E9 205 Machine Learning for Signal Processing

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Tutorial on Python

Outline

- Basics and control statements
- Functions and lists
- Tuples, Dictionaries, Strings, File i/o
- Modules
- Packages: Numpy, (Scipy, Matplotlib)

Reference: [1] Stanford university, Course: "Introduction to Scientific Python" Link: https://web.stanford.edu/~schmit/cme193/resources.html [2] : PythonLearn : http://www.pythonlearn.com/

How to install Python?

- Anaconda
 - https://store.continuum.io/cshop/anaconda/
- Very easy to install and also comes with a lot of packages.

How to use Python

- There are two ways to use Python:
 - command-line mode: talk directly to the interpreter
 - scripting-mode: write code in a file (called script) and run code by typing in the terminal

python Tutorial1.py

Tutorial1.py----> print "MLSP 2017"

Comments in Python

- Anything after a # is ignored by Python
- Why comment?
 - Describe what is going to happen in a sequence of code
 - Turn off a line of code perhaps temporarily
 - Easy to evaluate your assignments..

Control statements

- Control statements allow you to do more complicated tasks.
 - if
 - for
 - while

Indentation

- In Python, blocks of code are defined using *indentation*.
- Maintain indent to indicate the scope of the block
 - This means that everything indented after an if statement is only executed if the statement is True.
- Reduce indent back to the level of the if statement to indicate the end of the block
 - If the statement is False, the program skips all indented code and resumes at the first line of unindented code

```
if statement:
    # if statement is True, then all code here
    # gets executed but not if statement is False
    print "The statement is true"
    print "Else, this would not be printed"
# the next lines get executed either way
print "Hello, world,"
print "Bye, world!"
```

Control statements

```
if traffic_light == 'green':
    drive()
elif traffic_light == 'orange':
    accelerate()
else:
    stop()
```

```
for i in range(5):
    print i**2,
# 0 1 4 9 16
```

```
i = 1
while i < 100:
    print i**2,
    i += i**2 # a += b is short for a = a + b
# 1 4 36 1764</pre>
```

Functions and lists

Functions

- Much like a mathematical function, they take some input and then do something to find the result.
- Start a function definition with the keyword **def**
- Then comes the function name, with arguments in braces, and then a colon--> indentend(body of function) -->return (Specifies o/p)

```
x = 1
def add_one(x):
    x = x + 1  # local x
    return x

y = add_one(x)
# x = 1, y = 2
```

Lists

- Group variables together
- can mix element types
- Access items using square brackets: []

```
myList = [5, 2.3, 'hello']
myList[0]  # 5
myList[2]  # 'hello'
myList[3]  # ! IndexError
myList[-1]  # 'hello'
myList[-3]  # ?
```

Slicing and adding

- Lists can be sliced: [2:5]
- Lists can be multiplied
- Lists can be added

```
myList = [5, 2.3, 'hello']
myList[0:2]  # [5, 2.3]
mySecondList = ['a', '3']
concatList = myList + mySecondList
# [5, 2.3, 'hello', 'a', '3']
```

Lists are mutable

individual elements can be changed

```
myList = ['a', 43, 1.234]
myList[0] = -3
# [-3, 43, 1.234]
x = 2
myList[1:3] = [x, 2.3] # or: myList[1:] = [x, 2.3]
# [-3, 2, 2.3]
x = 4
# What is myList now?
```

More control over lists

- o len(xs)
- xs.append(x)
- ws.count(x)
- o xs.insert(i, x)
- xs.sort() and sorted(xs): what's the difference?
- ws.remove(x)
- xs.pop() or xs.pop(i)
- x in xs

All these can be found in the Python documentation, google: 'python list'

Or using dir(xs) / dir([])

• Tuples, dictionaries and strings

Tuples

- similar to lists
- Tuples are Immutable-->Unlike lists, we cannot change elements.

```
>>> myTuple = (1, 2, 3)
>>> myTuple[1]
2
>>> myTuple[1:3]
(2, 3)
```

```
>>> myTuple = ([1, 2], [2, 3])
>>> myTuple[0] = [3,4]
Traceback (most recent call last):
   File "<stdin>", line 1, in <module>
TypeError: 'tuple' object does not
support item assignment
>>> myTuple[0][1] = 3
>>> myTuple
([1, 3], [2, 3])
```

Dictionaries

- A dictionary is a collection of *key-value* pairs.
- An example: the keys are all words in the English language, and their corresponding values are the meanings.

```
>>> d = {}
>>> d[1] = "one"
>>> d[2] = "two"
>>> d
{1: 'one', 2: 'two'}
>>> e = {1: 'one', 'hello': True}
>>> e
{1: 'one', 'hello': True}
```

Print all key-value pairs of a dictionary

```
>>> d = {1: 'one', 2: 'two', 3: 'three'}
>>> for key, value in d.items():
...
print key, value
...
1 one
2 two
3 three
```

Strings

- Strings hold a sequence of characters.
- We can slice strings just like lists and tuples
- We can turn anything in Python into a string using **str**
- To split a string, for example, into seperate words, we can use split()

```
text = 'Hello, world!\n How are you?'
text.split()
# ['Hello,', 'world!', 'How', 'are', 'you?']
```

```
numbers = '1, 3, 2, 5'
numbers.split()
# ['1,', '3,', '2,', '5']
numbers.split(', ')
# ['1', '3', '2', '5']
[int(i) for i in numbers.split(', ')]
# [1, 3, 2, 5]
```

File I/O

- Interaction with the file system is pretty straightforward in Python.
- Done using file objects
- We can instantiate a file object using open or file
- f = open(filename, option)
 - filename: path and filename
 - Option: 'r' read file, 'w' write to file, 'a' append to file
- We need to close a file after we are done: f.close()
- read() Read entire line (or first n characters, if supplied)

with open('data/text_file.txt', 'r') as f:
 print f.read()

Writing to file

• Use write() to write to a file

with open(filename, 'w') as f:
 f.write("Hello, {}!\n".format(name))

Modules: Importing a module

- We can import a module by using **import**
 - E.g. import math
 - We can then access everything in math, for example the square root function, by: math.sqrt(2)
- We can rename imported modules
 - E.g. import math as m
 - Now we can write **m.sqrt(2)**
- In case we only need some part of a module
 - We can import only what we need using the from ... import ...syntax.
 - E.g. from math import sqrt
 - Now we can use **sqrt(2)** directly

• Numpy, Scipy, Matplotlib

Numpy

- Fundamental package for scientific computing with Python
- N-dimensional array object
- Linear algebra, Fourier transform, random number capabilities

```
import numpy as np
A = np.array([[1, 2, 3], [4, 5, 6]])
print A
# [[1 2 3]
# [4 5 6]]
Af = np.array([1, 2, 3], float)
```

```
np.arange(0, 1, 0.2)
\# \operatorname{array}([0., 0.2, 0.4, 0.6, 0.8])
np.linspace(0, 2*np.pi, 4)
# array([ 0.0, 2.09, 4.18, 6.28])
A = np.zeros((2,3))
# array([[ 0., 0., 0.],
# [0., 0., 0.]])
# np.ones, np.diag
A.shape
# (2, 3)
                    np.random.random((2,3))
```

```
# array([[ 0.78084261, 0.64328818, 0.55380341],
# [ 0.24611092, 0.37011213, 0.83313416]])
a = np.random.normal(loc=1.0, scale=2.0, size=(2,2))
# array([[ 2.87799514, 0.6284259 ],
# [ 3.10683164, 2.05324587]])
np.savetxt("a_out.txt", a)
# save to file
b = np.loadtxt("a_out.txt")
# read from file
```

Array attributes

Basic operations

```
a = np.arange(10).reshape((2,5))
```

a.ndim # 2 dimension a.shape # (2, 5) shape of array a.size # 10 # of elements a.T # transpose a.dtype # data type a = np.arange(4)
array([0, 1, 2, 3])
b = np.array([2, 3, 2, 4])
a * b # array([0, 3, 4, 12])
b - a # array([2, 2, 0, 1])
c = [2, 3, 4, 5]
a * c # array([0, 3, 8, 15])

Vector operations

- inner product
- outer product
- dot product (matrix multiplication)

```
# note: numpy automatically converts lists
u = [1, 2, 3]
v = [1, 1, 1]
np.inner(u, v)
# 6
np.outer(u, v)
# array([[1, 1, 1],
# [2, 2, 2],
# [3, 3, 3]])
np.dot(u, v)
# 6
```

Matrix operations

```
A = np.ones((3, 2))
# array([[ 1., 1.],
#        [ 1., 1.],
#        [ 1., 1.],
#        [ 1., 1.]])
A.T
#        [ 1., 1., 1.],
#        [ 1., 1., 1.]])
B = np.ones((2, 3))
# array([[ 1., 1., 1.],
#        [ 1., 1., 1.]])
```

```
np.dot(A, B)
# array([[ 2., 2., 2.],
# [ 2., 2., 2.],
# [ 2., 2., 2.],
# [ 2., 2., 2.]])
np.dot(B, A)
# array([[ 3., 3.],
# [ 3., 3.]])
```

Operations along axes

```
a = np.random.random((2,3))
# array([[ 0.9190687 , 0.36497813, 0.75644216],
# [ 0.91938241, 0.08599547, 0.49544003]])
a_sum()
# 3.5413068994445549
a.sum(axis=0) # column sum
# array([ 1.83845111, 0.4509736 , 1.25188219])
a.cumsum()
# array([ 0.9190687 , 1.28404683, 2.04048899, 2.9598714 ,
         3.04586687, 3.5413069])
#
a.cumsum(axis=1) # cumulative row sum
\# \operatorname{array}([0.9190687, 1.28404683, 2.04048899],
# [ 0.91938241, 1.00537788, 1.50081791]])
a.min()
# 0.0859954690403677
a.max(axis=0)
# array([ 0.91938241, 0.36497813, 0.75644216])
```

Slicing arrays

```
a = np.random.random((4,5))
a[2, :]
# third row, all columns
a[1:3]
# 2nd, 3rd row, all columns
a[:, 2:4]
# all rows, columns 3 and 4
```

Matrix operations

import numpy.linalg

eye(3)

Identity matrix

trace(A)

Trace

column_stack((A,B)) Stack column wise

row_stack((A,B,A)) Stack row wise

Linear algebra

import numpy.linalg

qr	Computes the QR decomposition
cholesky	Computes the Cholesky decomposition
inv(A)	Inverse
<pre>solve(A,b)</pre>	Solves $Ax = b$ for A full rank
lstsq(A,b)	Solves $\arg \min_x Ax - b _2$
eig(A)	Eigenvalue decomposition
eig(A)	Eigenvalue decomposition for symmetric or hermitian
eigvals(A)	Computes eigenvalues.
<pre>svd(A, full)</pre>	Singular value decomposition
pinv(A)	Computes pseudo-inverse of A

Fourier transform

import numpy.fft

- fft 1-dimensional DFT
- Ift2 2-dimensional DFT
- Iftn N-dimensional DFT
- ifft 1-dimensional inverse DFT (etc.)
- rfft Real DFT (1-dim)
- ifft Imaginary DFT (1-dim)

Random sampling

import numpy.random

- rand(d0,d1,...,dn)
- randn(d0, d1, ...,dn)
- randint(lo, hi, size)
- choice(a, size, repl, p)

shuffle(a)

permutation(a)

Random values in a given shape Random standard normal Random integers [lo, hi) Sample from a Permutation (in-place) Permutation (new array)

Matplotlib

- What is Matplotlib?
 - Plotting library for Python
 - Works well with Numpy
 - Syntax similar to Matlab

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
x = np.linspace(0, 10, 1000)
y = np.power(x, 2)
plt.plot(x, y)
plt.show()
```



What is SciPy?

- SciPy is a library of algorithms and mathematical tools built to work with NumPy arrays.
 - linear algebra scipy.linalg
 - statistics scipy.stats
 - optimization scipy.optimize
 - sparse matrices scipy.sparse
 - signal processing scipy.signal
 - image processing: skimage
 - pip install -U scikit-image
 - http://www.scipy-lectures.org/packages/scikit-image/index.html

Thank you!