Introduction to Python

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Python

What is Python

Python is a programming language that lets you work quickly and integrate systems more effectively. It's simple, easy to learn syntax emphasizes readability and therefore reduces the cost of program maintenance. Python supports modules and packages, which encourages program modularity and code reuse. The Python interpreter and the extensive standard library are available in source or binary form without charge for all major platforms, and can be freely distributed. Python.org (<a href="https://www.python.org/):

Why should you know Python?

In academic and research we can use Python and its lots of labraries to do wide range of things from basic calculation to complex statistical methods including machine learning. Since it is open-source, it is free to use.

Where can I get it?

There are many ways to install Python and its libraries. However, we are going to guide you in a pretty stable way, Anaconda.

Anaconda Cloud

Anaconda is a platform for Python that bundles most of the essential packages required in research. You can download it from https://www.anaconda.com/download/). *Note that we will use Python version 3.x in this course.*

Anaconda.org (https://anaconda.org/):

Anaconda Cloud is where data scientists share their work. You can search and download popular Python and R packages and notebooks to jumpstart your data science work.

Package Managers

We do not need to install any additional packages in this class. However, when you need it in the future recall these two names: pip and conda. For more information you can check this. (https://conda.io/docs/user-guide/tasks/manage-pkgs.html#installing-packages) out.

Jupyter Notebook

What is Jupyter Notebook?

The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code, equations, visualizations and narrative text. Uses include: data cleaning and transformation, numerical simulation, statistical modeling, data visualization, machine learning, and much more. Here are a few examples of Jupyter Notebooks projects:

- Machine Learning (https://nbviewer.jupyter.org/github/rhiever/Data-Analysis-and-Machine-Learning-Projects/blob/master/example-data-sciencenotebook/Example%2520Machine%2520Learning%2520Notebook.ipynb)
- Signal Processing (https://www.gw-openscience.org/s/events/GW150914/GW150914_tutorial.html)

How to use it?

This section will list the most used command shortcuts. For full version of docs you can visit <u>The Jupyter Notebook documentation (http://jupyter-notebook.readthedocs.io/en/latest/notebook.html)</u>.

```
arrows: move cursor
enter: enter edit mode
esc: quit edit mode
y: change cell to code mode
m: change cell to markdown mode
ctrl + enter: run selected cell
shift + enter: run selected cell and move cursor to the next cell
h: open help panel, list of shortcuts
```

Let's code

Introduction to Python

print("a+2=",a+2)

Here we will try to write some Python built-in functions to get how it works. This is brought from official Python website. For more information follow the <u>official tutorial (https://docs.python.org/3/tutorial/)</u>.

```
In [ ]: # print function
print("Hello, I'm Python!")
```

Numbers

```
In [ ]: # arithmatic
    print(1+3) # plus
    print(2-1) # minus
    print(3*6) # multiply
    print(2**3) # power
    print(17/3) # divide
    print(17//3) # floor divide
    print(10%3) #modulo
    print(abs(-5)) #absolute
    print(max(6,5)) #max value
In [ ]: a = 3
```

Strings

```
In []: phrase = "I am your father"
    phrase1 = "Nooo!"
    print(phrase.upper()+ phrase1.lower())
    print(len(phrase))
    print(phrase[5])
    print(phrase[-1]) # Last character
    print(phrase[-2]) # second-Last character
    print(phrase[0:4])# characters from position 0 (included) to 4 (excluded)
    print(phrase[10:]) # characters from position 10 (included) to the end
    print(len(phrase)) # returns the length of a string
```

Lists

letters

```
In [ ]: fruits = ['Banana', 'Apple', 'Lime', 'Watermelon', 'Papaya', 'Banana']
        print(fruits)
        print(fruits[0]) # indexing returns the item
        print(fruits[-1]) # last item
        print(fruits[1:4]) # return new list contain items index 1 (included) to 4 (ex
        cLuded)
        print(fruits[:3])
In [ ]: | fruits = ['Banana', 'Apple', 'Lime', 'Watermelon', 'Papaya', 'Banana']
        fruits.pop() # remove Last items of the List
        print("pop:", fruits)
        fruits.append('Pineapple') # add new items at the end of the list
        print("append:", fruits)
        fruits.insert(1, 'Mango') # insert item into the list
        print("insert:", fruits)
        fruits.remove('Apple')
        print("remove:", fruits)
In []: cubes = [1, 8, 27, 65, 125]
        cubes[3] = 64 # replace item at index 3
        cubes
In [ ]: letters = ['a', 'b', 'c', 'd', 'e', 'f', 'g']
        letters[2:5] = ['C', 'D', 'E'] # replace some values
```

```
In [ ]: # concatenated (glued together) with the + operator
fruits = ['Banana', 'Apple', 'Lime', 'Watermelon', 'Papaya', 'Banana']
vegetables = ['Tomato', 'Broccoli']
print(fruits+vegetables)
```

While statements

```
In [ ]: # Fibonacci series:
    # the sum of two elements defines the next
    a, b = 0, 1
    while a < 10:
        print(a, end = ' ')
        a, b = b, a+b</pre>
```

If statements

For statements

```
In []: words = ['cat', 'window', 'defenestrate']
for w in words:
    print(w, len(w))

# Loop over a slice copy of the entire list.
# if length of being looped word longer than 6
# insert this word to index 0 of the list
for w in words[:]:
    if len(w) > 6:
        words.insert(0, w)
    print("inserted list:", words)
```

```
In [ ]: | for i in range(5):
             print(i)
         a = ['Mary', 'had', 'a', 'little', 'lamb']
        for i in range(len(a)):
             print(i, a[i])
In [ ]: | for n in range(2, 10):
             for x in range(2, n):
                 if n % x == 0:
                     print(n, 'equals', x, '*', n//x)
                     break
             else:
                 print(n, 'is a prime number')
In [ ]: for num in range(2, 10):
             if num % 2 == 0:
                 print("Found an even number", num)
                 continue
             print("Found a number", num)
```

Defining functions

```
In [ ]: # Python 3: Fibonacci series up to n
def fib(n):
    a, b = 0, 1
    while a < n:
        print(a, end=' ')
        a, b = b, a+b
    print()
fib(1000)</pre>
```

```
In [ ]: def ask_ok(prompt, retries=4, reminder='Please try again!'):
    while True:
        ok = input(prompt)
        if ok in ('y', 'ye', 'yes'):
            return 'YES'
        if ok in ('n', 'no', 'nop', 'nope'):
            return 'NO'
        retries = retries - 1
        if retries < 0:
            return 'invalid user response'
        print(reminder)

ask_ok('Do you really want to quit?: ') # giving only the mandatory argument</pre>
```

```
In [ ]: ask_ok('OK to overwrite the file?: ', 2, 'Come on, only yes or no!') # giving
    all arguments
```

Mathematics and Statistics packages

basic arithmetics

print("\nsum, 1+2+3+4+5:", np.sum([1,2,3,4,5]))

print("power, 2^8:", np.power(2, 8))

In research we will inevitably have to do some statistics. It is good to know how we can do mathematics and statistics operation with Python. As we are using Anaconda which already installed most required packages, so you do not need to install additional package here.

• The packages we will learn today are numpy, pandas and matplotlib.

NumPy

NumPy is a python library to help dealing with mathematics especially array. As it can do so many things, the examples listed below may not be able to cover all. For more information, you can visit NumPy's official tutorial (https://docs.scipy.org/doc/numpy-1.15.0/user/quickstart.html).

```
In [ ]: import numpy as np
        # create array of range [0, 15)
         a = np.arange(15)
         print("basic array:", a)
         # reshape array
         print("\nreshaped:", a.reshape([3, 5]))
         # or do it at once
         print("\nall at once:", np.arange(15).reshape([3, 5]))
In [ ]: | # create array from list
         mylist = [1,2,3,4]
         print("array from list:", np.array(mylist))
         # create 2d array
         print("\n2d array:", np.array([mylist, mylist]))
         print("\ncheck dimension:", np.array([mylist, mylist]).shape, "<- 2 rows 4 col</pre>
         umns")
In [ ]: | # create array of ones
        print("array of ones:", np.ones([3,3]))
         # create array of zeros
         print("\narray of zeros:", np.zeros([5,5]))
         # some constants
         print("\npi:", np.pi)
         print("e:", np.e)
```

Pandas

pandas is an easy-to-use python library for data structures and data analysis. pandas implements a number of statistical functions that can be used in research. You can learn more about pandas in these pandas's official quick start tutorials:

- 10 mins tutorials (https://pandas.pydata.org/pandas-docs/stable/10min.html#min)
- pandas tutorials (https://pandas.pydata.org/pandas-docs/stable/tutorials.html)

```
In [ ]: import pandas as pd
        #create dataframe
        df for scatter = pd.DataFrame(np.random.rand(5,2), columns=['col1','col2'])
        df for bar = pd.DataFrame(np.random.rand(5, 1), columns=['col1'])
        df_for_line = pd.DataFrame(np.random.rand(100, 1), columns=['col1'])
        print("scatter dataframe:\n",df for scatter,"\n")
        print("bar dataframe:\n",df_for_bar,"\n")
        print("line dataframe:\n",df_for_line,"\n")
In [ ]: # we can also see dataframe rendered as HTML
        df_for_scatter
In [ ]: | # create new dataframe by concatenate existing ones
        new_df_concat_down = pd.concat([df_for_scatter[3:], df_for_scatter[3:]]) # use
        [3:1 to shorten
        # notice index start from 3 instead of 0
        new_df_concat_down
In [ ]: new_df_concat_side = pd.concat([df_for_scatter, df_for_scatter], axis=1)
        new_df_concat_side
```

Matplotlib

Matplotlib is a popular Python 2D plotting library. You can do most of data visualization using Matplotlib. pandas plotting methods also use Matplotlib as a based tool.

- Pyplot tutorial by matplotlib.org (https://matplotlib.org/tutorials/introductory/pyplot.html)
- Matplotlib tutorial by Nicolas P. Rougier (https://www.labri.fr/perso/nrougier/teaching/matplotlib/)

```
In []: import matplotlib.pyplot as plt
%matplotlib inline

plt.plot(df_for_line)
plt.title('Line plot')
plt.figure()

plt.scatter(df_for_scatter.col1, df_for_scatter.col2)
plt.title('Scatter plot')
plt.figure()

plt.bar(df_for_bar.index, df_for_bar.col1)
plt.title('Bar plot')
plt.figure()

from skimage import data

plt.imshow(data.chelsea())
```

Simple Linear Regression: Basic of Machine Learning and Statistics

y = ax + b

A linear function has one independent variable and one dependent variable. The independent variable is x and the dependent variable is y.

- a is the constant term or the y intercept. It is the value of the dependent variable when x = 0.
- b is the coefficient of the independent variable. It is also known as the slope and gives the rate of change of the dependent variable.

"The scenario is you are a HR officer, you got a candidate with 5 years of experience. Then what is the best salary you should offer to him?"

source: <u>towardsdatascience (https://towardsdatascience.com/machine-learning-simple-linear-regression-with-python-f04ecfdadc13)</u>

```
In [ ]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        from scipy import stats
        from sklearn.metrics import r2 score
        import seaborn as sns
        color = sns.color palette()
        sns.set()
        sns.set(color_codes=True)
        # Importing the dataset
        df = pd.read_csv('./datasets/salary_data.csv')
        x = df['YearsExperience']
        y = df['Salary']*0.45
        df
In [ ]: plt.xlabel('YearsExperience (yr)')
        plt.ylabel('Salary (Baht)')
        plt.title('Salary Data on Scatter Plot')
        plt.scatter(x,y)
        plt.show()
In [ ]: | stdy = y.std()
        meany = y.mean()
        plt.hist(y)
        plt.xlabel('Salary (Baht)')
        plt.ylabel('Quantity')
        plt.title('Salary Distribution Histogram')
        print ("The std of y is: " + str(stdy) + " The mean of y is: " + str(meany))
In [ ]: stdx = x.std()
        meanx = x.mean()
        plt.hist(x)
        plt.xlabel('YearsExperience (yr)')
        plt.ylabel('Quantity')
        plt.title('YearsExperience Distribution Histogram')
        print ("The std of x is: " + str(stdx) + " The mean of x is: " + str(meanx))
```

Parametric and Nonparametric Statistics

Parametric tests

assume underlying statistical distributions in the data. Therefore, several conditions of validity must be met so that the result of a parametric test is reliable. For example, Student's t-test for two independent samples is reliable only if each sample follows a normal distribution and if sample variances are homogeneous.

The advantage of using a parametric test instead of a nonparametric equivalent is that the former will have more statistical power than the latter. In other words, a parametric test is more able to lead to a rejection of H0. Most of the time, the p-value associated to a parametric test will be lower than the p-value associated to a nonparametric equivalent that is run on the same data.



Nonparametric tests

do not rely on any distribution. They can thus be applied even if parametric conditions of validity are not met. Nonparametric tests are more robust than parametric tests. In other words, they are valid in a broader range of situations (fewer conditions of validity).

Machine Learning

- numpy.poly1d : one-dimensional polynomial class
- numpy.polyfit: Least squares polynomial fit

```
In [ ]:    p = np.poly1d(np.polyfit(x,y,1))
    p
In [ ]:    plt.xlabel('YearsExperience (yr)')
    plt.ylabel('Salary (Baht)')
    plt.title('Salary Data on Scatter Plot - Linear Regression')

    xp = np.linspace(0, 11, 100)
    plt.plot(xp,p(xp),c = 'r')
    plt.scatter(x,y)
    plt.show()

In [ ]:    r2_train = r2_score(y, p(x))
    print('The rsquared value is: ' + str(r2_train))
```

scipy.stats.linregress: Calculate a linear least-squares regression for two sets of measurements.

```
In [ ]: slope, intercept, r_value, p_value, std_err = stats.linregress(x,y)
intercept,slope
```

```
In [ ]: def linefitline(b):
            return intercept + slope * b
        line = linefitline(x)
        plt.scatter(x,y)
        plt.xlabel('YearsExperience (yr)')
        plt.ylabel('Salary (Baht)')
        plt.title('Salary Data on Scatter Plot - Linear Regression')
        plt.plot(x,line, c = 'g')
        plt.show()
In [ ]: | r2_lin = r_value * r_value
        print('The rsquared value is: ' + str(r2_lin))
In [ ]: | shuffled = df.sample(frac=1).reset_index(drop=True)
        shuffx = shuffled['YearsExperience']
        shuffy = shuffled['Salary']
        trainX = shuffx[:20]
        testX = shuffx[20:]
        trainY = shuffy[:20]
        testY = shuffy[20:]
        plt.scatter(trainX, trainY)
        plt.xlabel('YearsExperience (yr)')
        plt.ylabel('Salary (Baht)')
        plt.title('Training Data - Salary Data on Scatter Plot')
        plt.show()
        plt.scatter(testX, testY)
        plt.xlabel('YearsExperience (yr)')
        plt.ylabel('Salary (Baht)')
        plt.title('Testing Data - Salary Data on Scatter Plot')
        plt.show()
In [ ]: | p1 = np.poly1d(np.polyfit(trainX, trainY, 1))
        xp = np.linspace(0, 10, 100)
        axes = plt.axes()
        axes.set_xlim([0,10])
        plt.scatter(trainX, trainY)
        plt.xlabel('Age (yr)')
        plt.ylabel('Length (cm)')
        plt.title('Training Linear Regression')
        plt.plot(xp, p1(xp), c='r')
        plt.show()
In [ ]: xp = np.linspace(0, 10, 100)
        axes = plt.axes()
        plt.scatter(testX, testY)
        plt.xlabel('Age (yr)')
        plt.ylabel('Length (cm)')
        plt.title('Testing Linear Regression')
        plt.plot(xp, p1(xp), c='r')
        plt.show()
```

FAQs

- · How should I search when I am stuck?
 - You can use google with specific keywords. For example, if you are doing some table processing using pandas, you should include pandas in keyword instead of python.
- What sources from search results should I set higher priority to check out?
 - This can be various, but first top focus should include StackOverflow, GitHub, and official sites of Python or packages in question.

Mini challenge

• Click here to open mini challenge notebook. (MiniChallenge.ipynb)

Assignment

Click here to open assignment notebook. (Assignment.ipynb)

Learning Materials

- Official Python tutorial (https://docs.python.org/3/tutorial/)
- Chula's Python 101 (https://www.cp.eng.chula.ac.th/books/python101/)