#### Lecture 0 Python Tutorial

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#### Software Setup

- Working on cloud (recommended)
  - ► GitHub Codespace, Google Colaboratory
  - free access to hardware with preinstalled packages
- Working on local machine

```
$ cd path/to/homework/directory
$ python3 -m venv nn  # virtual environment
    (isolated from system-wide packages)
$ source nn/bin/activate
$ pip3 install numpy
$ python3 src/main.py  # run script or just $
        python3 to enter interactive mode
$ deactivate
```

The Road Ahead...

1 Built-in Functionality

2 External Packages

## Basic Data Type: Numbers

```
x = 3
print(type(x)) # print "<class 'int'>"
print(x) # print "3"
print(x + 1)  # addition; print "4"
print(x - 1) # subtraction; print "2"
print(x * 2) # multiplication; print "6"
print(x ** 2) # exponentiation; print "9"
x += 1
print(x) # print "4"
x *= 2
print(x) # print "8"
v = 2.5
print(type(y)) # print "<class 'float'>"
print(y, y + 1, y * 2, y ** 2) # print "2.5 3.5
   5.0 6.25"
```

## Basic Data Type: Booleans

```
t = True
f = False
print(type(t)) # print "<class 'bool'>"
print(t and f) # logical AND; print "False"
print(t or f) # logical OR; print "True"
print(not t) # logical NOT; print "False"
print(t != f) # logical XOR; print "True"
```

## Basic Data Type: Strings

```
a = 'hello' # single quotes string literals
b = "world"  # double quotes string literals
print(a) # print "hello"
print(len(a)) # string length; print "5"
ab = a + ' ' + b # concatenation
print(ab) # print "hello world"
ab12 = f'\{a\} \{b\} \{12\}' \# format
print(ab12) # print "hello world 12"
print(a.capitalize()) # capitalize; print "Hello"
print(a.rjust(7))  # right-justify with space
  padding; print " hello"
print(a.center(7))  # center with space padding
   ; print " hello "
print(a.replace('l', '(ell)')) # replace all
   instances of one substring with another; print
    "he(ell)(ell)o"
print(' world '.strip())  # strip leading
   and trailing whitespace; print "world"
```

#### **Immutables**

```
x = 5 # primitives/immutables (cannot be
   modified in-place)
y = x # deep copy/pass by value
print(y is x) # test if x and y refer to same
   object; print "True"
y = 10 # y points to new object
print(x) # x is not affected; print "5"
print(y is x) # x and y refer to different
   objects; print "False"
print(id(x)) # check address of int object (on
   heap) that x points to, but no direct way to
   check address of x as reference (on stack)
print(id(y)) # x and y store different addresses
x += 1 # augmented assignment (+=, -=, *=,
   /=, etc.); same as x = x + 1
print(x) # print "6"
print(id(x)) # x points to new object
```

#### Container: Lists

```
xs = [3, 1, 2]  # create list (ordered)
print(xs[2])  # print "2"
print(xs[-1])  # negative index counts from
    list end; print "2"

xs[2] = 'foo'  # can contain different types
print(xs)  # print "[3, 1, 'foo']"
xs.append('bar')  # add new element to list end
print(xs)  # print "[3, 1, 'foo', 'bar']"
x = xs.pop()  # remove and return last element
print(x)  # print "bar"
```

### Container: List Slicing

```
nums = list(range(5)) # create list of integers
print(nums)
                # print "[0, 1, 2, 3, 4]"
print(nums[2:4])  # slice from index 2 to 4
   (exclusive); print "[2, 3]"
print(nums[2:]) # slice from index 2 to
   end; print "[2, 3, 4]"
print(nums[:2])  # slice from start to
   index 2 (exclusive); print "[0, 1]"
print(nums[:]) # slice of whole list;
   print "[0, 1, 2, 3, 4]"
print(nums[:-1])  # slice indices can be
   negative; print "[0, 1, 2, 3]"
nums[2:4] = [8, 9] # assign new sublist to a
   slice
print(nums)
                      # print "[0, 1, 8, 9, 4]"
```

### Container: List Comprehension

```
animals = ['cat', 'dog', 'monkey']
for animal in animals:
    print(animal) # print "cat", "dog", "monkey",
        each on its own line

nums = [0, 1, 2, 3, 4]
even_squares = [x ** 2 for x in nums if x % 2 ==
        0] # list comprehension
print(even_squares) # print "[0, 4, 16]"
```

#### Container: Dictionaries

```
d = {'cat': 'cute', 'dog': 'furry'} # create new
   dictionary (key-value pairs)
print(d['cat']) # get entry; print "cute"
print('cat' in d)  # check if dictionary has a
   given key; print "True"
d['fish'] = 'wet' # set entry
print(d['monkey']) # KeyError: 'monkey' not a
  key of d
print(d.get('monkey', 'N/A')) # get element with
   default; print "N/A"
print(d.get('fish', 'N/A')) # print "wet"
del d['fish'] # remove element
print(d.get('fish', 'N/A')) # "fish" is no
   longer a key; print "N/A"
```

## Container: Dictionary Comprehension

```
d = {'person': 2, 'cat': 4, 'spider': 8}
for animal in d:
    legs = d[animal]
    print(f'A {animal} has {legs} legs')
# print "A person has 2 legs", "A cat has 4 legs",
    "A spider has 8 legs"
nums = [0, 1, 2, 3, 4]
even_num_to_square = {x: x ** 2 for x in nums if x
    % 2 == 0}
print(even_num_to_square) # print "{0: 0, 2: 4,
   4: 16}"
```

#### Container: Sets

```
animals = {'cat', 'dog'} # create set (unordered
   collection of distinct elements)
print('cat' in animals) # check if element is in
    a set; print "True"
print('fish' in animals) # print "False"
animals.add('fish')
                  # add element
print('fish' in animals) # print "True"
print(len(animals))
                        # number of elements;
  print "3"
animals.add('cat')
                         # adding existing
   element does nothing
print(len(animals))
                         # print "3"
animals.remove('cat')
                         # remove element
print(len(animals))
                         # print "2"
```

#### Container: Set Comprehension

```
animals = {'cat', 'dog', 'fish'}
for idx, animal in enumerate(animals):
    print(f'#{idx+1}: {animal}')
# print "#1: fish", "#2: dog", "#3: cat"

from math import sqrt
nums = {int(sqrt(x)) for x in range(30)}
print(nums) # print "{0, 1, 2, 3, 4, 5}"
```

#### Container: Tuples

#### Mutables

```
num1 = [1, 2, 3, 4, 5] # non-primitives/mutables
   (can be modified in-place)
num2 = num1  # shallow copy/pass by reference
   (indeed still pass by value of 'reference')
print(num2 is num1) # num1 and num2 refer to
   same object; print "True"
              # also affect num1
num2[0] = 10
print(num2 is num1) # print "True"
print(id(num1))
print(id(num2)) # num1 and num2 are different
   references (holding same address on stack,
   stack overflow?) pointing to same object on
   heap (buffer overflow?)
num1 += [6] # in-place augmented assignment
print(num1) # print "[10, 2, 3, 4, 5, 6]"
print(id(num1)) # num1 points to same object
num1 = num1 + [7]  # create new object
print(id(num1)) # num1 points to new object
```

#### **Function**

```
def hello(name, loud=False):
    if loud:
        print(f'HELLO, {name.upper()}!')
    else:
        print(f'Hello, {name}')

hello('Bob')  # print "Hello, Bob"
hello('Fred', loud=True) # print "HELLO, FRED!"
```

#### Class: Encapsulation

```
class Vehicle: # superclass; encapsulation
    def __init__(self, make, model, year):
        self.make = make
        self.model = model
        self.year = year
    def display_info(self):
        print(f"Vehicle Info: {self.year} {self.
            make } { self . model } ")
    def start_engine(self):
        print("Engine started")
```

### Class: Inheritance & Polymorphism

```
class Car(Vehicle): # subclass; inheritance
   def __init__(self, make, model, year, doors):
        super().__init__(make, model, year)
        self.doors = doors
   def display_info(self): # method overriding;
       polymorphism
        super().display_info()
        print(f"Number of doors: {self.doors}")
   def honk(self):
       print("Car honks: Beep beep!")
car = Car("Toyota", "Camry", 2020, 4)
car.display_info() # print "Vehicle Info: 2020
   Toyota Camry", "Number of doors: 4"
car.start_engine() # print "Engine started"
car.honk() # print "Car honks: Beep beep!"
```

The Road Ahead...

Built-in Functionality

2 External Packages

## NumPy: Arrays

```
import numpy as np
a = np.array([1, 2, 3], dtype=np.int64) # create
   1-D array (neither row nor column vector)
print(type(a))
                          # print "<class 'numpy.</pre>
   ndarray'>"
print(a.shape)
                   # print "(3,)"
print(a[0], a[1], a[2]) # print "1 2 3"
a[0] = 5
                          # in-place modification
print(a)
                          # print "[5, 2, 3]"
b = np.array([[1,2,3],[4,5,6]]) # create 2-D
   array (including row or column vector)
print(b.shape)
                                  # print "(2, 3)"
print(b[0, 0], b[0, 1], b[1, 0]) # print "1 2 4"
c = np.array([[1, 2, 3]])
                                  # row vector
print(c.shape)
                                  # print "(1, 3)"
d = np.array([[1], [2], [3]]) # column vector
print(d.shape)
                                  # print "(3, 1)"
```

## NumPy: Arrays (Cont'd)

```
a = np.zeros((2,2))
                    # array of all zeros
print(a)
                    # print "[[ 0. 0.]
                     [ 0. 0.1]"
b = np.ones((1,2)) # array of all ones
                    # print "[[ 1. 1.]]"
print(b)
c = np.full((2,2), 7) \# constant array
print(c)
                    # print "[[ 7. 7.]
                    # [7.7.]]"
d = np.eye(2)
             # identity matrix
print(d)
                    # print "[[ 1. 0.]
                    # [ 0. 1.]]"
e = np.random.random((2,2)) # random array
print(e)
              # "[[ 0.91940167  0.08143941]
              # [ 0.68744134  0.87236687]]"
print(e.dtype)
                    # print "float64"
```

## NumPy: Array Indexing

```
a = np.array([[1,2,3,4], [5,6,7,8], [9,10,11,12]])
b = a[:2, 1:3] # slice (:) to extract subarray
print(b) # print "[[2 3]
              # [6 7]]"
print(a[0, 1]) # print "2"
b[0, 0] = 77 # b[0, 0], a[0, 1] point to same
   data, so changing one will change the other
print(a[0, 1]) # print "77"
r1 = a[1, :] # integer indexing, 1-D array
r2 = a[1:2, :] # slice indexing, 2-D array
print(r1, r1.shape) # print "[5 6 7 8] (4,)"
print(r2, r2.shape) # print "[[5 6 7 8]] (1, 4)"
c1 = a[:, 1]
c2 = a[:, 1:2]
print(c1, c1.shape) # print "[77 6 10] (3,)"
print(c2, c2.shape) # print "[[77]
                     [ 6]
                             [10]] (3, 1)"
```

## NumPy: Array Indexing (Cont'd)

```
a = np.array([[1,2], [3, 4], [5, 6]])
print(a[[0, 0], [1, 1]]) # returned array has
   shape (2,); print "[2 2]"
print(np.array([a[0, 1], a[0, 1]])) # equivalent
b = np.array([0, 1, 0]) # index array
print(a[np.arange(3), b]) # select one element
   from each row; print "[ 1 4 5]"
a[np.arange(3), b] += 10 # mutate one element
   from each row
print(a)
          # print "array([[11, 2],
                                 [3, 14],
                                  [15, 6]])"
bool_idx = (a > 10) # find elements bigger than 10
print(bool_idx) # print "[[ True False]
                   # [False True]
                  # [ True Falsell"
print(a[bool_idx]) # print "[11 14 15]"
print(a[a > 10]) # equivalent
```

### NumPy: Array Math

```
x = np.array([[1,2],[3,4]], dtype=np.float64)
y = np.array([[5,6],[7,8]], dtype=np.float64)
            # elementwise sum
print(x + y)
print(np.add(x, y)) # print "[[ 6.0 8.0]
                    # [10.0 12.0]]"
print(x - y)
              # elementwise difference
print(np.subtract(x, y)) # print "[[-4. -4.]
                        # [-4. -4.11"
print(x * y)
            # elementwise product
print(np.multiply(x, y)) # print "[[ 5. 12.]
                        # [21. 32.]]"
print(x / y)
             # elementwise division
print(np.divide(x, y))# print "[[0.2 0.33333333]
                              [0.42857143 0.5]]"
print(np.sqrt(x))
                # elementwise square root
                    # print "[[1. 1.41421356]
                              [1.73205081 2.]]"
```

## NumPy: Array Math (Cont'd)

```
x = np.array([[1,2],[3,4]]) # 2-D arrays
y = np.array([[5,6],[7,8]])
v = np.array([9,10]) # 1-D arrays
w = np.array([11, 12])
print(v.dot(w)) # inner product
print(np.dot(v, w)) # print "219"
print(v @ w)
                    # equivalent; np.matmul(v, w)
print(x.dot(v))  # matrix-vector product
print(np.dot(x, v)) # print "[29 67]"
print(x @ v)
                    # equivalent; np.matmul(x, v)
print(v.dot(x))  # vector-matrix product
print(np.dot(v, x)) # print "[39 58]"
print(v @ x)
                    # equivalent; np.matmul(v, x)
print(x.dot(y))  # matrix-matrix product
print(np.dot(x, y))
                    # print "[[19 22]
                    # [43 50]]"
print(x @ y)
                    # equivalent; np.matmul(x, y)
```

## NumPy: Array Math (Cont'd)

```
x = np.array([[1,2],[3,4]])
print(np.sum(x))
                       # sum of all elements;
   print "10"
print(np.sum(x, axis=0)) # sum of each column;
   print "[4 6]"
print(np.sum(x, axis=1)) # sum of each row; print
    "[3 7]"
                         # print "[[1 3]
print(x.T)
                         # [2 4]]"
v = np.array([1,2,3])
                         # transpose of 1-D array
    does nothing
                         # print "[1 2 3]"
print(v.T)
```

## NumPy: Array Broadcasting

```
# Outer product of vectors
v = np.array([1,2,3]) # v has shape (3,)
w = np.array([4,5]) # w has shape (2,)
print(np.reshape(v, (3, 1)) * w) # "[[ 4 5]
                                 # [12 15]]"
# Add vector to each row of a matrix
x = np.array([[1,2,3], [4,5,6]])
                   # "[[2 4 6]
print(x + v)
                          [5 7 9]]"
# Add vector to each column of a matrix
print((x.T + w).T)
print(x + np.reshape(w, (2, 1))) # "[[ 5 6 7]
                                # [ 9 10 11]]"
# Multiply matrix by a constant
print(x * 2)
              # "[[ 2 4 6]
                      # [8 10 12]]"
```

### PyTorch: Tensors

```
import torch
V = torch.tensor([1, 2, 3]) # create 1-D
   tensor
                  # print "tensor([1, 2, 3])"
print(V)
print(V[0]) # print "tensor(1)"
print(V[0].item())  # print "1"
M = torch.tensor([[1, 2], [3, 4]]) # create 2-D
   tensor
print(M)
                    # print "tensor([[1, 2],
                                     [3, 4]])"
print(M[0])
                    # print "tensor([1, 2])";
   same as M[0, :]
print(M.view(1, 4)) # reshape; print "tensor
   ([[1, 2, 3, 4]])"
W = torch.tensor([5, 6, 7])
print(V + W)  # print "tensor([6, 8, 10])"
```

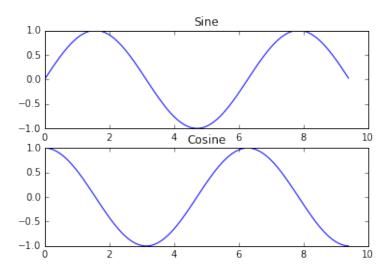
#### SciPy

```
from scipy import linalg, optimize, stats
# Solve linear system of equations
A = np.array([[3, 2, 0], [1, -1, 0], [0, 5, 1]])
b = np.array([2, 4, -1])
x = linalg.solve(A, b)
print(x) # print "[ 2. -2. 9.]"
# Optimization
def f(x):
   return x**2 + 5 * np.sin(x)
result = optimize.minimize(f, x0=0)
print(result.x) # print "[-1.11051081]"
# Statistics
data = np.array([1, 2, 2, 3, 4, 5, 6, 7, 8, 9])
mode = stats.mode(data)
print(mode.mode) # print "[2]"
```

## Matplotlib

```
import matplotlib.pyplot as plt
x = np.arange(0, 3 * np.pi, 0.1) # grid points
y_sin = np.sin(x)
y_{cos} = np.cos(x)
plt.subplot(2, 1, 1) # first subplot
plt.plot(x, y_sin)
plt.title('Sine')
plt.subplot(2, 1, 2) # second subplot
plt.plot(x, y_cos)
plt.title('Cosine')
plt.show()
                      # display plots
```

# Matplotlib (Cont'd)



#### References

- cs231n.stanford.edu CS231n: Deep Learning for Computer Vision
- numpy.org NumPy package for scientific computing
- pytorch.org PyTorch package for deep learning
- scipy.org SciPy package for scientific computing
- matplotlib.org Matplotlib package for visualization