06-scopes-closures-and-decorators

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0.0.1 Global and Local Scopes

In Python the **global** scope refers to the **module** scope.

The scope of a variable is normally defined by where it is (lexically) defined in the code.

```
[1]: a = 10
```

In this case, a is defined inside the main module, so it is a global variable.

In this case, **c** was defined inside the function **my_func**, so it is **local** to the function **my_func**. In this example, **n** is also **local** to **my_func**

Global variables can be accessed from any inner scope in the module, for example:

```
[3]: def my_func(n):
    print('global:', a)
    c = a ** n
    return c
```

```
[4]: my_func(2)
```

global: 10

[4]: 100

As you can see, my_func was able to reference the global variable a.

But remember that the scope of a variable is determined by where it is assigned. In particular, any variable defined (i.e. assigned a value) inside a function is local to that function, even if the variable name happens to be global too!

```
[5]: def my_func(n):
    a = 2
    c = a ** 2
    return c
```

```
[6]: print(a)
print(my_func(3))
print(a)
```

10

4

10

In order to change the value of a global variable within an inner scope, we can use the **global** keyword as follows:

```
[7]: def my_func(n):
    global a
    a = 2
    c = a ** 2
    return c
```

```
[8]: print(a)
 print(my_func(3))
 print(a)
```

10

4

2

As you can see, the value of the global variable a was changed from within my_func.

In fact, we can **create** global variables from within an inner function - Python will simply create the variable and place it in the **global** scope instead of the **local scope**:

```
[9]: def my_func(n):
    global var
    var = 'hello world'
    return n ** 2
```

Now, var does not exist yet, since the function has not run:

```
[10]: print(var)
```

```
NameError Traceback (most recent call last)
Cell In[10], line 1
----> 1 print(var)

NameError: name 'var' is not defined
```

Once we call the function though, it will create that global var:

```
[11]: my_func(2)
```

```
[11]: 4
```

```
[12]: print(var)
```

hello world

Beware!! Remember that whenever you assign a value to a variable without having specified the variable as global, it is local in the current scope. Moreover, it does not matter where the assignment in the code takes place, the variable is considered local in the entire scope - Python determines the scope of objects at compile-time, not at run-time.

Let's see an example of this:

```
[13]: a = 10

b = 100
```

```
[14]: def my_func():
    print(a)
    print(b)
```

```
[15]: my_func()
```

10 100

So, this works as expected - \mathbf{a} and \mathbf{b} are taken from the global scope since they are referenced **before** being assigned a value in the local scope.

But now consider the following example:

```
[16]: a = 10
b = 100

def my_func():
    print(a)
    print(b)
    b = 1000
```

```
[17]: my_func()
```

10

```
UnboundLocalError Traceback (most recent call last)
Cell In[17], line 1
----> 1 my_func()

Cell In[16], line 6, in my_func()
4 def my_func():
```

```
5 print(a)
----> 6 print(b)
7 b = 1000

UnboundLocalError: cannot access local variable 'b' where it is not associated
with a value
```

As you can see, **b** in the line **print(b)** is considered a **local** variable - that's because the **next** line **assigns** a value to **b** - hence **b** is scoped as local by Python for the **entire** function.

Of course, functions are also objects, and scoping applies equally to function objects too. For example, we can "mask" the built-in print Python function:

[18]: 'hello world!'

You may be wondering how we get our real print function back!

```
[19]: del print
```

```
[20]: print('hello')
```

hello

Yay!!

If you have experience in some other programming languages you may be wondering if loops and other code "blocks" have their own local scope too. For example in Java, the following would not work:

```
for (int i=0; i<10; i++) { int x = 2 * i; } system.out.println(x);
```

But in Python it works perfectly fine:

```
[21]: for i in range(10):
    x = 2 * i
print(x)
```

18

In this case, when we assigned a value to x, Python put it in the global (module) scope, so we can reference it after the for loop has finished running.

0.0.2 Nonlocal Scopes

Functions defined inside anther function can reference variables from that enclosing scope, just like functions can reference variables from the global scope.

```
[22]: def outer_func():
    x = 'hello'

    def inner_func():
        print(x)

    inner_func()
```

```
[23]: outer_func()
```

hello

In fact, any level of nesting is supported since Python just keeps looking in enclosing scopes until it finds what it needs (or fails to find it by the time it finishes looking in the built-in scope, in which case a runtime error occurrs.)

```
[24]: def outer_func():
    x = 'hello'
    def inner1():
        def inner2():
            print(x)
        inner2()
        inner1()
```

```
[25]: outer_func()
```

hello

But if we **assign** a value to a variable, it is considered part of the local scope, and potentially **masks** enclosin scope variable names:

```
[26]: def outer():
    x = 'hello'
    def inner():
        x = 'python'
    inner()
    print(x)
```

```
[27]: outer()
```

hello

As you can see, \mathbf{x} in **outer** was not changed.

To achieve this, we can use the **nonlocal** keyword:

```
[28]: def outer():
    x = 'hello'
    def inner():
        nonlocal x
        x = 'python'
    inner()
    print(x)
```

```
[29]: outer()
```

python

Of course, this can work at any level as well:

[31]: outer()

python

How far Python looks up the chain depends on the first occurrence of the variable name in an enclosing scope.

Consider the following example:

```
[32]: def outer():
    x = 'hello'
    def inner1():
        x = 'python'
        def inner2():
            nonlocal x
            x = 'monty'
        print('inner1 (before):', x)
        inner2()
        print('inner1 (after):', x)
    inner1()
    print('outer:', x)
```

```
[33]: outer()
```

```
inner1 (before): python
inner1 (after): monty
outer: hello
```

What happened here, is that x in inner1 masked x in outer. But inner2 indicated to Python that x was nonlocal, so the first local variable up in the enclosing scope chain Python found was the one in inner1, hence x in inner2 is actually referencing x that is local to inner1

We can change this behavior by making the variable x in inner nonlocal as well:

```
[34]: def outer():
          x = 'hello'
          def inner1():
             nonlocal x
              x = 'python'
              def inner2():
                  nonlocal x
                  x = 'monty'
              print('inner1 (before):', x)
              inner2()
              print('inner1 (after):', x)
          inner1()
          print('outer:', x)
[35]: outer()
     inner1 (before): python
     inner1 (after): monty
     outer: monty
[36]: x = 100
      def outer():
          x = 'python' # masks global x
          def inner1():
              nonlocal x # refers to x in outer
              x = 'monty' # changed x in outer scope
              def inner2():
                  global x # refers to x in global scope
                  x = 'hello'
              print('inner1 (before):', x)
              inner2()
              print('inner1 (after):', x)
          inner1()
          print('outer', x)
[37]: outer()
```

print(x)

```
inner1 (after): monty
outer monty
hello
```

But this will not work. In inner Python is looking for a local variable called x, outer has a label called x, but it is a global variable, not a local one - hence Python does not find a local variable in the scope chain.

```
[38]: x = 100
def outer():
    global x
    x = 'python'

    def inner():
        nonlocal x
        x = 'monty'
    inner()

Cell In[38], line 7
    nonlocal x

SyntaxError: no binding for nonlocal 'x' found
```

0.0.3 Closures

Let's examine that concept of a cell to create an indirect reference for variables that are in multiple scopes.

```
[39]: def outer():
    x = 'python'
    def inner():
        print(x)
    return inner
```

```
[40]: fn = outer()
```

```
[41]: fn.__code__.co_freevars
```

```
[41]: ('x',)
```

[]:

As we can see, x is a free variable in the closure.

```
[42]: fn.__closure__
```

[42]: (<cell at 0x000002417AAFFF70: str object at 0x000002417785EC70>,)

Here we see that the free variable x is actually a reference to a cell object that is itself a reference to a string object.

Let's see what the memory address of x is in the outer function and the inner function. To be sure string interning does not play a role, I am going to use an object that we know Python will not automatically intern, like a list.

```
[43]: def outer():
    x = [1, 2, 3]
    print('outer:', hex(id(x)))
    def inner():
        print('inner:', hex(id(x)))
        print(x)
    return inner

[44]: fn = outer()
    outer: 0x2417af1dd80

[45]: fn.__closure__

[45]: (<cell at 0x000002417A90A6B0: list object at 0x000002417AF1DD80>,)

[46]: fn()
    inner: 0x2417af1dd80
```

As you can see, each the memory address of x in outer, inner and the cell all point to the same object.

Modifying the Free Variable We know we can modify nonlocal variables by using the nonlocal keyword. So the following will work:

```
[47]: def counter():
    count = 0 # local variable

    def inc():
        nonlocal count # this is the count variable in counter
        count += 1
        return count
    return inc
```

```
[48]: c = counter()
```

```
[49]: c()
```

[49]: 1

[1, 2, 3]

```
[50]: c()
```

[50]: 2

[56]: fn2()

[56]: 3

Shared Extended Scopes As we saw in the lecture, we can set up nonlocal variables in different inner functions that reference the same outer scope variable, i.e. we have a free variable that is shared between two closures. This works because both non local variables and the outer local variable all point back to the same cell object.

```
[51]: def outer():
          count = 0
          def inc1():
              nonlocal count
              count += 1
              return count
          def inc2():
              nonlocal count
              count += 1
              return count
          return inc1, inc2
[52]: fn1, fn2 = outer()
[53]: fn1.__closure__, fn2.__closure__
[53]: ((<cell at 0x000002417AAFEDD0: int object at 0x00007FFE1CDEE308>,),
       (<cell at 0x000002417AAFEDD0: int object at 0x00007FFE1CDEE308>,))
     As you can see here, the count label points to the same cell.
[54]: fn1()
[54]: 1
[55]: fn1()
[55]: 2
```

0.0.4 Multiple Instances of Closures

Recall that **every** time a function is called, a **new** local scope is created.

```
[57]: from time import perf_counter
      def func():
          x = perf_counter()
          print(x, id(x))
[58]: func()
     34732.2120398 2480251053968
[59]: func()
     34732.662089 2480251053968
     The same thing happens with closures, they have their own extended scope every time the closure
     is created:
[60]: def pow(n):
           # n is local to pow
          def inner(x):
               # x is local to inner
               return x ** n
          return inner
     In this example, n, in the function inner is a free variable, so we have a closure that contains inner
     and the free variable n
[61]: square = pow(2)
[62]: square(5)
[62]: 25
[63]: cube = pow(3)
[64]: cube(5)
[64]: 125
     We can see that the cell used for the free variable in both cases is different:
[65]: square.__closure__
[65]: (<cell at 0x000002417AAFF4C0: int object at 0x00007FFE1CDEE348>,)
[66]: cube.__closure__
```

[66]: (<cell at 0x000002417AAFEB00: int object at 0x00007FFE1CDEE368>,)

In fact, these functions (square and cube) are **not** the same functions, even though they were "created" from the same power function:

```
[67]: id(square), id(cube)
```

```
[67]: (2480254295296, 2480254296576)
```

0.0.5 Beware!

Remember when I said the captured variable is a reference established when the closure is created, but the value is looked up only once the function is called?

This can create very subtle bugs in your program.

Consider the following example where we want to create some functions that can add 1, 2, 3, 4 and to whatever is passed to them.

We could do the following:

```
[68]: def adder(n):
    def inner(x):
        return x + n
    return inner
```

```
[69]: add_1 = adder(1)
add_2 = adder(2)
add_3 = adder(3)
add_4 = adder(4)
```

```
[70]: add_1(10), add_2(10), add_3(10), add_4(10)
```

```
[70]: (11, 12, 13, 14)
```

But suppose we want to get a little fancier and do it as follows:

```
[71]: def create_adders():
    adders = []
    for n in range(1, 5):
        adders.append(lambda x: x + n)
    return adders
```

```
[72]: adders = create_adders()
```

Now technically we have 4 functions in the adders list:

```
[73]: adders
```

```
The first one should add 1 to the value we pass it, the second should add 2, and so on.
[74]: adders[3](10)
[74]: 14
     Yep, that works for the 4th function.
[75]: adders[0](10)
[75]: 14
     Uh Oh - what happened? In fact we get the same behavior from every one of those functions:
[76]: adders[0](10), adders[1](10), adders[2](10), adders[3](10)
[76]: (14, 14, 14, 14)
     Remember what I said about when the variable is captured and when the value is looked up?
     When the lambdas are created their n is the n used in the loop - the same n!!
[77]: adders[0].__code__.co_freevars
[77]: ('n',)
      adders[0].__closure__
[78]: (<cell at 0x000002417AAFE590: int object at 0x00007FFE1CDEE388>,)
[79]: adders[1].__closure__
[79]: (<cell at 0x000002417AAFE590: int object at 0x00007FFE1CDEE388>,)
[80]: adders[2].__closure__
[80]: (<cell at 0x000002417AAFE590: int object at 0x00007FFE1CDEE388>,)
[81]: adders[3].__closure__
[81]: (<cell at 0x000002417AAFE590: int object at 0x00007FFE1CDEE388>,)
     So, by the time we call adder[i], the free variable n (shared between all adders) is set to 4.
[82]: hex(id(4))
[82]: '0x7ffe1cdee388'
```

<function __main__.create_adders.<locals>.<lambda>(x)>]

As we can see the memory address of the singleton integer 4, is what that cell is pointint to.

If you want to use a loop to do this and not end up using the same cell for each of the free variables, we can use a simple trick that forces the evaluation of n at the time the closure is **created**, instead of when the closure function is evaluated.

We can do this by creating a parameter for n in our lambda whose default value is the current value of n - remember from an earlier video that parameter defaults are avaluated when the function is created, not called.

```
[83]: def create_adders():
    adders = []
    for n in range(1, 5):
        adders.append(lambda x, step=n: x + step)
    return adders
```

```
[84]: adders = create_adders()
```

```
[85]: adders[0].__closure__
```

Why aren't we getting anything in the closure? What about free variables?

```
[86]: adders[0].__code__.co_freevars
```

[86]: ()

Hmm, nothing either... Why?

Well, look at the lambda in that loop. Does it reference the variable n (other than in the default value)? No. Hence, n is **not** a free variable in this case, and our lambda is just a plain lambda, not a closure.

And this code will now work as expected:

```
[87]: adders[0](10)

[87]: 11

[88]: adders[1](10)

[88]: 12

[89]: adders[2](10)

[89]: 13

[90]: adders[3](10)
```

You just need to understand that since the default values are evaluated when the function (lambda in this case) is **created**, the then-current **n** value is assigned to the local variable **step**. So **step** will not change every time the lambda is called, and since **n** is not referenced inside the function (and therefore evaluated when the lambda is called), **n** is not a free variable.

Nested Closures We can also nest closures, as can be seen in this example:

```
[91]: def incrementer(n):
          def inner(start):
              current = start
              def inc():
                  a = 10 # local var
                  nonlocal current
                  current += n
                  return current
              return inc
          return inner
[92]: fn = incrementer(2)
[93]:
     fn
[93]: <function __main__.incrementer.<locals>.inner(start)>
[94]: fn.__code__.co_freevars
[94]: ('n',)
[95]: fn.__closure__
[95]: (<cell at 0x000002417AAFB280: int object at 0x00007FFE1CDEE348>,)
[96]: inc_2 = fn(100)
[97]: inc 2
[97]: <function __main__.incrementer.<locals>.inner.<locals>.inc()>
[98]: inc_2.__code__.co_freevars
[98]: ('current', 'n')
[99]: inc_2.__closure__
[99]: (<cell at 0x000002417AAF8E20: int object at 0x00007FFE1CDEEF88>,
       <cell at 0x000002417AAFB280: int object at 0x00007FFE1CDEE348>)
```

Here you can see that the second free variable n, is pointing to the same cell as the free variable in fn.

Note that **a** is a local variable, and is not considered a free variable.

And we can call the closures as follows:

```
[100]: inc_2()
[100]: 102
[101]: inc_2()
[101]: 104
[102]: inc_3 = incrementer(3)(200)
[103]: inc_3()
[104]: 203
[104]: 206
```

0.0.6 Closure Applications (Part 1)

In this example we are going to build an averager function that can average multiple values.

The twist is that we want to simply be able to feed numbers to that function and get a running average over time, not average a list which requires performing the same calculations (sum and count) over and over again.

```
[105]: class Averager:
    def __init__(self):
        self.numbers = []

    def add(self, number):
        self.numbers.append(number)
        total = sum(self.numbers)
        count = len(self.numbers)
        return total / count

[106]: a = Averager()

[107]: a.add(10)
```

```
[108]: a.add(20)
[108]: 15.0
[109]: a.add(30)
[109]: 20.0
      We can do this using a closure as follows:
[110]: def averager():
           numbers = []
           def add(number):
               numbers.append(number)
                total = sum(numbers)
                count = len(numbers)
                return total / count
           return add
[111]: a = averager()
[112]: a(10)
[112]: 10.0
[113]: a(20)
[113]: 15.0
[114]: a(30)
[114]: 20.0
```

Now, instead of storing a list and reclaculating total and count every time wer need the new average, we are going to store the running total and count and update each value each time a new value is added to the running average, and then return total / count.

Let's start with a class approach first, where we will use instance variables to store the running total and count and provide an instance method to add a new number and return the current average.

```
[115]: class Averager:
    def __init__(self):
        self._count = 0
        self._total = 0

    def add(self, value):
        self._total += value
        self._count += 1
```

```
return self._total / self._count
[116]: a = Averager()
[117]: a.add(10)
[117]: 10.0
[118]: a.add(20)
[118]: 15.0
[119]: a.add(30)
[119]: 20.0
      Now, let's see how we might use a closure to achieve the same thing.
[120]: def averager():
           total = 0
           count = 0
           def add(value):
               nonlocal total, count
               total += value
               count += 1
               return 0 if count == 0 else total / count
           return add
[121]: a = averager()
[122]: a(10)
[122]: 10.0
[123]: a(20)
[123]: 15.0
[124]: a(30)
[124]: 20.0
```

Generalizing this example We saw that we were essentially able to convert a class to an equivalent functionality using closures. This is actually true in a much more general sense - very

often, classes that define a single method (other than initializers) can be implemented using a closure instead.

Let's look at another example of this.

[133]: x()

Suppose we want something that can keep track of the running elapsed time in seconds.

```
[125]: from time import perf_counter
[126]: class Timer:
           def __init__(self):
               self._start = perf_counter()
           def __call__(self):
               return (perf_counter() - self._start)
[127]: a = Timer()
      Now wait a bit before running the next line of code:
[128]: a()
[128]: 0.05063949999748729
      Let's start another "timer":
[129]: b = Timer()
[130]: print(a())
       print(b())
      0.12850159999652533
      0.03166890000284184
      Now let's rewrite this using a closure instead:
[131]: def timer():
           start = perf_counter()
           def elapsed():
                # we don't even need to make start nonlocal
                # since we are only reading it
               return perf_counter() - start
           return elapsed
[132]: x = timer()
```

```
[133]: 0.03930480000417447
[134]: y = timer()
[135]: print(x())
       print(y())
      0.17775740000070073
      0.031299300004320685
[136]: print(a())
       print(b())
       print(x())
       print(y())
      2.2515246999973897
      2.154861900002288
      0.21783770000183722
      0.07127410000248346
  []:
             Closure Applications (Part 2)
      Example 1 Let's write a small function that can increment a counter for us - we don't have an
      incrementor in Python (the ++ operator in Java or C++ for example):
[137]: def counter(initial_value):
           # initial_value is a local variable here
           def inc(increment=1):
               nonlocal initial value
               # initial_value is a nonlocal (captured) variable here
               initial_value += increment
               return initial_value
           return inc
[138]: counter1 = counter(0)
[139]: print(counter1(0))
```

1

[140]: print(counter1())

1222

return a + b

As you can see, each closure maintains a reference to the **initial_value** variable that was created when the **counter** function was **called** - each time that function was called, a new local variable **initial_value** was created (with a value assigned from the argument), and it became a nonlocal (captured) variable in the inner scope.

Example 2 Let's modify this example to now build something that can run, and maintain a count of how many times we have run some function.

```
[148]: def counter(fn):
    cnt = 0 # initially fn has been run zero times

def inner(*args, **kwargs):
    nonlocal cnt
    cnt = cnt + 1
    print('{0} has been called {1} times'.format(fn.__name__, cnt))
    return fn(*args, **kwargs)

    return inner
[149]: def add(a, b):
```

```
[150]: counted_add = counter(add)
      And the free variables are:
[151]: counted_add.__code__.co_freevars
[151]: ('cnt', 'fn')
      We can now call the counted_add function:
[152]: counted_add(1, 2)
      add has been called 1 times
[152]: 3
[153]: counted_add(2, 3)
      add has been called 2 times
[153]: 5
[154]: def mult(a, b, c):
           return a * b * c
[155]: counted_mult = counter(mult)
[156]: counted_mult(1, 2, 3)
      mult has been called 1 times
[156]: 6
[157]: counted_mult(2, 3, 4)
      mult has been called 2 times
[157]: 24
      Example 3 Let's take this one step further, and actually store the function name and the number
      of calls in a global dictionary instead of just printing it out all the time.
[158]: counters = dict()
       def counter(fn):
           cnt = 0 # initially fn has been run zero times
```

def inner(*args, **kwargs):

nonlocal cnt

```
cnt = cnt + 1
                counters[fn.__name__] = cnt # counters is global
               return fn(*args, **kwargs)
           return inner
[159]: counted_add = counter(add)
       counted_mult = counter(mult)
      Note that counters is a global variable, and therefore not a free variable:
[160]: counted_add.__code__.co_freevars
[160]: ('cnt', 'fn')
[161]: counted_mult.__code__.co_freevars
[161]: ('cnt', 'fn')
      We can now call out functions:
[162]: counted_add(1, 2)
[162]: 3
[163]: counted add(2, 3)
[163]: 5
[164]: counted_mult(1, 2, 'a')
[164]: 'aa'
[165]: counted_mult(2, 3, 'b')
[165]: 'bbbbbb'
[166]:
       counted_mult(1, 1, 'abc')
[166]: 'abc'
[167]: print(counters)
      {'add': 2, 'mult': 3}
```

Of course this relies on us creating the **counters** global variable first and making sure we are naming it that way, so instead, we're going to pass it as an argument to the **counter** function:

```
[168]: def counter(fn, counters):
           cnt = 0 # initially fn has been run zero times
           def inner(*args, **kwargs):
               nonlocal cnt
               cnt = cnt + 1
               counters[fn.__name__] = cnt # counters is nonlocal
               return fn(*args, **kwargs)
           return inner
[169]: func_counters = dict()
       counted_add = counter(add, func_counters)
       counted_mult = counter(mult, func_counters)
[170]: counted_add.__code__.co_freevars
[170]: ('cnt', 'counters', 'fn')
      As you can see, counters is now a free variable.
      We can now call our functions:
[171]: for i in range(5):
           counted_add(i, i)
       for i in range(10):
           counted_mult(i, i, i)
[172]: print(func_counters)
      {'add': 5, 'mult': 10}
      Of course, we don't have to assign the "counted" version of our functions a new name - we can
      simply assign it to the same name!
[173]: def fact(n):
           product = 1
           for i in range(2, n+1):
               product *= i
           return product
[174]: fact = counter(fact, func_counters)
[175]: fact(0)
[175]: 1
[176]: fact(3)
```

```
[176]: 6
[177]: fact(4)
[177]: 24
[178]: print(func_counters)
       {'add': 5, 'mult': 10, 'fact': 3}
       Notice, how we essentially added some functionality to our fact function, without modifying what
       the fact function actually returns.
       This leads us straight into our next topic: decorators!
  []:
       0.0.8 Decorators (Part 1)
       Recall the example in the last section where we wrote a simple closure to count how many times a
       function had been run:
[179]: def counter(fn):
            count = 0
            def inner(*args, **kwargs):
                nonlocal count
                count += 1
                print('Function {0} was called {1} times'.format(fn.__name__, count))
                return fn(*args, **kwargs)
            return inner
```

```
[180]: def add(a, b=0):
    """
    returns the sum of a and b
    """
    return a + b
```

```
[181]: help(add)
```

Help on function add in module __main__:
add(a, b=0)
 returns the sum of a and b

Here's the memory address that add points to:

```
[182]: id(add)
```

[182]: 2480254295616

Now we create a closure using the add function as an argument to the counter function:

```
[183]: add = counter(add)
```

And you'll note that add is no longer the same function as before. Indeed the memory address add points to is no longer the same:

```
[184]: id(add)
```

[184]: 2480254301376

```
[185]: add(1, 2)
```

Function add was called 1 times

[185]: 3

```
[186]: add(2, 2)
```

Function add was called 2 times

[186]: 4

What happened is that we put our **add** function 'through' the **counter** function - we usually say that we **decorated** our function **add**.

And we call that **counter** function a **decorator**.

There is a shorthand way of decorating our function without having to type:

func = counter(func)

```
[187]: @counter
def mult(a: float, b: float=1, c: float=1) -> float:
    """
    returns the product of a, b, and c
    """
    return a * b * c
```

```
[188]: mult(1, 2, 3)
```

Function mult was called 1 times

[188]: 6

```
[189]: mult(2, 2, 2)
```

Function mult was called 2 times

```
Let's do a little bit of introspection on our two decorated functions:
[190]: add.__name__
[190]: 'inner'
[191]: mult.__name__
[191]: 'inner'
      As you can see, the name of the function is no longer add or mult, but instead it is the name of
      that inner function in our decorator.
[192]: help(add)
      Help on function inner in module __main__:
      inner(*args, **kwargs)
[193]: help(mult)
      Help on function inner in module __main__:
      inner(*args, **kwargs)
      As you can see, we've also lost our docstring and parameter annotations!
      What about introspecting the parameters of add and mult:
[194]: import inspect
[195]: inspect.getsource(add)
[195]: "
            def inner(*args, **kwargs):\n
                                                    nonlocal count\n
       print('Function {0} was called {1} times'.format(fn.__name__, count))\n
       return fn(*args, **kwargs)\n"
[196]: inspect.getsource(mult)
            def inner(*args, **kwargs):\n
                                                    nonlocal count\n
                                                                              count += 1\n
[196]: "
       print('Function {0} was called {1} times'.format(fn.__name__, count))\n
       return fn(*args, **kwargs)\n"
      Even the signature is gone:
[197]: inspect.signature(add)
```

[189]: 8

```
[197]: <Signature (*args, **kwargs)>
[198]: inspect.signature(mult)
[198]: <Signature (*args, **kwargs)>
      Even the parameter defaults documentation is are gone:
[199]: inspect.signature(add).parameters
[199]: mappingproxy({'args': <Parameter "*args">, 'kwargs': <Parameter "**kwargs">})
      In general, when we create decorated functions, we end up "losing" a lot of the metadata of our
      original function!
      However, we can put that information back in - it can get quite complicated.
      Let's see how we might be able to do that for some simple things, like the docstring and the function
      name.
[200]: def counter(fn):
            count = 0
           def inner(*args, **kwargs):
                nonlocal count
                count += 1
                print("{0} was called {1} times".format(fn.__name__, count))
           inner.__name__ = fn.__name__
           inner.__doc__ = fn.__doc__
           return inner
[201]: @counter
       def add(a: int, b: int=10) -> int:
            returns sum of two integers
           return a + b
[202]: help(add)
      Help on function add in module __main__:
      add(*args, **kwargs)
           returns sum of two integers
       add.__name__
[203]:
```

[203]: 'add'

At least we have the docstring and function name back... But what about the parameters? Our real **add** function takes two positional parameters, but because the closure used a generic way of accepting *args and **kwargs, we lose this information

We can use a special function in the **functools** module, called **wraps**. In fact, that function is a decorator itself!

```
[204]: from functools import wraps
[205]: def counter(fn):
           count = 0
           @wraps(fn)
           def inner(*args, **kwargs):
               nonlocal count
               count += 1
               print("{0} was called {1} times".format(fn.__name__, count))
           return inner
[206]: @counter
       def add(a: int, b: int=10) -> int:
           returns sum of two integers
           return a + b
[207]: help(add)
      Help on function add in module __main__:
      add(a: int, b: int = 10) \rightarrow int
          returns sum of two integers
      Yay!!! Everything is back to normal.
[208]: inspect.getsource(add)
                                                            """\n
[208]: '@counter\ndef add(a: int, b: int=10) -> int:\n
                                                                     returns sum of two
       integers\n
                     """\n
                               return a + b\n'
[209]: inspect.signature(add)
[209]: <Signature (a: int, b: int = 10) -> int>
[210]: inspect.signature(add).parameters
[210]: mappingproxy({'a': <Parameter "a: int">, 'b': <Parameter "b: int = 10">})
```

0.0.9 Decorators Application (Timing)

Here we go back to an example we have seen in the past - timing how long it takes to run a certain function.

```
[211]: def timed(fn):
           from time import perf_counter
           from functools import wraps
           @wraps(fn)
           def inner(*args, **kwargs):
               start = perf_counter()
               result = fn(*args, **kwargs)
               end = perf_counter()
               elapsed = end - start
               args_ = [str(a) for a in args]
               kwargs_ = ['{0}={1}'.format(k, v) for (k, v) in kwargs.items()]
               all_args = args_ + kwargs_
               args_str = ','.join(all_args)
               print('{0}({1}) took {2:.6f}s to run.'.format(fn.__name__,
                                                                 args_str,
                                                                 elapsed))
               return result
           return inner
```

Let's write a function that calculates the n-th Fibonacci number:

```
1, 1, 2, 3, 5, 8, ...
```

We will implement this using three different methods: 1. recursion 2. a loop 3. functional programming (reduce)

We use a 1-based system, e.g. first Fibonnaci number has index 1, etc.

Using Recursion

```
[212]: def calc_recursive_fib(n):
    if n <=2:
        return 1
    else:
        return calc_recursive_fib(n-1) + calc_recursive_fib(n-2)</pre>
```

```
[213]: calc_recursive_fib(3)
```

[213]: 2

```
[214]: calc_recursive_fib(6)
```

```
[214]: 8
[215]: @timed
       def fib_recursed(n):
           return calc_recursive_fib(n)
[216]: fib recursed(33)
      fib_recursed(33) took 0.501208s to run.
[216]: 3524578
[217]: fib_recursed(34)
      fib_recursed(34) took 0.729796s to run.
[217]: 5702887
[218]: fib_recursed(35)
      fib_recursed(35) took 1.270354s to run.
[218]: 9227465
      There's a reason we did not decorate our recursive function directly!
[219]: Otimed
       def fib_recursed_2(n):
           if n <=2:
               return 1
           else:
               return fib_recursed_2(n-1) + fib_recursed_2(n-2)
[220]: fib_recursed_2(10)
      fib_recursed_2(2) took 0.000001s to run.
      fib_recursed_2(1) took 0.000001s to run.
      fib_recursed_2(3) took 0.000326s to run.
      fib_recursed_2(2) took 0.000000s to run.
      fib recursed 2(4) took 0.000335s to run.
      fib_recursed_2(2) took 0.000000s to run.
      fib_recursed_2(1) took 0.000000s to run.
      fib_recursed_2(3) took 0.000007s to run.
      fib_recursed_2(5) took 0.000350s to run.
      fib_recursed_2(2) took 0.000000s to run.
      fib_recursed_2(1) took 0.000000s to run.
      fib_recursed_2(3) took 0.000007s to run.
      fib_recursed_2(2) took 0.000000s to run.
```

```
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(6) took 0.000372s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib recursed 2(3) took 0.000007s to run.
fib recursed 2(2) took 0.000000s to run.
fib recursed 2(4) took 0.000029s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(5) took 0.000043s to run.
fib_recursed_2(7) took 0.000422s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib recursed 2(3) took 0.000007s to run.
fib recursed 2(5) took 0.000029s to run.
fib recursed 2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000010s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000017s to run.
fib_recursed_2(6) took 0.000053s to run.
fib_recursed_2(8) took 0.000482s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(2) took 0.000000s to run.
fib recursed 2(1) took 0.000000s to run.
fib recursed 2(3) took 0.000011s to run.
fib recursed 2(5) took 0.000032s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(6) took 0.000053s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
```

```
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(5) took 0.000027s to run.
fib recursed 2(7) took 0.000087s to run.
fib_recursed_2(9) took 0.000577s to run.
fib recursed 2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(5) took 0.000028s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib recursed 2(4) took 0.000014s to run.
fib recursed 2(6) took 0.000048s to run.
fib recursed 2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(5) took 0.000028s to run.
fib_recursed_2(7) took 0.000083s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib recursed 2(2) took 0.000000s to run.
fib recursed 2(4) took 0.000014s to run.
fib recursed 2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(5) took 0.000027s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(1) took 0.000000s to run.
fib_recursed_2(3) took 0.000007s to run.
fib_recursed_2(2) took 0.000000s to run.
fib_recursed_2(4) took 0.000014s to run.
fib_recursed_2(6) took 0.000048s to run.
fib_recursed_2(8) took 0.000137s to run.
fib_recursed_2(10) took 0.000722s to run.
```

[220]: 55

Since we are calling the function recursively, we are actually calling the **decorated** function recursively. In this case I wanted the total time to calculate the n-th number, not the time for each recursion.

You will notice from the above how inefficient the recursive method is: the same fibonacci numbers are calculated repeatedly! This is why as the value of n start increasing beyond 30 we start seeing considerable slow downs.

Using a Loop

[225]: 9227465

```
[221]: @timed
       def fib_loop(n):
           fib_1 = 1
           fib_2 = 1
           for i in range(3, n+1):
               fib_1, fib_2 = fib_2, fib_1 + fib_2
           return fib_2
[222]: fib_loop(3)
      fib_loop(3) took 0.000002s to run.
[222]: 2
[223]: fib_loop(6)
      fib_loop(6) took 0.000002s to run.
[223]: 8
[224]: fib_loop(34)
      fib_loop(34) took 0.000004s to run.
[224]: 5702887
[225]: fib_loop(35)
      fib_loop(35) took 0.000003s to run.
```

As you can see this method is much more efficient!

Using Reduce We first need to understand how we are going to calculate the Fibonnaci sequence using reduce:

In general each step in the reduction is as follows:

If we start our reduction with an initial value of (1, 0), we need to run our "loop" n times.

We therefore use a "dummy" sequence of length n to create n steps in our reduce.

```
[226]: from functools import reduce
       @timed
       def fib_reduce(n):
           initial = (1, 0)
           dummy = range(n-1)
           fib_n = reduce(lambda prev, n: (prev[0] + prev[1], prev[0]),
                           dummy,
                           initial)
           return fib_n[0]
[227]: fib_reduce(3)
      fib_reduce(3) took 0.000004s to run.
[227]: 2
[228]: fib_reduce(6)
      fib_reduce(6) took 0.000004s to run.
[228]: 8
[229]: fib_reduce(34)
      fib_reduce(34) took 0.000009s to run.
[229]: 5702887
[230]: fib_reduce(35)
      fib_reduce(35) took 0.000008s to run.
[230]: 9227465
      Now we can run a quick comparison between the various timed implementations:
[231]: fib_recursed(35)
       fib_loop(35)
       fib_reduce(35)
      fib_recursed(35) took 1.266078s to run.
      fib_loop(35) took 0.000004s to run.
      fib_reduce(35) took 0.000007s to run.
```

[231]: 9227465

Even though the recursive algorithm is by far the easiest to understand, it is also the slowest. We'll see how to fix this in an upcoming video using a technique called **memoization**.

First let's focus on the loop and reduce variants. Our timing is not very effective since we only time a single calculation for each - there could be some variance if we run these tests multiple times:

```
[232]: for i in range(10):
           result = fib_loop(10000)
      fib_loop(10000) took 0.001650s to run.
      fib loop(10000) took 0.001626s to run.
      fib_loop(10000) took 0.001648s to run.
      fib_loop(10000) took 0.001623s to run.
      fib_loop(10000) took 0.001547s to run.
      fib_loop(10000) took 0.001898s to run.
      fib_loop(10000) took 0.001542s to run.
      fib_loop(10000) took 0.001585s to run.
      fib_loop(10000) took 0.001544s to run.
      fib_loop(10000) took 0.001544s to run.
[233]: for i in range(10):
           result = fib_reduce(10000)
      fib_reduce(10000) took 0.002337s to run.
      fib reduce(10000) took 0.002334s to run.
      fib_reduce(10000) took 0.002291s to run.
      fib_reduce(10000) took 0.002278s to run.
      fib_reduce(10000) took 0.002265s to run.
      fib_reduce(10000) took 0.002273s to run.
      fib_reduce(10000) took 0.002270s to run.
      fib_reduce(10000) took 0.002453s to run.
      fib reduce(10000) took 0.002373s to run.
      fib_reduce(10000) took 0.002308s to run.
```

In general it is better to time the same function call multiple times and generate and average of the run times.

We'll see in an upcoming video how we can do this from within our decorator.

In the meantime observe that the simple loop approach seems to perform about twice as fast as the reduce approach!!

The moral of this side note is that simply because you **can** do something in Python using some fancy or cool technique does not mean you **should!**

We technically could write our reduce-based function as a one liner:

```
[234]: from functools import reduce
```

```
[235]: fib_loop(100)
```

fib_loop(100) took 0.000007s to run.

[235]: 354224848179261915075

```
[236]: fib_1(100)
```

<lambda>(100) took 0.000017s to run.

[236]: 354224848179261915075

So yes, it's cool that you can write this using a single line of code, but consider two things here: 1. Is it as efficient as another method? 2. Is the code **readable**?

Code readability is something I cannot emphasize enough. Given similar efficiencies (cpu / memory), give preference to code that is more easily understandable!

Sometimes, if the efficiency is not greatly impacted (or does not matter in absolute terms), I might even give preference to less efficient, but more readable (i.e. understanbdable), code.

But enough of the soapbox already:-)

0.0.10 Decorators Application (Logger, Stacked Decorators)

In this example we're going to create a utility decorator that will log function calls (to the console, but in practice you would be writing your logs to a file (e.g. using Python's built-in logger), or to a database, etc.

```
[237]: def logged(fn):
    from functools import wraps
    from datetime import datetime, timezone

    @wraps(fn)
    def inner(*args, **kwargs):
        run_dt = datetime.now(timezone.utc)
        result = fn(*args, **kwargs)
        print('{0}: called {1}'.format(fn.__name__, run_dt))
        return result

    return inner
```

```
[238]: @logged def func_1(): pass
```

```
[239]: @logged def func_2(): pass
```

```
[240]: func_1()
```

func_1: called 2023-06-14 15:51:39.112419+00:00

```
[241]: func_2()
```

```
func_2: called 2023-06-14 15:51:39.214969+00:00
```

Now we may additionally also want to time the function. We can certainly include the code to do so in our logged decorator, but we could also just use the @timed decorator we already wrote by stacking our decorators.

```
[242]: def timed(fn):
    from functools import wraps
    from time import perf_counter

    @wraps(fn)
    def inner(*args, **kwargs):
        start = perf_counter()
        result = fn(*args, **kwargs)
        end = perf_counter()
        print('{0} ran for {1:.6f}s'.format(fn.__name__, end-start))
        return result

    return inner
```

```
[243]: @timed
    @logged
    def factorial(n):
        from operator import mul
        from functools import reduce
        return reduce(mul, range(1, n+1))
```

```
[244]: factorial(10)
```

```
factorial: called 2023-06-14 15:51:40.670672+00:00 factorial ran for 0.000179s
```

[244]: 3628800

Note that the order in which we stack the decorators can make a difference!

Remember that this is because our stacked decorators essentially amounted to:

```
[245]: def factorial(n):
    from operator import mul
    from functools import reduce
    return reduce(mul, range(1, n+1))

factorial = timed(logged(factorial))
```

So in this case the timed decorator will be called first, followed by the logged decorator.

You may wonder why the printed output seems reversed. Look at how the decorators were defined - they first ran the function passed in, and **then** printed the result.

So in the above example, a simplified look at what happens in each decorator:

```
• timed(fn)(*args, **kwargs):
```

- 1. calls fn(*args, **kwargs)
- 2. prints timing
- logged(fn)(*args, **kwargs):
 - 1. calls fn(*args, **kwargs)
 - 2. prints log info

So, calling factorial = timed(logged(factorial))

is equivalent to:

So as you can see, the timed decorator ran first, but it called the logged decorated function first, then printed the result - hence why the print output seems reversed.

```
[246]: factorial(10)
```

factorial: called 2023-06-14 15:51:42.880885+00:00 factorial ran for 0.000269s

[246]: 3628800

But in the following case, the logged decorator will run first, followed by the timed decorator:

```
[247]: def factorial(n):
    from operator import mul
    from functools import reduce

    return reduce(mul, range(1, n+1))

factorial = logged(timed(factorial))
```

```
[248]: factorial(10)
```

factorial ran for 0.000011s factorial: called 2023-06-14 15:51:45.042794+00:00

```
[248]: 3628800
      Or, using the **@** notation:
[249]: @logged
       @timed
       def factorial(n):
           from operator import mul
           from functools import reduce
           return reduce(mul, range(1, n+1))
[250]: factorial(10)
      factorial ran for 0.000011s
      factorial: called 2023-06-14 15:51:46.412749+00:00
[250]: 3628800
[251]: Otimed
       @logged
       def factorial(n):
           from operator import mul
           from functools import reduce
           return reduce(mul, range(1, n+1))
[252]: factorial(10)
      factorial: called 2023-06-14 15:51:47.288395+00:00
      factorial ran for 0.000203s
[252]: 3628800
      To make this clearer, let's write two very simple decorators as follows:
[253]: def dec_1(fn):
           def inner():
               print('running dec_1')
               return fn()
           return inner
[254]: def dec_2(fn):
           def inner():
               print('running dec_2')
               return fn()
           return inner
```

```
[255]: @dec_1
    @dec_2
    def my_func():
        print('running my_func')
```

```
[256]: my_func()
```

```
running dec_1
running dec_2
running my_func
```

But if we change the order of the decorators:

```
[257]: @dec_2
@dec_1
def my_func():
    print('running my_func')
```

```
[258]: my_func()
```

```
running dec_2
running dec_1
running my_func
```

You may wonder whether this really matters in practice. And yes, it can.

Consider an API that contains various functions that can be called. However, endpoints are secured and can only be run by authenticated users who have some specific role(s). If they do not have the role you want to return an unauthorized error. But if they do, then you want to log that they called the endpoint.

In this case you may have one decorator that is used to check authentication and permissions (and immediately return an unauthorized error from the API if applicable), and the other to log the call.

If you decorated it this way:

then the call would always be logged.

But, in this instance:

your endpoint would only get logged if the user passed the authorize test.

0.0.11 Decorators Application (Memoization)

Let's go back to our Fibonacci example:

```
[259]: def fib(n):
    print ('Calculating fib({0})'.format(n))
    return 1 if n < 3 else fib(n-1) + fib(n-2)</pre>
```

When we run this, we see that it is quite inefficient, as the same Fibonacci numbers get calculated multiple times:

```
[260]: fib(6)
      Calculating fib(6)
      Calculating fib(5)
      Calculating fib(4)
      Calculating fib(3)
      Calculating fib(2)
      Calculating fib(1)
      Calculating fib(2)
      Calculating fib(3)
      Calculating fib(2)
      Calculating fib(1)
      Calculating fib(4)
      Calculating fib(3)
      Calculating fib(2)
      Calculating fib(1)
      Calculating fib(2)
[260]: 8
      It would be better if we could somehow "store" these results, so if we have calculated fib(4)
```

It would be better if we could somehow "store" these results, so if we have calculated fib(4) and fib(3) before, we could simply recall the these values when calculating fib(5) = fib(4) + fib(3) instead of recalculating them.

This concept of improving the efficiency of our code by caching pre-calculated values so they do not need to be re-calculated every time, is called "memoization"

We can approach this using a simple class and a dictionary that stores any Fibonacci number that's already been calculated:

```
[261]: class Fib:
    def __init__(self):
        self.cache = {1: 1, 2: 1}

    def fib(self, n):
        if n not in self.cache:
            print('Calculating fib({0})'.format(n))
            self.cache[n] = self.fib(n-1) + self.fib(n-2)
        return self.cache[n]

[262]: f = Fib()

[263]: f.fib(1)

[263]: f.fib(6)
```

Calculating fib(6)

```
Calculating fib(5)
      Calculating fib(4)
      Calculating fib(3)
[264]: 8
[265]: f.fib(7)
      Calculating fib(7)
[265]: 13
      Let's see how we could rewrite this using a closure:
[266]: def fib():
           cache = \{1: 1, 2: 2\}
           def calc_fib(n):
                if n not in cache:
                    print('Calculating fib({0})'.format(n))
                    cache[n] = calc_fib(n-1) + calc_fib(n-2)
               return cache[n]
           return calc_fib
[267]: f = fib()
[268]: f(10)
      Calculating fib(10)
      Calculating fib(9)
      Calculating fib(8)
      Calculating fib(7)
      Calculating fib(6)
      Calculating fib(5)
      Calculating fib(4)
      Calculating fib(3)
[268]: 89
      Now let's see how we would implement this using a decorator:
[269]: from functools import wraps
       def memoize_fib(fn):
           cache = dict()
           @wraps(fn)
```

```
def inner(n):
               if n not in cache:
                   cache[n] = fn(n)
               return cache[n]
           return inner
[270]: @memoize_fib
       def fib(n):
           print ('Calculating fib({0})'.format(n))
           return 1 if n < 3 else fib(n-1) + fib(n-2)
[271]: fib(3)
      Calculating fib(3)
      Calculating fib(2)
      Calculating fib(1)
[271]: 2
[272]: fib(10)
      Calculating fib(10)
      Calculating fib(9)
      Calculating fib(8)
      Calculating fib(7)
      Calculating fib(6)
      Calculating fib(5)
      Calculating fib(4)
[272]: 55
[273]: fib(6)
[273]: 8
```

As you can see, we are hitting the cache when the values are available.

Now, we made our memoization decorator "hardcoded" to single argument functions - we could make it more generic.

For example, to handle an arbitrary number of positional arguments and keyword-only arguments we could do the following:

```
if args not in cache:
                    cache[args] = fn(*args)
               return cache[args]
           return inner
[275]: @memoize
       def fib(n):
           print ('Calculating fib({0})'.format(n))
           return 1 if n < 3 else fib(n-1) + fib(n-2)
[276]: fib(6)
      Calculating fib(6)
      Calculating fib(5)
      Calculating fib(4)
      Calculating fib(3)
      Calculating fib(2)
      Calculating fib(1)
[276]: 8
[277]: fib(7)
      Calculating fib(7)
[277]: 13
      Of course, with this rather generic memoization decorator we can memoize other functions too:
[278]: def fact(n):
           print('Calculating {0}!'.format(n))
           return 1 if n < 2 else n * fact(n-1)
[279]: fact(5)
      Calculating 5!
      Calculating 4!
      Calculating 3!
      Calculating 2!
      Calculating 1!
[279]: 120
[280]: fact(5)
      Calculating 5!
      Calculating 4!
```

```
Calculating 3!
      Calculating 2!
      Calculating 1!
[280]: 120
      And memoizing it:
[281]: @memoize
       def fact(n):
           print('Calculating {0}!'.format(n))
           return 1 if n < 2 else n * fact(n-1)
[282]: fact(6)
      Calculating 6!
      Calculating 5!
      Calculating 4!
      Calculating 3!
      Calculating 2!
      Calculating 1!
[282]: 720
[283]: fact(6)
[283]: 720
```

Our simple memoizer has a drawback however: * the cache size is unbounded - probably not a good thing! In general we want to limit the cache to a certain number of entries, balancing computational

efficiency vs memory utilization. * we are not handling **kwargs

Memoization is such a common thing to do that Python actually has a memoization decorator built for us!

It's in the, you guessed it, **functools** module, and is called **lru_cache** and is going to be quite a bit more efficient compared to the rudimentary memoization example we did above.

[LRU Cache = Least Recently Used caching: since the cache is not unlimited, at some point cached entries need to be discarded, and the least recently used entries are discarded first]

```
[284]: from functools import lru_cache

[285]: @lru_cache()
    def fact(n):
        print("Calculating fact({0})".format(n))
        return 1 if n < 2 else n * fact(n-1)</pre>
[286]: fact(5)
```

```
Calculating fact(5)
      Calculating fact(4)
      Calculating fact(3)
      Calculating fact(2)
      Calculating fact(1)
[286]: 120
[287]: fact(4)
[287]: 24
      As you can see, fact(4) was returned via a cached entry!
      Same thing with our Fibonacci function:
[288]: 0lru_cache()
       def fib(n):
           print("Calculating fib({0})".format(n))
           return 1 if n < 3 else fib(n-1) + fib(n-2)
[289]: fib(6)
      Calculating fib(6)
      Calculating fib(5)
      Calculating fib(4)
      Calculating fib(3)
      Calculating fib(2)
      Calculating fib(1)
[289]: 8
[290]: fib(5)
[290]: 5
      Recall from a few videos back that we timed the calculation for Fibonacci numbers. Calculating
      fib(35) took several seconds - every time...
[291]: from time import perf_counter
[292]: def fib_no_memo(n):
           return 1 if n < 3 else fib_no_memo(n-1) + fib_no_memo(n-2)
[293]: start = perf_counter()
       result = fib no memo(35)
       print("result={0}, elapsed: {1}s".format(result, perf_counter() - start))
```

result=9227465, elapsed: 1.2736941000039224s

```
[294]: @lru_cache()
def fib_memo(n):
    return 1 if n < 3 else fib_memo(n-1) + fib_memo(n-2)</pre>
```

```
[295]: start = perf_counter()
    result = fib_memo(35)
    print("result={0}, elapsed: {1}s".format(result, perf_counter() - start))
```

result=9227465, elapsed: 7.630000618519261e-05s

And if we make the calls again:

```
[296]: start = perf_counter()
    result = fib_no_memo(35)
    print("result={0}, elapsed: {1}s".format(result, perf_counter() - start))
```

result=9227465, elapsed: 1.3363394999978482s

```
[297]: start = perf_counter()
    result = fib_memo(35)
    print("result={0}, elapsed: {1}s".format(result, perf_counter() - start))
```

result=9227465, elapsed: 6.919999577803537e-05s

You may have noticed that the <code>lru_cache</code> decorator was implemented using () - we'll see more on this later, but that's because decorators can themselves have parameters (beyond the function being decorated).

One of the arguments to the <code>lru_cache</code> decorator is the size of the cache - it defaults to 128 items, but we can easily change this - for performance reasons use powers of 2 for the cache size (or None for unbounded cache):

```
[298]: @lru_cache(maxsize=8)
def fib(n):
    print("Calculating fib({0})".format(n))
    return 1 if n < 3 else fib(n-1) + fib(n-2)</pre>
```

```
[299]: fib(10)
```

```
Calculating fib(10)
Calculating fib(9)
Calculating fib(8)
Calculating fib(7)
Calculating fib(6)
Calculating fib(5)
Calculating fib(4)
Calculating fib(3)
Calculating fib(2)
Calculating fib(1)
```

[299]: 55 [300]: fib(20) Calculating fib(20) Calculating fib(19) Calculating fib(18) Calculating fib(17) Calculating fib(16) Calculating fib(15) Calculating fib(14) Calculating fib(13) Calculating fib(12) Calculating fib(11) [300]: 6765 [301]: fib(10) Calculating fib(10) Calculating fib(9) Calculating fib(8) Calculating fib(7) Calculating fib(6) Calculating fib(5) Calculating fib(4) Calculating fib(3) Calculating fib(2) Calculating fib(1) [301]: 55

Note how Python had to recalculate fib for 10, 9, etc. This is because the cache can only contain 10 items, so when we calculated fib(20), it stored fib for 20, 19, ..., 11 (10 items) and therefore the oldest items fib 10, 9, ..., 1 were removed from the cache to make space.

[]:

0.0.12 Decorators 2

We have seen how to create some simple and not so simple decorators.

However we have also been using built-in decorators that can accept parameters, such as wraps and lru_cache.

This can be quite useful and we can accomplish the same thing ourselves.

First recall our original timer decorator from an earlier video (Decorator Application - Timer):

```
[302]: def timed(fn):
    from time import perf_counter

    def inner(*args, **kwargs):
        start = perf_counter()
        result = fn(*args, **kwargs)
        end = perf_counter()
        elapsed = end - start
        print('Run time: {0:.6f}s'.format(elapsed))
        return result

return inner
```

```
[303]: def calc_fib_recurse(n):
    return 1 if n < 3 else calc_fib_recurse(n-1) + calc_fib_recurse(n-2)

def fib(n):
    return calc_fib_recurse(n)</pre>
```

We can decorate our Fibonacci function using the **@** syntax, or the longer syntax as follows:

```
[304]: fib = timed(fib)
```

```
[305]: fib(30)
```

Run time: 0.109675s

[305]: 832040

Let's modify this so the timer runs the function multiple times and calculates the average run time:

```
[306]: def timed(fn):
    from time import perf_counter

    def inner(*args, **kwargs):
        total_elapsed = 0
        for i in range(10):
            start = perf_counter()
            result = fn(*args, **kwargs)
            end = perf_counter()
            total_elapsed += (perf_counter() - start)
            avg_elapsed = total_elapsed / 10
            print('Avg Run time: {0:.6f}s'.format(avg_elapsed))
            return inner
```

And again we decorate it using the long syntax:

```
[307]: def fib(n):
    return calc_fib_recurse(n)

fib = timed(fib)
```

```
[308]: fib(28)
```

Avg Run time: 0.043116s

[308]: 317811

But that value of 10 has been hardcoded. Let's make it a parameter instead.

Now to decorate our Fibonacci function we **have** to use the long syntax (as we saw in the lecture, the **@** syntax will not work):

```
[310]: def fib(n):
    return calc_fib_recurse(n)

fib = timed(fib, 5)
```

```
[311]: fib(28)
```

Avg Run time: 0.041800s (5 reps)

[311]: 317811

The problem is that we cannot use the @ decorator syntax because when using that syntax Python passes a **single** argument to the decorator: the function we are decorating - nothing else.

Of course we could just use what we did above, but the decorator syntax is kind of neat, so it would be nice to retain the ability to use it.

We just need to change our thinking a little bit to do this:

First, when we see the following syntax:

```
@dec def my_func(): pass
```

we see that dec must be a function that takes a single argument, the function being decorated.

You'll note that dec is just a function, but we do not call dec when we decorate my_func, we simply use the label dec.

Then Python does:

```
my_func = dec(my_func)
```

Let's try a concrete example:

```
[312]: def dec(fn):
    print ("running dec")

    def inner(*args, **kwargs):
        print("running inner")
        return fn(*args, **kwargs)

    return inner
```

```
[313]: @dec
def my_func():
    print('running my_func')
```

running dec

As we can see, when we decorated my_func, the dec function was called at that time.

(Because Python did this:

```
my_func = dec(my_func)
```

so dec was called)

And when we now call my_func, we see that the inner function is called, followed by the original my_func

```
[314]: my_func()
```

running inner
running my_func

But what if dec was not the decorator itself, but instead created and returned a decorator?

Let's see how we might do this:

```
[315]: def dec_factory():
    print('running dec_factory')
    def dec(fn):
```

```
print('running dec')
  def inner(*args, **kwargs):
      print('running inner')
      return fn(*args, **kwargs)
  return inner
return dec
```

So as you can see, calling dec_generator() will return that dec function which is our decorator:

```
[316]: @dec_factory()
def my_func(a, b):
    print(a, b)
```

running dec_factory
running dec

You can see that both dec_generator and dec were already called.

```
[317]: my_func(10, 20)
```

running inner 10 20

And there you go, all we did is basically create a decorator by calling a function (dec_factory) and use the return value of that call (the dec function) as our actual decorator.

We could have done the decoration this way too:

```
[318]: dec = dec_factory()
```

running dec_factory

```
[319]: @dec
def my_func():
    print('running my_func')
```

running dec

```
[320]: my_func()
```

running inner
running my_func

Or even this way:

```
[321]: dec = dec_factory()

def my_func():
    print('running my_func')
```

```
my_func = dec(my_func)
      running dec_factory
      running dec
[322]: my_func()
      running inner
      running my_func
      Of course we could even decorate it this way using a single statement:
[323]: def my_func():
           print('running my_func')
       my_func = dec_factory()(my_func)
      running dec_factory
      running dec
[324]: my_func()
      running inner
      running my_func
      OK, so now we have decorated our function using, not a decorator, but a decorator factory as
      follows:
[325]: def dec_factory():
           def dec(fn):
               def inner(*args, **kwargs):
                    print('running decorator inner')
                    return fn(*args, **kwargs)
               return inner
           return dec
[326]: @dec_factory()
       def my_func(a, b):
           return a + b
[327]: my_func(10, 20)
      running decorator inner
```

[327]: 30

You should note that in this approach, we are **calling dec_factory()**, [note the parentheses ()], and **then** using the return value (a decorator) to decorate our function.

So, we could pass arguments as we do so without affecting the final outcome. In fact we can even access them from anywhere inside dec_factory, including any of the nested functions!

Let's try this:

```
[328]: def dec_factory(a, b):
    def dec(fn):
        def inner(*args, **kwargs):
            print('running decorator inner')
            print('free vars: ', a, b) # a and b are free variables!
            return fn(*args, **kwargs)
        return inner
    return dec
```

```
[329]: @dec_factory(10, 20)
def my_func():
    print('python rocks')
```

```
[330]: my_func()
```

```
running decorator inner free vars: 10 20 python rocks
```

And this is how we can create decorators with parameters. We do not directly create a decorator, instead we use an outer function that returns a decorator when called, and pass arguments to that outer function, which the decorator and its inner function can of course access as nonlocal (free) variables.

So now, let's go back to our original problem where we wanted our timing decorator to run a number of loops which could be specified as a parameter when decorating the function we want to time.

Here it is again:

```
return inner
```

So, all we need to do is create an outer function around our timed decorator, and pass the num_reps argument to that outer function instead:

```
[332]: def timed_factory(num_reps=1):
           def timed(fn):
               from time import perf counter
               def inner(*args, **kwargs):
                   total_elapsed = 0
                   for i in range(num_reps):
                       start = perf_counter()
                       result = fn(*args, **kwargs)
                       end = perf_counter()
                       total_elapsed += (perf_counter() - start)
                   avg_elapsed = total_elapsed / num_reps
                   print('Avg Run time: {0:.6f}s ({1} reps)'.format(avg_elapsed,
                                                                    num_reps))
                   return result
               return inner
           return timed
```

```
[333]: @timed_factory(5)
def fib(n):
    return calc_fib_recurse(n)
```

```
[334]: fib(30)
```

Avg Run time: 0.113834s (5 reps)

[334]: 832040

Just to put the finishing touch on this, we probably don't want to have our outer function named the way it is (timed_factory). Instead we probably just want to call it timed. So lets just do this final part:

```
[336]: @timed(5)
def fib(n):
    return calc_fib_recurse(n)
```

```
[337]: fib(30)
```

Avg Run time: 0.110944s (5 reps)

[337]: 832040

0.0.13 Decorator Application (Decorator Class)

If you recalls how we wrote a parameterized decorator, we had to write a decorator factory that took in the arguments for our decorator and then returned the decorator (which could reference the arguments as free variables).

Very simply:

```
[338]: def my_dec(a, b):
    def dec(fn):
        def inner(*args, **kwargs):
            print('decorated function called: a={0}, b={1}'.format(a, b))
            return fn(*args, **kwargs)
        return inner
    return dec
```

```
[339]: @my_dec(10, 20)
def my_func(s):
    print('hello {0}'.format(s))
```

```
[340]: my_func('world')
```

```
decorated function called: a=10, b=20 hello world
```

So, our decorator factory was passed some arguments, and returned a callable which took one single parameter, the function being decorated, but also had access to the arguments passed to the factory.

Now, recall that we can make our class instances callable, simply by implementing the __call__ method.

Here's a simple example:

```
[341]: class MyClass:
    def __init__(self, a, b):
        self.a = a
        self.b = b

    def __call__(self):
        print('MyClass instance called: a={0}, b={1}'.format(self.a, self.b))
```

```
[342]: my_class = MyClass(10, 20)
```

```
[343]: my_class()
```

MyClass instance called: a=10, b=20

So let's modify this just a bit, and have the __call__ method be our decorator!

So, we can decorate our functions this way:

```
[345]: OMyClass(10, 20)
def my_func(s):
    print('Hello {0}!'.format(s))
```

Remember that <code>@MyClass(10, 20)</code> returned an object of type <code>MyClass</code>. But that object is itself callable, so we could do something like:

```
my_func = MyClass(10, 20)(my_func)
or, more simply
@MyClass(10, 20) def my_func(s): print(s)

[346]: my_func('Python')
```

```
MyClass instance called: a=10, b=20 Hello Python!
```

So as you can see, we can also use callable classes to decorate functions!

0.0.14 Decorator Application: Decorating Classes

We have so far worked with decorating functions. This means we can decorate functions defined with a def statement (we can use the @ syntax, or the long form). Since class methods are functions, they can be decorated too. Lambda expressions can also be decorated (using the long form).

But if you think about how our decorators work, they take a single parameter, a function, and return some other function - usually a closure that uses the original function that was passed as an argument.

We could use the same concept to accept, not a function, but a class instead. We could reference that class inside our decorator, modify it, and then return that modified class.

First we look at something called **monkey patching**. It boils down to modifying or extending our code at **run time**.

For example we can modify or add attributes to classes at run time. Modules too.

In Python, many of the classes we use can be modified at run time (built-ins like strings, lists, and so on, cannot).

But classes written in Python, such as the ones we write, and even library classes, as long as they are written in Python, not C, can. For example Fraction in the fractions module can be monkey patched.

Just because we can do something however, does not mean we should! Monkey patching can be extremely useful, but don't do it just because you can - as always there should be a real reason to do it, as we'll see in a bit.

Also, in general it is a bad idea to monkey patch the special methods __???__ (such as __len__) as this will often not work due to how these methods are searched for by Python.

```
[347]: from fractions import Fraction
[348]: Fraction.speak = lambda self: 'This is a late parrot.'
[349]: f = Fraction(2, 3)
[350]: f
[350]: Fraction(2, 3)
[351]: f.speak()
[351]: 'This is a late parrot.'
```

Yes, this is obviously nonsense, but you get the idea that you can add attributes to classes even if you do not have direct control over the class, or after your class has been defined.

If you want a more useful method, how about one that tells us if the Fraction is an integral number? (i.e. denominator is 1)

```
[352]: Fraction.is_integral = lambda self: self.denominator == 1
[353]: f1 = Fraction(1, 2)
       f2 = Fraction(10, 5)
[354]: f1.is_integral()
[354]: False
[355]: f2.is integral()
[355]: True
      Now, we can make this change to the class by calling a function to do it instead:
[356]: def dec speak(cls):
           cls.speak = lambda self: 'This is a very late parrot.'
           return cls
[357]: Fraction = dec_speak(Fraction)
       (Hopefully the above code reminds you of decorators.)
[358]: f = Fraction(10, 2)
[359]: f.speak()
[359]: 'This is a very late parrot.'
      We can use that function to decorate our custom classes too, using the short **@** syntax too.
[360]: @dec_speak
       class Parrot:
           def __init__(self):
                self.state = 'late'
[361]: polly = Parrot()
[362]: polly.speak()
[362]: 'This is a very late parrot.'
```

Using this technique we could for example add a useful *reciprocal* attribute to the Fraction class, but of course since it would probably be a one time kind of thing (how many Fraction classes are there that you will want to add a reciprocal to after all), there's no need for decorators. Decorators are useful when they are able to be reused in more general ways.

```
[363]: Fraction.recip = lambda self: Fraction(self.denominator, self.numerator)

[364]: f = Fraction(2,3)

[365]: f

[366]: Fraction(2, 3)

[366]: Fraction(3, 2)
```

These example are quite trivial, and not very useful.

So why bring this up?

Because this same technique can be used for more interesting things.

As a first example, let's say you typically like to inspect various properties of an object for debugging purposes, maybe the memory address, it's current state (property values), and the time at which the debug info was generated.

```
[367]: from datetime import datetime, timezone
[368]: def debug_info(cls):
           def info(self):
               results = []
               results.append('time: {0}'.format(datetime.now(timezone.utc)))
               results.append('class: {0}'.format(self.__class__.__name__))
               results.append('id: {0}'.format(hex(id(self))))
               if vars(self):
                   for k, v in vars(self).items():
                       results.append('{0}: {1}'.format(k, v))
               # we have not covered lists, the extend method and generators,
               # but note that a more Pythonic way to do this would be:
               #if vars(self):
                    results.extend('\{0\}: \{1\}'.format(k, v)
                                   for k, v in vars(self).items())
               return results
           cls.debug = info
```

```
return cls
[369]: @debug info
       class Person:
           def __init__(self, name, birth_year):
               self.name = name
               self.birth_year = birth_year
           def say_hi():
               return 'Hello there!'
[370]: p1 = Person('John', 1939)
[371]: p1.debug()
[371]: ['time: 2023-06-14 15:52:35.983201+00:00',
        'class: Person',
        'id: 0x2417af4b110',
        'name: John',
        'birth_year: 1939']
      And of course we can decorate other classes this way too, not just a single class:
[372]: @debug_info
       class Automobile:
           def __init__(self, make, model, year, top_speed_mph):
               self.make = make
               self.model = model
               self.year = year
               self.top_speed_mph = top_speed_mph
               self.current_speed = 0
           @property
           def speed(self):
               return self.current_speed
           @speed.setter
           def speed(self, new_speed):
               self.current_speed = new_speed
[373]: s = Automobile('Ford', 'Model T', 1908, 45)
[374]: s.debug()
[374]: ['time: 2023-06-14 15:52:36.159056+00:00',
        'class: Automobile',
```

```
'id: 0x2417af28d50',
        'make: Ford',
        'model: Model T',
        'year: 1908',
        'top_speed_mph: 45',
        'current_speed: 0']
[375]: s.speed = 20
[376]: s.debug()
[376]: ['time: 2023-06-14 15:52:36.284360+00:00',
        'class: Automobile',
        'id: 0x2417af28d50',
        'make: Ford',
        'model: Model T',
        'year: 1908',
        'top_speed_mph: 45',
        'current_speed: 20']
      Let's look at another example where decorating an entire class could be useful.
[377]: from math import sqrt
[378]: class Point:
           def __init__(self, x, y):
               self.x = x
               self.y = y
           def __abs__(self):
               return sqrt(self.x**2 + self.y**2)
           def __repr__(self):
               return 'Point({0},{1})'.format(self.x, self.y)
[379]: p1, p2, p3 = Point(2, 3), Point(2, 3), Point(0,0)
[380]: abs(p1)
[380]: 3.605551275463989
[381]: p1, p2
[381]: (Point(2,3), Point(2,3))
[382]: p1 == p2
```

[382]: False

Hmm, we probably would have expected p1 to be equal to p2 since it has the same coordinates. But by default Python will compare memory addresses, since our class does not implement the __eq__ method used for == comparisons.

So, that class does not support the comparison operators such as <, <=, etc.

Even == does not work as expected - it will use the memory address instead of using a comparison of the x and y coordinates as we might probably expect.

For the < operator, we need our class to implement the __lt__ method, and for == we need the __eq__ method.

Other comparison operators are supported by implementing a variety of functions such as $_{le}_{-}$ (<=), $_{le}_{-}$ (>), $_{le}_{-}$ (>=).

We are going to add the __lt__ and __eq__ methods to our Point class.

We will consider a Point object to be smaller than another one if it is closer to the origin (i.e. smaller magnitude).

```
[385]: del Point

class Point:
    def __init__(self, x, y):
        self.x = x
        self.y = y

    def __abs__(self):
        return sqrt(self.x**2 + self.y**2)

    def __eq__(self, other):
        if isinstance(other, Point):
            return self.x == other.x and self.y == other.y
        else:
```

```
return NotImplemented
           def __lt__(self, other):
                if isinstance(other, Point):
                    return abs(self) < abs(other)</pre>
                else:
                    return NotImplemented
           def __repr__(self):
                return '{0}({1},{2})'.format(self.__class__.__name__, self.x, self.y)
[386]: p1, p2, p3 = Point(2, 3), Point(2, 3), Point(0,0)
[387]: p1, p2, p1==p2
[387]: (Point(2,3), Point(2,3), True)
[388]: p2, p3, p2==p3
[388]: (Point(2,3), Point(0,0), False)
      As we can see, == now works as expected
[389]: p4 = Point(1, 2)
[390]: abs(p1), abs(p4), p1 < p4
[390]: (3.605551275463989, 2.23606797749979, False)
      Great, so now we have < and == implemented. What about the rest of the operators: <=, >, >=?
[391]: p1 > p4
[391]: True
      Ooh, since we have implemented < and ==, does this mean Python magically implemented a >
      operator (i.e. not < and not ==)?
      Not exactly! What happened is that since p1 and p4 are both points, running the comparison p1
      > p4 is really the same as evaluating p4 < p1 - and Python did do that automatically for us.
      But it has not implemented any of the others, such as >= and <=:
[392]: p1 <= p4
        TypeError
                                                     Traceback (most recent call last)
        Cell In[392], line 1
```

----> 1 p1 <= p4

```
TypeError: '<=' not supported between instances of 'Point' and 'Point'</pre>
```

Now, although we could proceed in a similar way and define >=, <= and > using the same technique, observe that if < and == is defined then:

```
a <= b iff a < b or a == b</li>
a > b iff not(a<b) and a != b</li>
a >= b iff not(a<b)</li>
```

So, to be quite generic we could create a decorator that will implement these last three operators as long as == and < are defined. We could then decorate **any** class that implements just those two operators.

```
[393]: def complete_ordering(cls):
    if '__eq__' in dir(cls) and '__lt__' in dir(cls):
        cls.__le__ = lambda self, other: self < other or self == other
        cls.__gt__ = lambda self, other: not(self < other) and not (self ==_u
        cother)
        cls.__ge__ = lambda self, other: not (self < other)
        return cls</pre>
```

In reality, the code above is **NOT** a good implementation at all. We are not checking that the types are compatible and returning a NotImplemented result if appropriate. I am also using inline operators (< and ==) instead of the dunder functions (__lt__ and __eq__). I just kept it simple because we'll use a better alternative in a bit.

For example, a better way to implement __ge__ would be as follows:

```
[394]: def ge_from_lt(self, other):
    # self >= other iff not(other < self)
    result = self.__lt__(other)
    if result is NotImplemented:
        return NotImplemented
    else:
        return not result</pre>
```

You may be wondering why I used __lt__ instead of just using the < operator. This is because I want to actually look at the result of the operation without raising an exception if the operation is not implemented. The way I have the total ordering decorator implemented could cause an infinite loop because when I evaluate self < other, if an exception is raised, Python will reflect the evaluation to other > self, and if that raises an error as well, Python will try to reflect that operation too, and we get into an infinite loop (which eventually terminates in a stack overflow). This was actually a bug in Python's standard library implementation of a complete_ordering decorator (called total_ordering) that was resolved in 3.4.

```
[395]: class Point:
    def __init__(self, x, y):
        self.x = x
```

```
self.y = y
           def __abs__(self):
               return sqrt(self.x**2 + self.y**2)
           def __eq__(self, other):
               if isinstance(other, Point):
                   return self.x == other.x and self.y == other.y
               else:
                   return NotImplemented
           def __lt__(self, other):
               if isinstance(other, Point):
                   return abs(self) < abs(other)</pre>
               else:
                   return NotImplemented
           def __repr__(self):
               return '{0}({1},{2})'.format(self.__class__, self.x, self.y)
[396]: Point = complete_ordering(Point)
[397]: p1, p2, p3 = Point(1, 1), Point(3, 4), Point(3, 4)
[398]: abs(p1), abs(p2), abs(p3)
[398]: (1.4142135623730951, 5.0, 5.0)
[399]: p1 < p2, p1 <= p2, p1 >= p2, p1 >= p2, p2 >= p3
[399]: (True, True, False, False, False, True)
      Now the complete_ordering decorator can also be directly applied to any class that defines __eq__
      and __lt__.
[400]: @complete_ordering
       class Grade:
           def __init__(self, score, max_score):
               self.score = score
               self.max_score = max_score
               self.score_percent = round(score / max_score * 100)
           def __repr__(self):
               return 'Grade({0}, {1})'.format(self.score, self.max_score)
           def __eq__(self, other):
               if isinstance(other, Grade):
```

```
return self.score_percent == other.score_percent
else:
    return NotImplemented

def __lt__(self, other):
    if isinstance(other, Grade):
        return self.score_percent < other.score_percent
else:
        return NotImplemented</pre>
```

```
[401]: g1 = Grade(10, 100)
g2 = Grade(20, 30)
g3 = Grade(5, 50)
```

```
[402]: g1 <= g2, g1 == g3, g2 > g3
```

[402]: (True, True, True)

Often, given the == operator and just **one** of the other comparison operators (<, <=, >=), then all the rest can be derived.

Our decorator insisted on == and <. but we could make it better by insisting on == and any one of the other operators. This will of course make our decorator more complicated, and in fact, Python has this precise functionality built in to the, you guessed it, functools module!

It is a decorator called total_ordering.

Let's see it in action:

```
[403]: from functools import total_ordering
```

```
return self.score_percent < other.score_percent
else:
   return NotImplemented</pre>
```

```
[405]: g1, g2 = Grade(80, 100), Grade(60, 100)
```

```
[406]: g1 >= g2, g1 > g2
```

[406]: (True, True)

Or we could also do it this way:

```
[407]: @total_ordering
       class Grade:
           def __init__(self, score, max_score):
               self.score = score
               self.max_score = max_score
               self.score_percent = round(score / max_score * 100)
           def __repr__(self):
               return 'Grade({0}, {1})'.format(self.score, self.max_score)
           def __eq__(self, other):
               if isinstance(other, Grade):
                   return self.score_percent == other.score_percent
               else:
                   return NotImplemented
           def __gt__(self, other):
               if isinstance(other, Grade):
                   return self.score_percent > other.score_percent
               else:
                   return NotImplemented
```

```
[408]: g1, g2 = Grade(80, 100), Grade(60, 100)
```

```
[409]: g1 \ge g2, g1 \ge g2, g1 \le g2, g1 \le g2
```

[409]: (True, True, False, False)

0.0.15 Decorator Application: Single Dispatch Generic Functions

Consider an application where we want to provide similar functionality but that varies slightly depending on the argument types passed in.

In this set of examples we consider this problem where functionality differs based on a single argument's type (hence single dispatch) instead of the type of multiple arguments (which would be multi dispatch)

If you have a background in some other OO languages such as Java or C#, you'll know that we can easily do something like this by basically **overloading** functions: using a different data type for the function parameter, hence changing the function signature. Then although the name of the function is the same, calling do_something(100) and do_something('java') would call a different function, the first one would call the do_something(int) function, and the second would call the do_something(String) function.

Of course, Python is not statically typed, so even if Python had function overloading built-in, we would not be able to make such a distinction in our function signatures since there is nothing that says that a parameter must be of a specific type, so in a best case scenario we would have to "distinguish" functions with the same name only by the number of parameters they take. And then we'd have to somehow deal with variable numbers of positional and keyword arguments too... Uuugh! In any event, single dispatch could never work.

Instead we have to come up with a different solution.

Let's say we want to display various data types in html format, with different presentations for integers (we want both base 10 and hex values), floats (we always want it rounded to 2 decimal points), strings (we want the string html-escaped, and all newline characters replaced by
br/>), lists and tuples should be implemented using bulleted lists, and the same with dictionaries except we want the name/value pair to be displayed in the bulleted list.

For starters, let's just implement individual functions to do each of those things.

I am going to keep the functions very simple, but in practice you should handle situations like None objects, empty lists and dictionaries, possibly the wrong type being passed to the function, etc.

```
[410]: from html import escape
      def html_escape(arg):
          return escape(str(arg))
      def html_int(a):
          return '{0}(<i>{1}</i)'.format(a, str(hex(a)))
      def html real(a):
          return '{0:.2f}'.format(round(a, 2))
      def html_str(s):
          return html_escape(s).replace('\n', '<br/>\n')
      def html list(l):
          items = ('{0}'.format(html_escape(item))
                   for item in 1)
          return '\n' + '\n'.join(items) + '\n'
      def html dict(d):
          items = ('{0}={1}'.format(html_escape(k), html_escape(v))
                   for k, v in d.items())
          return '\n' + '\n'.join(items) + '\n'
```

```
[411]: print(html_str("""this is
    a multi line string
    with special characters: 10 < 100"""))

this is <br/>   a multi line string<br/>   with special characters: 10 &lt; 100

[412]: print(html_int(255))

    255(<i>0xff</i)

[413]: print(html_escape(3+10j))

    (3+10j)</pre>
```

Ideally we would want to just have to call a single function, maybe htmlize that would figure out which particular flavor of the html_xxx function to call depending on the argument type.

We could try it as follows:

```
[414]: from decimal import Decimal

def htmlize(arg):
    if isinstance(arg, int):
        return html_int(arg)
    elif isinstance(arg, float) or isinstance(arg, Decimal):
        return html_real(arg)
    elif isinstance(arg, str):
        return html_str(arg)
    elif isinstance(arg, list) or isinstance(arg, tuple):
        return html_list(arg)
    elif isinstance(arg, dict):
        return html_dict(arg)
    else:
        # default behavior - just html escape string representation
        return html_escape(str(arg))
```

Now we can essentially use the same function call to handle different types - the htmlize function is a dispatcher - it dispatches the request to a different function based on the argument type. (There's a much better way to do some of this, but we'll have to wait until we cover abstract base classes to do so).

```
[415]: print(htmlize([1, 2, 3]))

    1>1
    2
```

```
<1i>3</1i>
      [416]: print(htmlize(dict(key1=1, key2=2)))
      <l
      key1=1
      key2=2
      [417]: print(htmlize(255))
      255(<i>0xff</i)
      But there are a number of shortcomings here:
[418]: print(htmlize(["""first element is
      a multi-line string""", (1, 2, 3)]))
      <l
      first element is
      a multi-line string
      (1, 2, 3)
      As you can see, the multi-line string did not get the newline characters replaced, the tuple was not
      rendered as an html list, and the integers do not have their hex representation.
      So we just need to redefine the html_list and html_dict functions to use the htmlize function:
[419]: def html_list(1):
          items = ['{0}'.format(htmlize(item)) for item in 1]
          return '\langle ul \rangle \ ' + ' \ ' \cdot join(items) + ' \ ' \cdot ' \ ' \cdot ' 
[420]: def html_dict(d):
          items = ['{0}={1}'.format(html_escape(k), htmlize(v)) for k, v in_{l}
        →d.items()]
          return '\n' + '\n'.join(items) + '\n'
[421]: print(htmlize(["""first element is
      a multi-line string""", (1, 2, 3)]))
      <l
      first element is <br/>
      a multi-line string
      1(<i>0x1</i)</li>
      2(<i>0x2</i)</li>
      3(<i>0x3</i)</li>
```

Much better, but hopefully you spotted something that might seem problematic!

Do we not have a circular reference?

In order to define html_list and html_dict we needed to call htmlize, but in order to define htmlize we needed to call html_list and html_dict.

Remember that in Python we can reference a function **inside** the body of another function **before** the function has been defined, as long as by the time we **call** the first function, the second one has been defined. SO this is actually OK.

If you don't believe me and want to make sure of this yourself, go ahead and reset your Kernel (click on the Kernel | Restart menu option), and run the following code without running anything prior to this.

The htmlize function body makes calls to other functions such as html_escape, html_int, etc that have not actually been defined yet

```
[422]: from html import escape
      from decimal import Decimal
      def htmlize(arg):
           if isinstance(arg, int):
               return html_int(arg)
           elif isinstance(arg, float) or isinstance(arg, Decimal):
               return html_real(arg)
           elif isinstance(arg, str):
               return html_str(arg)
           elif isinstance(arg, list) or isinstance(arg, tuple) or isinstance(arg, u
        ⇔set):
               return html_list(arg)
           elif isinstance(arg, dict):
               return html_dict(arg)
           else:
               # default behavior - just html escape string representation
               return html_escape(str(arg))
```

Now we define all the functions that htmlize uses before we actually call htmlize and all is good:

```
[423]: def html_escape(arg):
    return escape(str(arg))

def html_int(a):
    return '{0}(<i>{1}</i)'.format(a, str(hex(a)))

def html_real(a):
    return '{0:.2f}'.format(round(a, 2))</pre>
```

```
def html_str(s):
    return html_escape(s).replace('\n', '<br/>\n')

def html_list(l):
    items = ['{0}'.format(htmlize(item)) for item in l]
    return '\n' + '\n'.join(items) + '\n'

def html_dict(d):
    items = ['{0}={1}'.format(html_escape(k), htmlize(v)) for k, v in_u
d.items()]
    return '\n' + '\n'.join(items) + '\n'
```

```
[424]: print(htmlize(["""first element is a multi-line string""", (1, 2, 3)]))
```

```
  first element is <br/>
a multi-line string

  (i)
  1(<i>0x1</i)</li>
  (i)
  (i)
  (i)
  (ul>

  (ul>
```

As you can see this works just fine.

But we still have something undesirable. You'll notice that the dispatch function htmlize needs to have this big if...elif...else statement that will just keep growing as we need to handle more and more types (including potentially custom types).

This will just get unwieldy, and not very flexible (every time someone creates a new type that has to have a special html representation they will need to go into the htmlize function and modify it.

So instead, we are going to try a more flexible approach using decorators.

The way we are going to approach this is to create a dispatcher function, and then separately "register" each type-specific function with the dispatcher.

First, we are going to create a decorator that will do something that may seem kind of silly - it is going to take the decorated function and store it in a dictionary, using a key consisting of the **type** object.

Then when the returned closure is called, the closure will call the function stored in that dictionary.

```
[425]: def singledispatch(fn):
    registry = dict()
    registry[object] = fn

    def inner(arg):
        return registry[object](arg)
```

```
return inner

[426]: @singledispatch
    def htmlizer(arg):
        return escape(str(arg))

[427]: htmlizer('a < 10')</pre>
```

[427]: 'a < 10'

Next, we are going to add some functions to that **registry** dictionary, and modify our inner function to choose the correct function from the registry, or pick a default based on the type of the argument:

```
[428]: def singledispatch(fn):
    registry = dict()

    registry[object] = fn
    registry[int] = lambda arg: '{0}(<i>{1}</i)'.format(arg, str(hex(arg)))
    registry[float] = lambda arg: '{0:.2f}'.format(round(arg, 2))

    def inner(arg):
        fn = registry.get(type(arg), registry[object])
        return fn(arg)
    return inner</pre>
```

```
[429]: @singledispatch
def htmlize(a):
    return escape(str(a))
```

```
[430]: htmlize(10)
```

[430]: '10(<i>0xa</i)'

```
[431]: htmlize(3.1415)
```

[431]: '3.14'

Now, we want a way to add the specialized functions to the registry dictionary from **outside** the singledispatch function - to do so we will create a parametrized decorator that will (1) take the type as a parameter, and (2) return a closure that will decorate the function associated with the type:

```
[432]: def singledispatch(fn):
    registry = dict()

    registry[object] = fn
```

```
def register(type_):
    def inner(fn):
        registry[type_] = fn
    return inner

def decorator(arg):
    fn = registry.get(type(arg), registry[object])
    return fn(arg)

return decorator
```

But of course this is not good enough - how do we get a hold of the register function from outside singledispatch? Remember, singledispatch is a decorator that returns the decorated closure, not the register closure.

We can do this by adding the **register** function as an **attribute** of the **decorated** function before we return it.

While we're at it we're also going to:

- add the registry dictionary as an attribute as so we can look into it to see what it contains.
- add another function that given a type will return the function associated with that type (or the default function if the type is not found in the dictionary)

```
[433]: def singledispatch(fn):
           registry = dict()
           registry[object] = fn
           def register(type_):
               def inner(fn):
                   registry[type_] = fn
                   return fn # we do this so we can stack register decorators!
               return inner
           def decorator(arg):
               fn = registry.get(type(arg), registry[object])
               return fn(arg)
           def dispatch(type_):
               return registry.get(type_, registry[object])
           decorator.register = register
           decorator.registry = registry.keys()
           decorator.dispatch = dispatch
           return decorator
```

```
[434]: @singledispatch
def htmlize(arg):
    return escape(str(arg))
```

And we can see that htmlize (that returned inner) function has an attribute called register:

```
[435]: htmlize.register
```

```
[435]: <function __main__.singledispatch.<locals>.register(type_)>
```

as well as that registry attribute that we put in just we could see what keys are in the registry dictionary:

```
[436]: htmlize.registry
```

```
[436]: dict_keys([<class 'object'>])
```

We can also ask it what function it is going to use for any specific type (currently we only have one registered, the default, for the most general object type):

```
[437]: htmlize.dispatch(str)
```

```
[437]: <function __main__.htmlize(arg)>
```

And you'll note that the extended scope of register and dispatch is the same as the extended scope of htmlize.

So now we can register some functions (it will store the function with associated data type in the registry dictionary):

```
[438]: @htmlize.register(int)
def html_int(a):
    return '{0}(<i>{1}</i)'.format(a, str(hex(a)))</pre>
```

We can peek into the registered types:

```
[439]: htmlize.registry
```

```
[439]: dict_keys([<class 'object'>, <class 'int'>])
```

and we can ask the decorated htmlize function what function it is going to use for the int type:

```
[440]: htmlize.dispatch(int)
```

```
[440]: <function __main__.html_int(a)>
```

and we can actually call it as well:

```
[441]: htmlize(100)
```

```
[441]: '100(<i>0x64</i)'
```

The huge advantage now is that we can keep registering new handlers from anywhere in our module, or even from outside our module!

```
[442]: Ohtmlize.register(float)
      def html_real(a):
          return '{0:.2f}'.format(round(a, 2))
      @htmlize.register(str)
      def html_str(s):
          return escape(s).replace('\n', '<br/>\n')
      Ohtmlize.register(tuple)
      Ohtmlize.register(list)
      def html_list(1):
          items = ['{0}'.format(htmlize(item)) for item in 1]
          return '\n' + '\n'.join(items) + '\n'
      Ohtmlize.register(dict)
      def html_dict(d):
          items = ['{0}={1}'.format(htmlize(k), htmlize(v)) for k, v in d.
       →items()]
          return '\n' + '\n'.join(items) + '\n'
[443]: htmlize.registry
[443]: dict_keys([<class 'object'>, <class 'int'>, <class 'float'>, <class 'str'>,
      <class 'list'>, <class 'tuple'>, <class 'dict'>])
[444]: print(htmlize([1, 2, 3]))
     <l
     1(<i>0x1</i)</li>
     2(<i>0x2</i)</li>
      3(<i>0x3</i)</li>
     [445]: print(htmlize((1, 2, 3)))
     ul>
      1(<i>0x1</i)</li>
      2(<i>0x2</i)</li>
      3(<i>0x3</i)</li>
     </111>
[446]: print(htmlize("""this
      is a multi line string with
```

```
a < 10"""))
```

this
is a multi line string with
a < 10

Our single dispatch decorator works quite well - but it has some limitations. For example it cannot handle functions that take in more than one argument (in which case dispatching would be based on the type of the **first** argument), and we also are not allowing for types based on parent classes - for example, integers and booleans are both integral numbers - i.e. they both inherit from the Integral base class. Similarly lists and tuples are both more generic Sequence types. We'll see this in more detail when we get to the topic of abstract base classes (ABC's).

```
[447]: from numbers import Integral
[448]:
       isinstance(100, Integral)
[448]: True
       isinstance(True, Integral)
[449]:
[449]: True
[450]: isinstance(100.5, Integral)
[450]: False
[451]: type(100) is Integral
[451]: False
[452]: type(True) is Integral
[452]: False
[453]: (100).__class__
[453]: int
      (True).__class__
[454]:
[454]: bool
```

The way we have implement our decorator, if we register an Integral generic function, it won't pick up either integers or Booleans.

We can certainly fix this shortcoming ourselves, but of course...

We can use Python's built-in single dispatch support, in ...

```
you guessed it!
```

the functools module.

```
[455]: from functools import singledispatch from numbers import Integral from collections.abc import Sequence
```

```
[456]: @singledispatch
def htmlize(a):
    return escape(str(a))
```

The singledispatch returned closure has a few attributes we can use: 1. A register decorator (just like ours did) 2. A registry property that is the registry dictionary 3. A dispatch function that can be used to determine which registry key (registered type) it will use for the specified type.

```
[457]: Ohtmlize.register(Integral)
      def htmlize_int(a):
          return '{0}(<i>{1}</i)'.format(a, str(hex(a)))</pre>
[458]: htmlize.dispatch(int)
[458]: <function __main__.htmlize_int(a)>
[459]: htmlize.dispatch(bool)
[459]: <function __main__.htmlize_int(a)>
[460]: htmlize(100)
[460]: '100(<i>0x64</i)'
[461]: htmlize(True)
[461]: 'True(<i>0x1</i)'
[462]: Ohtmlize.register(Sequence)
      def html_sequence(1):
           items = ['{0}'.format(htmlize(item)) for item in 1]
          return '\n' + '\n'.join(items) + '\n'
[463]: htmlize.dispatch(list)
[463]: <function __main__.html_sequence(1)>
[464]: htmlize.dispatch(tuple)
[464]: <function __main__.html_sequence(1)>
```

```
[465]: htmlize.dispatch(str)
```

```
[465]: <function __main__.html_sequence(1)>
```

You'll note that a string is also a sequence type, hence our dispatcher will call the html_sequence function on a string.

In fact, at this point things would not even run properly.

If we were to call

htmlize('abc')

we'd get an infinite recursion!

The call to htmlize the string abc would treat it as a sequence, which would call htmlize character by character. But each character is itself just a string of length 1, so it will htmlize for that single character, which would treat it as a sequence, which would call htmlize for that single character again, and so on, in an infinite loop.

[466]: htmlize('abc')

```
Traceback (most recent call last)
RecursionError
Cell In[466], line 1
----> 1 htmlize('abc')
File ~\AppData\Local\Programs\Python\Python311\Lib\functools.py:909, in_
 ⇔singledispatch.<locals>.wrapper(*args, **kw)
    905 if not args:
           raise TypeError(f'{funcname} requires at least '
    906
    907
                           '1 positional argument')
--> 909 return dispatch(args[0].__class__)(*args, **kw)
Cell In[462], line 3, in html_sequence(1)
      1 @htmlize.register(Sequence)
     2 def html sequence(1):
           items = ['{0}'.format(htmlize(item)) for item in 1]
----> 3
           return '\n' + '\n'.join(items) + '\n'
Cell In[462], line 3, in stcomp>(.0)
     1 0htmlize.register(Sequence)
     2 def html_sequence(1):
----> 3
           items = ['{0}'.format(htmlize(item)) for item in 1]
           return '\n' + '\n'.join(items) + '\n'
File ~\AppData\Local\Programs\Python\Python311\Lib\functools.py:909, in_
 →singledispatch.<locals>.wrapper(*args, **kw)
    905 if not args:
           raise TypeError(f'{funcname} requires at least '
    906
```

```
907
                           '1 positional argument')
--> 909 return dispatch(args[0].__class__)(*args, **kw)
Cell In[462], line 3, in html_sequence(1)
      1 0htmlize.register(Sequence)
     2 def html sequence(1):
           items = ['{0}'.format(htmlize(item)) for item in 1]
           return '\n' + '\n'.join(items) + '\n'
Cell In[462], line 3, in stcomp>(.0)
      1 Ohtmlize.register(Sequence)
     2 def html_sequence(1):
           items = ['{0}'.format(htmlize(item)) for item in 1]
----> 3
           return '\n' + '\n'.join(items) + '\n'
    [... skipping similar frames: singledispatch.<locals>.wrapper at line 909 (98)
 times), times), times) at line 3 (987 times), html_sequence at line 3 (987 times)
Cell In[462], line 3, in html_sequence(1)
     1 @htmlize.register(Sequence)
     2 def html sequence(1):
           items = ['[0]'.format(htmlize(item)) for item in 1]
           return '\n' + '\n'.join(items) + '\n'
Cell In[462], line 3, in stcomp>(.0)
      1 Ohtmlize.register(Sequence)
     2 def html_sequence(1):
----> 3
           items = ['{0}'.format(htmlize(item)) for item in 1]
           return '\n' + '\n'.join(items) + '\n'
File ~\AppData\Local\Programs\Python\Python311\Lib\functools.py:909, in_
 ⇔singledispatch.<locals>.wrapper(*args, **kw)
    905 if not args:
   906
           raise TypeError(f'{funcname} requires at least '
                           '1 positional argument')
    907
--> 909 return dispatch(args[0].__class__)(*args, **kw)
File ~\AppData\Local\Programs\Python\Python311\Lib\functools.py:832, in_
 ⇒singledispatch.<locals>.dispatch(cls)
               cache_token = current_token
   830
   831 try:
           impl = dispatch_cache[cls]
--> 832
   833 except KeyError:
   834
           try:
File ~\AppData\Local\Programs\Python\Python311\Lib\weakref.py:415, in_
 →WeakKeyDictionary.__getitem__(self, key)
   414 def __getitem__(self, key):
```

```
--> 415 return self.data[ref(key)]

RecursionError: maximum recursion depth exceeded in comparison
```

Instead, we are going to register a string handler specifically - that way we will avoid that problem entirely:

```
[467]: @htmlize.register(str)
def html_str(s):
    return escape(s).replace('\n', '<br/>\n')
```

```
[468]: htmlize.dispatch(str)
```

```
[468]: <function __main__.html_str(s)>
```

So, even though a string is both an **str** instance and in general a sequence type, the "closest" type will be picked by the dispatcher (again something our own implementation did not do).

This means, we have something for generic sequences, but something specific for more specialized strings.

```
[469]: htmlize('abc')

[469]: 'abc'
```

We can do the same thing with sequences - right now html_sequence will be used for both lists and tuples.

But suppose we want slightly different handling of tuples:

```
[470]: @htmlize.register(tuple)
def html_tuple(t):
    items = [escape(str(item)) for item in t]
    return '({0})'.format(', '.join(items))
```

```
[471]: htmlize.dispatch(list)
```

[471]: <function __main__.html_sequence(1)>

```
[472]: htmlize.dispatch(tuple)
```

[472]: <function __main__.html_tuple(t)>

```
[473]: print(htmlize(['a', 100, 3.14]))
```

```
a
100(<i>0x64</i)</li>
```

```
3.14
```

```
[474]: print(htmlize(('a', 100, 3.14)))
```

```
(a, 100, 3.14)
```

One thing of note is that we started our decoration with a @singledispatch decorator - you'll notice that no specific type was indicated here - and in fact this means the dispatcher will use the generic object type.

This means that any object type not specifically handled by our dispatcher will fall back on that object key - hence you can think of it as the default for the dispatcher.

```
[475]: type(None)
[476]: NoneType
[476]: htmlize.dispatch(type(None))
[476]: <function __main__.htmlize(a)>
[477]: type(1+1j)
[477]: complex
[478]: htmlize.dispatch(complex)
[478]: <function __main__.htmlize(a)>
[479]: type(3)
[479]: int
[480]: htmlize.dispatch(int)
[480]: <function __main__.htmlize_int(a)>
```

Lastly, because the name of the individual specialized functions does not really matter to us (the dispatcher will pick the appropriate function), it is quite common for an underscore character (_) to be used for the function name - the memory address of each specialized function will be stored in the registry dictionary, and the function name does not matter - in fact we can even add lambdas to the registry.

```
[481]: @singledispatch
def htmlize(a):
    return escape(str(a))
```

```
[482]: Ohtmlize.register(int)
       def _(a):
           return '{0}({1})'.format(a, str(hex(a)))
[483]: Ohtmlize.register(str)
       def _(s):
           return escape(s).replace('\n', '<br/>\n')
[484]: htmlize.register(float)(lambda f: '{0:.2f}'.format(f))
[484]: <function __main__.<lambda>(f)>
[485]: htmlize.registry
[485]: mappingproxy({object: <function __main__.htmlize(a)>,
                     int: <function __main__._(a)>,
                     str: <function __main__._(s)>,
                     float: <function __main__.<lambda>(f)>})
      But note that the __main__._ function for int and str are not the same functions (even tough
      they have the same name):
[486]: id(htmlize.registry[str])
[486]: 2480264504288
[487]: id(htmlize.registry[int])
[487]: 2480264502368
      And everything works as expected:
[488]: htmlize(100)
[488]: '100(0x64)'
[489]: htmlize(3.1415)
[489]: '3.14'
[490]: print(htmlize("""this
       is a multi-line string
       a < 10"""))
      this<br/>
      is a multi-line string<br/>
      a < 10
```

If this same name but different function thing has you confused, look at it this way:

```
[491]: def my_func():
            print('my_func initial')
[492]:
       id(my_func)
[492]: 2480259974944
[493]: f = my_func
[494]: id(f)
[494]: 2480259974944
       So, f and my_func point to the same function in memory.
       Let's go ahead and "redefine" the function my func:
[495]: def my_func():
            print('second my_func')
       In fact, we did not "redefine" the previous my_func, it still exists in memory (and f still points to
       it). Instead we have re-assigned the function that my_func points to:
[496]: id(my_func)
[496]: 2480264503328
       But the original my_func is still around, and 'f' still has a reference to it:
       id(f)
[497]:
[497]: 2480259974944
       So, we can call each one:
[498]: f()
       my_func initial
[499]: my_func()
       second my_func
       But the function __name__ have the same value:
[500]: f.__name__
[500]: 'my_func'
```

```
[501]: my_func.__name__
```

[501]: 'my_func'

Just always keep in mind that labels point to something in memory, it is not the object itself. So in this case we have two distinct objects (functions) which happen to have the same name, but are two very different objects - f points to the first one we created, and my_func points to the second.