

Data Analysis and Visualisation with Python

Numerical Python (NumPy)

- NumPy is the most foundational package for numerical computing in Python.
- If you are going to work on data analysis or machine learning projects, then having a solid understanding of NumPy is nearly mandatory.
- Indeed, many other libraries, such as pandas and scikit-learn, use NumPy's array objects as the *lingua franca* for data exchange.
- One of the reasons as to why NumPy is so important for numerical computations is because it is designed for efficiency with large arrays of data. The reasons for this include:
 - It stores data internally in a continuous block of memory, independent of other in-built Python objects.
 - It performs complex computations on entire arrays without the need for for loops.

What you'll find in NumPy

- ndarray: an efficient multidimensional array providing fast array-orientated arithmetic operations and flexible *broadcasting* capabilities.
- Mathematical functions for fast operations on entire arrays of data without having to write loops.
- Tools for reading/writing array data to disk and working with memory-mapped files.
- Linear algebra, random number generation, and Fourier transform capabilities.
- A C API for connecting NumPy with libraries written in C, C++, and FORTRAN. This is why Python is the language of choice for wrapping legacy codebases.

The NumPy ndarray: A multi-dimensional array object

- The NumPy ndarray object is a fast and flexible container for large data sets in Python.
- NumPy arrays are a bit like Python lists, but are still a very different beast at the same time.
- Arrays enable you to store multiple items of the same data type. It is the facilities around the array object that makes NumPy so convenient for performing math and data manipulations.

Ndarray vs. lists

- By now, you are familiar with Python lists and how incredibly useful they are.
- So, you may be asking yourself:

“I can store numbers and other objects in a Python list and do all sorts of computations and manipulations through list comprehensions, for-loops etc. What do I need a NumPy array for?”

- There are very significant advantages of using NumPy arrays over lists.

Creating a NumPy array

- To understand these advantages, let's create an array.
- One of the most common, of the many, ways to create a NumPy array is to create one from a list by passing it to the `np.array()` function.

```
In: import numpy as np
    list1 = [0, 1, 2, 3, 4]
    arr = np.array(list1)

    print(type(arr))
    print(arr)
```

```
Out: In [1]: runfile('C:/Users,
          wdir='C:/Users/Lew_laptop,
          <type 'numpy.ndarray'>
          [0 1 2 3 4]
```

Differences between lists and ndarrays

- The key difference between an array and a list is that arrays are designed to handle vectorised operations while a python lists are not.
- That means, if you apply a function, it is performed on every item in the array, rather than on the whole array object.

- Let's suppose you want to add the number 2 to every item in the list. The intuitive way to do this is something like this:

```
In: import numpy as np
    list1 = [0, 1, 2, 3, 4]
    list1 = list1+2
```

Out: File "C:/Users/Lew_laptop/.spyder-py3/temp.py", line 9, in
list1 = list1+2
TypeError: can only concatenate list (not "int") to list

- That was not possible with a list, but you can do that on an array:

```
In: import numpy as np
    list1 = [0, 1, 2, 3, 4]
    arr = np.array(list1)
    print(arr)
    arr = arr+2
    print(arr)
```

Out: In [7]: runfile('C:/Users
Lew_laptop/.spyder-py3')
[0 1 2 3 4]
[2 3 4 5 6]

- It should be noted here that, once a Numpy array is created, you cannot increase its size.
- To do so, you will have to create a new array.

Create a 2d array from a list of list

- You can pass a list of lists to create a matrix-like a 2d array.

```
In: import numpy as np
list2 = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
arr2=np.array(list2)
print(arr2)
```

```
Out: [[0 1 2]
      [3 4 5]
      [6 7 8]]
```

The dtype argument

- You can specify the data-type by setting the dtype() argument.
- Some of the most commonly used NumPy dtypes are: float, int, bool, str, and object.

```
In: import numpy as np
list2 = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
arr3=np.array(list2, dtype='float')
print(arr3)
```

```
Out: [[0.  1.  2.]
      [3.  4.  5.]
      [6.  7.  8.]
```

The astype argument

- You can also convert it to a different data-type using the `astype` method.

```
In: import numpy as np
list2 = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
arr3=np.array(list2, dtype='float')
print(arr3)
arr3_s = arr3.astype('int').astype('str')
print(arr3_s)
```

```
Out: [[0.  1.  2.]
      [3.  4.  5.]
      [6.  7.  8.]]
      [['0' '1' '2']
      ['3' '4' '5']
      ['6' '7' '8']]
```

- Remember that, unlike lists, all items in an array have to be of the same type.

dtype='object'

- However, if you are uncertain about what data type your array will hold, or if you want to hold characters and numbers in the same array, you can set the dtype as 'object'.

```
In: arr_obj = np.array([1, 'a'], dtype='object')  
    print(arr_obj)
```

```
Out: [1 'a']
```

The tolist() function

- You can always convert an array into a list using the tolist() command.

```
In: arr_list = arr_obj.tolist()  
    print(arr_list)
```

```
Out: [1, 'a']
```

Inspecting a NumPy array

- There are a range of functions built into NumPy that allow you to inspect different aspects of an array:

```
In: import numpy as np
list2 = [[0, 1, 2], [3, 4, 5], [6, 7, 8]] Out:
arr3=np.array(list2, dtype='float')           Shape: (3, 3)
print('Shape:', arr3.shape)                  Data type: float64
print('Data type:', arr3.dtype)              Size: 9
print('Size:', arr3.size)                    Num dimensions: 2
print('Num dimensions:', arr3.ndim)
```

Extracting specific items from an array

- You can extract portions of the array using indices, much like when you're working with lists.
- Unlike lists, however, arrays can optionally accept as many parameters in the square brackets as there are number of dimensions

```
In: import numpy as np
list2 = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
arr3=np.array(list2, dtype='float')
print("whole:", arr3)
print("Part:", arr3[:2, :2])
```

```
Out: whole: [[0. 1. 2.]
 [3. 4. 5.]
 [6. 7. 8.]]
Part: [[0. 1.]
 [3. 4.]]
```


Boolean indexing

- A boolean index array is of the same shape as the array-to-be-filtered, but it only contains TRUE and FALSE values.

```
In: import numpy as np
    list2 = [[0, 1, 2], [3, 4, 5], [6, 7, 8]]
    arr3=np.array(list2, dtype='float')
    boo = arr3>2
    print(boo)
```

Out:

[[False	False	False]
[True	True	True]
[True	True	True]]

Pandas

- Pandas, like NumPy, is one of the most popular Python libraries for data analysis.
- It is a high-level abstraction over low-level NumPy, which is written in pure C.
- Pandas provides high-performance, easy-to-use data structures and data analysis tools.
- There are two main structures used by pandas; *data frames* and *series*.

Indices in a pandas series

- A pandas series is similar to a list, but differs in the fact that a series associates a label with each element. This makes it look like a dictionary.
- If an index is not explicitly provided by the user, pandas creates a RangeIndex ranging from 0 to $N-1$.
- Each series object also has a data type.

```
In: import pandas as pd
    new_series = pd.Series([5, 6, 7, 8, 9, 10])
    print(new_series)
```

```
Out: 0    5
     1    6
     2    7
     3    8
     4    9
     5   10
     dtype: int64
```

- As you may suspect by this point, a series has ways to extract all of the values in the series, as well as individual elements by index.

```
In: import pandas as pd
    new_series = pd.Series([5, 6, 7, 8, 9, 10])
    print(new_series.values)
    print('_____')
    print(new_series[4])
```

```
Out: [ 5  6  7  8  9 10]
      : _____
      9
```

- You can also provide an index manually.

```
In: import pandas as pd
    new_series = pd.Series([5, 6, 7, 8, 9, 10], index=['a', 'b', 'c', 'd', 'e', 'f'])
    print(new_series.values)
    print('_____')
    print(new_series['f'])
```

```
Out: [ 5  6  7  8  9 10]
      : _____
      10
```

- It is easy to retrieve several elements of a series by their indices or make group assignments.

```
In: import pandas as pd
new_series = pd.Series([5, 6, 7, 8, 9, 10], index=['a', 'b', 'c', 'd', 'e', 'f'])
print(new_series)
print('_____')
new_series[['a', 'b', 'f']] = 0
print(new_series)
```

Out:

a	5
b	6
c	7
d	8
e	9
f	10

dtype: int64

a	0
b	0
c	7
d	8
e	9
f	0

dtype: int64

Filtering and maths operations

- Filtering and maths operations are easy with Pandas as well.

```
In: import pandas as pd
    new_series = pd.Series([5, 6, 7, 8, 9, 10], index=['a', 'b', 'c', 'd', 'e', 'f'])
    new_series2 = new_series[new_series>0]
    print(new_series2)
    print('_____')
    new_series2[new_series2>0]*2
    print(new_series2)
```

```
Out: a      5
     b      6
     c      7
     d      8
     e      9
     f     10
     dtype: int64

_____
a      5
b      6
c      7
d      8
e      9
f     10
dtype: int64
```

Pandas data frame

- Simplistically, a data frame is a table, with rows and columns.
- Each column in a data frame is a series object.
- Rows consist of elements inside series.

Case ID	Variable one	Variable two	Variable 3
1	123	ABC	10
2	456	DEF	20
3	789	XYZ	30

Creating a Pandas data frame

- Pandas data frames can be constructed using Python dictionaries.

```
In: import pandas as pd
     df = pd.DataFrame({
         'country': ['Kazakhstan', 'Russia', 'Belarus', 'Ukraine'],
         'population': [17.04, 143.5, 9.5, 45.5],
         'square': [2724902, 17125191, 207600, 603628]})
     print(df)
```

```
Out:   country  population  square
0  Kazakhstan    17.04    2724902
1    Russia    143.50   17125191
2   Belarus     9.50    207600
3    Ukraine    45.50    603628
```


- You can also create a data frame from a list.

```
In: import pandas as pd
list2 = [[0,1,2],[3,4,5],[6,7,8]]
df = pd.DataFrame(list2)
print(df)
df.columns = ['V1', 'V2', 'V3']
print(df)
```

```
Out: In [12]: runfile('
      0  1  2
0  0  1  2
1  3  4  5
2  6  7  8
      V1  V2  V3
0  0  1  2
1  3  4  5
2  6  7  8
```

- You can ascertain the type of a column with the `type()` function.

```
In: print(type(df['country']))
```

```
Out: <class 'pandas.core.series.Series'>
```

- A Pandas data frame object as two indices; a column index and row index.
- Again, if you do not provide one, Pandas will create a RangeIndex from 0 to $N-1$.

```
In: import pandas as pd
     df = pd.DataFrame({
         'country': ['Kazakhstan', 'Russia', 'Belarus', 'Ukraine'],
         'population': [17.04, 143.5, 9.5, 45.5],
         'square': [2724902, 17125191, 207600, 603628]})
     print(df.columns)
     print('_____')
     print(df.index)
```

```
Out: Index(['country', 'population', 'square'], dtype='object')
      _____
      RangeIndex(start=0, stop=4, step=1)
```

- There are numerous ways to provide row indices explicitly.
- For example, you could provide an index when creating a data frame:

```
In: import pandas as pd
df = pd.DataFrame({
    'country': ['Kazakhstan', 'Russia', 'Belarus', 'Ukraine'],
    'population': [17.04, 143.5, 9.5, 45.5],
    'square': [2724902, 17125191, 207600, 603628]
}, index=['KZ', 'RU', 'BY', 'UA'])
print(df)
```

Out:

	country	population	square
KZ	Kazakhstan	17.04	2724902
RU	Russia	143.50	17125191
BY	Belarus	9.50	207600
UA	Ukraine	45.50	603628

- or do it during runtime.
- Here, I also named the index 'country code'.

```
In: import pandas as pd
df = pd.DataFrame({
    'country': ['Kazakhstan', 'Russia', 'Belarus', 'Ukraine'],
    'population': [17.04, 143.5, 9.5, 45.5],
    'square': [2724902, 17125191, 207600, 603628]
})
print(df)
print('_____')
df.index = ['KZ', 'RU', 'BY', 'UA']
df.index.name = 'Country Code'
print(df)
```

Out:

	country	population	square
0	Kazakhstan	17.04	2724902
1	Russia	143.50	17125191
2	Belarus	9.50	207600
3	Ukraine	45.50	603628

	country	population	square
Country Code			
KZ	Kazakhstan	17.04	2724902
RU	Russia	143.50	17125191
BY	Belarus	9.50	207600
UA	Ukraine	45.50	603628

- Row access using index can be performed in several ways.
- First, you could use `.loc()` and provide an index label.

```
In: print(df.loc['KZ'])
```

```
Out: country      Kazakhstan  
      population    17.04  
      square      2724902  
      Name: KZ, dtype: object
```

- Second, you could use `.iloc()` and provide an index number

```
In: print(df.iloc[0])
```

```
Out: country      Kazakhstan  
      population    17.04  
      square      2724902  
      Name: KZ, dtype: object
```

- A selection of particular rows and columns can be selected this way.

```
In: print(df.loc[['KZ', 'RU'], 'population'])
```

```
Out: Country Code  
KZ      17.04  
RU     143.50  
Name: population, dtype: float64
```

- You can feed `.loc()` two arguments, index list and column list, slicing operation is supported as well:

```
In: print(df.loc['KZ':'BY', :])
```

```
Out:
```

	country	population	square
Country Code			
KZ	Kazakhstan	17.04	2724902
RU	Russia	143.50	17125191
BY	Belarus	9.50	207600

Filtering

- Filtering is performed using so-called Boolean arrays.

```
print(df[df.population > 10][['country', 'square']])
```

	country	square
Country Code		
KZ	Kazakhstan	2724902
RU	Russia	17125191
UA	Ukraine	603628

Deleting columns

- You can delete a column using the `drop()` function.

```
In: print(df)
```

```
Out: Country Code      country  population  square
      KZ      Kazakhstan    17.04    2724902
      RU      Russia      143.50    17125191
      BY      Belarus     9.50     207600
      UA      Ukraine     45.50     603628
```

```
In: df = df.drop(['population'], axis='columns')
    print(df)
```

```
Out: Country Code      country  square
      KZ      Kazakhstan    2724902
      RU      Russia      17125191
      BY      Belarus     207600
      UA      Ukraine     603628
```


Reading from and writing to a file

- Pandas supports many popular file formats including CSV, XML, HTML, Excel, SQL, JSON, etc.
- Out of all of these, CSV is the file format that you will work with the most.
- You can read in the data from a CSV file using the `read_csv()` function.

```
df = pd.read_csv('filename.csv', sep=',')
```

- Similarly, you can write a data frame to a csv file with the `to_csv()` function.

```
df.to_csv('filename.csv')
```

- Pandas has the capacity to do much more than what we have covered here, such as grouping data and even data visualisation.
- However, as with NumPy, we don't have enough time to cover every aspect of pandas here.

Any questions?