Something cool with higher order functions ¶

In the last lecture, we looked at higher order functions: the idea that you can write functions that use *other functions* as arguments or as return values. One cool consequence of higher order functions: you don't need multi-argument functions anymore: you only ever need functions that accept one argument.

You might think this is trivial: if I want to write a function that takes two integers, just write a function that accepts a structure with two integer fields. This is more subtle than that: we will not use any notion of "tuples": pieces of data that actually represent multiple pieces of data.

Consider a simple function of two arguments:

```
In [1]: def myFun(x, y) :
    return 3 * x + y

print myFun(3, 8)
```

We can write a version of this function that only ever accepts one argument at a time. What we're going to do is take advantage of *closures* (remember Lecture 3) to write a function that takes the *first* argument, then returns a *new function* that incorporates the first argument and accepts the second argument. We can then call this new function on the second argument to produce the same result as the original function.

```
In [2]: def myFunCurry(x) : #note that this only takes one argument!
    def inner(y) : #this function takes the second argument!
        return 3 * x + y
    return inner

print myFunCurry(3)(8)
```

Let's deconstruct what happened. When we call myFun(3), we're getting back a new function that *closed* over 3:

```
In [3]: inter = myFunCurry(3)
print inter

<function inner at 0x103ba9938>
```

This function is the same as if we had written a function that substituted in 3 for x:

```
In [4]: def inter2(y) :
    return 3 * 3 + y
print inter2

<function inter2 at 0x103ba96e0>
```

These new functions can then accept y as their argument to finish the computation:

We can generalize this to functions of 3 arguments:

```
In [6]:
        def myFun3(x, y, z) :
            return x ** 2 + 3 * y + z
        print myFun3(3, 4, 5)
        26
In [7]: def myFun3Curry(x) :
            def inner1(y) :
                def inner2(z) :
                    return x ** 2 + 3 * y + z
                return inner2
            return inner1
        print myFun3Curry(3)(4)(5)
        26
In [8]: for i in range (1, 100) :
            for j in range (1, 100) :
                for k in range (1, 100):
                    assert(myFun3(i, j, k) == myFun3Curry(i)(j)(k))
```

We call this process (moving from a function that takes k arguments to a series of functions that each take 1 argument) *Currying*. "Currying" is named after Haskell Curry -- and so is the Haskell programming language!

Data structures

We have already seen two basic data structures in python. First, we saw lists:

```
In [9]: list1 = [0, 2, 4, 6, 8]
    print type(list1)
    print list1
    print list1[2:4]
    list2 = list1 + [10]
    print list2

<type 'list'>
    [0, 2, 4, 6, 8]
    [4, 6]
    [0, 2, 4, 6, 8, 10]
```

Wait, two data structures? Yes! Strings in Python are a data structure too. In fact, like lists, strings are a *sequence* data structure, that supports several of the same operations as lists:

```
In [10]:
         string1 = 'Hello'
         print type(string1)
         print len(string1)
         print string1[1:4]
         string2 = string1 + '!'
         print string2
         for s in string2:
              print s
         <type 'str'>
         ell
         Hello!
         Η
         е
         1
         1
         O
```

Tuples

Another sequence type in Python is the *tuple*. These look a lot like lists, with a few exceptions. First, you define them with () instead of []. Second, tuples are *immutable*. Once you define them, you cannot add or remove items from them. Think of tuples as a way of defining structures. You can get at the elements of tuples by indexing into them, just like lists or strings:

```
In [11]: tuple1 = ('Hello', 3.14, 2)
    print "{} {}".format(tuple1, type(tuple1))
    print "{} {}".format(tuple1[1], type(tuple1[1]))

    ('Hello', 3.14, 2) <type 'tuple'>
    3.14 <type 'float'>
```

And you can get at elements of a tuple by iterating over them (again, just like lists or strings)

Here's a fancier way to iterate over a tuple:

```
In [13]: for i, t in enumerate(tuple1):
    print "{} {}".format(t, type(t))
    print "{} {}".format(tuple1[i], type(tuple1[i]))

Hello <type 'str'>
    Hello <type 'str'>
    3.14 <type 'float'>
    3.14 <type 'float'>
    2 <type 'int'>
    2 <type 'int'>
```

What's going on with enumerate up there? That's a special function for iterating through sequence types (meaning you can use it on strings and lists, too) that emits *tuples* as its output. The tuples it emits are of the form (index, value). The looping code takes advantage of a handy Python trick called *unpacking* that lets you get at the elements of a tuple without having to index them.

```
In [14]: s, f, i = tuple1
print s
print f
print i

Hello
3.14
2
```

Using tuples as your replacement for C-like structs can be tricky, if the tuples get complicated (think about how hard it might be to remember the organization of the tuple). Python provides *named tuples* as a way around this, which we will get to when we talk about objects.

Sets

Python includes *sets* as a built-in data type. They operate just like Java sets or STL sets: unordered groups of elements that maintain a *uniqueness* property, where each value only appears once in the set

```
In [16]: | set1 = {'a', 'b', 'c'}
         print set1 #note the ordering!
         set(['a', 'c', 'b'])
In [17]: | set2 = {'a', 'b', 'c', 'a'}
         print set2
         set(['a', 'c', 'b'])
         set2.add('d')
In [18]:
         print set2
         set2.remove('a')
         print set2
         set(['a', 'c', 'b', 'd'])
         set(['c', 'b', 'd'])
In [19]:
         set3 = set() #empty set initialization
         print set3
         set3.add('a')
         set3.add('b')
         set3.add('a')
         print set3
         set([])
         set(['a', 'b'])
```

```
In [20]: for d in set2:
    print d

c
b
d
```

Comprehensions

Python provides set and list *comprehensions*, which are efficient ways of processing sets and lists to produce new sets and lists (think mathematical set notation)

```
In [21]: import numpy as np
    data = list(np.random.randint(-4, 4, 25))
    print data

[-1, 3, 2, -2, 3, 2, 0, 3, -2, 2, -2, -3, 3, -2, -1, -2, 0, -4, -2,
    0, -4, 3, -2, -4, 3]

In [22]: absdata = [abs(d) for d in data]
    print absdata

[1, 3, 2, 2, 3, 2, 0, 3, 2, 2, 2, 3, 3, 2, 1, 2, 0, 4, 2, 0, 4, 3, 2, 4, 3]

In [23]: absset = {abs(d) for d in data}
    print absset
    set([0, 1, 2, 3, 4])
```

Dictionaries

The final "basic" data structure in Python is the *dictionary*. (Other languages call them "associative arrays." You probably know them as "maps"): data structures that let you map *keys* to *values*. Each key in a Python dictionary is unique, and that key maps to a certain value.

```
In [24]: dict1 = {'a': 0, 'b': 1, 'c': 3}
    print dict1['a'], dict1['c']

0 3
In [25]: dict1['a'] = 10
    print dict1['a'], dict1['c']

10 3
```

When iterating over a dictionary, you iterate over the keys. If you want to iterate over both the keys and the values, use iteritems

Wait, what's going on with iteritems? We're not calling it like we do other functions like len or min or max. iteritems is a method of the dict class. dict1 in the above example (like all Python data) is an object. (We saw similar ways of calling methods when we append items to lists, or add items to sets.)

Classes and Objects

This is not a particularly formal introduction to the Python data model and object model. For that, please refer to documentation on the Python data model (https://docs.python.org/2/reference/datamodel.html) and Python classes (https://docs.python.org/2/tutorial/classes.html).

Python, like C++ and Java, is *object oriented*. The basic data model in Python is that everything is an object of some sort. An object combines data and methods. *Everything in Python is an object*, including "simple" data like integers and floats.

A *class* in python defines a set of *attributes*: these can be variables or methods. This defines a set of properties that you want all objects of a certain type to have. An *object* in Python is an *instance* of a class: it shares attributes with all other classes, but can also have attributes (think: member data) that is different from other instances. This lets you have objects with their own "local" data.

Methods for a class take an extra self argument. When you invoke a method on an object (think myList.append(x)), this self argument refers to the object you invoked the method on (in the example, myList).

```
In [27]: class Counter (object):
             totalCount = 0 #shared number across all instances
             def __init__(self) : #constructor for the class.
                 self.count = 0 #local count for each instance
             def incr(self) :
                 Counter.totalCount += 1
                 self.count += 1
             def str (self) : #special function like "toString" in Java
                 return "Total count: {}, Local count: {}".format(Counter.total
         Count, self.count)
In [28]: c1 = Counter()
         c2 = Counter()
         print c1
         print c2
         Total count: 0, Local count: 0
         Total count: 0, Local count: 0
In [29]: for i in range(0,5):
             c1.incr()
             c2.incr()
         print c1
         print c2
         Total count: 10, Local count: 5
         Total count: 10, Local count: 5
```

Classes themselves, like functions, are just objects, as are the methods inside them:

Unsurprisingly, like with functions, Python lets you create new classes dynamically and return them. This gives us a handy way to create things that behave like structures, using the namedtuple method:

```
In [31]: import collections
Point = collections.namedtuple('Point',['x', 'y', 'color'])
```

```
In [32]: p = Point(2.4, 3.7, 'red')
print p

Point(x=2.4, y=3.7, color='red')

In [33]: print p.x, p.y, p.color
2.4 3.7 red
```

Pandas

The place where you will probably be using classes the most is when manipulating pandas *dataframes*: this is the key class provided by pandas (in addition to *series*), and it provides a number of instance methods for manipulating data. We will not spend a lot of time deconstructing pandas dataframes in class -- we will explain as much as is needed in relevant homeworks. You can also look at the <u>docs</u> (https://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.html)

```
In [ ]: import pandas as pd
    data = pd.read_csv('hw02_problem3.csv', header=None, skipinitialspace=
    True)
    print type(data)
    data = data[(data[2] == 'white')]
    print type(data1)
    data1
In [ ]: data2 = data[(data[2] == 'white')][[0, 1]]
data2
```