Big Data and Automated Content Analysis

Week 7 – Monday »Statistics with Python«

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Plotting and calculating with Pandas

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Statistics in Python **General considerations**

General considerations

After having done all your nice text processing (and got numbers instead of text!), you probably want to analyse this further. You can always export to .csv and use R or Stata or SPSS or whatever.

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General considerations

After having done all your nice text processing (and got numbers instead of text!), you probably want to analyse this further. You can always export to .csv and use R or Stata or SPSS or whatever...

BUT:



- the dataset might be too big
- it's cumbersome and wastes your time
- it may introduce errors and makes it harder to reproduce

What statistics capabilities does Python have?

- Basically all standard stuff (bivariate and multivariate statistics) you know from SPSS
- Some advanced stuff (e.g., time series analysis)
- However, for some fancy statistical modelling (e.g., structural equation modelling), you can better look somewhere else (R)

Statistics in Python **Useful packages**

Statistics in Python

OOOOOOOO

Useful packages

```
numpy (numerical python) Provides a lot of frequently used functions, like mean, standard deviation, correlation, ... scipy (scientic python) More of that ;-) statsmodels Statistical models (e.g., regression or time series) matplotlib Plotting seaborn Even nicer plotting
```

Example 1: basic numpy

```
import numpy as np
  x = [1,2,3,4,3,2]
  y = [2,2,4,3,4,2]
  z = [9.7, 10.2, 1.2, 3.3, 2.2, 55.6]
  np.mean(x)
  2.5
  np.std(x)
  0.9574271077563381
  np.corrcoef([x,y,z])
  array([[ 1. , 0.67883359, -0.37256219],
         [ 0.67883359, 1.
                                , -0.56886529],
2
         [-0.37256219, -0.56886529, 1.
                                            ]])
3
```

Characteristics

- Operates (also) on simple lists
- Returns output in standard datatypes (you can print it, store it, calculate with it, ...)
- it's fast! np.mean(x) is faster than sum(x)/len(x)
- it is more accurate (less rounding errors)



Example 2: basic plotting

```
import matplotlib.pyplot as plt
x = [1,2,3,4,3,2]
y = [2,2,4,3,4,2]
plt.hist(x)
plt.plot(x,y)
plt.scatter(x,y)
```

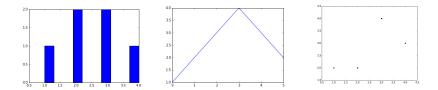


Figure: Examples of plots generated with matplotlib

Pandas Working with dataframes

When to use dataframes

Native Python data structures (lists, dicts, generators)

pro:

- flexible (especially dicts!)
- fast
- straightforward and easy to understand

con:

- if your data is a table, modeling this as, e.g., lists of lists feels unintuitive
- very low-level: you need to do much stuff 'by hand'



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Pandas dataframes

pro:

- like an R dataframe or a STATA or SPSS dataset
- many convenience functions (descriptive statistics, plotting over time, grouping and subsetting, ...)

con:

- not always necessary ('overkill')
- if you deal with really large datasets, you don't want to load them fully into memory (which pandas does)

Plotting and calculating

Plotting and calculating with Pandas

OLS regression in pandas

```
import pandas as pd
import statsmodels.formula.api as smf

df = pd.DataFrame({'income': [10,20,30,40,50], 'age': [20, 30, 10, 40, 50], 'facebooklikes': [32, 234, 23, 23, 42523]})

# alternative: read from CSV file (or stata...):
# df = pd.read_csv('mydata.csv')

myfittedregression = smf.ols(formula='income ~ age + facebooklikes', data=df).fit()
print(myfittedregression.summary())
```

```
OLS Regression Results
                                                              0.579
    Dep. Variable:
                             income R-squared:
    Model:
                               OLS Adj. R-squared:
                                                              0.158
    Method:
                      Least Squares F-statistic:
                                                             1.375
                     Mon. 05 Mar 2018 Prob (F-statistic):
    Date:
                                                             0.421
    Time:
                           18:07:29 Log-Likelihood:
                                                          -18.178
    No. Observations:
                                 5 AIC:
                                                             42.36
                                                              41 19
    Df Residuals:
                                 2 BIC:
10
    Df Model:
11
    Covariance Type:
                          nonrobust
    coef std err
                              P>ItI
                                      [95.0% Conf. Int.]
14
    Intercept 14.9525 17.764 0.842 0.489
                                                       -61.481 91.386
16
        0.4012 0.650 0.617 0.600 -2.394 3.197
    age
    facebooklikes 0.0004 0.001 0.650 0.583
17
                                                       -0.002 0.003
18
19
    Omnibus:
                               nan Durbin-Watson:
                                                             1.061
20
    Prob(Omnibus):
                               nan Jarque-Bera (JB):
                                                             0.498
21
    Skew:
                            -0.123 Prob(JB):
                                                             0.780
22
                              1 474 Cond No.
                                                            5 21e+04
    Kurtosis:
23
```

Other cool df operations

```
df['age'].plot() to plot a column
df['age'].describe() to get descriptive statistics
df['age'].value_counts() to get a frequency table
and MUCH more...
```

Recoding and transforming

```
To transform your data, you can use .apply(), .applymap(), and .map() or the .str.XXX() methods:
```

```
df['is_center'] = df['hood'].str.contains('[cC]enter')
```

or define your own function:

```
def is_center(x):
    return int(x.lower().find('center') > -1)

df['is_center'] = df['hood'].map(is_center)
```

or use a throwaway-function:

Recap:

• [0:5] to get elements 0, 1, 2, 3, 4 (works with lists, dataframes . . .)



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 df[['col1', 'col2']] to get only these two columns of a dataset



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- df[['col1', 'col2']] to get only these two columns of a dataset
- df[df['col1']=='whatever'] to get only the rows in which col1 is identical to the string 'whatever'



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And therefore, also:

- df[['col1', 'col2']] to get only these two columns of a dataset
- df[df['col1']=='whatever'] to get only the rows in which col1 is identical to the string 'whatever'
- df [df ['col2']>0] to get only the rows in which col2 is a number bigger than 0



More subsetting

To get a apecific row and/or column, you can use .iloc[] and .loc[]

• .iloc[] takes an int (the row/column numbers, .loc[] the names)



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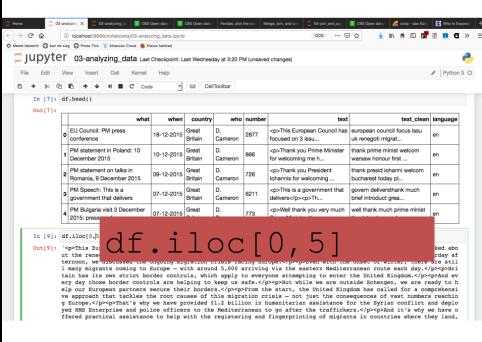
- .iloc[] takes an int (the row/column numbers, .loc[] the names)
- df.iloc[0,5] to get row 0, column 5

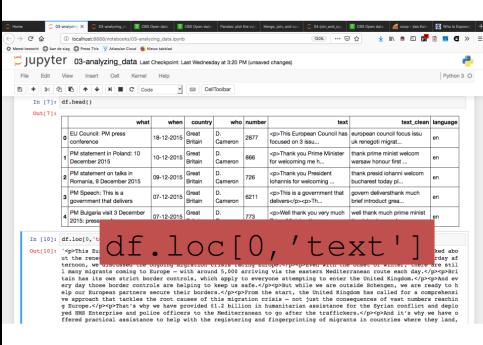
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- .iloc[] takes an int (the row/column numbers, .loc[] the names)
- df.iloc[0,5] to get row 0, column 5
- df.loc[0,'what'] to get row 0, column 'what'







Advanced Example

Out of a dataset with 1,000 speeches, get the one that talks most about [Tt]error

• We create a new column to count how many a word is mentioned:

```
df['terror'] =
df['speech'].str.count('[Tt]error')
```

Advanced Example

Out of a dataset with 1,000 speeches, get the one that talks most about [Tt]error

• We create a new column to count how many a word is mentioned:

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df['terror'] =
df['speech'].str.count('[Tt]error')
```

We do
 df.iloc[df['terror'].idxmax()]

Advanced Example

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• We create a new column to count how many a word is mentioned:

```
df['terror'] =
df['speech'].str.count('[Tt]error')
```

- We do
 df.iloc[df['terror'].idxmax()]
- That works because df.iloc[] expects an integer to identify the row number, and df ['terror'].idxmax() returns an integer (687 in our case)

```
df['terrorrefs'].idxmax()
687
df.iloc[687]
what
                Permanent Link to Press conference in Islamabad
                                                       14-12-2008
when
                                                   Great Britain
country
who
                                                         G. Brown
                                                             2954
number
text
              Transcript of a press conference given by t...
text clean
              transcript press confer given prime minist mr ...
language
                                                               en
terrorrefs
                                                               44
```

Name: 687, dtype: object

Joining and Merging

Typical scenario

- You have two datasets that share one column
- For instance, data from www.cbs.nl: one with economic indicators, one with social indicators
- You want to make one dataframe



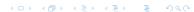
economie = pd.read_csv('82800ENG_UntypedDataSet_15112018_205454.csv', delimiter=';')
economie.head()

	ID	EconomicSectorsSIC2008	Regions	Periods	GDPVolumeChanges_1
0	132	T001081	PV20	1996JJ00	9.3
1	133	T001081	PV20	1997JJ00	-2.0
2	134	T001081	PV20	1998JJ00	-0.9
3	135	T001081	PV20	1999JJ00	-0.7
4	136	T001081	PV20	2000JJ00	1.5

population = pd.read_csv('37259eng_UntypedDataSet_15112018_204553.csv', delimiter=';')
population.head()

_					
	ID	Sex	Regions	Periods	LiveBornChildrenRatio_3
0	290	T001038	PV20	1960JJ00	18.6
1	291	T001038	PV20	1961JJ00	18.9
2	292	T001038	PV20	1962JJ00	18.9
3	293	T001038	PV20	1963JJ00	19.5
4	294	T001038	PV20	1964JJ00	19.6

What do you think: How could/should a joined table look like?



```
# remove unnecessary columns
economie.drop('ID',axis=1,inplace=True)
population.drop('ID',axis=1,inplace=True)
# remove differentiation by sex
population = population[population['Sex']=='T001038']
population.drop('Sex',axis=1,inplace = True)
# keep only rows of economic dataframe that contain the total economic activity
```

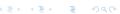
```
# remove those evil spaces at the end of the names of the provinces
population['Regions'] = population['Regions'].map(lambda x: x.strip())
economie['Regions'] = economie['Regions'].map(lambda x: x.strip())
```

```
population.merge(economie, on=['Periods', 'Regions'], how='inner')
```

economie = economie[economie['EconomicSectorsSIC2008']=='T001081
economie.drop('EconomicSectorsSIC2008', axis=1, inplace=True)

	Regions	Periods	LiveBornChildrenRatio_3	GDPVolumeChanges_1
0	PV20	1996JJ00	11.0	9.3
1	PV20	1997JJ00	11.4	-2.0
2	PV20	1998JJ00	11.6	-0.9
3	PV20	1999JJ00	11.6	-0.7
4	PV20	2000JJ00	11.5	1.5
5	PV20	2001JJ00	11.7	3.9
6	PV20	2002JJ00	11.4	2.1

Then merge



On what do you want to merge/join?

Standard behavior of.join(): on the row index (i.e., the row number, unless you changed it to sth else like a date)

```
df3 = df1.join(df2)
```



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But that's only meaningful if the indices of df1 and df2 mean the same. Therefore you can also join on a column if both dfs have it:

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df3 = df1.merge(df2, on='Regions')
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.merge() is the more powerful tool, .join() is a bit easier when joining on indices.



Inner, Outer, Left, and Right

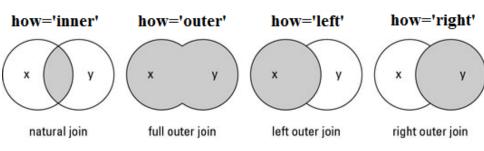
Main question: What do you want to do with keys that exist only in one of the dataframes?



Inner, Outer, Left, and Right

Main question: What do you want to do with keys that exist only in one of the dataframes?

df3 = df1.join(df2, how='xxx')



Aggregation

An example

- Suppose you have two dataframes, both containing information on something per region per year.
- You want to merge (join) the two, however, in one of them, the information is also split up by age groups. You don't want that.
- How do you bring these rows back to one row? With .agg()!

Aggregation

• Very useful after a .groupby()

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- Takes a function as argument: df2 = df.groupby('region').agg(sum)

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- Takes a function as argument: df2 = df.groupby('region').agg(sum)
- Or multiple functions:
 df2 = df.groupby('region').agg([sum, np.mean])
- ullet o yes, you could do .describe(), but .agg() is more flexible



An example

wijken

How do housing prices (WOZ-waarde) develop over time in different neighborhoods?

		Heighborho	Jus:						
	0	Burgwallen-Oude Zijde	263417.0	273525.0	289984.0	339548.0	400010.0	A00	Centrum
	1			281193.0	296762.0	351214.0	391011.0	A01	Centrum
	2			502230.0	560841.0	674610.0	755091.0	A02	Centrum
	3	Grachtengordel-Zuid	469946.0	478371.0	531225.0	627625.0	697576.0	A03	Centrum
pu	t; double	click to hide arkt/Lastage	295239.0	303500.0	340364.0	386716.0	438942.0	A04	Centrum
	5	Haarlemmerbuurt	304924.0	311743.0	345189.0	403267.0	458522.0	A05	Centrum
	6	Jordaan	270390.0	285877.0	307344.0	347740.0	402186.0	A06	Centrum
	7	De Weteringschans	344649.0	359119.0	399942.0	458010.0	515192.0	A07	Centrum
	8	Weesperbuurt/Plantage	307440.0	322276.0	353628.0	413388.0	473643.0	A08	Centrum
	9	Oostelijke Eilanden/Kadijken	253990.0	256421.0	276481.0	316261.0	381774.0	A09	Centrum
	11	Westelijk Havengebied	NaN	189402.0	224491.0	NaN	NaN	B10	Westpoort
	13	Houthavens	164263.0	167242.0	188360.0	349525.0	483318.0	E12	West
	14	Spaarndammer- en Zeeheldenbuurt	207439.0	209713.0	222371.0	256300.0	322981.0	E13	West
	15	Staatsliedenbuurt	209792.0	222070.0	241366.0	277214.0	325787.0	E14	West

Steps

- Get it into a tidy format (1 row = 1 observation) ("long" format)
- Optionally, but more neat (also for automatically getting correct plot labels): index rows by year
- 3 use .groupby() and .agg() to aggregate the data

```
wijken long = wijken.melt(id vars=['wijk', 'stadsdeel'],
                        value_vars=['2014', '2015', '2016', '2017', '2018'],
                        value name='woz-waarde',
                        var name = 'year')
                  .melt() transforms a df from wide to
wijken_long
                  long
0
    Burgwallen-Oude Ziide
                              Centrum
                                      2014 263417.0
                              id vars: what are the
1
    Burgwallen-Nieuwe Ziide
2
    Grachtengordel-West
                                 cases?
3
    Grachtengordel-Zuid
                              Cen
4
    Nieuwmarkt/Lastage
                              cer value vars: which vars
5
    Haarlemmerbuurt
                              cer contain the values?
    Jordaan
6
    De Weteringschans
                              Centrum
                                      2014 344649.0
    Weesperbuurt/Plantage
                                      2014 307440.0
8
                              Centrum
                                      2014 253990.0
9
    Oostelijke Eilanden/Kadijken
                              Centrum
10
    Westeliik Havengebied
                              Westpoort 2014 NaN
```

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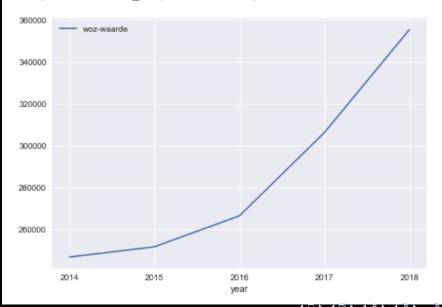
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- Group by:
 - Group only by year
 - ② Group by year and 'stadsdeel'
- Aggregation function
 - 1 mean
 - **2** Possibly also min, max, or even lambda x: max(x)-min(x)

wijken_long.groupby('year').agg(np.mean).plot(xticks=[0,1,2,3,4])

<matplotlib.axes._subplots.AxesSubplot at 0x1191a4128>



wijken_long.groupby(['year','stadsdeel']).agg(np.mean).unstack().plot(
 figsize=[10|,7], xticks=range(5))



What's unstacking?

Nieuw-West

Noord

Oost

2000

2224

2015

2016

228636.000000

wijken_long.groupby(['year','stadsdeel']).agg(np.mean)

		woz-v	waarde	-> lurn hierararc						
year	stadsdeel									
2014	Centrum	32681	4.100000	indices into non-						
	Nieuw-West	20045	3.500000	hierarchical struc						
	Noord	21587	9.500000							
	Oost	22182	8.142857							
	West	2358	wijken_lo	ong.group	pby(['year','	.agg(np.mean).unstack()				
	Westpoort	NaN		I	4-					
	Zuid	3382	stadsdeel	woz-waa	Nieuw-West	Noord	Oost	West	w	
	Zuidoost	158€	year	Contruin	Micuw-West	Hoolu	Ouat	11031	-	
2015	Centrum	3374	2014	326814.1	200453.500000	215879.500000	221828.142857	235801.0	N	

337425.5

200028.000000

370176.0 208002.428571

222417.200000

229650.466667

archical nructure

228636.000000

244608.428571

238568.8 189402.0

235801.0 NaN

Westpoort Zuid

3382

3465

3559

There are example datasets and notebooks on Canvas!

Find an exercise here: https://github.com/annekroon/bdaca/ blob/master/ipynb/basic_statistics.ipynb or on Canvas (under 'modules')

Next steps

Thursday: Final lecture

You are now able to read and write Python code. Therefore, we can now introduce some advanced analysis topics, mainly machine learning.

Monday 20/05

Not mandatory: In case you need (individual) consultation regarding final project.

