

**Pandas**

## Pandas Data Types

Pandas dtype	Python type	Usage
object	str	Text
int64	int	Integer numbers
float64	float	Floating point numbers
bool	bool	True/False values
datetime64	NA	Date and time values
timedelta[ns]	NA	Differences between two datetimes
category	NA	Finite list of text values

## Pandas Data Structure: Series

One-dimensional labeled array

```
In [5]: pd.Series([1.0, 2.4, 3.7])
```

```
Out[5]: 0    1.0  
        1    2.4  
        2    3.7  
        dtype: float64
```

```
In [6]: pd.Series({'a' : 1.0, 'b' : 2.4, 'c' : 3.7}, dtype="int64")
```

```
Out[6]: a    1  
        b    2  
        c    3  
        dtype: int64
```

## Pandas Data Structure: DataFrame

- Two-dimensional, like a spreadsheet or SQL table
- Rows have an *index*
- Columns have a *label* and a *data type*
- Commonly used variable name: `df`

```
In [9]: df = pd.DataFrame({'somecat' : ['a', 'b', 'b', 'a'],
    'somedate': [pd.datetime.utcnow(), pd.datetime(2000,1,1), None, None],
    'somefloat' : [1.4, 1.23456789e-5, None, np.pi],
    'someint' : [-1, 0, 42, 1000],
    'sometext': ['foo', 'bar', 'baz', None]])
df
```

Out[9]:

	somecat	somedate	somefloat	someint	sometext
0	a	2018-08-18 06:09:04.811527	1.400000	-1	foo
1	b	2000-01-01 00:00:00.000000	0.000012	0	bar
2	b	NaT	NaN	42	baz
3	a	NaT	3.141593	1000	None

## DataFrame: data types

In [15]: `df.dtypes`

```
Out[15]: somecat          object
          somedate      datetime64[ns]
          somefloat     float64
          someint       int64
          sometext      object
          dtype: object
```

## DataFrame: head, tail, sample

```
In [17]: df.head(2)
```

```
Out[17]:
```

	<b>somecat</b>	<b>somedate</b>	<b>somefloat</b>	<b>someint</b>	<b>sometext</b>
<b>0</b>	a	2018-08-15 06:43:48.751782	1.400000	-1	foo
<b>1</b>	b	2000-01-01 00:00:00.000000	0.000012	0	bar

```
In [56]: df.tail(2)
```

```
Out[56]:
```

	<b>somecat</b>	<b>somedate</b>	<b>somefloat</b>	<b>someint</b>	<b>sometext</b>
<b>2</b>	b	NaT	NaN	42	baz
<b>3</b>	a	NaT	3.141593	1000	None

```
In [57]: df.sample(2)
```

```
Out[57]:
```

	<b>somecat</b>	<b>somedate</b>	<b>somefloat</b>	<b>someint</b>	<b>sometext</b>
<b>3</b>	a	NaT	3.141593	1000	None
<b>0</b>	a	2018-05-06 17:58:40.283808	1.400000	-1	foo

## DataFrame: dimensions

- `shape`: number of rows and columns
- `count()`: number of values per column (excluding None)

```
In [58]: df.shape
```

```
Out[58]: (4, 5)
```

```
In [59]: df.count()
```

```
Out[59]: somecat      4  
         somedate     2  
         somefloat    3  
         someint      4  
         sometext     3  
         dtype: int64
```

## **Data Import and Export**



## Data Export

- Pandas DataFrame has many to\_\* functions for exporting data
- Writes to file if path is provided
- Many options to specify format:
  - JSON: orient, date formatting, ...
  - CSV: separator, column order, ...
  - SQL: replace, append, data type mapping, ...

```
In [11]: df.to_json('data/demo.json', orient='columns')
df.to_json()
```

```
Out[11]: '{"somecat":{"0":"a","1":"b","2":"b","3":"a"},"somedate":{"0":1534572544811,"1":946684800000,"2":null,"3":null},"somefloat":{"0":1.4,"1":0.0000123457,"2":null,"3":3.1415926536},"someint":{"0":-1,"1":0,"2":42,"3":1000},"sometext":{"0":"foo","1":"bar","2":"baz","3":null}}'
```

```
In [30]: df.to_csv('data/demo.csv', index=False)
df.to_csv()
```

```
Out[30]: ',somecat,somedate,somefloat,someint,sometext\n0,a,2018-08-15 06:43:48.751782,1.4,-1,foo\n1,b,2000-01-01 00:00:00.000000,1.23456789e-05,0,bar\n2,b,,,42,baz\n3,a,,3.141592653589793,1000,\n'
```

## Data Import

- Pandas has many `read_*` functions for importing data into a DataFrame
  - CSV, SQL (query or table), JSON, XML
- CSV:
  - Tries to automatically detect: separators, column names, data types
  - Fails sometimes, but can be defined explicitly
  - Parsing of dates needs to be defined explicitly
- SQL:
  - Requires connection to database

```
In [36]: df = pd.read_csv('data/demo.csv', parse_dates=['somedate'])
```

## Exercise 3 - Pandas overview & Basic functions

Goals:

- Be able to read data from into a Pandas DataFrame
- Get a brief overview about the data

Tasks:

- Open and inspect the Rossmann sample data with the Jupyter web interface
- Import store.csv and sales.csv data as Data Frame
  - Observe the warning when reading *sales* data
- Explore the data briefly:
  - Print the ten first/last/random stores and sales
  - What are the data types of the columns?
  - How many rows and columns are in the data sets?
  - How many null/NaN values are in the data sets?
- Write *store* data to a JSON file

## Exercise 3 - Bonus Tasks

- Fix the warning when reading *sales* data
- Write to and read from SQL (sqlite)
- Why are some numeric values converted to int64 while others are converted to float64?

**Explore Data**

## Selecting columns

If only some columns are required for further processing.

```
In [15]: df[['somecat','someint']]
```

Out[15]:

	somecat	someint
0	a	-1
1	b	0

```
In [16]: df[df.columns[2:4]].head(2)
```

Out[16]:

	somefloat	someint
0	1.400000	-1
1	0.000012	0

## Filter Rows by Boolean Indexing

<https://pandas.pydata.org/pandas-docs/stable/indexing.html#boolean-indexing>  
(<https://pandas.pydata.org/pandas-docs/stable/indexing.html#boolean-indexing>)

```
In [64]: df.loc[(df.somecat == 'b') | (df.someint > 100)]
```

Out[64]:

	somecat	somedate	somefloat	someint	sometext
1	b	2000-01-01	0.000012	0	bar
2	b	NaT	NaN	42	baz
3	a	NaT	3.141593	1000	NaN

## Filter Rows by Query

<https://pandas.pydata.org/pandas-docs/stable/indexing.html#the-query-method>  
(<https://pandas.pydata.org/pandas-docs/stable/indexing.html#the-query-method>).

```
In [65]: df.query('somecat == "b" or someint > 100')
```

Out[65]:

	somecat	somedate	somefloat	someint	sometext
1	b	2000-01-01	0.000012	0	bar
2	b	NaT	NaN	42	baz
3	a	NaT	3.141593	1000	NaN



## N largest / smallest

- Get the rows with the **n** largest or smallest values of a column.

```
In [66]: df.nlargest(2, 'someint')
```

```
Out[66]:
```

	somecat	somedate	somefloat	someint	sometext
3	a	NaT	3.141593	1000	NaN
2	b	NaT	NaN	42	baz

```
In [67]: df.nsmallest(2, 'somefloat')
```

```
Out[67]:
```

	somecat	somedate	somefloat	someint	sometext
1	b	2000-01-01 00:00:00.000000	0.000012	0	bar
0	a	2018-05-06 17:58:40.283808	1.400000	-1	foo

# Descriptive Statistics

## Measures

- Location
  - Mean: Sum up all values and divide them by the number of values,  
$$\bar{x} = 1/n \sum_{i=1}^n x_i$$
  - Median: Order values and take the one from the middle
- Variability
  - Variance:  $\frac{\sum (x - \bar{x})^2}{n-1}$
  - Standard Deviation:  $\sqrt{variance}$
- Example: [1,2,3,10,100]
  - mean:  $(1+2+3+10+100)/5=23.2$ , median: 3
  - variance:  $\frac{(1-23.2)^2 + (2-23.2)^2 + (3-23.2)^2 + (10-23.2)^2 + (100-23.2)^2}{4}$ , std: 43  
$$= \frac{493+449+408+174+5898}{4} = \frac{7422}{4} = 1855,5$$

## Describe

- Generates various summary statistics, excluding NaN.
- By default numeric columns only
- Can be changed by defining `include`

In [68]: `df.describe()`

Out[68]:

	<b>somefloat</b>	<b>someint</b>
<b>count</b>	3.000000	4.000000
<b>mean</b>	1.513868	260.250000
<b>std</b>	1.573883	493.573618
<b>min</b>	0.000012	-1.000000
<b>25%</b>	0.700006	-0.250000
<b>50%</b>	1.400000	21.000000
<b>75%</b>	2.270796	281.500000
<b>max</b>	3.141593	1000.000000

## Describe

- Generates various summary statistics, excluding NaN.
- By default numeric columns only
- Can be changed by defining `include`

```
In [69]: df.describe(include=[np.object,np.bool,np.datetime64])
```

Out[69]:

	<b>somecat</b>	<b>somedate</b>	<b>sometext</b>
<b>count</b>	4	2	3
<b>unique</b>	2	2	3
<b>top</b>	a	2018-05-06 17:58:40.283808	bar
<b>freq</b>	2	1	1
<b>first</b>	NaN	2000-01-01 00:00:00	NaN
<b>last</b>	NaN	2018-05-06 17:58:40.283808	NaN

## Single Summaries

- Many functions to calculate summaries
- mean, min, max, median, quantiles, std, var, count, sum, ...

```
In [70]: df.sum()
```

```
Out[70]: somefloat      4.541605  
         someint      1041.000000  
         dtype: float64
```

## Grouped Summaries

- Works also for grouped data
- mean, min, max, median, quantiles, std, var, count, sum, ...

```
In [71]: df.groupby('somecat').sum()
```

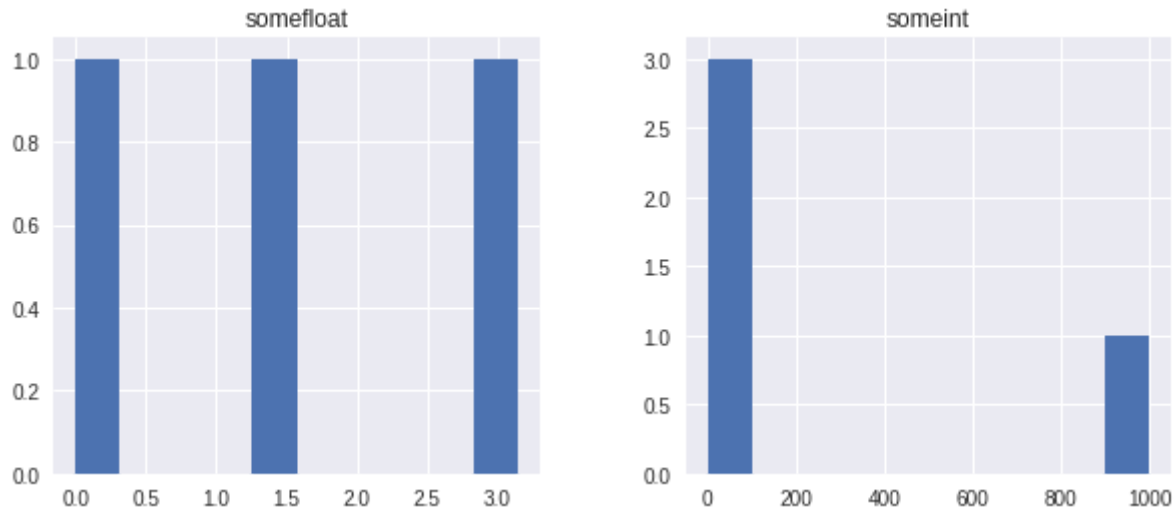
```
Out[71]:
```

	<b>somefloat</b>	<b>someint</b>
<b>somecat</b>		
<b>a</b>	4.541593	999
<b>b</b>	0.000012	42

# Histograms

- Draw histogram of the DataFrame's series using matplotlib / pylab.
- By default 10 bins.
- Frequency of values within the dataset
- Hint: trailing semicolon supresses output

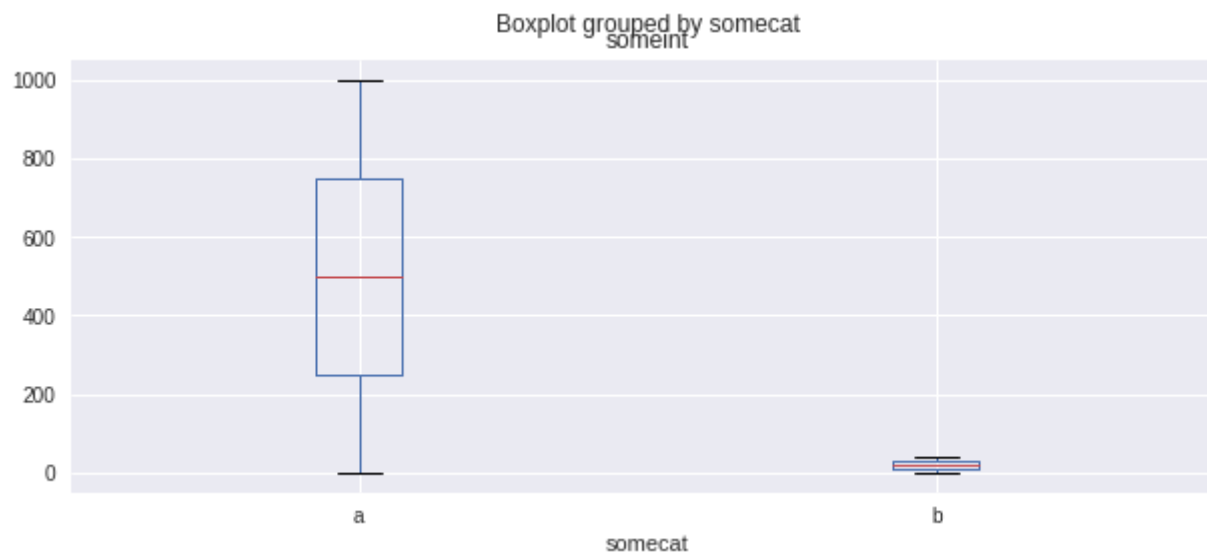
In [72]: `df.hist();`



## Boxplot

- Visualizes the distribution of values
- Median and first and third quantile

```
In [73]: df.boxplot(column='someint', by='somecat');
```





## Exercise 4 - Descriptive Statistics

Goals:

- Explore a data set
- Determine descriptive statistics

Tasks:

- Get the number of stores per store type
- Get the number of stores per assortment
- Determine the number of stores which have a competitor within 5 km
- Calculate min, max, and mean distance to the nearest competitor
- Plot a histogram of the distance to competitors
- Plot sales and number of customers per weekday
- Boxplot with sales per weekday

## Exercise 4 - Bonus Tasks

- Sales per store: min, avg, max, median
- Plot a histogram with sales per store, using 50 bins
- Play with the describe() function

## **Modifying a DataFrame**

## Modifying a DataFrame

- Selection often return a *view*
  - When modifying such a *view* a warning is printed
  - Explicit `copy ( )` possible
- Functions that modify data don't modify the DataFrame but return a copy
  - Many functions have an `inplace` parameter which is `False` by default

## Data Cleaning

- Fill gaps:
  - Interpolate
  - Padding or backfill
  - Fill with default vlaue
  - Drop row
- Drop duplicates

```
In [17]: df = pd.DataFrame({'somecat' : ['a', 'b', 'b', 'a'],
    'somedate': [pd.datetime.utcnow(), pd.datetime(2000,1,1), None, None],
    'somefloat' : [1.4, 1.23456789e-5, None, np.pi],
    'someint' : [-1, 0, 42, 1000], 'sometext': ['foo', 'bar', 'baz', None]}
df
```

Out[17]:

	somecat	somedate	somefloat	someint	sometext
0	a	2018-08-18 13:45:43.329279	1.400000	-1	foo
1	b	2000-01-01 00:00:00.000000	0.000012	0	bar
2	b	NaT	NaN	42	baz
3	a	NaT	3.141593	1000	None

```
In [18]: df2 = df.copy()
df2.somefloat.interpolate(inplace=True)
df2.sometext.fillna(method='ffill', inplace=True)
df2
```

Out[18]:

	somecat	somedate	somefloat	someint	sometext
0	a	2018-08-18 13:45:43.329279	1.400000	-1	foo
1	b	2000-01-01 00:00:00.000000	0.000012	0	bar
2	b	NaT	1.570802	42	baz
3	a	NaT	3.141593	1000	baz

## Data Cleaning

In [19]: df2

Out[19]:

	somecat	somedate	somefloat	someint	sometext
0	a	2018-08-18 13:45:43.329279	1.400000	-1	foo
1	b	2000-01-01 00:00:00.000000	0.000012	0	bar
2	b	NaT	1.570802	42	baz
3	a	NaT	3.141593	1000	baz

```
In [20]: df2.fillna(0, inplace=True)
df2.drop_duplicates('somedate', inplace=True)
df2
```

Out[20]:

	somecat	somedate	somefloat	someint	sometext
0	a	2018-08-18 13:45:43.329279	1.400000	-1	foo
1	b	2000-01-01 00:00:00	0.000012	0	bar
2	b	0	1.570802	42	baz

## Add columns

```
In [22]: df3 = df2.copy()
df3['newtext'] = 'abc'
df3['newint'] = df3['somefloat'].pow(df3.someint).astype(np.int64)
df3
```

Out[22]:

	somecat	somedate	somefloat	someint	sometext	newtext	
0	a	2018-08-18 13:45:43.329279	1.400000	-1	foo	abc	0
1	b	2000-01-01 00:00:00	0.000012	0	bar	abc	1
2	b	0	1.570802	42	baz	abc	1726



## Combine DataFrames

In [27]: df3

Out[27]:

	somecat	somedate	somefloat	someint	sometext	newtext	
0	a	2018-08-18 13:45:43.329279	1.400000	-1	foo	abc	0
1	b	2000-01-01 00:00:00	0.000012	0	bar	abc	1
2	b	0	1.570802	42	baz	abc	1726

```
In [26]: other_df = pd.DataFrame({
    'date': [pd.datetime(2000,1,1), pd.datetime(2000,1,2), pd.datetime(2000,1,3),
             pd.datetime(2000,1,1), pd.datetime(2000,1,2), pd.datetime(2000,1,3),
             pd.datetime(2000,1,1), pd.datetime(2000,1,2), pd.datetime(2000,1,3)],
    'key' : ['foo', 'foo', 'foo', 'bar', 'bar', 'bar', 'baz', 'baz', 'baz']})
other_df
```

Out[26]:

	date	key
0	2000-01-01	foo
1	2000-01-02	foo
2	2000-01-03	foo
3	2000-01-01	bar
4	2000-01-02	bar
5	2000-01-03	bar
6	2000-01-01	baz
7	2000-01-02	baz
8	2000-01-03	baz

## Combine Data Frames

```
In [28]: merged_df = pd.merge(left=other_df, right=df3[['somecat','somefloat','someint',  
            'sometext']], left_on='key', right_on='sometext')  
merged_df
```

Out[28]:

	date	key	somecat	somefloat	someint	sometext
0	2000-01-01	foo	a	1.400000	-1	foo
1	2000-01-02	foo	a	1.400000	-1	foo
2	2000-01-03	foo	a	1.400000	-1	foo
3	2000-01-01	bar	b	0.000012	0	bar
4	2000-01-02	bar	b	0.000012	0	bar
5	2000-01-03	bar	b	0.000012	0	bar
6	2000-01-01	baz	b	1.570802	42	baz
7	2000-01-02	baz	b	1.570802	42	baz
8	2000-01-03	baz	b	1.570802	42	baz

## Exercise 5 - Modifying a DataFrame

Goals:

- Data cleaning
- Add columns
- Combine data frames

Tasks:

- Read date column in *sales* data as date (datetime64)
- Calculate CompetitionOpenSince date from month and year columns (datetime64)
- Fill missing CompetitionDistance with average value of all distances
- Fill missing Competition date with today's date
- Replace StateHoliday categories a, b, c with 1 and convert to int32
- Join store type, assortment, distance and open since from store df to sales df
- Plot average sales and number of customers per store type

## **Exercise 5 - Bonus Task**

- Find correlations and plot a heat map