



Python Data Analytics

OBA 410/510

Lundquist College of Business



Strings

- Objects that contain sequences of character data

```
In [52]: name="Jessica"
```

```
In [53]: name
```

```
Out[53]: 'Jessica'
```

```
In [54]: address="345 W 13th St"  
address
```

```
Out[54]: '345 W 13th St'
```

- We can access individual characters in a string using the letter's index
 - first character having an index value of 0

```
In [55]: name[0]
```

```
Out[55]: 'J'
```

```
In [57]: name[4]
```

```
Out[57]: 'i'
```

```
In [58]: address[:3]
```

```
Out[58]: '345'
```



Methods

- Functions that belong to objects
- For instance *string* is an object
 - Therefore, there are methods for *strings*
- For instance:
- *upper()*
 - Return a copy of the string converted to uppercase.
- *count(sub)*
 - Return the number of occurrences of substring sub

```
In [59]: name.upper()
```

```
Out[59]: 'JESSICA'
```

```
In [63]: address.count('3')
```

```
Out[63]: 2
```



Lists

- Lists are a sequence of values that is similar to a string
- Differences with strings:
 - list is a sequence of any type, while a string is only a sequence of characters
 - lists are mutable and we can change elements
- We create lists using square brackets
- Lists can even contain other lists as an element

```
In [66]: list1=[3,5,7,9]
         list1
Out[66]: [3, 5, 7, 9]

In [69]: list2=[3,'hello']
         list2
Out[69]: [3, 'hello']

In [74]: list3=['hi',45,True,[33,1]]
         list3
Out[74]: ['hi', 45, True, [33, 1]]

In [75]: list2[1]='bye'
         list2
Out[75]: [3, 'bye']
```



Slicing and dicing lists

- We use indexes similar to strings

```
In [76]: list1[2:]
```

```
Out[76]: [7, 9]
```

```
In [77]: list3[3]
```

```
Out[77]: [33, 1]
```

```
In [78]: list3[2:4]
```

```
Out[78]: [True, [33, 1]]
```

```
In [80]: list2[1]
```

```
Out[80]: 'bye'
```



Adding/removing elements to/from lists

- Listname.append(new elements)
- Listname.remove(elements)

```
In [15]: 1 list3.remove('Hi')  
        2 list3
```

```
Out[15]: [45, 54, False, 54, 'Hello']
```

```
In [16]: 1 list3.remove(list3[0])  
        2 list3
```

```
Out[16]: [54, False, 54, 'Hello']
```

```
1 list3=[45,54,'Hi',False,54]
```

```
1 list3.append('Hello')  
2 list3
```

```
[45, 54, 'Hi', False, 54, 'Hello']
```



Defining new Functions-Example

- Defining a function that takes a list of numbers as input and adds up the first and last elements of the list

```
In [20]: 1 def list_add(l1):  
         2     a=l1[0]+l1[-1]  
         3     return a
```

```
In [27]: 1 list_add([3,5,7,132])
```

```
Out[27]: 135
```

```
In [28]: 1 my_list=[3,5,7,132]  
         2 list_add(my_list)
```

```
Out[28]: 135
```




Dictionaries

- A dictionary is similar to a list where the indices are not limited to only integers
- The dictionary is a set of key-value pairs where the key is the index to its associated value
- We create dictionaries using curly brackets

```
In [218]: dictionary_1={"a":3,5:"California", 'c':'Oregon'}  
  
In [220]: dictionary_1[5]  
Out[220]: 'California'
```

```
In [90]: dictionary_1={"a":3,'b':"California", 'c':'Oregon'}  
  
In [92]: dictionary_1["c"]  
Out[92]: 'Oregon'  
  
In [87]: dictionary_1['a']  
Out[87]: 3  
  
In [93]: dictionary_1['b']='Washington'  
          dictionary_1  
Out[93]: {'a': 3, 'b': 'Washington', 'c': 'Oregon'}
```




Python Packages

- Methods and functions are very useful, but they are not enough
- If we do not use packages, we need to write long codes for many simple things
- We do not need to be worried about maintaining every single line in our code, if we use packages
- Packages make our life easier!!
- Before using any package, we have to install and then import it



Python Packages

- scikit-learn
 - Most commonly used library for predictive modeling both in industry and academia
 - It contains most of the machine learning algorithms
 - Comprehensive documentation about each algorithm:
 - <http://scikit-learn.org/stable/documentation.html>
- NumPy
- SciPy
 - fundamental packages for scientific computing
- Matplotlib
 - Primary scientific plotting library in Python
- Pandas
 - Library for data manipulation and preparation



Python Packages

- NumPy
 - One of the fundamental packages for scientific computing
 - Contains functionalities for multidimensional arrays and linear algebra operations
 - scikit-learn takes in data in the form of Numpy arrays
 - So, any data we use, have to be converted to Numpy arrays
 - The core functionality of Numpy is the *ndarray* class, n-dimensional array
 - *import numpy as np*
 - *x=np.array([[3,5,7],[1,3,9]])*

```
In [4]: x
```

```
Out[4]: array([[3, 5, 7],  
              [1, 3, 9]])
```



Installing other packages

- You can use *pip* to install packages for Python
- *pip* (PIP Installs Packages or Preferred Installer Program) is a command-line utility in your console that allows you to install, reinstall, or uninstall PyPI (Python Package Index) packages
- To install a package use the following command (in your console):
 - *pip install* package-name
- To list all installed packages:
 - *pip list*
- To upgrade an outdated package:
 - *pip install --upgrade* SomeProject
- For more information see the following link:
 - <https://packaging.python.org/tutorials/installing-packages/>



NumPy

- Start using NumPy by **importing** it!

```
In [107]: import numpy as np
```

- NumPy *array* is an alternative to Python *list*
- We can do calculations over the entire array
- Operations are very fast
- But NumPy *arrays* can contain only one type of data!
- If you try to create an array with different data types, some of the element's type will change to have a homogeneous array



NumPy arrays

```
In [108]: # Creating one-dimension array  
a1=np.array([4, 8, 1, 0])  
a1
```

```
Out[108]: array([4, 8, 1, 0])
```

```
In [110]: a1.shape
```

```
Out[110]: (4,)
```

```
In [111]: #creating two-dimension array  
a2=np.array([[3,5,7],[1,1,1]])  
a2
```

```
Out[111]: array([[3, 5, 7],  
                [1, 1, 1]])
```

```
In [112]: a2.shape
```

```
Out[112]: (2, 3)
```

```
In [118]: # creating an array from a list  
l1=[[11,2,5,7,99],[6,4,4,0,90],[1,0,0,0,32]]  
a4=np.array(l1)  
a4
```

```
Out[118]: array([[11,  2,  5,  7, 99],  
                [ 6,  4,  4,  0, 90],  
                [ 1,  0,  0,  0, 32]])
```

```
In [113]: # type coercion  
a3=np.array([4,'Hi', True])  
a3
```

```
Out[113]: array(['4', 'Hi', 'True'], dtype='<U11')
```



Max, min, ...

- *max()* finds the maximum element in an array
- *min()* finds the minimum element in an array
- *argmax()* finds the index of the maximum element in an array
- *argmin()* finds the index of the minimum element in an array

```
In [140]: a6=np.array([2,6,0,36,14])  
a6  
Out[140]: array([ 2,  6,  0, 36, 14])  
  
In [141]: a6.max()  
Out[141]: 36  
  
In [142]: a6.argmax()  
Out[142]: 3  
  
In [143]: a6.min()  
Out[143]: 0  
  
In [144]: a6.argmin()  
Out[144]: 2
```




Mean, median, etc.

```
In [33]: list1=[11,4,2,88,0,2,5,-11,2,4]  
list1
```

```
Out[33]: [11, 4, 2, 88, 0, 2, 5, -11, 2, 4]
```

```
In [25]: np.min(list1)
```

```
Out[25]: -11
```

```
In [26]: np.argmin(list1)
```

```
Out[26]: 7
```

```
In [27]: np.max(list1)
```

```
Out[27]: 88
```

```
In [28]: np.argmax(list1)
```

```
Out[28]: 3
```

```
In [29]: np.median(list1)
```

```
Out[29]: 3.0
```

```
In [31]: np.std(list1)
```

```
Out[31]: 26.287069064465896
```



pandas

- <https://pandas.pydata.org/>
- Python library for data cleaning and pre-processing
- It is built around a data structure called DataFrame
- DataFrame is a tabular, column-oriented data structure with both column and row labels
- DataFrame is a data table similar to an Excel spreadsheet
- pandas provides a great range of methods to modify and operate on the structures
- In contrast to NumPy arrays, DataFrame allows each column to have a separate data type (for example integer, float, dates, and string)



panda's Data Structures

- **Series**

- One-dimensional ndarray with axis labels
- A cross between a list and a dictionary
- Items are all stored in an order and there's labels with which you can retrieve them
- We can create Series from
 - *lists* (the index will be incrementing integers),
 - *dictionaries* (the index is automatically assigned from the keys of the dictionary),
 - Or from scratch by passing a list of data, and index to the series (If you don't give an index to the series, the index will be incrementing integers)

Series



```
In [18]: import pandas as pd
```

```
In [32]: sports=['football','soccer','Taekwondo']
```

```
In [33]: s1=pd.Series(sports)
s1
```

```
Out[33]: 0    football
         1     soccer
         2   Taekwondo
         dtype: object
```

```
In [36]: sports_dic={'US':'football','Argentina':'soccer','South Korea':'Taekwondo'}
```

```
In [38]: s2=pd.Series(sports_dic)
s2
```

```
Out[38]: US          football
         Argentina    soccer
         South Korea  Taekwondo
         dtype: object
```

```
In [39]: s3 = pd.Series(['football','soccer','Taekwondo'], index=['US','Argentina','South Korea'])
s3
```

```
Out[39]: US          football
         Argentina    soccer
         South Korea  Taekwondo
         dtype: object
```



Querying Series

- A panda Series can be queried, either by the index position or the index label
- To query by numeric location, starting at zero, use the ***iloc*** attribute. To query by the index label, you can use the ***loc*** attribute.
- Keep in mind that *iloc* and *loc* are not methods, they are attributes. So you don't use parentheses to query them, but square brackets instead, which we'll call the indexing operator.

```
In [40]: s3.iloc[0]
```

```
Out[40]: 'football'
```

```
In [43]: s3.loc['US']
```

```
Out[43]: 'football'
```

```
In [42]: s3.loc['Argentina']
```

```
Out[42]: 'soccer'
```



Modify/Adding New Data

- *loc* attribute lets you modify data in place and also add new data
- If the value you pass in as the index doesn't exist, then a new entry is added
- Indices can have mixed types

```
In [57]: s3.loc['US']='Baseball'  
s3
```

```
Out[57]: US          Baseball  
Argentina          soccer  
South Korea        Taekwondo  
dtype: object
```

```
In [58]: s3.loc['Canada']='hockey'  
s3
```

```
Out[58]: US          Baseball  
Argentina          soccer  
South Korea        Taekwondo  
Canada             hockey  
dtype: object
```



Deleting data from Series

- `Series.drop(labels=None, inplace=False)`
 - `labels` : single label or list-like
 - Index or column labels to drop.
 - `inplace` : bool, default False
 - If True, do operation inplace and return None.

```
In [9]: 1 s3 = pd.Series(['football', 'soccer', 'Taekwondo'], index=['US', 'Argentina', 'South Korea'])  
        2 s3
```

```
Out[9]: US          football  
        Argentina    soccer  
        South Korea  Taekwondo  
        dtype: object
```

```
In [10]: 1 s3.drop('US')
```

```
Out[10]: Argentina    soccer  
        South Korea  Taekwondo  
        dtype: object
```




panda's Data Structures

- **DataFrame**

- The *DataFrame* data structure is the heart of the panda's library
- The *DataFrame* is conceptually a two-dimensional *Series* object
- There is an index and multiple columns of content, with each column having a label

```
In [10]: pd.DataFrame([[ 'Milk', 'Bread', 'Beer'],[12,8,25]]).T
```

```
Out[10]:
```

	0	1
0	Milk	12
1	Bread	8
2	Beer	25

```
In [64]: df = pd.DataFrame(data={'Product':['Milk','Bread','Beer'],'Cost':[12, 8 ,25]})  
df
```

```
Out[64]:
```

	Product	Cost
0	Milk	12
1	Bread	8
2	Beer	25

```
df2=pd.DataFrame(data={'Product':['Milk','Bread','Beer'],'Cost':[12,8,25]},  
                  index=['Trans1','Trans2','Trans3'])  
df2
```

	Product	Cost
Trans1	Milk	12
Trans2	Bread	8
Trans3	Beer	25