QF627 Programming and Computational Finance

S0305: Data Manipulation and Visualization

Learning Outcomes:

1. ☑ (DNT) Introduction to Moving Average Crossover strategy
2. ☑ Download CC3.SI.csv from eLearn
3. ☑ Sample Python code has 29 lines (including comments and empty lines).
4. ☑ Sample MATLAB code has 30 lines (including comments and empty lines).
5. After running the sample Python code, 6 *Types* of variables are listed in the "Variable explorer":

▪ axes.support ▪ DataFrame ▪ figure.Figure	▪ DatatimeIndex ▪ module ▪ Series

6. After running the sample MATLAB code, 3 *classes* of variables are listed in the "Workspace":

- datatime
- table
- double

7. Dr. Z first explains the ☑ Python / ☐ MATLAB sample code.
8. There are 4 **`import`** statements in the sample Python code, which are:
   - `import pandas as pd`
   - `import numpy as np`
   - `import matplotlib.pyplot as plt`
   - `import matplotlib.dates as mdates`
9. ☑ True / ☐ False **`read_csv`** can be found in the list of attributes of **`pd`**.
10. In the documentation of **`pandas.read_csv`**, we find the following default values of some parameters:

Parameter	Default Value	Parameter	Default Value
`sep`	`','`	`header`	`'infer'`
`names`	`None`	`index_col`	`None`
`usecols`	`None`	`mangle_dupe_cols`	`True`
`skiprows`	`None`	`skip_blank_lines`	`True`
`parse_dates`	`False`		

11. ☑ True / ☐ False In the function header of function `pandas.read_csv`, there is no *- or **-parameter.
12. A few examples are used to demonstrate the use of parameters "**`header`**", "**`names`**", "**`index_col`**" and "**`parse_dates`**".

header	names	index_col	parse_dates
`'infer'` or `0`			
`None`			
	`range(7,0,-1)`		
`0`	`range(7,0,-1)`		
		`0`	
		`0`	`True`

13. ☑ True / ☐ False **`pandas.read_csv`** read CSV file into a **`DataFrame`**.
14. ☑ True / ☐ False In a **`DataFrame`**, say **`data`**, **`data.columns`** returns the column labels of the **`DataFrame`**, **`data.index`** returns the index/row labels of the **`DataFrame`**. Both are list-like objects (or 1D array-like objects).
15. ☑ True / ☐ False **`Pandas.DataFrame`** is a two-dimensional size-mutable, tabular data structure with labeled axes (rows and columns), which can be thought of as a dict-like container for Series objects. **`Pandas.Series`** is a one-dimensional (Numpy) **`ndarray`** with axis labels.
16. ☑ True / ☐ False In the example on slide 27, both **`data['Open']`** and **`data.Open`** (the syntactic sugar of **`data['Open']`**) can be used to reference the data in the column labeled 'Open'. However, for the column labeled 'Adj Close', only **`data['Adj Close']`** is valid. Such a dict-like indexing method on a DataFrame, i.e. **`DataFrame[colname]`**, results a Series.
17. ☑ True / ☐ False In the example on slide 32, **`data.index`** is **`Index`** type and elements in **`data.index`** are **`str`** type.
18. ☑ True / ☐ False In the example on slide 33, with **`parse_dates=True`** in the **`read_csv`** function, **`data.index`** is **`DatetimeIndex`** type and elements in **`data.index`** are **`Timestamp`** type.
19. ☑ True / ☐ False In the example on slide 34, **`pandas.DataFrame.values`** returns a Numpy representation of the **`DataFrame`**. Only the values in the **`DataFrame`** will be returned, the axes labels will be removed.
20. ☑ True / ☐ False In the file CC3.SI.csv, there is a row where the 'Volume' eqauls 0. In the corresponding **`DataFrame`**, we also see this row. The following statement will remove this row:

```
data.drop(data.index[data['Volume']==0],inplace=True)
```

21. ☑ True / ☐ False In the example on slide 36, **`data['Volume']`** is a **`Series`**.
22. ☑ True / ☐ False Comparison between a **`Series`** and a scalar results a **`Series`**. For example, **`data['Volume']==0`** returns a **`Series`** of **`True`** and **`False`** with the same labels.
23. ☑ True / ☐ False Numpy **`ndarrays`** and Pandas **`Series`**es support boolean indexing, i.e. to use a list or an array of **`True`** or **`False`** as the index. The size of a boolean index must match the indexed array's size.
24. ☑ True / ☐ False **`data.index[data['Volume']==0]`** returns the row indices corresponding to those rows where Volume equals zero.
25. ☑ True / ☐ False If using "**`inplace=False`**", **`DataFrame.drop`** will result a copy, original data remains unchanged.
26. ☑ True / ☐ False **`DataFrame.drop`** uses **`axis=0`** as the default parameter value. It will remove rows with the specified labels. We can use **`axis=1`** and column labels to remove columns.
27. ☑ True / ☐ False **`Series.rolling`** results a Rolling object. Methods available to a Rolling object include: **`.sum()`**, **`.mean()`**, **`.std()`**, **`.max()`**, etc.
28. ☑ True / ☐ False **`RollingObject.mean()`** results a Series. Test the following:
    `data['Adj Close'].rolling(15).mean()`
29. ☑ True / ☐ False We can add a new column to a **`DataFrame`** as adding a new item to a dictionary with the keyword index. We cannot use the "dot notation" to add a column to a DataFrame.
30. ☑ Dr. Z explains why there are many **`NaN`** in the two newly added columns.
31. ☑ True / ☐ False **`numpy.round(Series, 3)`** is equivalent to **`Series.round(3)`**.
32. ☑ True / ☐ False Subtraction of two **`Series`**es (having the same labels) results a **`Series`** with the same labels.
33. ☑ True / ☐ False In the assignment to a slice of a **`Series`**, **`target=expression`**, **`expression`** can be a scalar or an iterable having the same number of items as the number of items in the target.
34. ☑ True / ☐ False **`Series.diff`** returns a **`Series`** of the same length (and same labels).
35. ☑ True / ☐ False For a Series **`y`**, **`y[(y>0) | (y<0)]`** returns a series with those rows in series **`y`** whose values are either greater than zero or less than zero.
36. ☐ True / ☑ False In the above expression, **`y[(y>0) | (y<0)]`**, we can omit the parenthesis and use **`y[y>0 | y<0]`**.
37. ☑ True / ☐ False Python logical OR operator (**`or`**) is **not** an element-wise operator.
38. ☑ True / ☐ False Python bitwise logical OR operator (**`|`**) is **not** an element-wise operator.
39. ☑ True / ☐ False Numpy **`ndarray`** does not have logical OR operation. Numpy has the **`logical_or`** function for the logical OR operation. It is an element-wise operation.

40. ☑ True / ☐ False Numpy **`ndarray`** has bitwise OR operation. It is an element-wise operation. It is a common practice to use the bitwise OR operator (**`|`**) on Numpy logical **`ndarray`**s for the element-wise OR operation on logical arrays.
41. Find the values in the column "crossSell" for the following rows (corresponding to `idxSell`):
    Use expression: `data.loc[idxSell,'crossSell']`

Date	Adj Close	crossSell	
		Before running In-Class Exercise 31	**After** running In-Class Exercise 31
2015-12-22	**3.278104**	NaN	3.278104
2016-01-08	**3.232448**	NaN	3.232448
2016-03-02	**3.086349**	NaN	3.086349
2016-08-31	**3.489341**	NaN	3.489341
2017-02-14	**2.701995**	NaN	2.701995
2017-04-26	**2.729855**	NaN	2.729855
2017-07-07	**2.700000**	NaN	2.700000

42. ☑ Dr. Z demonstrated what a Figure and an Axes are, respectively.
43. ☑ True / ☐ False **`matplotlib.pyplot.subplots`** creates a Figure and a set of subplots/axes.
44. ☑ True / ☐ False **`matplotlib.pyplot.figure`** creates a new Figure and make it the ***Current Figure***. **`matplotlib.pyplot.axes`** add an axes to the ***Current Figure*** and make it the ***Current Axes***.
45. ☑ True / ☐ False Almost all functions from **`pyplot`**, such as **`plt.plot()`**, are implicitly either referring the an existing ***Current Figure*** and ***Current Axes***, or creating them anew if none exist.
46. ☑ True / ☐ False For a complicated plot, it will be clearer and more convenient to use names for Figures and Axes.
47. ☑ True / ☐ False (DNT) **`matplotlib.pyplot.subplot2grid`** creates an axes at specific location inside a regular grid. This function has an input parameter to tell which figure to place axes in. Default figure is the ***Current Figure***.
48. ☑ True / ☐ False **`pandas.DataFrame.plot`** can specify axes, figure size and line styles through parameters **`ax`**, **`figsize`** and **`style`**, respectively.
49. Some color abbreviations supported in Matplotlib:

blue	green	red	cyan	magenta	yellow	black	white
`'b'`	`'g'`	`'r'`	`'c'`	`'m'`	`'y'`	`'k'`	`'w'`

50. Some line styles supported in Matplotlib:

solid line	dashed line	dash-dot line	dotted line
`'-'`	`'--'`	`'-.'`	`':'`

51. ☑ True / ☐ False The format string **`'g-'`** is for **green solid line** and **`'ro'`** is for **red circles**.
52. ☑ True / ☐ False **`DataFrame[list_colnames]`** returns a DataFrame corresponding to **`list_colnames`**.
53. ☑ True / ☐ False There are two **`fill_between`** functions:
    - **`matplotlib.pyplot.fill_between`** and
    - **`matplotlib.axes.Axes.fill_between`**
54. ☑ True / ☐ False In the **`fill_between`** function, **`x`** and **`y1`** are compulsory arguments, **`y2`** is an optional argument. **`x`** can be a 1D array or a sequence type. **`y1`** and **`y2`** can be 1D arrays, sequences of the same length or a scalar.
55. ☑ For slide 79, Dr. Z demonstrated different choices for **`x`** and explained the **`xy`** in the sample code.
56. ☑ True / ☐ False In **`fill_between`**, we can set the fill color by the parameter **`color`**, and the transparency of the fill color by the parameter **`alpha`**.
57. ☑ Sample code for plotting the **moving average** and **moving standard deviation** using **`matplotlib.pyplot.fill_between`**.
58. ☑ (DNT) Sample code for the GUI that contains an Axes using PyQt5.
59. ☑ True / ☐ False There is a library function **`Matplotlib.finance.candlestick_ohlc`** for the candlestick graph. It has a parameter **`ax`** to specify the Axes instance to plot to. The parameter **`quotes`** needs to be a sequence of **`(time, open, high, low, close, …)`** sequences, as long as the first 5 elements are these values, the record can be as long as you want (e.g., it may store volume). **`time`** must be in float days format (see **`date2num`**).
60. Complete the following code as given in In-Class Exercise 35.

```
%234567890123456789012345678901234567890123456789012345678901234567890
from matplotlib.finance import candlestick_ohlc
from matplotlib.dates import date2num
r=data.iloc[:15, :]
fig,ax=plt.subplots()
d=date2num(r.index.date)
candlestick_ohlc(ax, zip(d,r.Open,r.High,r.Low,r.Close)
                 width=0.5, colorup='g', colordown='r', alpha=1)
plt.setp(ax.get_xticklabels(), rotation=30)
ax.xaxis_date()
plt.show()
```

61. ☑ (DNT) Dr. Z demonstrated the use of **`QFileDialog.getOpenFileName`**.
62. ☑ True / ☐ False Numpy arrays use a different structure which uses less pace, runs faster and has optimized functions. A Python list's advantage is its flexibility.
63. ☑ True / ☐ False Python Lists do not support element-wise operations. Element-wise operations can be realized through the list comprehension.

64. Introduction to two Python built-in functions: **`all(iterable)`** and **`any(iterable)`**.

	Return value		Return value
`all([1, 2, 3])`	`True`	**`any([1, 2, 3])`**	`True`
`all([1, 0, 3])`	`False`	**`any([1, 0, 3])`**	`True`
`all([])`	`True`	**`any([])`**	`False`
`all([[]])`	`False`	**`any([[]])`**	`False`

65. Use list comprehension to solve the following problems.

`x=[1, 2, 3, 4]`

Compare `x>1` element by element.	**`[i>1 for i in x]`**
Is there any element of `x` greater than `1`?	**`any([i-1 for i in x])`**
Are all elements of `x` greater than `1`?	**`all([i-1 for i in x])`**
Compute `x+1` element by element.	**`[i+1 for i in x]`**
Compute `x**2` element by element.	**`[i**2 for i in x]`**
Test `x in range(5)` element by element.	**`[i in range(5) for i in x]`**
Is there any element of `x in range(5)`?	**`any([i in range(5) for i in x])`**
Are all elements of `x in range(5)`?	**`all([i in range(5) for i in x])`**

66. ☑ True / ☐ False Numpy arrays support some vectorized computations, such as comparison, addition, power, etc. The membership test operation (**`in`**) is not among these operations. Elementwise membership test is done by the library function **`numpy.isin`**.
67. ☑ True / ☐ False **`numpy.array`** creates an array (**`numpy.ndarray`** type). It can convert a list of numbers to a 1D array, and a nested list of numbers (rectangular shape) to a 2D array.
68. ☑ True / ☐ False Pandas DataFrame supports some vectorized computations, such as comparison, addition, power, etc. Elementwise membership test is done by the library function **`pandas.DataFrame.isin`**.
69. ☑ True / ☐ False **`pandas.DataFrame`** creates/is the primary pandas data structure. The parameter **`data`** can be a Numpy ndarray, a dictionary, or a list-like object, etc. It is a two-dimensional size-mutable, potentially heterogeneous tabular data structure with labeled axes (rows and columns). Parameters **`index`** and **`columns`** are used to specify row labels and column labels, respectively, which will default to **`np.arange(n)`**.
70. ☑ True / ☐ False Python built-in functions **`any`** and **`all`** only work on 1D arrays. For 2D arrays, we can use Numpy library functions **`numpy.any`** and **`numpy.all`**.
71. ☑ True / ☐ False Numpy library function **`numpy.any`** returns single Boolean unless **`axis`** is not **`None`**. For a 2D array, for **`axis=0`**, logical OR is applied along the first index, i.e. on the elements in the same column; for **`axis=1`**, logical OR is applied along the second index, i.e. on the elements in the same row.

72. Return values of **`numpy.any`**:

	Return Value
`np.any([[True,False],[True,False]])`	`True`
`np.any([[True,False],[True,False]], axis=0)`	`[True,False]`
`np.any([[True,False],[True,False]], axis=1)`	`[True,True]`
`np.any([[True,False],[True,False]], axis=-1)`	`[True,True]`

73. Indexing and Slicing of a nested list.

```
x=[[0,1,2],[3,4,5],[6,7,8]]
```

Expression	Value (or Error)
`x`	`[[0,1,2],[3,4,5],[6,7,8]]`
`x[0]`	`[0,1,2]`
`x[:1]`	`[[0,1,2]]`
`x[1][1]`	`4`
`x[:1][:1]`	`[[0,1,2]]`
`x[1,1]`	`error`
`x[:1,:1]`	`error`

74. Indexing and Slicing of a Numpy 2D array.

```
import numpy as np
x=np.arange(9).reshape(3,-1)
```

Expression	Value (or Error)

`x`	`array([[0,1,2]`
	`[3,4,5]`
	`[6,7,8]])`

`x[0]`	`array([0,1,2])`
`x[:1]`	`array([[0,1,2]])`
`x[1][1]`	`4`
`x[:1][:1]`	`array([[0,1,2]])`
`x[1,1]`	`4`
`x[:1,:1]`	`array([[0]])`

75. ☑ True / ☐ False List indices must be integers or slices, not tuple or list.
76. ☑ On slide 115, more examples on Numpy arrays' indexing and slicing are given, including boolean indexing (using a list of True or False) and fancy indexing (using a list of indices).
77. ☑ True / ☐ False For a Numpy 2D array , say **`x`**, **`x[1]`** and **`x[[1]]`** are of different structures though they contain the same numbers.
78. ☑ True / ☐ False (Review) Slicing of a list is a shallow copy.
79. ☑ True / ☐ False (Review) Only iterables can be assigned to the slicing of a list. The iterable assigned to extended slicing must match in size.

80. ☑ True / ☐ False Slicing of a Numpy array is a view. Fancy indexing returns a copy. We can use Numpy library function **`numpy.ndarray.base`** to tell whether memory is from some other object. For example,

```
import numpy as np
x=np.arange(4)
y=x[:2]
z=x[[0,1]]
print(y.base is x, z.base is x)  #True False
x[0]=100
print(y, z) #[100 1] [0 1]
```

81. ☑ True / ☐ False Comparison of single indexing and slicing via **`[]`** on Pandas DataFrame and Numpy 2D arrays. Slicing will return a DataFrame or a 2D array on rows, respectively. Indexing and fancy indexing on Numpy 2D arrays will return rows. Indexing and fancy indexing on DataFrames will return columns. Numpy arrays can use multiple indexing/slicing. Pandas DataFrame does not have multiple indexing/slicing via **`[]`**.

```
import numpy as np
import pandas as pd
x=np.arange(9).reshape(3,-1)
y=pd.DataFrame(x)
```

`x[0]`	**`y[0]`**	`x[:1]`	`y[:1]`
`[0 1 2]`	**`0  0`** **`1  3`** **`2  6`** **`Name: 0, dtype: int32`**	`[[0 1 2]]`	`  0  1  2` `0  0  1  2`
`x[[0]]`	**`y[[0]]`**	`x[1,1]`	**`y[1,1]`**
`[[0 1 2]]`	**`  0`** **`0  0`** **`1  3`** **`2  6`**	`4`	**`ERROR`**

82. ☑ True / ☐ False For Pandas DataFrame data selection via **`.iloc`**, which is integer position based, **`df.iloc[1]`** is equivalent to **`df.iloc[1,:]`**. All single indexing and slicing via **`.iloc`** return rows.
83. ☑ True / ☐ False For Pandas DataFrame data selection via **`.loc`**, which is label based, **`df.loc[1]`** is equivalent to **`df.loc[1,:]`**. All single indexing and slicing via **`.loc`** return rows. **IMPORTANT:** The slice object with labels **includes both the start and the stop**. This is different from usually Python slices.
84. ☑ True / ☐ False Numpy library function **`numpy.arange`** returns evenly spaced values within a given interval. When using a non-integer step, such as 0.1, the results will often not be consistent. It is better to use **`numpy.linspace`** for these cases.
85. ☑ True / ☐ False Numpy library function **`numpy.ndarray.reshape`** returns an array containing the **same data** with a new shape.

86. ☑ True / ☐ False With Numpy ndarray's broadcasting, we can perform binary operations on arrays of different sizes. For example,

Expression	Value (or Error)
`np.arange(3)+5`	`[5 6 7]`

`np.ones((3,3))+np.arange(3))`	`[[1. 2. 3.]`
	` [1. 2. 3.]`
	` [1. 2. 3.]]`

`np.arange(3).reshape((3,1))+np.arange(3)`	`[[0 1 2]`
	` [1 2 3]`
	` [2 3 4]]`

87. ☑ True / ☐ False Numpy library function **`numpy.ones`** returns a new array of given shape and type, filled with ones.
88. ☑ True / ☐ False Pandas DataFrame does not have Numpy ndarray's broadcasting.
89. ☑ Dr. Z demonstrated an example of subtraction of two Pandas Series with different labels.
90. Implementation of the algorithm on slide 139 (binomial tree method for option pricing):

Variable Names							
S	K	r	q	t	T	σ	N
S	K	r	q	t	T	sigma	N
Δt	u	d	p	V	i	j	$f_{i,j}$
dt	u	d	p	V	i	j	???

91. ☑ True / ☐ False Numpy library function **`numpy.maximum(x1,x2)`** compares two arrays and returns a new array containing the element-wise maxima. If one of the elements being compared is a NaN, then that element (NaN) is returned. If both elements are NaNs then the first (NaN) is returned. When comparing two arrays of different sizes, Numpy array's broadcasting will apply.

92. **Binomial Tree Algorithm Implentations 1&2**: $f_{i,j} \rightarrow$ `f[i][j]` using a nested list.

```
#234567890123456789012345678901234567890123456789012345678901234567890123456789012345
from math import exp, sqrt
def BTA(S,K,r,q,tau,sigma,N=100):
   #1
   deltaT=tau/N
   u=exp(sigma*sqrt(deltaT))
   d=1/u
   p=(exp((r-q)*deltaT)-d)/(u-d)
   #2
   fc=[[0.0 for j in range(i+1)] for i in range(N+1)]
   fp=[[0.0 for j in range(i+1)] for i in range(N+1)]
   #if using list comprehension
   #fc=[0]*(N+1)
   #fp=[0]*(N+1)

   for j in range(N+1):
      fc[N][j]=max(0,S*(u**j)*(d**(N-j))-K)
      fp[n][j]=max(0,K-S*(u**j)*(d**(N-j)))
   #if using list comprehension
   #fc[N]=[max(0, S*(u**j)*(d**(N-j))-K) for j in range(N+1)]
   #fp[N]=[max(0, K-S*(u**j)*(d**(N-j))) for j in range(N+1)]

   #3
   ert=exp(-r*deltaT)
   p1=1-p
   for i in range(N-1,0-1,-1):
      for j in range(i+1):
         fc[i][j]=ert*(p*fc[i+1][j+1]+p1*fc[i+1][j])
         fp[i][j]=ert*(p*fp[i+1][j+1]+p1*fp[i+1][j])
      #if using list comprehension
      #fc[i]=[ert*(p*fc[i+1][j+1]+p1*fc[i+1][j]) for j in range(i+1)]
      #fp[i]=[ert*(p*fp[i+1][j+1]+p1*fp[i+1][j]) for j in range(i+1)]

   #4
   return (fc[0][0], fp[0][0])

if __name__=='__main__':
   S=50.0
   K=50.0
   t=0
   T=183/365
   sigma=0.4
   r=0.04
   q=0.01
   N=100
   print(BTA(S,K,r,q,t,T,sigma,N))
```

93. **Binomial Tree Algorithm Implentation 3**: $f_{i,j} \rightarrow$ `f[i,j]` using a Numpy 2D array.

```
#234567890123456789012345678901234567890123456789012345678901234567890123456789012345
import numpy as np
def BTA(S, K, r, q, tau, sigma, N=100):
    #1
    deltaT=tau/N
    u=np.exp(sigma*np.sqrt(deltaT)
    d=1/u
    p=(np.exp((r-q)*deltaT)-d)/(u-d)
    #2
    p1=1-p
    ert=np.exp(-r*deltaT)
    for i in range(N-1, 0-1, -1):
        fc[i, 0:i+1]=ert*(p*fc[i+1, 0+1:i+1+1]+p1*fc[i+1,0:i+1])
        fp[i, 0:i+1]=ert*(p*fp[i+1, 0+1:i+1+1]+p1*fp[i+1,0:i+1])

    #3
    p1=1-p
    ert=np.exp(-r*deltaT)
    for i in range(N-1,0-1,-1):
       fc[i,0:i+1]=ert*(p*fc[i+1,0+1:i+1+1]+p1*fc[i+1,0:i+1])
       fp[i,0:i+1]=ert*(p*fp[i+1,0+1:i+1+1]+p1*fp[i+1,0:i+1])

    #4
    return (fc[0, 0], fp[0, 0])

if __name__=='__main__':
    S=50.0
    K=50.0
    t=0
    T=183/365
    sigma=0.4
    r=0.04
    q=0.01
    N=100
    print(BTA(S,K,r,q,t,T,sigma,N))
```

94. ☑ True / ☐ False The **`numpy.random`** module has library functions to generate random numbers from various distributions, e.g. **`numpy.random.randn`** returns a sample (or samples) from the standard normal distribution, **`numpy.random.random`** returns random floats in the half-open interval [0.0, 1.0) (a.k.a. the continuous uniform distribution).

Generate a 3-by-4 2D ndarray of random numbers from the normal distribution, and name this array x.	`import numpy as np`
	`x=np.random.randn(3,4)`
	`#or`
	`#x=np.random.standard_normal((3,4))`

Generate a 1D array of 10 random numbers from the normal distribution, and name this array x.	`import numpy as np`
	`x=np.random.randn(1,10)`
	`#or`
	`#x=np.random.standard_normal((1,10))`

Generate a 3-by-4 2D ndarray of random numbers from the U[0,1) distribution, and name this array x.	`import numpy as np` `x=np.random.random((3,4))`
Generate a 1D array of 10 random numbers from the U[0,1) distribution, and name this array x.	`import numpy as np` `x=np.random.random((1,10))`

95. ☑ True / ☐ False Numpy library function **`numpy.append(`*`arr`*`,`*`values`*`,`*`axis=None`*`)`** appends ***`values`*** to a copy of ***`arr`***. If the parameter **`axis`** is **`None`**, the function returns a flattened array. ***`values`*** must be of the correct shape (same shape as ***`arr`***, excluding ***`axis`***). For two 2D arrays, with **`axis=0`** it will append the second array with shape **`(m2, n)`** to the first array with shape **`(m1, n)`** as new rows, and with **`axis=1`** it will append the second array with shape **`(m, n2)`** to the first array with shape **`(m, n1)`** as new columns.
96. ☑ True / ☐ False Numpy library function **`numpy.concatenate`** join a sequence of arrays along an existing **`axis`**. The default value of **`axis`** is 0. **`numpy.concatenate(([1,2],[3,4]))`** is equivalent to **`numpy.append([1,2],[3,4])`**.
97. ☑ True / ☐ False Numpy library function **`numpy.random.seed`** seeds the random number generator. It can be called again to re-seed the generator.
    Q: What is printed by the third print statement?

Code

```
import numpy as np
numpy.random.seed(0)
print(np.random.randn(2))    #1
print(np.random.randn(2))    #2
numpy.random.seed(0)
print(np.random.randn(4))    #3
```

Output

```
[1.76405235 0.40015721]
[0.97873798 2.2408932]
[1.76405235 0.40015721 0.97873798 2.2408932]
```

98. **Monte Carlo Simulation Algorithm Implementation (Sample Exam, Question 6)**

`import numpy as np`
Given $S = 50, K = 50, t = 0, T = 183/365, \sigma = 0.4, r = 0.04, q = 0.01, N = 10000000.$
`S=50; K=50, T=183/365, sigma=0.4, r=0.04, q=0.01, N=10_000_000`
Use one command to create a vector of $\frac{N}{2}$ random numbers with standard normal distribution, denoted as d: $d = \left(d_1, d_2, \dots, d_{\frac{N}{2}}\right)$
`d=np.random.randn(N/2)`
Use one command and the library function **`numpy.concatenate`** to create a vector of N elements by using d as the first half, and $-d$ as the second half. Name this vector as p. $$p = (p_1, p_2, \dots, p_N) = \left(d_1, d_2, \dots, d_{\frac{N}{2}}, -d_1, -d_2, \dots, -d_{\frac{N}{2}}\right)$$
`p=np.concatenate((d,-d))`
Use one command to compute the mean and standard deviation of the N numbers in the vector p, and denote them as μ_p and σ_p, respectively.
`mu,sigma=np.mean(p),np.std(p)`
Use one command to update every element in the vector p as follows: $$p_i \leftarrow \frac{p_i - \mu_p}{\sigma_p} \text{ for } i = 1, 2, \dots, N.$$
`p[:,:]=(p[:,:]-mu)/sigma`
Use one command to obtain the vector $Y = (Y_1, Y_2, \dots, Y_N)$ computed as follows: $$Y_i = S \cdot e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma \cdot \sqrt{T} \cdot p_i} \text{ for } i = 1, 2, \dots, N.$$
`Y=S*np.exp((r-(sigma**2)/2)*T+sigma*mp.sqrt(T)*p)`
Use one command to obtain the vector $h = (h_1, h_2, \dots, h_N)$ computed as follows: $$h_i = \max(Y_i - K, 0) \text{ for } i = 1, 2, \dots, N.$$
`h=np.maxmize(Y-K,0)`
Use one command to calculate and return V as follows: $$V = \frac{1}{N} e^{-r \cdot T} \sum_{i=1}^{N} h_i$$
`V=np.exp(-r*T)*np.mean(h)`