
Python Language Essentials

Knowledge is a treasure, but practice is the key to it.

—Thomas Fuller

People often ask me about good resources for learning Python for data-centric applications. While there are many excellent Python language books, I am usually hesitant to recommend some of them as they are intended for a general audience rather than tailored for someone who wants to load in some data sets, do some computations, and plot some of the results. There are actually a couple of books on “scientific programming in Python”, but they are geared toward numerical computing and engineering applications: solving differential equations, computing integrals, doing Monte Carlo simulations, and various topics that are more mathematically-oriented rather than being about data analysis and statistics. As this is a book about becoming proficient at working with data in Python, I think it is valuable to spend some time highlighting the most important features of Python’s built-in data structures and libraries from the perspective of processing and manipulating structured and unstructured data. As such, I will only present roughly enough information to enable you to follow along with the rest of the book.

This chapter is not intended to be an exhaustive introduction to the Python language but rather a biased, no-frills overview of features which are used repeatedly throughout this book. For new Python programmers, I recommend that you supplement this chapter with the official Python tutorial (<http://docs.python.org>) and potentially one of the many excellent (and much longer) books on general purpose Python programming. In my opinion, it is *not* necessary to become proficient at building good software in Python to be able to productively do data analysis. I encourage you to use IPython to experiment with the code examples and to explore the documentation for the various types, functions, and methods. Note that some of the code used in the examples may not necessarily be fully-introduced at this point.

Much of this book focuses on high performance array-based computing tools for working with large data sets. In order to use those tools you must often first do some munging to corral messy data into a more nicely structured form. Fortunately, Python is one of

the easiest-to-use languages for rapidly whipping your data into shape. The greater your facility with Python, the language, the easier it will be for you to prepare new data sets for analysis.

The Python Interpreter

Python is an *interpreted* language. The Python interpreter runs a program by executing one statement at a time. The standard interactive Python interpreter can be invoked on the command line with the `python` command:

```
$ python
Python 2.7.2 (default, Oct  4 2011, 20:06:09)
[GCC 4.6.1] on linux2
Type "help", "copyright", "credits" or "license" for more information.
>>> a = 5
>>> print a
5
```

The `>>>` you see is the *prompt* where you'll type expressions. To exit the Python interpreter and return to the command prompt, you can either type `exit()` or press **Ctrl-D**.

Running Python programs is as simple as calling `python` with a `.py` file as its first argument. Suppose we had created `hello_world.py` with these contents:

```
print 'Hello world'
```

This can be run from the terminal simply as:

```
$ python hello_world.py
Hello world
```

While many Python programmers execute all of their Python code in this way, many *scientific* Python programmers make use of IPython, an enhanced interactive Python interpreter. [Chapter 3](#) is dedicated to the IPython system. By using the `%run` command, IPython executes the code in the specified file in the same process, enabling you to explore the results interactively when it's done.

```
$ ipython
Python 2.7.2 |EPD 7.1-2 (64-bit)| (default, Jul  3 2011, 15:17:51)
Type "copyright", "credits" or "license" for more information.

IPython 0.12 -- An enhanced Interactive Python.
?                -> Introduction and overview of IPython's features.
%quickref        -> Quick reference.
help             -> Python's own help system.
object?         -> Details about 'object', use 'object??' for extra details.

In [1]: %run hello_world.py
Hello world

In [2]:
```

The default IPython prompt adopts the numbered In [2]: style compared with the standard >>> prompt.

The Basics

Language Semantics

The Python language design is distinguished by its emphasis on readability, simplicity, and explicitness. Some people go so far as to liken it to “executable pseudocode”.

Indentation, not braces

Python uses whitespace (tabs or spaces) to structure code instead of using braces as in many other languages like R, C++, Java, and Perl. Take the for loop in the above quicksort algorithm:

```
for x in array:
    if x < pivot:
        less.append(x)
    else:
        greater.append(x)
```

A colon denotes the start of an indented code block after which all of the code must be indented by the same amount until the end of the block. In another language, you might instead have something like:

```
for x in array {
    if x < pivot {
        less.append(x)
    } else {
        greater.append(x)
    }
}
```

One major reason that whitespace matters is that it results in most Python code looking cosmetically similar, which means less cognitive dissonance when you read a piece of code that you didn’t write yourself (or wrote in a hurry a year ago!). In a language without significant whitespace, you might stumble on some differently formatted code like:

```
for x in array
{
    if x < pivot
    {
        less.append(x)
    }
    else
    {
        greater.append(x)
    }
}
```

```
}  
}
```

Love it or hate it, significant whitespace is a fact of life for Python programmers, and in my experience it helps make Python code a lot more readable than other languages I've used. While it may seem foreign at first, I suspect that it will grow on you after a while.



I strongly recommend that you use *4 spaces* to as your default indentation and that your editor replace tabs with 4 spaces. Many text editors have a setting that will replace tab stops with spaces automatically (do this!). Some people use tabs or a different number of spaces, with 2 spaces not being terribly uncommon. 4 spaces is by and large the standard adopted by the vast majority of Python programmers, so I recommend doing that in the absence of a compelling reason otherwise.

As you can see by now, Python statements also do not need to be terminated by semicolons. Semicolons can be used, however, to separate multiple statements on a single line:

```
a = 5; b = 6; c = 7
```

Putting multiple statements on one line is generally discouraged in Python as it often makes code less readable.

Everything is an object

An important characteristic of the Python language is the consistency of its *object model*. Every number, string, data structure, function, class, module, and so on exists in the Python interpreter in its own “box” which is referred to as a *Python object*. Each object has an associated *type* (for example, *string* or *function*) and internal data. In practice this makes the language very flexible, as even functions can be treated just like any other object.

Comments

Any text preceded by the hash mark (pound sign) # is ignored by the Python interpreter. This is often used to add comments to code. At times you may also want to exclude certain blocks of code without deleting them. An easy solution is to *comment out* the code:

```
results = []  
for line in file_handle:  
    # keep the empty lines for now  
    # if len(line) == 0:  
    #     continue  
    results.append(line.replace('foo', 'bar'))
```

Function and object method calls

Functions are called using parentheses and passing zero or more arguments, optionally assigning the returned value to a variable:

```
result = f(x, y, z)
g()
```

Almost every object in Python has attached functions, known as *methods*, that have access to the object's internal contents. They can be called using the syntax:

```
obj.some_method(x, y, z)
```

Functions can take both *positional* and *keyword* arguments:

```
result = f(a, b, c, d=5, e='foo')
```

More on this later.

Variables and pass-by-reference

When assigning a variable (or *name*) in Python, you are creating a *reference* to the object on the right hand side of the equals sign. In practical terms, consider a list of integers:

```
In [241]: a = [1, 2, 3]
```

Suppose we assign *a* to a new variable *b*:

```
In [242]: b = a
```

In some languages, this assignment would cause the data `[1, 2, 3]` to be copied. In Python, *a* and *b* actually now refer to the same object, the original list `[1, 2, 3]` (see [Figure A-1](#) for a mockup). You can prove this to yourself by appending an element to *a* and then examining *b*:

```
In [243]: a.append(4)
```

```
In [244]: b
Out[244]: [1, 2, 3, 4]
```

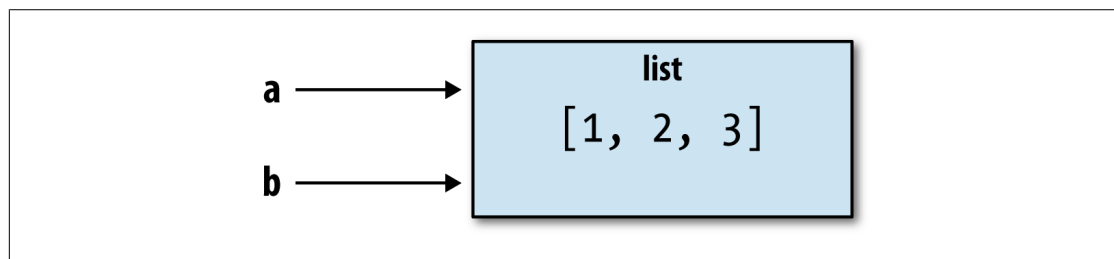


Figure A-1. Two references for the same object

Understanding the semantics of references in Python and when, how, and why data is copied is especially critical when working with larger data sets in Python.



Assignment is also referred to as *binding*, as we are binding a name to an object. Variable names that have been assigned may occasionally be referred to as bound variables.

When you pass objects as arguments to a function, you are only passing references; no copying occurs. Thus, Python is said to *pass by reference*, whereas some other languages support both pass by value (creating copies) and pass by reference. This means that a function can mutate the internals of its arguments. Suppose we had the following function:

```
def append_element(some_list, element):
    some_list.append(element)
```

Then given what's been said, this should not come as a surprise:

```
In [2]: data = [1, 2, 3]
```

```
In [3]: append_element(data, 4)
```

```
In [4]: data
```

```
Out[4]: [1, 2, 3, 4]
```

Dynamic references, strong types

In contrast with many compiled languages, such as Java and C++, object *references* in Python have no type associated with them. There is no problem with the following:

```
In [245]: a = 5          In [246]: type(a)
                                Out[246]: int
```

```
In [247]: a = 'foo'     In [248]: type(a)
                                Out[248]: str
```

Variables are names for objects within a particular namespace; the type information is stored in the object itself. Some observers might hastily conclude that Python is not a “typed language”. This is not true; consider this example:

```
In [249]: '5' + 5
-----
TypeError                                Traceback (most recent call last)
<ipython-input-249-f9dbf5f0b234> in <module>()
----> 1 '5' + 5
TypeError: cannot concatenate 'str' and 'int' objects
```

In some languages, such as Visual Basic, the string '5' might get implicitly converted (or *casted*) to an integer, thus yielding 10. Yet in other languages, such as JavaScript, the integer 5 might be casted to a string, yielding the concatenated string '55'. In this regard Python is considered a *strongly-typed* language, which means that every object has a specific type (or *class*), and implicit conversions will occur only in certain obvious circumstances, such as the following:

```
In [250]: a = 4.5
```

```
In [251]: b = 2
```

```
# String formatting, to be visited later
In [252]: print 'a is %s, b is %s' % (type(a), type(b))
a is <type 'float'>, b is <type 'int'>
```

```
In [253]: a / b
Out[253]: 2.25
```

Knowing the type of an object is important, and it's useful to be able to write functions that can handle many different kinds of input. You can check that an object is an instance of a particular type using the `isinstance` function:

```
In [254]: a = 5          In [255]: isinstance(a, int)
Out[255]: True
```

`isinstance` can accept a tuple of types if you want to check that an object's type is among those present in the tuple:

```
In [256]: a = 5; b = 4.5
```

```
In [257]: isinstance(a, (int, float))    In [258]: isinstance(b, (int, float))
Out[257]: True                          Out[258]: True
```

Attributes and methods

Objects in Python typically have both attributes, other Python objects stored “inside” the object, and methods, functions associated with an object which can have access to the object's internal data. Both of them are accessed via the syntax *obj.attribute_name*:

```
In [1]: a = 'foo'
```

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

Attributes and methods can also be accessed by name using the `getattr` function:

```
>>> getattr(a, 'split')
<function split>
```

While we will not extensively use the functions `getattr` and related functions `hasattr` and `setattr` in this book, they can be used very effectively to write generic, reusable code.

“Duck” typing

Often you may not care about the type of an object but rather only whether it has certain methods or behavior. For example, you can verify that an object is iterable if it implemented the *iterator protocol*. For many objects, this means it has a `__iter__` “magic method”, though an alternative and better way to check is to try using the `iter` function:

```
def isiterable(obj):
    try:
        iter(obj)
        return True
    except TypeError: # not iterable
        return False
```

This function would return `True` for strings as well as most Python collection types:

```
In [260]: isiterable('a string')      In [261]: isiterable([1, 2, 3])
Out[260]: True                        Out[261]: True

In [262]: isiterable(5)
Out[262]: False
```

A place where I use this functionality all the time is to write functions that can accept multiple kinds of input. A common case is writing a function that can accept any kind of sequence (list, tuple, ndarray) or even an iterator. You can first check if the object is a list (or a NumPy array) and, if it is not, convert it to be one:

```
if not isinstance(x, list) and isiterable(x):
    x = list(x)
```

Imports

In Python a *module* is simply a `.py` file containing function and variable definitions along with such things imported from other `.py` files. Suppose that we had the following module:

```
# some_module.py
PI = 3.14159

def f(x):
    return x + 2

def g(a, b):
    return a + b
```

If we wanted to access the variables and functions defined in `some_module.py`, from another file in the same directory we could do:

```
import some_module
result = some_module.f(5)
pi = some_module.PI
```

Or equivalently:

```
from some_module import f, g, PI
result = g(5, PI)
```


By using the `as` keyword you can give imports different variable names:

```
import some_module as sm
from some_module import PI as pi, g as gf

r1 = sm.f(pi)
r2 = gf(6, pi)
```

Binary operators and comparisons

Most of the binary math operations and comparisons are as you might expect:

```
In [263]: 5 - 7          In [264]: 12 + 21.5
Out[263]: -2            Out[264]: 33.5

In [265]: 5 <= 2
Out[265]: False
```

See [Table A-1](#) for all of the available binary operators.

To check if two references refer to the same object, use the `is` keyword. `is not` is also perfectly valid if you want to check that two objects are not the same:

```
In [266]: a = [1, 2, 3]

In [267]: b = a

# Note, the list function always creates a new list
In [268]: c = list(a)

In [269]: a is b          In [270]: a is not c
Out[269]: True           Out[270]: True
```

Note this is not the same thing as comparing with `==`, because in this case we have:

```
In [271]: a == c
Out[271]: True
```

A very common use of `is` and `is not` is to check if a variable is `None`, since there is only one instance of `None`:

```
In [272]: a = None

In [273]: a is None
Out[273]: True
```

Table A-1. Binary operators

Operation	Description
<code>a + b</code>	Add a and b
<code>a - b</code>	Subtract b from a
<code>a * b</code>	Multiply a by b
<code>a / b</code>	Divide a by b
<code>a // b</code>	Floor-divide a by b, dropping any fractional remainder

Operation	Description
<code>a ** b</code>	Raise a to the b power
<code>a & b</code>	True if both a and b are True. For integers, take the bitwise AND.
<code>a b</code>	True if either a or b is True. For integers, take the bitwise OR.
<code>a ^ b</code>	For booleans, True if a or b is True, but not both. For integers, take the bitwise EXCLUSIVE-OR.
<code>a == b</code>	True if a equals b
<code>a != b</code>	True if a is not equal to b
<code>a <= b, a < b</code>	True if a is less than (less than or equal) to b
<code>a > b, a >= b</code>	True if a is greater than (greater than or equal) to b
<code>a is b</code>	True if a and b reference same Python object
<code>a is not b</code>	True if a and b reference different Python objects

Strictness versus laziness

When using any programming language, it's important to understand *when* expressions are evaluated. Consider the simple expression:

```
a = b = c = 5
d = a + b * c
```

In Python, once these statements are evaluated, the calculation is immediately (or *strictly*) carried out, setting the value of `d` to 30. In another programming paradigm, such as in a pure functional programming language like Haskell, the value of `d` might not be evaluated until it is actually used elsewhere. The idea of deferring computations in this way is commonly known as *lazy evaluation*. Python, on the other hand, is a very *strict* (or *eager*) language. Nearly all of the time, computations and expressions are evaluated immediately. Even in the above simple expression, the result of `b * c` is computed as a separate step before adding it to `a`.

There are Python techniques, especially using iterators and generators, which can be used to achieve laziness. When performing very expensive computations which are only necessary some of the time, this can be an important technique in data-intensive applications.

Mutable and immutable objects

Most objects in Python are mutable, such as lists, dicts, NumPy arrays, or most user-defined types (classes). This means that the object or values that they contain can be modified.

```
In [274]: a_list = ['foo', 2, [4, 5]]
```

```
In [275]: a_list[2] = (3, 4)
```

```
In [276]: a_list
Out[276]: ['foo', 2, (3, 4)]
```

Others, like strings and tuples, are immutable:

```
In [277]: a_tuple = (3, 5, (4, 5))

In [278]: a_tuple[1] = 'four'
-----
TypeError                                 Traceback (most recent call last)
<ipython-input-278-b7966a9ae0f1> in <module>()
----> 1 a_tuple[1] = 'four'
TypeError: 'tuple' object does not support item assignment
```

Remember that just because you *can* mutate an object does not mean that you always *should*. Such actions are known in programming as *side effects*. For example, when writing a function, any side effects should be explicitly communicated to the user in the function’s documentation or comments. If possible, I recommend trying to avoid side effects and *favor immutability*, even though there may be mutable objects involved.

Scalar Types

Python has a small set of built-in types for handling numerical data, strings, boolean (True or False) values, and dates and time. See [Table A-2](#) for a list of the main scalar types. Date and time handling will be discussed separately as these are provided by the `datetime` module in the standard library.

Table A-2. Standard Python Scalar Types

Type	Description
None	The Python “null” value (only one instance of the None object exists)
str	String type. ASCII-valued only in Python 2.x and Unicode in Python 3
unicode	Unicode string type
float	Double-precision (64-bit) floating point number. Note there is no separate double type.
bool	A True or False value
int	Signed integer with maximum value determined by the platform.
long	Arbitrary precision signed integer. Large int values are automatically converted to long.

Numeric types

The primary Python types for numbers are `int` and `float`. The size of the integer which can be stored as an `int` is dependent on your platform (whether 32 or 64-bit), but Python will transparently convert a very large integer to `long`, which can store arbitrarily large integers.

```
In [279]: ival = 17239871

In [280]: ival ** 6
Out[280]: 26254519291092456596965462913230729701102721L
```

Floating point numbers are represented with the Python `float` type. Under the hood each one is a double-precision (64 bits) value. They can also be expressed using scientific notation:

```
In [281]: fval = 7.243
```

```
In [282]: fval2 = 6.78e-5
```

In Python 3, integer division not resulting in a whole number will always yield a floating point number:

```
In [284]: 3 / 2
Out[284]: 1.5
```

In Python 2.7 and below (which some readers will likely be using), you can enable this behavior by default by putting the following cryptic-looking statement at the top of your module:

```
from __future__ import division
```

Without this in place, you can always explicitly convert the denominator into a floating point number:

```
In [285]: 3 / float(2)
Out[285]: 1.5
```

To get C-style integer division (which drops the fractional part if the result is not a whole number), use the floor division operator `//`:

```
In [286]: 3 // 2
Out[286]: 1
```

Complex numbers are written using `j` for the imaginary part:

```
In [287]: cval = 1 + 2j
```

```
In [288]: cval * (1 - 2j)
Out[288]: (5+0j)
```

Strings

Many people use Python for its powerful and flexible built-in string processing capabilities. You can write *string literal* using either single quotes `'` or double quotes `"`:

```
a = 'one way of writing a string'
b = "another way"
```

For multiline strings with line breaks, you can use triple quotes, either `'''` or `"""`:

```
c = """
This is a longer string that
spans multiple lines
"""
```

Python strings are immutable; you cannot modify a string without creating a new string:

```
In [289]: a = 'this is a string'
```

```
In [290]: a[10] = 'f'
```

```
-----
TypeError                                Traceback (most recent call last)
<ipython-input-290-5ca625d1e504> in <module>()
----> 1 a[10] = 'f'
TypeError: 'str' object does not support item assignment
```

```
In [291]: b = a.replace('string', 'longer string')
```

```
In [292]: b
```

```
Out[292]: 'this is a longer string'
```

Many Python objects can be converted to a string using the `str` function:

```
In [293]: a = 5.6          In [294]: s = str(a)
```

```
In [295]: s
```

```
Out[295]: '5.6'
```

Strings are a sequence of characters and therefore can be treated like other sequences, such as lists and tuples:

```
In [296]: s = 'python'
```

```
In [297]: list(s)
```

```
Out[297]: ['p', 'y', 't', 'h', 'o', 'n']
```

```
In [298]: s[:3]
```

```
Out[298]: 'pyt'
```

The backslash character `\` is an *escape character*, meaning that it is used to specify special characters like newline `\n` or unicode characters. To write a string literal with backslashes, you need to escape them:

```
In [299]: s = '12\\34'
```

```
In [300]: print s
```

```
12\34
```

If you have a string with a lot of backslashes and no special characters, you might find this a bit annoying. Fortunately you can preface the leading quote of the string with `r` which means that the characters should be interpreted as is:

```
In [301]: s = r'this\has\no\special\characters'
```

```
In [302]: s
```

```
Out[302]: 'this\\has\\no\\special\\characters'
```

Adding two strings together concatenates them and produces a new string:

```
In [303]: a = 'this is the first half '
```

```
In [304]: b = 'and this is the second half'
```

```
In [305]: a + b
```

```
Out[305]: 'this is the first half and this is the second half'
```

String templating or formatting is another important topic. The number of ways to do so has expanded with the advent of Python 3, here I will briefly describe the mechanics of one of the main interfaces. Strings with a % followed by one or more format characters is a target for inserting a value into that string (this is quite similar to the `printf` function in C). As an example, consider this string:

```
In [306]: template = '%.2f %s are worth $%d'
```

In this string, %s means to format an argument as a string, %.2f a number with 2 decimal places, and %d an integer. To substitute arguments for these format parameters, use the binary operator % with a tuple of values:

```
In [307]: template % (4.5560, 'Argentine Pesos', 1)
Out[307]: '4.56 Argentine Pesos are worth $1'
```

String formatting is a broad topic; there are multiple methods and numerous options and tweaks available to control how values are formatted in the resulting string. To learn more, I recommend you seek out more information on the web.

I discuss general string processing as it relates to data analysis in more detail in [Chapter 7](#).

Booleans

The two boolean values in Python are written as `True` and `False`. Comparisons and other conditional expressions evaluate to either `True` or `False`. Boolean values are combined with the `and` and `or` keywords:

```
In [308]: True and True
Out[308]: True
```

```
In [309]: False or True
Out[309]: True
```

Almost all built-in Python types and any class defining the `__nonzero__` magic method have a `True` or `False` interpretation in an `if` statement:

```
In [310]: a = [1, 2, 3]
.....: if a:
.....:     print 'I found something!'
.....:
I found something!
```

```
In [311]: b = []
.....: if not b:
.....:     print 'Empty!'
.....:
Empty!
```

Most objects in Python have a notion of true- or falseness. For example, empty sequences (lists, dicts, tuples, etc.) are treated as `False` if used in control flow (as above with the empty list `b`). You can see exactly what boolean value an object coerces to by invoking `bool` on it:

```
In [312]: bool([]), bool([1, 2, 3])
Out[312]: (False, True)
```

```
In [313]: bool('Hello world!'), bool('')
Out[313]: (True, False)
```

```
In [314]: bool(0), bool(1)
Out[314]: (False, True)
```

Type casting

The `str`, `bool`, `int` and `float` types are also functions which can be used to cast values to those types:

```
In [315]: s = '3.14159'
```

```
In [316]: fval = float(s)          In [317]: type(fval)
Out[317]: float
```

```
In [318]: int(fval)          In [319]: bool(fval)          In [320]: bool(0)
Out[318]: 3                  Out[319]: True                 Out[320]: False
```

None

`None` is the Python null value type. If a function does not explicitly return a value, it implicitly returns `None`.

```
In [321]: a = None          In [322]: a is None
Out[322]: True
```

```
In [323]: b = 5             In [324]: b is not None
Out[324]: True
```

`None` is also a common default value for optional function arguments:

```
def add_and_maybe_multiply(a, b, c=None):
    result = a + b

    if c is not None:
        result = result * c

    return result
```

While a technical point, it's worth bearing in mind that `None` is not a reserved keyword but rather a unique instance of `NoneType`.

Dates and times

The built-in Python `datetime` module provides `datetime`, `date`, and `time` types. The `datetime` type as you may imagine combines the information stored in `date` and `time` and is the most commonly used:

```
In [325]: from datetime import datetime, date, time
```

```
In [326]: dt = datetime(2011, 10, 29, 20, 30, 21)
```

```
In [327]: dt.day      In [328]: dt.minute
Out[327]: 29         Out[328]: 30
```

Given a `datetime` instance, you can extract the equivalent `date` and `time` objects by calling methods on the `datetime` of the same name:

```
In [329]: dt.date()      In [330]: dt.time()
Out[329]: datetime.date(2011, 10, 29)  Out[330]: datetime.time(20, 30, 21)
```

The `strftime` method formats a `datetime` as a string:

```
In [331]: dt.strftime('%m/%d/%Y %H:%M')
Out[331]: '10/29/2011 20:30'
```

Strings can be converted (parsed) into `datetime` objects using the `strptime` function:

```
In [332]: datetime.strptime('20091031', '%Y%m%d')
Out[332]: datetime.datetime(2009, 10, 31, 0, 0)
```

See [Table 10-2](#) for a full list of format specifications.

When aggregating or otherwise grouping time series data, it will occasionally be useful to replace fields of a series of datetimes, for example replacing the minute and second fields with zero, producing a new object:

```
In [333]: dt.replace(minute=0, second=0)
Out[333]: datetime.datetime(2011, 10, 29, 20, 0)
```

The difference of two `datetime` objects produces a `datetime.timedelta` type:

```
In [334]: dt2 = datetime(2011, 11, 15, 22, 30)

In [335]: delta = dt2 - dt

In [336]: delta      In [337]: type(delta)
Out[336]: datetime.timedelta(17, 7179)  Out[337]: datetime.timedelta
```

Adding a `timedelta` to a `datetime` produces a new shifted `datetime`:

```
In [338]: dt
Out[338]: datetime.datetime(2011, 10, 29, 20, 30, 21)

In [339]: dt + delta
Out[339]: datetime.datetime(2011, 11, 15, 22, 30)
```

Control Flow

`if`, `elif`, and `else`

The `if` statement is one of the most well-known control flow statement types. It checks a condition which, if `True`, evaluates the code in the block that follows:

```
if x < 0:
    print 'It's negative'
```


An `if` statement can be optionally followed by one or more `elif` blocks and a catch-all `else` block if all of the conditions are `False`:

```
if x < 0:
    print 'It's negative'
elif x == 0:
    print 'Equal to zero'
elif 0 < x < 5:
    print 'Positive but smaller than 5'
else:
    print 'Positive and larger than or equal to 5'
```

If any of the conditions is `True`, no further `elif` or `else` blocks will be reached. With a compound condition using `and` or `or`, conditions are evaluated left-to-right and will short circuit:

```
In [340]: a = 5; b = 7

In [341]: c = 8; d = 4

In [342]: if a < b or c > d:
.....:     print 'Made it'
Made it
```

In this example, the comparison `c > d` never gets evaluated because the first comparison was `True`.

for loops

`for` loops are for iterating over a collection (like a list or tuple) or an iterator. The standard syntax for a `for` loop is:

```
for value in collection:
    # do something with value
```

A `for` loop can be advanced to the next iteration, skipping the remainder of the block, using the `continue` keyword. Consider this code which sums up integers in a list and skips `None` values:

```
sequence = [1, 2, None, 4, None, 5]
total = 0
for value in sequence:
    if value is None:
        continue
    total += value
```

A `for` loop can be exited altogether using the `break` keyword. This code sums elements of the list until a 5 is reached:

```
sequence = [1, 2, 0, 4, 6, 5, 2, 1]
total_until_5 = 0
for value in sequence:
    if value == 5:
        break
    total_until_5 += value
```

As we will see in more detail, if the elements in the collection or iterator are sequences (tuples or lists, say), they can be conveniently *unpacked* into variables in the `for` loop statement:

```
for a, b, c in iterator:
    # do something
```

while loops

A `while` loop specifies a condition and a block of code that is to be executed until the condition evaluates to `False` or the loop is explicitly ended with `break`:

```
x = 256
total = 0
while x > 0:
    if total > 500:
        break
    total += x
    x = x // 2
```

pass

`pass` is the “no-op” statement in Python. It can be used in blocks where no action is to be taken; it is only required because Python uses whitespace to delimit blocks:

```
if x < 0:
    print 'negative!'
elif x == 0:
    # TODO: put something smart here
    pass
else:
    print 'positive!'
```

It’s common to use `pass` as a place-holder in code while working on a new piece of functionality:

```
def f(x, y, z):
    # TODO: implement this function!
    pass
```

Exception handling

Handling Python errors or *exceptions* gracefully is an important part of building robust programs. In data analysis applications, many functions only work on certain kinds of input. As an example, Python’s `float` function is capable of casting a string to a floating point number, but fails with `ValueError` on improper inputs:

```
In [343]: float('1.2345')
Out[343]: 1.2345
```

```
In [344]: float('something')
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-344-439904410854> in <module>()
```

```
----> 1 float('something')
ValueError: could not convert string to float: something
```

Suppose we wanted a version of `float` that fails gracefully, returning the input argument. We can do this by writing a function that encloses the call to `float` in a `try/except` block:

```
def attempt_float(x):
    try:
        return float(x)
    except:
        return x
```

The code in the `except` part of the block will only be executed if `float(x)` raises an exception:

```
In [346]: attempt_float('1.2345')
Out[346]: 1.2345

In [347]: attempt_float('something')
Out[347]: 'something'
```

You might notice that `float` can raise exceptions other than `ValueError`:

```
In [348]: float((1, 2))
-----
TypeError                                 Traceback (most recent call last)
<ipython-input-348-842079ebb635> in <module>()
----> 1 float((1, 2))
TypeError: float() argument must be a string or a number
```

You might want to only suppress `ValueError`, since a `TypeError` (the input was not a string or numeric value) might indicate a legitimate bug in your program. To do that, write the exception type after `except`:

```
def attempt_float(x):
    try:
        return float(x)
    except ValueError:
        return x
```

We have then:

```
In [350]: attempt_float((1, 2))
-----
TypeError                                 Traceback (most recent call last)
<ipython-input-350-9bdfd730cead> in <module>()
----> 1 attempt_float((1, 2))
<ipython-input-349-3e06b8379b6b> in attempt_float(x)
      1 def attempt_float(x):
      2     try:
----> 3         return float(x)
      4     except ValueError:
      5         return x
TypeError: float() argument must be a string or a number
```

You can catch multiple exception types by writing a tuple of exception types instead (the parentheses are required):

```
def attempt_float(x):
    try:
        return float(x)
    except (TypeError, ValueError):
        return x
```

In some cases, you may not want to suppress an exception, but you want some code to be executed regardless of whether the code in the `try` block succeeds or not. To do this, use `finally`:

```
f = open(path, 'w')

try:
    write_to_file(f)
finally:
    f.close()
```

Here, the file handle `f` will *always* get closed. Similarly, you can have code that executes only if the `try` block succeeds using `else`:

```
f = open(path, 'w')

try:
    write_to_file(f)
except:
    print 'Failed'
else:
    print 'Succeeded'
finally:
    f.close()
```

range and xrange

The `range` function produces a list of evenly-spaced integers:

```
In [352]: range(10)
Out[352]: [0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
```

Both a start, end, and step can be given:

```
In [353]: range(0, 20, 2)
Out[353]: [0, 2, 4, 6, 8, 10, 12, 14, 16, 18]
```

As you can see, `range` produces integers up to but not including the endpoint. A common use of `range` is for iterating through sequences by index:

```
seq = [1, 2, 3, 4]
for i in range(len(seq)):
    val = seq[i]
```

For very long ranges, it's recommended to use `xrange`, which takes the same arguments as `range` but returns an iterator that generates integers one by one rather than generating

all of them up-front and storing them in a (potentially very large) list. This snippet sums all numbers from 0 to 9999 that are multiples of 3 or 5:

```
sum = 0
for i in xrange(10000):
    # % is the modulo operator
    if i % 3 == 0 or i % 5 == 0:
        sum += i
```



In Python 3, `range` always returns an iterator, and thus it is not necessary to use the `xrange` function

Ternary Expressions

A *ternary expression* in Python allows you combine an `if-else` block which produces a value into a single line or expression. The syntax for this in Python is

```
value = true-expr if condition else
       false-expr
```

Here, *true-expr* and *false-expr* can be any Python expressions. It has the identical effect as the more verbose

```
if condition:
    value = true-expr
else:
    value = false-expr
```

This is a more concrete example:

```
In [354]: x = 5

In [355]: 'Non-negative' if x >= 0 else 'Negative'
Out[355]: 'Non-negative'
```

As with `if-else` blocks, only one of the expressions will be evaluated. While it may be tempting to always use ternary expressions to condense your code, realize that you may sacrifice readability if the condition as well and the true and false expressions are very complex.

Data Structures and Sequences

Python's data structures are simple, but powerful. Mastering their use is a critical part of becoming a proficient Python programmer.

Tuple

A tuple is a one-dimensional, fixed-length, *immutable* sequence of Python objects. The easiest way to create one is with a comma-separated sequence of values:

```
In [356]: tup = 4, 5, 6
```

```
In [357]: tup
Out[357]: (4, 5, 6)
```

When defining tuples in more complicated expressions, it's often necessary to enclose the values in parentheses, as in this example of creating a tuple of tuples:

```
In [358]: nested_tup = (4, 5, 6), (7, 8)
```

```
In [359]: nested_tup
Out[359]: ((4, 5, 6), (7, 8))
```

Any sequence or iterator can be converted to a tuple by invoking `tuple`:

```
In [360]: tuple([4, 0, 2])
Out[360]: (4, 0, 2)
```

```
In [361]: tup = tuple('string')
```

```
In [362]: tup
Out[362]: ('s', 't', 'r', 'i', 'n', 'g')
```

Elements can be accessed with square brackets `[]` as with most other sequence types. Like C, C++, Java, and many other languages, sequences are 0-indexed in Python:

```
In [363]: tup[0]
Out[363]: 's'
```

While the objects stored in a tuple may be mutable themselves, once created it's not possible to modify which object is stored in each slot:

```
In [364]: tup = tuple(['foo', [1, 2], True])
```

```
In [365]: tup[2] = False
```

```
-----
TypeError                                Traceback (most recent call last)
<ipython-input-365-c7308343b841> in <module>()
----> 1 tup[2] = False
TypeError: 'tuple' object does not support item assignment
```

```
# however
```

```
In [366]: tup[1].append(3)
```

```
In [367]: tup
Out[367]: ('foo', [1, 2, 3], True)
```

Tuples can be concatenated using the `+` operator to produce longer tuples:

```
In [368]: (4, None, 'foo') + (6, 0) + ('bar',)
Out[368]: (4, None, 'foo', 6, 0, 'bar')
```

Multiplying a tuple by an integer, as with lists, has the effect of concatenating together that many copies of the tuple.

```
In [369]: ('foo', 'bar') * 4
Out[369]: ('foo', 'bar', 'foo', 'bar', 'foo', 'bar', 'foo', 'bar')
```

Note that the objects themselves are not copied, only the references to them.

Unpacking tuples

If you try to *assign* to a tuple-like expression of variables, Python will attempt to *unpack* the value on the right-hand side of the equals sign:

```
In [370]: tup = (4, 5, 6)

In [371]: a, b, c = tup

In [372]: b
Out[372]: 5
```

Even sequences with nested tuples can be unpacked:

```
In [373]: tup = 4, 5, (6, 7)

In [374]: a, b, (c, d) = tup

In [375]: d
Out[375]: 7
```

Using this functionality it's easy to swap variable names, a task which in many languages might look like:

```
tmp = a
a = b
b = tmp

b, a = a, b
```

One of the most common uses of variable unpacking when iterating over sequences of tuples or lists:

```
seq = [(1, 2, 3), (4, 5, 6), (7, 8, 9)]
for a, b, c in seq:
    pass
```

Another common use is for returning multiple values from a function. More on this later.

Tuple methods

Since the size and contents of a tuple cannot be modified, it is very light on instance methods. One particularly useful one (also available on lists) is `count`, which counts the number of occurrences of a value:

```
In [376]: a = (1, 2, 2, 2, 3, 4, 2)
```

```
In [377]: a.count(2)
Out[377]: 4
```

List

In contrast with tuples, lists are variable-length and their contents can be modified. They can be defined using square brackets `[]` or using the `list` type function:

```
In [378]: a_list = [2, 3, 7, None]

In [379]: tup = ('foo', 'bar', 'baz')

In [380]: b_list = list(tup)      In [381]: b_list
Out[381]: ['foo', 'bar', 'baz']

In [382]: b_list[1] = 'peekaboo' In [383]: b_list
Out[383]: ['foo', 'peekaboo', 'baz']
```

Lists and tuples are semantically similar as one-dimensional sequences of objects and thus can be used interchangeably in many functions.

Adding and removing elements

Elements can be appended to the end of the list with the `append` method:

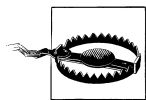
```
In [384]: b_list.append('dwarf')

In [385]: b_list
Out[385]: ['foo', 'peekaboo', 'baz', 'dwarf']
```

Using `insert` you can insert an element at a specific location in the list:

```
In [386]: b_list.insert(1, 'red')

In [387]: b_list
Out[387]: ['foo', 'red', 'peekaboo', 'baz', 'dwarf']
```



`insert` is computationally expensive compared with `append` as references to subsequent elements have to be shifted internally to make room for the new element.

The inverse operation to `insert` is `pop`, which removes and returns an element at a particular index:

```
In [388]: b_list.pop(2)
Out[388]: 'peekaboo'

In [389]: b_list
Out[389]: ['foo', 'red', 'baz', 'dwarf']
```

Elements can be removed by value using `remove`, which locates the first such value and removes it from the last:


```
In [390]: b_list.append('foo')
```

```
In [391]: b_list.remove('foo')
```

```
In [392]: b_list
```

```
Out[392]: ['red', 'baz', 'dwarf', 'foo']
```

If performance is not a concern, by using `append` and `remove`, a Python list can be used as a perfectly suitable “multi-set” data structure.

You can check if a list contains a value using the `in` keyword:

```
In [393]: 'dwarf' in b_list
```

```
Out[393]: True
```

Note that checking whether a list contains a value is a lot slower than dicts and sets as Python makes a linear scan across the values of the list, whereas the others (based on hash tables) can make the check in constant time.

Concatenating and combining lists

Similar to tuples, adding two lists together with `+` concatenates them:

```
In [394]: [4, None, 'foo'] + [7, 8, (2, 3)]
```

```
Out[394]: [4, None, 'foo', 7, 8, (2, 3)]
```

If you have a list already defined, you can append multiple elements to it using the `extend` method:

```
In [395]: x = [4, None, 'foo']
```

```
In [396]: x.extend([7, 8, (2, 3)])
```

```
In [397]: x
```

```
Out[397]: [4, None, 'foo', 7, 8, (2, 3)]
```

Note that list concatenation is a comparatively expensive operation since a new list must be created and the objects copied over. Using `extend` to append elements to an existing list, especially if you are building up a large list, is usually preferable. Thus,

```
everything = []
for chunk in list_of_lists:
    everything.extend(chunk)
```

is faster than the concatenative alternative

```
everything = []
for chunk in list_of_lists:
    everything = everything + chunk
```

Sorting

A list can be sorted in-place (without creating a new object) by calling its `sort` function:

```
In [398]: a = [7, 2, 5, 1, 3]
```

```
In [399]: a.sort()

In [400]: a
Out[400]: [1, 2, 3, 5, 7]
```

`sort` has a few options that will occasionally come in handy. One is the ability to pass a secondary *sort key*, i.e. a function that produces a value to use to sort the objects. For example, we could sort a collection of strings by their lengths:

```
In [401]: b = ['saw', 'small', 'He', 'foxes', 'six']

In [402]: b.sort(key=len)

In [403]: b
Out[403]: ['He', 'saw', 'six', 'small', 'foxes']
```

Binary search and maintaining a sorted list

The built-in `bisect` module implements binary-search and insertion into a sorted list. `bisect.bisect` finds the location where an element should be inserted to keep it sorted, while `bisect.insort` actually inserts the element into that location:

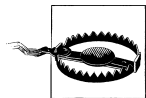
```
In [404]: import bisect

In [405]: c = [1, 2, 2, 2, 3, 4, 7]

In [406]: bisect.bisect(c, 2)      In [407]: bisect.bisect(c, 5)
Out[406]: 4                      Out[407]: 6

In [408]: bisect.insort(c, 6)

In [409]: c
Out[409]: [1, 2, 2, 2, 3, 4, 6, 7]
```



The `bisect` module functions do not check whether the list is sorted as doing so would be computationally expensive. Thus, using them with an unsorted list will succeed without error but may lead to incorrect results.

Slicing

You can select sections of list-like types (arrays, tuples, NumPy arrays) by using slice notation, which in its basic form consists of `start:stop` passed to the indexing operator `[]`:

```
In [410]: seq = [7, 2, 3, 7, 5, 6, 0, 1]

In [411]: seq[1:5]
Out[411]: [2, 3, 7, 5]
```

Slices can also be assigned to with a sequence:

```
In [412]: seq[3:4] = [6, 3]
```

```
In [413]: seq
Out[413]: [7, 2, 3, 6, 3, 5, 6, 0, 1]
```

While element at the **start** index is included, the **stop** index is not included, so that the number of elements in the result is **stop - start**.

Either the **start** or **stop** can be omitted in which case they default to the start of the sequence and the end of the sequence, respectively:

```
In [414]: seq[:5]          In [415]: seq[3:]
Out[414]: [7, 2, 3, 6, 3]  Out[415]: [6, 3, 5, 6, 0, 1]
```

Negative indices slice the sequence relative to the end:

```
In [416]: seq[-4:]         In [417]: seq[-6:-2]
Out[416]: [5, 6, 0, 1]    Out[417]: [6, 3, 5, 6]
```

Slicing semantics takes a bit of getting used to, especially if you're coming from R or MATLAB. See [Figure A-2](#) for a helpful illustrating of slicing with positive and negative integers.

A **step** can also be used after a second colon to, say, take every other element:

```
In [418]: seq[::2]
Out[418]: [7, 3, 3, 6, 1]
```

A clever use of this is to pass **-1** which has the useful effect of reversing a list or tuple:

```
In [419]: seq[::-1]
Out[419]: [1, 0, 6, 5, 3, 6, 3, 2, 7]
```

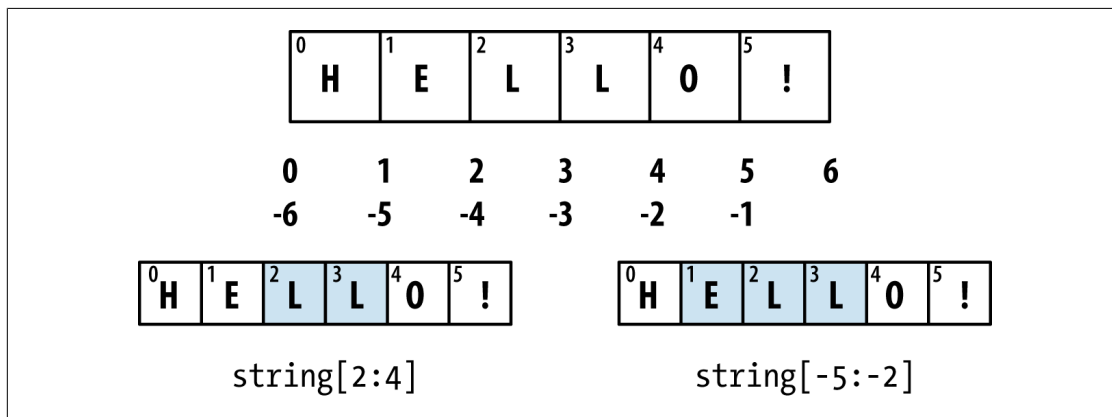


Figure A-2. Illustration of Python slicing conventions

Built-in Sequence Functions

Python has a handful of useful sequence functions that you should familiarize yourself with and use at any opportunity.

enumerate

It's common when iterating over a sequence to want to keep track of the index of the current item. A do-it-yourself approach would look like:

```
i = 0
for value in collection:
    # do something with value
    i += 1
```

Since this is so common, Python has a built-in function `enumerate` which returns a sequence of `(i, value)` tuples:

```
for i, value in enumerate(collection):
    # do something with value
```

When indexing data, a useful pattern that uses `enumerate` is computing a `dict` mapping the values of a sequence (which are assumed to be unique) to their locations in the sequence:

```
In [420]: some_list = ['foo', 'bar', 'baz']

In [421]: mapping = dict((v, i) for i, v in enumerate(some_list))

In [422]: mapping
Out[422]: {'bar': 1, 'baz': 2, 'foo': 0}
```

sorted

The `sorted` function returns a new sorted list from the elements of any sequence:

```
In [423]: sorted([7, 1, 2, 6, 0, 3, 2])
Out[423]: [0, 1, 2, 2, 3, 6, 7]

In [424]: sorted('horse race')
Out[424]: [' ', 'a', 'c', 'e', 'e', 'h', 'o', 'r', 'r', 's']
```

A common pattern for getting a sorted list of the unique elements in a sequence is to combine `sorted` with `set`:

```
In [425]: sorted(set('this is just some string'))
Out[425]: [' ', 'e', 'g', 'h', 'i', 'j', 'm', 'n', 'o', 'r', 's', 't', 'u']
```

zip

`zip` “pairs” up the elements of a number of lists, tuples, or other sequences, to create a list of tuples:

```
In [426]: seq1 = ['foo', 'bar', 'baz']

In [427]: seq2 = ['one', 'two', 'three']

In [428]: zip(seq1, seq2)
Out[428]: [('foo', 'one'), ('bar', 'two'), ('baz', 'three')]
```

`zip` can take an arbitrary number of sequences, and the number of elements it produces is determined by the *shortest* sequence:

```
In [429]: seq3 = [False, True]

In [430]: zip(seq1, seq2, seq3)
Out[430]: [('foo', 'one', False), ('bar', 'two', True)]
```

A very common use of `zip` is for simultaneously iterating over multiple sequences, possibly also combined with `enumerate`:

```
In [431]: for i, (a, b) in enumerate(zip(seq1, seq2)):
.....:     print('%d: %s, %s' % (i, a, b))
.....:
0: foo, one
1: bar, two
2: baz, three
```

Given a “zipped” sequence, `zip` can be applied in a clever way to “unzip” the sequence. Another way to think about this is converting a list of *rows* into a list of *columns*. The syntax, which looks a bit magical, is:

```
In [432]: pitchers = [('Nolan', 'Ryan'), ('Roger', 'Clemens'),
.....:                ('Schilling', 'Curt')]

In [433]: first_names, last_names = zip(*pitchers)

In [434]: first_names
Out[434]: ('Nolan', 'Roger', 'Schilling')

In [435]: last_names
Out[435]: ('Ryan', 'Clemens', 'Curt')
```

We’ll look in more detail at the use of `*` in a function call. It is equivalent to the following:

```
zip(seq[0], seq[1], ..., seq[len(seq) - 1])
```

reversed

`reversed` iterates over the elements of a sequence in reverse order:

```
In [436]: list(reversed(range(10)))
Out[436]: [9, 8, 7, 6, 5, 4, 3, 2, 1, 0]
```

Dict

`dict` is likely the most important built-in Python data structure. A more common name for it is *hash map* or *associative array*. It is a flexibly-sized collection of *key-value* pairs, where *key* and *value* are Python objects. One way to create one is by using curly braces `{}` and using colons to separate keys and values:

```
In [437]: empty_dict = {}

In [438]: d1 = {'a' : 'some value', 'b' : [1, 2, 3, 4]}
```

```
In [439]: d1
Out[439]: {'a': 'some value', 'b': [1, 2, 3, 4]}
```

Elements can be accessed and inserted or set using the same syntax as accessing elements of a list or tuple:

```
In [440]: d1[7] = 'an integer'

In [441]: d1
Out[441]: {7: 'an integer', 'a': 'some value', 'b': [1, 2, 3, 4]}

In [442]: d1['b']
Out[442]: [1, 2, 3, 4]
```

You can check if a dict contains a key using the same syntax as with checking whether a list or tuple contains a value:

```
In [443]: 'b' in d1
Out[443]: True
```

Values can be deleted either using the `del` keyword or the `pop` method (which simultaneously returns the value and deletes the key):

```
In [444]: d1[5] = 'some value'

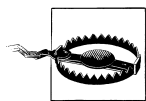
In [445]: d1['dummy'] = 'another value'

In [446]: del d1[5]

In [447]: ret = d1.pop('dummy')      In [448]: ret
Out[448]: 'another value'
```

The `keys` and `values` method give you lists of the keys and values, respectively. While the key-value pairs are not in any particular order, these functions output the keys and values in the same order:

```
In [449]: d1.keys()      In [450]: d1.values()
Out[449]: ['a', 'b', 7]  Out[450]: ['some value', [1, 2, 3, 4], 'an integer']
```



If you're using Python 3, `dict.keys()` and `dict.values()` are iterators instead of lists.

One dict can be merged into another using the `update` method:

```
In [451]: d1.update({'b' : 'foo', 'c' : 12})

In [452]: d1
Out[452]: {7: 'an integer', 'a': 'some value', 'b': 'foo', 'c': 12}
```

Creating dicts from sequences

It's common to occasionally end up with two sequences that you want to pair up element-wise in a dict. As a first cut, you might write code like this:

```
mapping = {}
for key, value in zip(key_list, value_list):
    mapping[key] = value
```

Since a dict is essentially a collection of 2-tuples, it should be no shock that the `dict` type function accepts a list of 2-tuples:

```
In [453]: mapping = dict(zip(range(5), reversed(range(5))))
```

```
In [454]: mapping
Out[454]: {0: 4, 1: 3, 2: 2, 3: 1, 4: 0}
```

In a later section we'll talk about *dict comprehensions*, another elegant way to construct dicts.

Default values

It's very common to have logic like:

```
if key in some_dict:
    value = some_dict[key]
else:
    value = default_value
```

Thus, the dict methods `get` and `pop` can take a default value to be returned, so that the above if-else block can be written simply as:

```
value = some_dict.get(key, default_value)
```

`get` by default will return `None` if the key is not present, while `pop` will raise an exception. With *setting* values, a common case is for the values in a dict to be other collections, like lists. For example, you could imagine categorizing a list of words by their first letters as a dict of lists:

```
In [455]: words = ['apple', 'bat', 'bar', 'atom', 'book']
```

```
In [456]: by_letter = {}
```

```
In [457]: for word in words:
.....:     letter = word[0]
.....:     if letter not in by_letter:
.....:         by_letter[letter] = [word]
.....:     else:
.....:         by_letter[letter].append(word)
.....:
```

```
In [458]: by_letter
Out[458]: {'a': ['apple', 'atom'], 'b': ['bat', 'bar', 'book']}
```

The `setdefault` dict method is for precisely this purpose. The if-else block above can be rewritten as:

```
by_letter.setdefault(letter, []).append(word)
```

The built-in `collections` module has a useful class, `defaultdict`, which makes this even easier. One is created by passing a type or function for generating the default value for each slot in the dict:

```
from collections import defaultdict
by_letter = defaultdict(list)
for word in words:
    by_letter[word[0]].append(word)
```

The initializer to `defaultdict` only needs to be a callable object (e.g. any function), not necessarily a type. Thus, if you wanted the default value to be 4 you could pass a function returning 4

```
counts = defaultdict(lambda: 4)
```

Valid dict key types

While the values of a dict can be any Python object, the keys have to be immutable objects like scalar types (int, float, string) or tuples (all the objects in the tuple need to be immutable, too). The technical term here is *hashability*. You can check whether an object is hashable (can be used as a key in a dict) with the `hash` function:

```
In [459]: hash('string')
Out[459]: -9167918882415130555
```

```
In [460]: hash((1, 2, (2, 3)))
Out[460]: 1097636502276347782
```

```
In [461]: hash((1, 2, [2, 3])) # fails because lists are mutable
```

```
-----
TypeError                                 Traceback (most recent call last)
<ipython-input-461-800cd14ba8be> in <module>()
----> 1 hash((1, 2, [2, 3])) # fails because lists are mutable
TypeError: unhashable type: 'list'
```

To use a list as a key, an easy fix is to convert it to a tuple:

```
In [462]: d = {}
```

```
In [463]: d[tuple([1, 2, 3])] = 5
```

```
In [464]: d
Out[464]: {(1, 2, 3): 5}
```

Set

A set is an unordered collection of unique elements. You can think of them like dicts, but keys only, no values. A set can be created in two ways: via the `set` function or using a *set literal* with curly braces:

```
In [465]: set([2, 2, 2, 1, 3, 3])
Out[465]: set([1, 2, 3])
```



```
In [466]: {2, 2, 2, 1, 3, 3}
Out[466]: set([1, 2, 3])
```

Sets support mathematical *set operations* like union, intersection, difference, and symmetric difference. See [Table A-3](#) for a list of commonly used set methods.

```
In [467]: a = {1, 2, 3, 4, 5}
```

```
In [468]: b = {3, 4, 5, 6, 7, 8}
```

```
In [469]: a | b # union (or)
Out[469]: set([1, 2, 3, 4, 5, 6, 7, 8])
```

```
In [470]: a & b # intersection (and)
Out[470]: set([3, 4, 5])
```

```
In [471]: a - b # difference
Out[471]: set([1, 2])
```

```
In [472]: a ^ b # symmetric difference (xor)
Out[472]: set([1, 2, 6, 7, 8])
```

You can also check if a set is a subset of (is contained in) or a superset of (contains all elements of) another set:

```
In [473]: a_set = {1, 2, 3, 4, 5}
```

```
In [474]: {1, 2, 3}.issubset(a_set)
Out[474]: True
```

```
In [475]: a_set.issuperset({1, 2, 3})
Out[475]: True
```

As you might guess, sets are equal if their contents are equal:

```
In [476]: {1, 2, 3} == {3, 2, 1}
Out[476]: True
```

Table A-3. Python Set Operations

Function	Alternate Syntax	Description
<code>a.add(x)</code>	N/A	Add element <code>x</code> to the set <code>a</code>
<code>a.remove(x)</code>	N/A	Remove element <code>x</code> from the set <code>a</code>
<code>a.union(b)</code>	<code>a b</code>	All of the unique elements in <code>a</code> and <code>b</code> .
<code>a.intersection(b)</code>	<code>a & b</code>	All of the elements in <i>both</i> <code>a</code> and <code>b</code> .
<code>a.difference(b)</code>	<code>a - b</code>	The elements in <code>a</code> that are not in <code>b</code> .
<code>a.symmetric_difference(b)</code>	<code>a ^ b</code>	All of the elements in <code>a</code> or <code>b</code> but <i>not both</i> .
<code>a.issubset(b)</code>	N/A	True if the elements of <code>a</code> are all contained in <code>b</code> .
<code>a.issuperset(b)</code>	N/A	True if the elements of <code>b</code> are all contained in <code>a</code> .
<code>a.isdisjoint(b)</code>	N/A	True if <code>a</code> and <code>b</code> have no elements in common.

List, Set, and Dict Comprehensions

List comprehensions are one of the most-loved Python language features. They allow you to concisely form a new list by filtering the elements of a collection and transforming the elements passing the filter in one concise expression. They take the basic form:

```
[expr for val in collection if condition]
```

This is equivalent to the following `for` loop:

```
result = []
for val in collection:
    if condition:
        result.append(expr)
```

The filter condition can be omitted, leaving only the expression. For example, given a list of strings, we could filter out strings with length 2 or less and also convert them to uppercase like this:

```
In [477]: strings = ['a', 'as', 'bat', 'car', 'dove', 'python']
```

```
In [478]: [x.upper() for x in strings if len(x) > 2]
Out[478]: ['BAT', 'CAR', 'DOVE', 'PYTHON']
```

Set and dict comprehensions are a natural extension, producing sets and dicts in a idiomatically similar way instead of lists. A dict comprehension looks like this:

```
dict_comp = {key-expr : value-expr for value in collection
              if condition}
```

A set comprehension looks like the equivalent list comprehension except with curly braces instead of square brackets:

```
set_comp = {expr for value in collection if condition}
```

Like list comprehensions, set and dict comprehensions are just syntactic sugar, but they similarly can make code both easier to write and read. Consider the list of strings above. Suppose we wanted a set containing just the lengths of the strings contained in the collection; this could be easily computed using a set comprehension:

```
In [479]: unique_lengths = {len(x) for x in strings}
```

```
In [480]: unique_lengths
Out[480]: set([1, 2, 3, 4, 6])
```

As a simple dict comprehension example, we could create a lookup map of these strings to their locations in the list:

```
In [481]: loc_mapping = {val : index for index, val in enumerate(strings)}
```

```
In [482]: loc_mapping
Out[482]: {'a': 0, 'as': 1, 'bat': 2, 'car': 3, 'dove': 4, 'python': 5}
```

Note that this dict could be equivalently constructed by:

```
loc_mapping = dict((val, idx) for idx, val in enumerate(strings))
```

The dict comprehension version is shorter and cleaner in my opinion.



Dict and set comprehensions were added to Python fairly recently in Python 2.7 and Python 3.1+.

Nested list comprehensions

Suppose we have a list of lists containing some boy and girl names:

```
In [483]: all_data = [['Tom', 'Billy', 'Jefferson', 'Andrew', 'Wesley', 'Steven', 'Joe'],
.....:               ['Susie', 'Casey', 'Jill', 'Ana', 'Eva', 'Jennifer', 'Stephanie']]
```

You might have gotten these names from a couple of files and decided to keep the boy and girl names separate. Now, suppose we wanted to get a single list containing all names with two or more e's in them. We could certainly do this with a simple `for` loop:

```
names_of_interest = []
for names in all_data:
    enough_es = [name for name in names if name.count('e') >= 2]
    names_of_interest.extend(enough_es)
```

You can actually wrap this whole operation up in a single *nested list comprehension*, which will look like:

```
In [484]: result = [name for names in all_data for name in names
.....:               if name.count('e') >= 2]

In [485]: result
Out[485]: ['Jefferson', 'Wesley', 'Steven', 'Jennifer', 'Stephanie']
```

At first, nested list comprehensions are a bit hard to wrap your head around. The `for` parts of the list comprehension are arranged according to the order of nesting, and any filter condition is put at the end as before. Here is another example where we “flatten” a list of tuples of integers into a simple list of integers:

```
In [486]: some_tuples = [(1, 2, 3), (4, 5, 6), (7, 8, 9)]

In [487]: flattened = [x for tup in some_tuples for x in tup]

In [488]: flattened
Out[488]: [1, 2, 3, 4, 5, 6, 7, 8, 9]
```

Keep in mind that the order of the `for` expressions would be the same if you wrote a nested `for` loop instead of a list comprehension:

```
flattened = []

for tup in some_tuples:
    for x in tup:
        flattened.append(x)
```

You can have arbitrarily many levels of nesting, though if you have more than two or three levels of nesting you should probably start to question your data structure design. It's important to distinguish the above syntax from a list comprehension inside a list comprehension, which is also perfectly valid:

```
In [229]: [[x for x in tup] for tup in some_tuples]
```

Functions

Functions are the primary and most important method of code organization and reuse in Python. There may not be such a thing as having too many functions. In fact, I would argue that most programmers doing data analysis don't write enough functions! As you have likely inferred from prior examples, functions are declared using the `def` keyword and returned from using the `return` keyword:

```
def my_function(x, y, z=1.5):
    if z > 1:
        return z * (x + y)
    else:
        return z / (x + y)
```

There is no issue with having multiple `return` statements. If the end of a function is reached without encountering a `return` statement, `None` is returned.

Each function can have some number of *positional* arguments and some number of *keyword* arguments. Keyword arguments are most commonly used to specify default values or optional arguments. In the above function, `x` and `y` are positional arguments while `z` is a keyword argument. This means that it can be called in either of these equivalent ways:

```
my_function(5, 6, z=0.7)
my_function(3.14, 7, 3.5)
```

The main restriction on function arguments is that the keyword arguments *must* follow the positional arguments (if any). You can specify keyword arguments in any order; this frees you from having to remember which order the function arguments were specified in and only what their names are.

Namespaces, Scope, and Local Functions

Functions can access variables in two different scopes: *global* and *local*. An alternate and more descriptive name describing a variable scope in Python is a *namespace*. Any variables that are assigned within a function by default are assigned to the local namespace. The local namespace is created when the function is called and immediately populated by the function's arguments. After the function is finished, the local namespace is destroyed (with some exceptions, see section on closures below). Consider the following function:

```
def func():
    a = []
    for i in range(5):
        a.append(i)
```

Upon calling `func()`, the empty list `a` is created, 5 elements are appended, then `a` is destroyed when the function exits. Suppose instead we had declared `a`

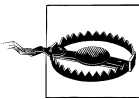
```
a = []
def func():
    for i in range(5):
        a.append(i)
```

Assigning global variables within a function is possible, but those variables must be declared as global using the `global` keyword:

```
In [489]: a = None
```

```
In [490]: def bind_a_variable():
.....:     global a
.....:     a = []
.....:     bind_a_variable()
.....:
```

```
In [491]: print a
[]
```



I generally discourage people from using the `global` keyword frequently. Typically global variables are used to store some kind of state in a system. If you find yourself using a lot of them, it's probably a sign that some object-oriented programming (using classes) is in order.

Functions can be declared anywhere, and there is no problem with having *local* functions that are dynamically created when a function is called:

```
def outer_function(x, y, z):
    def inner_function(a, b, c):
        pass
    pass
```

In the above code, the `inner_function` will not exist until `outer_function` is called. As soon as `outer_function` is done executing, the `inner_function` is destroyed.

Nested inner functions can access the local namespace of the enclosing function, but they cannot bind new variables in it. I'll talk a bit more about this in the section on closures.

In a strict sense, all functions are local to some scope, that scope may just be the module level scope.

Returning Multiple Values

When I first programmed in Python after having programmed in Java and C++, one of my favorite features was the ability to return multiple values from a function. Here's a simple example:

```
def f():
    a = 5
    b = 6
    c = 7
    return a, b, c
```

```
a, b, c = f()
```

In data analysis and other scientific applications, you will likely find yourself doing this very often as many functions may have multiple outputs, whether those are data structures or other auxiliary data computed inside the function. If you think about tuple packing and unpacking from earlier in this chapter, you may realize that what's happening here is that the function is actually just returning *one* object, namely a tuple, which is then being unpacked into the result variables. In the above example, we could have done instead:

```
return_value = f()
```

In this case, `return_value` would be, as you may guess, a 3-tuple with the three returned variables. A potentially attractive alternative to returning multiple values like above might be to return a dict instead:

```
def f():
    a = 5
    b = 6
    c = 7
    return {'a' : a, 'b' : b, 'c' : c}
```

Functions Are Objects

Since Python functions are objects, many constructs can be easily expressed that are difficult to do in other languages. Suppose we were doing some data cleaning and needed to apply a bunch of transformations to the following list of strings:

```
states = [' Alabama ', 'Georgia!', 'Georgia', 'georgia', 'FlOrIda',
          'south carolina##', 'West virginia?']
```

Anyone who has ever worked with user-submitted survey data can expect messy results like these. Lots of things need to happen to make this list of strings uniform and ready for analysis: whitespace stripping, removing punctuation symbols, and proper capitalization. As a first pass, we might write some code like:

```
import re # Regular expression module

def clean_strings(strings):
    result = []
```

```

for value in strings:
    value = value.strip()
    value = re.sub('[!#?]', '', value) # remove punctuation
    value = value.title()
    result.append(value)
return result

```

The result looks like this:

```

In [15]: clean_strings(states)
Out[15]:
['Alabama',
 'Georgia',
 'Georgia',
 'Georgia',
 'Florida',
 'South Carolina',
 'West Virginia']

```

An alternate approach that you may find useful is to make a list of the operations you want to apply to a particular set of strings:

```

def remove_punctuation(value):
    return re.sub('[!#?]', '', value)

clean_ops = [str.strip, remove_punctuation, str.title]

def clean_strings(strings, ops):
    result = []
    for value in strings:
        for function in ops:
            value = function(value)
        result.append(value)
    return result

```

Then we have

```

In [22]: clean_strings(states, clean_ops)
Out[22]:
['Alabama',
 'Georgia',
 'Georgia',
 'Georgia',
 'Florida',
 'South Carolina',
 'West Virginia']

```

A more *functional* pattern like this enables you to easily modify how the strings are transformed at a very high level. The `clean_strings` function is also now more reusable!

You can naturally use functions as arguments to other functions like the built-in `map` function, which applies a function to a collection of some kind:

```

In [23]: map(remove_punctuation, states)
Out[23]:
[' Alabama ',
 'Georgia',

```

```
'Georgia',  
'georgia',  
'FLorIda',  
'south carolina',  
'West virginia']
```

Anonymous (lambda) Functions

Python has support for so-called *anonymous* or *lambda* functions, which are really just simple functions consisting of a single statement, the result of which is the return value. They are defined using the `lambda` keyword, which has no meaning other than “we are declaring an anonymous function.”

```
def short_function(x):  
    return x * 2
```

```
equiv_anon = lambda x: x * 2
```

I usually refer to these as lambda functions in the rest of the book. They are especially convenient in data analysis because, as you’ll see, there are many cases where data transformation functions will take functions as arguments. It’s often less typing (and clearer) to pass a lambda function as opposed to writing a full-out function declaration or even assigning the lambda function to a local variable. For example, consider this silly example:

```
def apply_to_list(some_list, f):  
    return [f(x) for x in some_list]
```

```
ints = [4, 0, 1, 5, 6]  
apply_to_list(ints, lambda x: x * 2)
```

You could also have written `[x * 2 for x in ints]`, but here we were able to succinctly pass a custom operator to the `apply_to_list` function.

As another example, suppose you wanted to sort a collection of strings by the number of distinct letters in each string:

```
In [492]: strings = ['foo', 'card', 'bar', 'aaaa', 'abab']
```

Here we could pass a lambda function to the list’s `sort` method:

```
In [493]: strings.sort(key=lambda x: len(set(list(x))))
```

```
In [494]: strings  
Out[494]: ['aaaa', 'foo', 'abab', 'bar', 'card']
```



One reason lambda functions are called anonymous functions is that the function object itself is never given a name attribute.

Closures: Functions that Return Functions

Closures are nothing to fear. They can actually be a very useful and powerful tool in the right circumstance! In a nutshell, a closure is any *dynamically-generated* function returned by another function. The key property is that the returned function has access to the variables in the local namespace where it was created. Here is a very simple example:

```
def make_closure(a):
    def closure():
        print('I know the secret: %d' % a)
    return closure

closure = make_closure(5)
```

The difference between a closure and a regular Python function is that the closure continues to have access to the namespace (the function) where it was created, even though that function is done executing. So in the above case, the returned closure will always print `I know the secret: 5` whenever you call it. While it's common to create closures whose internal state (in this example, only the value of `a`) is static, you can just as easily have a mutable object like a dict, set, or list that can be modified. For example, here's a function that returns a function that keeps track of arguments it has been called with:

```
def make_watcher():
    have_seen = {}

    def has_been_seen(x):
        if x in have_seen:
            return True
        else:
            have_seen[x] = True
            return False

    return has_been_seen
```

Using this on a sequence of integers I obtain:

```
In [496]: watcher = make_watcher()

In [497]: vals = [5, 6, 1, 5, 1, 6, 3, 5]

In [498]: [watcher(x) for x in vals]
Out[498]: [False, False, False, True, True, True, False, True]
```

However, one technical limitation to keep in mind is that while you can mutate any internal state objects (like adding key-value pairs to a dict), you cannot *bind* variables in the enclosing function scope. One way to work around this is to modify a dict or list rather than binding variables:

```
def make_counter():
    count = [0]
    def counter():
```

```

        # increment and return the current count
        count[0] += 1
        return count[0]
    return counter

counter = make_counter()

```

You might be wondering why this is useful. In practice, you can write very general functions with lots of options, then fabricate simpler, more specialized functions. Here's an example of creating a string formatting function:

```

def format_and_pad(template, space):
    def formatter(x):
        return (template % x).rjust(space)

    return formatter

```

You could then create a floating point formatter that always returns a length-15 string like so:

```

In [500]: fmt = format_and_pad('%.4f', 15)

In [501]: fmt(1.756)
Out[501]: '          1.7560'

```

If you learn more about object-oriented programming in Python, you might observe that these patterns also could be implemented (albeit more verbosely) using classes.

Extended Call Syntax with `*args`, `**kwargs`

The way that function arguments work under the hood in Python is actually very simple. When you write `func(a, b, c, d=some, e=value)`, the positional and keyword arguments are actually packed up into a tuple and dict, respectively. So the internal function receives a tuple `args` and dict `kwargs` and internally does the equivalent of:

```

a, b, c = args
d = kwargs.get('d', d_default_value)
e = kwargs.get('e', e_default_value)

```

This all happens nicely behind the scenes. Of course, it also does some error checking and allows you to specify some of the positional arguments as keywords also (even if they aren't keyword in the function declaration!).

```

def say_hello_then_call_f(f, *args, **kwargs):
    print 'args is', args
    print 'kwargs is', kwargs
    print("Hello! Now I'm going to call %s" % f)
    return f(*args, **kwargs)

def g(x, y, z=1):
    return (x + y) / z

```

Then if we call `g` with `say_hello_then_call_f` we get:

```
In [8]: say_hello_then_call_f(g, 1, 2, z=5.)
args is (1, 2)
kwargs is {'z': 5.0}
Hello! Now I'm going to call <function g at 0x2dd5cf8>
Out[8]: 0.6
```

Currying: Partial Argument Application

Currying is a fun computer science term which means deriving new functions from existing ones by *partial argument application*. For example, suppose we had a trivial function that adds two numbers together:

```
def add_numbers(x, y):
    return x + y
```

Using this function, we could derive a new function of one variable, `add_five`, that adds 5 to its argument:

```
add_five = lambda y: add_numbers(5, y)
```

The second argument to `add_numbers` is said to be *curried*. There's nothing very fancy here as we really only have defined a new function that calls an existing function. The built-in `functools` module can simplify this process using the `partial` function:

```
from functools import partial
add_five = partial(add_numbers, 5)
```

When discussing pandas and time series data, we'll use this technique to create specialized functions for transforming data series

```
# compute 60-day moving average of time series x
ma60 = lambda x: pandas.rolling_mean(x, 60)

# Take the 60-day moving average of all time series in data
data.apply(ma60)
```

Generators

Having a consistent way to iterate over sequences, like objects in a list or lines in a file, is an important Python feature. This is accomplished by means of the *iterator protocol*, a generic way to make objects iterable. For example, iterating over a dict yields the dict keys:

```
In [502]: some_dict = {'a': 1, 'b': 2, 'c': 3}

In [503]: for key in some_dict:
.....:     print key,
a c b
```

When you write `for key in some_dict`, the Python interpreter first attempts to create an iterator out of `some_dict`:

```
In [504]: dict_iterator = iter(some_dict)
```

```
In [505]: dict_iterator
Out[505]: <dictionary-keyiterator at 0x10a0a1578>
```

Any iterator is any object that will yield objects to the Python interpreter when used in a context like a `for` loop. Most methods expecting a list or list-like object will also accept any iterable object. This includes built-in methods such as `min`, `max`, and `sum`, and type constructors like `list` and `tuple`:

```
In [506]: list(dict_iterator)
Out[506]: ['a', 'c', 'b']
```

A *generator* is a simple way to construct a new iterable object. Whereas normal functions execute and return a single value, generators return a sequence of values lazily, pausing after each one until the next one is requested. To create a generator, use the `yield` keyword instead of `return` in a function:

```
def squares(n=10):
    print 'Generating squares from 1 to %d' % (n ** 2)
    for i in xrange(1, n + 1):
        yield i ** 2
```

When you actually call the generator, no code is immediately executed:

```
In [2]: gen = squares()

In [3]: gen
Out[3]: <generator object squares at 0x34c8280>
```

It is not until you request elements from the generator that it begins executing its code:

```
In [4]: for x in gen:
...:     print x,
...:
Generating squares from 0 to 100
1 4 9 16 25 36 49 64 81 100
```

As a less trivial example, suppose we wished to find all unique ways to make change for \$1 (100 cents) using an arbitrary set of coins. You can probably think of various ways to implement this and how to store the unique combinations as you come up with them. One way is to write a generator that yields lists of coins (represented as integers):

```
def make_change(amount, coins=[1, 5, 10, 25], hand=None):
    hand = [] if hand is None else hand
    if amount == 0:
        yield hand
    for coin in coins:
        # ensures we don't give too much change, and combinations are unique
        if coin > amount or (len(hand) > 0 and hand[-1] < coin):
            continue

        for result in make_change(amount - coin, coins=coins,
                                  hand=hand + [coin]):
            yield result
```

The details of the algorithm are not that important (can you think of a shorter way?). Then we can write:

```
In [508]: for way in make_change(100, coins=[10, 25, 50]):
.....:     print way
[10, 10, 10, 10, 10, 10, 10, 10, 10, 10]
[25, 25, 10, 10, 10, 10, 10]
[25, 25, 25, 25]
[50, 10, 10, 10, 10, 10]
[50, 25, 25]
[50, 50]
```

```
In [509]: len(list(make_change(100)))
Out[509]: 242
```

Generator expressions

A simple way to make a generator is by using a *generator expression*. This is a generator analogue to list, dict and set comprehensions; to create one, enclose what would otherwise be a list comprehension with parenthesis instead of brackets:

```
In [510]: gen = (x ** 2 for x in xrange(100))
```

```
In [511]: gen
Out[511]: <generator object <genexpr> at 0x10a0a31e0>
```

This is completely equivalent to the following more verbose generator:

```
def _make_gen():
    for x in xrange(100):
        yield x ** 2
gen = _make_gen()
```

Generator expressions can be used inside any Python function that will accept a generator:

```
In [512]: sum(x ** 2 for x in xrange(100))
Out[512]: 328350
```

```
In [513]: dict((i, i ** 2) for i in xrange(5))
Out[513]: {0: 0, 1: 1, 2: 4, 3: 9, 4: 16}
```

itertools module

The standard library `itertools` module has a collection of generators for many common data algorithms. For example, `groupby` takes any sequence and a function; this groups consecutive elements in the sequence by return value of the function. Here's an example:

```
In [514]: import itertools
```

```
In [515]: first_letter = lambda x: x[0]
```

```
In [516]: names = ['Alan', 'Adam', 'Wes', 'Will', 'Albert', 'Steven']
```

```
In [517]: for letter, names in itertools.groupby(names, first_letter):
.....:     print letter, list(names) # names is a generator
A ['Alan', 'Adam']
```

```
W ['Wes', 'Will']
A ['Albert']
S ['Steven']
```

See [Table A-4](#) for a list of a few other `itertools` functions I’ve frequently found useful.

Table A-4. Some useful `itertools` functions

Function	Description
<code>imap(func, *iterables)</code>	Generator version of the built-in <code>map</code> ; applies <code>func</code> to each zipped tuple of the passed sequences.
<code>ifilter(func, iterable)</code>	Generator version of the built-in <code>filter</code> ; yields elements <code>x</code> for which <code>func(x)</code> is <code>True</code> .
<code>combinations(iterable, k)</code>	Generates a sequence of all possible <code>k</code> -tuples of elements in the iterable, ignoring order.
<code>permutations(iterable, k)</code>	Generates a sequence of all possible <code>k</code> -tuples of elements in the iterable, respecting order.
<code>groupby(iterable[, keyfunc])</code>	Generates (key, sub-iterator) for each unique key



In Python 3, several built-in functions (`zip`, `map`, `filter`) producing lists have been replaced by their generator versions found in `itertools` in Python 2.

Files and the operating system

Most of this book uses high-level tools like `pandas.read_csv` to read data files from disk into Python data structures. However, it’s important to understand the basics of how to work with files in Python. Fortunately, it’s very simple, which is part of why Python is so popular for text and file munging.

To open a file for reading or writing, use the built-in `open` function with either a relative or absolute file path:

```
In [518]: path = 'ch13/segismundo.txt'
```

```
In [519]: f = open(path)
```

By default, the file is opened in read-only mode `'r'`. We can then treat the file handle `f` like a list and iterate over the lines like so

```
for line in f:
    pass
```

The lines come out of the file with the end-of-line (EOL) markers intact, so you’ll often see code to get an EOL-free list of lines in a file like

```
In [520]: lines = [x.rstrip() for x in open(path)]
```

```
In [521]: lines
```

```

Out[521]:
['Sue\x03\xb1a el rico en su riqueza,',
 'que m\x03\xa1s cuidados le ofrece;',
 '',
 'sue\x03\xb1a el pobre que padece',
 'su miseria y su pobreza;',
 '',
 'sue\x03\xb1a el que a medrar empieza,',
 'sue\x03\xb1a el que afana y pretende,',
 'sue\x03\xb1a el que agravia y ofende,',
 '',
 'y en el mundo, en conclusi\x03\xb3n,',
 'todos sue\x03\xb1an lo que son,',
 'aunque ninguno lo entiende.',
 '']

```

If we had typed `f = open(path, 'w')`, a *new file* at `ch13/segismundo.txt` would have been created, overwriting any one in its place. See below for a list of all valid file read/write modes.

Table A-5. Python file modes

Mode	Description
r	Read-only mode
w	Write-only mode. Creates a new file (deleting any file with the same name)
a	Append to existing file (create it if it does not exist)
r+	Read and write
b	Add to mode for binary files, that is 'rb' or 'wb'
U	Use universal newline mode. Pass by itself 'U' or appended to one of the read modes like 'rU'

To write text to a file, you can use either the file's `write` or `writelines` methods. For example, we could create a version of `prof_mod.py` with no blank lines like so:

```

In [522]: with open('tmp.txt', 'w') as handle:
.....:     handle.writelines(x for x in open(path) if len(x) > 1)

In [523]: open('tmp.txt').readlines()
Out[523]:
['Sue\x03\xb1a el rico en su riqueza,\n',
 'que m\x03\xa1s cuidados le ofrece;\n',
 'sue\x03\xb1a el pobre que padece\n',
 'su miseria y su pobreza;\n',
 'sue\x03\xb1a el que a medrar empieza,\n',
 'sue\x03\xb1a el que afana y pretende,\n',
 'sue\x03\xb1a el que agravia y ofende,\n',
 'y en el mundo, en conclusi\x03\xb3n,\n',
 'todos sue\x03\xb1an lo que son,\n',
 'aunque ninguno lo entiende.\n']

```

See [Table A-6](#) for many of the most commonly-used file methods.

Table A-6. Important Python file methods or attributes

Method	Description
<code>read([size])</code>	Return data from file as a string, with optional <code>size</code> argument indicating the number of bytes to read
<code>readlines([size])</code>	Return list of lines in the file, with optional <code>size</code> argument
<code>readlines([size])</code>	Return list of lines (as strings) in the file
<code>write(str)</code>	Write passed string to file.
<code>writelines(strings)</code>	Write passed sequence of strings to the file.
<code>close()</code>	Close the handle
<code>flush()</code>	Flush the internal I/O buffer to disk
<code>seek(pos)</code>	Move to indicated file position (integer).
<code>tell()</code>	Return current file position as integer.
<code>closed</code>	True if the file is closed.