Scientific Programming Practical 8

Introduction

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Pandas

Pandas (**panel-data**) is a very efficient library to deal with **numerical tables** and time series

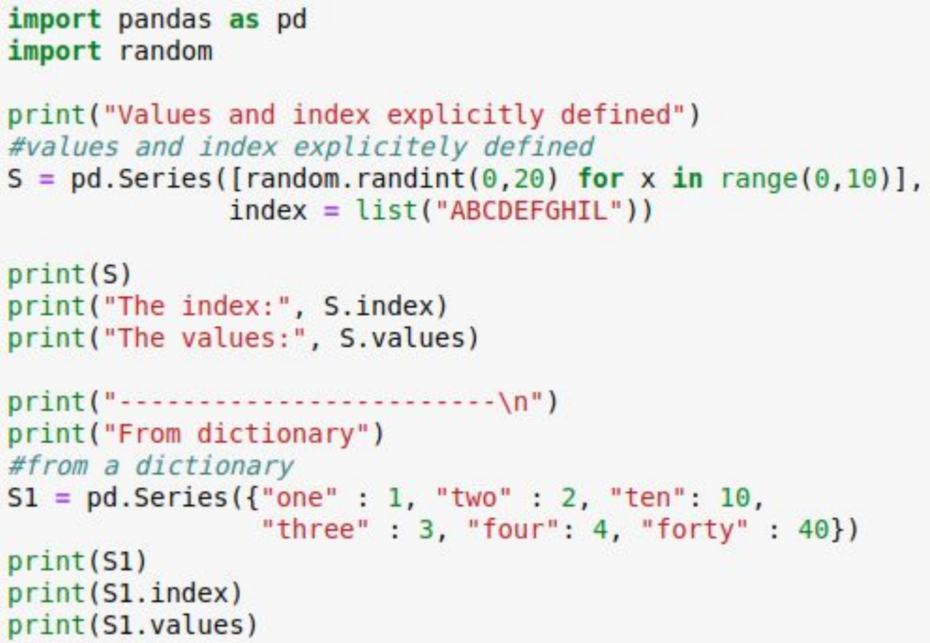
Two data structures:

Series: 1D tables
DataFrames: 2D tables

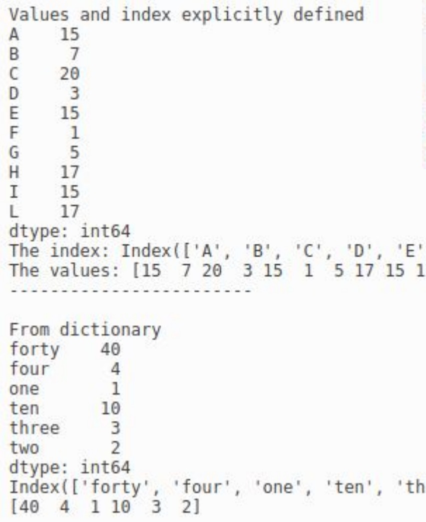
https://pandas.pydata.org/

Series

Series are 1-dimensional structures (like lists) containing data. Series are characterized by two types of information: the **values** and the **index** (a list of labels associated to the data). A bit like **list** and a bit like **dictionary**!



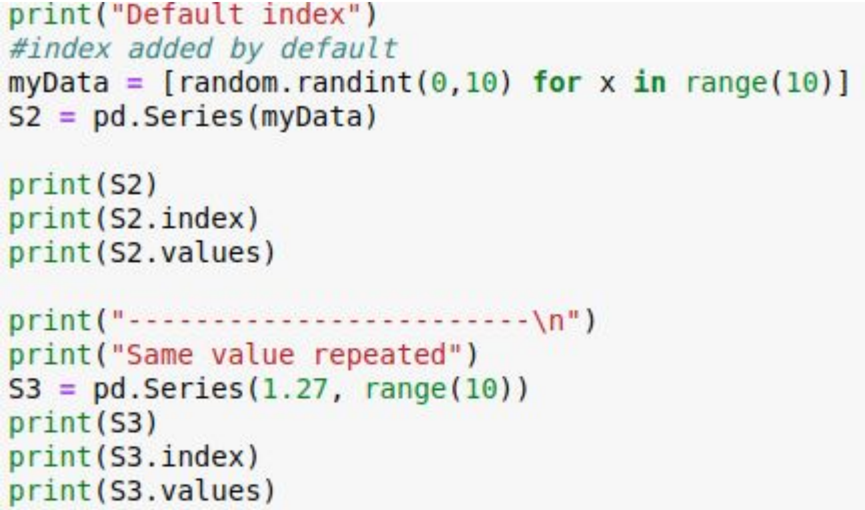
```python
import pandas as pd
import random

print("Values and index explicitly defined")
#values and index explicitely defined
S = pd.Series([random.randint(0,20) for x in range(0,10)],
              index = list("ABCDEFGHIL"))

print(S)
print("The index:", S.index)
print("The values:", S.values)

print("------------------------\n")
print("From dictionary")
#from a dictionary
S1 = pd.Series({"one" : 1, "two" : 2, "ten": 10,
                "three" : 3, "four": 4, "forty" : 40})
print(S1)
print(S1.index)
print(S1.values)
```



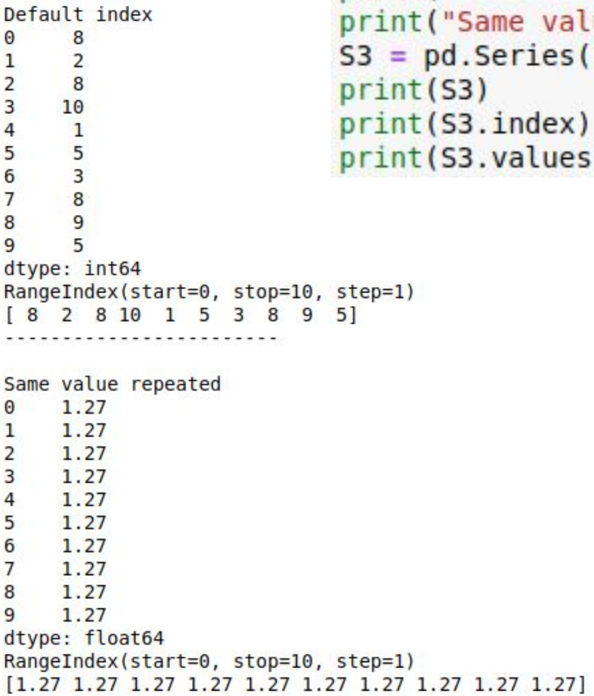
```
Values and index explicitly defined
A    15
B     7
C    20
D     3
E    15
F     1
G     5
H    17
I    15
L    17
dtype: int64
The index: Index(['A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'L'], dtype='object')
The values: [15  7 20  3 15  1  5 17 15 17]
------------------------

From dictionary
forty    40
four      4
one       1
ten      10
three     3
two       2
dtype: int64
Index(['forty', 'four', 'one', 'ten', 'three', 'two'], dtype='object')
[40  4  1 10  3  2]
```

Series

Series are 1-dimensional structures (like lists) containing data. Series are characterized by two types of information: the **values** and the **index** (a list of labels associated to the data). A bit like **list** and a bit like **dictionary**!



```python
print("Default index")
#index added by default
myData = [random.randint(0,10) for x in range(10)]
S2 = pd.Series(myData)

print(S2)
print(S2.index)
print(S2.values)

print("-----------------------\n")
print("Same value repeated")
S3 = pd.Series(1.27, range(10))
print(S3)
print(S3.index)
print(S3.values)
```

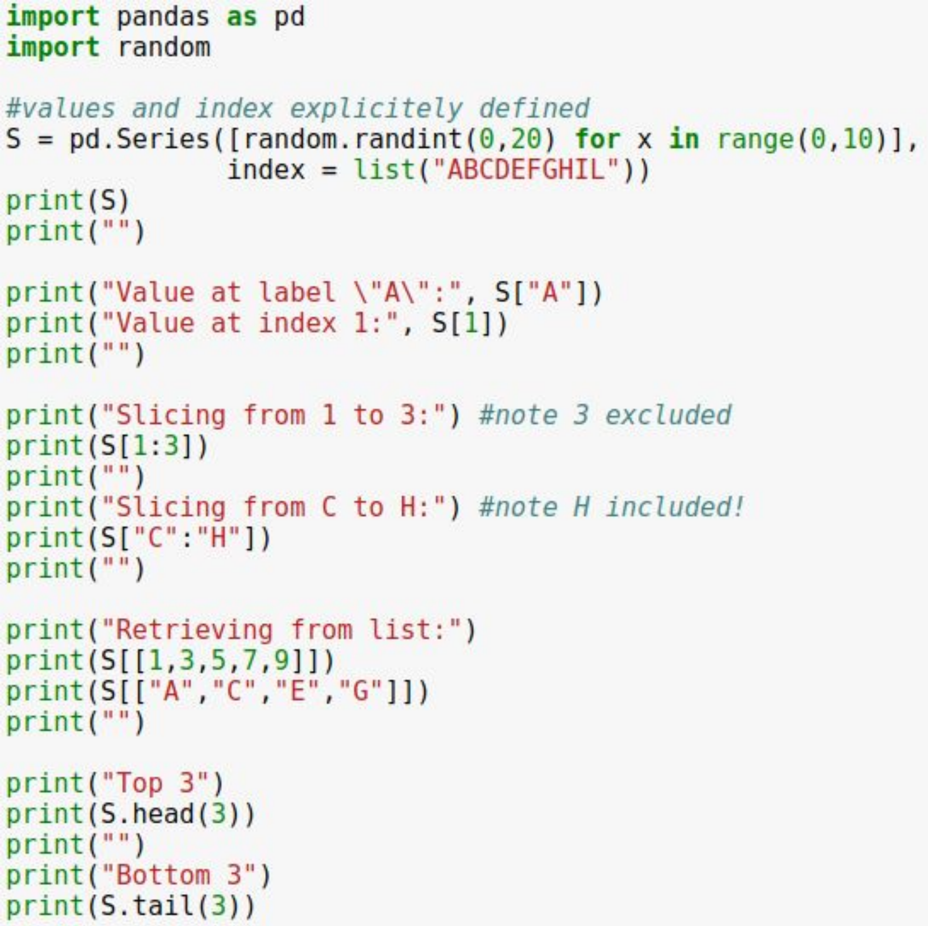


```
Default index
0     8
1     2
2     8
3    10
4     1
5     5
6     3
7     8
8     9
9     5
dtype: int64
RangeIndex(start=0, stop=10, step=1)
[ 8  2  8 10  1  5  3  8  9  5]
------------------------

Same value repeated
0    1.27
1    1.27
2    1.27
3    1.27
4    1.27
5    1.27
6    1.27
7    1.27
8    1.27
9    1.27
dtype: float64
RangeIndex(start=0, stop=10, step=1)
[1.27 1.27 1.27 1.27 1.27 1.27 1.27 1.27 1.27 1.27]
```

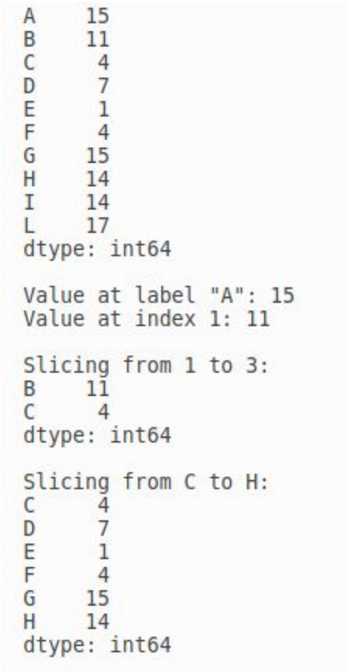
Series

Data in a series can be accessed by using the **label** (i.e. the index) as in a dictionary or through its **position** as in a list. Slicing is also allowed both by **position** and **index**.

In the latter case, `Series[S:E]` with **S and E indexes**, both **S and E are included**.



```python
import pandas as pd
import random

#values and index explicitely defined
S = pd.Series([random.randint(0,20) for x in range(0,10)],
              index = list("ABCDEFGHIL"))
print(S)
print("")

print("Value at label \"A\":", S["A"])
print("Value at index 1:", S[1])
print("")

print("Slicing from 1 to 3:") #note 3 excluded
print(S[1:3])
print("")
print("Slicing from C to H:") #note H included!
print(S["C":"H"])
print("")

print("Retrieving from list:")
print(S[[1,3,5,7,9]])
print(S[["A","C","E","G"]])
print("")

print("Top 3")
print(S.head(3))
print("")
print("Bottom 3")
print(S.tail(3))
```

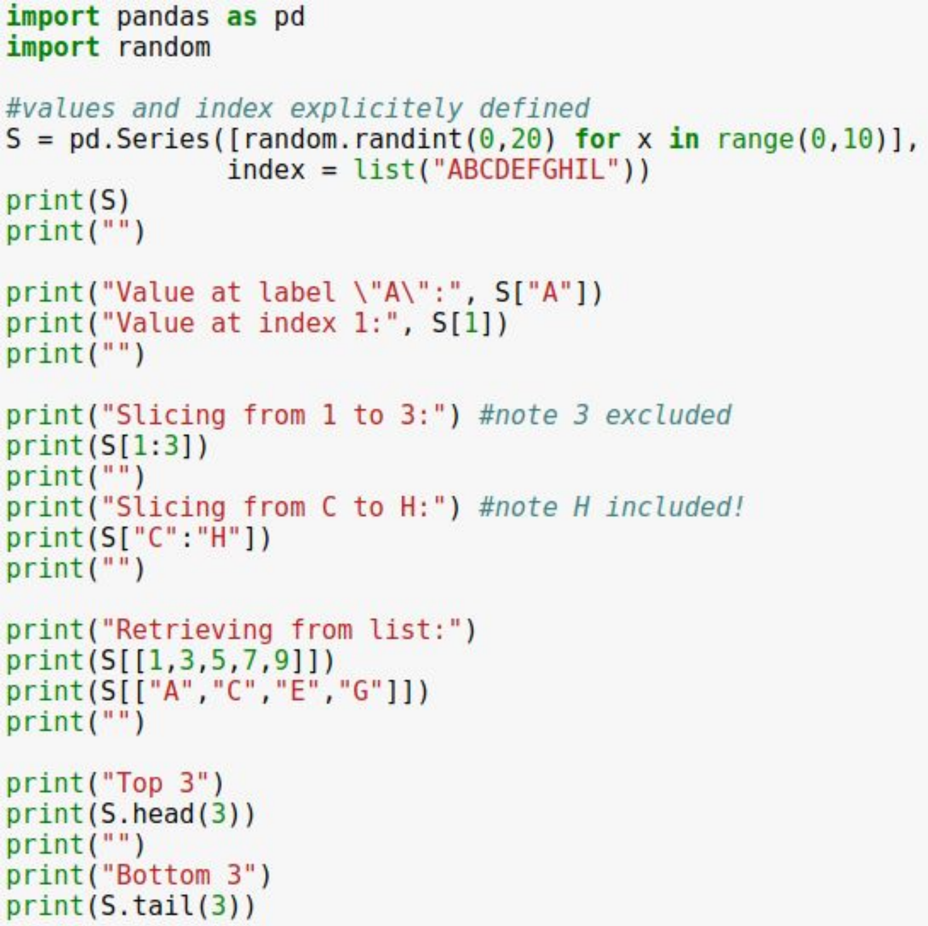


```
A    15
B    11
C     4
D     7
E     1
F     4
G    15
H    14
I    14
L    17
dtype: int64

Value at label "A": 15
Value at index 1: 11

Slicing from 1 to 3:
B    11
C     4
dtype: int64

Slicing from C to H:
C     4
D     7
E     1
F     4
G    15
H    14
dtype: int64
```

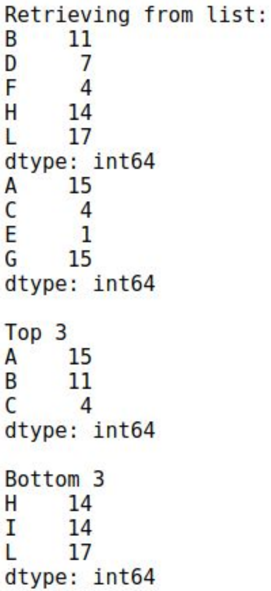
Series

Data in a series can be accessed by using the **label** (i.e. the index) as in a dictionary or through its **position** as in a list. Slicing is also allowed both by **position** and **index**.

In the latter case, `Series[S:E]` with **S and E labels**, both **S and E are included**.



```python
import pandas as pd
import random

#values and index explicitely defined
S = pd.Series([random.randint(0,20) for x in range(0,10)],
              index = list("ABCDEFGHIL"))
print(S)
print("")

print("Value at label \"A\":", S["A"])
print("Value at index 1:", S[1])
print("")

print("Slicing from 1 to 3:") #note 3 excluded
print(S[1:3])
print("")
print("Slicing from C to H:") #note H included!
print(S["C":"H"])
print("")

print("Retrieving from list:")
print(S[[1,3,5,7,9]])
print(S[["A","C","E","G"]])
print("")

print("Top 3")
print(S.head(3))
print("")
print("Bottom 3")
print(S.tail(3))
```



```
Retrieving from list:
B    11
D     7
F     4
H    14
L    17
dtype: int64
A    15
C     4
E     1
G    15
dtype: int64

Top 3
A    15
B    11
C     4
dtype: int64

Bottom 3
H    14
I    14
L    17
dtype: int64
```

Series

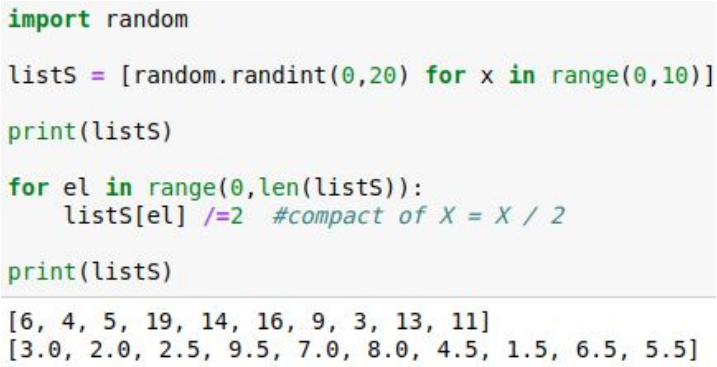
Important operations on series:

Operator broadcasting

Operations can automatically be broadcast to the entire Series. This is a quite cool feature and **saves us from looping through the elements of the Series**.

Example: Given a list of 10 integers and we want to divide them by 2.

Without pandas:



```
import random

listS = [random.randint(0,20) for x in range(0,10)]

print(listS)

for el in range(0,len(listS)):
    listS[el] /=2  #compact of X = X / 2

print(listS)
```

```
[6, 4, 5, 19, 14, 16, 9, 3, 13, 11]
[3.0, 2.0, 2.5, 9.5, 7.0, 8.0, 4.5, 1.5, 6.5, 5.5]
```

Series

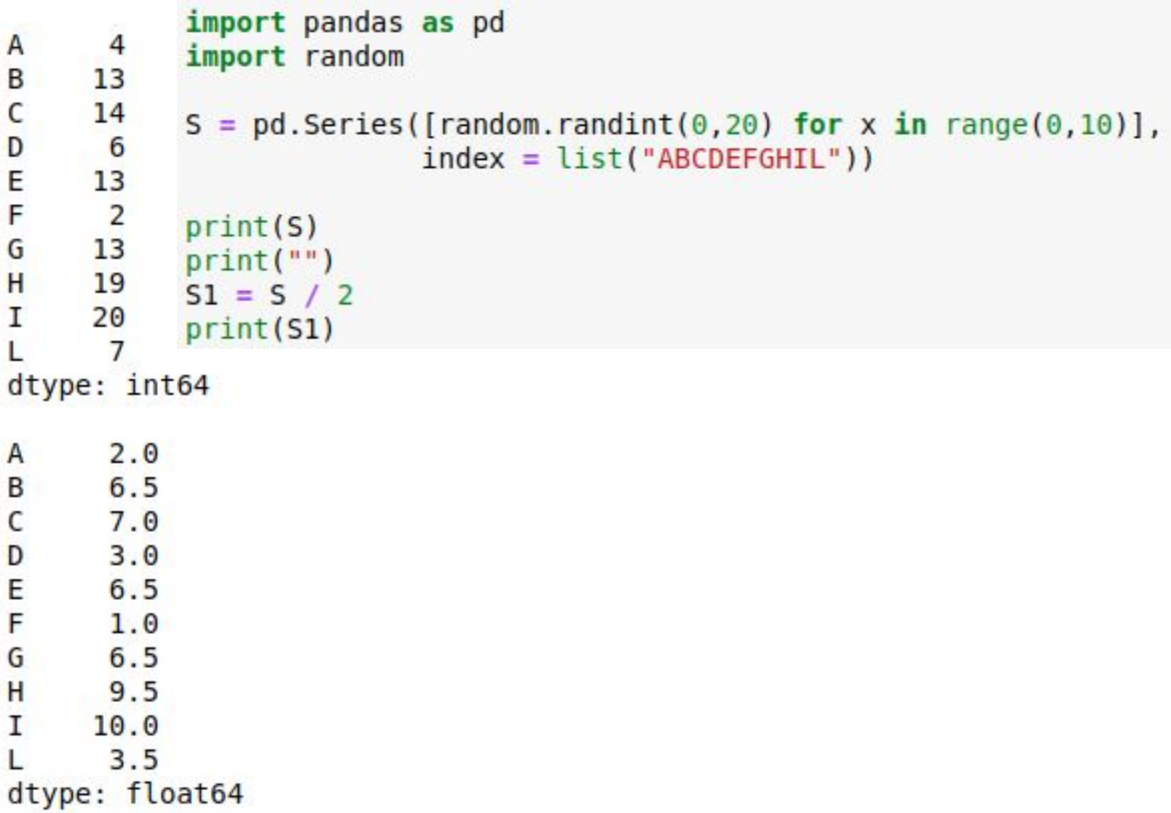
Example: Given a list of 10 integers and we want to divide them by 2.

Important operations on series:

Operator broadcasting

Operations can automatically be broadcast to the entire Series. This is a quite cool feature and **saves us from looping through the elements of the Series**.

With pandas (operator broadcasting):



```
import pandas as pd
import random

S = pd.Series([random.randint(0,20) for x in range(0,10)],
                   index = list("ABCDEFGHIL"))


print(S)
print("")
S1 = S / 2
print(S1)
```

```
A     4
B    13
C    14
D     6
E    13
F     2
G    13
H    19
I    20
L     7
dtype: int64

A     2.0
B     6.5
C     7.0
D     3.0
E     6.5
F     1.0
G     6.5
H     9.5
I    10.0
L     3.5
dtype: float64
```

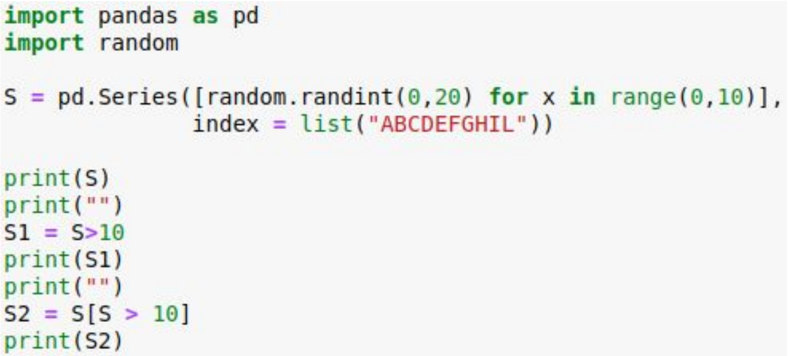
Series

Important operations on series:

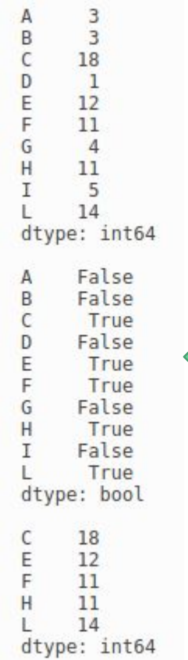
Operator broadcasting

Filtering

We can also apply boolean operators to obtain only the **sub-Series** with all the values satisfying a specific condition. This allows us to **filter** the Series.



```
import pandas as pd
import random

S = pd.Series([random.randint(0,20) for x in range(0,10)],
              index = list("ABCDEFGHIL"))

print(S)
print("")
S1 = S>10
print(S1)
print("")
S2 = S[S > 10]
print(S2)
```



```
A     3
B     3
C    18
D     1
E    12
F    11
G     4
H    11
I     5
L    14
dtype: int64

A    False
B    False
C     True
D    False
E     True
F     True
G    False
H     True
I    False
L     True
dtype: bool

C    18
E    12
F    11
H    11
L    14
dtype: int64
```

series of True and False where condition is/is not met

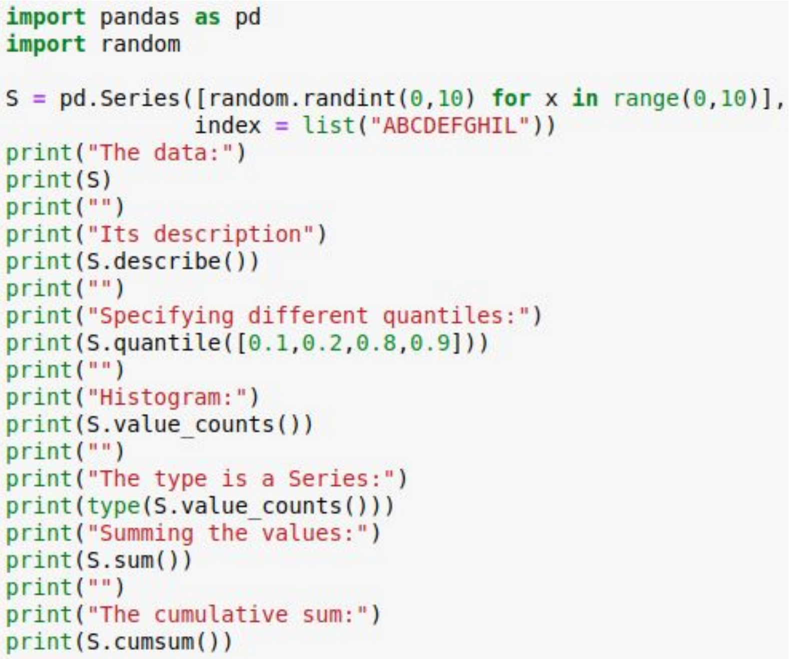
Series

Important operations on series:

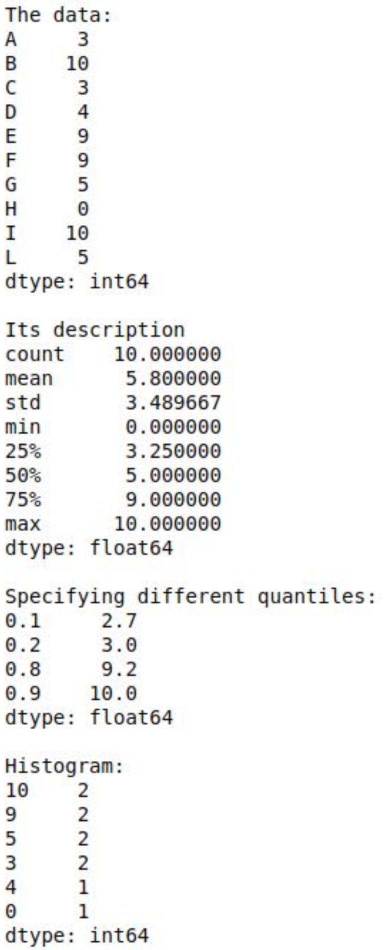
Operator broadcasting

Filtering

Computing stats



```
import pandas as pd
import random

S = pd.Series([random.randint(0,10) for x in range(0,10)],
              index = list("ABCDEFGHIL"))
print("The data:")
print(S)
print("")
print("Its description")
print(S.describe())
print("")
print("Specifying different quantiles:")
print(S.quantile([0.1,0.2,0.8,0.9]))
print("")
print("Histogram:")
print(S.value_counts())
print("")
print("The type is a Series:")
print(type(S.value_counts()))
print("Summing the values:")
print(S.sum())
print("")
print("The cumulative sum:")
print(S.cumsum())
```

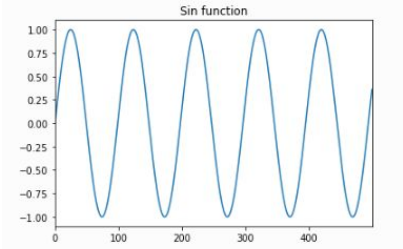


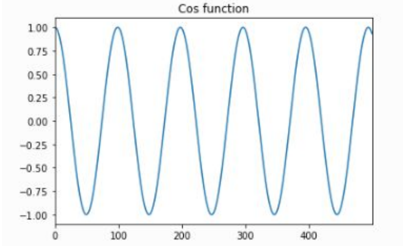
```
The data:
A     3
B    10
C     3
D     4
E     9
F     9
G     5
H     0
I    10
L     5
dtype: int64

Its description
count    10.000000
mean      5.800000
std       3.489667
min       0.000000
25%       3.250000
50%       5.000000
75%       9.000000
max      10.000000
dtype: float64

Specifying different quantiles:
0.1     2.7
0.2     3.0
0.8     9.2
0.9    10.0
dtype: float64

Histogram:
10    2
9     2
5     2
3     2
4     1
0     1
dtype: int64
```

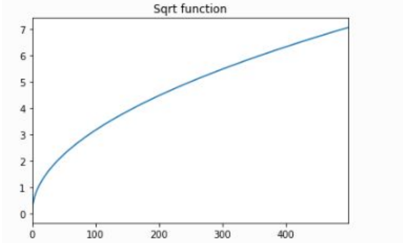
see notes for the complete results and other features like `Series.fillna(values)`

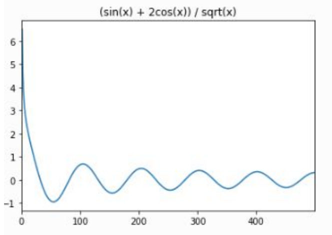
Plotting data

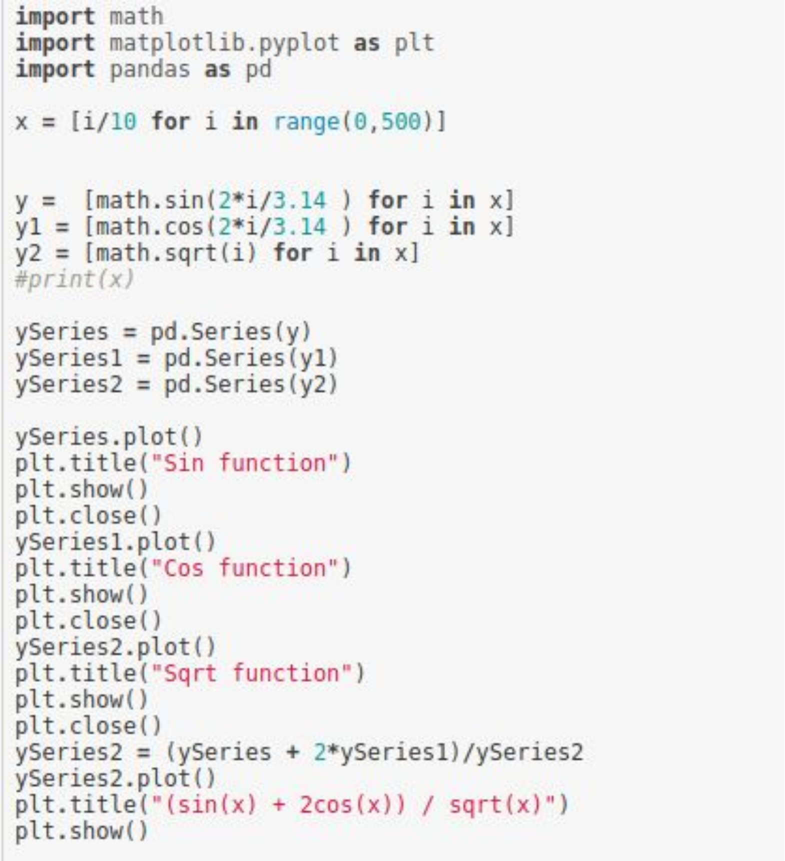
It is quite easy to plot data in Series and DataFrames thanks to matplotlib











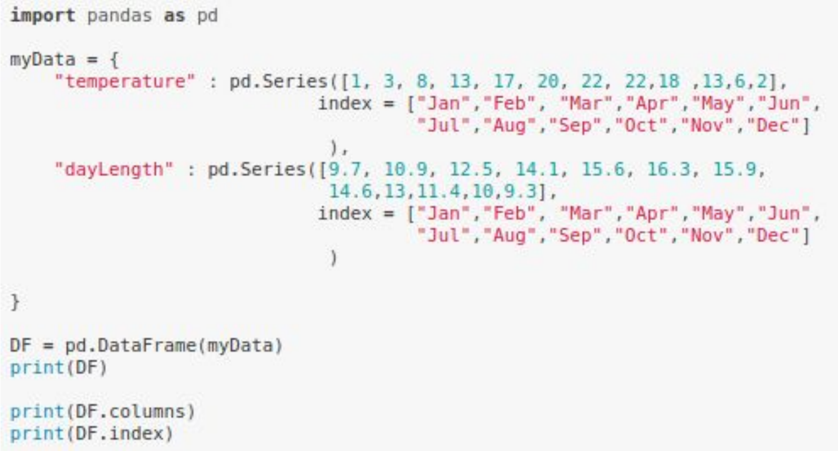
```python
import math
import matplotlib.pyplot as plt
import pandas as pd

x = [i/10 for i in range(0,500)]


y =  [math.sin(2*i/3.14 ) for i in x]
y1 = [math.cos(2*i/3.14 ) for i in x]
y2 = [math.sqrt(i) for i in x]
#print(x)

ySeries = pd.Series(y)
ySeries1 = pd.Series(y1)
ySeries2 = pd.Series(y2)

ySeries.plot()
plt.title("Sin function")
plt.show()
plt.close()
ySeries1.plot()
plt.title("Cos function")
plt.show()
plt.close()
ySeries2.plot()
plt.title("Sqrt function")
plt.show()
plt.close()
ySeries2 = (ySeries + 2*ySeries1)/ySeries2
ySeries2.plot()
plt.title("(sin(x) + 2cos(x)) / sqrt(x)")
plt.show()
```

DataFrames

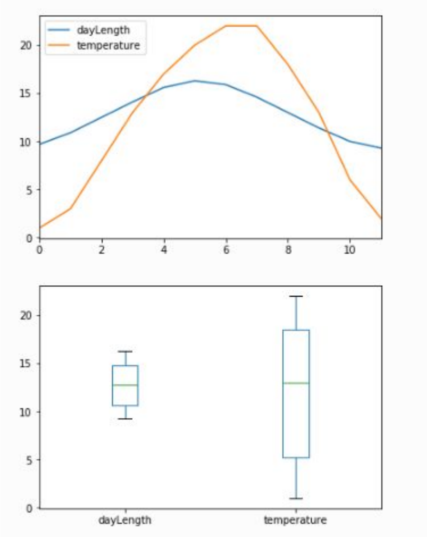
2D analogous of Series.
They have an **index** and several **columns**.

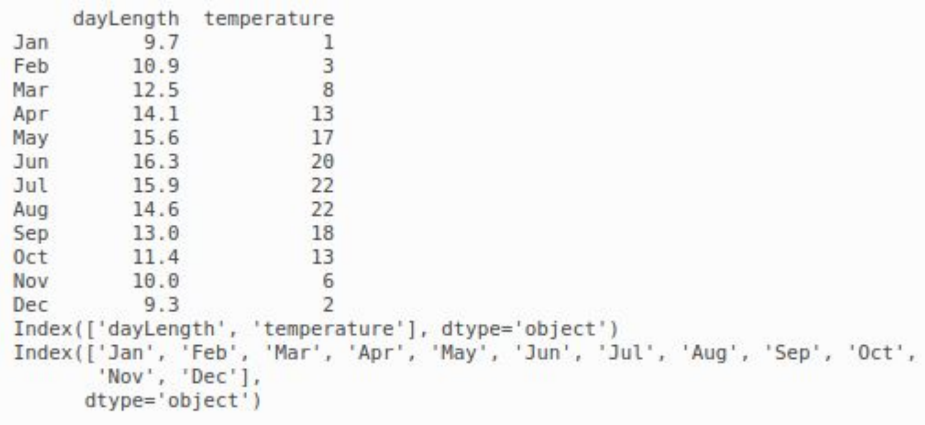
Data can be dishomogeneous.

Most of the the things seen for Series apply to DataFrames



```
import pandas as pd

myData = {
    "temperature" : pd.Series([1, 3, 8, 13, 17, 20, 22, 22,18 ,13,6,2],
                             index = ["Jan","Feb", "Mar","Apr","May","Jun",
                                      "Jul","Aug","Sep","Oct","Nov","Dec"]
                             ),
    "dayLength" : pd.Series([9.7, 10.9, 12.5, 14.1, 15.6, 16.3, 15.9,
                             14.6,13,11.4,10,9.3],
                             index = ["Jan","Feb", "Mar","Apr","May","Jun",
                                      "Jul","Aug","Sep","Oct","Nov","Dec"]
                             )

}

DF = pd.DataFrame(myData)
print(DF)

print(DF.columns)
print(DF.index)
```



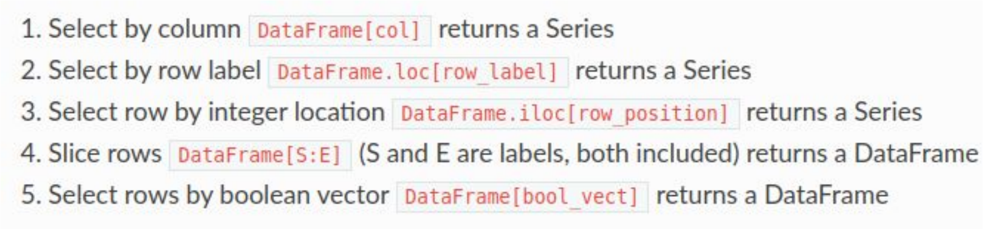


```
     dayLength  temperature
Jan        9.7            1
Feb       10.9            3
Mar       12.5            8
Apr       14.1           13
May       15.6           17
Jun       16.3           20
Jul       15.9           22
Aug       14.6           22
Sep       13.0           18
Oct       11.4           13
Nov       10.0            6
Dec        9.3            2
Index(['dayLength', 'temperature'], dtype='object')
Index(['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug', 'Sep', 'Oct',
       'Nov', 'Dec'],
      dtype='object')
```

DataFrames

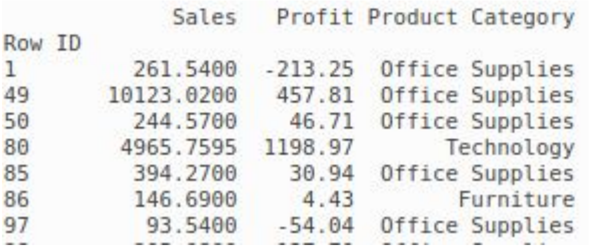
Load from file

We can **load external files**, extract info and apply operators, broadcasting and filtering...



1. Select by column `DataFrame[col]` returns a Series
2. Select by row label `DataFrame.loc[row_label]` returns a Series
3. Select row by integer location `DataFrame.iloc[row_position]` returns a Series
4. Slice rows `DataFrame[S:E]` (S and E are labels, both included) returns a DataFrame
5. Select rows by boolean vector `DataFrame[bool_vect]` returns a DataFrame

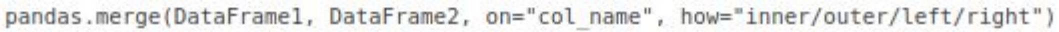
```
import pandas as pd

orders = pd.read_csv("file_samples/sampledata_orders.csv", sep=",",
index_col =0, header=0)

print("The Order Quantity column (top 5)")
print(orders["Order Quantity"].head(5))
print("")
print("The Sales column (top 10)")
print(orders.Sales.head(10))
print("")
print("The row with ID:50")
r50 = orders.loc[50]
print(r50)
print("")
print("The third row:")
print(orders.iloc[3])

print("The Order Quantity, Sales, Discount and Profit of the 2nd,
4th, 6th and 8th row:")
print(orders[1:8:2][["Order Quantity", "Sales","Discount", "Profit"]])
print("The Order Quantity, Sales, Discount and  Profit of orders with
discount > 10%:")
print(orders[orders["Discount"] > 0.1][["Order Quantity", "Sales",
"Discount", "Profit"]])
```



```
              Sales    Profit Product Category
Row ID
1          261.5400   -213.25  Office Supplies
49       10123.0200    457.81  Office Supplies
50         244.5700     46.71  Office Supplies
80        4965.7595   1198.97       Technology
85         394.2700     30.94  Office Supplies
86         146.6900      4.43        Furniture
97          93.5400    -54.04  Office Supplies
```

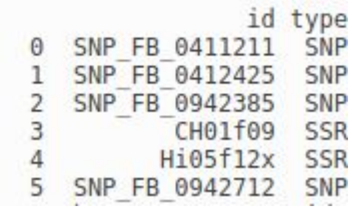
see notes for results

Merging DataFrames



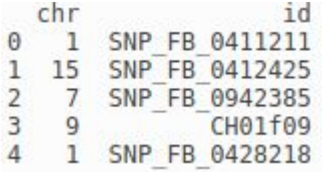
```
pandas.merge(DataFrame1, DataFrame2, on="col_name", how="inner/outer/left/right")
```

DFs1



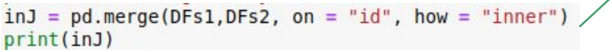
```
               id type
0  SNP_FB_0411211  SNP
1  SNP_FB_0412425  SNP
2  SNP_FB_0942385  SNP
3        CH01f09  SSR
4        Hi05f12x  SSR
5  SNP_FB_0942712  SNP
```

DFs2

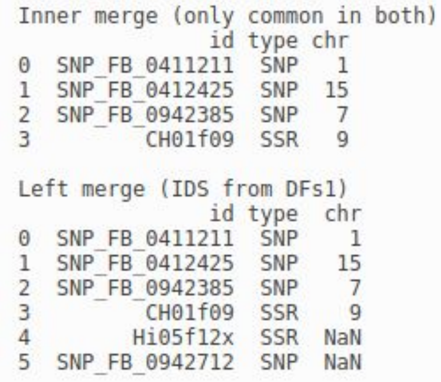


```
  chr              id
0   1  SNP_FB_0411211
1  15  SNP_FB_0412425
2   7  SNP_FB_0942385
3   9         CH01f09
4   1  SNP_FB_0428218
```

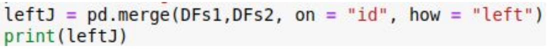
1. how = inner : non-matching entries are discarded;
2. how = left : ids are taken from the first DataFrame;
3. how = right : ids are taken from the second DataFrame;
4. how = outer : ids from both are retained.



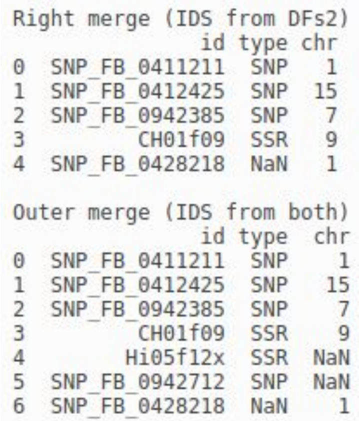
```
inJ = pd.merge(DFs1,DFs2, on = "id", how = "inner")
print(inJ)
```



```
Inner merge (only common in both)
               id type chr
0  SNP_FB_0411211  SNP   1
1  SNP_FB_0412425  SNP  15
2  SNP_FB_0942385  SNP   7
3         CH01f09  SSR   9

Left merge (IDS from DFs1)
               id type  chr
0  SNP_FB_0411211  SNP    1
1  SNP_FB_0412425  SNP   15
2  SNP_FB_0942385  SNP    7
3         CH01f09  SSR    9
4        Hi05f12x  SSR  NaN
5  SNP_FB_0942712  SNP  NaN
```



```
leftJ = pd.merge(DFs1,DFs2, on = "id", how = "left")
print(leftJ)
```

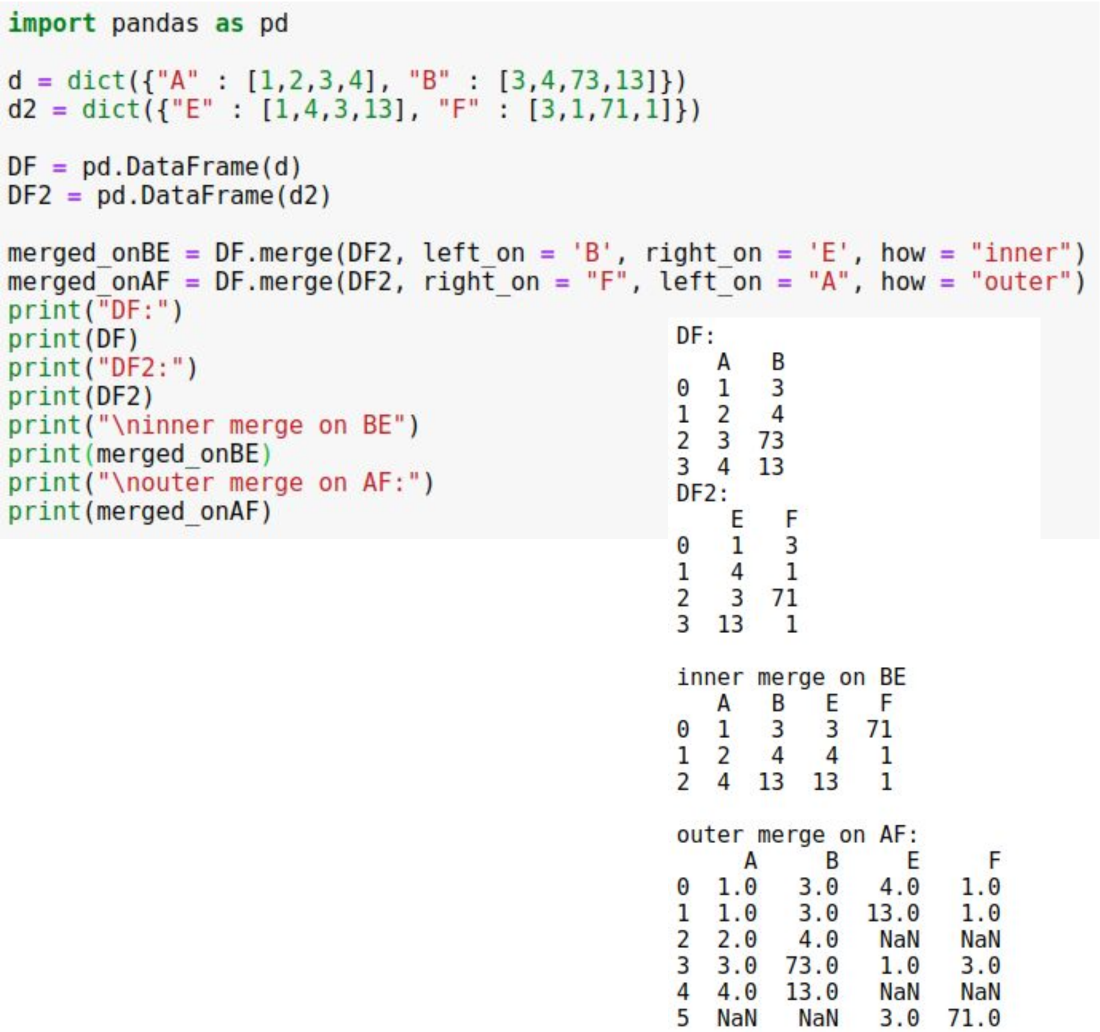


```
Right merge (IDS from DFs2)
               id type chr
0  SNP_FB_0411211  SNP   1
1  SNP_FB_0412425  SNP  15
2  SNP_FB_0942385  SNP   7
3         CH01f09  SSR   9
4  SNP_FB_0428218  NaN   1

Outer merge (IDS from both)
               id type  chr
0  SNP_FB_0411211  SNP    1
1  SNP_FB_0412425  SNP   15
2  SNP_FB_0942385  SNP    7
3         CH01f09  SSR    9
4        Hi05f12x  SSR  NaN
5  SNP_FB_0942712  SNP  NaN
6  SNP_FB_0428218  NaN    1
```

Merging DataFrames

Columns we merge on do not necessarily need to be the same, we can specify a correspondence between the row of the first dataframe (the one on the left) and the second dataframe (the one on the right) specifying which columns must have the same values to perform the merge.

This can be done by using the parameters right_on = column_name and left_on = column_name

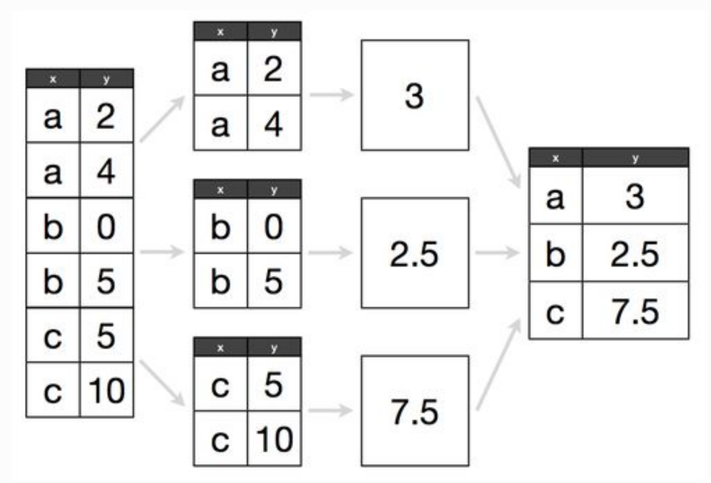


```python
import pandas as pd

d = dict({"A" : [1,2,3,4], "B" : [3,4,73,13]})
d2 = dict({"E" : [1,4,3,13], "F" : [3,1,71,1]})

DF = pd.DataFrame(d)
DF2 = pd.DataFrame(d2)

merged_onBE = DF.merge(DF2, left_on = 'B', right_on = 'E', how = "inner")
merged_onAF = DF.merge(DF2, right_on = "F", left_on = "A", how = "outer")
print("DF:")
print(DF)
print("DF2:")
print(DF2)
print("\ninner merge on BE")
print(merged_onBE)
print("\nouter merge on AF:")
print(merged_onAF)
```

```
DF:
   A   B
0  1   3
1  2   4
2  3  73
3  4  13
DF2:
    E   F
0   1   3
1   4   1
2   3  71
3  13   1

inner merge on BE
   A   B   E   F
0  1   3   3  71
1  2   4   4   1
2  4  13  13   1

outer merge on AF:
     A     B     E     F
0  1.0   3.0   4.0   1.0
1  1.0   3.0  13.0   1.0
2  2.0   4.0   NaN   NaN
3  3.0  73.0   1.0   3.0
4  4.0  13.0   NaN   NaN
5  NaN   NaN   3.0  71.0
```

Grouping DataFrames

The split-apply-aggregate paradigm



x	y
a	2
a	4
b	0
b	5
c	5
c	10

x	y
a	2
a	4

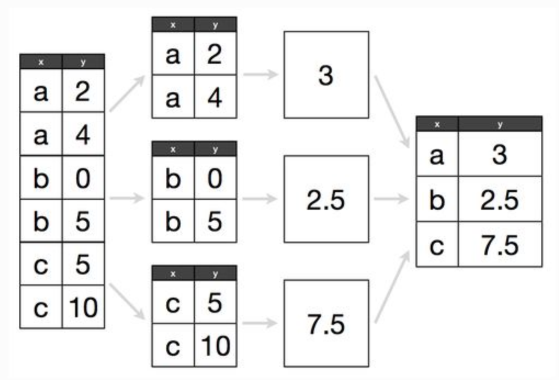
x	y
b	0
b	5

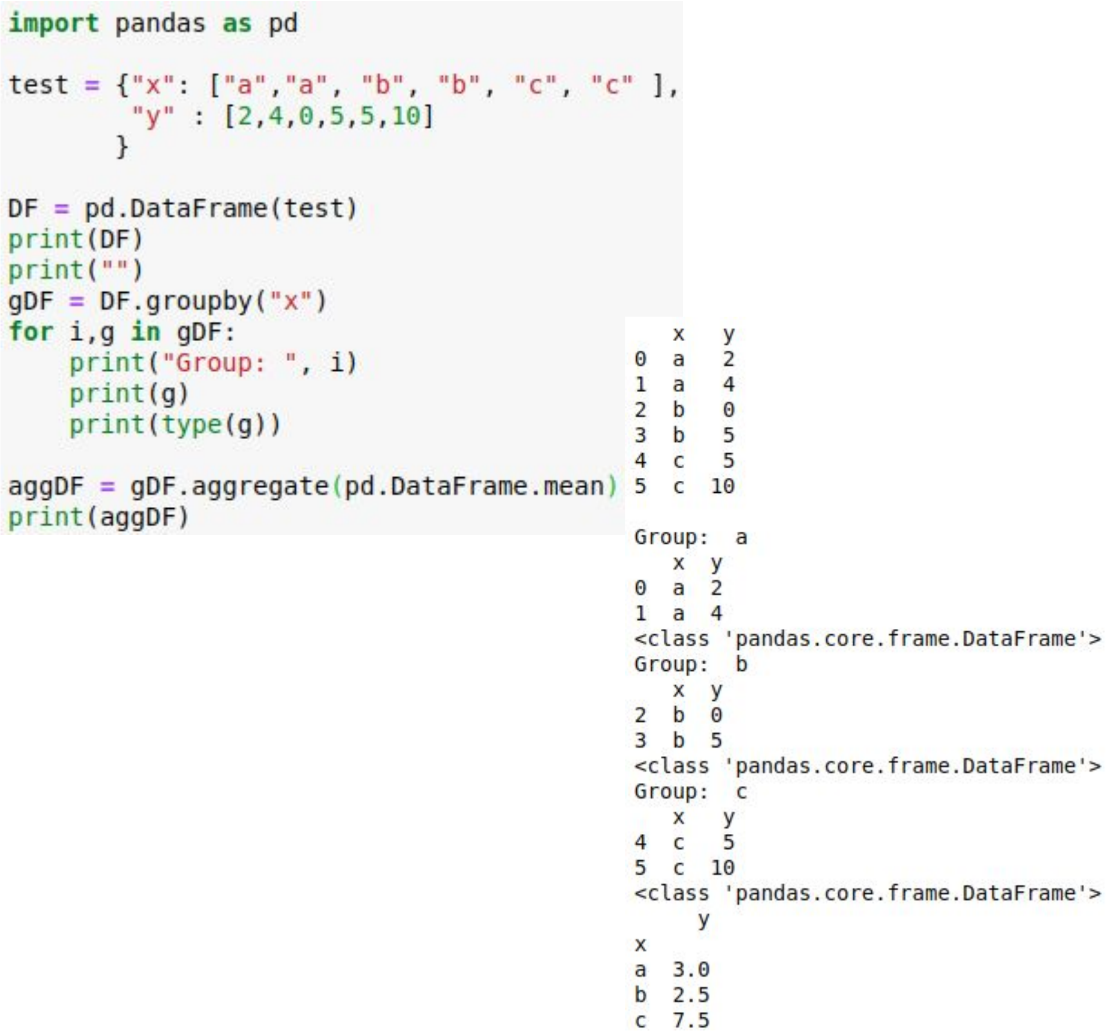
x	y
c	5
c	10

x	y
a	3
b	2.5
c	7.5

Grouping DataFrames

The split-apply-aggregate paradigm



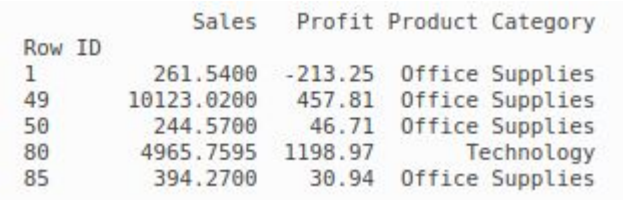


```
import pandas as pd

test = {"x": ["a","a", "b", "b", "c", "c" ],
        "y" : [2,4,0,5,5,10]
       }

DF = pd.DataFrame(test)
print(DF)
print("")
gDF = DF.groupby("x")
for i,g in gDF:
    print("Group: ", i)
    print(g)
    print(type(g))

aggDF = gDF.aggregate(pd.DataFrame.mean)
print(aggDF)
```

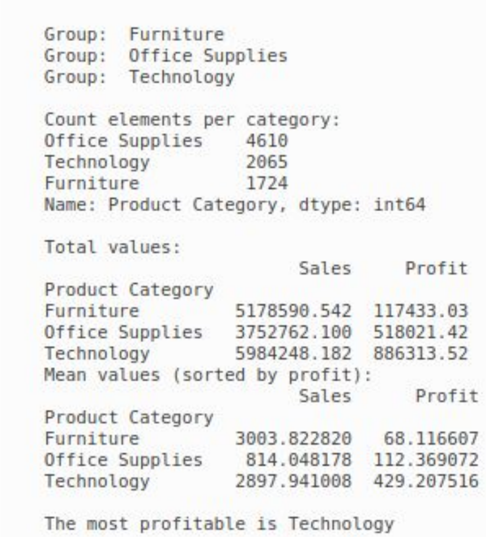
```
   x   y
0  a   2
1  a   4
2  b   0
3  b   5
4  c   5
5  c  10

Group:  a
   x  y
0  a  2
1  a  4
<class 'pandas.core.frame.DataFrame'>
Group:  b
   x  y
2  b  0
3  b  5
<class 'pandas.core.frame.DataFrame'>
Group:  c
   x   y
4  c   5
5  c  10
<class 'pandas.core.frame.DataFrame'>
     y
x
a  3.0
b  2.5
c  7.5
```

Grouping DataFrames

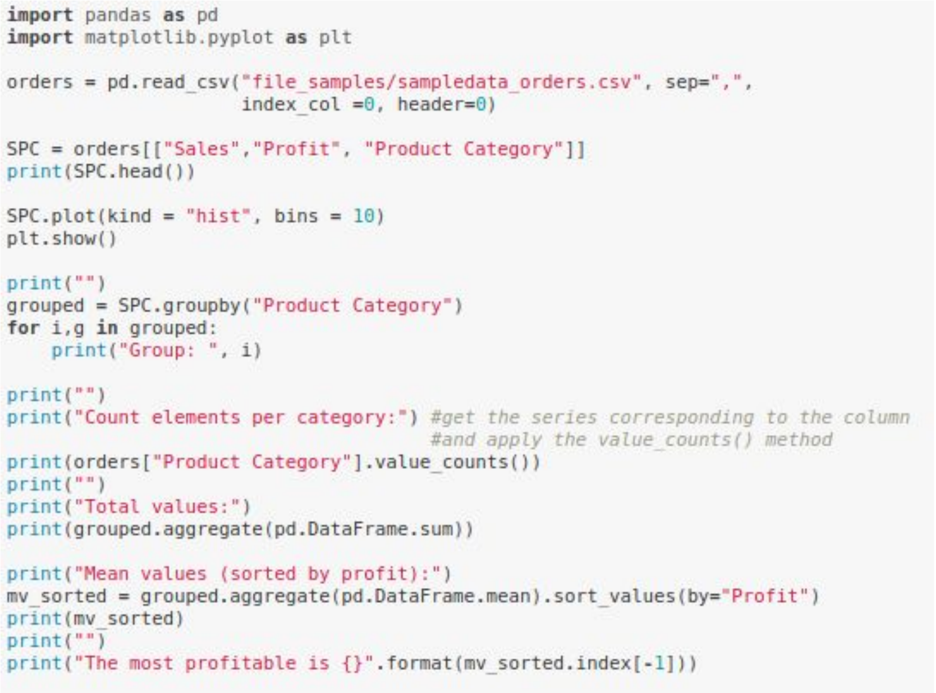
Questions:
How many Product categories?
Total sales and profits per category?
What is the most profitable category?



```
              Sales   Profit Product Category
Row ID
1          261.5400  -213.25  Office Supplies
49       10123.0200   457.81  Office Supplies
50         244.5700    46.71  Office Supplies
80        4965.7595  1198.97       Technology
85         394.2700    30.94  Office Supplies
```

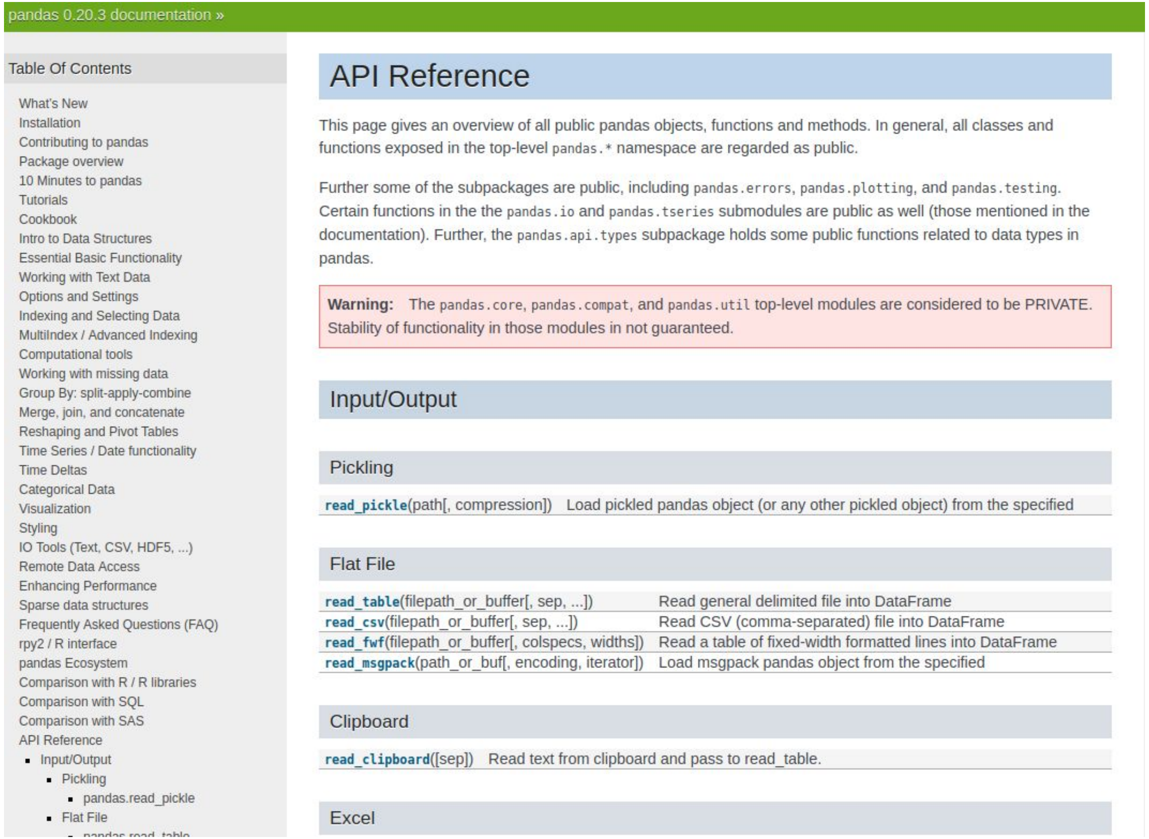


```
Group:  Furniture
Group:  Office Supplies
Group:  Technology

Count elements per category:
Office Supplies    4610
Technology         2065
Furniture          1724
Name: Product Category, dtype: int64

Total values:
                       Sales     Profit
Product Category
Furniture        5178590.542  117433.03
Office Supplies  3752762.100  518021.42
Technology       5984248.182  886313.52
Mean values (sorted by profit):
                       Sales      Profit
Product Category
Furniture        3003.822820   68.116607
Office Supplies   814.048178  112.369072
Technology       2897.941008  429.207516

The most profitable is Technology
```



```
import pandas as pd
import matplotlib.pyplot as plt

orders = pd.read_csv("file_samples/sampledata_orders.csv", sep=",",
                     index_col =0, header=0)

SPC = orders[["Sales","Profit", "Product Category"]]
print(SPC.head())

SPC.plot(kind = "hist", bins = 10)
plt.show()

print("")
grouped = SPC.groupby("Product Category")
for i,g in grouped:
    print("Group: ", i)

print("")
print("Count elements per category:") #get the series corresponding to the column
                                      #and apply the value_counts() method
print(orders["Product Category"].value_counts())
print("")
print("Total values:")
print(grouped.aggregate(pd.DataFrame.sum))

print("Mean values (sorted by profit):")
mv_sorted = grouped.aggregate(pd.DataFrame.mean).sort_values(by="Profit")
print(mv_sorted)
print("")
print("The most profitable is {}".format(mv_sorted.index[-1]))
```

http://pandas.pydata.org/pandas-docs/stable/api.html



pandas 0.20.3 documentation »

Table Of Contents

API Reference

This page gives an overview of all public pandas objects, functions and methods. In general, all classes and functions exposed in the top-level `pandas.*` namespace are regarded as public.

Further some of the subpackages are public, including `pandas.errors`, `pandas.plotting`, and `pandas.testing`. Certain functions in the the `pandas.io` and `pandas.tseries` submodules are public as well (those mentioned in the documentation). Further, the `pandas.api.types` subpackage holds some public functions related to data types in pandas.

Warning: The `pandas.core`, `pandas.compat`, and `pandas.util` top-level modules are considered to be PRIVATE. Stability of functionality in those modules in not guaranteed.

Input/Output

Pickling

read_pickle(path[, compression])	Load pickled pandas object (or any other pickled object) from the specified

Flat File

read_table(filepath_or_buffer[, sep, ...])	Read general delimited file into DataFrame
read_csv(filepath_or_buffer[, sep, ...])	Read CSV (comma-separated) file into DataFrame
read_fwf(filepath_or_buffer[, colspecs, widths])	Read a table of fixed-width formatted lines into DataFrame
read_msgpack(path_or_buf[, encoding, iterator])	Load msgpack pandas object from the specified

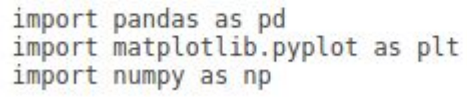
Clipboard

read_clipboard([sep])	Read text from clipboard and pass to read_table.

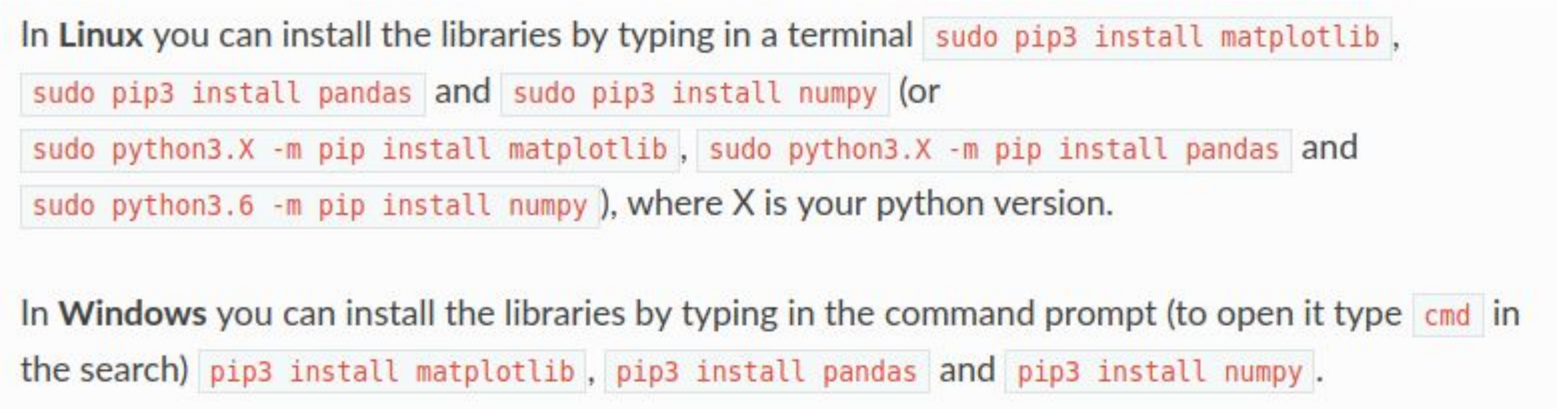
Excel

First things first

We are going to need some libraries



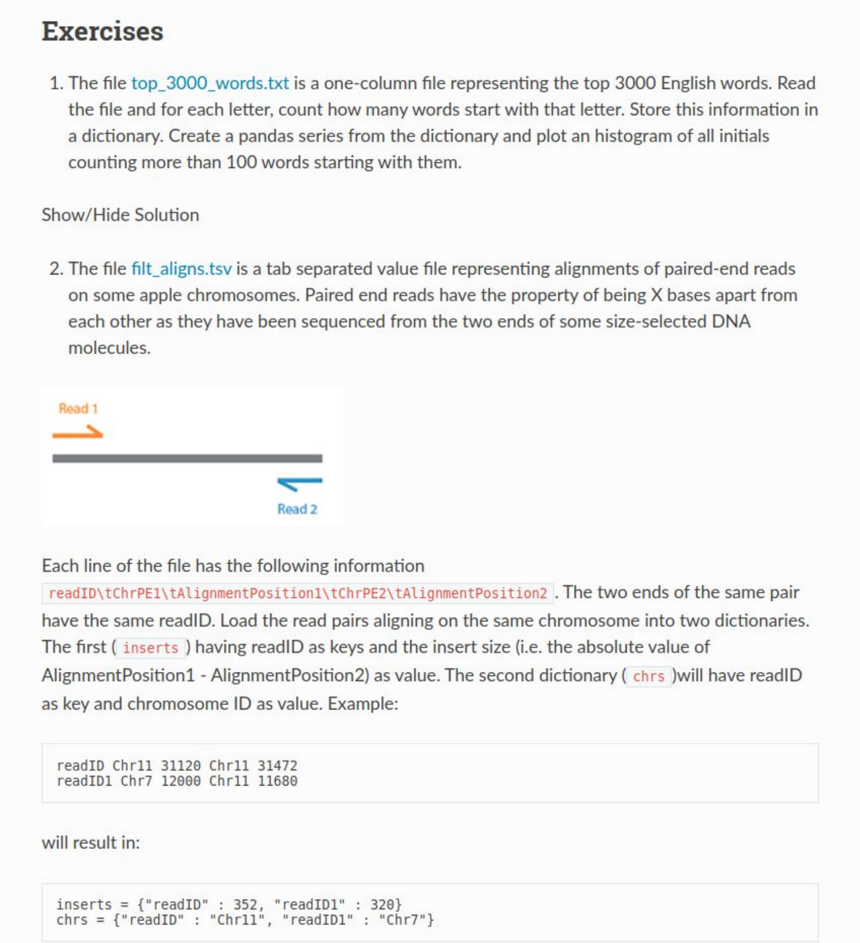
```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```



In **Linux** you can install the libraries by typing in a terminal `sudo pip3 install matplotlib`, `sudo pip3 install pandas` and `sudo pip3 install numpy` (or `sudo python3.X -m pip install matplotlib`, `sudo python3.X -m pip install pandas` and `sudo python3.6 -m pip install numpy`), where X is your python version.

In **Windows** you can install the libraries by typing in the command prompt (to open it type `cmd` in the search) `pip3 install matplotlib`, `pip3 install pandas` and `pip3 install numpy`.

http://qcbsciprolab2019.readthedocs.io/en/latest/practical8.html



Exercises

1. The file top_3000_words.txt is a one-column file representing the top 3000 English words. Read the file and for each letter, count how many words start with that letter. Store this information in a dictionary. Create a pandas series from the dictionary and plot an histogram of all initials counting more than 100 words starting with them.

Show/Hide Solution

2. The file filt_aligns.tsv is a tab separated value file representing alignments of paired-end reads on some apple chromosomes. Paired end reads have the property of being X bases apart from each other as they have been sequenced from the two ends of some size-selected DNA molecules.

Each line of the file has the following information `readID\tChrPE1\tAlignmentPosition1\tChrPE2\tAlignmentPosition2`. The two ends of the same pair have the same readID. Load the read pairs aligning on the same chromosome into two dictionaries. The first (`inserts`) having readID as keys and the insert size (i.e. the absolute value of AlignmentPosition1 - AlignmentPosition2) as value. The second dictionary (`chrs`)will have readID as key and chromosome ID as value. Example:

```
readID Chr11 31120 Chr11 31472
readID1 Chr7 12000 Chr11 11680
```

will result in:

```
inserts = {"readID" : 352, "readID1" : 320}
chrs = {"readID" : "Chr11", "readID1" : "Chr7"}
```