Python Primer

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1 Python

1.1 Lists and Tuples

1.1.1 Indexing into list

Out[1]: 2

1.1.2 Appending to a list

1.1.3 Deleting an element

```
In [3]: del 1[1]
```

1.1.4 Inserting an element

```
Out[4]: [1, 3, 3, 4]
```

1.1.5 Tuples

```
In [5]: t = (1, 3, 3, 4) # make a tuple
    1 == t
```

Out[5]: False

1.1.6 List to Tuple

Out[6]: True

1.2 Dictionaries

```
In [7]: Dict = {}
        Dict[1] = 2
        Dict['one'] = 'two'
        Dict['1'] = '2'
        Dict
Out[7]: {1: 2, '1': '2', 'one': 'two'}
1.2.1 Keys in Dictionary
In [8]: print "Dictionary keys"
        print Dict.keys()
        print "\nValue at 1 :"
        print Dict['1']
        print "\nValue at one"
        print Dict['one']
        one = 1
        print "\nValue at 1"
        print Dict[one]
        print "\nIterate over keys"
        for key in Dict.keys():
            print key
        print "\nDelete key : 1"
        del Dict[1]
        print Dict
Dictionary keys
['1', 1, 'one']
Value at 1 :
Value at one
two
Value at 1
Iterate over keys
1
1
one
Delete key : 1
{'1': '2', 'one': 'two'}
```

2 Classes and Function

2.1 Functions

```
In [9]: def printer(x):
            print x
        def adder(x,y):
            return x+y
        def square(x):
            return x**2
        a = 2
        b = 3
        print "Lets print a:"
        printer(a)
        print "\nLets print a + b"
        printer(adder(a,b))
        print "\n So you can pass the return of a function to another function just like everywhere. \n
        printer(square(adder(a,b)))
Lets print a:
Lets print a + b
So you can pass the return of a function to another function just like everywhere.
Lets take it another step further
25
2.2
      Classes
In [10]: class student(object):
             def __init__(self,name = None ,age = None):
                 if name == None:
                     self.name = "Amartya"
                 else:
                     self.name = name
                 if age == None:
                     self.age = 20
                 else:
                     self.age = age
             def update_name(self,name):
                 self.name = name
             def update_age(self,age):
                 self.age = age
             def inc_age(self):
                 self.age = self.age + 1
```

```
def return_info(self):
                 temp = [self.name, self.age]
                 return temp
In [11]: Amartya = student()
         print"Amartya:"
         print vars(Amartya)
         Bhuvesh = student("Bhuvesh", 21)
         print "\nBhuvesh:"
         print vars(Bhuvesh)
         print "\nIncrementing Bhuvesh's age"
         Bhuvesh.inc_age()
         print vars(Bhuvesh)
         print "\nMake Amartya a baby"
         Amartya.update_age(1)
         print vars(Amartya)
         print "\nA list of attributes of Amartya(Just to show what lists are)"
         print Amartya.return_info()
Amartya:
{'age': 20, 'name': 'Amartya'}
Bhuvesh:
{'age': 21, 'name': 'Bhuvesh'}
Incrementing Bhuvesh's age
{'age': 22, 'name': 'Bhuvesh'}
Make Amartya a baby
{'age': 1, 'name': 'Amartya'}
A list of attributes of Amartya(Just to show what lists are)
['Amartya', 1]
    Exceptions
3
In [12]: print "Adding 2 and 3"
         printer(adder(2,3))
         print "\nAdding 'Amartya' and 'Bhuvesh'"
         printer(adder("amartya", "bhuvesh"))
         print "\nBut say we want to practical and only add numbers , not people."
         def adder(x,y):
             try:
                 if type(x) != 'int' or type(x) != 'float' or type(y) != 'int' or type(y) != 'float':
                     raise ValueError()
```

```
else:
                   return x+y
            except ValueError:
                print "Error!! Error!! You cant add people\n"
        print "\nAdding 'Amartya' and 'Bhuvesh'"
        printer(adder("amartya", "bhuvesh"))
Adding 2 and 3
Adding 'Amartya' and 'Bhuvesh'
amartyabhuvesh
But say we want to practical and only add numbers , not people.
Adding 'Amartya' and 'Bhuvesh'
Error!! Error!! You cant add people
None
    Starting Numpy
In [13]: import numpy as np #Please don't forget this
    Basic types of arrays and matrices
4.1.1 Zero Array and Zero Matrix
In [14]: zeroArray = np.zeros(5)
        print "Zero Array"
        print zeroArray
        print "\nZero Matrix:"
        zeroArray = np.zeros([5,10])
        print zeroArray
Zero Array
[ 0. 0. 0. 0. 0.]
Zero Matrix:
[[0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
[0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
[0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
[0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]]
4.1.2 Ones array and Ones Matrix
In [15]: oneArray = np.ones(5)
        print "Ones Array"
        print oneArray
        print "\nOnes Matrix:"
        oneArray = np.ones([5,10])
```

print oneArray

```
Ones Array
[1. 1. 1. 1.]
Ones Matrix:
[[ 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]
[1. 1. 1. 1. 1. 1. 1. 1. 1.]
 [ 1. 1. 1. 1. 1. 1. 1. 1. 1. ]
4.1.3 Identity Matrix
In [16]: I = np.identity(5)
        print "Identity Matrix"
        print I
Identity Matrix
[[ 1. 0. 0. 0. 0.]
[ 0. 1. 0. 0. 0.]
[ 0. 0. 1. 0. 0.]
[ 0. 0. 0. 1. 0.]
[ 0. 0. 0. 0. 1.]]
4.1.4 Basic vector stuff
In [17]: A = [1, 2, 3]
        B = np.asarray(A)
        C = [4,5,6]
        D = np.asarray(C)
In [18]: print "Elementwise Multiplication"
        print B*D
        print "\nElementwise Addition"
        print B+D
        print "\n Dot Product"
        print np.dot(B,D)
Elementwise Multiplication
[ 4 10 18]
Elementwise Addition
[5 7 9]
Dot Product
32
In [19]: print "Lets square each element in the array"
        print [x**2 for x in C]
        print "\n Lets do some more complicated function"
        def updateX(x):
            x = x + 2
            x = np.log(x)
            x = np.power(x,2)
            return x
        print [updateX(x) for x in C]
```

```
Lets square each element in the array
[16, 25, 36]
Lets do some more complicated function
[3.2104019955684011, 3.7865663081964716, 4.3240771252638117]
4.1.5 Useful stuffs that make your life easy when coding stuffs.
In [20]: print "Createing an array of numbers from 1 to 9"
         A = np.arange(1,10)
         print A
         print "\n Reshape an array to matrix"
         B = np.reshape(A,[3,3])
         print B
         print "\n Transpose the matrix"
         C = np.transpose(B)
         print C
         print "\n Make elements less than 5 0"
         C[C<5] = 0
         print C
Createing an array of numbers from 1 to 9
[1 2 3 4 5 6 7 8 9]
Reshape an array to matrix
[[1 2 3]
 [4 5 6]
 [7 8 9]]
Transpose the matrix
[[1 4 7]
 [2 5 8]
 [3 6 9]]
Make elements less than 5 0
[[0 0 7]
[0 5 8]
 [0 6 9]]
In [21]: print "Summing up elements"
         print "\n Each column"
         print np.sum(C,axis=0)
         print "\n Each row"
         print np.sum(C,axis=1)
Summing up elements
Each column
[ 0 11 24]
Each row
[ 7 13 15]
```

```
In [22]: print "Mean of elements"
        print "\n Each column"
        print np.mean(C,axis=0)
        print "\n Each row"
        print np.mean(C,axis=1)
Mean of elements
Each column
              3.66666667 8.
ΓО.
                                    ٦
Each row
[ 2.33333333  4.33333333  5.
In [23]: print "Product of elements"
        print "\n Each column"
        print np.prod(C,axis=0)
        print "\n Each row"
        print np.prod(C,axis=1)
Product of elements
Each column
[ 0 0 504]
Each row
[0 0 0]
    Finally Theano!
5
In [24]: import theano
         import theano.tensor as T
In [25]: # Create the scalars
         x = T.scalar()
        y = T.scalar()
In [26]: print "Add two numbers"
         temp1 = x + y
         # So this is how you add two "Symbolic variables"
         addTh = theano.function([x,y],temp1)
         theano.pp(addTh.maker.fgraph.outputs[0])
Add two numbers
Out[26]: '(<TensorType(float64, scalar)> + <TensorType(float64, scalar)>)'
In [27]: print addTh(1,2)
3.0
In [28]: print "Comparing two numbers"
         temp1 = T.le(x, y)
         compTh = theano.function([x,y],temp1)
```

```
theano.pp(compTh.maker.fgraph.outputs[0])
         print compTh(4,3)
Comparing two numbers
In [29]: print "If else operator in Theano"
         xgy = T.ge(x,y)
         res = 2*x*xgy + (1 - xgy)*3*x
         ifelse = theano.function([x,y],res)
         print theano.pp(compTh.maker.fgraph.outputs[0])
         print ""
         print ifelse(5,4)
If else operator in Theano
le(<TensorType(float64, scalar)>, <TensorType(float64, scalar)>)
10.0
In [30]: #Create the symbolic graph
        z = x + y
        M = X * X
         a = T.sqrt(w)
         b = T.exp(a)
         c = a ** b
         d = T.log(c)
         uselessFunc = theano.function([x,y],d)
         theano.pp(uselessFunc.maker.fgraph.outputs[0])
Out[30]: 'Elemwise{Composite{sqrt(((i0 + i1) * i0))}(i0, i1) ** exp(Composite{sqrt(((i0 -
in))))
In [31]: print uselessFunc(1,4)
7.52932798092
    Where's the vector stuff
In [32]: x = T.vector('x')
         y = T.vector('y')
         A = np.asarray([1,2,3])
         B = np.asarray([4,5,6])
In [33]: xdoty = T.dot(x,y)
         xaddy = T.sum(x+y)
         dotfn = theano.function([x,y], xdoty)
         print "Lets do dot product in theano"
        print A,B,dotfn(A,B)
         print "\nFunctions with more than one outputs"
```

```
dotaddfn = theano.function([x,y], [xdoty,xaddy])
        print dotaddfn(A,B)
        print "\n All element wise operations are similar to numpy"
Lets do dot product in theano
[1 2 3] [4 5 6] 32.0
Functions with more than one outputs
[array(32.0), array(21.0)]
All element wise operations are similar to numpy
5.1.1 The famous logistic function
In [34]: x = T.matrix('x')
        s = 1 / (1 + T.exp(-x))
        logistic = theano.function([x], s)
        print theano.pp(logistic.maker.fgraph.outputs[0])
        logistic([[0, 1], [-1, -2]])
sigmoid(x)
[ 0.26894142, 0.11920292]])
5.2
     The update comes in
In [35]: state = theano.shared(0)
        inc = T.iscalar('inc')
        #Update the state by incrementing it with inc
        accumulator = theano.function([inc], state, updates=[(state, state+inc)])
In [36]: for i in range(0,10):
            accumulator(i)
            # In order to get the value of the accumulated
            print state.get_value()
        # We can also set the value of a shared variable
        state.set_value(0)
0
1
3
6
10
15
21
28
36
45
```

- 5.3 As you might have guessed ML is a lot about updating parameters to achieve lowest cost
- 5.4 But then we need to choose what to update it with
- 5.5 Gear up for some magic
- 5.6 Gradient Magic

```
In [37]: a = T.scalar('a')
        b = T.sqr(a)
        c = T.grad(b,a)
        gradfn = theano.function([a],c)
        print theano.pp(gradfn.maker.fgraph.outputs[0])
        print gradfn(4)
(TensorConstant{2.0} * a)
8.0
In [38]: B = theano.shared(np.asarray([1.,2.]))
        R = T.sqr(B).sum()
        A = T.grad(R, B)
        Z = theano.function([], R, updates={B: B - .1*A})
        for i in range(10):
            print('cost function = {}'.format(Z()))
            print('parameters = {}'.format(B.get_value()))
        # Try to change range to 100 to see what happens
cost function = 5.0
parameters = [0.8 1.6]
cost function = 3.2
parameters = [0.64 1.28]
cost function = 2.048
parameters = [ 0.512 1.024]
cost function = 1.31072
parameters = [0.4096 \ 0.8192]
cost function = 0.8388608
parameters = [0.32768 \ 0.65536]
cost function = 0.536870912
parameters = [0.262144 \ 0.524288]
cost function = 0.34359738368
parameters = [0.2097152 \ 0.4194304]
cost function = 0.219902325555
parameters = [0.16777216 \ 0.33554432]
cost function = 0.140737488355
parameters = [0.13421773 \ 0.26843546]
cost function = 0.0900719925474
parameters = [0.10737418 \ 0.21474836]
In []:
In []:
In []:
```

- In []:
- In []: