

Takenmind Assignment #1

Study the Complete Numpy and Pandas Lectures (Section 2 and 3 of intern-kit) and make a documentation in less than 500 words (Word limit is excluding the codes typed.) in a word document. A documentation can have codes, explanations and logical flowcharts.

Numpy Package

For scientific computing that provides n dimensional array class, tools and useful math operations e.g. linear algebra, Fourier transform... their objects are used on scipy, pandas and other packages.

numpy.array[<list>] <- Create array multidimensional from hardcoded data.

<object>.shape <- Dimensionality

.dtype <- data structure type.

.min() <- Mini.value

.max() <- Max.value

.copy() <- Generate a new data space variable.0

Method utils:

numpy.zeros(<dimensions>)

.ones(<dimensions>)

.empty(<dimensions>) <- Undefined Array

.eye(<dimensions>) <-Identity matrix. Diagonal.

Scalation and new version compatibility: $5/2 = 2.5$!!-> **from __future__ import division** -> $5/2 = 2.5$

Array Operations:

Assigination

Add/Substract

Multiplication

Slicing [:] and slice assigination, with tied memory space between identifier's data. Select rows, cols, dimensions, elements.

Copying arrays.

Loops along arrays. Utils:

`arr_rows = arr2d.shape [0] -> len(arr2d.shape)[0]`

`arr_cols = arr2d.shape [1] -> len(arr2d.shape)[0]`

Example for area access: `arr2d[1:,1:]`

Universal numpy functions

numpy.arange(<start>, <end>, <step>)

.add (<arrayA>, <arrayB>, ...)

.sqrt(<array>)

.maximum/minimum(<arrayA>, <arrayB>)

.sum() -> Sum of element values.

.mean()

.std()

.var()

.sort()

Other functions associated to numpy arrays on scripy.org

Save/load large data to/from numpy arrays

Save single array: **np.save**('<file_array>', <array>)

Load single array: **np.load**('<file_array>.npz')

Save several arrays: **np.savez**('<file_arrays>', x = <arrayA>, y = <arrayB>,...)

or **np.savez_compressed**('<file_arrays>', x = <arrayA>, y = <arrayB>,...)

Load several arrays: **arrays = np.load**('<file_arrays>.npz') ; **array['x']** and **array['y']** to fetch.

Save to text file: **np.savetxt**('<txt_array>', <array>, delimiter = {, ; : .})

Load from text file: **np.loadtxt**('<txt_array>', delimiter = {, ; : .}) **Note:** converts integer to float.

Conditional clause and boolean operations with numpy

Conditions and loops enclosed by []

np.where(<condition>, <valueYes>, <valueNo>)

Logic arrays True/False Values: **.all**() is AND, **.any**() is OR.

Inclusion: **numpy.in1d**(<array>, <array_numpy>) returns contained values.

Matplotlib/pyplot package

Matplotlib is an object oriented plotting library, generally applied to mathematical/statistical plot graphs.

- Define axes values with numpy array
- Examples with two variable functions: **np.meshgrid** and **np.cos**
- Define function values; print values.
- Plot values: **matplotlib.pyplot.imshow**(<function>)
- Add title; **matplotlib.pyplot.title**('<title>')
- Add scale bar: **matplotlib.pyplot.colorbar**()
- Save: **matplotlib.pyplot.savefig**('<image_file>')

Panda's to analyze data

Panda's library is specific to analyze data and perform math operations on datasets.

Panda's provides powerful structures for data analytics, time series, and statistics about.

Add package pandas: Settings ->Project Interpreter -> + -> Add 'pandas'. Example applications:

```
import pandas as pd
```

```
import numpy as np
```

```
from pandas import Series.
```

```
object = Series([ 5, 10, 15, 20 ])
```

```
print object -> index + data
```

```
object.index -> [ 0,1,2,3 ] = RangeIndex ( start=0, stop = 4, step = 1 )
```

```
object.values -> [ 5, 10, 15, 20 ]
```

- Numpy arrays to series

```
data_array = np.array ( [ 'a', 'b', 'c', 'd' ] )
```

```
s = Series (data_array)
```

```
object.index -> [ 0,1,2,3 ] = RangeIndex ( start=0, stop = 4, step = 1 )
```

```
object.values -> [ 'a', 'b', 'c', 'd' ]
```

- Custom index on series

```
s = Series( data_array, index = [ 'id1', 'id2', 'id3', 'id4' ] )
```

- Using real life example

```
revenue = Series([20, 80, 40, 35 ], index = [ 'ola', 'uber', 'grab', 'gojek' ])
```

```
revenue[ 'uber' ] -> 80
```

Boolean conditions:

```
revenue[ revenue >= 35 ] -> gojek, grab, uber
```

```
'ola' in revenue -> True
```

Convert to dictionary: **revenue_dict = revenue.to_dict**

NaN values:

```
index2 = [ 'ola', 'uber', 'grab', 'gojek', 'lyft' ]
```

```
revenue2 = Series( [ revenue, index2] ) -> lyft - NaN
```

```
pd.isnull( revenue2 ) -> True for NaN
```

```
pd.notnull( revenue2 ) -> not True for NaN
```

- Addition of series

```
add_revenues = revenue + revenue2 -> sum of values for each index class.
```

- Assigning names

```
revenue2.name = 'Co. revenues'
```

```
revenue2.index.name = 'Co. name'
```

- Dataframes (pandas.pydata.org). Is a functionality to analyze data, simulate a matrix with rows and cols along an index of rows created in addition.

import numpy as np

import pandas as pd

from pandas import Series, Dataframe

Exercise: from wikipedia find list of largest companies by revenue:

(https://en.wikipedia.org/wiki/List_of_largest_companies_by_revenue)

copy first six rows of list, included head. After load with pandas:

revenue_df = pd.read_clipboard()

print revenue_df

index and columns

print revenue_df.columns

print revenue_df ['Rank'] <- see data of column associated by label 'Rank'

print Dataframe(revenue_df = [<array of label columns included>]

- Example to create a new Dataframe object

new_df = Dataframe (revenue_df, columns = ['Rank', 'Name',...]

NaN values. Create new label 'Profit' without data at new Dataframe.

DataFrame_df2 = (revenue_df, columns=['Rank', 'Name', 'Profit',...]) <-New column profit with NaN.

print revenue_df2

head and tail: first and last rows.

revenue_df.head(2) -> index 0, 1 ; first two rows.

revenue_df.tail(2) -> index (n-1), n ; last two rows.

access rows in Dataframe

revenue_df.ix[0] -> first row

assign values to Dataframe. Two methods: from numpy, or from Series.

array1 = np.array([1, 2, 3, 4, 5, 6]) -> Create numpy array.

revenue_df2 ['Profit'] = array1 -> Assign values from numpy array,

profits = Series([900, 1000], index = [3, 5]) -> Create one pandas series

revenue_df2 ['Profit'] = profits -> all values NaN except index 3 and index 5 that was assigned.

Deletion of columns at Dataframe object

del revenue_df2 ['Profit'] -> Erase 'Profit' column.

- Dictionary functions to Dataframe:

sample = { 'Company': [A, B], 'Profit': [1000, 5000] }

sample_df = Dataframe (sample) -> enables Dataframe with indexes 0, 1; and Profit, Company data.

- Index objects

Performing index operations, as a set of series or Dataframes. Index as a label, but returns an array with positional 'u' labels. Then that indexes can be called by position.

```
series = Series([ 10, 20, 30, 40 ], index = list( 'abcd' ))
```

```
index1 = series.index -> index1[ 2] is 'c'
```

```
# negative indexes: index1[ -2: ] last two elements 'c', 'd' ; index1[ :-2 ] first two elements 'a', 'b'
```

```
# range of indexes: index[ 2:4] -> 'c', 'd'
```

Note: Indexes can't be modified, there are not mutable data.

- Reindexing methods

How can be reindex indexes and columns on Series and DataFrames.

```
from numpy.random import randn
```

```
series1 = Series( [ 1, 2, 3, 4 ], index = list( 'efgh' ))
```

```
# Creating new series with reindex.
```

```
series1 = series1.reindex( list( 'efghi' )) -> 'i' index doesn't have defined value (NaN).
```

```
# Using fill_value: series2 = series1.reindex( list( 'efghijk' ), fill_value = 10) -> 'j', 'k' values are ten.
```

```
# Using reindex methods: ffill
```

```
cars = Series([ 'Audi', 'Merc', 'BMW'], index = [0, 4, 8])
```

```
ranger = range(13) -> [ 0, 1, 2, ..., 11, 12 ]
```

```
cars.reindex( ranger, method = "ffill" ) -> Forward fill -> 0-3 is 'Audi', 4-7 is 'Merc', 8-12 is 'BMW'
```

```
# Create new dataframe using randn:
```

```
df1 = DataFrame (randn(25).reshape(5,5), index = list('abcde'), columns = [ 'c1', 'c2', 'c3', 'c4', c5])
```

```
# create new row f index: df2 = df1.reindex(list('abcdef'))
```

```
# create new col c6 index: df3 = df2.reindex(columns = ['c1', 'c2', 'c3', 'c4', 'c5', 'c6'])
```

```
# use .ix[] to reindex rows/colsw with one sentence
```

```
df4 = df1( list('abcdef' ), ['c1', 'c2', 'c3', 'c4', 'c5', 'c6'])
```

- Dropping entries from datatypes

```
#Series
```

```
cars = series(['BMW', 'Audi', 'Merc'], index = list('abc'))
```

```
cars = cars.drop('a') -> removes 'a' row.
```

```
# DataFrame
```

```
cars_df = DataFrame(np.arange(9).reshape(3,3), index =(['BMW', 'Audi', 'Merc'], index = ['a', 'b', 'c']),  
                    columns = ['revenue', 'profit', 'expenses'])
```

```
cars_df = cars.df.drop( 'BMW' ) -> Optional default parameter axis=0. Removes row with index 'BMW'
```

```
cars_df = cars.df.drop('profit', axis = 1 ) -> remove profit column.
```

- Handling null data: create series or dataframes with `np.nan = NaN`

series1.isnull() -> True if NaN

series1.dropna() -> Optional parameter axis: 0 rows, 1, columns. Remove indexes or columns that contains NaN

df1.dropna() -> The same for DataFrame.. parameters how = "all", thresh = n with max number of NaN to remove

fillna [0] fills NaN with values ; `fillna[{0:0, 1:50, 2:100, 3:200}]`

- Selecting and modifying data: `series(<labels>)`

Conditional indexes: examples: `series1[series1>30]` or `series1[series1 == 300]`

- Using DataFrame `df` and accessing

`df > 5` -> True or false depends of this condition

`df.ix['bike']` -> column 'bike' or `df.ix[1]` -> column n°1

- Data alignment

Sum of series **ser_a + ser_b** -> NaN if not defined. Same for subtraction **ser_a - ser_b**

- Sorting/Ranking Series and Dataframes.

Sort by index: **ser1.sort_index()**

Sort by values: **ser1.sort_values()**

Rank: meet position rank of values **ser1.rank()**

- Statistics& graph sketches with pandas

df.sum(axis = 0) -> sums along each columns (NaN is 0)

df.sum(axis = 1) -> sums along each row (NaN is 0)

df.min() **df.max()** -> minimum and maximum values

df.idxmax() index of row where value is max. Same for min.

df.cumsum() -> cumulative sum.

df.describe() -> statistical distribution values.

ser1.unique -> Number of unique values on Series.

ser1.value_counts -> frequency of presentation for values.