# Python Tutorial

March 7, 2020

## 1 Installation Prerequsites

### 1.1 Anaconda Installation

Anaconda / Conda

- Platform that allows you to manage python versions and packages through "environments"
- Includes pip python's package manager that can be used to expand the capabilities of a base python environment
- Each environment can run different versions of python
- Each environment has a different set of installed packages

Install Link: https://www.anaconda.com/distribution/

- Download Anaconda Python 3.7 version
- Use default settings for installation
- If your username has a space in it, anaconda will warn that this might cause install issues

### 1.2 Create Conda Environment

Windows: Open "Anaconda Prompt" through Start menu

Mac: Open "Terminal" through spotlight search

Enter the following command to create a new environment named "py37" with python version 3.7

conda create --name py37 python=3.7

Activate the environment you just created

conda activate py37

Conda Cheet Sheet is a useful reference: http://bit.ly/2xknudL

### 1.3 Install Packages

Many features are already included in python, but many packages can be installed to expand the capabilities of python

These packages can be installed using pip - python's package installer - or conda

## 1.3.1 jupyter-notebook

• a web-based interactive development environment (IDE) that allows you to interactively develop python scripts, analyze and visualize data

- A notebook is comprised of a series of cells; can either be a python code cell or a markdown text cell
- Can execute the contents of the active cell by clicking (Shift+Enter)
- Code cells are run in the order you execute them Install jupyter-notebook and add our conda environment as a python kernel (the python environment in which we'll run our code)

```
pip install jupyter
python -m ipykernel install --user --name=py37
```

### 1.3.2 Matplotlib

• A package for creating visualizations

pip install matplotlib

### 1.3.3 Pandas

- Used for working with datasets
- Load data into pandas DataFrame
- Can query large datasets

pip install pandas

## 1.3.4 Numpy

• Another useful tool for mathematical operations and analyzing data

pip install numpy

### 1.3.5 Sklearn

• Another useful tool for mathematical operations and analyzing data

pip install sklearn

## 1.4 Start-up jupyter

In Terminal/ Anaconda Prompt run:

jupyter-notebook

This will open the jupyter console in your web browser

## 1.4.1 Unix (mac terminal) Commands

Check current directory

pwd

List directory contents

ls

Change directory

cd < insert directory name without brackets >

Make new directory

mkdir < directory name >

## 2 Getting Started with python

Import packages that we'll be using

```
[1]: import numpy as np
  import pandas as pd
  # iPython magic to allow interactive plots
  %matplotlib notebook
  # Import for 3D plots
  from mpl_toolkits.mplot3d import Axes3D
  # Plotting library
  import matplotlib.pyplot as plt
  # Colormap Library
  from matplotlib import cm
  import sklearn
```

## 2.1 Variables

[6]: 6

## 2.1.1 Integers and Floating Point Numbers

```
[2]: # Define two integers
    num1 = 4
    num2 = 3
    total = num1 + num2
    product = num1 * num2
    ratio = num1 / num2

# int + float = float
    float1 = .125
    f1 = num1 + float1

[3]: f1

[3]: 4.125

[4]: num1

[4]: 4

[5]: num1 += 2
```

### **Printing**

```
[7]: print("sum:",total, "product:",product, "ratio:", ratio,"\n")
     print("sum: {}\nproduct: {}\nratio: {:.3f}".format(total, product, ratio))
     sum: 7
     product: 12
     ratio: 1.333
     2.1.2 Bool
 [8]: print("num1: ",num1, "\nnum1 > 3: ", num1 > 3)
     num1: 6
     num1 > 3: True
     2.1.3 Strings
     https://www.w3schools.com/python/python_ref_string.asp
 [9]: first_half = "the quick brown fox"
     second_half = "jumped over the lazy dog"
     full_sentence = first_half + " " + second_half
[10]: full_sentence
[10]: 'the quick brown fox jumped over the lazy dog'
[11]: first_half.replace("fox","bear")
[11]: 'the quick brown bear'
     Check the docs to see if the function returns the result or sets the value in-place
[12]: first_half
[12]: 'the quick brown fox'
[13]: first_half[0]
[13]: 't'
[14]: first_half[4:9]
[14]: 'quick'
[15]: second_half[second_half.find("the") + 4 : second_half.find("dog")]
```

```
[15]: 'lazy'
```

## 2.1.4 Lists, Tuples

https://www.w3schools.com/python/python\_ref\_list.asp

This is an example of an in-place function

You can index a list and get a subsequence

[20]: 2

[21]: [5, 4]

[22]: [3, 2]

Lists are mutable

[23]: [5, 4, 3, 2, 1]

Tuples are not mutable

```
[26]: # try running this code block and if an error gets raised do that
      try:
         tup_1[1] = 5
      except:
         print("TUPLES ARE NOT MUTABLE")
     TUPLES ARE NOT MUTABLE
     2.1.5 Sets
[27]: fruits = set(["apple", "orange", "banana"])
      vegetables = set(["broccoli", "carrot", "lettuce", "olives"])
      food_I_dont_like = set(["olives", "anchovy"])
[28]: fruits
[28]: {'apple', 'banana', 'orange'}
[29]: fruits.union(vegetables)
[29]: {'apple', 'banana', 'broccoli', 'carrot', 'lettuce', 'olives', 'orange'}
[30]: vegetables.difference(food_I_dont_like)
[30]: {'broccoli', 'carrot', 'lettuce'}
     2.1.6 Dictionaries
[31]: me = {"first name": "Dan", "last name": "Zeiberg", "age": 24}
      pedja = {"first name": "Predrag", "last name": "Radivojac"}
[32]: people = [me, pedja]
[33]: people
[33]: [{'first name': 'Dan', 'last name': 'Zeiberg', 'age': 24},
      {'first name': 'Predrag', 'last name': 'Radivojac'}]
     2.1.7 Nump Arrays, Matrices
[34]: arr_1 = np.array([6,2,6,8,1,6,22,6,8,999])
      arr_2 = np.random.randint(0,25,size=10)
[35]: arr 1
[35]: array([ 6, 2, 6, 8, 1, 6, 22,
                                                 6, 8, 999])
```

You can do arithmetic on arrays

3

0.03237

0.0

2.18

```
[36]: arr_2 / arr_1
[36]: array([3.5
                       , 6.5
                                    , 2.83333333, 2.375
                                                            , 1.
                       , 0.95454545, 3.83333333, 0.875
                                                            , 0.01101101])
     matrix_0 = np.array([[1,2,3],[4,5,6],[7,8,9]])
[38]: matrix_0
[38]: array([[1, 2, 3],
             [4, 5, 6],
             [7, 8, 9]])
     Indexing Matrices
[39]: matrix_0[1,2]
[39]: 6
     Numpy allows you to draw from statistical distributions
[40]: matrix 1 = np.random.normal(loc=0, scale=1, size=[5,4])
[41]: matrix_1
[41]: array([[ 1.65423992, 1.71262782, -1.13080366, 0.39229579],
             [-1.28035338, -0.27232278, -2.24187914,
                                                       0.37101365],
             [0.84828532, 0.64919246, -1.16023924,
                                                       0.14783276],
             [0.03972101, -0.73081974, -0.63169612, -0.9884438],
             [ 0.14377593, 1.58346939, 1.01394705, 0.39110502]])
         Pandas DataFrames
     3
     Load DataFrame from file
     boston housing dataset: https://www.kaggle.com/puxama/bostoncsv
[42]: housing = pd.read_csv("../data/BostonHousing.csv")
[43]: housing
[43]:
              crim
                          indus
                                                                   rad
                                                                         tax \
                      zn
                                 chas
                                          nox
                                                  rm
                                                       age
                                                               dis
           0.00632
                    18.0
                           2.31
                                       0.538
                                               6.575
                                                      65.2
                                                            4.0900
                                                                         296
      0
                                    0
                                                                      1
           0.02731
                                       0.469
      1
                     0.0
                           7.07
                                    0
                                               6.421
                                                      78.9
                                                            4.9671
                                                                      2
                                                                         242
      2
           0.02729
                     0.0
                           7.07
                                       0.469
                                              7.185
                                                      61.1
                                                            4.9671
                                                                      2
                                                                         242
                                    0
```

45.8

6.0622

222

0 0.458 6.998

```
222
      4
           0.06905
                     0.0
                           2.18
                                     0 0.458 7.147 54.2 6.0622
                                        0.573
      501
          0.06263
                     0.0 11.93
                                               6.593
                                                      69.1
                                                            2.4786
                                                                          273
          0.04527
                          11.93
                                        0.573
                                               6.120
                                                      76.7
                                                                          273
      502
                     0.0
                                                            2.2875
      503 0.06076
                     0.0 11.93
                                        0.573
                                               6.976
                                                      91.0
                                                            2.1675
                                                                          273
                                     0
                     0.0 11.93
                                        0.573
                                                                          273
      504 0.10959
                                     0
                                               6.794
                                                      89.3
                                                            2.3889
      505 0.04741
                     0.0 11.93
                                        0.573
                                               6.030
                                                      80.8
                                                            2.5050
                                                                          273
           ptratio
                            lstat medv
                         b
              15.3
                    396.90
                             4.98
                                    24.0
      0
      1
              17.8
                    396.90
                             9.14
                                   21.6
      2
              17.8 392.83
                             4.03
                                   34.7
      3
              18.7
                    394.63
                             2.94
                                   33.4
      4
              18.7
                    396.90
                             5.33
                                   36.2
                                    22.4
      501
              21.0
                    391.99
                             9.67
      502
              21.0 396.90
                             9.08
                                   20.6
      503
              21.0 396.90
                             5.64
                                   23.9
      504
              21.0 393.45
                              6.48
                                   22.0
      505
              21.0 396.90
                             7.88 11.9
      [506 rows x 14 columns]
[44]: people_df = pd.DataFrame(people)
[45]: people_df
[45]:
        first name
                    last name
                                 age
                      Zeiberg
      0
               Dan
                               24.0
      1
           Predrag
                   Radivojac
                                 {\tt NaN}
[46]: # each element in the dataframe is uniquely identified by it's index value
      # Lets set the index for the people DataFrame to "last name"
      people_df = people_df.set_index("last name")
[47]:
     people_df
[47]:
                first name
                              age
      last name
      Zeiberg
                       Dan
                            24.0
      Radivojac
                             NaN
                   Predrag
     Get the age of the person represented by the first row
     Can index a dataframe by row number
[48]: housing.iloc[0]["age"]
```

```
[48]: 65.2
```

Or you can index by the index (key) and column name

Get Dan's age

```
[49]: people_df.loc["Zeiberg","age"]
```

[49]: 24.0

Extract Column of DataFrame, convert to numpy array, limit to first 10 values

```
[50]: housing["age"].values[:10]
```

You can query a DataFrame

```
[51]: housing[housing["age"] > 33]
```

[51]:		crim	zn	indus	chas	nox	rm	age	dis	rad	tax	\
	0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	
	1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	
	2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	
	3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	
	4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	
		•••		•••	•••		•••	• •••				
	501	0.06263	0.0	11.93	0	0.573	6.593	69.1	2.4786	1	273	
	502	0.04527	0.0	11.93	0	0.573	6.120	76.7	2.2875	1	273	
	503	0.06076	0.0	11.93	0	0.573	6.976	91.0	2.1675	1	273	
	504	0.10959	0.0	11.93	0	0.573	6.794	89.3	2.3889	1	273	
	505	0.04741	0.0	11.93	0	0.573	6.030	80.8	2.5050	1	273	

```
ptratio
                     lstat
                             medv
                   b
0
        15.3
              396.90
                       4.98
                             24.0
1
        17.8
              396.90
                       9.14
                             21.6
2
                             34.7
        17.8
              392.83
                       4.03
        18.7
                       2.94
                             33.4
3
              394.63
4
        18.7
              396.90
                       5.33
                             36.2
                       9.67
501
        21.0
              391.99
                              22.4
502
        21.0
              396.90
                       9.08
                             20.6
        21.0
                       5.64
503
              396.90
                             23.9
504
        21.0
              393.45
                       6.48
                             22.0
```

396.90

7.88 11.9

[425 rows x 14 columns]

21.0

505

```
[52]: housing[(housing["age"] > 33) & (housing["tax"] <= 350)]
[52]:
                     zn indus
                                                                     tax \
             crim
                               chas
                                       nox
                                                            dis rad
                                               rm
                                                    age
     0
          0.00632
                   18.0
                          2.31
                                     0.538
                                            6.575
                                                   65.2
                                                         4.0900
                                                                  1
                                                                     296
                                  0
                                     0.469
     1
          0.02731
                    0.0
                          7.07
                                            6.421
                                                   78.9
                                                         4.9671
                                                                     242
     2
          0.02729
                    0.0
                          7.07
                                  0 0.469
                                            7.185
                                                   61.1
                                                        4.9671
                                                                     242
     3
          0.03237
                    0.0
                          2.18
                                     0.458
                                            6.998
                                                   45.8
                                                         6.0622
                                                                     222
     4
          0.06905
                    0.0
                          2.18
                                     0.458
                                            7.147
                                                   54.2
                                                         6.0622
                                                                     222
     501
         0.06263
                    0.0 11.93
                                  0
                                     0.573
                                            6.593 69.1
                                                        2.4786
                                                                  1
                                                                     273
                                                                     273
     502 0.04527
                    0.0 11.93
                                     0.573
                                            6.120
                                                   76.7
                                                        2.2875
                                                                  1
                                  0
     503 0.06076
                    0.0 11.93
                                     0.573
                                            6.976
                                                   91.0
                                                        2.1675
                                                                     273
     504 0.10959
                    0.0 11.93
                                     0.573
                                            6.794
                                                   89.3
                                                                     273
                                                         2.3889
                    0.0 11.93
                                     0.573 6.030
     505 0.04741
                                                   80.8
                                                        2.5050
                                                                     273
                          lstat medv
          ptratio
                       b
                           4.98
     0
             15.3 396.90
                                 24.0
     1
             17.8 396.90
                            9.14 21.6
     2
             17.8 392.83
                            4.03
                                 34.7
     3
             18.7 394.63
                            2.94
                                 33.4
     4
             18.7
                   396.90
                            5.33 36.2
     . .
                            •••
              •••
                    •••
     501
             21.0 391.99
                           9.67
                                 22.4
     502
             21.0 396.90
                           9.08
                                 20.6
     503
             21.0 396.90
                            5.64
                                 23.9
     504
             21.0 393.45
                            6.48 22.0
     505
             21.0 396.90
                            7.88 11.9
     [202 rows x 14 columns]
     3.1 Joining DataFrames
[53]: people_df
[53]:
               first name
                            age
     last name
     Zeiberg
                      Dan
                          24.0
     Radivojac
                  Predrag
                            NaN
[54]: publications_df = pd.DataFrame([
         →Positive-Unlabeled Classification Setting",
          "year": 2020,
          "first author": "Zeiberg"},
         {"title": "Prediction of boundaries between intrinsically ordered and \sqcup
      →disordered protein regions",
          "year": 2003,
```

```
"first author": "Radivojac"}])
[55]: publications_df
[55]:
                                                     title year first author
      O Fast Nonparametric Estimation of Class Proport... 2020
                                                                    Zeiberg
      1 Prediction of boundaries between intrinsically... 2003
                                                                  Radivojac
[56]: people_and_publication = people_df.merge(right=publications_df, how="left", __
       →left_on="last name", right_on="first author")
[57]: people_and_publication
[57]: first name
                     age
                                                                      title year \
               Dan 24.0 Fast Nonparametric Estimation of Class Proport... 2020
      1
           Predrag
                     NaN Prediction of boundaries between intrinsically... 2003
        first author
             Zeiberg
           Radivojac
[58]: # remove the first author column
      people_and_publication = people_and_publication.drop("first author",axis=1)
[59]: people_and_publication = people_and_publication.rename(columns={"title":
       →"publication name"})
[60]: people_and_publication
[60]:
       first name
                                                           publication name year
                     age
               Dan 24.0 Fast Nonparametric Estimation of Class Proport... 2020
      0
      1
                     NaN Prediction of boundaries between intrinsically... 2003
           Predrag
         Conditionals and Loops
[61]: randnum = np.random.choice([1,2,3,4])
      if randnum == 1:
          print("chose first value")
      elif randnum == 2:
          print("chose second value")
      elif randnum == 3:
         print("chose third value")
      else:
          print("choes fourth value")
```

#### chose second value

You can repeat a block of code using for loops and while loops

### For loops

iterate over a list of values, this can be loop indices or other data

```
[62]: for i in range(10):
          print(i)
     0
     1
     2
     3
     4
     5
     6
     7
     8
     9
[63]: for person in people:
          print(person["first name"])
     Dan
     Predrag
     Loop over matrix values
[64]: # Loop over each row
      for r in range(matrix_1.shape[0]):
          # Loop over each column
          for c in range(matrix_1.shape[1]):
              print(matrix_1[r,c])
          print()
     1.65423992031951
     1.712627822242182
     -1.1308036574487206
     0.3922957878874731
     -1.2803533825104598
     -0.2723227779931885
     -2.2418791430036276
     0.3710136540528735
     0.8482853165237828
     0.6491924645783096
     -1.1602392378006166
```

```
0.14783275760652517
```

```
0.039721006981576455
```

- -0.7308197379929903
- -0.6316961216431078
- -0.988443795217655
- 0.14377593402483918
- 1.5834693931259611
- 1.0139470508658355
- 0.39110501887619814

### While Loops

Continue executing a block of code while the specified condition is false

```
[65]: values = [1,7,1,6,888,221,5]
    idx = 0
    while values[idx] < 50:
        idx += 1
    print("first big number is ",values[idx])</pre>
```

first big number is 888

## 5 Functions

```
[66]: def add(a,b):
    return a+b

[67]: add(4,5)

[67]: 9

[68]: def factorial(x):
    if x > 0:
        return x * factorial(x-1)
    elif x == 0:
        return 1
    else:
        raise Exception("Input must be non-negative")
[69]: factorial(5)
```

[69]: 120

## 6 Classes

```
[70]: class Person:
          def __init__(self, name, hair_color, eye_color):
             self.name = name
              self.hair_color = hair_color
              self.eye_color = eye_color
          def introduce(self):
             print("Hi, my name is {}".format(self.name))
          def converse(self, other_person):
             print("Hi {}, how are you today?".format(other_person.name))
[71]: dan = Person("Dan", "brown", "green")
      emily = Person("Emily", "blond", "blue")
      dan.introduce()
      emily.converse(dan)
     Hi, my name is Dan
     Hi Dan, how are you today?
     7 Plotting
[72]: ph = pd.read_csv("../data/pH-example.txt")
[73]: ph
[73]:
          time
          0.00 7.000
      1
          0.01 6.999
      2
          0.02 6.998
          0.03 6.997
      3
      4
          0.04 6.996
      96
          0.96 6.991
      97
          0.97 6.990
      98
          0.98 6.989
      99
          0.99 6.988
      100 1.00 6.987
      [101 rows x 2 columns]
[74]: fig,ax = plt.subplots(1,1)
      ax.plot(ph["time"], ph["v"])
     <IPython.core.display.Javascript object>
```

```
<IPython.core.display.HTML object>
[74]: [<matplotlib.lines.Line2D at 0xa218ec978>]
[75]: fig, ax = plt.subplots(1,1)
     hist = ax.hist(housing["age"],bins=30)
     <IPython.core.display.Javascript object>
     <IPython.core.display.HTML object>
     3D-Plot
[76]: from scipy.stats import multivariate_normal as mvn
[77]: grid = np.zeros((20,20))
      for i in range(grid.shape[0]):
          for j in range(grid.shape[1]):
              grid[i,j] = mvn.pdf([i,j],[9,9],[[5,0],[0,5]])
[78]: fig = plt.figure()
      ax = fig.gca(projection='3d')
      X,Y = np.meshgrid(list(range(grid.shape[0])), list(range(grid.shape[1])))
      surface = ax.plot_surface(X,Y,grid, cmap=cm.coolwarm)
     <IPython.core.display.Javascript object>
     <IPython.core.display.HTML object>
        Example
[79]: ph_arr = ph.values
[80]: delta = ph_arr[1:,1] - ph_arr[:-1,1]
```

```
axes[1].set_ylabel("$\Delta$ pH")
xlab = axes[1].set_xlabel("Time, h")

<IPython.core.display.Javascript object>

<IPython.core.display.HTML object>

[]:
```