Today: Review of OOP, numpy and matplotlib

I. Review of OOP

With one addition... Inheritance!

- There is a hierarchy of classes. A subclass is a part of a superclass.
- Thus, classes can <u>inherit</u> from other classes = A subclass can inherit attributes and methods from its superclass
- Syntax for defining a subclass

class SubClassName(SuperClassName):

- a subclass has access to all instance variables and methods of superclass
- In subclass, we can refer to superclass as super ()
- See Person and Student.py for examples

II. numpy and matplotlib

A. Numpy

Numpy is an external package in Python for making multidimensional arrays (very similar to Matlab)

- Why do we use it? **Speed!**
- To learn more about numpy: http://www.numpy.org/
- If you've downloaded Anaconda, you already have the numpy package downloaded.
- Two rules in Numpy help accomplish this speed
 - 1. Homogeneous: all elements of an array are of same type
 - 2. Fixed length

1) First import numpy: import numpy

2) 6 ways to create an array in numpy

- 1. function array (var): can put any one sequential type variable (list, tuple, set) in the parentheses to cast it into numpy array
- 2. function zeros(x): return an array of length x (all elements 0) Use a tuple as an argument instead of int to create multidimensional arrays

- 3. function ones (x): same as above with 1.0s instead of 0.s
- 4. function arange() : calling the array function on range()

```
In: import numpy as np
In: np.arange(3)
Out: array([0, 1, 2])
```

```
5. function linspace(starting point, ending point, number of points I want
      in between): carves up the range into n-1 pieces for you
In: np.linspace(1,10,2)
Out: array([ 1., 10.])
In: np.linspace(1,10,5)
Out: array([ 1. , 3.25, 5.5 , 7.75, 10. ])
   6. function random.rand(n): array of n random numbers from 0 to 1
In: np.random.rand(3)
Out: array([ 0.63040386, 0.27029223, 0.46125486])
In: np.random.rand(3,3)
Out: [[ 0.61925984  0.97407503  0.02655378]
      [ 0.51168271  0.46085259  0.27524427]
      [ 0.77434942  0.64746699  0.49072881]] (2D array of random#)
  3) Some constants for numpy arrays
   1. dtype: returns datatype of the array elements
   2. shape: returns a tuple that shows length in each dimension
   3. ndim: returns dimension
   4. size: returns number of elements
e.g.
a = numpy.zeros((3,4))
In: a.shape
Out: (3,4)
In: a.ndim
Out: 2
In: a.size
Out: 12
In: a.dtype
Out: dtype('int64')
e.g.
In: d1 = numpy.zeros(3)
In: dl.shape
Out: (3,) # because one dimensional
In: d2 = numpy.zeros((1,3))
In: d2.shape
Out: (1,3) # now 2D although d1 and d2 look same to us
4) Indexing through numpy arrays and assigning elements
In: import numpy as np
In: s3 = np.zeros((2,2,2)) #3D array:2 floors of 2row*2column matrix
In: s3[0][0][0]
Out: 0.
In: s3[0,0,0] #different syntax for same meaning as above (both work)
Out: 0.
In: s3[1,:,:] #colon means everything in that dimension
Out: array([[[ 0., 0.],
              [ 0., 0.]])
```

```
In: s3[1,1,1] = 100 #assigning elements \rightarrow this modifies array
In: s3
Out: array([[
               0.,
                        0.],
                 0.,
                        0.]],
             [
            [[ 0., 0.],
                 0., 100.]])
            [
In: s3[0,:,:] = 1 #slicing works(set everything in first floor to 1)
Out: array([[ 1.,
                     1.],
            [ 1.,
                       1.]],
            [[ 0., 0.],
            [ 0., 100.]])
5) Methods
  • .astype (datatype): returns a brand new numpy array with elements casted
In: s3.astype(int)
Out: array([[ 1,
                   1],
                   1]],
            [ 1,
            [[ 0, 0],
             [ 0, 100]])
  • .reshape (tuple for size): returns new, reformatted version of original
           ** should keep size (number of elements) the same like below
In: s3.reshape((2,1,4))
Out: array([[[ 1., 1., 1., 1.]],
            [[ 0., 0., 100.]]])
In: s3
                 1., 1.],
1., 1.]],
Out: array([[
            [
            [ [
                 0., 0.],
                 0., 100.]]])  # s3 unmodified
  • .resize(): same as reshape but modifies the original array instead of creating new one
In: s3.resize((2,1,4))
In: s3
Out: array([[[ 1., 1., 1., 1.]],
            [[ 0., 0., 100.]]])
  • perform arithmetic operations on all elements
In: s3 = s3/2
In: s3
Out: array([[[ 0.5, 0.5, 0.5, 0.5]],
```

[[0., 0., 50.]]

B. MATPLOTLIB

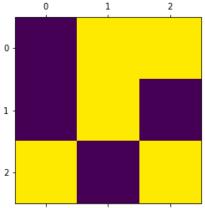
As always, import!

import matplotlib.pyplot as plt

1. Visual grid

```
plt.matshow(numpy_array): function to generate visual display from array
plt.show(): show visual display
```

e.g.



2. Plot on Graph

var1 = np.linspace(start, stop, #stops) #determine range and inverval of x

1) Generate 1 plot

```
plt.plot(var1,var2) # line plot of var2 vs var1
plt.plot(var2,var1,'o') # dot plot
```

2) Generate nice graph with title, axis title

3) Generate various subplots on one display

```
fig = plt.figure() # generate figure
gr1 = fig.add subplot(221)
      #Among 2x2 subplots of fig, gr1 is the 1st graph (top left)
gr2 = fig.add subplot(222) # gr2 is the 2<sup>nd</sup> graph (top right)
gr3 = fig.add_subplot(223) # gr2 is 3<sup>rd</sup> graph (bottom left)
gr4 = fig.add subplot(224) # gr2 is 4<sup>th</sup> graph (bottom right)
gr1.plot(var1, var2)
gr2.plot(var2, var3), etc. # must plot and title each subplot separately
e.g.
def f(t):
    return t**2
t = np.linspace(0,3,50)
y = np.zeros(len(t))
for i in range(len(t)):
    y[i] = f(t[i])
fig = plt.figure()
gr1 = fig.add subplot(221)
gr1.plot(t,y)
gr1.set title('t^2 graph')
gr2 = fig.add subplot(222)
gr2.plot(t,t)
gr2.set title('t graph')
gr3 = fig.add subplot(223)
gr4 = fig.add subplot(224) #gr3, gr4 are empty
plt.show()
```

Result:

