):

    def __init__(self, start=(0,0)):
        self.steps = ["start"]
        self.walk = np.array([start])
        self._code = {"up": (0,1),
                      "down": (0,-1),
                      "left": (-1,0),
                      "right": (1,0)}

    def __str__(self):
        current_position = self.position()[-1]
        total_steps = len(self.steps) - 1
        message = "After {0} steps you are at position: {1}"
        return message.format(total_steps, current_position)

    def step(self, nsteps=100):
        steps = np.random.choice(self._code.keys(), nsteps)
        walk = np.array([self._code[step] for step in steps])
        self.steps += steps
        self.walk = np.vstack([self.walk, walk])
        return None

    def position(self):
        return self.walk.cumsum(axis=0)
```

```

    def plot(self):
        xy = self.position()
        plt.plot(xy[:,0], xy[:,1])
        plt.savefig("test.png")
        return None

np.random.seed(2)
rw = Random2DWalk()
print rw
rw.step()
print rw
print rw.steps[:5]
print rw.walk[:5]
rw.plot()

## After 0 steps you are at position: [0 0]
## After 100 steps you are at position: [-9  7]
## ['start', 'down', 'left', 'right', 'down']
## [[ 0  0]
##   [ 0 -1]
##   [-1  0]
##   [ 1  0]
##   [ 0 -1]]

```

### 2.3.1 Exercise: Sierpinski triangle

Write a Python script to construct Sierpinski triangle using the following algorithm:

1. Choose 3 points in the plane (forming a triangle).
2. Choose another "starting" pointing (current position).
3. Randomly choose one of the corners of the triangle.
4. Move halfway from your current position to the selected corner.
5. Plot the new current position.
6. Repeat from step 3 (for 100 times).

## 2.4 SciPy

- <http://scipy-lectures.github.io/intro/scipy.html>

### 2.5 Exercise: State of the Union addresses

For this exercise, you will revisit the State of the Union Addresses. Load the data, use whatever tools and analysis you want, and use matplotlib to make plots.





## Chapter 3

# Scikit Image

*Introduction to image processing with **scikit-image** by Stéfan van der Walt. Stéfan just the Berkeley Institute for Data Science. Previously, he was a senior lecturer in applied mathematics at Stellenbosch University, South Africa, and an associate project scientist in the astronomy department, UC Berkeley. He has been involved in the development of scientific open source software since 2003 and enjoys teaching Python at workshops and conferences. Stéfan is the founder of scikit-image (an image-processing library written in the Python language) and a contributor to NumPy, SciPy and Dipy.*

### 3.1 Links

- <http://scikit-image.org/>