Handout 10

Introduction to Pandas data structures

Pandas library provides efficient operations on multidimensional arrays of heterogeneous data with attached row and column labels. Built on top of NumPy.
Described in https://pandas.pydata.org/pandas-docs/stable/index.html

```
import pandas
import pandas as pd
```

Main data structures are `Series` and `DataFrame`.

SERIES

One-dimensional array of *indexed* data. Indexing is not necessarily numeric; index may contain duplicate values.
https://pandas.pydata.org/pandas-docs/stable/dsintro.html#series

```
>>> s = pd.Series  ([ 7, 12, 3, 45])
>>> s  # will show index on the left and element on the right
0     7
1    12
2     3
3    45
dtype: int64

>>> type(s)
 pandas.core.series.Series

>>> s.values
 array([ 7, 12,  3, 45], dtype=int64)

>>> s.get_values()
 array([ 7, 12,  3, 45], dtype=int64)

>>> type(s.values)
 numpy.ndarray

>>> s.index
 RangeIndex(start=0, stop=4, step=1)
```

Can assign a non-numeric index:

```
  >>> si = pd.Series([210, 31, 40, 15],
                 index =['Porcupine','Chipmunk','Squirrel','Goat'])
  >>> si

  Porcupine    210
  Chipmunk      31
  Squirrel      40
  Goat          15
  dtype: int64
```

```
>>> si['Goat']
15

>>> si[0]  # can also use the numeric index!
 210
```

Series created from a dictionary:

```
>>> d = {'d': 45, 'b': 33, 'c':12, 'a': 34}
>>> d
 {'d': 45, 'b': 33, 'c': 12, 'a': 34}
>>> sd = pd.Series(d)
>>> sd
d    45
b    33
c    12
a    34
```

DATAFRAME

A two-dimensional array with labeled rows and columns.
Each column is of type Series.

CREATE DATAFRAME

There are a number of ways to create and initialize new DataFrames, for example from
    –   a file
    –   Python dictionaries, lists or tuples
    –   pd.Series, pd.DataFrame, NumPy data array

Example 1

```
>>> data
{'city': ['Boston', 'Paris', 'Lagos'],
 'area': [11700, 17174, 2706],
 'population': [4628910, 12405426, 21000000]}

>>> frame = pd.DataFrame(data)
>>> frame
    city    area  population
0  Boston  11700     4628910
1   Paris  17174    12405426
2   Lagos   2706    21000000

>>> frame = pd.DataFrame(data, columns = ['city', 'country','area','population'])
>>> frame
    city country   area  population
0  Boston     NaN  11700     4628910
1   Paris     NaN  17174    12405426
2   Lagos     NaN   2706    21000000
```

```
>>> frame.columns
 Index(['city', 'country', 'area', 'population'], dtype='object')

>>> frame.index
 RangeIndex(start=0, stop=3, step=1)
```

- Selection of rows.

  ```
  >>> frame.loc[1] # there will be more on loc and iloc attributes later
  city            Paris
  country           NaN
  area            17174
  population   12405426
  Name: 1, dtype: object

  >>> type(frame.loc[1])
  pandas.core.series.Series
  ```

- Selection of columns. Each column is a pd.Series:
  ```
  >>> frame['city']
  0    Boston
  1     Paris
  2     Lagos
  Name: city, dtype: object

  >>> type(frame['city'])
   pandas.core.series.Series
  >>> frame.city
  0    Boston
  1     Paris
  2     Lagos
  Name: city, dtype: object
  ```

- Assigning column values
  ```
  >>> frame.country = 'USA'
  >>> frame
       city country   area  population
  0  Boston     USA  11700     4628910
  1   Paris     USA  17174    12405426
  2   Lagos     USA   2706    21000000

  >>> frame.country = ['USA', 'France', 'Nigeria']
  >>> frame
       city  country   area  population
  0  Boston      USA  11700     4628910
  1   Paris   France  17174    12405426
  2   Lagos  Nigeria   2706    21000000
  ```

- Assign data
  ```
  >>> frame.area =  pd.Series ([30000000, 5000000], index = [2,3])
  >>> frame.area
  0          NaN
  1          NaN
  2   30000000.0
  ```

```
   Name: area, dtype: float64
   >>> frame
        city   country        area  population
   0  Boston      USA         NaN     4628910
   1   Paris   France         NaN    12405426
   2   Lagos  Nigeria  30000000.0    21000000
```

- Add a column:
```
>>> frame['mega'] =  (frame['population'] >= 10**7)
>>> frame
     city   country        area  population   mega
0  Boston      USA         NaN     4628910  False
1   Paris   France         NaN    12405426   True
2   Lagos  Nigeria  30000000.0    21000000   True
```

- Add a row : done by appending a dictionary, a DataFrame or Series object
```
>>> frame = frame.append({'city':'Yerevan', 'country': 'Armenia', 'population':1075000},
                          ignore_index  = True)
>>> frame

       city   country        area  population  mega
0    Boston      USA         NaN     4628910   0.0
1     Paris   France         NaN    12405426   1.0
2     Lagos  Nigeria  30000000.0    21000000   1.0
3   Yerevan  Armenia         NaN     1075000   NaN
```

Example 2
: suppose have the following content in `weather.xlsx`

Day	Description	High Temp	Low Temp	Precip	Humidity
1	PM Showers	50	51	50	77
2	PM Thundershowers	52	61	90	89
3	Mostly Sunny	45	60	0	55

```
>>> exdf = pd.read_excel("weather.xlsx") # also have pd.read_csv
>>> exdf
   Day        Description  High Temp  Low Temp  Precip  Humidity
0    1         PM Showers         50        51      50        77
1    2  PM Thundershowers         52        61      90        89
2    3       Mostly Sunny         45        60       0        55

>>> exdf.values
array([[1, 'PM Showers ', 50, 51, 50, 77],
       [2, 'PM Thundershowers', 52, 61, 90, 89],
       [3, 'Mostly Sunny', 45, 60, 0, 55]], dtype=object)

>>> type(exdf.values)
 numpy.ndarray

>>> exdf.values.shape
 (3, 6)
```

```
>>> exdf['Precip']
0    50
1    90
2     0
Name: Precip, dtype: int64

>>> exdf['Precip'][0]
50
```

INDEX
- Pandas Index objects hold axis labels and other metadata.
- Implemented as immutable ndarray implementing an ordered, sliceable set.
- Index can contain duplicates.

```
>>> exdf.columns
Index(['Day', 'Description', 'High Temp', 'Low Temp', 'Precip', 'Humidity '],
dtype='object')

>>> exdf.index
RangeIndex(start=0, stop=3, step=1)

>>> exdf.index = ['Mon', 'Tue', 'Wed']
>>> exdf
         Day       Description  Low Temp  High Temp  Precip  Humidity
Mon       1          PM Showers        50         51      50        77
Tue       2  PM Thundershowers        52         61      90        89
Wed       3        Mostly Sunny        45         60       0        55
```

Renaming, Adding and removing rows/columns

** Note - function descriptions below are not complete

```
dataframe.rename( index = index-dict,
              columns = column-dict,
              inplace = True )
```
index/columns param is a dictionary of form **{old:new}.** Function replaces the **old** values in the **index/columns** with the **new. inplace** controls whether the original dataframe is changed.

```
dataframe.reindex( index=None,
               columns=None,
              fill_value=NaN )
```
Returns a reindexed DataFrame. Conform DataFrame to new index with optional filling logic, placing NA/NaN in locations having no value in the previous index. A new object is produced, the original frame is unchanged

```
dataframe.drop(index=None,
               columns=None,
               inplace = False)
```
Remove rows or columns by specifying index or column names. index, columns : single label or list-like. A new object is produced, the original data frame is unchanged, **unless inplace parameter is True.**

```
Index.union(index_other)
Index.intersection(index_other)
Index.difference(index_other)
```
Form the union/intersect/difference of two Index objects and sorts if possible.

```
>>> exdf.drop(columns = ['Day'])
Out[155]:
          Description  Low Temp  High Temp  Precip  Humidity
Mon        PM Showers        50         51      50        77
Tue  PM Thundershowers       52         61      90        89
Wed      Mostly Sunny        45         60       0        55
>>> exdf
     Day        Description  Low Temp  High Temp  Precip  Humidity
Mon    1          PM Showers        50         51      50        77
Tue    2  PM Thundershowers        52         61      90        89
Wed    3        Mostly Sunny        45         60       0        55

>>> exdf.drop(columns = ['Day'], inplace=True)
>>> exdf
          Description  Low Temp  High Temp  Precip  Humidity
Mon        PM Showers        50         51      50        77
Tue  PM Thundershowers       52         61      90        89
Wed      Mostly Sunny        45         60       0        55

>>> exdf.reindex(columns = exdf.columns.union(['Wind']))
          Description  Low Temp  Humidity  High Temp  Precip  Wind
Mon        PM Showers        50        77         51      50   NaN
Tue  PM Thundershowers       52        89         61      90   NaN
Wed      Mostly Sunny        45        55         60       0   NaN

>>> exdf.rename(columns = {"Description":"Descr", "Low Temp":"LowTemp"}, inplace = True)
>>> exdf
  Day              Descr  LowTemp  High Temp  Precip  Humidity
0    1          PM Showers       50         51      50        77
1    2  PM Thundershowers       52         61      90        89
2    3        Mostly Sunny       45         60       0        55
```

Practice problem
1. Create the efdx table from the xlsx file as a pandas Dataframe.
2. Change the exdf column titles to all lower case
3. Change the index (row labels) to include the rest of the week, preserving the existing data.
4. Update the Wind column with values 45.6 , 12.5, 3
5. Rearrange the columns to list the low temp, then high temp, then other fields
6. Remove column 'Wind'
7. Add a column with the Average Temp, computed as the average of High and Low temp values
8. Create a NumPy table with low and high temperature values; compute min, max and median values in for each of the two parameters.

SELECTION, FILTERING, SORTING

'Standard' **slicing** (but `loc` and `iloc` methods are preferred, see below)

```
>>> frame
     city country   area  population
0  Boston     NaN  11700     4628910
1   Paris     NaN  17174    12405426
2   Lagos     NaN   2706    21000000

>>> frame[:2]
     city country   area  population
0  Boston     NaN  11700     4628910
1   Paris     NaN  17174    12405426


>>> frame[['city', 'area']]
     city   area
0  Boston  11700
1   Paris  17174
2   Lagos   2706

>>> frame[['city', 'area']][:2]
     city   area
0  Boston  11700
1   Paris  17174
```

Filtering by condition

```
frame[frame['area']>10000]
Out[202]:
     city country   area  population
0  Boston     NaN  11700     4628910
1   Paris     NaN  17174    12405426

frame[frame['city'] == 'Paris']
Out[206]:
    city country   area  population
1  Paris     NaN  17174    12405426
```

Sorting by index

```
>>> frame.sort_index(ascending=False, inplace = True)
>>> frame
     city   area  population
2   Lagos   2706    21000000
1   Paris  17174    12405426
0  Boston  11700     4628910
```

Sorting by value

```
>>> frame.sort_values(by=['city'])
     city   area  population
0  Boston  11700     4628910
2   Lagos   2706    21000000
1   Paris  17174    12405426
```

```
>>> frame.sort_values(by=['population'], ascending = False)
      city    area   population
2    Lagos    2706    21000000
1    Paris   17174    12405426
0   Boston   11700     4628910
```

INDEXING AND SELECTION USING LOC AND ILOC

iloc - integer based indexing

Pandas provides ways to get **purely integer based indexing**. The semantics follow closely Python and NumPy slicing. These are `0-based` indexing. When slicing, the start bounds is *included*, while the upper bound is *excluded*.

The `.iloc` attribute is the primary access method. The following are valid inputs:
- An integer e.g. `5`.
- A slice object with ints `1:7`.
- A list or array of integers `[4, 3, 0]`.
- A function with one argument that returns any of the above items

```
>>> df1 = pd.DataFrame(np.arange(1,16).reshape(5,3),  index=list('abcdf'),
columns=list('xyz'))
>>> df1
    x   y   z
a   1   2   3
b   4   5   6
c   7   8   9
d  10  11  12
f  13  14  15

>>> df1.iloc[:3, :2]
Out[215]:
    x  y
a   1  2
b   4  5
c   7  8
```

loc - label based indexing
Pandas provides access to groups of rows and columns by label(s) or a boolean array.
Allowed inputs are:
- A single label, e.g. 5 or `'a'`, (note that 5 is interpreted as a *label* of the index, and **never** as an integer position along the index).
- A list or array of labels, e.g. `['a', 'b', 'c']`.
- A slice object with labels, e.g. `'a':'f'`.

Note:
 Note that contrary to usual python slices, **both** the start and the stop are included

```
>>> df1.loc['a': 'c', 'x':'y']
    x  y
a   1  2
b   4  5
```

```
        c  7  8
        >>> df2 = pd.DataFrame(np.arange(20,35).reshape(5,3),  index=[3,4, 5,6, 7 ],
        columns=list('xyz'))
        >>> df2
            x   y   z
        3  20  21  22
        4  23  24  25
        5  26  27  28
        6  29  30  31
        7  32  33  34

        >>> df2.loc[:3]  # Note, 3 is evaluated as a label, not a 0-based index.
            x   y   z
        3  20  21  22
        >>> df2.loc[:,'x']
        Out[231]:
        3    20
        4    23
        5    26
        6    29
        7    32
        Name: x, dtype: int32
```

VECTORIZED OPERATIONS, UFUNCS, ALIGNMENT

The vectorized operations from NumPy are applicable to Series and DataFrame, e.g.

```
>>> df2.loc[:,'x']*2 + 5
Out[233]:
3    45
4    51
5    57
6    63
7    69
Name: x, dtype: int32
```

Note, that in operations involving two structures:
- operations are applied to elements with matching index values (**alignment**)
- the resulting structure contains the union of all indices
-
```
>>> s1 = pd.Series( range(10), [chr(code) for code in range (ord('a'), ord('a')+10)] )
>>> s2 = pd.Series( [20 for i in range(5)], [chr(code) for code in range (ord('c'),
ord('c')+5) ] )
>>> s3 = s1 + s2
>>> s3
Out[117]:
a    NaN
b    NaN
c    22.0
d    23.0
e    24.0
f    25.0
g    26.0
```

```
h     NaN
i     NaN
j     NaN
dtype: float64
```

In addition, there are vectorized methods for working with **text strings**.
http://pandas.pydata.org/pandas-docs/stable/text.html#text-string-methods
- • These methods exclude missing/NA values automatically.
- • They are accessed via the Series's `str` attribute and generally have names matching the analogous built-in string methods.
- • Slicing using [::] syntax is done with method `str.slice` () – works on strings AND lists
- • Indexing - `str.get()` – works on strings AND lists

Example
```
>>> exdf = pd.read_csv("weather.csv")
>>> exdf
    Day        Description  High Temp  Low Temp  Precip  Humidiy
0     1          PM Showers         50        51      50       77
1     2  PM Thundershowers         52        61      90       89
2     3        Mostly Sunny         45        60       0       55

>>> exdf['Description'].str.split()
Out[37]:
0         [PM, Showers]
1    [PM, Thundershowers]
2       [Mostly, Sunny]
Name: Description, dtype: object

>>> exdf['Description'].str.split().str.get(0)
Out[38]:
0        PM
1        PM
2    Mostly
Name: Description, dtype: object

>>> exdf['Description'].str.lower()
0          pm showers
1    pm thundershowers
2         mostly sunny

>>> exdf['DescrAbbrev'] = exdf['Description'].str.slice(0,5)
>>> exdf
    Day        Description  High Temp   ...    Precip  Humidiy   DescrAbbrev
0     1          PM Showers         50   ...        50       77         PM Sh
1     2  PM Thundershowers         52   ...        90       89         PM Th
2     3        Mostly Sunny         45   ...         0       55         Mostl
```

AGGREGATION

Recall that `DataFrame.values` produces a NumPy table. Hence, all aggregation methods and statistical functions that work for NumPy can be applied to `DataFrame.values`.

Also, the following Pandas aggregate functions are applicable to Series and DataFrames.

Function	Description
count	Number of non-NA observations
sum	Sum of values
mean	Mean of values
mad	Mean absolute deviation
median	Arithmetic median of values
min	Minimum
max	Maximum
mode	Mode
abs	Absolute Value
prod	Product of values
std	Bessel-corrected sample standard deviation
var	Unbiased variance
quantile	Sample quantile (value at %)

```
>>> frame.sum()
city          LagosParisBoston
area                     31580
population            38034336
dtype: object

>>> round(frame.mean(), 2)
area            10526.67
population   12678112.00
dtype: float64

>>> round(frame.std(), 2)
area             7305.02
population    8188950.80
dtype: float64
```

Practice problem: based on the data contained in the IMDB.csv and IMDB-dataDict.txt, compose python code using Panda's package for the following tasks.

1. Create a DataFrame containing data from the first columns up to and including Runtime.
2. Find how many movies are in the data set.
3. Select data for those movies that run for less than 2 hours
4. Find the most expensive and least expensive film, the mean and standard deviation of the budget.
5. Find the 10 top rated movies
6. Find out the range of years for the movies listed in the data set
7. List 10 other questions to answer based on this data and create solutions for these queries.

References

- Pandas documentation http://pandas.pydata.org/pandas-docs/stable/
- Series https://pandas.pydata.org/pandas-docs/stable/dsintro.html#series
- DataFrame https://pandas.pydata.org/pandas-docs/stable/dsintro.html#dataframe
- Complete list of Pandas functions https://pandas.pydata.org/pandas-docs/stable/api.html
- Unicode HOWTO https://docs.python.org/3/howto/unicode.html