Computer Simulations and Risk Assessment – Lecture 2

Fall 2019

Brandeis International Business School



Course Information - Schedule

Class Date	Text Chapters
Aug. 30, 2019 – L1	 Course Introduction/Python Installation Introduction to Quantitative Finance Career Python basics
Sep. 6, 2019 – L2	Advanced Python Topics
Sep. 13, 2019 – L3	Advanced Python Topics
Sep. 20, 2019 – L4	Sourcing and handling DataStylized financial data analysis using Python
Sep. 27, 2019 – L5	Value at Risk
Oct. 4, 2019 – L6	 Conditional Value at Risk (Expected Shortfall) + Mid-term Review
Oct. 11, 2019	Mid-term
Oct. 18, 2019 – L7	Modeling Volatility I
Oct. 25, 2019 – L8	Modeling Volatility II
Nov. 1, 2019 – L9	Practical application case Studies I
Nov. 8, 2019 – L10	Practical application case Studies II
Nov. 15, 2019 – L11	Back Testing + Conditional risk prediction
Nov. 22, 2019 – L12	Research project presentation
Dec. 6, 2019 – L13	Final Review



FIN 285A: Simulations & Risk, Fall 2019 By Steve Xia Introductio n to Python Programmi ng

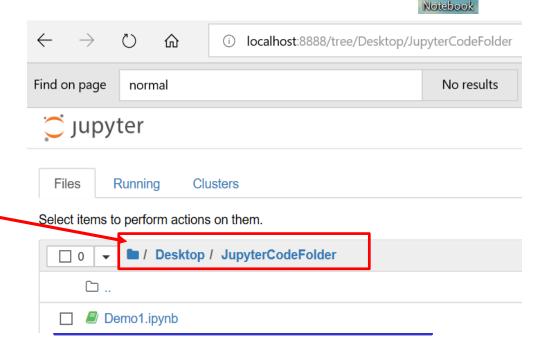
- Jupyter notebook introduction
- Packages
- Functions and Methods
- Matrix operation and manipulation
- Control structure
- Data I/Os
- Data types and structures
- Charting





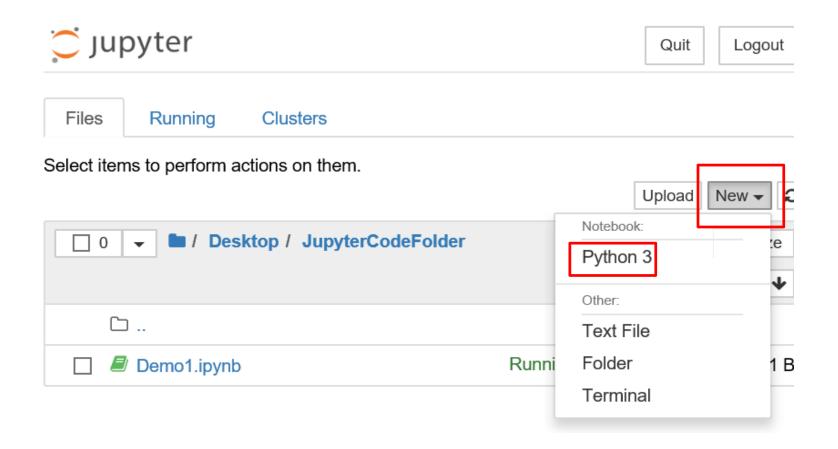
- **Jupyter Notebook** is an open-source web application that allows you to create and share documents that contain live code, equations, visualizations and narrative text
- For submitting homework, please use Jupyter Notebook to organize your code for each problem into cells and submit one notebook for each homework assignment (The TAs will help you on this)
- To begin Jupyter notework, double click the icon

• Then go to the folder where you want to have the code saved



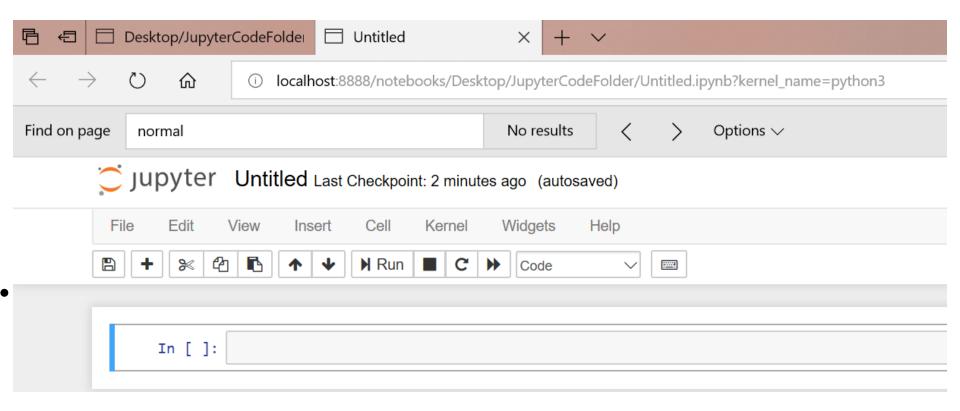
Jupyter

• Then click New, select Python 3





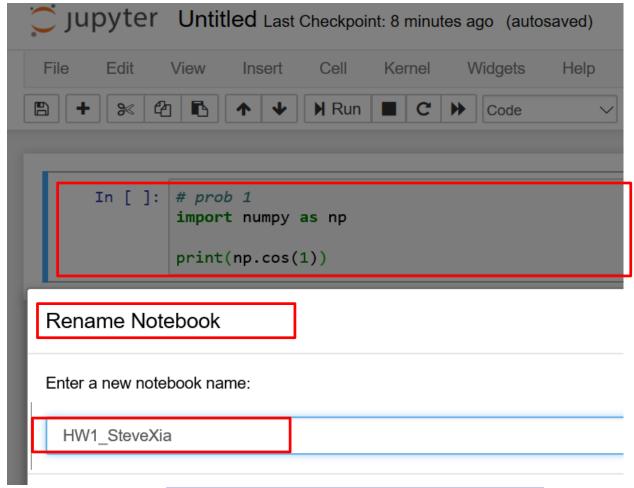
• A new window with name Untitled will pop up







• Copy the first portion of the code (e.g., hw1, problem 1) from your Spyder editor here as the first cell, then rename it HW1_YourName



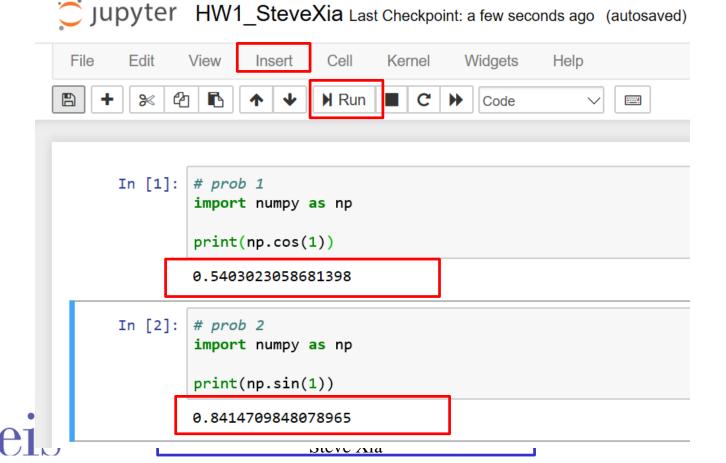


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• Insert a cell below and copy the second portion of the code (e.g., hw1 problem 2) from your Spyder here as the second cell, then repeat the process until you are done with copying all of your homework solutions here. Click run to show results for code in each

cell



• Click save and you should find the file HW1_SteveXia.ipynb at the folder where you started the jupyter session. Submit this file to the

TA ご Jupyter HW1_SteveXia Last Check Edit File View Cell Kernel Insert **N** Run In [1]: # prob 1 import numpy as np print(np.cos(1)) 0.5403023058681398 JupyterCodeFolder Name access .ipynb checkpoints top Demo1.ipynb iloads HW1_SteveXia.ipynb ments FIN 285A: Simulations & Risk, Fall 2019 By Steve Xia

Python Packages

- Python's power comes from a lot of the 'specialized' packages you can download to work with the main program
 - Numpy for matrix and numerical operations
 - Pandas for data analysis
 - SciPy for scientific calculations such as integration
 - o scikit-learn—for machine learning (sponsored by Google)
- A package is just a directory containing
 - files with Python code called modules in Python speak
 - o possibly some compiled code that can be accessed by Python (e.g., functions compiled from C or FORTRAN code)
 - a file called __init__.py that specifies what will be executed when we type import package_name (e.g., import numpy)
- use the import command to load the package into Python
 - o import numpy as np # Load the numpy package and name it np
 - \circ b = np.cos(a) # calling the cosine function contained in the numpy (np) package

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Python Packages

- To find what packages you have installed with Python:
 - Type in help("modules") in your Spyder console In [1]: help("modules")

```
cProfile
Crypto
                                          mistune
                                                               spyder io dcm
Cvthon
                     cachecontrol
                                          mk1
                                                               spyder io hdf5
                                                               spyder profiler
                    calendar
IPython
                                          mmap
OleFileIO PL
                    certifi
                                          mmapfile
                                                               spyder pylint
OpenSSL
                    cffi
                                                               sqlalchemy
                                          mmsystem
PIL
                                          modulefinder
                    cgi
                                                               sglite3
PyQt5
                    cgitb
                                          mpmath
                                                               sre compile
future
                    chardet
                                          msgpack
                                                               sre constants
ast
                    chunk
                                          msilib
                                                               sre parse
                    click
                                                               ss1
asyncio
                                          msvcrt
                                          multipledispatch
bisect
                    cloudpickle
                                                               sspi
blake2
                                          multiprocessing
                    clyent
                                                               sspicon
bootlocale
                    cmath
                                          navigator updater
                                                               stat
bz2
                    cmd
                                          nbconvert
                                                               statistics
cffi backend
                    code
                                          nbformat
                                                               statsmodels
codecs
                    codecs
                                          netbios
                                                               storemagic
codecs cn
                    codeop
                                          netro
                                                               string
codecs hk
                    collections
                                                               stringprep
                                          networkx
codecs iso2022
                    colorama
                                          nltk
                                                               struct
codecs jp
                    colorsys
                                          nntplib
                                                               subprocess
codecs kr
                    commctrl
                                          nose
                                                               sunau
                    compileall
codecs tw
                                          notebook
                                                               symbol
collections
                    comtypes
                                          nt
                                                               sympy
collections abc
                    concurrent
                                          ntpath
                                                               sympyprinting
compat pickle
                    conda
                                          ntsecuritycon
                                                               symtable
                                          nturl2path
compression
                    conda build
                                                               sys
                    conda env
                                          numba
                                                               sysconfig
CSV
                    conda verify
                                          numbers
                                                               tables
ctypes
_ctypes_test
                    configparser
                                          numexpr
                                                               tabnanny
datetime
                    contextlib
                                                               tarfile
                                          numpy
```

you can find and explore the directory for NumPy on your computer

easily - C:\ProgramData\Anaconda3\pkgs\numpy
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Python Packages - SciPy

- The <u>SciPy</u> library is built on top of NumPy and provides additional functionality
- SciPy includes many of the standard routines used in
 - linear algebra
 - integration
 - interpolation
 - optimization
 - distributions and random number generation
 - signal processing
- For example, let's calculate $\int_{-2}^{2} \phi(z) dz$ where ϕ is the standard normal density:

from scipy.stats import norm from scipy.integrate import quad

```
fai = norm()
value, error = quad(fai.pdf, -2, 2) # Integrate using Gaussian quadrature
```



Find what version of Packages you have & update

- import scipy
- scipy.__version__ Out[30]: '1.1.0'
- To update your packages:
 - o In Anaconda prompt, type in: "conda update scipy"

```
Anaconda Prompt

(base) C:\Users\flyin>conda update scipy
```



Python Packages - NumPy

- NumPy is the fundamental package for scientific computing with Python.
- It contains among other things:
 - a powerful N-dimensional array object
 - sophisticated (broadcasting) functions
 - tools for integrating C/C++ and Fortran code
 - useful linear algebra, Fourier transform, and random number capabilities
- For example, create simple matrix using Numpy:

```
import numpy as np
# create a 3x3 matrix
X = np.matrix([[1, 2, 3], [4, 5, 6], [7, 8, 9]])

# use built in functions to initiate matrices
# create a matrix of zeros with three rows and two columns
Y = np.zeros((3,2))
# create a matrix of ones with three rows and two columns
Y1 = np.ones((3,2))
```



A few basics

- Python is a case-sensitive language
- How to add comment lines: use the # symbol
- How to comment out a whole block of code: begin with """, and end the block with """
- Line break: Simply type enter to break a long line into multiple lines
- del X, Y clear variable X and Y from the computer memory
- Click the 'x' button under the variable explorer window in Spyder to clear all the variables in memory
- Type in %clear to clear the Ipython console or 'control L' from the keyboard



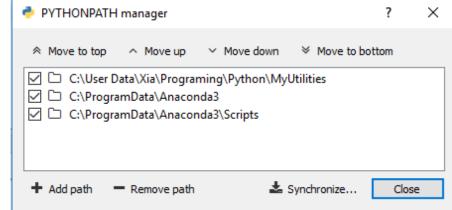
Setup: Defining file/function search path



Where does Python search for Python script files

or more importantly, Python functions?

- It always start with the current working folder (how do you find out what is the current working folder?)
- Then the path defined in the path setup
- > Functions in packages loaded in using the import call is automatically available



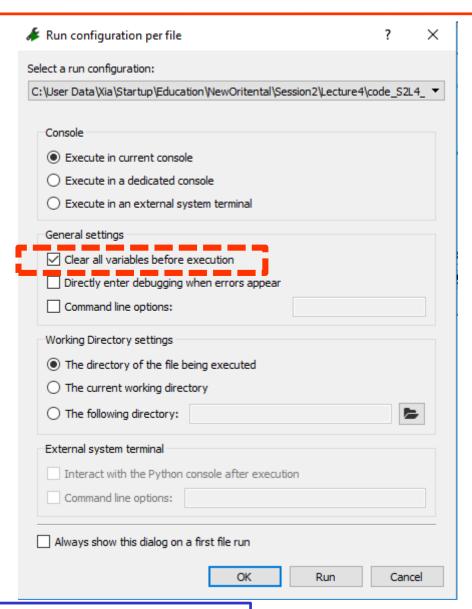
- The way to change the path file:
 - 1. In Spyder, click the Python path manager icon and add the folder you want to the list of folders
 - 2. In a python script: import sys sys.path.append("C:/User Data/Xia/Programing/Python/MyUtilities")



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Setup: Clear all variables before running a python script

- You might want to automatically clear all the variables in Python before you run a script.
- Easiest way to do this is to change the Spyder setting
- going to the Spyder menu >
 Run > Configuration per file
 > General settings and
 selecting the option called
 Clear all variables before
 execution





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Python Basics – function call and object method

- Python's powerful in that there are a lot of computing algorithms are built-in through available functions and methods associated with a class coming with the software
- Basic form of a function call:
 - \circ A, B = function function A, B = function (X,Y)
 - o A and B are the output variables; X and Y are the input variables
- Basic form a method call associated with an class/object

```
>> X = np.array([1, 2, 3])
```

>> X.mean



Example: Creating a function and calling a function

Creating a function and calling a function

```
def addfunc(x,y):
    return x+y

def addsubfunc(x,y):
    return x+y, x-y

a=addfunc(1,2)
c,d = addsubfunc(1,2)
```

```
IPython console

☐ Console 1/A 

In [55]: addfunc(1,2)
Out[55]: 3

In [56]: addsubfunc(1,2)
Out[56]: (3, -1)

In [57]: |
```

- One thing to remember:
 - Location of the function, need to either at the top of the script, or, use the import command to import from packages or saved python files where the functions are defined



Example: Creating a class and calling methods associated with the class

- Creating a class called **Point** and define a method **print** associated with it
- Then create an object belongs to the class Point and calls the method print

```
# defining a class called Point, with two methods, assign and print
class Point:
    def assign(self, x, y, z):
        self.x = x
        self.y = y
        self.z = z

    def print(self):
        print(self.x, self.y, self.z)

# define a point object
P1 = Point()
P1.assign(1,2,3)
P1.print()
```

- One thing to remember:
 - Location of the class definition, need to either at the top of the script, or,
 - use the import command to import from packages or saved python files where the classes are defined



Python: Lambda Functions

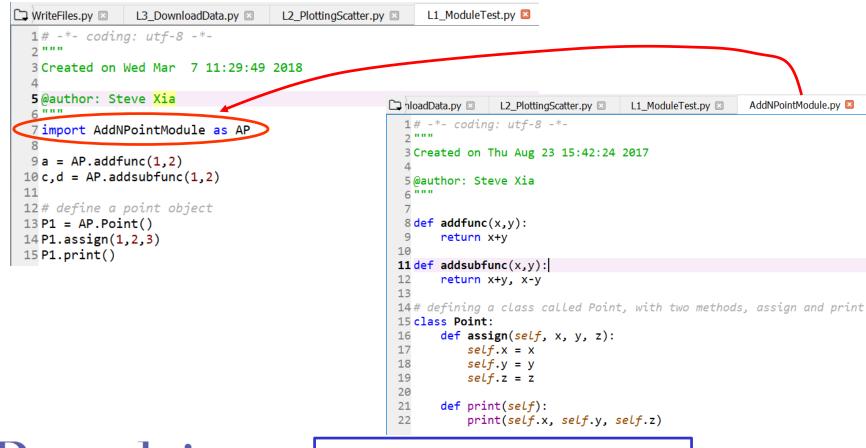
- Python supports the creation of anonymous functions (i.e. functions that are not bound to a name) at runtime, using a construct called "lambda":
- Here the function created by lambda is said to be anonymous, because it was never given a name. This is a very nice feature for applications where we don't want to create a separate named stand-alone function
- The benefit is that you have the flexibility to change the content of the function each time you use it.
- a) A code shows the difference between a normal function definition (f) and a Lambda function definition, and b) how it is used in our optimization code is shown below

quick and dirty way

```
>>> def f (x): return x**2
...
>>> print f(8)
64
>>>
>>> g = lambda x: x**2
>>>
>>> print g(8)
64
Steve Xia
```

Modules: importing user created functions

- A module is a file containing Python definitions and statements. The file name is the module name with the suffix .py appended
- Modules are a good way to organize user defined functions



Matrix Manipulation - basics

- Use the numpy package to help construct and manipulate matrices
 - o import numpy as np
- Use ',' to add new columns or new row elements.
 - o X=np.matrix([[1, 2, 3], [4, 5, 6], [7, 8, 9]])
- Or use sequencing: np.linspace(2, 4, 5) to create an 1d array
- Matrix indexing starts at zero, not one
 - X[0,2] the first row, third column element of matrix X
 - X[0,:] the first row of the matrix X
 - X[:,1] the second column of the matrix X
 - b1=X[1:3,0:2]— the 2nd and 3rd rows and 1^{1st} and 2nd column of the matrix X
 - Note: the python indexing convention is that 0:N means index of 1,2,...N-1(not N)
 - X[0,-1] first row, last column element of X. Index -1 can be used to reference the last element of an array



Matrix Initiation using built in functions

- Create a matrix of zeros with three rows and two columns
 - $\circ \quad Y = np.zeros((3,2))$
- create a matrix of ones with three rows and two columns
 - \circ Y1 = np.ones((3,2))
- create a matrix of nan with one rows and three columns
 - \circ Y3 = np.full((1,3), np.nan)



Matrix Manipulation - more

• How to append a matrix to the right, or below another matrix

>>
$$X = np.matrix([[1, 2, 3], [4, 5, 6], [7, 8, 9]])$$

>> $Z = np.ones((1,3))$
 $X = 1$ 2 3 $Z = 1$ 1 1
4 5 6
7 8 9

$$>>$$
Z1 = **np.vstack**((X,Z))

$$Z1 = 1$$
 2 3 1 4 5 6 1 7 8 9 1

 How to take out a row/column of a matrix

• Take out the first row of Z1

• Delete the second column of Z1



Matrix Operation & Boolean Indexing

- The algebraic operators +, -, *, / and ** all act element-wise on arrays
 - \circ a = np.array([1, 2, 3, 4]) b = np.array([5, 6, 7, 8])
 - \circ a + b -> array([6, 8, 10, 12])
- Sample matrix multiplication
 - \circ A = np.ones((2, 2)) B = np.ones((2, 2))
 - o A @ B -> [2., 2. 2., 2.]
 - Or np.dot(A,B) gives the same results
- Using Boolean outputs to select only a portion of a matrix
- Example 1 selecting only elements that are greater than 3
 - $\circ \quad Z2[Z2 > 3]$
- Example 2 select only the rows of y where the first element of each row is greater than 0
 - \circ # y = np.arange(35).reshape(5,7)
 - \circ # y1 = y[y[:,0]>0]



Logical and Comparison Operations

- Boolean data type True or False
- <, <=, >, >=, ==, != are operators that compare the values of two objects and returns True of False
 - >> X=2
 - >> X!= True
- 'and' for logical and; 'or' for logical or; 'not' for logical not

operators	descriptions
not x	Returns True if x is True, False otherwise
x and y	Returns x if x is False, y otherwise
x or y	Returns y if x is False, x otherwise

- '&' for bitwise and; '|' for bitwise or; '~' for bitwise not
- Membership operations
 - o 'in' evaluates to True if it finds a variable in a specified sequence and False otherwise. 'not in' does the opposite
 - $\circ >> \text{'good'}$ in 'this is a great show' \rightarrow False
 - $\circ >>$ 'good' not in 'this is a great show' \to True



Basic Control Structures

```
    For loops

    for x in range(3, 8, 2):
       print(x) # Prints out 3,5,7
       y=x+1
       print(y) # Prints out 4, 6, 8

    While loops

    # Prints out 0,1,2,3,4
    count = 0
    while count < 5:
       print(count)
      count += 1
• If statements
     var1 = 100
    if var1>100:
      print("Var Value {0:8.4f} is greater than 100".format(var1))
    else:
      print("Var Value {0:8.4f} is equal or less than 100".format(var1))
```



Debugging - Example

• The Spyder IDE has ok debugging capability

• The key basic debugging function is done by setting break points in your code so you can step through certain parts of your code to find

out what is happening

```
Editor
🗀 py 🗵
          L2_DebugDemo.py
                             L2_matrixNIndexing.py
  1# -*- codina: utf-8 -*-
  3 Created on Wed Sep 5 23:38:00 2018
  5@author: Steve Xia
  7 import numpy as np
  8 def demo(x):
        for i in range(5):
            print("i={}, x={}".format(i, x))
            x = x + 1
        return x
13 a = np.ones((3,1))
 14 a1 = 3
15 b=demo(0)
 16 c=2
```



Python – how to get help information

1. Highlight the function you need help on, then press 'control+ i' in Spyder Editor

2. Use the help() function from IPython console

```
# read in data from
df3 = pd.read_excel(
```

read_excel

Definition: read_excel(*args, **kwargs) **Type**: Function of pandas.io.excel module

Read an Excel table into a pandas DataFrame

Parameters

io : string, path object (pathlib.Path or py._path.lc file-like object, pandas ExcelFile, or xlrd worl

```
In [14]: help(pd.read_excel)
Help on function read_excel in module pandas.io.excel:
```

read_excel(io, sheet_name=0, header=0, names=None, inde squeeze=False, dtype=None, engine=None, converters=None false_values=None, skiprows=None, nrows=None, na_values date_parser=None, thousands=None, comment=None, skipfoon Read an Excel table into a pandas DataFrame

Parameters

io : string, path object (pathlib.Path or py._path. file-like object, pandas ExcelFile, or xlrd wor The string could be a URL. Valid URL schemes in and file For file URLs a host is expected. For



Python – how to get help information

- 3. Just search the internet. You will find a lot of answers. This is my favorite way.
- 4. To see content of a method associated predefined classes: ??np.mean() or ??X.mean() with X a numpy array

Lcoation of souce code: c:\programdata\anaconda3\lib\site-packages\numpy\core\fromnumeric.py

```
In [15]: ??np.mean()
Signature: np.mean(a, axis=None, dtype=None, out=None, keepdims=<class 'numpy, globals, NoValue'>
Source:
def mean(a, axis=None, dtype=None, out=None, keepdims=np._NoValue):
    Compute the arithmetic mean along the specified axis.
    Returns the average of the array elements. The average is taken over
    the flattened array by default, otherwise over the specified axis.
    `float64` intermediate and return values are used for integer inputs.
    Parameters
    a : array like
        Array containing numbers whose mean is desired. If `a` is not an
        array, a conversion is attempted.
    axis: None or int or tuple of ints, optional
        Axis or axes along which the means are computed. The default is to
        compute the mean of the flattened array.
        .. versionadded:: 1.7.0
        If this is a tuple of ints, a mean is performed over multiple axes,
        instead of a single axis or all the axes as before.
File:
           c:\programdata\anaconda3\lib\site-packages\numpy\core\fromnumeric.py
           function
Type:
```

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Homework Assignment

See separate assignment doc on Latte – assignment #1



Open and save data

- Read from a csv file $df = pd.read_csv('InputDataExample.csv', index_col=0, parse_dates=True)$
- Read from from an xlsx file $df3 = pd.read_excel('InputDataExampleXlsx.xlsx', index_col=0, sheetname='Sheet1')$
- Save to an xlsx file

 writer = pd.ExcelWriter('OutputDataExampleXlsx.xlsx', engine='xlsxwriter')

 # Write each dataframe to a different worksheet.

 df3.to_excel(writer, sheet_name='OriginalData', startrow=1, startcol=0,

 header=True, index=True)
- Open Matlab data file
 import scipy.io as spio mat = spio.loadmat('EF_MonteCarlo_InputData.mat', squeeze_me=True) stock = mat['port']

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Open and save data

Open and write a text file

```
file1 = open("myfile.txt","w")

L = ["This is Delhi \n","This is Paris \n","This is London \n"]

#\n is placed to indicate EOL (End of Line)
file1.write("Hello \n")
file1.writelines(L)
file1.close() #to change file access modes

file1 = open("myfile.txt","r+")
content_1st = file1.readline()
content_2nd = file1.readline()
content_rest = file1.readlines()
```



Python Graphics – line plots and bar charts

```
# line plots
figure count = 1
plt.figure(figure count)
plt.plot(ret1)
plt.ylabel('R(t)')
# bar chart
figure count = figure count+1
plt.figure(figure count)
plt.hist( ret1, normed=True, bins=50,
histtype='stepfilled', alpha=0.5, label='ret')
plt.legend(loc='upper left', bbox to anchor=(0.05,
0.9), shadow=True, ncol=1)
xmin, xmax = -0.1, 0.1
plt.xlim( (xmin, xmax) )
plt.xlabel('returns')
plt.ylabel('frequency')
```



Plotting methods associated with Series/Dataframes

• Example: simple line plot of all data in a dataframe/Series

```
Ret_Dow.plot(kind='hist', bins=bins, normed=True, alpha=0.5, color='blue')
```

• Example: areaplot of data in a dataframe pw = pd.DataFrame(portfolios) pw.columns = df.columns.values pw.index = returns pw.plot.area()



Useful Python Charting Resources

https://www.datacamp.com/community/tutorials/matplot lib-tutorial-python

https://python-graph-gallery.com/

https://matplotlib.org/users/pyplot tutorial.html

https://plot.ly/python/



Extra Resources for Learning Python

- https://www.w3schools.com/python/default.asp
- https://developers.google.com/edu/python/introduction
- https://docs.python.org/3/tutorial/index.html
- https://lectures.quantecon.org/py/index_learning_python.html
- https://pandas.pydata.org/pandas-docs/stable/dsintro.html
- http://www.scipy-lectures.org/intro/numpy/numpy.html
- Book
 - <u>Python for Data Analysis</u>, Wes McKinney, O'Reilly Media, Inc., 2018.

