# Python R training course - Session 4

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### August 27, 2017

#### Lecture list

- 1. Introduction to Python
- 2. The Python Standard Library, if/loops statements
- 3. Vector/matrix structures, numpy library
- 4. Python data types, File Processing, Pandas library
  - Structure of Python files. Read/create a file.
  - Read csv file using Pandas library
- 5. Functions in Python, Debugging.
- 6. Introduction to R, R for Python programmers.
- 7. Import data, plot data.
- 8. Data Mining in Python/R.

## 1 File Processing

Python provides basic functions and methods necessary to manipulate files by default. You can do most of the file manipulation using a file object.

#### 1.1 Opening and Closing Files

Before you can read or write a file, you have to open it using Python's built-in **open()** function. This function creates a file object, which would be utilized to call other support methods associated with it.

- f = open(file\_name [, access\_mode][, buffering])
  - **file\_name**: The file\_name argument is a string value that contains the name of the file that you want to access.
  - access\_mode: The access\_mode determines the mode in which the file has to be opened, i.e., read, write, append, etc. A complete list of possible values is given below in the table. This is optional parameter and the default file access mode is read (r).

• **buffering**: If the buffering value is set to 0, no buffering takes place. If the buffering value is 1, line buffering is performed while accessing a file. If you specify the buffering value as an integer greater than 1, then buffering action is performed with the indicated buffer size. If negative, the buffer size is the system default(default behavior).

Here is a list of the different modes of opening a file:

### Mode Sescription

- r Opens a file for reading only. The file pointer is placed at the beginning of the file. This is the default mode.
- rb Opens a file for reading only in binary format. The file pointer is placed at the beginning of the file. This is the default mode.
- r+ Opens a file for both reading and writing. The file pointer placed at the beginning of the file.
- rb+ Opens a file for both reading and writing in binary format. The file pointer placed at the beginning of the file.
- w Opens a file for writing only. Overwrites the file if the file exists. If the file does not exist, creates a new file for writing.
- wb Opens a file for writing only in binary format. Overwrites the file if the file exists. If the file does not exist, creates a new file for writing.
- w+ Opens a file for both writing and reading. Overwrites the existing file if the file exists. If the file does not exist, creates a new file for reading and writing.
- wb+ Opens a file for both writing and reading in binary format. Overwrites the existing file if the file exists. If the file does not exist, creates a new file for reading and writing.
- a Opens a file for appending. The file pointer is at the end of the file if the file exists. That is, the file is in the append mode. If the file does not exist, it creates a new file for writing.
- ab Opens a file for appending in binary format. The file pointer is at the end of the file if the file exists. That is, the file is in the append mode. If the file does not exist, it creates a new file for writing.
- a+ Opens a file for both appending and reading. The file pointer is at the end of the file if the file exists. The file opens in the append mode. If the file does not exist, it creates a new file for reading and writing.
- ab+ Opens a file for both appending and reading in binary format. The file pointer is at the end of the file if the file exists. The file opens in the append mode. If the file does not exist, it creates a new file for reading and writing.

Once a file is opened and you have one file object, you can get various information related to that file.

Here is a list of all attributes related to file object:

Attribute	Description
file.closed	Returns true if file is closed, false otherwise.
file.mode	Returns access mode with which file was opened.
file.name	Returns name of the file.
file.softspace	Returns false if space explicitly required with print, true otherwise.

The **close()** method of a file object flushes any unwritten information and closes the file object, after which no more writing can be done.

Python automatically closes a file when the reference object of a file is reassigned to another file. It is a good practice to use the close() method to close a file.

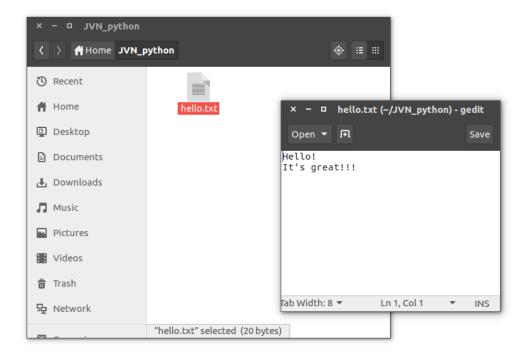
Syntax:

#### 1.2 Reading and Writing Files

The file object provides a set of access methods to make our lives easier. We would see how to use **read()** and **write()** methods to read and write files.

The **write()** method writes any string to an open file. It is important to note that Python strings can have binary data and not just text.

Example:



The **read()** method reads a string from an open file. It is important to note that Python strings can have binary data. apart from text data.

**Syntax** 

```
fileObject.read([count]);
```

Here, passed parameter is the number of bytes to be read from the opened file. This method starts reading from the beginning of the file and if count is missing, then it tries to read as much as possible, maybe until the end of file.

Example

Let's take a file hello.txt, which we created above.

#### 1.3 Renaming and Deleting Files

Python **os** module provides methods that help you perform file-processing operations, such as renaming and deleting files.

To use this module you need to import it first and then you can call any related functions.

#### 1.3.1 The rename() Method

The **rename()** method takes two arguments, the current filename and the new filename:

#### 1.3.2 The remove() Method

You can use the **remove()** method to delete files by supplying the name of the file to be deleted as the argument.

```
Syntax
```

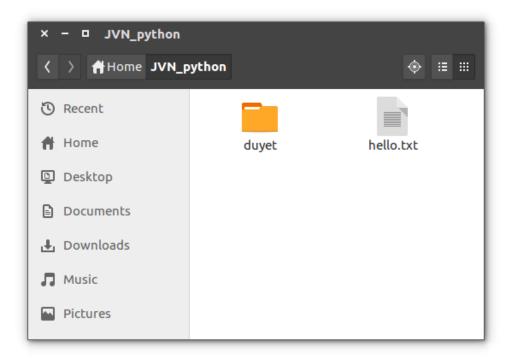
### 1.4 Directories in Python

The **os** module has several methods that help you create, remove, and change directories.

### 1.4.1 The mkdir() Method

You can use the mkdir() method of the os module to create directories in the current directory

```
os.mkdir("newdir")
In [39]: os.mkdir("duyet")
```



#### 1.4.2 The chdir() Method

You can use the **chdir()** method to change the current directory.

```
os.chdir("newdir")
```

### 1.4.3 The getcwd() Method

The **getcwd()** method displays the current working directory.

```
In [41]: print os.getcwd()
/home/duyetdev/project/Python-R-Basic
```

## 1.4.4 The rmdir() Method

The rmdir() method deletes the directory, which is passed as an argument in the method.

```
os.rmdir('dirname')
In [50]: os.rmdir("./duyet")
In []:
In []:
```

Ex: Creates the following folder structure

#### 2 Pandas

Pandas is a package of fast, efficient data analysis tools for Python

Just as NumPy provides the basic array data type plus core array operations, pandas

- defines fundamental structures for working with data and
- endows them with methods that facilitate operations such as
  - reading in data
  - adjusting indices
  - working with dates and time series
  - sorting, grouping, re-ordering and general data munging [1]
  - dealing with missing values, etc., etc.

More sophisticated statistical functionality is left to other packages, such as statsmodels and scikit-learn, which are built on top of pandas

```
In [52]: import numpy as np
     import pandas as pd
```

#### 2.1 Series

Two important data types defined by pandas are Series and DataFrame

You can think of a Series as a "column" of data, such as a collection of observations on a single variable

A DataFrame is an object for storing related columns of data Let's start with Series

Here you can imagine the indices 0, 1, 2, 3 as **indexing** four listed companies, and the values being daily returns on their shares.

Pandas Series are built on top of NumPy arrays, and support many similar operations

```
Out [55]: 0 0.109632
             0.125849
        1
        2
             1.593836
         3
              1.826518
        Name: daily returns, dtype: float64
In [57]: s[s >= 1]
Out[57]: 2 1.593836
        Name: daily returns, dtype: float64
In [58]: s.describe()
                  4.000000
Out[58]: count
                  0.000700
        mean
         std
                1.402894
        min
               -1.826518
        25%
               -0.374405
        50%
                  0.117741
        75%
                  0.492846
                  1.593836
        max
        Name: daily returns, dtype: float64
  But their indices are more flexible
In [59]: s.index = ['AMZN', 'AAPL', 'MSFT', 'GOOG']
Out [59]: AMZN 0.109632
               0.125849
        AAPL
        MSFT
               1.593836
         GOOG -1.826518
        Name: daily returns, dtype: float64
In [60]: s['GOOG']
Out[60]: -1.8265181959162498
```

#### 2.2 DataFrames

While a Series is a single column of data, a DataFrame is several columns, one for each variable. Here's the contents of *test\_pwt.csv* 

```
"country", "country isocode", "year", "POP", "XRAT", "tcgdp", "cc", "cg"

"Argentina", "ARG", "2000", "37335.653", "0.9995", "295072.21869", "75.716805379", "5.5788042896"

"Australia", "AUS", "2000", "19053.186", "1.72483", "541804.6521", "67.759025993", "6.7200975332"

"India", "IND", "2000", "1006300.297", "44.9416", "1728144.3748", "64.575551328", "14.072205773"

"Israel", "ISR", "2000", "6114.57", "4.07733", "129253.89423", "64.436450847", "10.266688415"

"Malawi", "MWI", "2000", "11801.505", "59.543808333", "5026.2217836", "74.707624181", "11.658954494"

"South Africa", "ZAF", "2000", "45064.098", "6.93983", "227242.36949", "72.718710427", "5.7265463933"

"United States", "USA", "2000", "282171.957", "1", "9898700", "72.347054303", "6.0324539789"

"Uruguay", "URY", "2000", "3219.793", "12.099591667", "25255.961693", "78.978740282", "5.108067988"
```

```
In [61]: df = pd.read_csv('https://github.com/QuantEcon/QuantEcon.lectures.code/raw/master/panda
In [62]: df.head()
Out[62]:
                                                      POP
              country country isocode
                                                                XRAT
                                                                              tcgdp
                                        year
            Argentina
                                   ARG
                                        2000
                                                37335.653
                                                            0.999500
                                                                       2.950722e+05
         1
            Australia
                                   AUS
                                        2000
                                                                      5.418047e+05
                                                19053.186
                                                            1.724830
         2
                India
                                   IND
                                        2000
                                              1006300.297
                                                           44.941600
                                                                       1.728144e+06
         3
               Israel
                                   ISR
                                        2000
                                                 6114.570
                                                            4.077330
                                                                      1.292539e+05
         4
               Malawi
                                  MWI
                                        2000
                                                11801.505 59.543808 5.026222e+03
                   СС
                               cg
            75.716805
                        5.578804
            67.759026
                        6.720098
         2 64.575551 14.072206
         3 64.436451 10.266688
         4 74.707624 11.658954
In [63]: df[2:4]
Out [63]:
           country country isocode
                                                   POP
                                    year
                                                            XRAT
                                                                          tcgdp \
         2
             India
                                IND
                                     2000
                                           1006300.297 44.94160
                                                                  1.728144e+06
         3 Israel
                                ISR
                                     2000
                                              6114.570
                                                         4.07733 1.292539e+05
                   СС
                               cg
           64.575551
                      14.072206
         3 64.436451
                      10.266688
```

To select columns, we can pass a list containing the names of the desired columns represented as strings

```
In [64]: df[['country', 'tcgdp']]
```

```
Out [64]:
                  country
                                  tcgdp
         0
                Argentina 2.950722e+05
         1
                Australia 5.418047e+05
         2
                    India 1.728144e+06
                   Israel 1.292539e+05
         3
         4
                   Malawi 5.026222e+03
         5
             South Africa 2.272424e+05
         6
            United States 9.898700e+06
         7
                  Uruguay 2.525596e+04
```

To select both rows and columns using integers, the iloc attribute should be used with the format .iloc[rows,columns]

```
In [65]: df.iloc[2:5,0:4]
Out [65]:
           country country isocode
                                     year
                                                   POP
         2
             India
                                IND
                                     2000
                                           1006300.297
         3 Israel
                                ISR
                                     2000
                                              6114.570
         4 Malawi
                                MWI
                                     2000
                                             11801.505
```

To select rows and columns using a mixture of integers and labels, the loc attribute can be used in a similar way

Let's imagine that we're only interested in population and total GDP (tcgdp).

One way to strip the data frame df down to only these variables is to overwrite the dataframe using the selection method described above

```
In [71]: df = df[['country', 'POP', 'tcgdp']]
         df
Out [71]:
                                   POP
                  country
                                               tcgdp
         0
                Argentina
                             37335.653 2.950722e+05
         1
                Australia
                             19053.186 5.418047e+05
         2
                    India 1006300.297 1.728144e+06
         3
                   Israel
                              6114.570 1.292539e+05
         4
                   Malawi
                             11801.505 5.026222e+03
         5
             South Africa
                            45064.098 2.272424e+05
         6
           United States
                            282171.957 9.898700e+06
         7
                              3219.793 2.525596e+04
                  Uruguay
```

Here the index 0, 1,..., 7 is redundant, because we can use the country names as an index To do this, we set the index to be the country variable in the dataframe

```
In [72]: df = df.set_index('country')
         df
Out [72]:
                                POP
                                            tcgdp
         country
         Argentina
                          37335.653 2.950722e+05
         Australia
                          19053.186 5.418047e+05
         India
                        1006300.297 1.728144e+06
         Israel
                           6114.570 1.292539e+05
         Malawi
                          11801.505 5.026222e+03
         South Africa
                         45064.098 2.272424e+05
                         282171.957 9.898700e+06
         United States
         Uruguay
                           3219.793 2.525596e+04
In [73]: df.columns = 'population', 'total GDP'
         df
Out [73]:
                         population
                                        total GDP
         country
```

Argentina	37335.653	2.950722e+05
Australia	19053.186	5.418047e+05
India	1006300.297	1.728144e+06
Israel	6114.570	1.292539e+05
Malawi	11801.505	5.026222e+03
South Africa	45064.098	2.272424e+05
United States	282171.957	9.898700e+06
Uruguay	3219.793	2.525596e+04

Population is in thousands, let's revert to single units

```
In [74]: df['population'] = df['population'] * 1e3
Out [74]:
                                        total GDP
                         population
         country
         Argentina
                       3.733565e+07 2.950722e+05
                       1.905319e+07 5.418047e+05
         Australia
         India
                       1.006300e+09 1.728144e+06
         Israel
                       6.114570e+06 1.292539e+05
        Malawi
                       1.180150e+07 5.026222e+03
         South Africa 4.506410e+07 2.272424e+05
        United States 2.821720e+08 9.898700e+06
         Uruguay
                       3.219793e+06 2.525596e+04
```

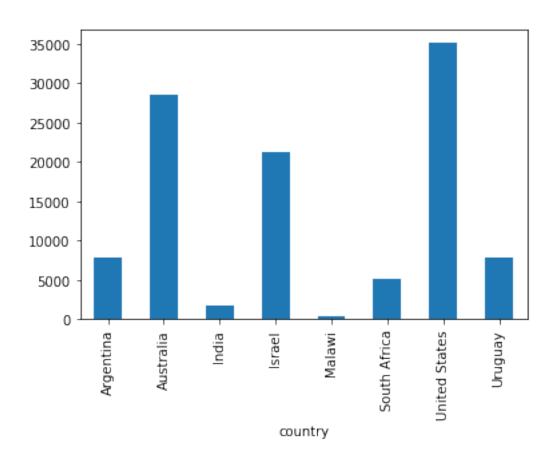
Next we're going to add a column showing real GDP per capita, multiplying by 1,000,000 as we go because total GDP is in millions

```
In [75]: df['GDP percap'] = df['total GDP'] * 1e6 / df['population']
        df
Out [75]:
                         population
                                        total GDP
                                                     GDP percap
        country
        Argentina
                       3.733565e+07 2.950722e+05
                                                    7903.229085
        Australia
                       1.905319e+07 5.418047e+05 28436.433261
        India
                       1.006300e+09 1.728144e+06
                                                   1717.324719
        Israel
                       6.114570e+06 1.292539e+05 21138.672749
        Malawi
                       1.180150e+07 5.026222e+03
                                                     425.896679
        South Africa
                       4.506410e+07 2.272424e+05
                                                    5042.647686
        United States 2.821720e+08 9.898700e+06 35080.381854
                       3.219793e+06 2.525596e+04
                                                    7843.970620
        Uruguay
```

One of the nice things about pandas DataFrame and Series objects is that they have methods for plotting and visualization that work through Matplotlib

```
In [77]: df['GDP percap'].plot(kind='bar')
Out[77]: <matplotlib.axes._subplots.AxesSubplot at 0x7f230ff42650>
```

In [78]: import matplotlib.pyplot as plt
 plt.show()



### 3 References

- 10 Minutes to pandas (https://pandas.pydata.org/pandas-docs/stable/10min.html)[https://pandas.pydata.org/pandas-docs/stable/10min.html]
- Pandas Tutorial: DataFrames in Python https://www.datacamp.com/community/tutorials/pandas-tutorial-dataframe-python
- https://www.tutorialspoint.com/python/python\_files\_io.htm

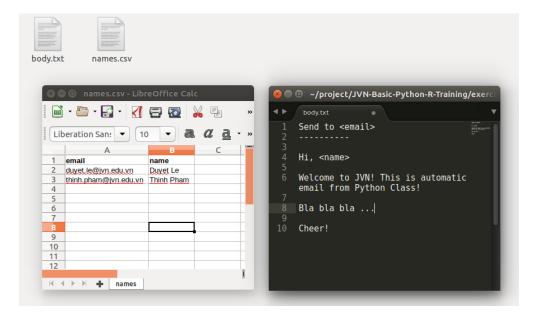
### 4 Excercises

### **Python Program to Merge Mails**

When we want to send the same invitations to many people, the body of the mail does not change. Only the name (and maybe address) needs to be changed.

Mail merge is a process of doing this. Instead of writing each mail separately, we have a template for body of the mail and a list of names that we merge together to form all the mails.

For this program, we have written all the names and emails in separate lines in the file "names.csv". The body is in the "body.txt" file.



figs/session\_4/merge\_mails.png

We open both the files in reading mode and iterate over each name. A new file with the name "[name].txt" is created, where name is the name of that person. Replace the and in body email.

### In []: