

# Data structuring, part 1

**The Pandas way**

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# Recap

*Which Python containers have learned about so far?*

-

*Which containers can we turn into a `numpy` array?*

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# Agenda

1. [motivation](#)
2. [numpy and pandas overview](#)
3. the pandas series
  - [working with series](#) and [numeric procedures](#)
  - [boolean series](#)
4. more tools:
  - [inspecting and selecting observations](#)
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**Why we structure data**

# Motivation

*Why do we want to learn data structuring?*

- Data never comes in the form of our model. We need to 'wrangle' our data.

*Can our machine learning models not do this for us?*

- Not yet :). The current version needs **tidy** data. What is tidy?

One row per observation.

## Loading the software

```
In [ ]: import numpy as np  
import pandas as pd
```

# Numpy and Pandas

# Numpy overview

What is the numpy (<http://www.numpy.org/>) module?

numpy is a Python module similar to matlab

- fast and versatile for manipulating arrays
- linear algebra tools available
- used in some machine learning and statistics packages

Example from yesterday

```
In [ ]: table = [[1,2],[3,4]]  
        arr = np.array(table)
```



# Pandas motivation

*Why use Pandas?*

1. simplicity - Pandas is built with Python's simplicity
2. powerful and fast tools for manipulating data from numpy
3. flexibility and new data tools compared with numpy (more info follows)
4. development - breathtaking speed of new tools coming

# Pandas data types

*How do we work with data in Pandas?*

- We use two fundamental data structures: Series and DataFrame.

# Pandas data frames (1)

*What is a DataFrame ?*

- A matrix with labelled columns and rows (which are called indices). Example:

```
In [ ]: df = pd.DataFrame(arr,  
                           index=['i', 'ii'],  
                           columns=['A', 'B'])  
  
print(df)
```

- An object with powerful data tools.

## Pandas data frames (2)

*How are pandas dataframes built?*

Pandas dataframes can be thought of as numpy arrays with some additional stuff.

Most functions from `numpy` can be applied directly to Pandas. We can convert a `DataFrame` to a `numpy` array with `values` attribute.

```
In [ ]: df.values # .tolist()
```

*To note:* In Python we can describe it as a list of lists of a dict of dicts.

# Pandas series

*What is a Series?*

- A vector/list with labels for each entry. Example:

```
In [ ]: L = [1, 1.2, 'abc', True]

my_series = pd.Series(L)
my_series.to_dict()
```

*What data structure does this remind us of?*

- A mix of Python list and dictionary (more info follows)

# Series vs DataFrames

*How are Series related to DataFrames?*

Every column is a series. Example, access as key:

```
In [ ]: print(df.A)
```

Another option is access as object method:

```
In [ ]: print(df['B'])
```

*To note:* The latter option more robust as variables named same as methods, e.g. `count` , cannot be accessed.

# Indices and column names

*Why don't we just use numpy arrays and matrices?*

- inspection of data is quicker
- keep track of rows after deletion
- indices may contain fundamentally different data structures
  - e.g. time series, hierarchical groups
- facilitates complex operation:
  - merging datasets
  - split-apply-combine

# Working with pandas Series



# Generating a Series (1)

Let's revisit our series

```
In [ ]: my_series
```

Components in series

- index: label for each observation
- values: observation data
- dtype: the format of the series - object means any data type is allowed
  - note: the object dtype is SLOW!

## Generating a Series (2)

*How do we set custom index?*

Example:

```
In [ ]: num_data = range(0,3)
        indices = ['B', 'C', 'A']
        my_series2 = pd.Series(num_data, index=indices)
        my_series2
```

## Generating a Series (3)

*Can a dictionary be converted to a series?*

Yes, we just put into the Series class constructor. Example:

```
In [ ]: d = {'yesterday': 0, 'today': 1, 'tomorrow': 3}
        my_series3 = pd.Series(d)
        my_series3
```

Note: Same is true for DataFrames which requires that each value in the dictionary is also a dictionary.

## Generating a Series (4)

*Can we convert series to dictionaries?*

- Yes, in most cases.

```
In [ ]: my_series3.to_dict()
```

- **WARNING!#@:** Series indices are NOT unique

```
In [ ]: s = pd.Series(range(3),index=['A', 'A', 'A'])  
s.index.duplicated().sum()
```

# The power of pandas

*How is the series different from a dict?*

- We will see that pandas Series have powerful methods and operations.
- It is both key and index based (i.e. sequential).

## Converting data types

The data type of a series can be converted with the **astype** method:

```
In [ ]: my_series3.astype(np.float64) # np.str
```

# **Numeric procedures**

# Numeric operations (1)

*How can we basic arithmetic operations with arrays, series and dataframes?*

Like Python data! An example:

```
In [ ]: my_arr1 = np.array([2,3,2])  
        my_arr2 = my_arr1 ** 2  
        my_arr2
```



## Numeric operations (2)

*Are other numeric python operators the same??*

Numeric operators work `/` , `//` , `-` , `*` , `**` as expected.

*Why is this useful?*

- vectorized operations are VERY fast;
- requires very little code.

## Numeric operations (3)

*Can we do the same with two vectors?*

- Yes, we can also do elementwise addition, multiplication, subtractions etc. of series.  
Example:

```
In [ ]: my_arr1 + my_arr2
```

# Numeric methods (1)

Pandas series has powerful numeric methods built-in. Example of 10 mil. obs:

```
In [ ]: arr = np.random.normal(size=10**7)
        s2 = pd.Series(arr)
        s2.median()
```

Other useful methods include: mean, quantile, min, max, std, describe, quantile and many more.

```
In [ ]: my_series2.describe()
```

## Numeric methods (2)

An important method is `value_counts`. This counts number for each observation.

Example:

```
In [ ]: my_series4 = pd.Series(my_arr2)
        my_series4.value_counts() # .unique
```

What is observation in the `value_counts` output - index or data?

## Numeric methods (3)

*Are there other powerful numeric methods?*

Yes: examples include

- `unique, nunique`: the unique elements and the count of unique elements
- `cut, qcut`: partition series into bins
- `diff`: difference every two consecutive observations
- `cumsum`: cumulative sum
- `nlargest, nsmallest`: the n largest elements
- `idxmin, idxmax`: index which is minimal/maximal
- `corr`: correlation matrix

Check [series documentation \(https://pandas.pydata.org/pandas-docs/stable/generated/pandas.Series.html\)](https://pandas.pydata.org/pandas-docs/stable/generated/pandas.Series.html) for more information.

# Boolean Series

## Logical expression for Series

*Can we test an expression for all elements?*

Yes: `==`, `!=` work for a single object or Series with same indices. Example:

```
In [ ]: my_series3 == 0
```

What datatype is returned?

## Logical expression in Series (2)

*Can we check if elements in a series equal some element in a container?*

Yes, the `isin` method. Example:

```
In [ ]: my_rng = range(2)
        print(list(my_rng))
        my_series3.isin(my_rng)
```



# Power of boolean series (1)

*Can we combine boolean Series?*

Yes, we can use:

- the & operator (*and*)
- the | operator (*or*)

```
In [ ]: print((my_series3 > 0) & (my_series3 == 1))
```

```
In [ ]: (df.A > 2) & (df.B<=3) # selection by multiple columns
```

What datatype is returned?

## Power of boolean series (2)

*Why do we care for boolean series (and arrays)?*

Because we can use the to select rows based on their content.

```
In [ ]: my_series3[my_series3<3]
```

NOTE: Boolean selection is extremely useful for dataframes!!

**Inspecting and selecting observations**

# Viewing series and dataframes

*How can we view the contents in our dataset?*

- We can use `print` our dataset
- We can visualize patterns by plotting (from tomorrow)

## The head and tail

We select the *first* rows in a DataFrame or Series with the `head` method.

```
In [ ]: arr = np.random.normal(size=[100])  
my_series7 = pd.Series(arr)  
my_series7.head(3)
```

The `tail` method selects the last observations in a DataFrame.

## Row selection (1)

*How can we select certain rows in a Series when for given index **keys**?*

With the `loc` attribute. Example:

```
In [ ]: # my_loc = 'tomorrow'
         my_loc = ['today', 'tomorrow']
         my_series3.loc[my_loc]
```

## Row selection (2)

*How can we select certain rows in a Series when for given index **integers**?*

The `iloc` method selects rows for provided index integers.

```
In [ ]: # print(my_series3.iloc[1])  
        # print(my_series3)  
        # arr = np.random.normal(size=10**6)  
        pd.Series(arr).iloc[500010:500013]
```

## Row selection (3)

*Do our tools for viewing specific rows, i.e. `loc`, `iLoc` work for DataFrames?*

- Yes, we can use both `loc` and `iloc`. As default they work the same.

```
In [ ]: my_idx = ['i', 'ii', 'iii']
my_cols = ['a', 'b']
my_data = [[1, 2], [3, 4], [5, 6]]
my_df = pd.DataFrame(my_data, columns=my_cols, index=my_idx)
# print(my_df)
print(my_df.loc['i'])
```



## Row selection (4)

*How are Loc , iLoc different for DataFrames?*

- For DataFrames we can also specify columns.

```
In [ ]: idx_keep = ['i', 'ii']  
        cols_keep = ['a']  
        print(my_df.loc[idx_keep, cols_keep])
```

# Columns selection

*How can we select columns in a DataFrame?*

- Option 1: using the `[]` and providing a list of columns.
- Option 2: using `loc` and setting row selection as `:`.

```
In [ ]: print(my_df.loc[:, ['b']])
```

## Selection quiz

*What does : do in iLoc or Loc?*

Select all rows (columns).

# Modifying DataFrames

# Modifying DataFrames

*Why do we want to modify DataFrames?*

- Because data rarely comes in the form we want it.

## Chaging the index (1)

*How can we change the index of a DataFrame?*

We change set a DataFrame's index index using its method `set_index`. Example:

```
In [ ]: print(my_df.set_index('a'))
```

## Chaging the index (2)

*Is our DataFrame changed? I.e. does it have a new index?*

No, we must overwrite it or make it into a new object:

```
In [ ]: print(my_df)
        my_df_a = my_df.set_index('a').copy()
        print(my_df_a)
```

## Chaging the index (3)

Sometimes we wish to remove the index. This is done with the `reset_index` method:

```
In [ ]: print(my_df_a.reset_index(drop=True))  
        # print(my_df)
```

By specifying the keyword `drop = True` we delete the index.

*To note:* Indices can have multiple levels, in this case `level` can be specified to delete a specific level.



## Chaging the column names

Column names can be changed with

```
In [ ]: print(my_df)
        my_df.columns = ['A', 'B']
        print(my_df)
```

DataFrame's also have the function called `rename` .

# Chaging all column values

*How can we can update values in a DataFrame?*

```
In [ ]: print(my_df)

# set uniform value
my_df['B'] = 3
print(my_df)

# set different values
my_df['B'] = [2,17,0]
print(my_df)
```

# Chaging specific column values

*How can we can update values in a DataFrame?*

```
In [ ]: print(my_df)

# loc, iloc
my_loc2 = ['i', 'iii']
my_df.loc[my_loc2, 'A'] = 10
print(my_df)
```

## Sorting data

A DataFrame can be sorted with `sort_values` ; this method takes one or more columns to sort by.

```
In [ ]: print(my_df.sort_values(by='A', ascending=True))
```

*To note:* Many key word arguments are possible for `sort_values`, including `ascending` if for one or more valuable we want descending values. Sorting by index is possible with `sort_index` .

## **DataFrame IO: loading and storing**

# Reading DataFrames (1)

Download the file from [URL \(https://api.statbank.dk/v1/data/FOLK1A/CSV?lang=en&Tid=\\*\)](https://api.statbank.dk/v1/data/FOLK1A/CSV?lang=en&Tid=*). Open directly in Pandas.

```
In [ ]: url = 'https://api.statbank.dk/v1/data/FOLK1A/CSV?lang=en&Tid=*'  
df = pd.read_csv(url, sep=';') # open the file as dataframe  
print(df.head(2))
```

## Reading DataFrames (2)

Now let's try opening the file from the [URL](https://api.statbank.dk/v1/data/FOLK1A/CSV?lang=en&Tid=*) ([https://api.statbank.dk/v1/data/FOLK1A/CSV?lang=en&Tid=\\*](https://api.statbank.dk/v1/data/FOLK1A/CSV?lang=en&Tid=*)) as a local file:

```
In [ ]: abs_path = 'C:/Users/bvq720/Downloads/FOLK1A.csv' # absolute path
        rel_path = 'FOLK1A.csv' # relative path

        df = pd.read_csv(abs_path, sep=';') # open the file as dataframe
        print(df.head(2))
```

- absolute path: entire path starting from which disk etc.
- relative paths: from where your program, i.e. Jupyter is

## Reading other data types

Other pandas readers include: excel, sql, sas, stata and many more.



## Storing data

Data can be stored in a particular format with `to_(FORMAT)` where (FORMAT) is the file type such as csv. Let's try with `to_csv`:

```
In [ ]: df.to_csv('DST_people_count.csv', index=False)
```

Should we always set `index=False` . Yes, unless time series!!! Otherwise the index will be exported too!

# The end

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