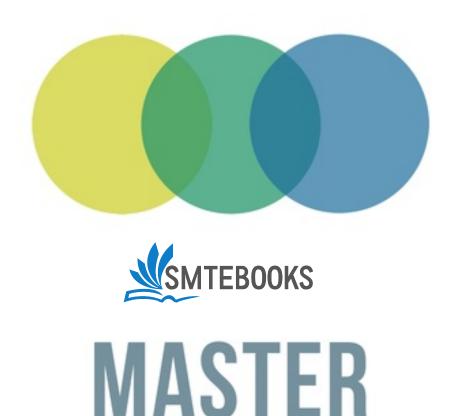
ROBERT SMALLSHIRE & AUSTIN BINGHAM

THE PYTHON



SixtyN®RTH

The Python Master

Robert Smallshire and Austin Bingham

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 $\ \odot$ 2013 - 2019 Robert Smallshire and Austin Bingham ISBN for EPUB version: 978-82-93483-08-3

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Preface

In 2013, when we incorporated our Norway-based software consultancy and training business *Sixty North*, we were courted by *Pluralsight*, a publisher of online video training material, to produce Python training videos for the rapidly growing MOOC market. At the time, we had no experience of producing video training material, but we were sure we wanted to carefully structure our introductory Python content to respect certain constraints. For example, we wanted an absolute minimum of forward references since those would be very inconvenient for our viewers. We're both men of words who live by Turing Award winner Leslie Lamport's maxim "*If you're thinking without writing you only think you're thinking*", so it was natural for us to attack video course production by first writing a script.

In the intervening years we worked on three courses with *Pluralsight*: <u>Python Fundamentals</u>, <u>Python – Beyond the Basics</u>, and <u>Advanced Python</u>. These three online courses have been transformed into three books <u>The Python Apprentice</u>, <u>The Python Journeyman</u>, and this one, <u>The Python Master</u>.

You can read *The Python Master* either as a standalone Python tutorial, or as the companion volume to the corresponding *Advanced Python* video course, depending on which style of learning suits you best. In either case we assume that you're up to speed with the material covered in the preceding books or courses.

Errata and Suggestions

All the material in this book has been thoroughly reviewed and tested; nevertheless, it's inevitable that some mistakes have crept in. If you do spot a mistake, we'd really appreciate it if you'd let us know via the *Leanpub* Python Master Discussion page so we can make amends and deploy a new version.

Conventions Used in This Book

Code examples in this book are shown in a fixed-width text which is colored with syntax highlighting:

```
>>> def square(x):
... return x * x
```

Some of our examples show code saved in files, and others — such as the one above — are from interactive Python sessions. In such interactive cases, we include the prompts from the Python session such as the triple-arrow (>>>) and triple-dot (...) prompts. You don't need to type these arrows or dots. Similarly, for operating system shell-commands we will use a dollar prompt (\$) for Linux, macOS and other Unixes, or where the particular operating system is unimportant for the task at hand:

```
$ python3 words.py
```

In this case, you don't need to type the \$ character.

For Windows-specific commands we will use a leading greater-than prompt:

```
> python words.py
```

Again, there's no need to type the > character.

For code blocks which need to be placed in a file, rather than entered interactively, we show code without any leading prompts:

We've worked hard to make sure that our lines of code are short enough so that each single logical line of code corresponds to a single physical line of code in your book. However, the vagaries of publishing e-books to different devices and the very genuine need for occasional long lines of code mean we can't guarantee that lines don't wrap. What we can guarantee, however, is that where a line does wrap, the publisher has inserted a backslash character \ in the final column. You need to use your judgement to determine whether this character is legitimate part of the code or has been added by the e-book platform.

```
>>> print("This is a single line of code which is very long. Too long, in fact, to fit on\ a single physical line of code in the book.")
```

If you see a backslash at the end of the line within the above quoted string, it is *not* part of the code, and should not be entered.

Occasionally, we'll number lines of code so we can refer to them easily from the narrative next. These line numbers should not be entered as part of the code. Numbered code blocks look like this:

```
1 def write_grayscale(filename, pixels):
       height = len(pixels)
       width = len(pixels[0])
 4
 5
      with open(filename, 'wb') as bmp:
 6
          # BMP Header
 7
           bmp.write(b'BM')
 8
           # The next four bytes hold the filesize as a 32-bit
9
10
           # little-endian integer. Zero placeholder for now.
           size_bookmark = bmp.tell()
11
           bmp.write(b'\x00\x00\x00\x00')
```

Sometimes we need to present code snippets which are incomplete. Usually this is for brevity where we are adding code to an existing block, and where we want to be clear about the block structure without repeating all existing contents of the block. In such cases we use a Python comment containing three dots # ... to indicate the elided code:

```
class Flight:
    # ...

def make_boarding_cards(self, card_printer):
    for passenger, seat in sorted(self._passenger_seats()):
        card_printer(passenger, seat, self.number(), self.aircraft_model())
```

Here it is implied that some other code already exists within the Flight class block before the make_boarding_cards() function.

Finally, within the text of the book, when we are referring to an identifier which is also a function we will use the identifier with empty parentheses, just as we did with make_boarding_cards() in the preceding paragraph.

Welcome!

Welcome to *The Python Master*. This the third in Sixty North's trilogy of books which cover the core Python language, and it builds directly on the knowledge we impart in the first two

books, *The Python Apprentice* and *The Python Journeyman*.

Our books follow a thoughtfully designed spiral curriculum. We visit the same, or closely related, topics several times in increasing depth, sometimes multiple times in the same book. For example, in *The Python Apprentice* we cover single class inheritance. Then in *The Python Journeyman* we cover multiple class inheritance. In this book we'll cover metaclasses to give you the ultimate power over class construction.

The Python Master covers some aspects of Python which you may use relatively infrequently. Mastery of the Python language calls for felicity with these features — at least to the extent that you can identify their use in existing code and appreciate their intent. We'll go all the way in this book to show you how to use the most powerful of Python's language features to greatest

effect...with occasional reminders that sometimes there is a simpler way.

Knowing how to use advanced features is what we will teach in this book. Knowing *when* to use advanced features — demonstrating a level of skilfulness that can only be achieved through time and experience — is the mark of the true master.

Specifically, we'll look at:

- Advanced flow control: loop-else clauses, try..else and switch emulation
- **Byte-oriented programming**: interpreting and manipulating data at the lowest level in Python
- **Object internals**: how objects are represented under the hood in Python

- **Descriptors**: gaining complete control over attribute access using a crucial mechanism in Python which is usually behind the scenes
- **Custom object allocation**: making the most efficient use of memory with tactics like interning of immutable objects
- **Metaclasses**: understanding the process by which class declarations are transformed into class objects and writing your own metaclasses to customise class construction
- **Class decorators**: these can be a simpler, although less powerful, alternative to metaclasses in many cases
- **Abstract base classes**: specifying and detecting class interface protocols, including how you can make built-in types become subclasses of your own types

That's a lot to cover, but over the course of this book you'll begin to see how many of these pieces fit together.

Prerequisites

A functioning Python 3.5+ runtime

First and foremost, you will need access to a working Python 3.5+1 system. Any version from 3.5 onward will suffice, and we have tried to avoid any dependencies on minor versions. With that said, more recent Python 3 versions have lots of exciting new features and standard library functionality, so if you have a choice you should probably get the most recent stable version.

At a minimum, you need to be able to run a Python 3 REPL. You can, of course, use an IDE if you wish, but we won't require anything beyond what comes with the standard Python distribution.

Experience with concepts from the previous books

In this book we'll assume that you have knowledge of — and ideally first-hand experience using — the concepts and techniques covered in *The Python Apprentice* and *The Python Journeyman*. The material in this book will be accessible to anyone who's read the previous books, but it may not be very meaningful without associated practical application of that knowledge. In other words, this book is really intended for people who have worked on a few Python projects, developers seasoned enough to appreciate *how* and *why* some of these advanced topics might be applied.

If you don't have much Python experience yet, by all means read on. But be aware that some concepts (*e.g.* metaclasses) often don't make complete sense until you've personally encountered a situation where they could be helpful. Take what you can from this book for now, continue programming, and refer back to here when you encounter such a situation.

The Road Goes On Forever

When Alexander saw the breadth of his domain, he wept for there were no more worlds to conquer.

– Plutarch (popular misquote)

This book completes our trilogy on core Python, and after reading it you will know a great deal about the language. Unlike poor Alexander the Great, though, you can save your tears: there is still a lot to Python that we haven't covered, so you can keep learning for years to come!

The material in *The Python Master* can be tricky, and much of it, by its nature, requires experience to use well. So take your time, and try to correlate what you learn with your past and ongoing Python programming experience. And once you think you understand an idea, take the ultimate test: *teach it to others*. Give conference talks, write blog posts (or books!), speak at user groups, or just show what you've learned to a colleague.

If writing these books has taught us one thing, it's that explaining things to others is the best way to find out what you don't understand. This book can help you on the way to Python mastery, but going the full distance is up to you. Enjoy the journey!

Chapter 1 - Advanced Flow Control

In this chapter we'll be looking at some more advanced, and possibly even obscure, flow-control techniques used in Python programs with which you should be familiar at the advanced level.

Specifically, we'll cover:

- else clauses on loops
- else clauses on try blocks
- emulating switch statements
- dispatching on type

By understanding these unusual language features you'll be able to understand more code that you may encounter. You'll also be able to reduce the complexity of your own code by eliminating unnecessary conditionals.

else clauses on loops

You doubtless associate the else keyword with optional clauses complementary to the conditional clause introduced by the if statement. But did you know that else can also be used to associate optional code blocks with loops? That sounds pretty weird, and to be honest it *is* an unusual language feature. That's why it's only rarely seen in the wild and most definitely deserves some explanation.

while .. else

We'll start out by saying that Guido van Rossum, inventor and benevolent dictator for life of Python, has admitted that he would not include this feature if he developed Python again. Back in our timeline though, the feature is there, and we need to understand it.

Apparently, the original motivation for using the else keyword this way in conjunction with while loops comes from Donald Knuth in early efforts to rid structured programming languages of goto. Although the choice of keyword is at

first perplexing, it's possible to rationalise the use of else in this way through comparison with the if...else construct:

```
if condition:
    execute_condition_is_true()
else:
    execute_condition_is_false()
```

In the if...else structure, the code in the if clause is executed if the condition evaluates to True when converted to bool — in other words, if the condition is 'truthy'. On the other hand, if the condition is 'falsey', the code in the else clause is executed instead.

Now look at the while . . else construct:

```
while condition:
    execute_condition_is_true()
else:
    execute_condition_is_false()
```

We can, perhaps, glimpse the logic behind choosing the else keyword. The else clause will be executed when, and only when, the condition evaluates to False. The condition may already be False when execution first reaches the while statement, so it may branch immediately into the else clause. Or there may be any number of cycles of the while-loop before the condition becomes False and execution transfers to the else block.

Fair enough, you say, but isn't this equivalent to putting the else code after the loop, rather than in a special else block, like this?:

```
while condition:
    execute_condition_is_true()
execute_condition_is_false()
```

You would be right in this simple case. But if we place a break statement within the loop body it becomes possible to exit the loop without the loop conditional ever becoming false. In that case the execute_condition_is_false() call happens even though condition *is not* False:

```
while condition:
    flag = execute_condition_is_true()
    if flag:
        break
execute_condition_is_false()
```

To fix this, we could use a second test with an if statement to provide the desired behaviour:

```
while condition:
    flag = execute_condition_is_true()
    if flag:
        break

if not condition:
    execute_condition_is_false()
```

The drawback with this approach is that the test is duplicated, which violates the Don't Repeat Yourself — or DRY — guideline, which hampers maintainability.

The while..else construct in Python allows us to avoid this second redundant test:

```
while condition:
    flag = execute_condition_is_true()
    if flag:
        break
else:
    execute_condition_is_false()
```

Now, the else block *only* executes when the main loop condition evaluates to False. If we jump out of the loop another way, such as with the break statement, execution jumps over the else clause. There's no doubt that, however you rationalise it, the choice of keyword here is confusing, and it would have been better all round if a nobreak keyword had been used to introduce this block. In lieu of such a keyword, we heartily recommend that, if you *are* tempted to use this obscure and little-used language feature, you include such a nobreak comment like this:

```
while condition:
    flag = execute_condition_is_true()
    if flag:
        break
else: # nobreak
    execute_condition_is_false()
```

So much for the theory; is this any use in practice?

We must admit that neither of the authors of this book have used while..else in practice. Almost every example we've seen could be implemented better by another, more easily understood, construct, which we'll look at later. That said, let's look at an example in evaluator.py:

```
def is_comment(item):
    return isinstance(item, str) and item.startswith('#')
def execute(program):
    """Execute a stack program.
   Args:
       program: Any stack-like collection where each item in the stack
           is a callable operator or non-callable operand. The top-most
            items on the stack may be strings beginning with '#' for
            the purposes of documentation. Stack-like means support for:
              item = stack.pop() # Remove and return the top item
              stack.append(item) # Push an item to the top
              if stack:
                                  # False in a boolean context when empty
   # Find the start of the 'program' by skipping
   # any item which is a comment.
   while program:
        item = program.pop()
        if not is comment(item):
            program.append(item)
           break
   else: # nobreak
       print("Empty program!")
        return
   # Evaluate the program
   pending = []
   while program:
        item = program.pop()
        if callable(item):
            try:
                result = item(*pending)
            except Exception as e:
                print("Error: ", e)
                break
            program.append(result)
            pending.clear()
        else:
            pending.append(item)
   else: # nobreak
        print("Program successful.")
        print("Result: ", pending)
   print("Finished")
if __name__ == '__main__':
    import operator
   program = list(reversed((
        "# A short stack program to add",
        "# and multiply some constants",
        9,
        13,
        operator.mul,
        operator add)))
   execute(program)
```

This code evaluates simple 'stack programs'. Such programs are specified as a stack of items where each item is either a callable function (for these we use any regular Python function) or an argument to that function. So to evaluate 5 + 2, we would set up the stack like this:

5 2

When the plus operator is evaluated, its result is pushed onto the stack. This allows us to perform more complex operations such as (5 + 2) * 3:

As the stack contains the expression in reverse Polish notation, the parentheses we needed in the infix version aren't required. In reality, the stack will be a Python list. The operators will be callables from the Python standard library operators module, which provides named-function equivalents of every Python infix operator. Finally, when we use Python lists as stacks the top of the stack is at the end of the list, so to get everything in the right order we'll need to reverse our list:

```
program = list(reversed([5, 2, operator.add, 3, operator.mul]))
```

For added interest, our little stack language also supports comments as strings beginning with a hash symbol, just like Python. However, such comments are only allowed at the beginning of the program, which is at the top of the stack:

```
import operator

program = list(reversed((
    "# A short stack program to add",
    "# and multiply some constants",
    5,
    2,
    operator.add,
    3,
    operator.mul)))
```

We'd like to run our little stack program by passing it to a function execute(), like this:

```
execute(program)
```

Let's see what such a function might look like and how it can use the while..else construct to good effect. The first thing our execute() function needs to do is pop all the comment strings from the top of the stack and discard them. To help with this, we'll define a simple predicate which identifies stack items as comments:

```
def is_comment(item):
    return isinstance(item, str) and item.startswith('#')
```

Notice that this function relies on an important Python feature called *boolean short-circuiting*. If item is not a string then the call to startswith() raises an AttributeError. However, when evaluating the boolean operators and and or Python will only evaluate the second operand if it is necessary for computing the result. When item is *not* as string (meaning the first operand evaluates to False) then the result of the boolean and must also be False; in this case there's no need to evaluate the second operand.

Given this useful predicate, we'll now use a while-loop to clear comment items from the top of the stack:

```
while program:
    item = program.pop()
    if not is_comment(item):
        program.append(item)
        break
else: # nobreak
    print("Empty program!")
    return
```

The conditional expression for the while statement is the program stack object itself. Remember that using a collection in a boolean context like this evaluates to True if the collection is non-empty or False if it is empty. Or put another way, empty collections are 'falsey'. So this statement reads as "While there are items remaining in the program."

The while-loop has an associated else clause where execution will jump if the while condition should ever evaluate to False. This happens when there are no more items remaining in the program. In this clause we print a warning that the program was found to be logically empty, then returning early from the execute() function.

Within the while block, we pop() an item from the stack — recall that regular Python lists have this method which removes and returns the last item from a list. We use logical negation of our is_comment() predicate to determine if the

just-popped item is *not* a comment. If the loop has reached a non-comment item, we push it back onto the stack using a call to append(), which leaves the stack with the first non-comment item on top, and then break from the loop. Remember that the while-loop else clause is best thought of as the "no break" clause, so when we break from the loop execution skips the else block and proceeds with the first statement after.

This loop executes the else block in the case of search failure — in this example if we fail to locate the first non-comment item because there isn't one. Search failure handling is probably the most widespread use of loop else clauses.

Now we know that all remaining items on the stack comprise the actual program. We'll use another while-loop to evaluate it:

```
pending = []
while program:
    item = program.pop()
    if callable(item):
        try:
            result = item(*pending)
        except Exception as e:
            print("Error: ", e)
            break
        program.append(result)
        pending.clear()
    else:
        pending.append(item)
else: # nobreak
    print("Program successful.")
    print("Result: ", pending)
```

Before this loop we set up an empty list called pending. This will be used to accumulate arguments to functions in the stack, which we'll look at shortly.

As before, the condition on the while-loop is the program stack itself, so this loop will complete, and control will be transferred to the while-loop else-clause, when the program stack is empty. This will happen when program execution is complete.

Within the while-loop we pop the top item from the stack and inspect it with the built-in callable() predicate to decide if it is a function. For clarity, we'll look at the else clause first. That's the else clause associated with the if, not the else clause associated with the while!

If the popped item is *not* callable, we append it to the pending list, and go around the loop again if the program is not yet empty.

If the item *is* callable, we try to call it, passing any pending arguments to the function using the star-args extended call syntax. Should the function call fail, we catch the exception, print an error message, and break from the while-loop. Remember this will bypass the loop else clause. Should the function call succeed, we assign the return value to result, push this value back onto the program stack, and clear the list of pending arguments.

When the program stack is empty the else block associated with the while-loop is entered. This prints "Program successful" followed by any contents of the pending list. This way a program can "return" a result by leaving non-callable values at the bottom of the stack; these will be swept up into the pending list and displayed at the end.

for-else loops

Now we understand the while..else construct we can look at the analogous for..else construct. The for..else construct may seem even more odd than while..else, given the absence of an explicit condition in the for statement, but you need to remember that the else clause is really the no-break clause. In the case of the for-loop, that's exactly when it is called — when the loop is exited without breaking. This includes the case when the iterable series over which the loop is iterating is empty.⁴

```
for item in iterable:
    if match(item):
        result = item
        break
else: # nobreak
    # No match found
    result = None

# Always come here
print(result)
```

The typical pattern of use is like this: We use a for-loop to examine each item of an iterable series, and test each item. If the item matches, we break from the loop. If we fail to find a match the code in the else block is executed, which handles the 'no match found' case.

For example, here is a code fragment which ensures that a list of integers contains at least one integer divisible by a specified value. If the supplied list does not contain a multiple of the divisor, the divisor itself is appended to the list to establish the invariant:

```
items = [2, 25, 9, 37, 28, 14]
divisor = 12

for item in items:
    if item % divisor == 0:
        found = item
        break
else: # nobreak
    items.append(divisor)
    found = divisor

print("{items} contains {found} which is a multiple of {divisor}"
    .format(**locals()))
```

We set up a list of numeric items and a divisor, which will be 12 in this case. Our for-loop iterates through the items, testing each in turn for divisibility by the divisor. If a multiple of the divisor is located, the variable found is set to the current item, and we break from the loop — skipping over the loop-else clause — and print the list of items. Should the for-loop complete without encountering a multiple of 12, the loop-else clause will be entered, which appends the divisor itself to the list, thereby ensuring that the list contains an item divisible by the divisor.

For-else clauses are more common than while-else clauses, although we must emphasise that neither are common, and both are widely misunderstood. So although we want *you* to understand them, we can't really recommend using them unless you're sure that everyone who needs to read your code is familiar with their use.

In a survey undertaken at PyCon 2011, a majority of those interviewed could not properly understand code which used loop-else clauses. Proceed with caution!

An alternative to loop else clauses

Having pointed out that loop else clauses are best avoided, it's only fair that we provide you with an alternative technique, which we think is better for several reasons, beyond avoiding an obscure Python construct.

Almost any time you see a loop else clause you can refactor it by extracting the loop into a named function, and instead of break-ing from the loop, prefer to return directly from the function. The search failure part of the code, which was in the else clause, can then be dedented a level and placed after the loop body. Doing so, our new ensure_has_divisible() function would look like this:

```
def ensure_has_divisible(items, divisor):
    for item in items:
        if item % divisor == 0:
            return item
    items.append(divisor)
    return divisor
```

which is simple enough to be understood by any Python programmer. We can use it, like this:

This is easier to understand, because it doesn't use any obscure and advanced Python flow-control techniques. It's easier to test because it is extracted into a standalone function. It's reusable because it's not mixed in with other code, and we can give it a useful and meaningful name, rather than having to put a comment in our code to explain the block.

The try..except..else construct

The third slightly oddball place we can use else blocks is as part of the try..except exception handling structure. In this case, the else clause is executed if **no** exception is raised:

```
try:
    # This code might raise an exception
    do_something()
except ValueError:
    # ValueError caught and handled
    handle_value_error()
else:
    # No exception was raised
    # We know that do_something() succeeded, so
    do_something_else()
```

Looking at this, you might wonder why we don't call do_something_else() on the line after do_something(), like this:

```
try:
    # This code might raise an exception
    do_something()
    do_something_else()
except ValueError:
    # ValueError caught and handled
    handle_value_error()
```

The downside of this approach, is that we now have no way of telling in the except block whether it was do_something() or do_something_else() which raised the exception. The enlarged scope of the try block also obscures our intent with catching the exception; where are we expecting the exception to come from?

Although rarely seen in the wild, it is useful, particularly when you have a series of operations which may raise the same exception type, but where you only want to handle exceptions from the first such, operation, as commonly happens when working with files:

```
try:
    f = open(filename, 'r')
except OSError: # OSError replaces IOError from Python 3.3 onwards
    print("File could not be opened for read")
else:
    # Now we're sure the file is open
    print("Number of lines", sum(1 for line in f))
    f.close()
```

In this example, both opening the file and iterating over the file can raise an OSError, but we're only interested in handling the exception from the call to open().

It's possible to have both an else clause and a finally clause. The else block will only be executed if there was no exception, whereas the finally clause will always be executed.

Emulating switch

Most imperative programming languages include a switch or case statement which implements a multiway branch based on the value of an expression. Here's an example for the C programming language, where different functions are called depending on the value of a menu_option variable. There's also handing for the case of 'no such option':

```
switch (menu_option) {
   case 1: single_player(); break;
   case 2: multi_player(); break;
   case 3: load_game(); break;
   case 4: save_game(); break;
   case 5: reset_high_score(); break;
   default:
        printf("No such option!");
        break;
}
```

Although switch can be emulated in Python by a chain of if..elif..else blocks, this can be tedious to write, and it's error prone because the condition must be repeated multiple times.

An alternative in Python is to use a mapping of callables. Depending on what you want to achieve, these callables may be lambdas or named functions.

We'll look at a simple adventure game you cannot win in kafka.py, which we'll refactor from using if..elif..else to using dictionaries of callables. Along the way, we'll also use try..else:

```
"""Kafka - the adventure game you cannot win."""
def play():
   position = (0, 0)
   alive = True
   while position:
        if position == (0, 0):
            print("You are in a maze of twisty passages, all alike.")
        elif position == (1, 0):
           print("You are on a road in a dark forest. To the north you can see a tower.")
        elif position == (1, 1):
            print("There is a tall tower here, with no obvious door. A path leads east.")
            print("There is nothing here.")
        command = input()
        i, j = position
        if command == "N":
           position = (i, j + 1)
        elif command == "E":
           position = (i + 1, j)
        elif command == "S":
           position = (i, j - 1)
        elif command == "W":
           position = (i - 1, j)
        elif command == "L":
        elif command == "Q":
            position = None
            print("I don't understand")
   print("Game over")
if __name__ == '__main__':
   play()
```

The game-loop uses two if..elif..else chains. The first prints information dependent on the players current position. Then, after accepting a command

from the user, the second if..elif..else chain takes action based upon the command.

Let's refactor this code to avoid those long if..elif..else chains, both of which feature repeated comparisons of the same variable against different values.

The first chain describes our current location. Fortunately in Python 3, although not in Python 2, print() is a function, and can therefore be used in an expression. We'll leverage this to build a mapping of position to callables called locations:

```
locations = {
    (0, 0): lambda: print("You are in a maze of twisty passages, all alike."),
    (1, 0): lambda: print("You are on a road in a dark forest. To the north you can see a\
tower."),
    (1, 1): lambda: print("There is a tall tower here, with no obvious door. A path leads\
east.")
    }
```

We'll look up a callable using our position in locations as a key, and call the resulting callable in a try block:

```
try:
    locations[position]()
except KeyError:
    print("There is nothing here.")
```

In fact, we don't really intend to be catching KeyError from the *callable*, only from the dictionary lookup, so this also gives us opportunity to narrow the scope of the try block using the try..else construct we learned about earlier. Here's the improved code:

```
try:
    location_action = locations[position]
except KeyError:
    print("There is nothing here.")
else:
    location_action()
```

We separate the lookup and the call into separate statements, and move the call into the else block.

Similarly, we can refactor the if..elif..else chain which handles user input into dictionary lookup for a callable. This time, though, we used named

functions rather than lambdas to avoid the restriction that lambdas can only contain expressions and not statements. Here's the branching construct:

```
actions = {
    'N': go_north,
    'E': go_east,
    'S': go_south,
    'W': go_west,
    'L': look,
    'Q': quit,
}

try:
    command_action = actions[command]
except KeyError:
    print("I don't understand")
else:
    position = command_action(position)
```

Again we split the lookup of the command action from the call to the command action.

Here are five callables referred to in the dictionary values:

```
def go_north(position):
    i, j = position
    new_position = (i, j + 1)
    return new_position
def go_east(position):
    i, j = position
    new_position = (i + 1, j)
    return new_position
def go_south(position):
    i, j = position
    new_position = (i, j - 1)
    return new_position
def go_west(position):
    i, j = position
    new_position = (i - 1, j)
    return new_position
def look(position):
    return position
def quit(position):
    return None
```

Notice that using this technique forces us into a more functional style of programming. Not only is our code broken down into many more functions, but

the bodies of those functions can't modify the state of the position variable. Instead, we pass in this value explicitly and return the new value. In the new version mutation of this variable only happens in one place, not five.

Although the new version is larger overall, we'd claim it's much more maintainable. For example, if a new piece of game state, such as the players inventory, were to be added, all command actions would be required to accept and return this value. This makes it much harder to forget to update the state than it would be in chained if..elif..else blocks.

Let's add a new "rabbit hole" location which, when the user unwittingly moves into it, leads back to the starting position of the game. To make such a change, we need to change *all* of our callables in the location mapping to accept and return a position and a liveness status. Although this may seem onerous, we think it's a good thing. Anyone maintaining the code for a particular location can now see what state needs to be maintained. Here are the location functions:

```
def labyrinth(position, alive):
    print("You are in a maze of twisty passages, all alike.")
    return position, alive

def dark_forest_road(position, alive):
    print("You are on a road in a dark forest. To the north you can see a tower.")
    return position, alive

def tall_tower(position, alive):
    print("There is a tall tower here, with no obvious door. A path leads east.")
    return position, alive

def rabbit_hole(position, alive):
    print("You fall down a rabbit hole into a labyrinth.")
    return (0, 0), alive
```

The corresponding switch in the while-loop now looks like this:

```
locations = {
    (0, 0): labyrinth,
    (1, 0): dark_forest_road,
    (1, 1): tall_tower,
    (2, 1): rabbit_hole,
    }

try:
    location_action = locations[position]
except KeyError:
    print("There is nothing here.")
else:
    position, alive = location_action(position, alive)
```

We must also update the call to location_action() to pass the current state and receive the modified state.

Now let's make the game a little more morbid by adding a deadly lava pit location which returns False for the alive status. Here's the function for the lava pit location:

```
def lava_pit(position, alive):
    print("You fall into a lava pit.")
    return position, False
```

And we must remember to add this to the location dictionary:

```
locations = {
    (0, 0): labyrinth,
    (1, 0): dark_forest_road,
    (1, 1): tall_tower,
    (2, 1): rabbit_hole,
    (1, 2): lava_pit,
}
```

We'll also add an extra conditional block after we visit the location to deal with deadly situations:

```
if not alive:
    print("You're dead!")
    break
```

Now when we die, we break out of the while loop which is the main game loop. This gives us an opportunity to use a while..else clause to handle non-lethal game loop exits, such as choosing to exit the game. Exits like this set the position variable to None, which is 'falsey':

```
while position:
    # ...
else: # nobreak
    print("You have chosen to leave the game.")
```

Now when we quit deliberately, setting position to None and causing the while-loop to terminate, we see the message from the else block associated with the while-loop:

```
You are in a maze of twisty passages, all alike.

E
You are on a road in a dark forest. To the north you can see a tower.

N
There is a tall tower here, with no obvious door. A path leads east.

O
```

```
You have chosen to leave the game.
```

But when we die by falling into the lava alive gets set to False. This causes execution to break from the loop, but we don't see the "You have chosen to leave" message as the else block is skipped:

```
You are in a maze of twisty passages, all alike.

E
You are on a road in a dark forest. To the north you can see a tower.

N
There is a tall tower here, with no obvious door. A path leads east.

N
You fall into a lava pit.
You're dead!
Game over
```

Dispatching on Type

To "dispatch" on type means that the code which will be executed depends in some way on the type of an object or objects. Python dispatches on type whenever we call a method on an object; there may be several implementations of that method in different classes, and the one that is selected depends on the type of the self object.

Ordinarily, we can't use this sort of polymorphism with regular functions. One solution is to resort to switch-emulation to route calls to the appropriate implementation by using type objects as dictionary keys. This is ungainly, and it's tricky to make it respect inheritance relationships as well as exact type matches.

singledispatch

The singledispatch decorator, which we'll introduce in this section, provides a more elegant solution to this problem.

Consider the following code which implements a simple inheritance hierarchy of shapes, specifically a circle, a parallelogram and a triangle, all of which inherit from a base class called Shape:

```
class Shape:
    def __init__(self, solid):
        self.solid = solid

class Circle(Shape):
```

```
def __init__(self, center, radius, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.center = center
       self.radius = radius
   def draw(self):
       print("\u25CF" if self.solid else "\u25A1")
class Parallelogram(Shape):
   def __init__(self, pa, pb, pc, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.pa = pa
       self.pb = pb
       self.pc = pc
   def draw(self):
        print("\u25B0" if self.solid else "\u25B1")
class Triangle(Shape):
   def __init__(self, pa, pb, pc, *args, **kwargs):
        super().__init__(*args, **kwargs)
       self.pa = pa
       self.pb = pb
       self.pc = pc
   def draw(self):
        print("\u25B2" if self.solid else "\u25B3")
def main():
    shapes = [Circle(center=(0, 0), radius=5, solid=False),
              Parallelogram(pa=(0, 0), pb=(2, 0), pc=(1, 1), solid=False),
              Triangle(pa=(0, 0), pb=(1, 2), pc=(2, 0), solid=True)]
   for shape in shapes:
        shape.draw()
if __name__ == '__main__':
   main()
```

Each class has an initializer and a draw() method. The initializers store any geometric information peculiar to that type of shape. They pass any further arguments up to the Shape base class, which stores a flag indicating whether the shape is solid.

When we say:

```
shape.draw()
```

in main(), the particular draw() that is invoked depends on whether shape is an instance of Circle, Parallelogram, or Triangle. On the receiving end, the object referred to by shape becomes referred to by the first formal parameter to

the method, which as we know is conventionally called self. So we say the call is "dispatched" to the method, depending on the type of the first argument.

This is all very well and is the way much object-oriented software is constructed, but this can lead to poor class design because it violates the single responsibility principle. Drawing isn't a behaviour inherent to shapes, still less drawing to a particular type of device. In other words, shape classes should be all about shape-ness, not about things you can do with shapes, such as drawing, serialising or clipping.

What we'd like to do is move the responsibilities which aren't intrinsic to shapes out of the shape classes. In our case, our shapes don't do anything else, so they become containers of data with no behaviour, like this:

```
class Circle(Shape):
    def __init__(self, center, radius, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.center = center
        self.radius = radius
class Parallelogram(Shape):
    def __init__(self, pa, pb, pc, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.pa = pa
        self.pb = pb
        self.pc = pc
class Triangle(Shape):
    def __init__(self, pa, pb, pc, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.pa = pa
        self.pb = pb
        self.pc = pc
```

With the drawing code removed there are several ways to implement the drawing responsibility outside of the classes. We could us a chain of if..elif..else tests using isinstance():

```
def draw(shape):
   if isinstance(shape, Circle):
        draw_circle(shape)
   elif isinstance(shape, Parallelogram):
        draw_parallelogram(shape)
   elif isinstance(shape, Triangle):
        draw_triangle(shape)
   else:
        raise TypeError("Can't draw shape")
```

In this version of draw() we test shape using up to three calls to isinstance() against Circle, Parallelogram and Triangle. If the shape object doesn't match any of those classes, we raise a TypeError. This is awkward to maintain and is rightly considered to be very poor programming style.

Another approach is to emulate a switch using dictionary look-up where the dictionary keys are types and the dictionary values are the functions which do the drawing:

```
def draw(shape):
    drawers = {
        Circle: draw_circle,
        Parallelogram: draw_parallelogram,
        Triangle: draw_triangle,
    }
    try:
        drawer = drawers(type(shape))
    except KeyError as e:
        raise TypeError("Can't draw shape") from e
    else:
        drawer(shape)
```

Here we lookup a drawer function by obtaining the type of shape in a try block, translating the KeyError to a TypeError if the lookup fails. If we're on the happy-path of no exceptions, we invoke the drawer with the shape in the else clause.

This looks better, but is actually much more fragile because we're doing *exact* type comparisons when we do the key lookup. This means that a subclass of, say, Circle wouldn't result in a call to draw_circle().

The solution to these problems arrived in Python 3.4 in the form of singledispatch, a decorator defined in the Python Standard Library functools module which performs dispatch on type. In earlier versions of Python, including Python 2, you can install the singledispatch package from the Python Package Index.

Functions which support multiple implementations dependent on the type of their arguments are called 'generic functions', and each version of the generic function is referred to as an 'overload' of the function. The act of providing another version of a generic function for different argument types is called *overloading* the function. These terms are common in statically typed languages such as C#, C++ or Java, but are rarely heard in the context of Python.

To use singledispatch we define a function decorated with the singledispatch decorator. Specifically, we define a particular version of the function which will be called if a more specific overload has not been provided. We'll come to the type-specific overloads in a moment.

At the top of the file we need to import singledispatch:

```
from functools import singledispatch
```

Then lower down we implement the generic draw() function:

```
@singledispatch
def draw(shape):
    raise TypeError("Don't know how to draw {!r}".format(shape))
```

In this case, our generic function will raise a TypeError.

Recall that decorators wrap the function to which they are applied and bind the resulting wrapper to the name of the original function. So in this case, the wrapper returned by the decorator is bound to the name draw. The draw wrapper has an attribute called register which is *also* a decorator; register() can be used to provide extra versions of the original function which work on different types. This is function overloading.

Since our overloads will all be associated with the name of the original function, draw, it doesn't matter what we call the overloads themselves. By convention we call them _, although this is by no means required. Here's an overload for Circle, another for Parallelogram, and a third for Triangle:

```
@draw.register(Circle):
    def _(shape):
        print("\u25CF" if shape.solid else "\u25A1")

@draw.register(Parallelogram)
def _(shape):
        print("\u25B0" if shape.solid else "\u25B1")

@draw.register(Triangle)
def _(shape):
    # Draw a triangle
    print("\u25B2" if shape.solid else "\u25B3")
```

By doing this we have cleanly separated concerns. Now drawing is dependent on shapes, but not shapes on drawing.

Our main function, which now looks like this, calls the global- scope generic draw function, and the singledispatch machinery will select the most specific overload if one exists, or fallback to the default implementation:

We could add other capabilities to Shape in a similar way, by defining other generic functions which behave polymorphically with respect to the shape types.

Using singledispatch with methods

You need to take care not to use the singledispatch decorator with methods. To see why, consider this attempt to implement a generic intersects() predicate method on the Circle class which can be used to determine whether a particular circle intersects instances of any of the three defined shapes:

```
class Circle(Shape):
    def __init__(self, center, radius, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.center = center
       self.radius = radius
   @singledispatch
    def intersects(self, shape):
       raise TypeError("Don't know how to compute intersection with {!r}".format(shape))
   @intersects.register(Circle)
    def _(self, shape):
        return circle_intersects_circle(self, shape)
   @intersects.register(Parallelogram)
    def _(self, shape):
       return circle_intersects_parallelogram(self, shape)
   @intersects.register(Triangle)
   def _(self, shape):
        return circle_intersects_triangle(self, shape)
```

At first sight, this looks like a reasonable approach, but there are a couple of problems here.

The first problem problem is that we can't register the type of the class currently being defined with the intersects generic function, because we have not yet finished defining it.

The second problem is more fundamental: Recall that singledispatch dispatches based only on the type of the *first* argument:

```
do_intersect = my_circle.intersects(my_parallelogram)
```

When we're calling our new method like this it's easy to forget that my_parallelogram is actually the *second* argument to Circle.intersects. my_circle is the first argument, and it's what gets bound to the self parameter. Because self will *always* be a Circle in this case our intersect() call will always dispatch to the first overload, irrespective of the type of the second argument.

This behaviour prevents the use of singledispatch with methods. All is not lost however. The solution is to move the generic function out of the class, and invoke it from a regular method which swaps the arguments. Let's take a look:

```
class Circle(Shape):
    def __init__(self, center, radius, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.center = center
       self.radius = radius
    def intersects(self, shape):
        # Delegate to the generic function, swapping arguments
       return intersects_with_circle(shape, self)
@singledispatch
def intersects_with_circle(shape, circle):
    raise TypeError("Don't know how to compute intersection of {!r} with {!r}"
                    .format(circle, shape))
@intersects_with_circle.register(Circle)
def _(shape, circle):
   return circle_intersects_circle(circle, shape)
@intersects.register(Parallelogram)
def _(shape, circle):
   return circle_intersects_parallelogram(circle, shape)
@intersects.register(Triangle)
def _(shape, circle):
   return circle_intersects_triangle(circle, shape)
```

We move the generic function intersects() out to the global scope and rename it to intersects_with_circle(). The replacement intersects() method of Circle, which accepts the formal arguments self and shape, now delegates to intersect_with_circle() with the actual arguments swapped to shape and self.

To complete this example, we would need to implement two other generic functions, intersects_with_parallelogram() and intersects_with_triangle(), although will leave that as an exercise.

Combining class-based polymorphism and overloading-based polymorphism in this way would give us a complete implementation of not just single-dispatch but *double*-dispatch, allowing us to do:

```
shape.intersects(other_shape)
```

This way, the function is selected based on the types of both shape *and* other_shape, without the shape classes themselves having any knowledge of each other, keeping coupling in the system manageable.

Summary

That just about wraps up this chapter on advanced flow control in Python 3. Let's summarise what we've covered:

- We looked at else clauses on while-loops, drawing an analogy with the
 much more well-known association between if and else. We showed how
 the else block is executed only when the while-loop condition evaluates to
 False. If the loop is exited by another means, such as via break or return,
 the else clause is *not* executed. As such, else clauses on while-loops are
 only ever useful if the loop contains a break statement somewhere within it.
- The loop-else clause is a somewhat obscure and little-used construct. We strongly advise commenting the else keyword with a "nobreak" remark so it is clear under what conditions the block is executed.
- The related for . .else clause works in an identical way: The else clause is effectively the "nobreak" clause, and is only useful if the loop contains a break statement. They are most useful with for-loops in searching. When an item is found while iterating we break from the loop skipping the else clause. If no items are found and the loop completes 'naturally' without breaking ,the else clause is executed and code handling the "not found" condition can be implemented.
- Many uses of loop else clauses particularly for searching may be better handled by extracting the loop into its own function. From here execution is returned directly when an item is found and code after the loop can handle the "not found" case. This is less obscure, more modular, more

- reusable, and more testable than using a loop else clause within a longer function.
- Next we looked at the try..except..else construct. In this case, the else clause is executed only if the try block completed successfully without any exception being raised. This allows the extent of the try block to be narrowed, making it clearer from where we are expecting exceptions to be raised.
- Python lacks a "switch" or "case" construct to implement multi-branch control flow. We showed the alternatives, including chained if..elif..else blocks, and dictionaries of callables. The latter approach also forces you to be more explicit and consistent about what is required and produced by each branch, since you must pass arguments and return values rather than mutating local state in each branch.
- We showed how to implement generic functions which dispatch on type using the singledispatch decorator available from Python 3.4. This decorator can be applied only to module scope functions, not methods, but by implementing forwarding methods and argument swapping, we can delegate to generic functions from methods. This gives us a way of implementing double-dispatch calls.
- In passing, we saw that the Python logical operators use short-circuit evaluation. This means that the operators only evaluate as many operands as are required to find the result. This can be used to 'protect' expressions from run time situations in which they would not make sense.

Chapter 2 - Byte Oriented Programming

In this chapter we'll be going low-level and look at byte-oriented programming in Python. At the bottom, everything is bits and bytes, and sometimes it's necessary to work at this level, particularly when dealing with binary data from other sources.

Remember that in Python 3 — although not in Python 2 — there is a very clear separation between text data, which are stored in the Unicode capable str type, and raw bytes, which are stored in the aptly named bytes type.

Specifically, we're going to:

- review the bitwise operators
- look at the binary representation of integers
- fill in some additional details about the bytes type
- introduce the bytearray type
- packing and unpacking binary data with the struct module
- look at the memoryview object
- memory-mapped files

Bitwise operators

Let's start right at the bottom with the bitwise operators. These will seem straightforward enough, but our exploration of them will lead us into some murky corners of Python's integer implementation.

We covered the bitwise-and, bitwise-or, and shift operators in Chapter 9 of <u>The Python Apprentice</u> when reading binary BMP image files. Now we'll complete the set by introducing the bitwise exclusive-or operator and the bitwise complement operator. Along the way, we'll also use bitwise shifts:

Operator	Description	
&	Bitwise AND	
	Bitwise OR	
\wedge	Bitwise XOR	

```
Bitwise NOTLeft shiftRight shift
```

We'll demonstrate how each of these work, but this will necessitate an interesting detour into Python's integer representation.

```
Recall that we can specify binary literals using the 0b prefix and we can display integers in binary using the built-in bin() function:

>>> 0b11110000
240
>>> bin(240)
'0b11110000'
```

Exclusive-or

The exclusive-OR operator behaves exactly as you would expect, setting bits in the output value if exactly one of the corresponding operand bits is set:

```
>>> bin(0b11100100 ^ 0b00100111)
```

Bitwise complement (bitwise not)

The bitwise complement — or NOT — operator is more unexpectedly difficult to demonstrate, although very easy to use:

```
>>> bin(~0b11110000)
```

You were probably expecting 0b00001111 as the result, although of course Python doesn't usually print leading zeros, so 0b1111 is perhaps more reasonable. Actually, leading zeros are part of the reason we get the surprising -0b1110001 result.

Two's-complement

Computer number systems typically represent negative integer numbers using a system called two's-complement. Here's a refresher on how two's complement works.

8-bit two's-complement can represent numbers from -128 to 127. Let's see how to represent the number -58 in 8-bit two's complement:

- 1. We start with the signed decimal -58, the value we want to represent.
- 2. We take the absolute value, which is 58, and represent this in 8-bit binary, which is 00111010. The same binary pattern can be interpreted as the *unsigned* decimal 58.
- 3. Because -58 is negative, we must apply additional steps 4 and 5, below. Positive numbers have the same representation in signed and unsigned representations, so don't require these additional steps.
- 4. Now we flip all the bits a bitwise-not operation to give 11000101. This is where the "two" in "two's complement" comes from: the complement operation is done in base two (or binary). This gives a bit-pattern that would be interpreted as 197 as an unsigned integer, although that's not too important here.
- 5. Finally, we add one, which gives the bit pattern 11000110 which, although it could be interpreted as the unsigned integer 198, is outside the range -128 to +127 allowed for *signed* 8-bit integers. This is the two's complement 8-bit representation for -58.

Two's complement has particular advantages over other representations, such as a sign-bit and magnitude scheme. For example, two's-complement works naturally with arithmetic operations such as addition and subtraction involving negative numbers.

Two's-complement and unlimited precision integers

Recall that Python 3 uses arbitrary precision integers. That is, Python integers are not restricted to one, two, four or eight bytes, but can use as many bytes as are necessary to store integers of any magnitude.

However, two's-complement present a particular problem when used with unlimited precision number. When we take the complement by flipping the bits, how many leading zeros should we flip? In a fixed-precision integer representation the answer is obvious. But what about variable bit-width integers. What if we flipped an unlimited (*i.e.* infinite) number of leading zeros to give an infinite number of leading ones? How would you interpret the result?

Of course, Python doesn't really represent negative integers using an infinite number of leading ones, but conceptually this is what is going on.

This is why, when asked to represent negative numbers in binary, Python actually uses a leading unary minus as part of magnitude representation, rather than a giving the unsigned binary two's-complement representation:

```
>>> bin(4)
'0b100'
>>> bin(-4)
'-0b100'
```

This means we don't easily get to see the internal bit representation of negative numbers using bin() — in fact, we can't even determine what internal representation scheme is used! This can make it tricky to inspect the behaviour of code which uses the bitwise operators.

When our use of the bitwise operators — such as the bitwise-not operator — results in a bit pattern that would normally represent a negative integer in two's complement format, Python displays that value in sign- magnitude format, obscuring the result we wanted. To get at the *actual* bit pattern, we need to do a little extra work. Let's return to our earlier example:

Binary two's complement (9-bit) Decimal Binary sign-magnitude

\mathbf{v}	0b011110000	240	+0b11110000
~v (0b100001111	-241	-0b11110001

Let's make v equal to the value 0b11110000:

```
>>> v = 0b11110000
```

This value is equal to 240 in signed decimal:

```
>>> v 240
```

Now we'll use the bitwise-not operator:

```
>>> ~v
-0b11110001
```

Here's the explanation of what's just happened. Working with the binary representation of 240 we need an 8-bit representation, although to accommodate the two's-complement representation for positive integers we need to accommodate at least one zero, so really we need at least a 9-bit representation.

It is the bits of this 9-bit representation to which the bitwise-not is applied, giving 100001111. This is the two-complement representation of -241.

Displayed back in the sign-magnitude representation Python uses when displaying binary numbers, -241 is displayed as -0b11110001.

A more intuitive two's-complement

So much for the explanation, but how do we see the flipped bits arising from our application of bitwise-not in a more intuitive way? One approach is to manually take the two's-complement of the *magnitude*, or absolute value of the number:

```
\begin{array}{ccc} & -0b11110001 \\ \text{flip} & 00001110 \\ \text{add 1} & 00001111 \end{array}
```

Unfortunately, this also uses the bitwise-not operator, and we end up chasing our tail trying to escape the cleverness of Python's arbitrary precision integers.

Another approach is to rely on the characteristics of two's complement arithmetic:

- 1. Take the signed interpretation of the two's complement binary value, in this case -241.
- 2. Add to it 2 raised to the power of the number of bits used in the representation, excluding the leading ones. In this case that's 2^8 or 256, and -241 + 256 is 15
- 3. 15 has the 00001111 bit pattern we're looking for the binary value we expected to get when we applied bitwise-not to 11110000.

Remember that the two's complement of a positive number is itself, so our function needs to take account of this:

```
>>> def twos_complement(x, num_bits):
...     if x < 0:
...         return x + (1 << num_bits)
...     return x</pre>
```

If you have difficulty understanding how this works — don't worry — it *is* tricky. Fifteen minutes with pencil, paper, and a Python interpreter will be well rewarded.

An intuitive two's-complement with bitwise-and

An less usable, but perhaps more obvious approach, is to use bitwise-and with a mask of 1s to discard all the leading 1s in negative results. This effectively specifies to how many bits precision we want the result to be presented. So for an 8 bit result we do this:

```
>>> bin(~0b11110000 & 0b11111111)
```

giving what we wanted. See how asking for a 9 bit result reveals the leading 1 of the two's complement representation:

```
>>> bin(~0b11110000 & 0b111111111)
```

In fact, since Python 3.2, we can ask Python how many bits are required to represent the integer value using the bit_length() method of the integer type, although notice that this excludes the sign:

```
>>> int(32).bit_length()
6
>>> int(240).bit_length()
8
>>> int(-241).bit_length()
8
>>> int(256).bit_length()
```

Examining bits with to_bytes()

We can also retrieve a byte-oriented representation directly as a bytes object using the to_bytes() method 5 :

```
>>> int(0xcafebabe).to_bytes(length=4, byteorder='big')
b'\xca\xfe\xba\xbe'
>>> int(0xcafebabe).to_bytes(length=4, byteorder='little')
b'\xbe\xba\xfe\xca'
```

If you want to use the native bytes order, you can retrieve the sys.byteorder value:

```
>>> import sys
>>> sys.byteorder
'little'
>>> little_cafebabe = int(0xcafebabe).to_bytes(length=4, byteorder=sys.byteorder)
>>> little_cafebabe
b'\xbe\xba\xfe\xca'
```

Of course, given some bytes we can also turn them back into an integer, using the complementary class method, from_bytes():

```
>>> int.from_bytes(little_cafebabe, byteorder=sys.byteorder)
3405691582
>>> hex(_)
'0xcafebabe'
```

Asking for the byte-oriented representation of an integer using to_bytes() will fail for negative integers with an OverflowError:

```
>>> int(-241).to_bytes(2, byteorder='big')
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
OverflowError: can't convert negative int to unsigned
```

However, if we set the optional signed argument to True, rather than its default value of False we can get a two's-complement representation back:

```
>>> int(-241).to_bytes(2, byteorder='little', signed=True)
b'\x0f\xff'
```

This indicates another way to answer the question that started this quest, by indexing into the result bytes object to retrieve the least significant byte, and converting that to a binary representation:

```
>>> bin((~0b11110000).to_bytes(2, byteorder='little', signed=True)[0])
'0b1111'
```

While this may be useful, it's certainly not concise.

The bytes type in depth

We first introduced the bytes type way back in Chapter 2 of <u>The Python Apprentice</u>. We did this to highlight the essential differences between str, which is the immutable sequence of Unicode codepoints and bytes which is the immutable sequence of bytes. This is particularly important if you're coming to Python 3 from Python 2, where str behaved differently.

Literals bytes`

At this point you should be comfortable with the bytes literal, which uses the b prefix. The default Python source code encoding is UTF-8, so the the characters used in a *literal* byte string are restricted to printable 7-bit ASCII characters — that is those with codes from 0 to 127 inclusive which aren't control codes:

```
>>> b"This is OK because it's 7-bit ASCII" b"This is OK because it's 7-bit ASCII"
```

Seven-bit control codes or characters which can't be encoded in 7-bit ASCII result in a SyntaxError:

```
>>> b"Norwegian characters like A and Ø are not 7-bit ASCII"
File "<stdin>", line 1
SyntaxError: bytes can only contain ASCII literal characters.
```

To represent other bytes with values equivalent to ASCII control codes and byte values from 128 to 255 inclusive we must use escape sequences:

```
>>> b"Norwegian characters like \xc5 and \xd8 are not 7-bit ASCII" b'Norwegian characters like \xc5 and \xd8 are not 7-bit ASCII'
```

Notice that Python echoes these back to us as escape sequences too. This is just a sequence of bytes, not a sequence of characters. If we want a sequence of Unicode code points we must decode the bytes into a text sequence of the str type, and for this we need to know the encoding. In this case I used Latin 1:

```
>>> norsk = b"Norwegian characters like \xc5 and \xd8 are not 7 bit ASCII"
>>> norsk.decode('latin1')
'Norwegian characters like Å and Ø are not 7 bit ASCII'
```

Indexing and slicing bytes

Notice that when we retrieve an item from the bytes object by indexing, we get an int object, not a one byte sequence:

```
>>> norsk[0]
78
>>> type(norsk[0])
<class 'int'>
```

This is another fundamental difference between bytes and str.

Slicing of bytes objects however *does* return a new bytes object:

```
>>> norsk[21:25]
b'like'
```

The bytes constructors

There are a few other forms of the bytes constructor it's good to be aware of. You can create a zero length bytes sequence simply by calling the constructor with no arguments:

```
>>> bytes()
```

You can create a zero-filled sequence of bytes by passing a single integer to the bytes constructor:

```
>>> bytes(5)
b'\x00\x00\x00\x00\x00'
```

You can also pass an iterable series of integers:

```
>>> bytes(range(65, 65+26))
b'ABCDEFGHIJKLMNOPQRSTUVWXYZ'
```

It's up to you to ensure that the values are non-negative and less than 256 to prevent a ValueError being raised:

```
>>> bytes([63, 127, 255, 511])
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
ValueError: bytes must be in range(0, 256)
```

One option if you need to construct a bytes object by encoding a Unicode str object is to use the two-argument form of the bytes constructor, which accepts a str in the first argument and and encoding for the second:

```
>>> bytes('Norwegian characters Å and Ø', 'utf16')
b'\xff\xfeN\x000\x00r\x00w\x00e\x00g\x00i\x00a\x00n\x00 \x00c\x00h\x00a\x00r\x00a\x00c\x0\
ot\x00e\x00r\x00s\x00 \x00\xc5\x00 \x00a\x00n\x00d\x00 \x00\xd8\x00'
```

Finally there is a class method fromhex() which is a factory method for creating a bytes object from a string consisting of concatenated two digit hexadecimal numbers:

```
>>> bytes.fromhex('54686520717569636b2062726f776e20666f78')
b'The quick brown fox'
```

There isn't a method to go in the other direction, so we have to use a generator expression to convert each byte to its hex representation, stripping the leading 0x from each resulting string using slicing:

```
>>> ''.join(hex(c)[2:] for c in b'The quick brown fox')
'54686520717569636b2062726f776e20666f78'
```

The mutable bytearray sequence

The bytes type we just looked at is only one of several so-called *binary* sequence types in Python. Whereas bytes is immutable, the bytearray type is

mutable. This means it supports many of the same operations as the other mutable sequence type with which you are already deeply familiar: list.

The bytearray constructors

The bytearray type supports the same constructors as bytes, so we won't go over them in detail here:

```
>>> bytearray()
bytearray(b'')
>>> bytearray(5)
bytearray(b'\x00\x00\x00\x00')
>>> bytearray(b'Construct from a sequence of bytes')
bytearray(b'Construct from a sequence of bytes')
>>> bytearray('Norwegian characters Å and Ø', 'utf16')
bytearray(b'\xff\xfeN\x000\x00r\x00w\x00e\x00g\x00i\x00a\x00n\x00 \x00c\x00h\x00a\x00r\x0\
0a\x00c\x00t\x00e\x00e\x00r\x00s\x00 \x00\xc5\x00 \x00a\x00n\x00d\x00 \x00\xf8\x00')
>>> bytearray.fromhex('54686520717569636b2062726f776e20666f78')
bytearray(b'The quick brown fox')
```

Being mutable we can use any of the mutable sequence operations to modify the bytearray, in place:

```
>>> pangram = bytearray(b'The quick brown fox')
>>> pangram.extend(b' jumps over the lazy dog')
>>> pangram
bytearray(b'The quick brown fox jumps over the lazy dog')
>>> pangram[40:43] = b'god'
>>> pangram
bytearray(b'The quick brown fox jumps over the lazy god')
```

String-like operations on bytearray

The bytearray type supports the same operations as list together with many of the string-like operations supported by bytes, such as upper(), split() and join():

```
>>> pangram.upper()
bytearray(b'THE QUICK BROWN FOX JUMPS OVER THE LAZY GOD')
>>>
>>> words = pangram.split()
>>> words
[bytearray(b'The'), bytearray(b'quick'), bytearray(b'brown'), bytearray(b'fox'), bytearra\y(b'jumps'), bytearray(b'over'), bytearray(b'the'), bytearray(b'lazy'), bytearray(b'god')\\]
>>> bytearray(b' ').join(words)
bytearray(b'The quick brown fox jumps over the lazy god')
```

When using string-like operations such as upper() or capitalize() on the bytes and bytearray types you must take particular care that only 7-bit ASCII

byte strings are in play.

Interpreting byte streams with the struct module

The Python Standard Library struct module is so called because it is capable of decoding the byte pattern of struct objects from the C and C++ languages. Structs — short for structures — are composite data types which, in binary form, consist of the byte patterns for C language primitives concatenated together. It may not be immediately obviously why this is necessary — or even useful — in a language like Python, but being very close to the machine level, C data types are something of a low-level *lingua franca* for binary data exchange between programs in many languages. Furthermore, as most operating systems are written in C or C++, the ability to interpret C structures into Python values and back again takes on even greater significance.

In this demo, we're going to:

- Write a binary file from a C program
- Read the file from a Python program
- Create Python classes which mirror C structures
- Demonstrate diagnostic techniques for dealing with binary data

Writing structs from C

This isn't a course on C, but we're sure you'll understand the following C structures for Vector, Color and Vertex in this program written in the C99 variant of the C language:

```
/**
  * colorpoints.c
  *
  * A C99 program to write a colored vertex
  * structures to a binary file.
  *
  */
#include <stdio.h>

struct Vector {
    float x;
    float y;
    float z;
};

struct Color {
    unsigned short int red;
    unsigned short int green;
    unsigned short int blue;
```

```
};
struct Vertex {
   struct Vector position;
    struct Color color;
};
int main(int argc, char** argv) {
    struct Vertex vertices[] = {
        \{ .position = { 3323.176, 6562.231, 9351.231 },
          .color = { 3040, 34423, 54321 } },
        { .position = { 7623.982, 2542.231, 9823.121 },
          .color = { 32736, 5342, 2321 } },
        { .position = { 6729.862, 2347.212, 3421.322 },
          .color = \{ 45263, 36291, 36701 \} \},
        { .position = { 6352.121, 3432.111, 9763.232 },
          .color = { 56222, 36612, 11214 } } };
    FILE* file = fopen("colors.bin", "wb");
    if (file == NULL) {
        return -1;
    fwrite(vertices, sizeof(struct Vertex), 4, file);
    fclose(file);
    return 0;
}
```

As you can see, we declare a Vector to be comprised of three float values. It's important to realise that a C float is actually a single-precision floating point value represented with 32 bits, which is different from a Python float, which is a double-precision floating point value represented with 64 bits.

We then declare a Color structure, which comprises three unsigned short integers. Again, this integer type is quite different from what we have available in Python, where we have arbitrary precision *signed* integers. In fact, having just 16 bits of precision, C's unsigned short int can only represent values from 0 through to 65535.

The third structure we declare is a Vertex, which combines a Vector and a Color together into one larger structure.

In the main() function, the program creates an array of four Vertex structures, and writes them to a file called colors.bin before exiting.

We'll now compile this C program into an executable. The details of how you do this are heavily system dependent, and require that you at least have access to a

C99 compiler. On our macOS system with the XCode development tools installed, we can simply use make from the command line:

```
$ make colorpoints
cc colorpoints.c -o colorpoints
```

This produces an executable called colorpoints:

```
$ ls
colorpoints colorpoints.c
```

When we run the executable, a colors.bin file is produced, as expected:

```
$ ./colorpoints
$ ls
colorpoints colorpoints.c colors.bin
```

Interpreting structs in Python

Now that we have a binary data file, let's try to make sense of it in Python. Our starting point will be a simple program called reader.py to read only the first vertex of data from the file:

```
import struct

def main():
    with open('colors.bin', 'rb') as f:
        buffer = f.read()
    items = struct.unpack_from('@fffHHH', buffer)
    print(repr(items))

if __name__ == '__main__':
    main()
```

In the main function, we open the file for read, being careful to remember to do so in binary mode. We then use the read() method of file objects to read the entire contents of the file in to a bytes object.

We use the struct.unpack_from() function to convert the raw byte sequence into more friendly types. This function accepts a special format string containing codes which specify how to interpret the bytes.

The leading @ character specifies that native byte order and alignment are to be used. Other characters can be used to force particular byte orderings, such as <

for little-endian, and > for big-endian. It's also possible to choose between native and no *alignment*, a topic we'll be revisiting shortly. If no byte-order character is specified, then @ is assumed.

There are also code letters corresponding to all of the common C data types, which are mostly variations on different precisions of signed and unsigned integers, together with 32- and 64-bit floating point numbers, byte arrays, pointers and fixed-length strings.

In our example, each of the three 'f' characters tells struct to expect a single-precision C float, and each of the 'H' characters tells struct to expect an unsigned short int, which is a 16-bit type.

Let's run our program:

```
$ python3 reader.py
(3323.176025390625, 6562.23095703125, 9351.2314453125, 3040, 34423, 54321)
```

We can see that struct.unpack_from returns a tuple. The reason our values don't look *exactly* the same as they did in the source code to our C program, is because we specified values in *decimal* in the source, and the values we chose are not representable exactly in binary. There has also been a conversion from the single-precision C float to the double-precision Python float, which is why the values we get back have so many more digits. Of course, the 16 bit unsigned short int values from C can be represented exactly as Python int objects.

TODO: We need to show the program output here!

Tuple unpacking

One obvious improvement to our program, given that unpack_from() returns a tuple, is to use tuple unpacking to place the values into named variables:

```
x, y, z, red, green, blue = struct.unpack_from('@fffHHH', buffer)
```

We can also shorten our format string slightly by using repeat counts. For example 3f means the same as fff:

```
x, y, z, red, green, blue = struct.unpack_from('@3f3H', buffer)
```

That's not a big win in this case, but it can be very useful for larger data series.

Reading all of the vertices

Finally, of course, we'd like to read all four Vertex structures from our file. We'll also make the example a bit more realistic by reading the data into Python classes which are equivalents of the Vector, Color, and Vertex structs we had in C. Here they are:

```
class Vector:
    def __init__(self, x, y, z):
        self.x = x
        self.y = y
        self.z = z
    def __repr__(self):
        return 'Vector({}, {}, {})'.format(self.x, self.y, self.z)
class Color:
    def __init__(self, red, green, blue):
        self.red = red
        self.green = green
        self.blue = blue
    def __repr__(self):
        return 'Color({}, {}, {})'.format(self.red, self.green, self.blue)
class Vertex:
    def __init__(self, vector, color):
        self.vector = vector
        self.color = color
    def __repr__(self):
        return 'Vertex({!r}, {!r})'.format(self.vector, self.color)
```

We'll also make a factory function to construct a type Vertex, which aggregates a Vector and a Color, being careful to use an argument order that is compatible with what we get back from the unpack function:

We'll add an import for pretty-printing at the top of the module:

```
Finally we'll re-work the main() function to use struct.iter_unpack():

def main():
    with open('colors.bin', 'rb') as f:
```

```
buffer = f.read()

vertices = []
for x, y, z, red, green, blue in struct.iter_unpack('@3f3H', buffer):
    vertex = make_colored_vertex(x, y, z, red, green, blue)
    vertices.append(vertex)

pp(vertices)
```

In fact, we can unwind one our earlier refactorings, and simply unpack the tuple directly into the arguments of make_colored_vertex() using extended call syntax:

```
def main():
    with open('colors.bin', 'rb') as f:
        buffer = f.read()

    vertices = []
    for fields in struct.iter_unpack('@3f3H', buffer):
        vertex = make_colored_vertex(*fields)
        vertices.append(vertex)

    pp(vertices)
```

In this code, fields will will be the tuple of three float and three int values returned for each structure. Rather than unpacking into named variables, we use extended call syntax to unpack the fields tuple directly into the arguments of make_colored_vertex(). When we've accumulated all vertices into a list, we pretty-print the resulting data structure. Let's try it!:

```
$ python3 reader.py
Traceback (most recent call last):
   File "examples/reader.py", line 59, in <module>
        main()
   File "examples/reader.py", line 52, in main
        for fields in struct.iter_unpack('@3f3H', buffer):
struct.error: iterative unpacking requires a bytes length multiple of 18
```

Oh dear! What happened?

Accounting for padding

The struct.iter_unpack() function is complaining because it expects the buffer byte sequence to be a multiple of 18 bytes long. Each of our structures consists of three 4-byte floats and three 2-byte unsigned short integers. We know that $3\times 4+3\times 2=18$. So how long is our buffer? Let's add some temporary diagnostic code to our program. After we read the file, we'll print the buffer length, and the buffer contents:

```
print("buffer: {} bytes".format(len(buffer)))
print(buffer)
```

When we run now, we get this output before the stack trace:

```
$ python3 reader.py
buffer: 80 bytes
b"\xd1\xb20E\xd9\x11\xcdE\xed\x1c\x12F\xe0\x0bw\x861\xd4\x00\x00\xdb?\xeeE\xb2\xe3\x1eE||\
\x19F\xe0\x7f\xde\x14\x11\t\x00\x00\xe5N\xd2Ed\xb3\x12E'\xd5UE\xcf\xb0\xc3\x8d]\x8f\x00\x\
00\xf8\x80\xc6E\xc7\x81VE\xee\x8c\x18F\x9e\xdb\x04\x8f\xce+\x00\x00"
Traceback (most recent call last):
    File "example/reader.py", line 56, in <module>
        main()
    File "example/reader.py", line 48, in main
        for fields in struct.iter_unpack('@3f3H', buffer):
struct.error: iterative unpacking requires a bytes length multiple of 18
```

Curiously, the buffer is reported as having 80 bytes, and checking at the command line, we can see that that is consistent with the file length:

```
$ ls -l
total 40
-rwxr-xr-x   1 rjs   staff   8732 11 Dec 12:32 colorpoints
-rw-r--r--   1 rjs   staff   1029 11 Dec 11:34 colorpoints.c
-rw-r--r--   1 rjs   staff   80 11 Dec 12:33 colors.bin
-rw-r--r--   1 rjs   staff   1068 11 Dec 17:39 reader.py
```

We're expecting $4\times 18=72$ bytes, so where are the extra eight bytes coming from?

Our diagnostic print statements above are a good start, but it's really awkward to read the standard bytes representation, especially when it contains a mix of ASCII characters and escape sequences. We can't directly convert a binary sequence into a readable hex string, but Python 3 has some tools in the standard library to help, in the form of the binascii module, which contains the oddly named hexlify() function. Let's import it:

```
from binascii import hexlify
and modify our diagnostic print statement to:
print(hexlify(buffer))
```

This gives us a long string, which is perhaps not even an improvement!

What we really want to do is to split pairs of digits with spaces. We can do this by slicing out consecutive pairs:

```
hex_buffer = hexlify(buffer).decode('ascii')
hex_pairs = ' '.join(hex_buffer[i:i+2] for i in range(0, len(hex_buffer), 2))
print(hex_pairs)
```

In this code, we hexlify then encode to an ASCII string. We then join() the successive two-digit slices with spaces, using a range() expression with a step of 2. We then print the hex pairs:

```
d1 b2 4f 45 d9 11 cd 45 ed 1c 12 46 e0 0b 77 86 31 d4 00 00 db 3f ee 45 b2 e3 1e 45 7c 7c\
19 46 e0 7f de 14 11 09 00 00 e5 4e d2 45 64 b3 12 45 27 d5 55 45 cf b0 c3 8d 5d 8f 00 0\
0 f8 80 c6 45 c7 81 56 45 ee 8c 18 46 9e db 04 8f ce 2b 00 00
```

This is a big improvement, but still leaves us counting bytes on the display. Let's precede our line of data with an integer count:

```
indexes = ' '.join(str(n).zfill(2) for n in range(len(buffer)))
print(indexes)
```

We generate integers using a range, convert then to strings and pad each number with leading zeros to a width of two; good enough for the first hundred bytes of our data.

Finally, we have something we can work with⁶:

```
buffer: 80 bytes
00 01 02 03 04 05 06 07 08 09 10 11 12 13 14 15 16 17 18 19
d1 b2 4f 45 d9 11 cd 45 ed 1c 12 46 e0 0b 77 86 31 d4 00 00

20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39
db 3f ee 45 b2 e3 1e 45 7c 7c 19 46 e0 7f de 14 11 09 00 00

40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59
e5 4e d2 45 64 b3 12 45 27 d5 55 45 cf b0 c3 8d 5d 8f 00 00

60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79
f8 80 c6 45 c7 81 56 45 ee 8c 18 46 9e db 04 8f ce 2b 00 00
```

Now we've got a useful way of viewing our bytes object, let's get back to diagnosing our problem of why we have 80 bytes rather than 72. Looking carefully, we can see that the first bytes at indices 0 to 17 inclusive contain legitimate data — and we know this to be the case because we decoded it earlier. Looking at bytes 18 and 19 though, we see two zero bytes. From bytes 20 to 37, we have another run of what looks like legitimate data, again followed by another two zero bytes at indices 38 and 39.

This pattern continues to the end of the file. What we're seeing is 'padding' added by the C compiler to align structures on four-byte boundaries. Our 18 byte structure needs to be padded with two bytes to take it to 20 bytes which is divisible by four. In order to skip this padding we can use 'x' which is the format code for pad bytes. In this case there are two pad bytes per structure, so we can add 'xx' to our format string:

```
vertices = []
for fields in struct.iter_unpack('@3f2Hxx', buffer):
    vertex = make_colored_vertex(*fields)
    vertices.append(vertex)
```

With this change in place, we can successfully read our structures from C into Python!:

```
buffer: 80 bytes
00 01 02 03 04 05 06 07 08 09 10 11 12 13 14 15 16 17 18 19
d1 b2 4f 45 d9 11 cd 45 ed 1c 12 46 e0 0b 77 86 31 d4 00 00
20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39
db 3f ee 45 b2 e3 1e 45 7c 7c 19 46 e0 7f de 14 11 09 00 00
40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59
e5 4e d2 45 64 b3 12 45 27 d5 55 45 cf b0 c3 8d 5d 8f 00 00
60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79
f8 80 c6 45 c7 81 56 45 ee 8c 18 46 9e db 04 8f ce 2b 00 00
[Vertex(Vector(3323.176025390625, 6562.23095703125, 9351.2314453125),
        Color(3040, 34423, 54321)),
Vertex(Vector(7623.98193359375, 2542.23095703125, 9823.12109375),
        Color(32736, 5342, 2321)),
Vertex(Vector(6729.86181640625, 2347.2119140625, 3421.322021484375),
        Color(45263, 36291, 36701)),
Vertex(Vector(6352.12109375, 3432.111083984375, 9763.232421875),
        Color(56222, 36612, 11214))]
```

Here's the complete code so far:

```
from binascii import hexlify
from pprint import pprint as pp
import struct

class Vector:

    def __init__(self, x, y, z):
        self.x = x
        self.y = y
        self.z = z

    def __repr__(self):
        return 'Vector({}, {}, {})'.format(self.x, self.y, self.z)

class Color:
    def __init__(self, red, green, blue):
```

```
self.red = red
        self.green = green
       self.blue = blue
   def __repr__(self):
        return 'Color({}, {}, {})'.format(self.red, self.green, self.blue)
class Vertex:
   def __init__(self, vector, color):
        self.vector = vector
       self.color = color
   def __repr__(self):
        return 'Vertex({!r}, {!r})'.format(self.vector, self.color)
def make_colored_vertex(x, y, z, red, green, blue):
   return Vertex(Vector(x, y, z),
                  Color(red, green, blue))
def main():
   with open('colors.bin', 'rb') as f:
       buffer = f.read()
   print("buffer: {} bytes".format(len(buffer)))
   indexes = ' '.join(str(n).zfill(2) for n in range(len(buffer)))
   print(indexes)
   hex_buffer = hexlify(buffer).decode('ascii')
   hex_pairs = ' '.join(hex_buffer[i:i+2] for i in range(0, len(hex_buffer), 2))
   print(hex_pairs)
   vertices = []
   for fields in struct.iter_unpack('@3f3Hxx', buffer):
        vertex = make_colored_vertex(*fields)
       vertices.append(vertex)
   pp(vertices)
if __name__ == '__main__':
   main()
```

From this point onwards, we'll no longer need the diagnostic print statements which helped us get this far, so consider them removed from this point onwards.

Understanding "native" in context

You might be wondering why, given that we used the @ character at the beginning of our format string to specify native byte-order, native size, and native alignment, this didn't work out of the box.

So did we!

Eventually, we traced this mismatch to the fact that our Python interpreter — which is itself implemented in C — and our little C program for writing vertices to a file, were compiled using different C compilers with different structure padding conventions. This just goes to show that when dealing with binary data you need to be *very* careful if you want your programs to be portable between systems and compilers.

Memory Views

Python has a built-in type called memoryview which wraps any existing, underlying collection of bytes and which supports something called the *buffer protocol*. The buffer protocol is implemented at the C level inside the Python interpreter, and isn't a protocol in the same sense that we use the word when talking about the Python-level *sequence* and *mapping* protocols. In fact, the memoryview type *implements* the Python-level *sequence* protocol, allowing us to view the underlying byte buffer as a sequence of Python objects.

Our previous example required that we read the data from the file into a byte array, and translate it with struct.unpack() into a tuple of numeric objects, effectively duplicating data. We're going to change that example now to use memoryviews, avoiding the duplication.

We can construct memoryview instances by passing any object that supports the buffer protocol C API to the constructor. The only built-in types which support the buffer protocol are bytes and bytearray. We'll construct a memory view from the buffer just after our diagnostic print statements, with this line of code:

```
mem = memoryview(buffer)
```

Exploring code at runtime with code.interact

To explore the capabilities of memoryview we could use a debugger such as PDB to stop the program just after this line, but we'll introduce you to another technique from the Python Standard Library code module: code.interact(). This function will suspend the program and drop us to the REPL. By passing a reference to the local namespace we can get access to the 'live' variables in our program, including our memoryview. With these two changes, our main() function now looks like this:

```
def main():
    with open('colors.bin', 'rb') as f:
        buffer = f.read()
```

```
mem = memoryview(buffer)
code.interact(local=locals()) # Temporary

vertices = []
for fields in struct.iter_unpack('@3f3Hxx', buffer):
    vertex = make_colored_vertex(*fields)
    vertices.append(vertex)

pp(vertices)
```

We use a call to the locals() built-in function to get a reference to the current namespace. Now, when we run our program, we get a REPL prompt at which we can access our new mem object:

```
$ python3 reader.py
>>> mem
<memory at 0x10214e5c0>
```

The memoryview object supports indexing, so retrieving the byte at index 21 and converted to hexadecimal gives us 3f, as we might expect:

```
>>> hex(mem[21])
```

Being bona-fide sequences, memoryview objects support slicing:

```
>>> mem[12:18]
<memory at 0x10214e750>
```

Crucially, memoryview slices are *also* memoryviews. There's no copying going on here! The zeroth byte of the slice has value e0:

```
>>> hex(mem[12:18][0])
```

Casting memoryview elements

Much in the same way we used the struct module to interpret binary data as native types, we can use the memoryview.cast() method to do something equivalent. This method accepts uses the same format codes as used by the struct module, except only a single element code is permitted. In other words the interpreted items must all be of the same type.

We know that the bytes in the [12:18] slice represent three unsigned short int values, so by passing 'H' to the cast() method we can interpret the values that way:

```
>>> mem[12:18].cast('H')
<memory at 0x10214e688>
```

Notice that this *also* returns a memoryview, but this time one that knows the type of its elements:

```
>>> mem[12:18].cast('H')[0]
3040
>>> mem[12:18].cast('H')[1]
34423
>>> mem[12:18].cast('H')[2]
54321
```

Alternatively, we can use the memoryview.tolist() method to convert the series of elements to a list:

```
>>> mem[12:18].cast('H').tolist()
[3040, 34423, 54321]
```

When you've finished with the interactive session, you can send an end-of-file character just as you normally would to terminate a REPL session with Ctrl-D on Unix or Ctrl-Z on Windows; your program will then continue executing from the first statement after code.interact().

There are other interesting features of memoryviews such as the ability to interpret multidimensional C arrays. If you need those facilities we recommend you consult the documentation.

Before moving on, remove the code.interact() line, so our program runs uninterrupted again.

Python classes built on memoryviews

Let's use the ability to slice and cast memoryviews to modify our Vector and Color types to use the bytes buffer as the underlying storage.

Our modified vector class now looks like this:

```
class Vector:
```

```
raise TypeError("Vector: memoryview must contain at least 3 floats")
self._mem = mem_float32

@property
def x(self):
    return self._mem[0]

@property
def y(self):
    return self._mem[1]

@property
def z(self):
    return self._mem[2]

def __repr__(self):
    return 'Vector({}, {}, {})'.format(self.x, self.y, self.z)
```

The initializer accepts a memoryview which we expect to expose floats. We validate this by checking the format code returned by the memoryview.format attribute against a string containing 'f' and 'd', those codes for single- and double-precision floats respectively. We also check that the memory view exposes at least three items; note than when used with a memoryview len() returns the number of *items* not the number of *bytes*. The only instance attribute now holds a reference to the memoryview.

Our old instance attributes are replaced by properies which perform the appropriate lookups in the memoryview. Since we can use properties to replace attributes, our __repr__() implementation can remain unmodified.

The modified Color class works exactly the same way, except now we check that we're wrapping unsigned integer types:

```
class Color:
```

```
def __init__(self, mem_uint16):
    if mem_uint16.format not in "HILQ":
        raise TypeError("Color: memoryview values must be unsigned integers")
    if len(mem_uint16) < 3:
        raise TypeError("Color: memoryview must contain at least 3 integers")
    self._mem = mem_uint16

@property
def red(self):
    return self._mem[0]

@property
def green(self):
    return self._mem[1]

@property
def blue(self):
    return self._mem[2]</pre>
```

```
def __repr__(self):
    return 'Color({}, {}, {})'.format(self.red, self.green, self.blue)
```

Our Vertex class, which simply combines a Vector and Color, can remain as before, although our make_colored_vertex() factory function needs to be changed to accept a memoryview — specifically one that is aligned with the beginning of a Vertex structure:

The function now slices the vertex memoryview into two parts, for the vector and color respectively, and casts each to a typed memoryview. These are used to construct the Vector and Color objects, which are then passed on to the Vertex constructor.

Back in our main function after our creation of the mem instance, we'll need to rework our main loop. We'll start by declaring a couple of constants describing the size of a Vertex structure, and the stride between successive Vertex structures, this allows us to take account of the two padding bytes between structures:

```
VERTEX_SIZE = 18
VERTEX_STRIDE = VERTEX_SIZE + 2
```

Next we'll set up a generator expression which yields successive memoryviews into whole Vertex structures:

```
vertex_mems = (mem[i:i + VERTEX_SIZE] for i in range(0, len(mem), VERTEX_STRIDE))
```

Remember that each slice into mem is itself a memoryview.

This time, rather than an explicit for-loop to build the list of vertices, we'll use a list comprehension to pass each vertex memory view in turn to make_colored_vertex():

```
vertices = [make_colored_vertex(vertex_mem) for vertex_mem in vertex_mems]
```

Running this program, we can see we get exactly the same results as before, except that now our Vector and Color objects are backed by the binary data we loaded from the file, with much reduced copying.

Here's the complete program as it now stands:

```
from binascii import hexlify
from pprint import pprint as pp
import struct
class Vector:
   def __init__(self, mem_float32):
       if mem_float32.format not in "fd":
           if len(mem_float32) < 3:</pre>
           raise TypeError("Vector: memoryview must contain at least 3 floats")
       self._mem = mem_float32
   @property
   def x(self):
       return self._mem[0]
   @property
   def y(self):
       return self._mem[1]
   @property
   def z(self):
       return self._mem[2]
   def __repr__(self):
       return 'Vector({}, {}, {})'.format(self.x, self.y, self.z)
class Color:
   def __init__(self, mem_uint16):
       if mem_uint16.format not in "HILQ":
           raise TypeError("Color: memoryview values must be unsigned integers")
       if len(mem_uint16) < 3:</pre>
           raise TypeError("Color: memoryview must contain at least 3 integers")
       self._mem = mem_uint16
   @property
   def red(self):
       return self._mem[0]
   @property
   def green(self):
       return self._mem[1]
   @property
   def blue(self):
       return self._mem[2]
   def __repr__(self):
       return 'Color({}, {}, {})'.format(self.red, self.green, self.blue)
class Vertex:
   def __init__(self, vector, color):
       self.vector = vector
       self.color = color
```

```
def __repr__(self):
       return 'Vertex({!r}, {!r})'.format(self.vector, self.color)
def make_colored_vertex(mem_vertex):
   mem_vector = mem_vertex[0:12].cast('f')
   mem_color = mem_vertex[12:18].cast('H')
   return Vertex(Vector(mem_vector),
                  Color(mem_color))
VERTEX_SIZE = 18
VERTEX_STRIDE = VERTEX_SIZE + 2
def main():
   with open('colors.bin', 'rb') as f:
       buffer = f.read()
   mem = memoryview(buffer)
   VERTEX SIZE = 18
   VERTEX STRIDE = VERTEX SIZE + 2
   vertex_mems = (mem[i:i + VERTEX_SIZE]
                   for i in range(0, len(mem), VERTEX_STRIDE))
   vertices = [make_colored_vertex(vertex_mem)
                for vertex mem in vertex mems]
   pp(vertices)
if __name__ == '__main__':
   main()
```

Memory-mapped files

There's still one copy happening though: the transfer of bytes from the file into our buffer bytes object. This is not a problem for our trifling 80 bytes, but for very large files this could be prohibitive.

By using an operating system feature called *memory-mapped files* we can use the virtual memory system to make large files *appear* as if they are in memory. Behind the scenes the operating system will load, discard and sync pages of data from the file. The details are operating system dependent, but the pages are typically only 4 kilobytes in size, so this can be memory efficient if you need access to relatively small parts of large files.

Memory-mapped file support in Python is provided by the standard library mmap module. This module contains a single class, also called mmap, which behaves like a mix between a bytearray and a file-like object. The mmap instance is created around a file-handle on Windows or a file-descriptor on Unix, either of which can be obtained by calling the fileno() method of a regular file object.

In fact, mmap instances support the C API *buffer protocol* and so can be used as the argument we pass to our memoryview constructor.

Using mmap in our implementation

Let's modify our example to do so. All this requires is that we import the mmap module at the top of our file:

```
import mmap
```

We then need to modify our main() function to retrieve the file handle or descriptor, passing it to the mmap class constructor. Like other file-like objects, mmaps must be closed when we're done with them. We can either call the close() method explicity or, more conveniently, use the mmap object as a context manager. We'll go with the latter. Here's the resulting main() method:

The only difference here is that buffer is now a memory-mapped file rather than bytes sequence as it was previously. We've avoided reading the file into memory twice — once into the operating system file cache and once more into our own collection — by directly working on the operating system's view of the file.

Dangling memory references

This "works" after a fashion when we run it, insofar as our Vertex objects are created with the memory-mapped file backing store. However, we get a nasty failure when the mmap object is closed by the context manager, which tells us that "exported pointers exist":

```
Traceback (most recent call last):
    File "examples/reader.py", line 108, in <module>
        main()
    File "examples/reader.py", line 100, in main
        pp(vertices)
    BufferError: cannot close exported pointers exist
```

The cause is that at the point the mmap object is closed we still have a chain of extant memoryview objects which ultimately depend on the mmap. A reference counting mechanism in the buffer-protocol has tracked this, and knows that the mmap still has memoryview instances pointing to it.

There are a couple of approaches we could take here. We could arrange for the memoryview.release() method to be called on the memoryview objects inside our Vector and Color instances. This method deregisters the memoryview from any underlying buffers and invalidates the memoryview so any further operations with it raise a ValueError. This would just move the problem though, now we'd have zombie Vector and Color instances containing invalid memoryviews. Better, we think, to respect the constraint in our design that the lifetime of our memory-mapped file-backed objects must be shorter than the lifetime of our memory mapping.

Thinking about our live object-graph, there are two local variables which ultimately hold references to the memory-mapped file: mem which is our lowest level memoryview, and vertices which is the list of Vertex objects.

By explicitly removing these name bindings, using two invocations of the del statement, we can clean up the memoryviews, so the memory map can be torn down safely:

```
del mem
del vertices
```

With this change in place, the main function looks like this:

```
for vertex_mem in vertex_mems]
del vertices
del mem
```

And our program runs flawlessly, with minimal memory usage, even for huge sets of vertex objects:

Summary

Let's summarize what we've covered in this chapter:

- Our review of Python's bitwise operators including exlusive-or and complement led us down the path of understanding how Python internally represents and externally *presents* arbitrary precision integer values at the bit level.
- To print bit patterns for values which equate to negative numbers you can sign-extend the result by using bitwise-and with a binary value consisting of consecutive 1 bits.
- We reviewed the constructors and some key methods of the *immutable* binary sequence type bytes and the *mutable* binary sequence type bytearray.
- We showed how to interpret a series of bytes as native or C language types such as short unsigned integer, translating them into compatible Python types, using the struct module.
- We demonstrated how to display byte streams in an easily readable hexadecimal format using the hexlify module in conjuction with some simple formatting expressions.
- We explained that a detailed understanding of how C compilers align data structures on certain byte or word boundaries can be crucial to correctly reading data file produced by C or C++ programs.
- We introduced memoryview for obtaining copy-free views, slices, and type casts from underlying byte sequences.
- We touched on the use of the interact() function from the standard library code module, which allows us to drop to the REPL at any point in a

- program, and resume execution later.
- We showed that memory-mapped files implemented with the mmap module implement the buffer-protocol, and so can be used as the basis for memoryview objects, further reducing the amount of data copying when reading files.

There's a lot more to be learned about low-level byte-oriented programming in Python, including writeable memoryviews, shared memory with mmap, and interfacing to native C and C++ code. We can't possibly cover all of that in *this* book, but we've shown you the starting points for your own explorations.

Chapter 3 - Object Internals and Custom Attributes

In this chapter we'll show how objects are represented internally in Python, as a dictionary called __dict__. We'll also demonstrate how to directly query and manipulate objects through the internal __dict__. We'll then use this knowledge to implement custom attribute access, by overriding the special __getattr__(), __getattrribute__(), __setattr__(), and __delattr__() methods. We'll round off by seeing how to make objects more memory efficient by defining __slots__.

How are Python objects stored?

We'll start with a simple class to represent two-dimensional vectors, called simply Vector:

Let's instantiate an object of that class in the REPL:

In the list returned by dir() we see the two named attributes x and y along with many of Python's special attributes, quite a few of which we've explained previously in *The Python Apprentice* and *The Python Journeyman*. One attribute in particular is of interest to us today, and that is __dict__. Let's see what it is:

```
>>> v.__dict__
{'x': 5, 'y': 3}
```

As its name indicates, __dict__ is indeed a dictionary, one which contains the names of our object's attributes as keys, and the values of our object's attributes as, well, values. Here's further proof, if any were needed, that __dict__ is a Python dictionary:

```
>>> type(v.__dict__)
<class 'dict'>
```

We can retrieve attributes values directly from __dict__:

```
>>> v.__dict__['x']
```

Modify values:

```
>>> v.__dict__['x'] = 17
>>> v.x
17
```

and even remove them!:

```
>>> del v.__dict__['x']
>>> v.x
Traceback (most recent call last):
   File "<input>", line 1, in <module>
AttributeError: 'Vector' object has no attribute 'x'
```

As you might by now suspect, we can test for their existence:

```
>>> 'x' in v.__dict__
False
>>> 'y' in v.__dict__
```

and insert new attributes into the dictionary:

```
>>> v.__dict__['z'] = 13
>>> v.z
13
```

Although all of these direct queries and manipulations of __dict__ are possible, for the most part you should prefer to use the built-in functions getattr(), hasattr(), delattr() and setattr():

```
>>> getattr(v, 'y')
```

```
>>> hasattr(v, 'x')
False
>>> delattr(v, 'z')
>>> setattr(v, 'x', 9)
```

Direct access to __dict__ does have legitimate uses though, so it's essential to be aware of its existence, and how and when to use it for advanced Python programming.

Our Vector class, like most vector classes, has hardwired attributes called \times and y to store the two components of the vector.

Many problems though, require us to deal with vectors in different coordinate systems within the same code. Or perhaps it's just convenient to use a different labelling scheme, such a u and v instead of x and y in a particular context. Let's see what we can come up with:

```
class Vector:
```

In this code, we accept a keyword argument, which is received into the coords dictionary, the contents of which we use to update the entries in __dict__. Remember that dictionaries are unordered⁸, so there's no way to ensure that the coordinates are stored in the order they are specified. Our __repr__() implementation must iterate over the dictionary, sorting by key, for convenience.

This allows us to provide arbitrary coordinate names:

This is all very well, but our coordinates are now essentially "public" attributes of our vector objects. What if we want our Vector class to be an immutable value type, so values provided to the constructor can't be subsequently changed? Ordinarily, we would do this by prefixing our attributes with an underscore to signal that they are implementation details, and then provide a property with only a getter, to prevent modification. In this case though, we don't know the attribute names in advance, so we can't declare a property getter which must be named at *class* definition time, not at *object* instantiation time. We'll show how to work around this using the special __getattr__ method, but first, let's change our __init__() method to store data in "private" attributes, and our __repr__() to report them correctly:

class Vector:

We can construct and represent instances as before:

```
>>> v = Vector(p=9, q=3)
>>> v
Vector(p=9, q=3)
```

But now the attributes are stored in "private" attributes called _p and _q:

So we can no longer access p directly, because it doesn't exist:

```
>>> v.p
Traceback (most recent call last):
   File "<input>", line 1, in <module>
AttributeError: 'Vector' object has no attribute 'p'
```

What we'd like to do is to fake the existence of p for read access, and to do that, we need to intercept attribute access before the AttributeError is raised. To do that, we can override __getattr__(). Notice that there are two very similarly named special methods: __getattr__() and __getattribute__(). The former is only called when regular attribute lookup fails. The latter is called for *all* attribute access irrespective of whether an attribute of the requested name exists or not. For our case, we'll be using __getattr__() to intercept lookup failures.

Here's our Vector definition with __getattr__() added. We've just added a simple stub that prints the attribute name:

class Vector:

When we request non-existent attributes, the name of the attribute is now printed:

```
>>> v = Vector(p=3, q=9)
>>> v.p
name = p
>>> v.q
name = q
```

But when we request an attribute that exists, we simply get the attribute value, indicating that __getattr__ isn't being called:

```
>>> v._q
```

Now we can modify __getattr__() to prepend the underscore to name, and retrieve the attribute for us:

class Vector:

```
def __init__(self, **coords):
    private_coords = {'_' + k: v for k, v in coords.items()}
    self.__dict__.update(private_coords)
```

At first sight, this appears to work just fine:

```
>>> from vector_05 import *
>>> v = Vector(p=5, q=10)
>>> v.p
5
>>> v.q
```

but there are some serious problems lurking here. The first, is that we can still *assign* to p and q:

```
>>> v.p = 13
```

Remember, there wasn't really an attribute called p, but Python has no qualms about creating it for us on demand. Worse, because we have unwittingly brought p into existence, __getattr__() is not longer invoked for requests of p, even though our hidden attribute _p is still there behind the scenes, with a different value:

```
>>> v.p
13
>>> v._p
```

We can prevent this, by overriding __setattr__() too, to intercept attempts to write to our attributes:

class Vector:

```
def __init__(self, **coords):
    private_coords = {'_' + k: v for k, v in coords.items()}
    self.__dict__.update(private_coords)

def __getattr__(self, name):
    private_name = '_' + name
    return getattr(self, private_name)

def __setattr__(self, name, value):
    raise AttributeError("Can't set attribute {!r}".format(name))

def __repr__(self):
    return "{}({})".format(
```

```
type(self).__name__,
', '.join("{k}={v}".format(k=k[1:], v=self.__dict__[k])
    for k in sorted(self.__dict__.keys())))
```

This successfully prevents writing to any attribute:

```
>>> v = Vector(p=4, q=8)
>>> v.p = 7
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "examples/vector/vector.py", line 16, in __setattr__
        raise AttributeError("Can't set attribute {!r}".format(name))
AttributeError: Can't set attribute 'p'
```

We'll return to __setattr__ shortly, but first, we need to demonstrate another problem with our __getattr__(). Although __getattr__() works fine with attributes we've carefully faked:

```
>>> v.p
```

Look what happens when we try to access an attribute for which there is no faked support:

```
>>> v.x
Traceback (most recent call last):
    File "<input>", line 1, in <module>
    File "examples/vector/vector.py", line 13, in __getattr__
        return getattr(self, private_name)
File "examples/vector/vector.py", line 13, in __getattr__
        return getattr(self, private_name)
File "examples/vector/vector.py", line 13, in __getattr__
        return getattr(self, private_name)
...
File "examples/vector/vector.py", line 13, in __getattr__
        return getattr(self, private_name)
RuntimeError: maximum recursion depth exceeded while calling a Python object
```

This happens because our request for attribute x causes __getattr__() to look for an attribute _x, which doesn't exist, which invokes __getattr__() again, to lookup attribute __x, which doesn't exist. And so on recursively, until the Python interpreter exceeds its maximum recursion depth and raises a RuntimeError.

To prevent this happening, you might be tempted to check for existence of the private attribute using hasattr(), like this:

```
class Vector:
```

```
def __init__(self, **coords):
    private_coords = {'_' + k: v for k, v in coords.items()}
    self.__dict__.update(private_coords)
```

Unfortunately, this doesn't work either, since it turns out that hasattr() also ultimately calls __getattr__() in search of the attribute! What we need to do, is directly check for the presence of our attribute in __dict__:

class Vector:

This now works as we would wish:

```
>>> v = Vector(p=9, q=14)
>>> v
Vector(p=9, q=14)
>>> v.p
9
>>> v.x
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "examples/vector/vector.py", line 14, in __getattr__
        raise AttributeError('{!r} object has no attribute {!r}'.format(self.__class__, name))
AttributeError: <class 'Vector'> object has no attribute 'x'
```

In fact, attribute lookup in Python follows a pretty complex procedure, so instead of invoking that procedure again by calling getattr(), we can just directly return the attribute value from __dict__. This also enables us to switch to easier-to-ask-for-forgiveness-than permission (EAFP) rather than look-before-you-leap (LBYL) style:

```
class Vector:
    def __init__(self, **coords):
        private_coords = {'_' + k: v for k, v in coords.items()}
        self.__dict__.update(private_coords)
    def __getattr__(self, name):
        private_name = '_' + name
            return self.__dict__[private_name]
        except KeyError:
            raise AttributeError('{!r} object has no attribute {!r}'.format(
                type(self).__name__, name))
    def __setattr__(self, name, value):
        raise AttributeError("Can't set attribute {!r}".format(name))
    def __repr__(self):
        return "{}({})".format(
            type(self).__name_
               e(ser).__name__,
'.join("{k}={v}".format(k=k[1:], v=self.__dict__[k])
                for k in sorted(self.__dict__.keys())))
Giving:
>>> v = Vector(p=1, q=2)
Vector(p=1, q=2)
>>> v.p
>>> v.x
Traceback (most recent call last):
  File "<input>", line 1, in <module>
  File "examples/vector/vector.py", line 16, in __getattr_
    raise AttributeError('{!r} object has no attribute {!r}'.format(type(self).__name___, \
```

Of course, the EAFP version has the same behaviour as the LBYL version.

Customizing attribute deletion

AttributeError: 'Vector' object has no attribute 'x'

Deleting attributes is something that is very rarely seen in Python, although it is possible. Our Vector class also allows us to remove attributes by calling by calling delattr():

```
>>> v
Vector(p=1, q=2)
```

```
>>> delattr(v, '_p')

Or even:

>>> del v._q
>>> v
Vector()
```

Although a client is unlikely to attempt this inadvertently, we can prevent it by overriding __delattr__():

```
class Vector:
    def __init__(self, **coords):
        private_coords = {'_' + k: v for k, v in coords.items()}
        self.__dict__.update(private_coords)
    def __getattr__(self, name):
        private_name = '_' + name
            return self.__dict__[private_name]
        except KeyError:
            raise AttributeError('{!r} object has no attribute {!r}'.format(
                type(self).__name__, name))
    def __setattr__(self, name, value):
        raise AttributeError("Can't set attribute {!r}".format(name))
    def __delattr__(self, name):
        raise AttributeError("Can't delete attribute {!r}".format(name))
    def __repr__(self):
        return "{}({})".format(
            type(self).__name__,
', '.join("{k}={v}".format(k=k[1:], v=self.__dict__[k])
                for k in sorted(self.__dict__.keys())))
```

Here's how that code behaves when trying do delete either public or private attributes:

```
>>> v = Vector(p=9, q=12)
>>> v
Vector(p=9, q=12)
>>> del v.q
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
   File "example/vector/vector.py", line 19, in __delattr__
        raise AttributeError("Can't delete attribute {!r}".format(name))
AttributeError: Can't delete attribute 'q'
>>> del v._q
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
   File "example/vector/vector.py", line 19, in __delattr__
```

```
raise AttributeError("Can't delete attribute {!r}".format(name))
AttributeError: Can't delete attribute '_q'
```

Customising attribute storage

There's no requirement for us to store attributes directly in __dict__. Here's an example of a subclass of Vector that stores a *mutable* red-green-blue color along with the immutable Vector components:

```
class ColoredVector(Vector):
   COLOR_INDEXES = ('red', 'green', 'blue')
   def __init__(self, red, green, blue, **coords):
        super().__init__(**coords)
        self.__dict__['color'] = [red, green, blue]
   def __getattr__(self, name):
           channel = ColoredVector.COLOR_INDEXES.index(name)
       except ValueError:
           return super().__getattr__(name)
            return self.__dict__['color'][channel]
   def __setattr__(self, name, value):
        try:
           channel = ColoredVector.COLOR_INDEXES.index(name)
        except ValueError:
            super().__setattr__(name, value)
            self.__dict__['color'][channel] = value
```

Internally, we store the red-green-blue channels in a list. We override both __getattr__() and __setattr__() to provide read and write access to the color channels, being careful to forward requests to the superclass when necessary. So long as we are careful to only ever access our color attribute by accessing __dict__ directly, everything seems to work well:

```
'__subclasshook__', '__weakref__', '__p', '__q', 'color']
>>> cv.red = 50
>>> cv.red
50
>>> cv.p = 12
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "examples/vector/vector.py", line 53, in __setattr__
        super().__setattr__(name, value)
   File "examples/vector/vector.py", line 20, in __setattr__
        raise AttributeError("Can't set attribute {!r}".format(name))
AttributeError: Can't set attribute 'p'
```

There's a gremlin lurking here though: Our __repr__() implementation in the base class makes an assumption which is no longer valid. It assumes all attributes are prefixed with an underscore internally, and it doesn't know about color:

```
>>> cv
ColoredVector(p=9, q=14, olor=[50, 44, 238])
```

This is surprisingly hard to fix elegantly, without the derived ColoredVector class knowing too much about implementation details of the Vector base class; we believe it should be possible to derive from a class without knowing how it works. The base class <code>__repr__()</code> makes assumptions about the contents of its <code>__dict__</code> which it cannot reasonably expect to be respected by subclasses. As an exercise, we recommend changing Vector so it stores its components in a dedicated dictionary separate from <code>__dict__</code>, although of course, this dictionary itself will need to be stored in <code>__dict__</code>. Here is a fix which "works" by duplicating and modifying some of the logic in the base class:

```
def __repr__(self):
    keys = set(self.__dict__.keys())
    keys.discard('color')
    coords = ', '.join(
        "{k}={v}".format(k=k[1:], v=self.__dict__[k])
        for k in sorted(keys))

return "{cls}(red={red}, green={green}, blue={blue}, {coords})".format(
        cls=type(self).__name__,
        red=self.red,
        green=self.green,
        blue=self.blue,
        coords=coords)
```

This method override works by removing the attribute name 'color' from the list of keys before using the same logic as the superclass to produce the string of sorted co-ordinates. The color channel values are accessed in the normal way, which will invoke __getattr__().

It's worth bearing in mind that this example demonstrates the awkwardness of inheriting from classes which were not deliberately designed as base classes. Our code serves it's purpose in demonstrating customised attribute access, but we couldn't recommend such use of inheritance in production code.

Using vars() to access __dict__

Do you remember the vars() built-in function? Without an argument it returns a dictionary containing the current namespace, as so acts just like locals(). However, if we supply an argument it returns the __dict__ attribute of the argument, so instead of writing:

```
obj.__dict__
```

we can write:

```
vars(obj)
```

Arguably, this is more *Pythonic* than accessing __dict__ directly, for much the same reason that calling:

```
len(collection)
```

is definitely more idiomatic than calling:

```
collection.__len__()
```

That said, the length returned by len() is always immutable, whereas __dict__ is a mutable dictionary. In our opinion it us much clearer that the internal state of an object is being modified when we directly modify the __dict__ attribute, like this:

```
self.__dict__['color'] = [red, green, blue]
than we we go via vars():
vars(self)['color'] = [red, green, blue]
```

Whichever you use - and we don't feel strongly either way - you should be aware of this use of vars() with an argument.

Overriding __getattribute__()

Recall that __getattr__() is called only in cases when "normal attribute lookup fails"; it is our last chance for us to intervene before the Python runtime raises an AttributeError. But what if we want to intercept all attribute access? In that case, we can override __getattribute__(). I use the term "override" advisedly, because it is the implementation of __getattribute__() in the ultimate base class object that is responsible for the normal lookup behaviour, including calling __getattr__(). This level of control is seldom required, and you should always consider whether __getattr__() is sufficient for your needs. That said, __getattribute__() does have its uses.

Consider this class, which implements a LoggingProxy which logs every attribute retrieval made against the target object, supplied to the constructor:

class LoggingProxy:

Since __getattribute__() intercepts all attribute attribute access through the 'dot' operator, we must be very careful to never access attributes of the LoggingProxy through the dot. In the initialiser we use the __setattr__() implementation inherited from the object superclass to do our work. Inside __getattribute__() itself, we retrieve the attribute called "target" using a call to the superclass implementation of __getattribute__(), which implements the default lookup behaviour. Once we have a reference to the target class, we delegate to the getattr() built-in function. If attribute lookup on the target fails, we then raise an AttributeError with an informative message. Note how careful we must be even to return our object's __class__!

If attribute retrieval was successful, we report as much before returning the attribute value.

Let's see it in action. If you're following along at the keyboard, we recommend using the regular Python REPL in a terminal, rather than using the more advanced REPLs such as ipython or the PyCharm REPL, because these advanced REPLs performs a great many invocations of __getattribute__() themselves, to support automatic code-completion:

```
>>> cv = ColoredVector(red=23, green=44, blue=238, p=9, q=14)
>>> cw = LoggingProxy(cv)
>>> cw.p
Retrieved attribute 'p' = 9 from ColoredVector(p=9, q=14, olor=[23, 44, 238])
9
>>> cw.red
Retrieved attribute 'red' = 23 from ColoredVector(p=9, q=14, olor=[23, 44, 238])
23
```

So far, so good. But what happens when we *write* to an attribute through the proxy? In this example both writes *appear* to be accepted without error, although only one of them should be:

```
>>> cw.p = 19
>>> cw.red = 5
```

Although in fact, neither of the writes succeeded:

```
>>> cw.p
Retrieved attribute 'p' = 9 from ColoredVector(p=9, q=14, olor=[23, 44, 238])
9
>>> cw.red
Retrieved attribute 'red' = 23 from ColoredVector(p=9, q=14, olor=[23, 44, 238])
23
```

What's happening here is that our attribute *writes* to the cw proxy are invoking __setattr__() on the object base class, which is actually creating new attributes in the LoggingProxy instance __dict__. However, *reads* through the proxy correctly bypass this __dict__ and are redirected to the target. In effect, the proxy __dict__ has become write-only!

The solution to this is to override __setattr__() on the LoggingProxy too:

class LoggingProxy:

```
def __init__(self, target):
    super().__setattr__('target', target)

def __getattribute__(self, name):
    target = super().__getattribute__('target')

try:
    value = getattr(target, name)
    except AttributeError as e:
```

```
raise AttributeError("{} could not forward request {} to {}"
                super().__getattribute__('__class__').__name___,
                name,
                target)
            ) from e
    print("Retrieved attribute {!r} = {!r} from {!r}"
          .format(name, value, target))
    return value
def __setattr__(self, name, value):
    target = super().__getattribute__('target')
        setattr(target, name, value)
    except AttributeError as e:
        raise AttributeError("{} could not forward request {} to {}"
            .format(
                super().__getattribute__('__class__').__name__,
                target)
    print("Set attribute {!r} = {!r} on {!r}"
          .format(name, value, target))
```

With __setattr__() in place we can successfully write through the logging proxy to modify mutable attributes – here we update red to 55, and attempted writes to immutable attributes, such as p are rejected as planned:

```
>>> from vector import *
>>> from loggingproxy import *
>>>
>>> cw = ColoredVector(red=23, green=44, blue=238, p=9, q=14)
ColoredVector(red=23, green=44, blue=238, p=9, q=14)
>>> cw.red = 55
>>>
ColoredVector(red=55, green=44, blue=238, p=9, q=14)
>>> cw.p = 19
Traceback (most recent call last):
  File "examples/vector/vector.py", line 48, in __setattr__
   channel = ColoredVector.COLOR_INDEXES.index(name)
ValueError: tuple.index(x): x not in tuple
During handling of the above exception, another exception occurred:
Traceback (most recent call last):
 File "<stdin>", line 1, in <module>
  File "examples/vector/vector.py", line 50, in __setattr__
   super().__setattr__(name, value)
  File "examples/vector/vector.py", line 16, in __setattr_
   raise AttributeError("Can't set attribute {!r}".format(name))
AttributeError: Can't set attribute 'p'
```

Special methods which bypass __getattribute__()

It's important to realise that <u>__getattribute__</u>() only intercepts attribute lookup through the dot operator. Let's create a ColoredVector called cv and a LoggingProxy for it called cw:

```
>>> from vector import *
>>> from loggingproxy import *
>>> cv = ColoredVector(red=39, green=22, blue=89, s=45, t=12)
>>> cv
ColoredVector(red=39, green=22, blue=89, s=45, t=12)
>>> cw = LoggingProxy(cv)
>>> cw
<loggingproxy.LoggingProxy object at 0x101976ba8>
```

If we call the __repr__() method directly, the call is routed via the proxy and is dispatched successfully:

```
>>> cw.__repr__()
Retrieved attribute '__repr__' = <bound method ColoredVector.__repr__ of ColoredVector(re\
d=39, green=22, blue=89, s=45, t=12)> from ColoredVector(red=39, green=22, blue=89, s=45,\
t=12)
'ColoredVector(red=39, green=22, blue=89, s=45, t=12)'
```

However, if we request the repr() of cw in the conventional manner, using the built-in function, the __getattribute__() function of our LoggingProxy is not invoked, the call is not forwarded to the ColouredVector, and instead we get the default repr for LoggingProxy:

```
>>> repr(cw)
'<loggingproxy.LoggingProxy object at 0x101976ba8>'
```

This shows that __getattribute__() can only be used to intercept special method calls when the special method is retrieved *directly* – which is something we don't normally do. Normal access to facilities provided by special methods is through built-in functions such as len(), iter(), repr() and so on. These all bypass the __getattribute()__ override for performance reasons.

What this means in practice, is that if you want to write a proxy object such as LoggingProxy which transparently proxies an object including its repr or other special methods, it's up to you to provide an implementation of __repr__() that forwards the call appropriately:

```
def __repr__(self):
    target = super().__getattribute__('target')
    repr_callable = getattr(target, '__repr__')
    return repr_callable()
```

This now works when called via the built-in repr() function:

```
>>> from vector import *
>>> from loggingproxy import *
>>> cv = ColoredVector(red=39, green=22, blue=89, s=45, t=12)
>>> cv
ColoredVector(red=39, green=22, blue=89, s=45, t=12)
>>> cw = LoggingProxy(cv)
>>> repr(cw)
'ColoredVector(red=39, green=22, blue=89, s=45, t=12)'
```

The same goes for the other special methods.

Where are the methods?

One question you may have is "Where are the methods?". Why is it, that when we introspect the __dict__ of an object, we see only attributes and not methods, but that as we have just seen with __repr__ we can retrieve methods using getattr(). So, where *are* the methods?

The answer is that methods are attributes of another object, the class object associated with our instance. As we already know, we can get to the class object via the __class__ attribute, and sure enough it too has a __dict__ attribute which contains references to the callable objects which are the manifestations of the methods of our class.

Returning briefly to our Vector example:

```
>>> v = Vector(x=3, y=7)
>>> v.__dict__
{'_y': 7, '_x': 3}
>>> v.__class__
<class 'vector_09.Vector'>
>>> v.__class__.__dict__
mappingproxy({
'__delattr__': <function Vector.__delattr__ at 0x103013400>,
'__module__': 'vector_09',
'__setattr__': <function Vector.__setattr__ at 0x1030130d0>,
'__repr__': <function Vector.__repr__ at 0x103013488>,
'__init__': <function Vector.__init__ at 0x1030131e0>,
'__dict__': <attribute '__dict__' of 'Vector' objects>,
'__getattr__': <function Vector.__getattr__ at 0x103013158>,
'__weakref__': <attribute '__weakref__' of 'Vector' objects>,
'__doc__': None})
```

Of course we can retrieve the callable and pass our instance to it:

```
>>> v.__class__.__dict__['__repr__'](v)
'Vector(x=3, y=7)'
```

It's well worth spending some experimental time on your own poking around with these special attributes to get a good sense of how the Python object model hangs together.

It's worth noting that the __dict__ attribute of a *class* object is not a regular dict, but is instead of type mappingproxy, a special mapping type used internally in Python, which does not support item assignment:

```
>>> v.__class__.__dict__['a_vector_class_attribute'] = 5
TypeError: 'mappingproxy' object does not support item assignment
```

To add an attribute to a *class* you must use the setattr() function:

```
>>> setattr(v.__class__, 'a_vector_class_attribute', 5)
>>> Vector.a_vector_class_attribute
5
```

The machinery of setattr() knows how to insert attributes into the class dictionary:

```
>>> v._class_._dict__
mappingproxy({'__weakref__': <attribute '__weakref__' of 'Vector' objects>,
'__setattr__': <function Vector.__setattr__ at 0x101a5fa60>,
'a_vector_class_attribute': 5, '__doc__': None, '__repr__':
<function Vector.__repr__ at 0x101a5fb70>, '__init__':
<function Vector.__init__ at 0x101a5f620>, '__getattr__':
<function Vector.__getattr__ at 0x101a5f6a8>, '__module__':
'vector', '__dict__': <attribute '__dict__' of 'Vector' objects>,
'__delattr__': <function Vector.__delattr__ at 0x101a5fae8>})
```

Slots

We'll finish off this part of the course with a brief look at a mechanism in Python for reducing memory use: slots. As we've seen, each and every object stores its attributes in a dictionary. Even an empty Python dictionary is quite a hefty object, weighing in at 288 bytes:

```
>>> d = {}
>>> import sys
>>> sys.getsizeof(d)
288
```

If you have thousands or millions of objects this quickly adds up, causing your programs to need megabytes or gigabytes of memory. Given contemporary

computer architectures this tends to lead to reduced performance as CPU caches can hold relatively few objects.

Techniques to solve the high memory usage of Python programs can get pretty involved, such as implementing Python objects in lower-level languages such as C or C++, but fortunately Python provides the slots mechanism which can provide some big wins for low effort, with tradeoffs that will be acceptable in most cases.

Let's take a look! Consider the following class to describe the type of electronic component called a resistor:

class Resistor:

```
def __init__(self, resistance_ohms, tolerance_percent, power_watts):
    self.resistance_ohms = resistance_ohms
    self.tolerance_percent = tolerance_percent
    self.power_watts = power_watts
```

It's difficult to determine the size in memory of Python objects, but with care, we can use the <code>getsizeof()</code> function in the sys module. To get the size of an instance of Resistor, we need to account for the size of the Resistor object itself, and the size of its <code>__dict__</code>:

```
>>> from resistor import *
>>> r10 = Resistor(10, 5, 0.25)
>>>
>>> import sys
>>> sys.getsizeof(r10) + sys.getsizeof(r10.__dict__)
152
```

Python objects being the highly-dynamic dictionary-based objects they are we can add attributes to them a runtime, and see this reflected as an increased size:

```
>>> r10.cost_dollars = 0.02
>>> sys.getsizeof(r10) + sys.getsizeof(r10.__dict__)
248
```

This a quite a big object – especially when you consider that the equivalent struct in the C programming language would weight in at no more than 64 bytes with very generous precision on the number types.

Let's see if we can improve on this using slots. To use slots we must declare a class attribute called __slots__ to which we assign a sequence of strings

containing the fixed names of the attributes we want all instances of the class to contain:

class Resistor:

```
__slots__ = ['resistance_ohms', 'tolerance_percent', 'power_watts']

def __init__(self, resistance_ohms, tolerance_percent, power_watts):
    self.resistance_ohms = resistance_ohms
    self.tolerance_percent = tolerance_percent
    self.power_watts = power_watts
```

Now let's look at the space performance of this new class. We can instantiate Resistor just as before:

```
>>> from resistor import *
>>> r10 = Resistor(10, 5, 0.25)
```

And retrieve it's attributes in exactly the same way:

```
>>> r10.tolerance_percent
5
>>> r10.power_watts
0.25
```

However, it's size is much reduced, from 152 bytes down to 64 bytes, less than half the size:

```
>>> import sys
>>> sys.getsizeof(r10)
64
```

There's always a tradeoff though, as we can no longer dynamically add attributes to instances of Resistor:

```
>>> r10.cost_dollars = 0.02
Traceback (most recent call last):
   File "<input>", line 1, in <module>
AttributeError: 'Resistor' object has no attribute 'cost_dollars'
```

This is because the internal structure of Resistor no longer contains a dict:

```
>>> r10.__dict__
Traceback (most recent call last):
   File "<input>", line 1, in <module>
AttributeError: 'Resistor' object has no attribute '__dict__'
```

For most applications, slots won't be required, and you shouldn't use them unless measurements indicate that they may help, as slots can interact with other

Python features and diagnostic tools in surprising ways. In an ideal world, slots wouldn't be necessary and in our view they're quite an ugly language feature, but at the same time we've worked on applications where the simple addition of a __slots__ attribute has made the difference between the pleasure of programming in Python, and the pain of programming in a lower-level, but more efficient language. Use wisely!

Summary

Let's summarise what we've covered in this chapter:

- We discovered that Python objects store their attributes internally within a dictionary called __dict__ which maps attribute names to attribute values.
- We showed that instance attributes can be created, retrieved, updated and deleted by direct manipulation of __dict__.
- We showed how any failure to retrieve an attribute by normal means causes the __getattr__() special method to be invoked. The implementation of __getattr__() can use arbitrary logic to 'fake' the existence of attributes programatically.
- Similarly, assignment to, and deletion of, attributes can be customised by overriding __setattr__() and __delattr__().
- Calls to the hasattr() built-in function may also invoke __getattr__() so __getattr__() implementations need to be particularly careful to avoid non-terminating recursion.
- Occasionally, it's necessarily to customise *all* attribute lookup even for regular attributes. In these cases the default lookup machinery in the special __getattribute__() of the object base class may be overridden, taking care to delegate to the base class implementation as necessary, with a call via super().
- Method callables are stored in the __class__.__dict__ dictionary.
- Slots are a quick way to make Python objects more memory efficient at the cost of being less dynamic.

Chapter 4 - Descriptors

In this chapter we'll investigate a feature of Python you've been using — perhaps unknowingly — called *descriptors*. Descriptors are the mechanism used to implement properties in Python. We covered properties thoroughly in *The Python Journeyman*, but we'll start here with a short review.

In Chapter 4 of *The Python Journeyman* we showed how to create properties using the property decorator. In this chapter we'll dig deeper and show how to create properties in the raw using the property() constructor. Then we'll show how to create a specialised property by a defining a custom descriptor which implements the descriptor protocol. We'll round off by demonstrating that there are two categories of descriptor — data descriptors and non-data descriptors — and we'll show how these interact with Python's somewhat complicated attribute lookup rules.

A review of properties

As promised, we'll start with a very brief review of properties, our entry point into the world of descriptors. To explain descriptors, we'll be building a simple class to model planets, focusing on particular physical attributes such as size, mass, and temperature.

Let's start with this basic class definition for a planet in planet.py, consisting of little more than an initializer. There are no properties here yet, we'll add them in a moment:

This is simple enough to use $\frac{9}{2}$:

Unfortunately, our code also allows us to represent nonsensical situations, such as setting a negative radius, either by directly mutating an attribute, like this:

```
>>> pluto.radius_metres = -10000
```

Or by simply passing nonsense such as zero mass, negative orbital periods, or temperatures below absolute zero to the constructor:

```
>>> planet_x = Planet(name='X', radius_metres=10e3, mass_kilograms=0,
... orbital_period_seconds=-7293234, surface_temperature_kelvin=-5)
>>> planet_x.surface_temperature_kelvin
-5
```

We already know how to improve this: by wrapping our instance attribute in property getters and setters which perform validation by checking that the physical quantities are positive. We then assign through those properties in __init__ to get validation on construction for free:

```
# planet.py
class Planet:
    def __init__(self,
                 name,
                 radius_metres,
                 mass_kilograms,
                 orbital_period_seconds,
                 surface_temperature_kelvin):
        self.name = name
        self.radius_metres = radius_metres
        self.mass_kilograms = mass_kilograms
        self.orbital_period_seconds = orbital_period_seconds
        self.surface_temperature_kelvin = surface_temperature_kelvin
    @property
    def name(self):
        return self._name
    @name.setter
    def name(self, value):
        if not value:
           raise ValueError("Cannot set empty Planet.name")
        self._name = value
    @property
    def radius_metres(self):
```

```
return self._radius_metres
@radius_metres.setter
def radius_metres(self, value):
    if value <= 0:</pre>
        raise ValueError("radius_metres value {} is not "
                          "positive.".format(value))
    self._radius_metres = value
@property
def mass_kilograms(self):
    return self._mass_kilograms
@mass_kilograms.setter
def mass_kilograms(self, value):
    if value <= 0:</pre>
        raise ValueError("mass_kilograms value {} is not "
                          "positive.".format(value))
    self._mass_kilograms = value
@property
def orbital_period_seconds(self):
    return self._orbital_period_seconds
@orbital_period_seconds.setter
def orbital_period_seconds(self, value):
    if value <= 0:</pre>
        raise ValueError("orbital_period_seconds value {} is not "
                          "positive.".format(value))
    self._orbital_period_seconds = value
@property
def surface_temperature_kelvin(self):
    return self._surface_temperature_kelvin
@surface_temperature_kelvin.setter
def surface_temperature_kelvin(self, value):
    if value <= 0:</pre>
        raise ValueError("surface_temperature_kelvin value {} is not "
                          "positive.".format(value))
    self._surface_temperature_kelvin = value
```

From a robustness standpoint, this code is much better. For example, we can no longer construct massless planets:

The trade-off though, is that the amount of code has exploded, and worse, there is a lot of duplicated code checking that all those numeric attribute values are non-negative.

Descriptors will ultimately provide a way out of this, but first we need to do a little more unravelling of properties to aid our understanding.

Unravelling the property function

In *The Python Journeyman* we introduced property as a function decorator for property getters. To briefly recap, a getter method, which encapsulates direct attribute access, is decorated by the property decorator which creates a property object. The getter function is bound to an attribute of the property object called fget, and the original name of the getter is conceptually re-bound to the property object.

The property object also has an attribute called setter which is in fact another decorator. When a setter method is decorator by the setter, the original property object is modifed to bind an attribute called fset to the setter method.

The property object effectively aggregates the getter and setter into a single property object which behaves like an attribute, and it behaves like an attribute because it is a *descriptor*. Shortly, we'll learn how it is able to appear so attribute-like, but first let's unravel properties a bit more.

Remember that function decorators are just regular functions which process an existing function and return a new object — usually a new function which wraps the decorated function.

Instead of using decorator syntax to apply the decorator to a function, we can define a regular undecorated function and then pass the function to the decorator, rebinding it to the same, or indeed any other, name.

Given that decorators are functions, let's rework our code to apply property explicitly using regular function-call syntax, avoiding the special decorator application syntax using the @ symbol. When doing this, it's important to note that the property() function supports several arguments for simultaneously supplying the getter, setter and deleter, functions along with a docstring value. In fact, help(property) makes this quite clear:

```
fget is a function to be used for getting an attribute value, and likewise
fset is a function for setting, and fdel a function for del'ing, an
attribute. Typical use is to define a managed attribute x:
class C(object):
    def getx(self): return self._x
def setx(self, value): self._x = value
def delx(self): del self._x
    x = property(getx, setx, delx, "I'm the 'x' property.")
Decorators make defining new properties or modifying existing ones easy:
class C(object):
    @property
    def x(self):
         "I am the 'x' property."
         return self. x
    @x.setter
    def x(self, value):
        self._x = value
    @x.deleter
    def x(self):
         del self._x
```

As you can see, we can separately define our getter and setter functions, then call the property() constructor within the *class* definition to produce a class attribute. Let's use this form in our Planet class for the numerical attributes. We'll leave the name attribute using the decorator form so you can compare side-by-side.

Implementing properties without decorators

First we'll remove the property decorator from the getter functions, before removing the setter decorator from the setter functions. Then we'll prefix the names of the getter functions with <code>_get_</code> and the names of the setter functions with <code>_set_</code>. We use single-underscore prefixes here because these are not special methods. Finally, we create a property object using the property function — you can think of it as a constructor call in this context — passing the pair of getter and setter functions. Continuing, we need to do the same for the <code>mass_kilograms</code> property, the <code>orbital_period_seconds</code> property, and the <code>surface_temperature_kelvin</code> property:

class Planet:

```
self.orbital_period_seconds = orbital_period_seconds
    self.surface_temperature_kelvin = surface_temperature_kelvin
@property
def name(self):
    return self._name
@name.setter
def name(self, value):
    if not value:
        raise ValueError("Cannot set empty Planet.name")
    self._name = value
def _get_radius_metres(self):
    return self._radius_metres
def _set_radius_metres(self, value):
    if value <= 0:</pre>
        raise ValueError("radius_metres value {} is not "
                          "positive.".format(value))
    self._radius_metres = value
radius_metres = property(fget=_get_radius_metres,
                          fset=_set_radius_metres)
def _get_mass_kilograms(self):
    return self._mass_kilograms
def _set_mass_kilograms(self, value):
    if value <= 0:</pre>
        raise ValueError("mass_kilograms value {} is not "
                          "positive.".format(value))
    self._mass_kilograms = value
mass_kilograms = property(fget=_get_mass_kilograms,
                          fset=_set_mass_kilograms)
def _get_orbital_period_seconds(self):
    return self._orbital_period_seconds
def _set_orbital_period_seconds(self, value):
    if value <= 0:</pre>
        raise ValueError("orbital_period_seconds value {} is not "
                          "positive.".format(value))
    self._orbital_period_seconds = value
orbital_period_seconds = property(fget=_get_orbital_period_seconds,
                                   fset=_set_orbital_period_seconds)
def _get_surface_temperature_kelvin(self):
    return self._surface_temperature_kelvin
def _set_surface_temperature_kelvin(self, value):
    if value <= 0:</pre>
        raise ValueError("surface_temperature_kelvin value {} is not "
                          "positive.".format(value))
    self._surface_temperature_kelvin = value
surface_temperature_kelvin = property(fget=_get_surface_temperature_kelvin,
                                       fset=_set_surface_temperature_kelvin)
```

If you think this form of property set up is a retrograde step compared to the decorator form, we'd agree with you; there's a good reason we introduce properties as decorators first. Nevertheless, seeing this alternative style reinforces the notion that property is simply a function, which returns an object called a descriptor, which is in turn bound to a class attribute.

The runtime behaviour of this code hasn't changed at all. We can still create objects, retrieve attribute values through properties, and attempt to set attributes values through properties with rejection of nonsensical values:

We already know what sort of operations we can perform with a descriptor. We can *get* a value, *set* a value and *delete* a value. In the case of property, these getting, setting and deleting operations call functions we supply, which query and manipulate instance attributes. In general though, the descriptor operations can be implemented to do almost anything.

Implementing a descriptor

We've seen that property is a descriptor which wraps three functions. Let's create a more specialised descriptor useful for modelling the strictly positive numeric values in our Planet class.

Here is a simple descriptor, called Positive:

```
from weakref import WeakKeyDictionary

class Positive:

    def __init__(self):
        self._instance_data = WeakKeyDictionary()

    def __get__(self, instance, owner):
        return self._instance_data[instance]

    def __set__(self, instance, value):
        if value <= 0:</pre>
```

```
raise ValueError("Value {} is not positive".format(value))
self._instance_data[instance] = value

def __delete__(self, instance):
    raise AttributeError("Cannot delete attribute")
```

The descriptor class implements the three functions which comprise the descriptor protocol: __get__(), __set__(), and __delete__(). These are called when we get a value from a descriptor, set a value through a descriptor, or delete a value through a descriptor, respectively. In addition, the Positive class implements __init__() to configure new instances of the descriptor. Before we look in more detail at each of these methods, let's make use of our new descriptor to refactor our Planet class.

We remove the setters and getters for radius_meters and replace the call to the property constructor with a call to the Positive constructor. We do the same for the mass_kilograms, orbital_period_seconds and surface_temperature_kelvin quantities:

```
class Planet:
```

```
def __init__(self,
             name,
             radius_metres,
             mass_kilograms,
             orbital_period_seconds.
             surface_temperature_kelvin):
    self.name = name
    self.radius_metres = radius_metres
    self.mass_kilograms = mass_kilograms
    self.orbital_period_seconds = orbital_period_seconds
    self.surface temperature kelvin = surface temperature kelvin
@property
def name(self):
    return self._name
@name.setter
def name(self, value):
    if not value:
        raise ValueError("Cannot set empty Planet.name")
    self._name = value
radius_metres = Positive()
mass_kilograms = Positive()
orbital_period_seconds = Positive()
surface_temperature_kelvin = Positive()
```

With the Positive descriptor on hand, the Planet class shrinks by a huge amount!

At first sight, this may appear confusing. It looks like we're assigning to radius_metres twice — once in the initializer and once in the body of the class. In fact, the call in the body of the class is binding a instance of a Positive descriptor object to a *class* attribute of Planet. The call in __init__ is then apparently assigning to an *instance* attribute — although as we'll see in a moment, this assignment is actually invoking a method on the descriptor object.

Allow us to explain the machinery we've created: As we said earlier Postive implements *descriptor protocol* consisting in full of the three methods __get__, __set__ and __delete__.

Let's start with an instance of Planet:

When we retrieve an attribute like this:

```
>>> pluto.mass_kilograms
1.305e+22
```

The Positive.__get__() function is called. The instance argument is set to pluto and the owner argument is set to the object which *owns* the descriptor, in this case the *class* Planet. So our attribute retrieval above, is equivalent to:

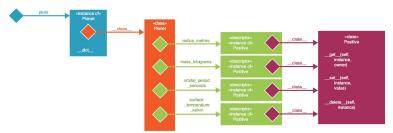
```
Positive.__get__(pluto, Planet)
```

Similarly, our assignments to the descriptors in the Planet.__init__() function resolve to calls equivalent to:

```
Positive.__set__(self, radius_metres)
```

At this point the new Planet instance isn't yet bound to the name pluto, only to self.

Here's what we have so far in a graphical format:



Positive descriptors on Pluto

We have a reference named pluto bound to an instance of Planet. The __class__ reference of the pluto instance points to the Planet class object. The Planet class object contains four attributes, each of which references a distinct instance of the Positive descriptor object. The __class__ reference of each descriptor object refers to the Positive class. When you put it all together, the call m = pluto.mass_kilograms boils down to m = Positive.__get__(pluto, Planet).

Storing instance data

Notice that because the descriptor is owned by the Planet *class*, rather than by the pluto *instance*, we can't just store the value in the descriptor object. If we did that, the value would be shared between all Planet instances. Instead, we need to somehow associate the attribute value with the instance.

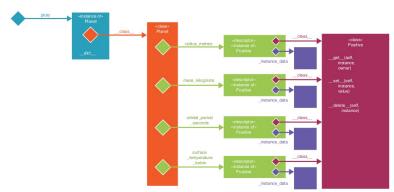
At first, this seems easy. The __get__() call is handed a reference to the instance - pluto in this case - so why not store the value in pluto's __dict__?

There is a problem though: Within the descriptor class Positive we have no way of knowing to which attribute name the descriptor is bound in the Planet class. We can distinguish between descriptor instances in __get__ using the self argument, but in Python objects do not know which names have been bound to them. This means we have no way of correlating between descriptor instances and the descriptor names embedded in the Planet class. In other words, none of our descriptor objects know which attribute they represent. This in turn means that we can't store the descriptor value in the __dict__ of Pluto, because we wouldn't know what dictionary key to use.

This apparent shortcoming of descriptors, not knowing the name of the class attribute to which they are bound, is also evident in the fact that our value validation error message in __set__ no

longer mentions the attribute name; this is a clear regression in capabilities. This is fixable, but the solution will have to wait to the *next* chapter when we look at metaclasses.

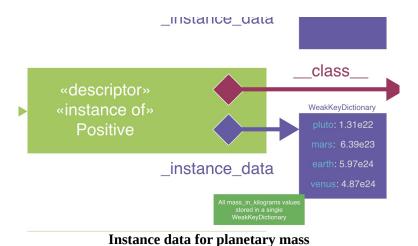
So, how to associate the values with this instances? Let's look at the solution pictorially first, then we'll reiterate with the code.



Instance data for Pluto

We use a special collection type from the Python Standard Library's weakref package called WeakKeyDictionary. This works pretty much like a regular dictionary except that it won't retain value objects which are referred to only by the dictionary key references; we say the references are weak. A WeakKeyDictionary owned by each descriptor instance is used to associate planet instances with the values of the quantity represented by that descriptor — although the descriptor itself doesn't know which quantity is being represented.

For example, the WeakKeyDictionary shown here associates planet instances with mass_in_kilograms values.



Similar WeakKeyDictionary instances associate planet instances with radius_in_metres, orbital_period_seconds, and surface_temperature_kelvin values.

Because the dictionary keys are *weak* references, if a planet instance is destroyed — let's pretend the Earth is vapourised to make way for a hyperspace bypass — the corresponding entries in all the weak-key-dictionaries are also removed.

Examining the implementation

Let's look at how this is all put together in code. A WeakKeyDictionary instance called _instance_data is created in the descriptor initializer:

```
class Positive:
    def __init__(self):
        self._instance_data = WeakKeyDictionary()
```

As a result, our Planet class indirectly aggregates *four* such dictionaries, one in each of the four descriptors:

```
class Planet:
    # . . .
    radius_metres = Positive()
    mass_kilograms = Positive()
    orbital_period_seconds = Positive()
    surface_temperature_kelvin = Positive()
```

Within the __set__() method we associate the attribute value with the planet instance by inserting a mapping from the instance as key to the attribute as value:

```
class Positive:
    ...
    def __set__(self, instance, value):
        if value <= 0:
            raise ValueError("Value {} is not positive".format(value))
        self._instance_data[instance] = value</pre>
```

As such, a single dictionary will contain all the radius_metres values for all Planet instances. Another dictionary will contain all mass_kilograms values for all Planet instances, and so on. We're storing the instance attribute values completely outside the instances, but in such a way that we can reliably retrieve them in __get__.

Retrieving descriptors on classes

One aspect of the descriptor protocol we haven't yet addressed is what happens when we retrieve a descriptor from a class. As we've seen instance attribute retrieval works fine:

```
>>> mercury, venus, earth, mars = main()
>>> mars.radius_metres
3389500.0
```

However, class attribute retrieval does not:

```
>>> Planet.radius_metres
Traceback (most recent call last):
    File "<input>", line 1, in <module>
    File "examples/descriptors/properties_05.py", line 10, in __get__
        return self._instance_data[instance]
    File "/Library/Frameworks/Python.framework/Versions/3.4/lib/python3.4/weakref.py", line\
345, in __getitem__
    return self.data[ref(key)]
TypeError: cannot create weak reference to 'NoneType' object
```

In such cases, the instance argument of __get__() will be set to None. Because we cannot create a weak reference to None, this causes a failure with the WeakKeyDictionary used for attribute storage.

By testing the instance argument against None we can detect when a descriptor value is being retrieved via a class attribute.

Here's a modified Positive descriptor:

```
class Positive:
    def __init__(self):
        self._instance_data = WeakKeyDictionary()
```

```
def __get__(self, instance, owner):
    if instance is None:
        return self
    return self._instance_data[instance]

def __set__(self, instance, value):
    if value <= 0:
        raise ValueError("Value {} is not positive".format(value))
    self._instance_data[instance] = value

def __delete__(self, instance):
    raise AttributeError("Cannot delete attribute")</pre>
```

In the __get__() implementation when instance is None we do what seems most natural, returning the descriptor object itself rather than performing the attribute lookup. With the revised code we see more helpful behaviour:

```
>>> Planet.radius_metres
cproperties_06.Positive object at 0x103012d68>
```

If you need to query or manipulate the class which contains the descriptor objects you can get hold of this through the owner argument of the __get__ method, which in this case will contain a reference to the Planet class. In many cases though, you won't need to use owner, so you can do as we've done and just ignore it.

Data versus non-data descriptors

You may have heard the terms "data descriptor" and "non-data descriptor", but what do they mean?

A *non-data descriptor* is a descriptor which implements only the __get__() method and so is read-only. So called *data descriptors* sport both the __get__() and __set__() methods and are writable.

The distinction is important because of the precedence of attribute lookup. A precedence chain controls attribute lookup according to the following rules:

- If an *instance*'s __dict__ has an entry with the same name as a *data descriptor*, the data descriptor takes precedence.
- If an *instance*'s __dict__ has an entry with the same name as a *non-data descriptor*, the dictionary entry takes precedence.

These statements are true, but quite a mouthful.

Another way of looking at it is that attribute lookup proceeds first to data descriptors (such as properties defined in the class), then to instance attributes in **dict**, and then on to non-data descriptors in the class again.

Approaching the difference experimentally

A simple experiment should make things clearer. Here we define a data descriptor, a non-data descriptor and a class which uses them both:

Let's try this in a REPL session. After we've created an instance of Owner, we'll retrieve the attribute a, set an item in the instance dictionary with the same name, and retrieve a again:

Since a is a data descriptor, the first rule applies and the data descriptor takes precedence when we reference obj.a.

Now let's try that with the non-data attribute b:

The first time we access obj.b there is no entry of the same name in the instance dictionary. As a result, the non-data descriptor takes precedence. After we've added a b entry into __dict__, the second rule applies and the dictionary entry takes precedence over the non-data descriptor.

Summary

We have seen that the descriptor protocol is itself very simple, which can lead to very concise and declarative code, which hugely reduces duplication. At the same time, implementing descriptors correctly can be tricky and requires careful testing.

Let's review the key points:

- We demonstrated how to create property descriptors without using decorator syntax by passing property getters and setters directly to the property() function. This reinforced the notion that properties create objects called *descriptors* which are bound to *class* attributes.
- We showed how to implement a simple descriptor to perform a basic attribute validation check by creating a class that implemented the descriptor protocol.
- We explained how descriptor instances have no knowledge of to which class attributes they are bound. This means that each descriptor instance must store the instance attributes for all descriptor owner instances. This can be achieved using a WeakKeyDictionary from the Python Standard Library weakref module.
- We saw how to determine whether descriptors are retrieved from their owning *classes* rather than via *instances*, by detecting when the instance argument to __get__() is set to None. A natural course of action in such as case is to return the descriptor instance itself.
- We explained the distinction between data and non-data descriptors, and how this relates to attribute lookup precedence.

We're not quite done with descriptors yet, and in particular we'd like descriptor instances to know the name of the class attributes to which they have been bound. To solve that problem we need sharper tools, in the form of Python's metaclasses, which we'll be covering in the next chapter.

Chapter 5 - Instance Creation

In this chapter module we'll be taking a deep-dive into exactly what happens when we create a new object. With this knowledge in hand, we'll be able to exercise fine control over instance creation, which allows us to customize Python objects in powerful ways.

Instance Creation

So what *does* happen when you create an object?

To illustrate, we'll an 8×8 chess board. Consider this simple class which represents a coordinate on a chess board consisting of a *file* (column) letter from 'a' to 'h' inclusive and a *rank* (row) number from 1 to 8 inclusive:

class ChessCoordinate:

```
def __init__(self, file, rank):
    if len(file) != 1:
        raise ValueError("{} component file {!r} does not have a length of one."
                         .format(type(self).__name__, file))
    if file not in 'abcdefgh':
        raise ValueError("{} component file {!r} is out of range."
                         .format(type(self).__name__, file))
    if rank not in range(1, 9):
        raise ValueError("{} component rank {!r} is out of range."
                         .format(type(self).__name__, rank))
    self._file = file
    self.\_rank = rank
@property
def file(self):
    return self._file
@property
def rank(self):
    return self._rank
def __repr__(self):
    return "{}(file={}, rank={})".format(
        type(self).__name__, self.file, self.rank)
def __str__(self):
    return '{}{}'.format(self.file, self.rank)
```

This class implements an immutable value type; the initialiser establishes the invariants and the property accessors prevent inadvertent modification of the encapsulated data.

init is not a constructor

It's easy to think of __init__() as the *constructor* implementation, but let's look closely at what happens when we create an instance of ChessCoordinate by calling the constructor. We'll use a debugger — PDB in this case, but any debugger will do — to pause on the first line of __init__(). Since __init__() accepts self as an argument, it's clear that the object referred to by self *already exists*. That is to say, the object has already been constructed and the job of __init__() really is just to initialise the object:

```
>>> white_queen = ChessCoordinate('d', 4)
> /Users/sixtynorth/sandbox/chess_coordinate.py(6)__init__()
-> if len(file) != 1:
(Pdb) self
*** AttributeError: 'ChessCoordinate' object has no attribute '_file'
```

The AttributeError indicates that our debugger is having difficulty displaying the uninitialised object. This is because PDB is using repr to display it, and our __repr__() implementation, quite reasonably, expects that the object has been initialised. However, we can check type(self) to see that self already has the required type, and we can look at self.__dict__ to see that the instance dictionary is empty:

```
(Pdb) type(self)
<class '__main__.ChessCoordinate'>
(Pdb) self.__dict__
{}
```

As we continue to step through the initializer, we can watch the instance dictionary get populated through assignments to attributes of self. From chapter we know that behind the scenes object.__setattr__() is being called:

```
(Pdb) next
> /Users/sixtynorth/sandbox/chess_coordinate.py(10)__init__()
-> if file not in 'abcdefgh':
(Pdb) next
> /Users/sixtynorth/sandbox/chess_coordinate.py(14)__init__()
-> if rank not in range(1, 9):
(Pdb) next
> /Users/sixtynorth/sandbox/chess_coordinate.py(18)__init__()
-> self._file = file
(Pdb) next
> /Users/sixtynorth/sandbox/chess_coordinate.py(19)__init__()
-> self._rank = rank
```

```
(Pdb) self.__dict__
{'_file': 'd'}
(Pdb) next
--Return--
> /Users/sixtynorth/sandbox/chess_coordinate.py(19)__init__()->None
-> self._rank = rank
(Pdb) self.__dict__
{'_file': 'd', '_rank': 4}
```

Note also that __init__() doesn't return anything; it simply mutates the instance it has been given.

Construction and __new__

So, if __init__() isn't responsible for creating the instance, what is? If we use dir() to look at the special methods of ChessCoordinate, we see one called __new__():

```
>>> dir(ChessCoordinate)
['_class__', '__delattr__', '__dict__', '__dir__', '__doc__',
'_eq__', '__format__', '__ge__', '__getattribute__', '__gt__',
'_hash__', '__init__', '__le__', '__lt__', '__module__', '__ne__',
'_new__', '_reduce__', '__reduce_ex__', '__repr__', '__setattr__',
'_sizeof__', '__str__', '__subclasshook__', '__weakref__', 'file',
'rank']
```

We haven't defined __new__() ourselves, but we do inherit an implementation from the universal base class, object. It is this base-class implementation of __new__() which is responsible for allocating our object in this case:

```
>>> ChessCoordinate.__new__ is object.__new__
True
```

But what is the signature of __new__()? Don't bother looking in the help() because, frankly, the answer isn't very helpful!:

```
>>> help(object.__new__)
Help on built-in function __new__:
    __new__(*args, **kwargs) method of builtins.type instance
    Create and return a new object. See help(type) for accurate signature
```

Instead, let's figure out the signature of __new__() by instrumentation. We'll implement the most basic override of __new__ which simply delegates to the base-class implementation. We'll add a few print statements so we can easily inspect the arguments and return value:

```
class ChessCoordinate:
```

```
def __new__(cls, *args, **kwargs):
   print("cls =", cls)
    print("args =", repr(args))
    print("kwargs =", repr(kwargs))
    obj = super().__new__(cls)
    print("id(obj) =", id(obj))
    return obj
def __init__(self, file, rank):
    print("id(self) =", id(self))
    if len(file) != 1:
        raise ValueError("{} component file {!r} does not have a length of one."
                         .format(type(self).__name__, file))
    if file not in 'abcdefgh':
        raise ValueError("{} component file {!r} is out of range."
                         .format(type(self).__name__, file))
    if rank not in range(1, 9):
        raise ValueError("{} component rank {!r} is out of range."
                         .format(type(self).__name__, rank))
    self._file = file
    self._rank = rank
# Other implementation omitted...
```

Notice that <code>__new__()</code> appears to be implicitly a *class* method. It accepts cls as it's first argument, rather than self. In fact <code>__new__()</code> is a specially-cased static method that happens to take the type of the class as its first argument — but the distinction isn't important here.

Here's what it looks like when we use this in a REPL:

```
>>> white_queen = ChessCoordinate('d', 4)
cls = <class '__main__.ChessCoordinate'>
args = ('d', 4)
kwargs = {}
id(obj) = 4598210744
id(self) = 4598210744
```

The cls argument is the *class* of the new object which will be be allocated. In our case, that will be ChessCoordinate, but in the presence of inheritance, it isn't necessarily the case that the cls argument will be set to the class enclosing the __new__() definition.

In general, besides the class object, __new__() also accepts whatever arguments have been passed to the constructor. In this case, we've soaked up any such arguments using *args and **kwargs although we could have used specific argument names here, just as we have with __init__(). We print these additional argument values to the console.

Remember that the purpose of __new__() is to allocate a new object. There's no special command for that in Python — all object allocation must ultimately be done by object.__new__(). Rather than call the object.__new__() implementation directly, we'll call it via super(). Should our immediate base class change in the future, this is more maintainable. The return value from the call to object.__new__() is a new, uninitialised instance of ChessCoordinate. We print its id() (remember we can't expect repr() to work yet) and then return.

It is this returned object which is then passed as the self argument to __init__(). We have printed the id() of self here to demonstrate that this is indeed the case.

Customising allocation

We've shown the essential mechanics of overriding __new__(): we accept the class type and the constructor arguments and return an instance of the correct type. Ultimately the only means we have for creating new instances is by calling object.__new__(). This is all well and good, but what are some practical uses?

One use for controlling instance creation is a technique called *interning*, which can dramatically reduce memory consumption. We'll demonstrate this by extending our program to allocate some chess boards in the start-of-game configuration. In our implementation each board is represented as a dictionary mapping the name of a piece to one of our ChessCoordinate objects. For fun, we've used the Unicode chess code points to represent our pieces. In our program 'ww' means "black queen's rook" and 'drad' mean "white king's bishop's pawn". We need to be this specific because, as you'll remember, dictionaries require that keys are distinct.

We'll revise our program by first removing ChessCoordinate.__new__ and removing the call to print in __init__. We'll also add a starting_board() function that creates a board in the starting configuration:

```
'wïa': ChessCoordinate('a', 2),
'∰∆∆': ChessCoordinate('b', 2),
'må&': ChessCoordinate('c', 2),
'wwa': ChessCoordinate('d', 2),
'��&': ChessCoordinate('e', 2),
'this': ChessCoordinate('f', 2),
'this': ChessCoordinate('f', 2),
'this': ChessCoordinate('g', 2),
'this': ChessCoordinate('h', 2),
''≝': ChessCoordinate('a', 8),
 '∰&': ChessCoordinate(̀'b', 8),
 '∰i': ChessCoordinate('c', 8),
'ww': ChessCoordinate('d', 8),
 ••': ChessCoordinate('e',
 '•1': ChessCoordinate('f',
'•▲': ChessCoordinate('g', 8),
 '•ێ': ChessCoordinate('h', 8),
 'wx▲': ChessCoordinate('a', 7),
 '∰‱∆': ChessCoordinate('b', 7),
 '***': ChessCoordinate('c', 7),
'***': ChessCoordinate('d', 7),
'**A': ChessCoordinate('e', 7),
'**A': ChessCoordinate('f', 7),
'**A': ChessCoordinate('g', 7),
 ♠X▲': ChessCoordinate('h', 7),
```

Let's go to the REPL and create a board using our new function:

```
>>> board = starting_board()
```

After creating a single board this way, our system reports that our Python process is using about 7 MB of memory. Creating 10,000 chess-boards utilises some 75 MB of memory to store the 320,000 instances of ChessCoordinate contained by the 10,000 dictionaries that have been created to represent all the boards:

```
>>> boards = [starting_board() for _ in range(10000)]
```

Bear in mind though, that there are only 64 distinct positions on a chess board, and given that our ChessCoordinate objects are deliberately immutable value types, we should never need more than 64 instances. In our specific case, in fact, we should never need more that 32 instances.

Interning allocated objects

Here are updated definitions for __new__() and __init__() which allow us to limit allocations to just the necessary ones, one per coordinate:

```
class ChessCoordinate:
   _interned = {}
```

```
def __new__(cls, file, rank):
    if len(file) != 1:
        raise ValueError("{} component file {!r} does not have a length of one."
                         .format(cls.__name__, file))
    if file not in 'abcdefgh':
        raise ValueError("{} component file {!r} is out of range."
                         .format(cls.__name__, file))
    if rank not in range(1, 9):
        raise ValueError("{} component rank {!r} is out of range."
                         .format(cls.__name__, rank))
    key = (file, rank)
    if key not in cls._interned:
        obj = super().__new__(cls)
        obj._file = file
        obj.\_rank = rank
        cls._interned[key] = obj
    return cls._interned[key]
def __init__(self, file, rank):
    pass
```

We're now using named positional arguments for the file and rank arguments to __new__(), and we've moved the validation logic from __init__() to __new__().

Once the arguments are validated, we use them to create a single tuple object from file and rank. We use this tuple as a key and check if there is an entry against this key tuple in a dictionary called _interned which we've attached as a class-attribute. Only if the tuple key is not present in the dictionary do we allocate a new instance; we do this by calling object.__new__(). We then configure the new instance — doing the remainder of the work that used to be done in __init__() — and insert the newly minted instance into the dictionary. Of course, the instance we return is an element in the dictionary, either the one we just created or one created earlier.

Our __init__() method is now empty, and in fact we could remove it entirely.

With these changes in place allocating 10,000 boards takes significantly less memory than before — we're down to about 23 MB.

Interning is a powerful tool for managing memory usage — in fact Python uses it internally for integers and strings — but it should only be used with immutable value types such as our ChessCoordinate where instances can safely be shared between data structures.

Now that we understand __new__(), in the next chapter we can move on to another advanced Python topic: metaclasses.

Summary

We've covered an crucial topic in this short chapter: the distinction between the allocation and initialization of instances. Let's summarise what we covered:

- We showed how the static method __new__() is called to allocate and return a new instance.
 - __new___() is implicitly a static method which accepts the class of the new instance as its first argument, and it doesn't require the either the @classmethod or @staticmethod decorators
- Ultimately, object.__new__() is responsible for allocating all instances.
- One use for overriding __new__() is to support instance interning which can be useful when certain values of immutable value types are very common, or when the domain of values is small and finite, such as with the squares of a chess board.

This isn't the whole story though, and Python offers yet more control over instance creation at the class level. Before we get to that, though, we need to understand metaclasses.

Chapter 6 - Metaclasses

We've mentioned metaclasses several times. We owe you an explanation!

In this chapter, we'll look at:

- the concept of the class of class objects *metaclass* for short
- the default metaclass, called type
- specifying metaclasses in class definitions
- defining metaclasses, including the special methods of metaclasses
- practical examples of metaclasses which solve some problems we discovered earlier in the book
- how metaclasses interact with inheritance

The class of class objects

To assist us on our journey to understand metaclasses, we need a simple class. We'll use Widget, which will be empty:

```
class Widget:
    pass
```

We'll instantiate some widgets:

```
>>> w = Widget()
>>> w
<__main__.Widget object at 0x1006747b8>
```

and introspect their types:

```
>>> type(w)
<class '__main__.Widget'>
```

We all know that in Python, the type of an instance is its class. So what is the type of its *class*?:

```
>>> type(Widget)
<class 'type'>
```

Using type as a function and a value

The type of a class is type. We've covered this before in chapter 12 of The Python Journeyman, but it bears repeating here. One potentially confusing aspect here is that we are using type() as a function to determine the type of an object, but type is also being used as a value — it is the type of a class object. However, this duality isn't so unusual; we see exactly the same situation with, say, list where it is used as a constructor and a value:

```
>>> a = list("A list")
>>> a
['A', ' ', 'l', 'i', 'i', 's', 't']
>>> type(a)
<class 'list'>
```

Given that the type of the Widget class is type we can say that the *metaclass* of Widget is type. In general, the type of any class object in Python is its metaclass, and the default metaclass is type.

Going further, we can discover that the type of type is type:

```
>>> type(type)
<class 'type'>
```

We can get the same results by drilling down through special attributes:

```
>>> w.__class__
<class '__main__.Widget'>
>>> w.__class__._class__
<class 'type'>
>>> w.__class__._class__._class__
<class 'type'>
```

Metaclasses and class creation

To understand where metaclasses fit into Python, we must consider how class objects are created. When we define a class in a Python source file, like this:

```
class Widget:
    pass
```

this is actually shorthand for:

```
class Widget(object, metaclass=type)
    pass
```

In addition to passing base classes as positional arguments we can pass the metaclass keyword argument. In the same way that the base class is implicitly object, the metaclass is implicitly type.

Class allocation and initialisation

Roughly speaking when we write a class block in our Python source code, it is syntactic sugar for creating a dictionary which is passed to a metaclass to convert the dictionary into a class object. To see how that works, it's import to understand the several tasks that metaclasses perform during creation of a new class. When we write:

```
class Widget:
    pass
```

what is actually happening is something like this:

```
name = 'Widget'
metaclass = type
bases = ()
kwargs = {}
namespace = metaclass.__prepare__(name, bases, **kwargs)
Widget = metaclass.__new__(metaclass, name, bases, namespace, **kwargs)
metaclass.__init__(Widget, name, bases, namespace, **kwargs)
```

Which is to say that:

- The class name is 'Widget'
- The metaclass is type
- The class has no base classes, other than the implicit object
- No keyword arguments were passed to the metaclass we'll cover what this means later
- The metaclass's __prepare__() method is called to create a new namespace object, which behaves like a dictionary
- Behind the scenes, the Python runtime populates the namespace dictionary while reading the contents of the class-block
- The metaclass's __new__() method is called to allocate the class object
- The metaclass's __init__() is called to initialise the class object.

The name, bases and namespace arguments contain the information collected during execution of the class definition — normally the class attributes and method definitions inside the class block — although in our case the class-block is logically empty.

Tracing metaclass invocations

By providing our own metaclass we can customise these behaviours. We'll start with a very simple metaclass called TracingMeta in a module tracing.py.

TracingMeta prints its method invocations and return values to the console for each of __prepare__(), __new__() and __init__(). Notice that our metaclass needs to be a subclass of an existing metaclass, so we will subclass type. Each of our overrides delegates to the base-class type to do the actual work required, via a call to super():

```
class TracingMeta(type):
```

```
@classmethod
def __prepare__(mcs, name, bases, **kwargs):
    print("TracingMeta.__prepare__(name, bases, **kwargs)")
    print("mcs =", mcs)
    print("name =", name)
    print("bases =", bases)
print("kwargs =", kwargs)
    namespace = super().__prepare__(name, bases)
    print("<-- namespace =", namespace)</pre>
    print()
    return namespace
def __new__(mcs, name, bases, namespace, **kwargs):
    print("TracingMeta.__new__(mcs, name, bases, namespace)")
    print("mcs =", mcs)
print("name =", name)
print("bases =", bases)
    print("namespace =", namespace)
    print("kwargs =", kwargs)
    cls = super().__new__(mcs, name, bases, namespace)
    print("<-- cls =", cls)</pre>
    print()
    return cls
def __init__(cls, name, bases, namespace, **kwargs):
    print("TracingMeta.__init__(cls, name, bases, namespace)")
    print("cls =", cls)
    print("name =", name)
    print("bases =", bases)
    print("namespace =", namespace)
print("kwargs =", kwargs)
    super().__init__(name, bases, namespace)
    print()
```

Notice that although __new__() is *implictly* a class-method, we must explicitly decorate the __prepare__() class-method with the appropriate decorator.

Now we'll define a class containing a simple method called action() which prints a message and a single class attribute the_answer with the <u>cosmically inevitable value 42</u>. We'll do this at the REPL so you can see clearly *when* the metaclass machinery is invoked:

```
>>> from tracing import TracingMeta
>>> class Widget(metaclass=TracingMeta):
```

```
... def action(message):
... print(message)
         the\_answer = 42
TracingMeta.__prepare__(name, bases, **kwargs)
mcs = <class 'tracing.TracingMeta'>
name = Widget
bases = ()
kwargs = \{\}
<-- namespace = \{\}
TracingMeta.__new__(mcs, name, bases, namespace, **kwargs)
mcs = <class 'tracing.TracingMeta'>
name = Widget
bases = ()
namespace = {'__qualname__': 'Widget', 'action': <function Widget.action at 0x1030162f0>,\
   'the_answer': 42, '__module__': 'builtins'}
<-- cls = <class 'Widget'>
TracingMeta.__init__(cls, name, bases, namespace, **kwargs)
cls = <class 'Widget'>
name = Widget
bases = ()
namespace = {'__qualname__': 'Widget', 'action': <function Widget.action at 0x1030162f0>,\
   'the_answer': 42, '__module__': 'builtins'}
  kwargs = \{\}
```

The instant we complete the class definition we can see from the tracing output that Python executes the __prepare__(), __new__() and __init__() methods in turns.

The __prepare__ method

First of all, lets look at __prepare__(), the purpose of which is to produce an initial mapping object to contain the class namespace. The mcs argument is a reference to the metaclass itself. This first argument is analagous to the self argument passed to instance methods and the cls argument passed to classmethods. For metaclass methods, it is conventionally called mcs.

The name argument contains the name of our Widget class as a string.

The bases argument is an empty tuple. We didn't declare any base classes for Widget, and the ultimate object base-class is implicit.

The kwargs argument is an empty dictionary. We'll cover the significance of this shortly.

The most important aspect of __prepare__ is that when it calls its superclass implementation in type, the return value is a dictionary — or more generally a

mapping type. In this case it's a regular empty dictionary. This dictionary will be the namespace associated with the nascent class.

The __new__ method

Next we'll look in detail at __new__(), the purpose of which is to allocate the new class object. The mcs argument is a reference to the metaclass as before. The name and bases arguments are still the string name of the new class and the tuple of base classes.

The mapping object that we returned from __prepare__() is passed as the namespace argument to __new__(). The Python runtime has populated this dictionary with several entries as it has processed the class definition of widget. Two of the items are our action() method and our the_answer class attribute. The other two items are __module__ and __qualname__ which the Python runtime has added.

The __module__ attribute is mapped to the name of the module in which the class was defined. Because we used the REPL, this is builtins in this case. The __qualname__ attribute contains the fully-qualified name of the class including parent modules and packages. In this case, it contains just the class name, as the builtins module used by the REPL — being the last namespace in the LEGB lookup hierarchy — is available everywhere, .

The kwargs dictionary passed to __new__() is also still empty.

Within __new__() we delegate to the base class type.__new__() via a call to super(), forwarding the mcs, name, bases, and namespace arguments. The object returned by this call *is* the new widget class object that we are in the process of allocating and configuring. The new widget class is what we return from __new__(). Note that any changes we wish to make to the contents of the namespace object must have been made *before* this call, as this is the point at which the class object is created. To change the contents of the class namespace after this call, the class object must be manipulated directly.

The __init__ method

Finally we come to __init__() the purpose of which is to *configure* the newly created class object. Note that __init__ here is an *instance method* of the metaclass, not an explicit class-method like __prepare__() or an implicit class-

method like __new__(). As such it accepts cls as its first argument, which is one level less- meta than mcs in the same way that self is one level less meta than cls. The name, bases, namespace and kwargs arguments are all as before. Again we delegate to the type base class via super() although __init__() doesn't return anything as it is expected to modify the existing class object that was handed to it. Note that although the namespace object is passed to __init__, its contents will already have been used upstream by __new__ when allocating the class object. Changes to namespace will be inneffectual at this juncture, and any changes to the class object must be made by manipulating cls directly.

Putting it all together

The key here is that metaclasses give us the opportunity to modify the dictionary of class attributes, which includes methods, before the class is instantiated. We even get opportunity to modify the list of base classes or produce an entirely different class if required, although such uses are rare. We'll look at some complex examples soon to make this clear.

You may be wondering which out of __prepare__(), __new__(), and __init__() you should override. If you don't override __prepare__(), the default implementation in type will produce a regular dictionary for the namespace object, so you only need to override it if you need the behaviour provided by another mapping type.

Usually, it will only be necessary to override either __new__() or __init__(), but of these two, only __new__() can make decisions before the new class is allocated. The distinction between __new__() and __init__() for metaclasses is exactly the same as it is for regular classes. Later we'll see that it might be wise to prefer configuration in __init__() rather than __new__() so that metaclasses are more composable.

Passing additional arguments to the metaclass

Earlier, we skipped over the fact that that __prepare__(), __new__(), and __init__() all sport a **kwargs parameter to accept arbitrary keyword arguments. These can be supplied in the parameter list when defining a class, and any keyword arguments provided over and above the a named metaclass argument will be forwarded to the three special methods.

Here's an example where we've passed a tension keyword argument in the parameter list of our Reticulator class:

```
>>> class Reticulator(metaclass=TracingMeta, tension=496):
... def reticulate(self, spline):
           print(spline)
      cubic = True
. . .
TracingMeta.__prepare__(name, bases, **kwargs)
mcs = <class 'tracing.TracingMeta'>
name = Reticulator
bases = ()
kwargs = {'tension': 496}
<-- namespace = \{\}
TracingMeta.__new__(mcs, name, bases, namespace)
mcs = <class 'tracing.TracingMeta'>
name = Reticulator
bases = ()
namespace = {'reticulate': <function Reticulator.reticulate at 0x103016510>, '__qualname_\
_': 'Reticulator', 'cubic': True, '__module__': 'builtins'}
kwargs = {'tension': 496}
<-- cls = <class 'Reticulator'>
TracingMeta.__init__(cls, name, bases, namespace)
cls = <class 'Reticulator'>
name = Reticulator
bases = ()
namespace = {'reticulate': <function Reticulator.reticulate at 0x103016510>, '__qualname_\
_': 'Reticulator', 'cubic': True, '__module__': 'builtins'}
kwargs = {'tension': 496}
```

As you can see, the arguments are dutifully forwarded to the metaclass methods, and the argument values could have been used to configure the class object. This allows the class statement to be used as a kind of class *factory*.

An example of metaclass keywords

Here's an interesting example which uses **kwargs. The __new__() method of the EntriesMeta metaclass expects kwargs to contain a num_entries key which maps to an integer value. This is used to populate the namespace with named entries using the letters of the alphabet. It looks like we only need to override __new__() to achieve our aims:

```
class EntriesMeta(type):

def __new__(mcs, name, bases, namespace, **kwargs):
    print("Entries.__new__(mcs, name, bases, namespace, **kwargs)")
    print(" kwargs =", kwargs)
    num_entries = kwargs['num_entries']
    print(" num_entries =", num_entries)
    namespace.update({chr(i): i for i in range(ord('a'), ord('a')+num_entries)})
    cls = super().__new__(mcs, name, bases, namespace)
    return cls
```

In __new__() we:

- Print kwargs
- Extract the num_entries value from kwargs
- Print num_entries
- Use a dictionary comprehension to generate a diction mapping letters to number values
- Update the namespace dictionary with our dictionary items
- Pass the modified namespace object on to type.__new__() via a call to super()

Let's try to use it:

```
>>> from entries import EntriesMeta
>>> class AtoZ(metaclass=EntriesMeta, num_entries=26):
...    pass
...
Entries.__new__(mcs, name, bases, namespace, **kwargs)
    kwargs = {'num_entries': 26}
    num_entries = 26
Traceback (most recent call last):
    File "<input>", line 1, in <module>
TypeError: type.__init__() takes no keyword arguments
```

The problem we've set up here is that both __init__() and __new__() must accept any additional arguments — they must have the same signature. We need to add a do-nothing __init__() to keep Python happy:

```
def __init__(cls, name, bases, namespace, **kwargs):
    super().__init__(name, bases, namespace)
```

With this in place, everything works as intended:

```
>>> from entries import EntriesMeta
>>> class AtoZ(metaclass=EntriesMeta, num_entries=26):
       pass
Entries.__prepare__(name, bases, **kwargs)
 mcs = <class 'entries.EntriesMeta'>
 name = AtoZ
 bases = ()
 kwargs = {'num_entries': 26}
<-- namespace = {}
Entries.__new__(mcs, name, bases, namespace, **kwargs)
 mcs = <class 'entries.EntriesMeta'>
 name = AtoZ
 bases = ()
 namespace = {'__module__': 'builtins', '__qualname__': 'AtoZ'}
 kwargs = {'num_entries': 26}
 num_entries = 26
```

```
<-- cls = <class 'AtoZ'>
Entries.__init__(cls, name, bases, namespace, **kwargs)
   cls = <class 'AtoZ'>
   name = AtoZ
   bases = ()
   namespace = {'__qualname__': 'AtoZ', 'j': 106, 'q': 113, 't': 116, 'v': 118, 'g': 103, \
'z': 122, '__module__': 'builtins', 'e': 101, 'k': 107, 'a': 97, 'b': 98, 'c': 99, 'y': 1\
21, 'l': 108, 'i': 105, 'n': 110, 's': 115, 'h': 104, 'm': 109, 'o': 111, 'p': 112, 'w': \
119, 'd': 100, 'r': 114, 'f': 102, 'u': 117, 'x': 120}
   kwargs = {'num_entries': 26}
```

We can see the num_entries item arrive in kwargs, and by the time we get to __init__() we can see the namespace object with additional entries.

We can use a regular argument as well as or instead of **kwargs. Let's convert num_entries to a proper argument:

```
class EntriesMeta(type):
```

```
def __new__(mcs, name, bases, namespace, num_entries, **kwargs):
    print("Entries.__new__(mcs, name, bases, namespace, **kwargs)")
    print(" kwargs =", kwargs)
    print(" num_entries =", num_entries)
    namespace.update({chr(i): i for i in range(ord('a'), ord('a')+num_entries)})
    cls = super().__new__(mcs, name, bases, namespace)
    return cls
```

This works just as expected:

```
>>> from entries import EntriesMeta
>>> class AtoZ(metaclass=EntriesMeta, num_entries=10):
Entries.__prepare__(name, bases, **kwargs)
 mcs = <class 'entries.EntriesMeta'>
 name = AtoZ
 bases = ()
 kwargs = {'num_entries': 10}
<-- namespace = {}
Entries.__new__(mcs, name, bases, namespace, **kwargs)
 mcs = <class 'entries.EntriesMeta'>
 name = AtoZ
 bases = ()
 namespace = {'__module__': 'builtins', '__qualname__': 'AtoZ'}
 kwargs = {}
 num entries = 10
<-- cls = <class 'AtoZ'>
Entries.__init__(cls, name, bases, namespace, **kwargs)
 cls = <class 'AtoZ'>
 name = AtoZ
 bases = ()
 namespace = {'j': 106, 'f': 102, '__module__': 'builtins', 'd': 100, 'b': 98, 'g': 103,\
 'c': 99, '__qualname__': 'AtoZ', 'i': 105, 'a': 97, 'e': 101, 'h': 104}
 kwargs = {'num_entries': 10}
```

Metaclass methods and visibility

We've covered three important special methods for metaclasses, __prepare__(), __new__() and __init__(), but what happens if we include other methods in the metaclass? Let's see, by adding a method called metamethod to the TracingMeta metaclass we were experimenting with earlier:

```
class TracingMeta(type):
     @classmethod
     def __prepare__(mcs, name, bases, **kwargs):
          print("TracingMeta.__prepare__(name, bases, **kwargs)")
          print("mcs =", mcs)
         print("name =", name)
print("bases =", bases)
print("kwargs =", kwargs)
          namespace = super().__prepare__(name, bases)
          print("<-- namespace =", namespace)</pre>
          print()
          return namespace
     def __new__(mcs, name, bases, namespace, **kwargs):
          print("TracingMeta.__new__(mcs, name, bases, namespace)")
          print("mcs =", mcs)
print("name =", name)
          print("bases =", bases)
          print("namespace =", namespace)
print("kwargs =", kwargs)
          cls = super().__new__(mcs, name, bases, namespace)
          print("<-- cls =", cls)</pre>
          print()
          return cls
     def __init__(cls, name, bases, namespace, **kwargs):
          print("TracingMeta.__init__(cls, name, bases, namespace)")
          print("cls =", cls)
         print( cls = , cls)
print("name =", name)
print("bases =", bases)
print("namespace =", namespace)
print("kwargs =", kwargs)
super().__init__(name, bases, namespace)
          print()
     def metamethod(cls):
          print("TracingMeta.metamethod(cls)")
          print("cls = ", cls)
          print()
```

Let's create the class Widget with its TracingMeta metaclass:

```
>>> from tracing import *
>>> class Widget(metaclass=TracingMeta):
...    pass
...
TracingMeta.__prepare__(name, bases, **kwargs)
mcs = <class 'tracing.TracingMeta'>
name = Widget
bases = ()
```

```
kwargs = {}
<-- namespace = {}

TracingMeta.__new__(mcs, name, bases, namespace)
mcs = <class 'tracing.TracingMeta'>
name = Widget
bases = ()
namespace = {'__qualname__': 'Widget', '__module__': 'builtins'}
kwargs = {}
<-- cls = <class 'Widget'>

TracingMeta.__init__(cls, name, bases, namespace)
cls = <class 'Widget'>
name = Widget
bases = ()
namespace = {'__qualname__': 'Widget', '__module__': 'builtins'}
kwargs = {}
```

It turns out that regular methods of the metaclass can be accessed similarly to *class* methods of the Widget class, and they will have the class passed to them as the first (and implicit) argument:

```
>>> Widget.metamethod()
TracingMeta.metamethod(cls)
cls = <class 'Widget'>
```

However, unlike regular class methods we create with the @classmethod decorator, we cannot access so-called metamethods via the widget instance:

```
>>> w = Widget()
>>> w.metamethod()
Traceback (most recent call last):
   File "<input>", line 1, in <module>
AttributeError: 'Widget' object has no attribute 'metamethod'
```

Metamethods are rarely used in practice, although one metamethod in particular has an interesting use: __call__(), which we'll look at shortly.

Class method definitions in the regular class and its base classes will take precedence over looking up a method in its metaclass.

Regular methods of the metaclass accept cls as their first argument. This makes sense because cls (the class) is the 'instance' of the metaclass; it is analogous to self. On the other hand, class- methods of the metaclass accept mcs as their first argument — the metaclass — analogous to the cls argument of a class method in an actual class.

Fine-grained instantiation control with metaclass __call__()

For a moment, return your thoughts to the material we covered in chapter 5 when we looked a *instance* allocation with __new___(). We know that in order to create instances we call the constructor of the desired class:

```
>>> w = Widget()
```

We have learned that behind the scenes this will call Widget.__new__() to allocate a Widget followed by Widget.__init__() to do any further initialisation. Let's pull back the curtain, and see exactly what *is* "behind the scenes".

The behaviour of calling __new__() followed by __init__() when we call a constructor is actually the responsibility of __call__() on the metaclass. This makes sense when we remember that __call__() is a metamethod and therefore can be called like a classmethod, and that __call__() makes the objects on which it is defined *callable*, like functions. This is the mechanism by which *classes* in Python become callable, and what we have been referring hitherto as a *constructor* call is in fact the __call__() metamethod.

Let's see this in action by overriding __call__() in our TracingMeta example:

class TracingMeta(type):

```
@classmethod
def __prepare__(mcs, name, bases, **kwargs):
     print("TracingMeta.__prepare__(name, bases, **kwargs)")
     print(" mcs =", mcs)
print(" name =", name)
print(" bases = ", bases)
print(" kwargs = ", kwargs)
     namespace = super().__prepare__(name, bases)
     print("<-- namespace =", namespace)</pre>
     return namespace
def __new__(mcs, name, bases, namespace, **kwargs):
     print("TracingMeta.__new__(mcs, name, bases, namespace)")
print(" mcs =", mcs)
print(" name =", name)
print(" bases =", bases)
     print(" namespace =", namespace)
print(" kwargs =", kwargs)
     cls = super().__new__(mcs, name, bases, namespace)
     print("<-- cls =", cls)</pre>
     print()
     return cls
def __init__(cls, name, bases, namespace, **kwargs):
     print("TracingMeta.__init__(cls, name, bases, namespace)")
     print(" cls =", cls)
print(" name =", name)
print(" bases =", bases)
```

```
print(" namespace =", namespace)
    print(" kwargs =", kwargs)
    super().__init__(name, bases, namespace)
    print()
def metamethod(cls):
    print("TracingMeta.metamethod(cls)")
print(" cls = ", cls)
    print()
def __call__(cls, *args, **kwargs):
    print("TracingMeta.__call__(cls, *args, **kwargs)")
    print(" cls =", cls)
print(" args =", args)
    print(" kwargs =", kwargs)
    print(" About to call type.__call__()")
    obj = super().__call__(*args, **kwargs)
    print(" Returned from type.__call__()")
    print("<-- obj =", obj)</pre>
    print()
    return obj
```

We'll also implement a TracingClass which will use TracingMeta as its metaclass. TracingClass overrides __new__() and __init__() so we can see when they're called:

```
class TracingClass(metaclass=TracingMeta):

    def __new__(cls, *args, **kwargs):
        print(" TracingClass.__new__(cls, args, kwargs")
        print(" cls =", cls)
        print(" args =", args)
        print(" kwargs =", kwargs)
        obj = super().__new__(cls)
        print(" <-- obj =", obj)
        print()
        return obj

def __init__(self, *args, **kwargs):
        print(" TracingClass.__init__(self, *args, **kwargs")
        print(" self =", self)
        print(" args =", args)
        print(" kwargs =", kwargs)
        print()</pre>
```

Notice that when we import the module into the REPL, the metaclass trifecta __prepare__(), __new__(), and __init__() are invoked when the TracingClass is defined:

```
>>> from tracing import *
TracingMeta.__prepare__(name, bases, **kwargs)
   mcs = <class 'tracing.TracingMeta'>
   name = TracingClass
   bases = ()
   kwargs = {}
<-- namespace = {}
TracingMeta.__new__(mcs, name, bases, namespace)</pre>
```

```
mcs = <class 'tracing.TracingMeta'>
 name = TracingClass
 bases = ()
 namespace = {'__new__': <function TracingClass.__new__ at 0x103016488>, '__init__': <fu\</pre>
nction TracingClass.__init__ at 0x103016510>, '__qualname__': 'TracingClass', '__module__\
': 'tracing'}
 kwargs = \{\}
<-- cls = <class 'tracing.TracingClass'>
TracingMeta.__init__(cls, name, bases, namespace)
 cls = <class 'tracing.TracingClass'>
 name = TracingClass
 bases = ()
 namespace = {'__new__': <function TracingClass.__new__ at 0x103016488>, '__init__': <fu\</pre>
nction TracingClass.__init__ at 0x103016510>, '__qualname__': 'TracingClass', '__module__\
': 'tracing'}
 kwargs = \{\}
```

Now we'll instantiate TracingClass with a positional argument and a keyword argument:

```
>>> t = TracingClass(42, keyword="clef")
TracingMeta.__call__(cls, *args, **kwargs)
 cls = <class 'tracing.TracingClass'>
 args = (42,)
 kwargs = {'keyword': 'clef'}
 About to call type.__call__()
TracingClass.__new__(cls, args, kwargs
 cls = <class 'tracing.TracingClass'>
 args = (<class 'tracing.TracingClass'>, 42)
 kwargs = {'keyword': 'clef'}
<-- obj = <tracing.TracingClass object at 0x103012ef0>
TracingClass.__init__(self, *args, **kwargs
 self = <tracing.TracingClass object at 0x103012ef0>
  args = (<class 'tracing.TracingClass'>, 42)
  kwargs = {'keyword': 'clef'}
 Returned from type.__call__()
<-- obj = <tracing.TracingClass object at 0x103012ef0>
```

Look carefully at the control flow here. Our call to the constructor invokes __call__() on the metaclass, which receives the arguments we passed to the constructor in addition to the type we're trying to construct.

Our __call__() override calls the superclass implementation, which is type.__call__() in this case. See how type.__call__() in turn calls TracingClass.__new__() followed by TracingClass.__init__(). In other words, it is type.__call__() which orchestrates the default class allocation and initialisation behaviour.

It's very rare to see the __call__() metamethod overridden. It's pretty low level in Python terms and provides some of the most basic Python machinery. That

said, it can be powerful.

An example of overriding __call__ on a metaclass

In keywordmeta.py we have an example of a metaclass overriding __call__() to prevent classes which use the metaclass accepting positional arguments to their constructors:

```
class KeywordsOnlyMeta(type):
    def __call__(cls, *args, **kwargs):
        if args:
            raise TypeError("Constructor for class {!r} does not accept positional argume\
nts.".format(cls))
        return super().__call__(cls, **kwargs)

class ConstrainedToKeywords(metaclass=KeywordsOnlyMeta):
    def __init__(self, *args, **kwargs):
        print("args =", args)
        print("kwargs =", kwargs)
```

Even though the __init__() method in ConstrainedToKeywords accepts positional arguments through *args, execution never gets this far as non-empty positional argument lists are intercepted by __call__() in the KeywordsOnlyMeta metaclass, which causes a TypeError to be raised:

```
>>> from keywordmeta import *
>>>
c = ConstrainedToKeywords(23, 45, 96, color='white')
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "/Users/rjs/training/tmp/metaclass/keywordmeta.py", line 6, in __call__
        raise TypeError("Constructor for class {!r} does not accept positional arguments.".fo\
rmat(cls))
TypeError: Constructor for class <class 'keywordmeta.ConstrainedToKeywords'> does not acc\
ept positional arguments.
```

Constructor calls which contain only keyword arguments are permitted as designed:

```
>>> c = ConstrainedToKeywords(color='white')
args = (<class 'keywordmeta.ConstrainedToKeywords'>,)
kwargs = {'color': 'white'}
```

We've covered a lot of the theory and practice behind metaclasses. Now we'll build on those ideas with some useful applications.

Practical metaclass examples

Python metaclasses can seem very...well...meta! So it's time for a section on metaclasses which solve some actual problems you've probably encountered.

Preventing duplicate class attributes

At some point, you've probably run into the issue that duplicate class attributes names aren't flagged by Python as errors:

```
class Dodgy:
    def method(self):
        return "first definition"

def method(self):
    return "second definition"
```

In fact, the second definition takes precedence because it overwrites the first entry in the namespace dictionary as the class definition is processed:

```
>>> from duplicatesmeta import *
>>> dodgy = Dodgy()
>>> dodgy.method()
'second definition
```

Let's write a metaclass which detects and prevents this unfortunate situation from occurring. To do this, rather than using a regular dictionary as the namespace object used during class construction, we need a dictionary which raises an error when we try to assign to an existing key.

Here is just such a dictionary, OneShotDict, which is implemented by specialising the built-in dict type and overriding the __init__() and __setitem__() methods. Note that the built-in dict has quite a sophisticated initializer which accepts many different forms of arguments, but something much simpler is sufficient in our case:

Before we use the OneShotDict in a metaclass, let's just check that it's working as expected:

```
>>> d = OneShotDict()
>>> d['A'] = 65
>>> d['B'] = 66
>>> d['A'] = 32
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "/Users/sixtynorth/sandbox/metaclasses/orderedmeta.py", line 14, in __setitem__
        raise KeyError("Cannot assign to existing key {!r} in {!r}".format(key, type(self).__\
name__))
KeyError: Cannot assign to existing key 'A' in 'OneShotDict'
```

We can now design a very simple metaclass which uses OneShotDict for the namespace object:

```
class ProhibitDuplicatesMeta(type):
    @classmethod
    def __prepare__(mcs, name, bases):
        return OneShotDict()
```

All we need to do is override the __prepare__() classmethod and returning an instance of our specialised dictionary.

If we try to define a class with duplicate methods using this metaclass, we get an error:

The main shortcoming here is that the error message isn't hugely informative. Unfortunately, we don't have access to the part of the runtime machinery which reads our class definition and populates the dictionary, so we can't intercept the KeyError and emit a more useful error instead. The best we can do — rather than using a general purpose collection like OneShotDict — is to create a functional equivalent called something like OneShotClassNamespace with a more specific error message. This has the benefit that we can pass in additional diagnostic information, such as the name of the class currently being defined, into the namespace object on construction, which helps us emit a more useful message:

```
class OneShotClassNamespace(dict):
   def __init__(self, name, existing=None):
       super().__init__()
        self._name = name
        if existing is not None:
            for k, v in existing:
               self[k] = v
   def __setitem__(self, key, value):
        if key in self:
            raise TypeError("Cannot reassign existing class "
                             "attribute {!r} of {!r}".format(key, self._name))
        super().__setitem__(key, value)
class ProhibitDuplicatesMeta(type):
   @classmethod
   def __prepare__(mcs, name, bases):
       return OneShotClassNamespace(name)
```

After renaming OneShotDict to OneShotClassNamespace we adjust its initializer to accept a positional name argument which we store as an instance attribute _name. In the guard clause of __setitem__ we change the exception type from ValueError to TypeError and edit the error message to make it both more germane and more informative. Lastly, we need to remember to forward the class name which is passed to __prepare__() to the OneShotClassNamespace constructor.

When we try to execute the module containing the class with the duplicate method definition, we get a much more useful error message "Cannot reassign existing class attribute 'method' of 'Dodgy'":

```
>>> from duplicatesmeta import *
Traceback (most recent call last):
   File "/Users/sixtynorth/sandbox/metaclasses/duplicatesmeta.py", line 38, in <module>
        class Dodgy(metaclass=ProhibitDuplicatesMeta):
   File "/Users/sixtynorth/sandbox/metaclasses/duplicatesmeta.py", line 43, in Dodgy
        def method(self):
   File "/Users/sixtynorth/sandbox/metaclasses/duplicatesmeta.py", line 27, in __setitem__
        raise TypeError("Cannot reassign existing class attribute {!r} of {!r}".format(key, s\
elf._name))
TypeError: Cannot reassign existing class attribute 'method' of 'Dodgy'
```

Much better!

Naming Descriptors using Metaclasses

For our next metaclass example, let's return to the planet example we used in <u>chapter 4</u> to illustrate descriptors. Here's a reminder of the code we ended up

with:

```
from weakref import WeakKeyDictionary
class Positive:
    def __init__(self):
        self._instance_data = WeakKeyDictionary()
    def __get__(self, instance, owner):
        if instance is None:
            return self
        return self._instance_data[instance]
    def __set__(self, instance, value):
        if value <= 0:</pre>
            raise ValueError("Value {} is not positive".format(value))
        self._instance_data[instance] = value
    def __delete__(self, instance):
        raise AttributeError("Cannot delete attribute")
class Planet:
    def __init__(self,
                 name,
                 radius_metres,
                 mass_kilograms,
                 orbital_period_seconds,
                 surface_temperature_kelvin):
        self.name = name
        self.radius metres = radius metres
        self.mass_kilograms = mass_kilograms
        self.orbital_period_seconds = orbital_period_seconds
        self.surface_temperature_kelvin = surface_temperature_kelvin
    @property
    def name(self):
        return self._name
    @name.setter
    def name(self, value):
        if not value:
            raise ValueError("Cannot set empty Planet.name")
        self._name = value
    radius_metres = Positive()
    mass kilograms = Positive()
    orbital_period_seconds = Positive()
    surface_temperature_kelvin = Positive()
```

Recall that we implemented a new descriptor type called Positive which would only admit positive numeric values. This saved a lot of boilerplate code in the definition of our Planet class, but we lost an important capability along the way, because there is no way for a descriptor instance to know to which class attribute it has been bound. One of the instances of Positive is bound to

Planet.radius_metres, but it has no way of knowing that. The default Python machinery for processing class definitions just doesn't set up that association.

The shortcoming is revealed when we trigger a ValueError by trying to assign a non-positive value to one of the attributes. Here we try to give the planet Mercury a nonsensical negative mass:

```
>>> from planet import *
>>> mercury, venus, earth, mars = make_planets()
>>> mercury.mass_kilograms = -10000
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "/Users/rjs/training/tmp/metaclass/planet.py", line 16, in __set__
        raise ValueError("Value {} is not positive".format(value))
ValueError: Value -10000 is not positive
```

The error message doesn't — and in fact *can't* — tell us which attribute triggered the exception.

Now we'll show how we can modify the class creation machinery by defining metaclasses which should be able to intervene in the process of defining the Planet class in order to give descriptor instances the right name.

We'll start by introducing a new base-class for our descriptors called Named. This is very simple and just has name as a public instance attribute. The constructor defines a default value of None because we won't be in a position to assign the attribute value until after the descriptor object has been constructed:

```
class Named:
    def __init__(self, name=None):
        self.name = name
```

We'll modify our existing Positive descriptor so it becomes a subclass of Named and therefore gains the name attribute. Again the constructor arguments defines a default of None:

```
class Positive(Named):
    def __init__(self, name=None):
        super().__init__(name)
        self._instance_data = WeakKeyDictionary()

def __get__(self, instance, owner):
    if instance is None:
        return self
    return self._instance_data[instance]

def __set__(self, instance, value):
```

We've modfied the argument list to __init__(), ensured that the superclass initialiser is called, and made use of the new name attribute in the error message raised by __set__() when we try to assign a non-positive number to the descriptor.

Now we need a metaclass which can detect the presence of descriptors which are Named and assign class attribute names to them:

```
class DescriptorNamingMeta(type):

    def __new__(mcs, name, bases, namespace):
        for name, attr in namespace.items():
            if isinstance(attr, Named):
                attr.name = name
        return super().__new__(mcs, name, bases, namespace)
```

Again, this is fairly straightforward. In __new__ we iterate over the names and attributes in the namespace dictionary, and if the attribute is an instance of Named we assign the name of the current item to its public name attribute.

Having modified the contents of the namespace, we then call the superclass implementation of __new__() to actually allocate the new class object.

The only change we need to make to our Planet class is to refer to the metaclass on the opening line. There's not need for us to modify our uses of the Positive descriptor, the optional name argument will default to None when the class definition is read, before metaclass __new__() is invoked:

```
def name(self):
    return self._name

@name.setter
def name(self, value):
    if not value:
        raise ValueError("Cannot set empty Planet.name")
    self._name = value

radius_metres = Positive()
mass_kilograms = Positive()
orbital_period_seconds = Positive()
surface_temperature_kelvin = Positive()
```

By trying to set a non-positive mass for the planet Mercury, we can see that each descriptor objects now knows the name of the attribute to which it has been bound, so can it emit *much* more helpful diagnostic message:

```
>>> from planet import *
>>> mercury, venus, earth, mars = make_planets()
>>> mercury.mass_kilograms
3.3022e+23
>>> mercury.mass_kilograms = -10000
Traceback (most recent call last):
  File "<input>", line 1, in <module>
   File "/Users/sixtynorth/sandbox/metaclasses/planet_08.py", line 23, in __set__
        raise ValueError("Attribute value {} {} is not positive".format(self.name, value))
ValueError: Attribute value mass_kilograms -10000 is not positive
```

Metaclasses and Inheritance

We'll finish off this module by looking at how metaclasses interact with inheritance.

In metainheritance.py let's define two metaclasses related only by the fact that they both subclass type:

```
# metainheritance.py
class MetaA(type):
    pass
class MetaB(type):
    pass
```

We'll also define two regular classes, A and B which use the MetaA and MetaB as their respective metaclasses:

```
class A(metaclass=MetaA):
    pass
```

```
class B(metaclass=MetaB):
    pass
```

Now we'll introduce a third regular class D which derives from class A:

```
class D(A):
    pass
```

Now, let's introspect the *class* D itself — not an instance of D — to determine what its metaclass is:

```
>>> type(D)
<class 'metainheritance.MetaA'>
```

The metaclass of class D is MetaA, which was inherited from regular class A. So, metaclasses are inherited.

What happens if we try to create a new class C which inherits from both regular classes A and B with their different metaclasses? Let's give it a go:

```
class C(A, B):
    pass
```

When we try to execute this code by importing it at the REPL, we get a type error:

```
>>> from metainheritance import *
Traceback (most recent call last):
   File "<input>", line 1, in <module>
    File "/Applications/PyCharm.app/Contents/helpers/pydev/pydev_import_hook.py", line 21, \
in do_import
    module = self._system_import(name, *args, **kwargs)
   File "/Users/rjs/training/p4/courses/pluralsight/advanced-python/source/advanced-python\
-m06-metaclasses/examples/metainheritance.py", line 22, in <module>
        class C(A, B):
TypeError: metaclass conflict: the metaclass of a derived class must be a (non-strict) su\
bclass of the metaclasses of all its bases
```

With the message "metaclass conflict: the metaclass of a derived class must be a (non-strict) subclass of the metaclasses of all its bases" Python is telling us that it doesn't know what to do with the unrelated metaclasses. Which metaclass __new__() should be used to allocate the class object?

To resolve this we need a single metaclass — let's call it MetaC — which we can create by inheriting from both MetaA and MetaB:

```
class MetaC(MetaA, MetaB):
    pass
```

In the definition of C we must override the metaclass to specify MetaC:

```
class C(A, B, metaclass=MetaC):
   pass
```

With these changes we can successfully import C and check that it's metaclass is MetaC:

```
>>> from metainheritance import *
>>> type(C)
<class 'metainheritance.MetaC'>
```

So we've persuaded Python to accept our code, but our metaclasses are empty so they combine trivially.

Sometimes, metaclasses will combine in straightforward ways. For example, our ProhibitDuplicatesMeta which overrides only __prepare__(), and our KeyWordsOnlyMeta which overrides only __call__(), can be combined into the conceptually simple (but horribly named) ProhibitDuplicatesAndKeyWordsOnlyMeta:

```
class ProhibitDuplicatesAndKeyWordsOnlyMeta(
    ProhibitDuplicatesMeta,
    KeyWordsOnlyMeta):
nass
```

Designing metaclasses for composition

To cooperate gracefully, non-trivial metaclasses must be designed with this in mind. This isn't always straightforward — or even possible — if the combination makes no sense. An important step in designing cooperative metaclasses is to diligently use super() when delegating to "base" classes, because as we learnt in chapter 8 of The Python Journeyman super() actually delegates to the next class in the Method Resolution Order (a.k.a. MRO) which accounts for multiple inheritance.

Even though both TracingMeta (defined earlier in this chapter) and DescriptorNamingMeta (which we used with the planet attributes) both override __new__(), they combine in either order because our implementations delegate to the next class in the MRO chain using calls to super(). In planet.py we can import TracingMeta and combine our two metaclasses into TracingDescriptorNamingMeta with multiple inheritance:

```
class TracingDescriptorNamingMeta(TracingMeta, DescriptorNamingMeta):
    pass
```

The Planet class definition is executed when we import the module. This is when the metaclasses to their work, and we can see that the tracing works as expected:

```
>>> from planet import *
TracingMeta.__prepare__(name, bases, **kwargs)
  mcs = <class 'tracing.TracingMeta'>
  name = TracingClass
  bases = ()
  kwargs = \{\}
<-- namespace = \{\}
TracingMeta.__new__(mcs, name, bases, namespace)
  mcs = <class 'tracing.TracingMeta'>
  name = TracingClass
  bases = ()
namespace = {'__module__': 'tracing', '__qualname__': 'TracingClass', '__init__': <func\tion TracingClass.__init__ at 0x1032c89d8>, '__new__': <function TracingClass.__new__ at \
0x1032c8950>}
 kwargs = \{\}
<-- cls = <class 'tracing.TracingClass'>
TracingMeta.__init__(cls, name, bases, namespace)
  cls = <class 'tracing.TracingClass'>
  name = TracingClass
  bases = ()
  namespace = {'__module__': 'tracing', '__qualname__': 'TracingClass', '__init__': <func\</pre>
tion TracingClass.__init__ at 0x1032c89d8>, '__new__': <function TracingClass.__new__ at \
0x1032c8950>}
  kwargs = \{\}
```

We can also confirm that the descriptor naming is working by causing a ValueError and checking for the descriptor name in the error message:

```
>>> mercury.mass_kilograms = -10000
Traceback (most recent call last):
    File "<input>", line 1, in <module>
NameError: name 'mercury' is not defined
>>> mercury, venus, earth, mars = make_planets()
>>> mercury.mass_kilograms = -10000
Traceback (most recent call last):
    File "<input>", line 1, in <module>
    File "/Users/rjs/training/planet_09.py", line 24, in __set__
        raise ValueError("Attribute value {} {} is not positive".format(self.name, value))
ValueError: Attribute value mass_kilograms -10000 is not positive
```

The two metaclasses cooperate successfully by using super().

```
Prefer __init__ to __new__
```

Another tip for designing metaclasses which cooperate well is to prefer to put code which *configures* the class object in __init__() (which is handed to the

class object to be configured) rather than in __new__() (which is responsible for allocating the new class object).

If any of your cooperative metaclasses expect custom keyword arguments passed from the class statement, you'll need to ensure that these are forwarded through the MRO chain. You also need to ensure that they are consumed, if necessary, so that subsequent classes in the MRO don't receive unexpected arguments. All custom keyword arguments must have been consumed by the time type. __new__() is called, as it expects no additional arguments.

Ultimately though, if you're dependent on metaclasses from a third-party framework such as *SQLAlchemy* or *Qt*, there's a good chance that their metaclasses won't compose gracefully. You should be able to figure out whether they will compose by reading their source code, if you have access to it.

Summary

We've covered a lot of ground in this chapter, and you should now know more than the majority of Python developers about the customisation of class creation using metaclasses.

- All classes have a metaclass which is the type of the class object. The default type of class objects is type.
- The metaclass is responsible for processing the class definition from parsing from the source code into a class object.
- The __prepare__() metaclass method must return a mapping object which the Python runtime will populate with namespace items collected from parsing and executing the class definition.
- The __new__() metaclass method must allocate and return a class object and configure it using the contents of the class namespace, the list of base classes passed from the definition, and any additional keyword arguments passed in the definition.
- The __init__() metaclass method can be used to further configure a class object, and must have the same signature as __new__().
- The __call__() metaclass method in effect implements the *constructor* for class instances, and is invoked when we construct an instance.
- An important use case for metaclasses is to support so-called named descriptors, whereby we can configure descriptor objects such as properties with the name of the class attribute to which they are assigned.

• Strict rules control how muliple metaclasses interact in the presence of inheritance relationships between the regular classes which use them. Judicial metaclass design using super() to delegate via the MRO can yield metaclasses which compose gracefully.

```
<code></code><code class="nd">@my_class_decorator</code>
<code class="k">class</code> <code class="nc">Temperature</code> <code
class="p">:</code>
  <code class="k">def</code> <code class="fm">__init__</code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">kelvin</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n">_kelvin</code> <code class="o">=</code> <code
class="n">kelvin</code>
  <code class="k">def</code> <code class="nf">get kelvin</code> <code
class="p">(</code><code class="bp">self</code><code class="p">):</code>
  <code class="k">return</code> <code class="bp">self</code><code
class="o">.</code><code class="n"> kelvin</code>
  <code class="k">def</code> <code class="nf">set kelvin</code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">value</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> kelvin</code> <code class="o">=</code> <code
class="n">value</code>
<code></code><code class="k">def</code> <code
class="nf">my class decorator</code><code class="p">(</code><code
class="bp">cls</code><code class="p">):</code>
  <code class="k">for</code> <code class="n">name</code> <code
class="p">,</code> <code class="n">attr</code> <code class="ow">in</code>
```

```
<code class="nb">vars</code><code class="p">(</code><code
class="bp">cls</code><code class="p">)</code><code class="o">.</code>
<code class="n">items</code><code class="p">():</code>
  <code class="k">print</code><code class="p">(</code><code
class="n">name</code><code class="p">)</code>
  <code class="k">return</code> <code class="bp">cls</code>
<code></code><code class="gp">>>> </code><code class="kn">from</code>
<code class="nn">class decorators</code> <code class="kn">import</code>
<code class="o">*</code>
<code class="go">__weakref__</code>
<code class="go">get_kelvin</code>
<code class="go">set kelvin</code>
<code class="go"> dict </code>
<code class="go">__module__</code>
<code class="go">__init__</code>
<code></code><code class="k">def</code> <code class="nf">invariant</code>
<code class="p">(</code><code class="n">predicate</code><code class="p">):
</code>
  <code class="sd">"""Create a class decorator which checks a class invariant.
</code>
<code class="sd"> Args:</code>
<code class="sd"> predicate: A callable to which, after every method invocation,
</code>
<code class="sd"> the object on which the method was called will be passed.
</code>
```

```
<code class="sd"> The predicate should evaluate to True if the class
invariant</code>
<code class="sd"> has been maintained, or False if it has been violated.</code>
<code class="sd"> Returns:</code>
<code class="sd"> A class decorator for checking the class invariant tested
by</code>
<code class="sd"> the supplied predicate function.</code>
<code class="sd"> """"</code>
  <code class="k">def</code> <code
class="nf">invariant checking class decorator</code><code class="p">
(</code><code class="bp">cls</code><code class="p">):</code>
  <code class="sd">"""A class decorator for checking invariants."""</code>
  <code class="n">method names</code> <code class="o">=</code> <code
class="p">[</code><code class="n">name</code> <code class="k">for</code>
<code class="n">name</code><code class="p">,</code> <code
class="n">attr</code> <code class="ow">in</code> <code
class="nb">vars</code><code class="p">(</code><code
class="bp">cls</code><code class="p">)</code><code class="o">.</code>
<code class="n">items</code><code class="p">()</code> <code
class="k">if</code> <code class="nb">callable</code><code class="p">
(</code><code class="n">attr</code><code class="p">)]</code>
  <code class="k">for</code> <code class="n">name</code> <code
class="ow">in</code> <code class="n">method names</code> <code
class="p">:</code>
  <code class="n">_wrap_method_with_invariant_checking_proxy</code>
<code class="p">(</code><code class="bp">cls</code><code class="p">,
```

```
</code> <code class="n">name</code> <code class="p">,</code> <code
class="n">predicate</code><code class="p">)</code>
  <code class="k">return</code> <code class="bp">cls</code>
  <code class="k">return</code> <code
class="n">invariant checking class decorator</code>
<code class="k">def</code> <code
class="nf">_wrap_method_with_invariant_checking_proxy</code><code
class="p">(</code><code class="bp">cls</code><code class="p">,</code>
<code class="n">name</code><code class="p">,</code> <code
class="n">predicate</code><code class="p">):</code>
  <code class="n">method</code> <code class="o">=</code> <code
class="nb">getattr</code><code class="p">(</code><code
class="bp">cls</code><code class="p">,</code> <code
class="n">name</code><code class="p">)</code>
  <code class="k">assert</code> <code class="nb">callable</code><code
class="p">(</code><code class="n">method</code><code class="p">)</code>
  <code class="nd">@functools.wraps</code><code class="p">(</code><code
class="n">method</code><code class="p">)</code>
  <code class="k">def</code> <code
class="nf">invariant_checking_method_decorator</code><code class="p">
(</code><code class="bp">self</code><code class="p">,</code> <code
class="o">*</code><code class="n">args</code><code class="p">,</code>
<code class="o">**</code><code class="n">kwargs</code><code class="p">):
</code>
```

```
<code class="n">result</code> <code class="o">=</code> <code
class="n">method</code><code class="p">(</code><code
class="bp">self</code><code class="p">,</code> <code class="o">*</code>
<code class="n">args</code><code class="p">,</code> <code class="o">**
</code><code class="n">kwargs</code><code class="p">)</code>
  <code class="k">if</code> <code class="ow">not</code> <code
class="n">predicate</code><code class="p">(</code><code
class="bp">self</code><code class="p">):</code>
  <code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="n">predicate</code><code class="o">.
</code>\
<code class="vm">__doc__</code><code class="p">,</code> <code
class="bp">self</code><code class="p">))</code>
  <code class="k">return</code> <code class="n">result</code>
  <code class="nb">setattr</code><code class="p">(</code><code
class="bp">cls</code><code class="p">,</code> <code
class="n">name</code><code class="p">,</code> <code
class="n">invariant_checking_method_decorator</code><code class="p">)
</code>
<code></code> <code class="k">def</code> <code
class="nf">not below absolute zero</code><code class="p">(</code><code
class="n">temperature</code><code class="p">):</code>
  <code class="sd">"""Temperature not below absolute zero"""</code>
  <code class="k">return</code> <code class="n">temperature</code><code
class="o">.</code><code class="n"> kelvin</code> <code class="o">>=
</code> <code class="mi">0</code>
```

```
<code class="nd">@invariant</code><code class="p">(</code><code
class="n">not_below_absolute_zero</code><code class="p">)</code>
<code class="k">class</code> <code class="nc">Temperature</code> <code
class="p">:</code>
  <code class="k">def</code> <code class="fm"> init </code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">kelvin</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> kelvin</code> <code class="o">=</code> <code
class="n">kelvin</code>
  <code class="k">def</code> <code class="nf">get kelvin</code> <code
class="p">(</code><code class="bp">self</code><code class="p">):</code>
  <code class="k">return</code> <code class="bp">self</code><code
class="o">.</code><code class="n"> kelvin</code>
  <code class="k">def</code> <code class="nf">set_kelvin</code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">value</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> kelvin</code> <code class="o">=</code> <code
class="n">value</code>
<code></code><code class="gp">>>> </code><code class="kn">from</code>
<code class="nn">class_decorators</code> <code class="kn">import</code>
<code class="o">*</code>
<code class="gp">>>> </code><code class="n">t</code> <code class="o">=
</code> <code class="n">Temperature</code> <code class="p">(</code> <code
```

```
class="mf">5.0</code><code class="p">)</code>
<code></code><code class="gp">>>> </code><code class="n">t</code> <code
class="o">=</code> <code class="n">Temperature</code> <code class="p">
(</code><code class="o">-</code><code class="mf">1.0</code><code
class="p">)</code>
<code class="gt">Traceback (most recent call last):</code>
  File <code class="nb">"<input>"</code>, line <code class="m">1</code>, in
<code class="n"><module></code>
  File <code class="nb">"class decorators.py"</code>, line <code
class="m">47</code>, in <code
class="n">invariant_checking_method_decorator</code>
  <code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="n">predicate</code><code class="o">.
</code><code class="vm">__doc__</code><code class="p">,</code>\
<code class="gr"> self))</code>
<code class="gr">RuntimeError</code>: <code class="n">Class invariant
'Temperature not below absolute zero' violated for <class_d\</code>
<code class="go">ecorators.Temperature object at 0x103012940></code>
<code></code><code class="gp">>>> </code><code class="n">s</code>
<code class="o">=</code> <code class="n">Temperature</code> <code
class="p">(</code><code class="mf">42.0</code><code class="p">)</code>
<code class="gp">>>> </code><code class="n">s</code><code class="o">.
</code><code class="n">set_kelvin</code><code class="p">(</code><code
class="o">-</code><code class="mf">1.0</code><code class="p">)</code>
<code class="gt">Traceback (most recent call last):</code>
  File <code class="nb">"<input>"</code>, line <code class="m">1</code>, in
```

```
<code class="n"><module></code>
  File <code class="nb">"class decorators.py"</code>, line <code
class="m">47</code>, in <code
class="n">invariant checking method decorator</code>
  <code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="n">predicate</code><code class="o">.
</code><code class="vm"> doc </code><code class="p">,</code>\
<code class="gr"> self))</code>
<code class="gr">RuntimeError</code>: <code class="n">Class invariant
'Temperature not below absolute zero.' violated for <class \</code>
<code class="go">decorators.Temperature object at 0x10302ca20></code>
<code></code><code class="nd">@invariant</code><code class="p">(</code>
<code class="n">not below absolute zero</code><code class="p">)</code>
<code class="k">class</code> <code class="nc">Temperature</code> <code
class="p">:</code>
  <code class="k">def</code> <code class="fm"> init </code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">kelvin</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
```

<code class="k">def</code> <code class="nf">get_kelvin</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">):</code>

class="n"> kelvin</code> <code class="o">=</code> <code

class="n">kelvin</code>

```
<code class="k">return</code> <code class="bp">self</code><code
class="o">.</code> <code class="n">_kelvin</code>
```

<code class="k">def</code> <code class="nf">set_kelvin</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">value</code> <code class="p">):</code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_kelvin</code> <code class="o">=</code> <code
class="n">value</code>

<code class="nd">@property</code>

<code class="k">def</code> <code class="nf">celsius</code> <code class="p">(</code> <code class="bp">self</code> <code class="p">):</code>

<code class="k">return</code> <code class="bp">self</code><code
class="o">.</code> <code class="n">_kelvin</code> <code class="o">-</code>
<code class="mf">273.15</code>

<code class="nd">@celsius.setter</code>

<code class="k">def</code> <code class="nf">celsius</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">value</code> <code class="p">):</code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_kelvin</code> <code class="o">=</code> <code
class="n">value</code> <code class="o">+</code> <code
class="mf">273.15</code>

<code class="nd">@property</code>

```
<code class="k">def</code> <code class="nf">fahrenheit</code> <code
class="p">(</code><code class="bp">self</code><code class="p">):</code>
  <code class="k">return</code> <code class="bp">self</code><code
class="o">.</code><code class="n"> kelvin</code> <code class="o">*</code>
<code class="mi">9</code><code class="o">/</code><code
class="mi">5</code> <code class="o">-</code> <code
class="mf">459.67</code>
  <code class="nd">@fahrenheit.setter</code>
  <code class="k">def</code> <code class="nf">fahrenheit</code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">value</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> kelvin</code> <code class="o">=</code> <code class="p">
(</code><code class="n">value</code> <code class="o">+</code> <code
class="mf">459.67</code><code class="p">)</code> <code class="o">*
</code> <code class="mi">5</code> <code class="o">/</code> <code
class="mi">9</code>
<code></code><code class="gp">>>> </code><code class="kn">from</code>
<code class="nn">class_decorators</code> <code class="kn">import</code>
<code class="o">*</code>
<code class="gp">>>> </code><code class="n">t</code> <code class="o">=
</code> <code class="n">Temperature</code> <code class="p">(</code> <code
class="mf">42.0</code><code class="p">)</code>
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">celsius</code>
<code class="go">-231.14999999999998</code>
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">celsius</code> <code class="o">=</code> <code
class="o">-</code><code class="mi">100</code>
```

```
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">celsius</code>
<code class="go">-100.0</code>
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">celsius</code> <code class="o">=</code> <code
class="o">-</code><code class="mi">300</code>
<code></code><code class="gp">>>> </code><code class="kn">from</code>
<code class="nn">class decorator</code> <code class="kn">import</code>
<code class="o">*</code>
<code class="gp">>>> </code><code class="kn">from</code> <code
class="nn">pprint</code> <code class="kn">import</code> <code
class="n">pprint</code> <code class="k">as</code> <code
class="n">pp</code>
<code class="gp">>>> </code><code class="n">pp</code><code class="p">
(</code><code class="nb">vars</code><code class="p">(</code><code
class="n">Temperature</code><code class="p">))</code>
<code class="go">mappingproxy({'__dict__': <attribute '__dict__' of</pre>
'Temperature' objects>,</code>
<code class="go"> ' doc ': None,</code>
<code class="go"> '__init__': <function Temperature.__init__ at 0x1031bf598>,
</code>
<code class="go"> '__module__': 'class_decorators',</code>
<code class="go"> '__weakref__': <attribute '__weakref__' of 'Temperature'</pre>
objects>,</code>
<code class="go"> 'celsius': <property object at 0x1031bb138>,</code>
<code class="go"> 'fahrenheit': <property object at 0x1031bb188>,</code>
<code class="go"> 'get kelvin': <function Temperature.get kelvin at</pre>
```

```
0x1031bf158>,</code>
<code></code><code class="k">def</code> <code class="nf">invariant</code>
<code class="p">(</code><code class="n">predicate</code><code class="p">):
</code>
  <code class="sd">"""Create a class decorator which checks a class invariant.
</code>
<code class="sd"> Args:</code>
<code class="sd"> predicate: A callable to which, after every method invocation,
</code>
<code class="sd"> the object on which the method was called will be passed.
</code>
<code class="sd"> The predicate should evaluate to True if the class
invariant</code>
<code class="sd"> has been maintained, or False if it has been violated.</code>
<code class="sd"> Returns:</code>
<code class="sd"> A class decorator for checking the class invariant tested
by</code>
<code class="sd"> the supplied predicate function.</code>
<code class="sd"> """</code>
  <code class="k">def</code> <code
class="nf">invariant_checking_class_decorator</code><code class="p">
(</code><code class="bp">cls</code><code class="p">):</code>
  <code class="sd">"""A class decorator for checking invariants."""</code>
```

<code class="n">method_names</code> <code class="o">=</code> <code
class="p">[</code> <code class="n">name</code> <code class="k">for</code>
<code class="n">name</code> <code class="p">,</code> <code
class="n">attr</code> <code class="ow">in</code> <code
class="nb">vars</code> <code class="p">(</code> <code
class="bp">class="o">.</code> <code
class="p">)</code> <code class="o">.</code>
<code class="n">items</code> <code class="p">()</code> <code
class="k">if</code> <code class="nb">callable</code> <code class="p">(</code> <code class="p")</code></code></code></code class="p">(</code> <code class="p")>)</code></code></code>

<code class="k">for</code> <code class="n">name</code> <code
class="ow">in</code> <code class="n">method_names</code> <code
class="p">:</code>

<code class="n">_wrap_method_with_invariant_checking_proxy</code>
<code class="p">(</code><code class="bp">cls</code><code class="p">,
</code> <code class="n">,</code> <code
class="n">predicate</code><code class="p">)</code>

<code class="n">property_names</code> <code class="o">=</code> <code
class="p">[</code> <code class="n">name</code> <code class="k">for</code>
<code class="n">name</code> <code class="p">,</code> <code
class="n">attr</code> <code class="ow">in</code> <code
class="nb">vars</code> <code class="p">(</code> <code
class="bp">class="bp">cls</code> <code class="p">)</code> <code class="o">.</code>
<code class="n">items</code> <code class="p">()</code> <code
class="k">if</code> <code class="nb">isinstance</code> <code class="p">(</code> <code class="p")>(</code> <code class="p">,</code> <code
class="n">p">,</code> <code
class="n">p">,</code> <code
class="n">p">,</code> <code
class="n">p">,</code> <code
class="n">p">,</code> <code</code> <code
class="n">p">,</code> <code</code> <code</td>

<code class="n">operty</code><code class="p">)]</code>

<code class="k">for</code> <code class="n">name</code> <code class="ow">in</code> <code class="n">property_names</code> <code class="p">:</code>

```
<code class="n">_wrap_property_with_invariant_checking_proxy</code>
<code class="p">(</code><code class="bp">cls</code><code class="p">,
</code> <code class="p">,</code> <code
class="n">predicate</code><code class="p">)</code>
```

<code class="k">return</code> <code class="bp">cls</code>

<code class="k">return</code> <code
class="n">invariant checking class decorator</code>

<code></code><code class="k">def</code> <code
class="nf">_wrap_property_with_invariant_checking_proxy</code><code
class="p">(</code><code class="bp">cls</code><code class="p">,</code>
<code class="n">name</code><code class="p">,</code> <code
class="n">predicate</code><code class="p">):</code>

<code class="n">prop</code> <code class="o">=</code> <code
class="nb">getattr</code> <code class="p">(</code> <code
class="bp">cls</code> <code class="p">,</code> <code
class="n">name</code> <code class="p">)</code>

<code class="k">assert</code> <code class="nb">isinstance</code> <code
class="p">(</code> <code class="n">prop</code> <code class="p">,</code>
 <code class="p">)</code>

<code class="n">invariant_checking_proxy</code> <code class="o">=
</code> <code class="n">InvariantCheckingPropertyProxy</code> <code
class="p">(</code> <code class="n">prop</code> <code class="p">,</code>
<code class="n">predicate</code> <code class="p">)</code>

<code class="nb">setattr</code><code class="p">(</code><code
class="bp">cls</code><code class="p">,</code> <code</pre>

```
class="n">name</code><code class="p">,</code> <code
class="n">invariant_checking_proxy</code><code class="p">)</code>
<code></code><code class="k">class</code> <code
class="nc">InvariantCheckingPropertyProxy</code><code class="p">:</code>
  <code class="k">def</code> <code class="fm"> init </code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">referent</code><code class="p">,</code> <code
class="n">predicate</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> referent</code> <code class="o">=</code> <code
class="n">referent</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> predicate</code> <code class="o">=</code> <code
class="n">predicate</code>
  <code class="k">def</code> <code class="fm">__get__</code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">instance</code><code class="p">,</code> <code
class="n">owner</code><code class="p">):</code>
  <code class="k">if</code> <code class="n">instance</code> <code
class="ow">is</code> <code class="bp">None</code> <code class="p">:
</code>
  <code class="k">return</code> <code class="bp">self</code>
  <code class="n">result</code> <code class="o">=</code> <code
class="bp">self</code><code class="o">.</code><code
class="n"> referent</code><code class="o">.</code><code
class="fm"> get </code><code class="p">(</code><code
class="n">instance</code><code class="p">,</code> <code
```

class="n">owner</code><code class="p">)</code>

```
<code class="k">if</code> <code class="ow">not</code> <code
class="bp">self</code> <code class="o">.</code> <code
class="n">_predicate</code> <code class="p">(</code> <code
class="n">instance</code> <code class="p">):</code>
```

<code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code> <code class="s2">"Class invariant {!r} violated for {!r}"
</code> <code class="o">.</code> <code class="n">format</code> <code
class="p">(</code></code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_predicate</code><code class="o">.</code><code
class="vm">__doc__</code><code class="p">,</code> <code
class="n">instance</code><code class="p">))</code>

<code class="k">return</code> <code class="n">result</code>

<code class="k">def</code> <code class="fm">__set__</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">instance</code> <code class="p">,</code> <code
class="n">value</code> <code class="p">):</code>

<code class="n">result</code> <code class="o">=</code> <code
class="bp">self</code> <code class="o">.</code> <code
class="n">_referent</code> <code class="o">.</code> <code
class="fm">__set__</code> <code class="p">(</code> <code
class="n">instance</code> <code class="p">,</code> <code
class="n">value</code> <code class="p">)</code>

<code class="k">if</code> <code class="ow">not</code> <code
class="bp">self</code> <code class="o">.</code> <code
class="n">_predicate</code> <code class="p">(</code> <code
class="n">instance</code> <code class="p">):</code>

<code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code> <code class="s2">"Class invariant {!r} violated for {!r}"
</code> <code class="o">.</code> <code class="n">format</code> <code
class="p">(</code></code>

```
<code class="bp">self</code><code class="o">.</code><code
class="n">_predicate</code><code class="o">.</code><code
class="vm">__doc__</code><code class="p">,</code> <code
class="n">instance</code><code class="p">))</code>
```

<code class="k">return</code> <code class="n">result</code>

<code class="k">def</code> <code class="fm">__delete__</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">instance</code> <code class="p">):</code>

<code class="n">result</code> <code class="o">=</code> <code
class="bp">self</code> <code class="o">.</code> <code
class="n">_referent</code> <code class="o">.</code> <code
class="fm">__delete__ </code> <code class="p">(</code> <code
class="n">instance</code> <code class="p">)</code></code>

<code class="k">if</code> <code class="ow">not</code> <code
class="bp">self</code> <code class="o">.</code> <code
class="n">_predicate</code> <code class="p">(</code> <code
class="n">instance</code> <code class="p">):</code>

<code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code> <code class="s2">"Class invariant {!r} violated for {!r}"
</code> <code class="o">.</code> <code
class="n">format</code> <code
class="p">(</code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_predicate</code><code class="o">.</code><code
class="vm">__doc__</code><code class="p">,</code> <code
class="n">instance</code><code class="p">))</code>

<code class="k">return</code> <code class="n">result</code>

<code></code><code class="gp">>>> </code><code class="n">t</code> <code class="o">=</code> <code class="n">Temperature</code><code class="p"> (</code><code class="p">)</code>

```
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">celsius</code> <code class="o">=</code> <code
class="o">-</code><code class="mi">300</code>
<code class="gt">Traceback (most recent call last):</code>
  File <code class="nb">"<input>"</code>, line <code class="m">1</code>, in
<code class="n"><module></code>
  File <code class="nb">"class_decorators.py"</code>, line <code
class="m">64</code>, in <code class="n"> set </code>
  <code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="bp">self</code><code class="o">.</code>
<code class="n">_predicate</code><code class="o">.</code><code
class="n">__</code>\
<code class="err">doc , instance))</code>
<code class="go">RuntimeError: Class invariant 'Temperature not below
absolute zero' violated for <class_d\</code>
<code class="go">ecorators.Temperature object at 0x1030226d8></code>
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">fahrenheit</code> <code class="o">=</code> <code
class="mi">100</code>
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">celsius</code>
<code class="go">37.777777777783</code>
<code></code><code class="k">def</code> <code
class="nf">not below absolute zero</code><code class="p">(</code><code
class="n">temperature</code><code class="p">):</code>
  <code class="sd">"""Temperature not below absolute zero"""</code>
```

```
<code class="k">return</code> <code class="n">temperature</code><code
class="o">.</code><code class="n"> kelvin</code> <code class="o">>=
</code> <code class="mi">0</code>
<code class="k">def</code> <code class="nf">below_absolute_hot</code>
<code class="p">(</code><code class="n">temperature</code><code
class="p">):</code>
  <code class="sd">"""Temperature below absolute hot"""</code>
  <code class="c1"># See http://en.wikipedia.org/wiki/Absolute hot</code>
  <code class="k">return</code> <code class="n">temperature</code><code
class="o">.</code><code class="n"> kelvin</code> <code class="o"><=
</code> <code class="mf">1.416785e32</code>
<code class="nd">@invariant</code><code class="p">(</code><code
class="n">below absolute hot</code><code class="p">)</code>
<code class="nd">@invariant</code><code class="p">(</code><code
class="n">not_below_absolute_zero</code><code class="p">)</code>
<code class="k">class</code> <code class="nc">Temperature</code> <code
class="p">:</code>
  <code class="k">def</code> <code class="fm"> init </code> <code
class="p">(</code><code class="bp">self</code><code class="p">,</code>
<code class="n">kelvin</code><code class="p">):</code>
  <code class="bp">self</code><code class="o">.</code><code
class="n"> kelvin</code> <code class="o">=</code> <code
```

class="n">kelvin</code>

<code class="k">def</code> <code class="nf">get_kelvin</code><code
class="p">(</code> <code class="bp">self</code> <code class="p">):</code>

<code class="k">return</code> <code class="bp">self</code><code class="o">.</code> <code class="n">_kelvin</code>

<code class="k">def</code> <code class="nf">set_kelvin</code><code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">value</code> <code class="p">):</code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_kelvin</code> <code class="o">=</code> <code
class="n">value</code>

<code class="nd">@property</code>

<code class="k">def</code> <code class="nf">celsius</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">):</code>

<code class="k">return</code> <code class="bp">self</code><code class="o">.</code> <code class="o">-</code> <code class="o">-</code> <code class="o">-</code>

<code class="nd">@celsius.setter</code>

<code class="k">def</code> <code class="nf">celsius</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">value</code> <code class="p">):</code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_kelvin</code> <code class="o">=</code> <code
class="n">value</code> <code class="o">+</code> <code
class="mf">273.15</code>

```
<code class="nd">@property</code>
```

<code class="k">def</code> <code class="nf">fahrenheit</code> <code
class="p">(</code> <code class="bp">self</code> <code class="p">):</code>

<code class="k">return</code> <code class="bp">self</code> <code
class="o">.</code> <code class="n">_kelvin</code> <code class="o">*</code>
<code class="mi">9</code> <code class="o">/</code> <code
class="mi">5</code> <code class="o">-</code> <code
class="mf">459.67</code>

<code class="nd">@fahrenheit.setter</code>

<code class="k">def</code> <code class="nf">fahrenheit</code><code
class="p">(</code> <code class="bp">self</code> <code class="p">,</code>
<code class="n">value</code> <code class="p">):</code>

<code class="bp">self</code><code class="o">.</code><code
class="n">_kelvin</code> <code class="o">=</code> <code class="p">
(</code> <code class="o">+</code> <code
class="mf">459.67</code> <code class="p">)</code> <code class="o">*
</code> <code class="o">></code> <code class="o">></code> <code class="o">></code> <code class="o">></code> <code class="o">>/</code> <code clas

<code></code><code class="gp">>>> </code><code class="n">t</code> <code class="o">=</code> <code class="n">Temperature</code><code class="p"> (</code><code class="p">)</code>

<code class="gp">>>> </code><code class="n">t</code><code class="o">. </code><code class="n">set_kelvin</code><code class="p">(</code><code class="p">(</code><code class="p">)</code>

<code class="gt">Traceback (most recent call last):</code>

File <code class="nb">"<input>"</code>, line <code class="m">1</code>, in <code class="n"><module></code>

```
File <code class="nb">"class_decorators.py"</code>, line <code
class="m">39</code>, in <code
class="n">invariant_checking_method_decorator</code>
  <code class="n">result</code> <code class="o">=</code> <code
class="n">method</code><code class="p">(</code><code
class="bp">self</code><code class="p">,</code> <code class="o">*</code>
<code class="n">args</code><code class="p">,</code> <code class="o">**
</code><code class="n">kwargs</code><code class="p">)</code>
  File <code class="nb">"class decorators.py"</code>, line <code
class="m">41</code>, in <code
class="n">invariant checking method decorator</code>
  <code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="n">predicate</code><code class="o">.
</code><code class="vm">__doc__</code><code class="p">,</code>\
<code class="gr"> self))</code>
<code class="gr">RuntimeError</code>: <code class="n">Class invariant
'Temperature not below absolute zero' violated for <class d\</code>
<code class="go">ecorators.Temperature object at 0x103025780></code>
<code class="gp">>>> </code><code class="n">t</code><code class="o">.
</code><code class="n">set kelvin</code><code class="p">(</code><code
class="mf">1e33</code><code class="p">)</code>
<code class="gt">Traceback (most recent call last):</code>
  File <code class="nb">"<input>"</code>, line <code class="m">1</code>, in
<code class="n"><module></code>
  File <code class="nb">"class decorators.py"</code>, line <code
class="m">41</code>, in <code
class="n">invariant checking method decorator</code>
```

```
<code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="n">predicate</code><code class="o">.
</code><code class="vm">__doc__</code><code class="p">,</code>\
<code class="gr"> self))</code>
<code class="gr">RuntimeError</code>: <code class="n">Class invariant
'Temperature below absolute hot' violated for <class_decora\</code>
<code class="go">tors_4.Temperature object at 0x103025780></code>
<code></code><code class="gp">>>> </code><code class="n">t</code><code
class="o">.</code><code class="n">celsius</code> <code class="o">=</code>
<code class="o">-</code><code class="mi">300</code>
<code class="gt">Traceback (most recent call last):</code>
  File <code class="nb">"<input>"</code>, line <code class="m">1</code>, in
<code class="n"><module></code>
  File <code class="nb">"class decorators 4.py"</code>, line <code
class="m">64</code>, in <code class="n"> set </code>
  <code class="k">raise</code> <code class="ne">RuntimeError</code> <code
class="p">(</code><code class="s2">"Class invariant {!r} violated for {!r}"
</code><code class="o">.</code><code class="n">format</code><code
class="p">(</code><code class="bp">self</code><code class="o">.</code>
<code class="n">_predicate</code><code class="o">.</code><code
class="n"> </code>\
<code class="err">doc___, instance))</code>
<code class="go">RuntimeError: Class invariant 'Temperature not below
absolute zero' violated for <class d\</code>
<code class="go">ecorators_4.Temperature object at 0x103025780></code>
<code></code><code class="gp">>>> </code><code class="n">t</code><code
```

```
class="o">.</code><code class="n">celsius</code> <code class="o">=</code> <code class="mf">1e34</code>
```

```
<code class="go">>>></code>
```

<code></code><code class="n">property_names</code> <code class="o">=
</code> <code class="p">[</code> <code class="n">name</code> <code
class="k">for</code> <code class="n">name</code> <code class="p">,</code>
<code class="n">attr</code> <code class="ow">in</code> <code
class="nb">vars</code> <code class="p">(</code> <code
class="bp">cls</code> <code class="p">)</code> <code
class="bp">cls</code> <code class="p">)</code> <code class="o">.</code>
<code class="n">items</code> <code class="p">()</code> <code
class="k">if</code> <code class="nb">isinstance</code> <code
class="p">,</code> <code
class="p">,</code> <code
class="p">,</code> <code
class="p">,</code> <code
class="p">,</code> <code
class="p">,</code> <code
class="p">)]</code></code></code>

<code class="k">for</code> <code class="n">name</code> <code class="ow">in</code> <code class="n">property_names</code> <code class="p">:</code>

<code class="n">_wrap_property_with_invariant_checking_proxy</code>
<code class="p">(</code><code class="bp">cls</code><code class="p">,
</code> <code class="n">name</code><code class="p">,</code> <code
class="n">predicate</code><code class="p">)</code>

However, our InvariantCheckingPropertyProxy does not satisfy the isinstance() check, so our proxy which enforces the not_below_absolute_zero invariant is applied to the genuine property although the below_absolute_hot proxy is not applied.

For a solution to this problem we'll use Python's abstract base-class mechanism, a topic we'll explore in the next chapter.

Summary

- Class decorators provide a simpler alternative to metaclasses for processing class definitions prior to the definition being bound to the class name.
- Remember, though, that class decorators are less powerful than metaclasses.
- Use class decorators when you can, and metaclasses when you must.

Class decorators can be chained just like function decorators.

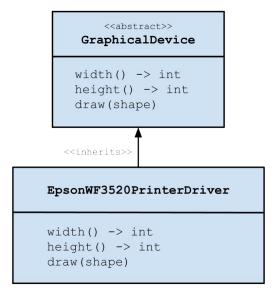
Chapter 8 - Abstract Base Classes

In this chapter we'll be investigating the abstract base-class mechanism in Python as originally defined in <u>PEP 3119</u>. In <u>The Python Journeyman</u> we used some abstract base classes, such as collections.abc.Sequence, when implementing a sorted-set collection type. This time, we'll look at the tools provided in the Python Standard Library <u>abc module</u> for creating abstract base-classes of your own design.

If you're coming to Python from another object-oriented language such as Java, C++, or C# you may have preconceived ideas of what an abstract base class is, and how to use one. Beware though! The abstract base-class mechanism in Python is much more flexible and can work in what may seem to be very surprising ways, so pay attention!

What is an abstract base-class?

Thinking for a moment beyond the confines of Python, what is, in general, an abstract base-class? The clue is in the name: *base* refers to the fact that the class is intended to be the target of an inheritance relationship; that is, we expect another class to derive from the base-class. For instance, GraphicalDevice here is intended as the base-class for other classes such as our EpsonWF3520PrinterDriver.



Graphical device inheritance

Abstract refers to the fact that the class cannot be instantiated in isolation; that is, it makes no sense to create an object of the type of the base-class alone. It only makes sense to instantiate the class as part of an object of a derived type. Ideally, it should should not be *possible* to instantiate an abstract base-class directly. The opposite of abstract is *concrete*, and in this example the printer driver is a concrete class, so it makes sense to instantiate it.

Why are abstract base-classes useful?

The rationale for any abstract base-class is to define an interface which derived classes *must* implement. This allows client code to be written against the base-class interface. In this example the printer driver must override the three abstract methods of GraphicalDevice. Done diligently, this leads to a highly desirable property of class hierarchies called *Liskov Substitutability*, a design principle which states that subclasses, from the point of view of client code, should be interchangeable. In other words, client code developed against an abstract interface should not require knowledge of specific concrete types, only of the capabilities as promised by the abstract base-class. So, for example, code written against the interface of GraphicalDevice should be able to render to an Epson WF3520 printer, or an 1080p LCD Display without modification — we can substitute one concrete class for another.

Abstract base-classes differ from pure interfaces such as those we have in languages like Java, insofar as they can also contain implementation code which is to be shared by all derived classes.

What about duck typing?

Why do we need to define named interfaces when we have duck typing? Isn't it sufficient to know whether a particular object responds to the interface of a duck, and behaves like a duck, without actually knowing that it *is* a duck?

In Python, this is true both in theory and practice, but determining whether a particular object supports the required interface in advance of exercising that interface can be quite awkward. For example, what does it mean in Python to be a *mutable sequence*? We know that list is a mutable sequence, but we cannot assume that all mutable sequences are lists. In fact, the mutable sequence protocol requires at least sixteen methods are implemented. When relying on duck-typing it can be difficult to be sure that you've met the requirements, and if clients *do* need to determine whether a particular object is admissable as a mutable sequence with a look-before-you-leap approach, the check is messy and awkward to perform robustly.

Abstract base-classes in Python

Abstract base-classes in Python serve two purposes: First of all they provide a mechanism for defining protocols or interfaces, and ensuring that implementers of those protocols meet some minimum requirements. Secondly they provide a means for easily determining whether an arbitrary class or instance meets the requirements of a specific protocol.

For instance, we can determine that list is mutable sequence using the built-in issubclass() function:

```
>>> from collections.abc import MutableSequence
>>> issubclass(list, MutableSequence)
True
```

This much may not be surprising, but let's look at the base-class of list. In fact, we'll look at the transitive base-classes of list by examining it's method resolution order using the __mro__ attribute:

```
>>> list.__mro__
(<class 'list'>, <class 'object'>)
```

This reveals that list has only one base-class, object, and that MutableSequence is nowhere to be seen. Further reflection — if you'll excuse the pun — might lead you to wonder how it is that such a fundamental type as Python's list can be a subclass of a type defined in a *library* module.

We've started out with this curious example so as to efficiently disabuse migrants from other programming languages of any existing notions of what abstract base-classes are, how they work, how they are used, and why they are useful! That done, we will dig further into the mechanism.

Abstract base-classes are indeed abstract

Let's establish that MutableSequence is indeed abstract, by attempting to directly instantiate it:

```
>>> ms = MutableSequence()
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
TypeError: Can't instantiate abstract class MutableSequence with abstract methods __delit\
em__, __getitem__, __len__, __setitem__, insert
```

This fails with a useful TypeError explaining the five methods we need to implement the *Mutable Sequence* protocol. The reason we don't need to implement all sixteen is that eleven of them can be implemented in terms of the other five, and the MutableSequence abstract base-class contains code to do exactly this. Note, however, that these implementations may not be the most efficient since they can't exploit knowledge of the concrete class, but must instead work entirely through the interface of the abstract class.

Defining subclasses with __subclasscheck__()

Now to the question of how issubclass(list, MutableSequence) returns

True. What happens is that when we call issubclass(list, MutableSequence)
the built-in issubclass() function checks for the existence of a method called

__subclasscheck__() on the *type* of MutableSequence (which is to say, on the
metaclass of MutableSequence), and if it is present it is called with the subclass
list as an argument. In other wordst:

```
issubclass(list, MutableSequence)
```

is roughly equivalent to:

```
if hasattr(type(MutableSequence), '__subclasscheck__'):
    return type(MutableSequence).__subclasscheck__(list)
# normal issubclass() behaviour...
```

It's up to the metaclass of MutableSequence to determine whether or not list is a subclass of MutableSequence, rather than list being required to know that MutableSequence is one of its base-classes. This allows list to be a subclass of MutableSequence without MutableSequence being a superclass of list (in the normal sense of being the target of an inheritance relationship). This unusual asymmetry between the superclass and subclass relationships in Python can be pretty mind-boggling if you're coming from, say, Java, but it's also incredibly powerful. We describe such base-classes as *virtual* base-classes, which is nothing at all to do with an identically named concept in C++.

A simple example

The __subclasscheck__() method on the metaclass of the virtual base class can do pretty much anything it likes to determine whether its argument is to be considered a subclass. Consider the code in weapons.py:

```
class SwordMeta(type):
    def __subclasscheck__(cls, sub):
        return (hasattr(sub, 'swipe') and callable(sub.swipe)
                hasattr(sub, 'sharpen') and callable(sub.sharpen))
class Sword(metaclass=SwordMeta):
    pass
class BroadSword:
    def swipe(self):
        print("Swoosh!")
    def sharpen(self):
        print("Shink!")
class SamuraiSword:
    def swipe(self):
        print("Slice!")
    def sharpen(self):
        print("Shink!")
class Rifle:
    def fire(self):
        print("Bang!")
```

In this module we have defined a Sword class with a metaclass SwordMeta. SwordMeta defines the __subclasscheck__() method to check for the existence of callable swipe and sharpen attributes on the class. In this situation Sword will play the role of a virtual base-class. A few simple tests at the REPL confirm that BroadSword and SamuraiSword are indeed considered subclasses of Sword even though there is no explicit relationship through inheritance:

```
>>> issubclass(BroadSword, Sword)
True
>>> issubclass(SamuraiSword, Sword)
True
>>> issubclass(Rifle, Sword)
False
```

__instancecheck__()

This isn't the whole story though, as tests of *instances* using isinstance() will return inconsistent results:

```
>>> samurai_sword = SamuraiSword()
>>> isinstance(samurai_sword, Sword)
False
```

This is because the isinstance() machinery checks for the existence of the __instancecheck__() metamethod which we have not yet implemented. Let's do so now:

Our __instancecheck__() implementation simply delegates to issubclass(). With this change in place, our call to isinstance() produces a result consistent with the result from issubclass():

```
>>> samurai_sword = SamuraiSword()
>>> isinstance(samurai_sword, Sword)
True
```

This surprising technique is used in Python for some of the collection abstract base-classes, including Sized:

After importing Sized from collections.abc we define a new class SizedCollection which is *not* related to Sized through inheritance. SizedCollection stores an integer size and has a __init__() to initialize the size, and __len__() to allow the size to be retrieved with the built-in len() function.

In this case, implementing a __len__() method is sufficient to be considered a subclass of the Sized abstract base-class.

It's worth bearing in mind that our implementations of SwordMeta.__instancecheck__() and SwordMeta.__subclasscheck__() are somewhat naïve, as they make no attempt to check the regular, non-virtual base-classes of the objects being tested. This could lead to some surprising behaviour. Bear in mind that correctly overriding __subclasscheck__() and __instancecheck__() on your own metaclasses is difficult. Don't worry though, we'll be presenting some more digestible alternatives shortly.

Non-transitivity of subclass relationships

Overriding __subclasscheck__() affords class implementers a great deal of flexibility. So much flexibility in fact that not only should you not expect symmetry between the superclass and subclass relationships, you shouldn't expect transitivity of the subclass relationship. What this means is that if C is a subclass of B, and B is subclass of A, it doesn't necessarily follow that C is a subclass of A.

One glaring example from Python revolves around the Hashable virtual baseclass from collections.abc:

```
>>> from collections.abc import Hashable
>>> issubclass(object, Hashable)
True
>>> issubclass(list, object)
True
```

```
>>> issubclass(list, Hashable)
False
```

This occurs because list — which, remember, is a mutable collection — disables hashing by removing __hash__(). This method would otherwise be inherited from object, and the Hashable abstract base-class checks for it through its __subclasscheck__():

```
>>> object.__hash__
<slot wrapper '__hash__' of 'object' objects>
>>>
>>> list.__hash__
>>>
    list.__hash__ is None
True
```

Some further investigation reveals that the list class sets the __hash__ attribute to None. The Hashable.__subclasscheck__() implementation checks for this eventuality and uses it to signal non-hashability.

This example is also interesting because it demonstrates the fact that even the ultimate *base-class* object can be considered a subclass of Hashable — underlying the lack of symmetry between superclass and subclass relationships in Python.

Method resolution and virtual base-classes

It's worth bearing in mind that, unlike regular base-classes, virtual base-classes don't play a role in method resolution. We'll demonstrate this by adding a thrust() method to the Sword virtual base-class:

```
class Sword(metaclass=SwordMeta):
    def thrust(self):
        print("Thrusting...")
```

Attempts to invoke this method on subclasses raise an AttributeError, and checking further, we can see that Sword is not present in the MRO for BroadSword:

```
>>> broad_sword = BroadSword()
>>> isinstance(broad_sword, Sword)
True
>>> broad_sword.swipe()
Swipe!
>>> broad_sword.thrust()
Traceback (most recent call last):
  File "<input>", line 1, in <module>
```

```
AttributeError: 'BroadSword' object has no attribute 'thrust'
>>>
>>> BroadSword.__mro__
(<class 'BroadSword'>, <class 'object'>)
```

For this reason, it's not possible to call virtual base-class methods using super() since super() works by searching the MRO.

Library support for abstract base-classes

We've pointed out that implementing __subclasscheck__() and __instancecheck__() correctly can be awkward to get right. Fortunately, the standard library provides support for implementing abstract base-classes with the abc module — and particularly the ABCMeta metaclass — along with some other useful pieces of infrastructure including the ABC base-class and the @abstractmethod decorator. We'll cover each of these in detail now.

The ABCMeta metaclass

The ABCMeta metaclass implements reliable __subclasscheck__() and __instancecheck__() methods along with some other handy capabilities we'll come to shortly. We'll eventually use ABCMeta in our Sword example, allowing us to dispose of SwordMeta. But first we're going to cannibalise ABCMeta so that we can see how it works:

```
from abc import ABCMeta

class Sword(metaclass=ABCMeta):
    pass
```

ABCMeta doesn't, of course, know what it means to be a sword, so the test that was previously in SwordMeta.__subclasscheck__() needs to be relocated elsewhere. The ABCMeta.__subclasscheck__() method calls the special __subclasshook__() method on our actual class to perform the test.

In fact, all Python objects have the __subclasshook__() classmethod which accepts the potential subclass as its only argument. The method should return True, False, or NotImplemented. We can see this by calling it on object:

```
>>> object.__subclasshook__()
NotImplemented
```

For classes where __subclasshook__() returns NotImplemented the subclass test continues with the usual mechanism of testing the non-virtual base-classes.

Boolean return values, on the other hand, definitively indicate whether the argument class is to be considered a subclass of the base-class or not.

Let's try implementing __subclasshook__() for our Sword class with pretty much the same definition we used for __subclasscheck__() previously:

This class method, in conjunction with ABCMeta, is sufficient to make issubclass() and isinstance() work:

```
>>> issubclass(SamuraiSword, Sword)
True
>>> issubclass(Rifle, Sword)
False
>>> broad_sword = BroadSword()
>>> isinstance(broad_sword, Sword)
True
```

SamuraiSword is a subclass of Sword. Rifle is not a subclass of Sword, and an instance of BroadSword is an instance of a Sword.

Virtual subclass registration

It's also possible to directly register a class as a virtual subclass of an abstract base-class whose metaclass is ABCMeta by using the register() metamethod. For example, let's create an abstract base-class Text and register the built-in type str as a virtual subclass:

```
>>> class Text(metaclass=ABCMeta):
... pass
...
>>> Text.register(str)
<class 'str'>
```

Notice that the register() metamethod returns the class which was registered — a point we'll return to in a moment. Now we can even retrofit base-classes (albeit virtual ones) to the built-in types:

```
>>> issubclass(str, Text)
True
>>> isinstance("Is this text?", Text)
True
```

Here were demonstrate that the built-in str class is now considered a subclass of our our Text class defined at the REPL, and str objects are instances of Text.

Using register as a decorator

Because the register() metamethod returns its argument, we can even use it as a class decorator. Let's register a class Prose as a virtual subclass of Text:

```
>>> @Text.register
... class Prose:
... pass
...
>>> issubclass(Prose, Text)
True
```

Combining subclass registration with __subclasshook__()

You need to take care when combining virtual subclass registration with the __subclasshook__() technique because the result of __subclasshook__() takes precedence over the subclass registry. Returning True or False from __subclasshook__() is taken as a definite answer. If you also want registration to be accounted for you should return NotImplemented to indicate "not sure".

To see this in action, let's add a LightSaber which has no sharpen() method to our example. This class won't satisfy the __subclasshook__() test we defined in Sword, but we still want it identified as a virtual subclass of Sword, so we've registered it using the decorator form of Sword.register:

```
@Sword.register
class LightSaber:

    def swipe(self):
        print("Ffffkrrrrshhzzzwooooom..woom..woooom..")
```

Even though we've registered LightSaber with Sword the subclass test returns False:

```
>>> issubclass(LightSaber, Sword)
False
```

To fix this, we need to ensure that __subclasshook__() never returns False because doing so causes the __subclasscheck__() implementation in ABCMeta to skip the check for registered subclasses. Instead, in the case of a negative result, we should return NotImplemented:

```
class Sword(metaclass=ABCMeta):
```

With this change in place — which exploits shortcut evaluation of the logical operators — subclass detection now works as expected for implicitly detected subclasses, explicitly registered subclasses, and non-subclasses:

```
>>> issubclass(BroadSword, Sword)
True
>>> issubclass(LightSaber, Sword)
True
>>> issubclass(Rifle, Sword)
False
```

Bear in mind that this somewhat contrived example is designed to demonstrate that you should take care with subclass registration. How useful is our virtual base-class Sword now? The answer is "not very". Being an instance of Sword is no longer a useful predicate for the object in question since we can't guarantee the presence of the sharpen() method, which was — if you'll excuse the pun — the whole point of the sword.

The ABC convenience base-class

The abc module contains a class ABC which is simply a regular class that has ABCMeta as its metaclass:

```
class ABC(metaclass=ABCMeta):
    """Helper class that provides a standard way to create an ABC using
    inheritance.
    """
    pass
```

This makes it even easier to declare abstract base-classes without having to put the metaclass mechanism on show. This may be an advantage when coding for audiences who haven't been exposed to the concept of metaclasses. Using ABC our Sword class becomes:

```
hasattr(sub, 'sharpen') and callable(sub.sharpen))
or NotImplemented)
```

The abstractmethod decorator

Finally, we reach the aspect of Python's abstract base-classes which most immediately springs to mind when we talk about abstract base-classes in general: The ability to declare *abstract methods*.

An abstract method is a method which is declared but which doesn't necessarily have a useful definition. Abstract methods *must* be overridden in derived concrete classes, and the presence of an abstract method will prevent its host class being instantiated.

Abstract methods should be decorated with the @abstractmethod decorator, and their abstractness will only be enforced if the metaclass of the host class is ABCMeta:

```
from abc import (ABC, abstractmethod)

class AbstractBaseClass(ABC): # metaclass is ABCMeta

@abstractmethod
    def an_abstract_method(self):
        raise NotImplementedError # Method body syntactically required.
```

Let's add abstract methods for swiping and thrusting to our Sword abstract base-class. We'll also update __subclasshook__() to match:

```
@abstractmethod
def parry(self):
    raise NotImplementedError
```

Python syntax requires that we provide an implementation for each method, so for swipe() and parry() we raise a NotImplementedError but for thrust() we provide a default behaviour of printing "Thrusting!".

We'll take this opportunity to remind you of the distinction between NotImplemented and NotImplementedError. NotImplemented is a value returnable from predicate functions which are unable to make a determination of True or False. On the other hand, NotImplementedError is an exception type to be raised in place of missing code.

We'll now make BroadSword a *real* subclass of Sword, by explicitly adding Sword to the list of base-classes:

```
class BroadSword(Sword):
    def swipe(self):
        print("Swipe!")

def sharpen(self):
    print("Shink!")
```

We cannot instantiate BroadSword as it too is still abstract:

```
>>> broad_sword = BroadSword()
Traceback (most recent call last):
   File "<input>", line 1, in <module>
TypeError: Can't instantiate abstract class BroadSword with abstract methods parry, thrust
```

We must implement all abstract methods for the class to be considered concrete:

```
class BroadSword(Sword):
    def swipe(self):
        print("Swoosh!")

    def thrust(self):
        super().thrust()

    def parry(self):
        print("Parry!")

    def sharpen(self):
        print("Shink!")
```

Notice that in thrust() we call the implementation provided in the abstract class, using super():

```
>>> broad_sword = BroadSword()
>>> broad_sword.swipe()
Swipe!
```

Note however, that the requirement to implement abstract methods doesn't extend to *virtual* subclasses, only to *real* subclasses. We can still instantiate SamuraiSword successfully, even though it doesn't implement any of the abstract methods:

```
>>> samurai_sword = SamuraiSword()
>>> samurai_sword
<SamuraiSword object at 0x103022160>
```

Combining abstractmethod with other decorators

What if you need to combine @abstractmethod with other decorators? The @abstractmethod decorator can be combined with the @staticmethod, @classmethod, and @property decorators, although care must be taken that @abstractmethod is the *innermost* decorator. For properties you can independently mark the getters and setters as abstract:

```
class AbstractBaseClass(ABC):
   @staticmethod
   @abstractmethod
   def an_abstact_static_method():
       raise NotImplementedError
   @classmethod
   @abstractmethod
   def an_abstract_class_method(cls):
       raise NotImplementedError
   @property
   @abstractmethod
    def an_abstract_property(self):
       raise NotImplementedError
   @an_abstract_property.setter
   @abstractmethod
   def an_abstract_property(self, value):
        raise NotImplementedError
```

Recall the properties are implemented using descriptors. When implementing your own descriptors you need to do a little extra work to ensure your descriptor plays nicely with @abstractmethod. Let's look at that now.

Propagating abstractness through descriptors

Any descriptors which are implemented in terms of abstract methods should identify themselves as abstract by exposing a __isabstractmethod__ attribute which should evaluate to True:

```
class MyDataDescriptor(ABC):
    @abstractmethod
    def __get__(self, instance, owner):
        # ...
        pass

    @abstractmethod
    def __set__(self, instance, value):
        # ...
        pass

    @abstractmethod
    def __delete__(self, instance):
        # ...
        pass

    @property
    def __isabstractmethod__(self):
        return True # or False if not abstract
```

This __isabstractmethod__ attribute can either be implemented as a class attribute of the descriptor class, or it can itself be implemented as a property, as we have done here. Implementing it as a property is useful in cases where abstractness needs to be determined at runtime, as we'll see in an example shortly.

Let's see this in action with a very simple example. We'll define a class called AbstractBaseClass which inherits from ABC. Within this class we'll define two properties called abstract_property and concrete_property using the appropriate combinations of decorators:

```
>>> from abc import (ABC, abstractmethod)
>>> class AbstractBaseClass(ABC):
...     @property
...     @abstractmethod
...     def abstract_property(self):
...         raise NotImplementedError
...     @property
...     def concrete_property(self):
...     return "sand, cement, water"
...
>>>
```

By querying the descriptor objects created by the property decorators we can inspect the __isabstractmethod__ attributes of the two properties, and see that they have been marked as abstract and non-abstract (or concrete) as necessary:

```
>>> AbstractBaseClass.abstract_property.__isabstractmethod__
True
>>> AbstractBaseClass.concrete_property.__isabstractmethod__
False
```

This happens because the __isabstractmethod__ flag is inspected by the ABCMeta implementation when the metaclass creates the actual class object which hosts the property.

So much for the theory, let's see this in practice.

Fixing our @invariant class decorator with ABCs

We rounded off <u>chapter 7</u> on class decorators by building a class decorator for checking class invariants after every method call and property access. This worked fine for both methods and properties with a single application of the decorator, but with chained @invariant decorators the checking didn't work as planned for properties; only the innermost @invariant was taking effect. Let's recap.

Although the innermost lower bound check works:

```
>>> t = Temperature(42)
>>> t.celsius = -300
Traceback (most recent call last):
   File "<input>", line 1, in <module>
   File "class_decorators_4.py", line 64, in __set__
        raise RuntimeError("Class invariant {!r} violated for {!r}".format(self._predicate.__\doc__, instance))
RuntimeError: Class invariant 'Temperature not below absolute zero' violated for <class_d\ecorators_4.Temperature object at 0x103025780>
```

Violating the upper bound is not detected as we had intended:

```
>>> t.celsius = 1e34
```

The problem here is that our class decorator is detecting specifically property instances with this fragment:

Because our property wrappers of type InvariantCheckingPropertyProxy are not detected as instances of property they are not wrapped a second time, and

the invariant specified in the outermost decorator is not enforced.

Abstract base-classes to the rescue

We promised to use abstract base-classes to fix this problem, so let's go! We'll introduce a new abstract base-class called PropertyDataDescriptor which inherits from the ABC convenience class and which contains three abstract methods which define the data descriptor protcol. It also includes the abstract property __isabstractmethod__() for correct propagation of abstractness:

```
class PropertyDataDescriptor(ABC):
```

```
@abstractmethod
def __get__(self, instance, owner):
    raise NotImplementedError

@abstractmethod
def __set__(self, instance, value):
    raise NotImplementedError

@abstractmethod
def __delete__(self, instance):
    raise NotImplementedError

@property
@abstractmethod
def __isabstractmethod__(self):
    raise NotImplementedError
```

Note that becuase __isabstractmethod__ needs to look like an abstract attribute, we have implemented it by applying the @abstractmethod and @property decorators in that order.

Having defined an abstract *base* class, we now need some subclasses. The first will be a *virtual* subclass — the built-in property class — which we'll register with the base-class:

```
PropertyDataDescriptor.register(property)
```

The second subclass will be a real subclass. We'll modify our existing property proxy InvariantCheckingPropertyProxy to inherit from PropertyDataDescriptor, which will also require that we override the __isabstractmethod__ property:

```
class InvariantCheckingPropertyProxy(PropertyDataDescriptor):
    def __init__(self, referent, predicate):
        self._referent = referents
        self._predicate = predicate
```

```
def __get__(self, instance, owner):
    if instance is None:
        return self
    result = self._referent.__get__(instance, owner)
    if not self._predicate(instance):
raise RuntimeError("Class invariant {!r} violated for {!r}".format(
                       self._predicate.__doc__, instance))
    return result
def __set__(self, instance, value):
    result = self._referent.__set__(instance, value)
    if not self._predicate(instance):
raise RuntimeError("Class invariant {!r} violated for {!r}".format(
                       self._predicate.__doc__, instance))
   return result
def __delete__(self, instance):
    result = self._referent.__delete__(instance)
    if not self._predicate(instance):
raise RuntimeError("Class invariant {!r} violated for {!r}".format(
                       self._predicate.__doc__, instance))
    return result
@property
def __isabstractmethod__(self):
   return self._referent.__isabstractmethod__
```

Finally, we need to update the search-and-wrap logic in invariant_checking_class_decorator() to use the more general test for instances of PropertyDataDescriptor rather than the more specific test for just property.

```
def invariant(predicate):
    """Create a class decorator which checks a class invariant.
   Args:
       predicate: A callable to which, after every method invocation,
            the object on which the method was called will be passed.
            The predicate should evaluate to True if the class invariant
            has been maintained, or False if it has been violated.
   Returns:
       A class decorator for checking the class invariant tested by
        the supplied predicate function.
   def invariant_checking_class_decorator(cls):
        """A class decorator for checking invariants."""
       method_names = [name for name, attr
                        in vars(cls).items()
                        if callable(attr)]
        for name in method_names:
            _wrap_method_with_invariant_checking_proxy(cls, name, predicate)
        property_names = [name for name, attr
                          in vars(cls).items()
                          if isinstance(attr, PropertyDataDescriptor)]
        for name in property_names:
            _wrap_property_with_invariant_checking_proxy(cls, name, predicate)
```

```
return cls
return invariant_checking_class_decorator
```

With these changes in place both invariants are enforced on property writes:

```
>>> t = Temperature(42)
>>> t.celsius = -300
Traceback (most recent call last):
  File "<input>", line 1, in <module>
  File "class_decorators.py", line 86, in __set_
  result = self._referent.__set__(instance, value)
File "class_decorators.py", line 88, in __set__
    raise RuntimeError("Class invariant {!r} violated for {!r}".format(self._predicate.__\
doc__, instance))
RuntimeError: Class invariant 'Temperature not below absolute zero' violated for <class d\
ecorators. Temperature object at 0x103012cc0>
>>> t.celsius = 1e34
Traceback (most recent call last):
 File "<input>", line 1, in <module>
  File "class_decorators.py", line 88, in __set
    raise RuntimeError("Class invariant {!r} violated for {!r}".format(self._predicate.__\
doc__, instance))
RuntimeError: Class invariant 'Temperature below absolute hot' violated for <class_decora\
tors. Temperature object at 0x103012cc0>
```

We create a temperature of 42 kelvin, then attempt to modify the temperature through the celsius setter with a temperature of -300 celsius (which is below absolute zero). This now fails as designed, signalling violation of the 'Temperature not below absolute zero' invariant. Then we test the other invariant, by assigning a temperature of 1 × 10^34 through the celsius setter. This also fails as designed, signalling violation 'Temperature below absolute hot' class invariant.

There's a lot going on in this code with decorators, metaclasses, abstract baseclasses, and descriptors, and it may seem somewhat complicated. All this complexity is well encapsulated in the invariant class decorator, however, so take a step back and enjoy the simplicity of the client code in the Temperature class.

Summary

In this chapter we've explained Python's system of abstract base-classes, which is rather somewhat more flexible that similar concepts in other languages. In particlar we covered these topics:

- Subclass/instance checking
 - How the behaviour of the built-in issubclass() and isinstance() functions can be specialised for a base-class by defining the __subclasscheck__() and __instancecheck__() methods on the metaclass of that base-class.
 - Specialised subclass checks allow us to centralize the definition of what it means to be a subclass by gathering look-before-you-leap protocol checks into one place. Any class which implements the required protocol will become at least a virtual subclass of a virtual base-class.
- The standard library abc module contains tools for assisting in the definition of abstract base-classes.
 - Most important amongst those tools is the ABCMeta metaclass which can be used as the metaclass for abstract base-classes.
 - Slightly more conveniently, you can simply inherit from the ABC class which has ABCMeta as its metaclass.
 - ABCMeta provides default implementations of both
 __subclasscheck__() and __instancecheck__() which support two
 means of identifying subclasses: A special __subclasshook__()
 classmethod on abstract base-classes and a registration method.
 - __subclasshook__() accepts a candidate subclass as its only argument and should return True or NotImplemented. False should only be returned if it is desired to disable subclass registration.
 - Passing any class even a built-in class to the register() metamethod of an abstract base-class will register the argument as a *virtual* subclass of the base-class.
 - An @abstractmethod decorator can be used to prevent instantiation of abstract classes. It requires methods marked as such to be overridden in real — although not virtual — subclasses.
 - The @abstractmethod decorator can be combined with other decorators such as @staticmethod, @classmethod, and @property, but @abstractmethod should always be the innermost decorator.
 - Descriptors should propagate abstractness from underlying methods by exposing the __isabstractmethod__ attribute.

Afterword: Continue the journey

Python is a large and complicated language with many moving parts. We find it remarkable that much of this complexity is hidden so well in Python. We hope in this book we've given you deeper insight into some important mechanisms in Python which, while a bit trickier to understand, can deliver great expressive power. You have reached the end of this book on advanced Python, and now is the time to take what you have learned and apply these powerful techniques in your work and play with Python. No matter where your journey goes, though, remember that, above all else, it's great *fun* to write Python software, so enjoy yourselves!

Notes

- **1** At least Python 3.5, though any version greater than that will work as well (e.g. 3.6). *←*
- **2** See https://mail.python.org/pipermail/python-ideas/2009-October/006157.html↔
- **3** See http://citeseerx.ist.psu.edu/viewdoc/download? doi=10.1.1.103.6084&rep=rep1&type=pdf <u>←</u>
- **4** A widespread misconception is that else clauses associated with for-loops are executed if and only if the iterator which the for-loop is consuming is initially empty. This is not the case, the else clause will be executed *when* the iterator is exhausted, but only if the for-loop was also exited 'naturally', without invoking break. As such, the else clause is useful to handle the "not found" case when searching for something:
- **5** Note we are required to pass the length parameter to specify how many bytes we want in the result and the byteorder parameter to specify whether we want the bytes returned in big-endian order with the most significant byte first, or little-endian order with the least significant byte first <u>←</u>
- **6** We've taken additional steps in this book to make the output even more readable by wrapping the interleaved lines of indices and buffer data to make it more readable. The code we present outputs all indices on one line, and all buffer bytes on the following line. <u>←</u>
- 7 The word *cast* is a synonym for *type-conversion* used in many different programming languages. The exact meaning is language dependent, but it typically refers to reinterpreting the bit-level data as a different type. We don't often talk about casting in Python, as Python is strongly typed, but this one place where the notion of casting crops up. <u>←</u>

- **8** From Python 3.7 dictionary insertion order will be preserved, according to dictat from Python's Benevolent Dictator for Life (BDFL), Guido van Rossum, on the Python mailing list in December 2017. https://mail.python.org/pipermail/python-dev/2017-December/151283.html <u>←</u>
- **9** You'll notice that in our view, whatever the International Astronomical Union says, Pluto **is** a planet!<u>←</u>
- **10** Of course your system will likely report different numbers. A lot of factors are involved in memory usage. The real point is that allocating 10,000 board should show a marked increase in memory usage. *⊆*
- **11** Recall from <u>chapter 4</u> that a descriptor is any object supporting any of the __get__(), __set__(), or __delete__() methods. ←