# Importação/Manipulação de Dados com Python

Parte 2

## Lendo CSV em Python, pandas

Primeiramente, import evoca pacotes e tem função similar a library() e/ou require() no R.

```
import numpy as np #para matrizes e arrays
import matplotlib.pyplot as plt
import pandas as pd #para dataframes
print(pd.__version__) # Checks pandas version, we need 0.18 at least
```

## 1.1.4

#### Baby names

Dados são da SSA (Social Security Agency), mas eu só consegui baixá-los de https://github.com/hadley/data-baby-names/blob/master/baby-names.csv.

```
#R
babyNamesR = readr::read_csv("../dados/baby-names.csv") %>% as.data.1
babyNamesR %>% head(n=3)
##
     year name
                   percent sex
## 1 1880
             John 0.081541 boy
## 2 1880 William 0.080511 boy
## 3 1880
            James 0.050057 boy
#python
babyNamesPY = pd.read_csv("../dados/baby-names.csv", header = 0)
babyNamesPY.head()
##
                      percent
      year
               name
                               sex
               John
## 0
     1880
                     0.081541
                               boy
           William 0.080511
     1880
## 1
                               boy
## 2
     1880
              James 0.050057
                               boy
            Charles 0.045167
## 3
      1880
                               boy
## 4
      1880
             George 0.043292
                               boy
```

#### Em R

```
class(babyNamesR)

## [1] "data.frame"

class(py$babyNamesPY)

## [1] "data.frame"
```

#### Em python

```
type(babyNamesPY)

## <class 'pandas.core.frame.DataFrame'>

type(r.babyNamesR)

## <class 'pandas.core.frame.DataFrame'>
```

#### Em python

```
babyNamesPY.shape
## (258000, 4)
babyNamesPY.info()
## <class 'pandas.core.frame.DataFrame'>
## RangeIndex: 258000 entries, 0 to 257999
## Data columns (total 4 columns):
## # Column Non-Null Count
                                Dtype
## ---
## 0 year 258000 non-null int64
## 1 name
                258000 non-null object
## 2 percent 258000 non-null float64
## 3
                258000 non-null object
       sex
## dtypes: float64(1), int64(1), object(2)
## memory usage: 7.9+ MB
```

#### Selecionando colunas

```
print(babyNamesPY.year) #ou print(babyNamesPY['year'])
## 0
             1880
## 1
             1880
## 2
             1880
## 3
             1880
             1880
## 4
##
## 257995
             2008
## 257996
             2008
## 257997
             2008
## 257998
             2008
             2008
## 257999
```

## Name: year, Length: 258000, dtype: int64

## Apply functions on DataFrames

```
df = pd.DataFrame(
  {'A': [1, 2, 3],
   'B':[4, 5, 6]
  })
df.apply(np.mean,axis=0)
## A 2.0
## B 5.0
## dtype: float64
df.apply(np.mean,axis=1)
## 0 2.5
## 1 3.5
## 2 4.5
## dtype: float64
```

Dados do SSA contêm somente os 1000 nomes mais comuns de cada ano....

```
print(babyNamesPY.groupby(['year','sex']).name.count())
## year
        sex
## 1880
        boy
                1000
        girl
##
                1000
## 1881 boy
                1000
        girl
##
                1000
## 1882 boy
                1000
##
                 . . .
## 2006 girl
                1000
## 2007
        boy
                1000
##
        girl
                1000
       boy
## 2008
                1000
        girl
                1000
##
## Name: name, Length: 258, dtype: int64
```

#### Alguns verbos coincidem com dplyr

```
print(babyNamesPY.groupby(['year','sex']).percent.sum())
## year sex
## 1880 boy 0.930746
   girl 0.934546
##
## 1881 boy 0.930439
    girl 0.932690
##
## 1882 boy 0.927532
##
## 2006 girl 0.684830
       boy 0.801105
## 2007
       girl 0.677453
##
## 2008 boy 0.795414
##
       girl 0.672516
## Name: percent, Length: 258, dtype: float64
```

## Indexando (linhas)

Não é possível indexar diretamente um DataFrame, você precisa acessar o atributo i loc

```
print(babyNamesPY.iloc[0]) # ou print(babyNamesPY.iloc[0,:])
## year
                  1880
                  John
## name
## percent 0.081541
## sex
                   boy
## Name: 0, dtype: object
print(babyNamesPY.iloc[0:3])
##
     year
               name
                      percent
                               sex
## 0
     1880
               John 0.081541
                               boy
     1880
           William 0.080511
                               boy
## 1
## 2
     1880
              James 0.050057
                               boy
```

# No (significant number of) boys named Sue...

```
print(babyNamesPY.loc[babyNamesPY.name == "Sue",])
```

```
##
           year name
                       percent
                                  sex
           1880
                 Sue
                       0.000666
                                 girl
## 129189
## 130185
           1881
                 Sue
                       0.000678
                                 girl
## 131171
          1882
                      0.000726 girl
                 Sue
## 132216
          1883
                 Sue
                       0.000566 girl
## 133194
           1884
                       0.000669 girl
                 Sue
## ...
           . . .
                  . . .
## 229543
                       0.000193 girl
           1980
                 Sue
                       0.000152 girl
           1981
## 230654
                 Sue
                                 girl
## 231777
           1982
                 Sue
                       0.000116
           1983
                                 girl
## 232885
                 Sue
                       0.000096
## 233984
           1984
                       0.000082
                                 girl
                 Sue
##
   [105 rows x 4 columns]
```

# Indexando (linhas e colunas)

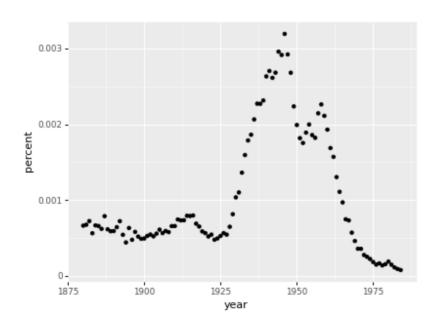
```
print(babyNamesPY.iloc[0,0])
## 1880
print(babyNamesPY.loc[0, 'name'])
## John
print(babyNamesPY.loc[[0,10,100],['name','year']])
##
          name
                year
## 0
          John 1880
        Edward 1880
## 10
## 100
         Perry
                1880
print(babyNamesPY.index)
## RangeIndex(start=0, stop=258000, step=1)
```

## Plot usando matplotlib

```
babySue = babyNamesPY.loc[babyNamesPY.name == "Sue",]
babySue.plot(kind = 'scatter', x = 'year', y = 'percent')
plt.show() # from matplotlib
```

## Plot usando ggplot

```
from plotnine import *
ggplot(babySue) + geom_point(aes(x = 'year', y = 'percent'))
## <ggplot: (-9223372036806471017)>
```



## Formatos Suportados pelo Pandas

Format Type	Data Description	Reader	Writer
text	CSV (*)	read_csv	to_csv
text	JSON	read_json	to_json
text	HTML	read_html	to_html
text	Local clipboard	read_clipboard	to_clipboard
binary	MS Excel	read_excel	to_excel
binary	OpenDocument	read_excel	

#### Observações:

- read\_csv possui o argumento delimiter, que pode ser ajustado para outros tipos de arquivos em texto plano retangulares;
- read\_csv também possui o argumento chunksize, que pode ser usado para leitura por partes.

# Formatos Suportados pelo Pandas

Format Type	Data Description	Reader	Writer
binary	HDF5 Format	read_hdf	to_hdf
binary	Feather Format	read_feather	to_feather
binary	Parquet Format	read_parquet	to_parquet
binary	Msgpack	read_msgpack	to_msgpack
binary	Stata	read_stata	to_stata
binary	SAS	read_sas	
binary	Python Pickle Format	read_pickle	to_pickle
SQL	SQL	read_sql	to_sql
SQL	Google Big Query	read_gbq	to_gbq

#### Pandas e Chunks

## 0.10973800000000006

## SQLite, Pandas e Python

```
import pandas as pd
import sqlite3
conn = sqlite3.connect("../dados/disco.db")
pd.read_sql_query("SELECT * FROM artists LIMIT 5", conn)
     ArtistId
##
                           Name
                          AC/DC
## 0
            2
## 1
                         Accept
                      Aerosmith
## 2
            3
            4 Alanis Morissette
## 3
            5
                 Alice In Chains
## 4
conn.close()
```

## MongoDB, Pandas e Python

```
from pymongo import MongoClient
import pprint
myurl = "mongodb+srv://fernandaBD:mongo123@cluster0.2ph3s.mongodb.net
client = MongoClient(myurl)
db = client['me315mongodb']
collection = db['diamantes']
collection
```

## Collection(Database(MongoClient(host=['cluster0-shard-00-01.2ph3s.mongodb.

#### MongoDB

```
doc = collection.find_one()
pprint.pprint(doc)

## {'_id': ObjectId('5fd034c6e17a0000d50063aa'),
## 'carat': 0.22,
## 'clarity': 'VS2',
## 'color': 'E',
## 'cut': 'Fair',
## 'depth': 65.1,
## 'price': 337,
## 'table': 61.0,
## 'x': 3.87,
## 'y': 3.78,
## 'z': 2.49}
```

#### MongoDB

```
doc = collection.find_one({"cut":"Premium"})
pprint.pprint(doc)

## {'_id': ObjectId('5fd034c6e17a0000d50063ae'),
## 'carat': 0.22,
## 'clarity': 'SI1',
## 'color': 'F',
## 'cut': 'Premium',
## 'depth': 60.4,
## 'price': 342,
## 'table': 61.0,
## 'x': 3.88,
## 'y': 3.84,
## 'z': 2.33}
```

## MongoDB

```
doc = collection.find({"cut":"Premium"}).limit(5)
for x in doc:
  pprint.pprint(x,width=10)
## {'_id': ObjectId('5fd034c6e17a0000d50063ae'),
## 'carat': 0.22,
## 'clarity': 'SI1',
## 'color': 'F',
## 'cut': 'Premium',
## 'depth': 60.4,
## 'price': 342,
## 'table': 61.0,
## 'x': 3.88,
## 'v': 3.84,
## 'z': 2.33}
## {'_id': ObjectId('5fd034c6e17a0000d50063da'),
## 'carat': 0.3,
## 'clarity': 'SI2',
## 'color': 'J',
## 'cut': 'Premium',
## 'depth': 59.3,
## 'price': 405,
```