

Mumpy functions	np.mean() np.median() np.sum()	Matplotlib	N stay switch 3 perc. perc. 4 perc. perc. ↳ legend: stay & switch y-axis: perc. x-axis: 3,4	Subsets
	np.log2() np.absolute() np.nan np.linspace(..., ...) np.arange(start, stop, step)	df.plot() df.plot.set() df.plot.scatter() df.plot.set_xlabel('...') df.plot.set_ylabel('...') df.col.hist()	sub = df[df.col == ...] sub2 = df[(df.col1 == ...) & (df.col2 > ...)] (df.col == ...).sum() / mean	
Pandas functions	Pandas series	df.col.unique()	% magic % time % timeit % ls	% cp % % time Linux command
Pd.isnull() Pd.notnull() Pd.read_csv() Pd.read_excel() Pd.concat ↳ can be used for chunks if chunk size is passed to Pd.read_csv Pd.crosstab() ↳ pass series for index, series for cols & margins = True for 'All' ↳ crossdf[...].unstack() for same result as pivot_table() Pandas dataframe functions	df.col.value_counts() df.col.dtype df.col.to_datetime() df.col.str.contains('...') ↳ boolean, dus kan nog .mean() / .sum() achter. df.col.str.strip() df.col.idxmax() df.col.astype(type) ↳ change type of col df.col.values ↳ get values from a col as an array df.col.index df.col.str.findall(pattern) df.col.min/max/sum/mean/mode median/detect_outliers_boxplot() df.drop() df.copy() df.iterrows() df.sort_index() df.set_index() df.reset_index() df.applymap(lambda x: *functie wat je met x doet en waar je x mee vervangt*) f.shape() f.pivot_table(index='colname', columns='colname') f.groupby('colname') f.col.dropna(subset='...') ↳ subset to define in which col to look for missing values f.sort_values() f.count() → non na values f.round(int/dict/series) f.mean() f.corr() f.to_pickle('filename') ↳ save df as pickle file f.describe() f.fillna() f.head() / df.tail()	Plotly df.col.iplot(kind=...) df.col.value_counts().iplot() df.crosstab.iplot() Seaborn sns.set() → to use seaborn on matplotlib.pyplot as plt sns.boxplot() sns.pairplot(df, hue='category', column) sns.distplot(df) ↳ combination of hist & kdeplot sns.lmplot() sns.heatmap(df.corr()) sns.jointplot() sns.scatterplot() sns.kdeplot() sns.FacetGrid(df, row, col) ↳ after this, use grid.map(plt.hist, bins=) sns.plot.add_legend() sns.factorplot()	Counter(list) % magic % time % timeit % ls	strings s.len s.get s.split('seperat') s.count('...') s.find('...') *'.join(list) s.lower() s.replace() s.splitlines() s.startswith() s.strip() s.upper()
Theory 1	Accuracy = $(TP+TN)/\text{alles}$ Precision = $TP/(TP+FP)$ ↳ $P(\text{ja} \text{uitslag ja})$ recall = $TP/(TP+FN)$ vb) accuracy 95% occurrence 1 op 1000 hoeveel mensen met positieve test zijn niet besmet? $TP = 0.95 * 0.001$ $FP = (1-0.95) * (1-0.001)$ Precision = $TP/(TP+FP)$	Theory 2 $P(A B) = \frac{P(A \cap B)}{P(B)}$ $= \frac{P(A B)P(B)}{P(B)}$ ↳ broadcasting	five different ways how many men and women were on board titanic.sex.value_counts() titanic.groupby('sex')[['survived']].count() titanic.pivot_table(index='sex', values='survived', aggfunc='len') pd.crosstab(titanic.sex, titanic.survived)	[...] ↳ choice [...] ↳ any except within [a-z] ↳ any characters between * → any single char. \\s → any white space \\S → any non white sp. \\d → any digit \\D → any non digit \\w → any word char. \\W → any non word cha (a b) → matches either a or b a? → 0 or one of a a* → 0 or more of a a+ → one or more of a a{3} → exactly 3 of a \$ end of string ^ start of string

str. contains str. extract str. startwith re.sub re.findall

\d	digit	{3}	exactly 3x
\w	letter	{2,4}	2 to 4 x
\s	whitespace	{3,}	3 or more x
\D	non-digit	*	0 or more x
\S	non-whitespace	?	once or none
[]	one of	^	start of string
[a-z]	a to z	\$	end of string

def strcount(file):

i = defaultdict(int)

with open(file) as f:

for line in f:

for letter in line:

i[letter] += 1

omgekeerd sort = hoog \rightarrow laag dus
ascending = False

nltk.tokenize.wordpunct_tokenize()

df[df[x] < df[x].quantile(.99)] \rightarrow verwijderde 1% outliers

actual values precision = $\frac{TP}{TP+FP}$

		actual values	
		P	N
values	P	TP	FP
	N	FN	TN

Recall = $\frac{TP}{TP+FN}$

import seaborn as sns.

y = sns.load_dataset("x")

y.groupby(["x"]).sum().count() → dataframe gegroepeerd

y.sort_values('z', ascending=False) → geef dataframe gesorteerd hoog-laag o.b.v. z

y['z'].value_counts() → hoeveel z voorkomt in y

y.loc[y['z'] == 'three'] → dataframe met alleen de regels waar 'three' in z voorkomt

y['z'].values → Returns array of values in z.

ony[[z, w]] → maak dataframe y with columns z and w

np.mean, np.median, np.std

Pd.DataFrame.from_dict(x, orient='index')

.sort_index()

.idxmax() → hoogste index waarop

pandas.Index(a) & pandas.Index(b) → intersect

pandas.Index(a) | pandas.Index(b) → union

pandas.Index(a) ^ pandas.Index(b) → symmetric difference

y[(y > 0.3)] = masking

y.T → transpose

y[y > 100]

.fillna(0)

y.dropna()

y.set_index("x")

pandas.read_csv('x', encoding='ISO-8859-1', sep=',', header=0)

.capitalize() → make a capital letter of the first letter of str

pivot voorbeeld: maak pivot die de meest voorvkomende materialen geeft van de 10 meest verkoopste kunstwerken. Index moet de naam kunstwerk zijn, kolommen materiaal en value aantal keer dat de kunstwerk het gebruikt heeft.

Antwoord: → for i in kunst: Kunst[i['Naam-kunstwerk'], 'Id'].groupby('Naam-kunstwerk').count(), sort_values('Id', ascending=False); i: id, index: value → Kunst.pivot_table([kunst['Naam-kunstwerk']], index='Naam-kunstwerk', columns='Materiaal')

↳ aggfunc=len is count, sum = values ogeteld

pandas.crosstab(y.z, y.w))

Count how many time each word occurs from collections import defaultdict with open(file) as f:

```
for l in f:  
    for c in l.split():  
        tellings[c] += 1
```

tellings = pd.Series(tellings).sort_values(ascending=False)
return tellings

axis=1 (cols) = 0 (rows)

Precision =
$$\frac{TP}{TP + FP}$$

Recall =
$$\frac{TP}{TP + FN}$$

Ry selecteren DF.ik[[...]]

Series naar DF: pd.DataFrame({col1: ser1, col2: ser2})

DF.describe() → count, mean, std, min, max

df = DataFrame
se = Series

xx = VarList()

pd.cstab?

functies

df.Series(data, index='index')
df.DataFrame(data, columns=[list], index=[list])

df[se].index / df.columns => Maanen van Jan C

df.size / shape / ndim / dtype

if (se > 0, 3) & (se < 0, 8) masking

if loc [df.col > x] & (df.col < y), [hist col] matching

if transpose => switch cols and index

sum(), mean(), median(), min(), max()

df.dropna().describe()

count() => totaal aantal

default dict() => methode in voorloop & (int / float / str of c)

with open file as f: regel voor regel

pd.read_excel(excel)(file)

df.value_counts()

df.sort_index / values

df.query()

pd.crosstab

de.sinc()

With B2.2 file (file) conform as &:

Groeperen

df.groupby('key').sum()

df.groupby('key').agg(aggregate['min', np.median]) maak e functies

df.groupby('key').apply(eigen function)

df.pivot_table('col', index='col', columns='col')

functies op DF

df.col [1 * x - y * z] df.col

df.subtract(df.col, axis=0) => op kolommen

None / np.nan

isnull(), notnull() = bool, alles niet null

dropna(), f.1nal method='ffill', axis=0/1)

Samenvoegen

pd.concat(dfr, dfr2), pd.merge(dfr, dfr2)

dfr.append(dfr2)

11-09

RegEx

se.str.contains(r'') => bool na .startswith()

re.findall(r'', data)

1. \D = Any general, geen getal

2. \S = Any spatie, geen spatie

3. \W = letter, geen letters

\b = alles behalve newline

+ * ? = H+, O+, loto keer

{3,3+} = 3x 3,3+
| or ^ = net iets
| ...) groep

Grouping & Pivot

planets.groupby('dom1').sum()

df.groupby('dom1')

• aggregate(['min', np.median, max]) → die methodes

• aggregate({'data1': 'min', 'data2': 'max'}) voor elke kolom

• filter(filter_func) → filteren of gem. gradijs dan niet van een cat.

• transform(lambda x: x - x.mean()) → houdtzelfde shape

• apply(lambda by: data2) → alles delen door som van kolom data2

df.groupby(mapping)

mapping = {'A': 'rowel', 'B': 'consonant', 'C': 'consonant'}

df.groupby([str.lower(), mapping]).mean() → hierarchical indexing

L = [0, 1, 0, 1, 2, 0]

df.groupby(L) → index 0, 2, 5 samen

titanic.groupby(['sex', 'class'])

[['survived'].aggregate('mean').method]

= titanic.pivot_table('survived', index='sex', columns='class')

indexen cols kunnen dan zijn voor hierarchical

Theorie

Onderstaan net grote beelden:

1 Met werkplaatsen, gezip file

(reg) voor regel

2 Clean memory object werkt

3 Verdediging over

versch computers.

Z-score

$$z_{\text{score}} = (t_{\text{bo}} - t_{\text{bo}.mean}) / t_{\text{bo}.std(1)}$$

Missing values:

1 Invullen met constante (NaN or None?)

2 Verwijgen door gem (numeriek)

of modus (categoriek)

3 Verwijgen door random waarde

4 Verwijgen door inferred waarde

Guthres herkennen:

1 histogram 2 scatterplot 3 |Z-score| < 3

Regex

soup.findall('Affiliation')

part- RegisteredName.txt

[a-zA-Z] een volgels
+ huren letters
* een of meer
? nul of meer
. nul of een

{3} {3,6} 3, 3-6
^ nul of start string
\$ eind string
| elke karakter
\\S, \\s (geen) whitespace
\\D, \\d (geen) cijfer
(...) wat je nu) capuren
re.findall, split, sub
\\W, \\w (geen) speciale letter

16 word boundary

dataL['2013'] wordt op datetime
str.lower(), str. **<TAB>**, str.capitalize()
Als 'info': ['BJCID', ...] dan maakt
df['info'].str.get_dummies('1') dan
dummy variabelen
str.contains('BJbreakfast').sum()
dummy_df[dummy_df.isin(frequent_sections)]
pd.get_dummies(categorical Series).apply(lambda

for file in tqdm_notebook(namelist):
f = archive.open(file), root = tree.getroot()
tree = etree.parse(f), root.find('handelingen')

.set_index('titel', inplace=True)

df['n-sections'] = df.max()

.value_counts → histogram

quantile(0.99)

& intersection | → union

| → symmetric diff.

pd.DataFrame.from_didict(df), orient=

'index'

df.sort_values(by=['kolom']), place=True,

ascending=False)

data[ddata.notnull()] → Null or None

df1.intersection(df2)

pd.merge(df1, df2, how='inner')

'outer' V intersection
union

1. pandas

Lees csv file in

```
import numpy as np  
import pandas as pd  
import seaborn as sns  
import re
```

```
pd.read_csv(bestand, sep = ',', index_col = ...  
, usecols = ['...', ...]) = df
```

```
from tqdm import  
tqdm.notebook
```

Sort op index of kdom

```
df.sort_index(axis = 0, ascending = T/F, inplace = True)  
df.sort_values('kdom', ascending = True, ...)
```

kdom als index

```
df = df.set_index('kdom')
```

unieke waarden tellen

```
df['kdom'].value_counts()
```

Dataframes aan elkaar plakken

```
s = pd.concat([df1, df2], join = inner, axis = 1)
```

specifieke kdom van specifieke rij

```
peiling.loc[['D66']]['Zetels']
```

```
peiling['Zetels'].loc['D66']
```

Partij met meer dan 25 zetels

```
zetels = pd.DataFrame(df['Zetels'] > 25)
```

Data Cleaning

Name columns

```
df.columns = ['...']
```

check voor null values

```
sd.isnull()
```

Drop rows with null values

```
pd df.dropna(axis = 1)
```

replace null values with mean

```
df.fillna(s.mean())
```

Groupby

Selecteer specifieke waarde in kdom

```
nl = df[df['original-lang'] = 'nl']
```

5 meest vertaalde boeken uit nl

```
meest = nl.groupby('target-lan').max()
```

```
['num-transl'].nlargest(5)
```

zach hier tuples van

```
lst = set(x for x in meest.items())
```

veel mannen en vrouwen

waren aan board

```
titanic.groupby('sex')['survived'].count()
```

```
titanic['sex'].value_counts()
```

Pivot tables

```
pd.pivot_table(df, index = ['sex'], columns =  
'class', values = 'survived', aggfunc = len)
```

```
titanic.pivot_table()
```

veel mensen zaten er per geslacht

in elke klasse.

```
titanic.pivot_table(index = 'sex', columns =
```

```
'class', values = 'survived') of :
```

```
pd.crosstab(titanic.sex, titanic.class)
```

gemiddelde leeftijd van mensen die het wel

of niet hebben overleefd

```
titanic.pivot_table(index = 'sex', columns =  
'survived', values = 'age', aggfunc = [np.mean,  
np.median])
```

```
wanneer precies is de variabele 'adult-male'  
waar titanic1 = titanic[titanic['adult-male']]>  
titanic1.pivot_table(index = 'age', values =  
['adult-male'], aggfunc = sum)] | preprocess:
```

regular expressions

```
-csv(bestand) Prince = pd.read
```

```
bestand inladen: prince.text = prince.text.astype
```

```
(str)
```

```
bestand = open('prince-lyrics.csv', encoding = 'utf8')
```

Hoeveel nummers bevatten het woord sex. (met of

```
print(prince.text.str.contains(r'\b[Ss]ex\b')) | sur
```

hoeveel nummers beginnen met sex?

```
print(prince.text.str.contains(r'^[Ss]ex\b')).sum() of  
prince.text.str.startswith('Sex').sum() + prince.text.startswith('sex')
```

hoeveel nummers bevatten sex maar niet letterlijk sex als woord

```
ef = prince.text.str.findall(r'[Ss]ex[w+}') | print(ef[ef.str.len() > 0].head())
```

```
print(ef[ef.str.len() > 0].count()) | from collections import Counter
```

```
print(Counter([s for l in ef.values for s in l]))
```

Regex * = 0 or more + = 1 or more ? = 0 or 1 {2} = exactly two

```
{2,5} = between 2 and 5 {2,} = 2 or more [ab-d] = one character of a,b,c,d
```

```
[l-b] = backspace \d = one digit \s = one space \w = one word ^ = start of str
```

```
$ = end of string \b = word boundary . = any character except newline \a = character a
```

```
\ = escapes special character. werkt op pandas Series (ach Nee)
```

```
String operations: str.capitalize() = hoofdletter monte.str.lower() monte.str.len()
```

```
monte.str.startswith('T') monte.str.split() Regex methods: match() = returns boolean
```

```
extract() = matched groups,.findall() = for each element, replace() = replace pattern with str
```

```
contains() = boolean, count() = count occurrences of patterns, rsplit() excepts regen
```

```
extract first name: monte.str.extract('([A-Za-z]+)', expand = False). Find the names
```

```
that start or end with a consonant monte.str.findall(r'^[AEIOU].*[aeiou]$')
```

```
Extract last name monte.str.split().str.get(-1) \w = [a-zA-Z0-9-]
```

```
\W = [^a-zA-Z0-9_] re.search = match pattern anywhere in string(pattern, string)
```

Wikipedia bestand per regel inlezen

```
with open(bestand, encoding = 'utf-8') as f:
```

```
for line in f: \n:
```

```
for l in tqdm.notebook():
```

```
split regels: gesplit = l.split('\t')[1:]
```

```
dictje = {}
```

```
for x in gesplit:
```

```
weer = x.split(' )
```

```
dictje[weer[0]] = int(weer[1])
```

```
for x, y in dictje.items():
```

```
iets[x] = 1
```

```
iets[x] = y
```

```
for iets in dictje.items():
```

handige overzichten: dir(str) / dir(pd) / dir(np)

DF uit dictionary: pd.DataFrame.from_dict(orient = 'index')

```
from collections import Counter
```

correlatie = corr1 = df['k'].corr(df['i'])

maak dict van lijsten

lege-dict [leeglijst] = [lijst, lijst]

reset index:

```
df.reset_index()
```

Series = 1 kolom uit df

voorbeld inlezen en letters tellen

```
bestand = 'prince-lyrics.csv'
```

```
def tellertje(bestand):
```

```
from collections import defaultdict
```

```
telling = defaultdict(int)
```

with open(bestand) as f:

```
for l in f:
```

```
for c in list(l):
```

```
telling[c] += 1
```

tellingseries = pd.Series(telling).sort_values(ascending = False)

```
return tellingseries
```

```
df = pd.DataFrame({key:  
[A,B,C,B,A,C], data:  
[3,5,7]})
```

df.groupby(key).sum()

key Data

A 3

B 5

C 7

meteen dir(pd), dir(np) \Rightarrow ls magic
with open('Bestand') as file:
for page in file:

Read - CSV()

df.sort_index()
df.sort_values('kolom')
df.set_index('kolom')
df.value_counts()

Pd.concat, pd.merge \rightarrow df1.merge(df2)

Selection

df['col'] \rightarrow returns series

df[['col1', 'col2']] \rightarrow nieuwe df.

df.iloc[0,:] \rightarrow first row

Cleaning

df.columns = ['a', 'b', 'c'] \rightarrow rename
pd.isnull(), pd.notnull() df.dropna()
df.dropna(axis=1) \rightarrow drop columns
df.fillna(x), so.astype(float)
s.replace(1, 'one')

Filter

df[df['col'] > 0.5]
df[(df['col'] > 0.5) & (df['col'] < 0.7)]
df.sort_values([col1, col2], ascending=[True, False])
df.groupby(col)
df.groupby([col1, col2])
df.groupby([col1][col2], mean)/sum/etc.
df.groupby(col).agg(np.mean)
df.apply(np.mean), df.apply(np.max, axis=dummy_df = pd.get_dummies(dummy_df)
df.pivot_table(index=col1, values=col2, aggfunc=mean)
~~df.crosstab?~~

Pd.DataFrame.from_dict?

\rightarrow Dict of Dicts to DataFrame

df.DataFrame(dict)

df.Series(dict)

titanic['sex'].value_counts()

titanic.pivot_table(index='sex', columns='class', values='survived')
pd.crosstab(titanic['sex'], titanic['class'])

index.name = 'naam'

titanic.groupby('sex')[['survived']].mean()

de dubbele paren geven een mood frame

titanic.groupby(['sex', 'class'])['survived'].agg(mean).unstack() \rightarrow zorgt voor mood frame

ge = pd.cut(titanic['age'], [0, 18, 80])

titanic.pivot_table('survived', ['sex', 'age'], 'class')

df.qcut(titanic['fare'], 2)

Regex **dir(str)**
df['kolom'].str.findall(r'bla bla').str.len()
len(), str.match(), replace(), count()
lower(), extract(), contains(), split()
Prince['text'].str.contains(r'sex').sum()
 \rightarrow alles behalve '\n' (newline)

W \rightarrow [a-zA-Z0-9_]

W \rightarrow tegen overgestelde

b boundary between words and not word

s single white space

S non white space character

t, \n, \r \rightarrow tab, newline, return

d \rightarrow [0-9] \rightarrow alles behalve

^ = start, \$ = end

+ = one or more occurrences

* = zero or more

? = match 0 or 1 of thing to the left

[abc] = a or b or c $\{^abc\}$ alles behalve

re.sub(regex, vervanging, string)

endswith(), str.startswith()

Plotting

$x =$ $y =$
df.plot(kind=[line, bar, hist, barh, box, scatter, pie], subplots=True or False)
use_index=True \rightarrow sticks for x axis
~~title = 'string'~~, grid=False, legend=False
logx=False, logy=False, loglog=False,
xticks=[values for xaxis], yticks=[]
sort_columns=False)

Frequent_sections = [1st met titels]

dummy_df = df.sections.apply(pd.Series).iloc

dummy_df = pd.get_dummies(dummy_df)

dummy_df = dummy_df[frequent_sec].groupby

(by='page_id').sum()

chance_dict = dict()

for b in frequent_sections:

chance_dict[b] = dict()

for a in frequent_sections:

if a == b:

chance_dict[b][a] = 1

else:

bkans = sum(dummy_df[b] == 1)

chance_dict[b][a] = len(dummy_df[C

dummydf[a] == 1] / (dummydf[b] == 1)) / bkans

Conditionele kans: Kans op A en B als je

weet dat B er is. $A+B/B$

Recall = $TP/(TP+FN)$ Je bent ziek + test goed

Precision = $TP/(TP+FP)$ Test is positief + je bent ziek

Languages = {l2.split(';')[0]: int(l2.split(';')[1])}

for l2 in l.split('t')[1:-1]

10 frequent-sections:
 change_dict [b] = dict()
 for a in frequent-sections:
 if dummy - df-test[b][a] > 0:
 change_dict[Test[b][a]] = dummy - df-test[b][a]

c[8]:
 change_di ... [b][a] = 0
 change_di ... [b] = True [a].sum() / dummy ... [a].sum()

nd_change = pd.DataFrame (change_dict).sort_index(axis=1)

4. iris-iris-species
 logistic function
 def logfunc(x):
 return 1/(1+np.exp(-x))

11 frequent-sections:
 change_dict [b] = dict()
 for a in frequent-sections:
 if dummy - df-test[b][a] > 0:
 change_dict[Test[b][a]] = dummy - df-test[b][a]

c[8]:
 change_di ... [b][a] = 0
 change_di ... [b] = True [a].sum() / dummy ... [a].sum()

nd_change = pd.DataFrame (change_dict).sort_index(axis=1)

pd.crosstab(churn[0], ch[0]) / 48
 churn[0].value_counts()
 pd.cust(churndf[0], 3, labels=[
 'low', 'medium', 'high'])
 file naar dict
 D = dict()
 With open('fib.tsv') as f:
 for line in f:
 ll = line.split('\t')
 D[[ll[0]]] = {i: split(',')[0]:
 :split(',')[1] for i in ll[1]}
 • nieuw kolom met aantal char.
 df['new'] = df.name.str.capitalize().str.len()
 alle recepten zonder breakfast
 A = recipe.dropna(subset=[description])
 A[[A.description.str.contains(r'(B|b)reakfast)')]]
 load planets dataset
 sns.load_dataset('planets')
 eerste liggende planet per methode
 indices = Planets.groupby('method')[distance]
 planets.loc[indices]
 loading in big files
 import gzip
 with gzip.open('input.gz', 'r') as fin:
 for line in fin:
 print 'Regel: ', Line[0:10]
 allines.append(line.split(':'))
 reader = pd.read_csv('person.csv.gz',
 sep=';',
 header=0,
 encoding='utf8',
 compression='gzip',
 skiprows=[1, 2, 3],
 index_col=0,
 na_values='NULL',
 low_memory=False,
 usecols=['mycols',
 chunksize=10000)
 df = pd.concat([churn.dropna(axis=1),
 'index', subset=[birthDate]]) for chunk
 in reader)

extract first names
 name.str.extract('([A-Z][a-z]+)',
 expand=False)
 start + end with cursor.
 monte.str.findall(r'[AEIOU][aeiou]')

- same as str.slice(0,3)
- last name each entry
- split column into two
- monte['info'].str.get_dummies(1)
- longest ingredient list
- recipes.name[np.argmax(recipes.ingr.
 show name cinnamon as ingr.
- rec.ingr.str.contains(CC) * np.sum()
- is ingredient in the list?
- import re
- spice = pd.DataFrame(dict((spice, rec.ingr.str.
 contains(spice, re.IGNORECASE))) for
 spice in spice-list))
- recipe with tomato, fish
- spiceadf.query('tomato&fish')
- recipes.name[[index]]
- match, extract, replace, contains,
- count

HS 2 oper kolom mean kolom
 corsdf.fillna(corsdf.mean())
 corsdf.notnull(), is null means
 ova data cleaning
 corsdf.country.replace('!', '')
 dec.name = 'decade'
 dec.name = 'decade'
 decade = Planets.groupby('method')[
 'decade'][['number']].sum()
 maxstack(1, fillna(0))
 Ooit bij pivot te gebruiken
 als index plus
 ooit heel tafde:
 A.groupby(['sex', 'class'])
 ['survived'].aggregate('mean').
 unstack
 A.pivot_table('survived',
 index='sex', columns=
 'class')
 bin age
 age = pd.cut(A['age'], [0, 18])
 A.pivot_table('survived', ['sex', 'age'],
 'class')
 ooit add taal, quantile cat
 fare = pd.qcut(A['fare'], 2)
 in aggregate kan zo'n n dicht
 'surv': sum, 'fare': 'mean'
 margins=True gaff total

name.met.median() in plants.
 for method, group in plants.
 groupby('method'):
 print ('{}: {} shape={}'.format(
 group, group.shape))
 format(method, group, shape)
 m = aggregate name (min, max)
 of E[data1:min, data2:max]
 after function
 def filter(x):
 return x[(data2 > x) &
 x.groupby(key).filter(filter)]
 subtract groupwise mean
 df.groupby(key).transform(lambda x: x - x.mean())
 apply
 df.groupby(key).apply(normalize)
 def normalize(x):
 x[(data1 < x) & x < data2].sum()
 return x
 split key
 L = grouping keys, len gelijk aan df
 df.groupby(L).sum()
 map index to group
 AF = df.set_index('key')
 mapping = AF['id': 'dick', 'etc'
 df2.groupby(mapping).sum()
 [, name str.lower, i]
 count discovered planets by
 method and decade
 dec = 10 #Planets['year'] // 10
 dec = dec.astype(str) + 't'
 dec.name = 'decade'
 Planets.groupby('method')[
 'decade'][['number']].sum()
 maxstack(1, fillna(0))
 HS 2 oper kolom mean kolom
 corsdf.fillna(corsdf.mean())
 corsdf.notnull(), is null means
 ova data cleaning
 corsdf.country.replace('!', '')
 dec.name = 'decade'
 decade = Planets.groupby('method')[
 'decade'][['number']].sum()
 maxstack(1, fillna(0))
 HS 3 clean columns = ll.split(':')[0].
 replace('!', '') for win columns
 [0:] + ['churn']
 churndf = pd.read_csv('file', header=
 None, names=cleancolumns)
 omdat True == 1
 churn.Corr() [churn.Corr(), round(2) == 1]
 omdat True kijgen
 churn[0].str.strip() == 'True', sum())
 set to boolean type
 churn[0]:
 ooit heel is true?
 churndf[0].sum() /count()
 ooit internationale plan
 churn[0].value_counts() /sum() * 100
 crossstab

titanic.sex.unique() [3].Count() ← naam niet median()
 for method, group in titanic.
 groupby('method'):
 print ('{}: {} shape={}'.format(
 group, group.shape))
 format(method, group, shape)
 m = aggregate name (min, max)
 of E[data1:min, data2:max]
 after function
 def filter(x):
 return x[(data2 > x) &
 x.groupby(key).filter(filter)]
 subtract groupwise mean
 df.groupby(key).transform(lambda x: x - x.mean())
 apply
 df.groupby(key).apply(normalize)
 def normalize(x):
 x[(data1 < x) & x < data2].sum()
 return x
 split key
 L = grouping keys, len gelijk aan df
 df.groupby(L).sum()
 map index to group
 AF = df.set_index('key')
 mapping = AF['id': 'dick', 'etc'
 df2.groupby(mapping).sum()
 [, name str.lower, i]
 count discovered planets by
 method and decade
 dec = 10 #Planets['year'] // 10
 dec = dec.astype(str) + 't'
 dec.name = 'decade'
 Planets.groupby('method')[
 'decade'][['number']].sum()
 maxstack(1, fillna(0))
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 corsdf.fillna(corsdf.mean())
 corsdf.notnull(), is null means
 ova data cleaning
 corsdf.country.replace('!', '')
 dec.name = 'decade'
 decade = Planets.groupby('method')[
 'decade'][['number']].sum()
 maxstack(1, fillna(0))
 HS 3 clean columns = ll.split(':')[0].
 replace('!', '') for win columns
 [0:] + ['churn']
 churndf = pd.read_csv('file', header=
 None, names=cleancolumns)
 omdat True == 1
 churn.Corr() [churn.Corr(), round(2) == 1]
 omdat True kijgen
 churn[0].str.strip() == 'True', sum())
 set to boolean type
 churn[0]:
 ooit heel is true?
 churndf[0].sum() /count()
 ooit internationale plan
 churn[0].value_counts() /sum() * 100
 crossstab

gendercount = EP: Counter(genderlist[p][N] for p in genderlist)}
 ker = pd.DataFrame.from_dict(gendercount, orient='index').fillna(0).astype(int)
 ker['vrouw'] = (ker['female'] / ker['sum']) * 100).round()
 ker.sort_values('vrouw', ascending=False, inplace=True)
 D = EP: secret[translate[p][df.loc[p].Zetels] for p in translate}
 t = pd.DataFrame.from_dict(D, orient='index')
 A.columns = ['vrouwen']
 df = df.join(A)
 df['man'] = df.Zetels - df.vrouwen

voorwaardelijke kans:
 P(A|B) = P(A ∩ B) / P(B)
 SOA-test 95%. accuraat, toch geen SOA? 1 in 1000
 je wil complement van $\frac{TP}{TP+FP}$ dus - $\frac{FP}{TP+FP}$
 $TP = P(ja, \text{geen SOA}) = P(\text{SOA}) * P(\text{test ja}) = P(ja|\text{SOA}) * P(\text{SOA} = 0,95) * 0,001$
 $FP = P(ja, \text{geen SOA}) = P(\text{geen SOA}) * P(ja) = P(ja|\text{geen SOA}) * P(\text{geen SOA}) = (1 - 0,95) * 0,999$

title = df.loc[df['n_sections'] != None].title
 sns.heatmap(churndf.Corr())
 with open('load_usaeng', encoding='utf8') as f:
 C = 0
 editorss = defaultdict(int)
 edits = defaultdict(int)
 multilingual = ... same
 edit_by_multilinguals = ...
 pairs = ...
 for l in tqm_notebook(f):
 ll = l.split('\t')
 talen = {ll.split(',')[0]: int(ll.split(',')[1])}
 for ll in l.split('\t')[1:]
 if sum(talen.values()) >= 5 and len(talen)
 talen = {x:y for x,y in talen.items() if y>=5}
 for x in talen:
 if talen[x] > 1:
 editorss[x] += 1
 edits[x] += talen[x]
 if len(talen) > 1:
 multilingual[x] += 1
 edits_by_multilinguals[x] += talen[x]
 for x in talen:
 if len(talen) > 1:
 for i in combinations(talen, 2):
 pairs[tuple(sorted(i))] += 1
 +2 v21

PIVOT TABLES

`df.pivot_table(index='...', columns='...', values='...', aggfunc='...')`

↳ Je kan ook twee indexen meegeven

↳ aggfunc = mean by default

Hoeveel mannen en vrouwen aan boord?

`titanic.pivot_table(index='sex', columns='class', values='survived', aggfunc='sum')`

Hoeveel mensen de ramp hebben overleefd per geslacht, per klasse.

① `titanic.pivot_table(index='sex', columns='class', values='survived', aggfunc='sum')`,
`titanic.pivot_table(index='sex', columns='class', values='survived', aggfunc='len')`

② `titanic.pivot_table(index='sex', columns='class', values='survived')`

Gemiddelde én median per geslacht van overlevende / niet-overlevende

`titanic.pivot_table(index='sex', columns='survived', values='age', aggfunc=[np.mean, np.median])`

Aantal 'woorden' in 'original' kolom

min-max normalization
 $(df - df.min()) / (df.max() - df.min())$

`raw['original'].str.count('woord').sum()`

Selecteren van waarde op basis van 2 kolommen

`df.loc[['kolom 1'], ['kolom 2']] of df.loc[kolom1, kolom2]`

Series maken

`Pd.Series(data='df[...])`

`df.sum(axis=0) → som per kolom`

`df.sum(axis=1) → som per rij`

Pandas Subsets

`df[['c011', 'c012']].columns`

`df['c011'] single column as series`

`df.loc['index-0:raw'] row`

`df.loc['index 2, index2'] rows`

`df.loc[:, rows][cols] rows & columns`

`df.loc['row 1', 'c012'] row & column`

Bestand per regel inladen

with open(bestand) as f:

for l in f:

Detective work

`Prince = pd.read_csv(bestand)`

`Prince['text'] = prince['text'].astype(str)`

`prince['text'].str.contains('...').sum()`

`prince['text'].str.startswith('...')`

`prince['text'].str.findall('...')`

5 ways to count men and women

`titanic.sex.value_counts()`

`titanic.groupby('sex')['survived'].count()`

{s: titanic[titanic.sex==s].sex.count() for s in titanic.unique()}

`titanic.pivot_table(index='sex', values='survived', aggfunc='len')`

`Pd.crosstab(titanic.sex, titanic.survived).sum(axis=1)`

Regex

\n Newline

[...] Range or character class

[^...] not " "

Any character except newline

\w Word character

\W nonword character

\b word boundary

\B non word boundary

i case insensitive matching

.. group subpattern into \1\2

Definitie

`def teller(bestand):`

from collections import defaultdict

telling = defaultdict(int)

With open(bestand) as f:

for l in f:

for c in list(l):

telling[c] += 1

`t-series = pd.Series(telling).sort_values(ascending=False)`

return t-series

By value-counts plot → sort-index

↳ explicit index
`data.loc[0]`

↳ implicit index / python index

`Pd.read_csv(bestand, index_col='...')`

`.set_index()`

`.sort_values(by=...)`

`df.columns=[...]`

↳ rename columns

`.value_counts()`

`.min()`

`.max()`

`.sum()`

`.mean()`

`.mode()`

`.median()`

`.abs()`

pd.series

`df[[col1, col2]].corr()`

`df[[col1]].mean()`

`df[[col1]].mode()`

`df[[col1]].median()`

`df.sort_values`

`df[[col1]].min/max/sum`

`astype(str/float/int)`

`df['colm'].nlargest(r, columns=...)`

precision = $\frac{TP}{TP+FP}$

Recall = $\frac{TP}{TP+FN}$

`Pd.crosstab(col1, col2)`

Min-max normalization

$X^* = \frac{X - \min(x)}{\text{range}(x)} =$

$\frac{X - \min(x)}{\max(x) - \min(x)}$

↳ is by min waarde gelijk aan 0

2-score

$X^* = \frac{X - \text{mean}(x)}{\text{SD}(x)}$

↳ is by mean waarde gelijk aan 0.

2-score is niet geschikt om outliers te identificeren omdat mean & SD worden beïnvloed door outliers.

Be carefull with using correlated variables

- at best → overemphasizes one data component

- at worst → unstable & unreliable results.

0f /series

`.count()`

Str methods

`len()` `startswith()`

`lower()` `endswith()`

`upper()` `isupper()`

`index()` `isnumeric()`

`lstrip()` `split()`

`rstrip()`

plots

Outliers vinden

`df.plot.scatter(x=col1, y=col2, title=...)`

Regressielijn → data=df, fit-reg=True

`sns.distplot(df)`

`af.plot(kind='bar')`

Missing Values

Replace by:

- Constant → creation of info

- mean / mode of field → overoptimistic confidence levels

- Random variable
 ↳ spread should remain closer to original
 ↳ spread will be reduced

Big files

van import gzip
 with gzip.open(bestand) as bla:

for line in bla:

Filling missing values

`df[[col1]].fillna(df[[col1]].`

`mean(axis=0)) → mean`

`df[[col1]].fillna(random.`

`random))`

Regex methods

\d digit character

\D nondigit character

\s whitespace character

\S non whitespace character

^ the start of the line of text

\$ the end of line of text

* match 0 or more items

+ match 1 or more items

? match 1 or 0 items

{n} match exactly n items

`re.match()`

`extract()`

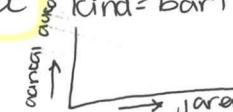
`findall()`

`replace()`

`contains()`

`count()`

`split()`





Data Sciencee

CheatSheet

kolom selecteren

< `DF[["..", "..."]], DF[..]"]`

rij selecteren

< `DF.ix[[...]], DF.ix[.."]`{ `DF.ix[[.., ..], [.., ..]]`

→ kan ook zonder list als enk.

→ put this in `pd.Series()` to get a series objectSeries naar DF => `pd.DataFrame({ "col1": seri, "col2": ser2 })`dict naar DF => `pd.DataFrame(my_dict), key = index`

DF.set_index("name")

DF.sort_values(by = "name") | DF.sort_index()

DF.add_suffix(str)

DF.value_counts() => tellt hoe vaak waarde voorkomt in Series

DF.describe() => count, mean, std, min, 25, 50, 75, max

↳ categorical series => count, unique, top, freq

DF.plot(x = column, y = column, kind = "") ← plotten!!!

pd.pivot_table(), pd.crosstab(row, column), DF.groupby()

↳ margins = True => column & row with "All"

aggfunc = len => ipr.count

get string from Series => .values[index]

remove items from serie1 not in serie2 => `serie1[~serie1.isin(serie2)]`get index with max value in Series => `serie.idxmax()` (also column)

loop door bestand => with open(file) as f:

dataset seaborn inladen => sns.load_dataset(...)

excel file lezen => pd.read_excel

CSV file lezen => pd.read_csv(.., sep = ..)

dict.items() => k, v

dict(zip(list1, list2)) <= dict van 2 lijsten

Als index tuple is, naar multi

pd.MultiIndex.from_tuples(index)

join 2 DF's `DF1.join(DF2)`precision =
 $TP / (TP + FP)$ recall =
 $TP / (TP + FN)$

<pre> dt.nam/shape/slice/astype LISTS DATA x[2,0]=12 x[:,2] x[::2] x[-1] x[:,0] first col x[0,:] first row var[x,y,z==1] # extract odd nos. np.concatenate([x,y,m]) p.vstack([]) p.hsplit() # array + s = ufuncs abs(x)/np.absolute() np.sum()/min()/max() np.any//np.all() np.argsort() np.sort(axis=0/1) np.log2(x) 2 log(8) > 2^3 = 8 np.percentile(series, q=[25,50,99]) np.argwhere(series, 3=0) </pre>	<p><u>LISTS DATA</u></p> <p>dir(pd) with open(file) as f: sns.load_dataset() pd.read_excel()</p> <p><u>viewing outliers</u></p> <p>df.loc[df[X]< df[X].quantile(0.99)]</p> <p><u>plot</u></p> <p>plt.hist(name)</p> <p>df.head(10).plot(kind='barh')</p> <p>df.plot.scatter(x=x, y=y, fit-reg=True, style='o')</p> <p>sns.distplot(np.log(x))</p> <p>df.plot(linewidth=1, loglog=True, useindex=False)</p> <p>histogram → value_counts()</p> <p><u>create set of tuples</u></p> <p>set(zip(df.x, df.y))</p>	<p><u>REGEX</u></p> <p>re.match() contains()</p> <p>extract() count()</p> <p>findall() split()</p> <p>replace() rsplit()</p> <p>word.split("") .title</p> <p>.startswith .capitalize</p> <p>.endswith .swapcase</p> <p>.replace .</p> <p>strip() .lstrip()</p> <p>(strip()) .isspace()</p> <p>Rstrip() .</p> <p>m=re.search('<p>.*?</p>', m.group)</p>
<pre> pd.DataFrame(data) NF→ser: .squeeze lf=pd.DataFrame({'col1':ser1, 'col2':ser2}) lf=pd.DataFrame(ser.values.reshape()) indA & indB = intersection A B = union A ^ B = symmetric difference data[['a','c']] data[['a','c']] data.loc[data.x>2, ['p','c']] .index[0] </pre>	<p>to_frame()</p> <p># relative tussen x en y</p> <p>pd.crosstab(df.x, df.y)</p>	<p>pd.crosstab(df.x, df.y).sum(axis=1)</p> <p>df = pd.DataFrame.from_dict(x, orient='index') re.sub(x,y,str)</p> <p>df.name.unique.size</p> <p>df.pivot_table(index= values= columns= aggfunc= [np.mean, np.median])</p> <p>precision = TP/(TP+FP)</p> <p>recall = TP/(TP+FN)</p>
<pre> df.dropna(axis='columns') df.fillna(0) pd.concat([x,y]) df.append(df2) pd.merge(df1, df2, on='employee', left_on='emp', right_on='emp', left_index=True, suffixes=['_1','_2']) df1.join(df2) </pre>	<p>with open(file) as f:</p> <p>c=0 e=defaultdict(int) for l in tgdm_notebook(f): c+=1.</p>	<p>df.column.contains(r'^')</p> <p>sum</p> <p>aggfunc(ipv count)</p>
<pre> groupby count() min() max() first() std() var() ast() mad() prod() mean() sum() median() sum() </pre>	<p>df.groupby('key').aggregate(['min', 'max'])</p> <p>margin=True first row iloc[0]</p> <p>filter(filter-function)</p> <p># drop data based on</p> <p>apply(function)</p>	<p>titanic.dropna(subset=['age'])</p> <p>r'\b[A-Z]{2}\b{2}\b{9}'</p> <p>prince.text.str.contains(r'\b{ss}\b{ex}\b{1}')</p> <p>percentage kliners:</p> <p>prince.text.str.findall(r'[AEIOUYaeiouy]').st</p> <p>descri</p>

Readfiles | help(pd.Series.loc); dir(pd/np); dir(str);

df = pd.read_csv('file.csv') | Regex.findall; match; extract;
pd.read_excel('file.xlsx') | df.text = Prince.text.astype(str)

with open('file') as file : | begin met sex → Prince.text.str.contains(r'\b[Ss]ex\b')
for page in file : | bevat sex → Prince.text.str.contains(r'\b[Ss]ex\b')
df.sort_values(by='col') | songs met woorden waar sex in voorkomt
if sort_index() | Prince.text.str.findall(r'[Ss]ex\w+')

with open('file') as f : | el[et.str.len() > 0] → laat zien of count()
for line in f: | Counter([s for i in et.values for s in i])
 regels = line.split('\t') | woorden met letter encipher erin
 languages = {i.split('!')[0]: int(i.split('!')[1]) | findall(r'\b[A-Za-z0-9]*[0-9]+[A-Za-z0-9]*[A-Za-z][A-Za-z0-9]*[0-9]+[A-Za-z0-9]*\b')}

for i in regels:

if sum(languages.values()) >= 5 and len(language) <= 5:

languages = {language: value for language, value in languages.items()}

if int(value) >= 23

for language, value in languages.items():

editors[language] += 1

edits[language] += value

pd.DataFrame.from_dict(dict, orient='index')

Samenvoegen = pd.concat([dataf1, dataf2], join='inner', axis=1)

merge = df1.merge(df2, how='inner', ...)

Selecteren |

df[df['col'] == 'voorwaarde'].sort_values('col1', ascending=True)

df.loc[row, col] → df[, col] of gewoon df['col']

integer: df.iloc[0] → df.iloc[0]

by label → df.loc[[0], 'Colname']

by label/posit → df.ix[[index], 'colname']

bool index → s[s > 1] = Series where val is not > 1

s[(s < -1) | (s > 2)] = s where val is < -1 or > 2

df[df['population'] > 1200000] → Filter df

df.loc[:, (df != any())] → sel col any

.all() → sel col all

Tellen | len(df[df['col1'] == 'value']) | df.groupby('value').size()

count_values() | df.pivot_table(index, val, aggfunc='len')

dictcom = {x: df[df.col == x].col1.size() for x in df.col.unique()}

voortbouw | pd.crosstab(df.col1, df.col2)

extclass | Sex class

per | df.groupby(['sex', 'class']).count().unstack()

hoeveel deel overleefd van sex per class

titanic.Pivot_table(index='Sex', columns='sex', values='survived')

gem en mede van geslacht van overlevers - niet overlevers

titanic.Pivot_table(index='sex', columns='survived', values='age', aggfunc=[np.mean, np.median])

gen met nan eruit:

df = titanic.dropna(subset=['age'])

lef teldeletters(bestand):

from collections import defaultdict

telling = defaultdict(int)

With open(bestand) as f:

for l in f:

for c in list(l):

telling[c] += 1

tellingseries = pd.Series(telling).sort_values()

return tellingseries

teldeletters(bestand).plot(kind='bar', figsize(15, 7), logy=True);

df.replace(1, one)
df.dropna(subset=...)
df.describe()
df.corr(col1, col2)
df.min/max/std
S['col'].str.lower()
df.held.str.replace
df.towl
df.shape
df.info

1\b[A-Za-z0-9]*[0-9]+[A-Za-z][A-Za-z0-9]*\b
1\b[A-Za-z0-9]*[A-Za-z][A-Za-z0-9]*[0-9]+
[A-Za-z0-9]*\b'

klinkers vinden:.findall(r'[AEIOUYaeiouy]')

. → any char except line break
|w → word char, digit, _ . + period
|s + white space
|d → digit 0-9
+ → one or more
* → zero or more
? → once or none
t\nr → tab, newline, return
[abc] → a/b/c [^ABC] → alles behalve
re.sub(reg, vervanging, string)

from pandas.util.testing
import

test if equal → assert_frame_equal
(df1, df2)
stack() → level → stack col to row
level0 → upper col is stack to row
unstack() → row to col
crosstab
pd.crosstab(df.col1, df.col2) Zet beide als
groupby([col1, 'col2']).count().unstack().&
grouping → pd.crosstab(df.col, [df.col1, col2])
query → df.query("col1 < col2 < col3")
| → and, or; col1 in [val, val, val]

concat(df1, df2), axis=0,1
append: df1.append(df2)

pd.merge(df1, df2)

merge on index → left_index=True

error plot vrg 4 w7

error = df.Percentage - df.Percentage[aa]

Percs = dt.Percentage

Percs.plot, bar(yerr=error)

gemiddelde aantal woorden per sectie

mean = (df['n_words']) / df['n_sect']

log = np.log2(means)

plot = sns.distplot(log)

boven 20

df.loc[(col['x'] > 20)]

gem zonder ran df.means(skipna=T)

gender count

{P: Counter(genderlist[P][:N]) for P
in genderlist}

PVU25 zetels, hoeveel vrouw

Counter(mvdict['PVU'][25]['female'])

Series. index	$\begin{aligned} \text{Index as ordered set:} \\ X = 1, 3, 5, 7, 9 \\ Y = 2, 3, 5, 7, 11 \\ \rightarrow X \& Y = 3, 5, 7 \\ \rightarrow X Y = 1, 2, 3, 5, 7, 9, 11 \\ \rightarrow X ^ Y = 1, 2, 9, 11 \end{aligned}$	<p><u>REGEX</u></p> <table border="0"> <tr><td>1 start</td><td>^</td></tr> <tr><td>\$ end</td><td>\$</td></tr> <tr><td>\w word</td><td>\w</td></tr> <tr><td>\W non-word</td><td>\W</td></tr> <tr><td>\S non-space</td><td>\S</td></tr> <tr><td>[abc]</td><td>[^abc]</td></tr> </table>	1 start	^	\$ end	\$	\w word	\w	\W non-word	\W	\S non-space	\S	[abc]	[^abc]	<p>$\sim \rightarrow \text{not}$</p> <p>defaultdict ↳ from collections</p>
1 start	^														
\$ end	\$														
\w word	\w														
\W non-word	\W														
\S non-space	\S														
[abc]	[^abc]														
Series. values															
DataFrame.columns															
data.items() → tuples															
data['a': 'b'] → includes b.															
data[['a', 'b']]															
data.loc[1:3] → index-based	<ul style="list-style-type: none"> • dropna / fillna() • notnull() → all non nan values. • query ("query") → @var • sort_values() by = • mean(axis='columns') → mean over rows • count() → number of items. • pivot_table('survived', index='sex', columns='class') • pivot_table(index='sex', columns='class', aggfunc='sum'). • str.match → bool • extract → matched groups • findall → list • replace • contains → bool • count → occurrences of pattern. • split • rsplit • get_dummies → indicator var split. • sort_index() • sort_values() • crosstab • hist • df.col.value_counts().plot.bash() • if.itertrows() • pd.merge • df.col.quantile • df. pd.DataFrame.from_dict(dict, orient=index, columns=[]) 	<p>elem.findall(text)</p> <p>f = archive.open(file)</p> <p>root = etree.parse(f).getroot</p> <p>node.find</p> <p>node.findall</p> <p>BS.find(')').text</p> <p>BS.findall</p>	<p>5 ways:</p> <ul style="list-style-type: none"> • value_counts() • groupby('')[].count() • pivot-table aggfunc='len' • crosstab().sum(axis=1) 												
data. iloc[1:3] → python-based															
data.values[0] → first col.															
data.loc[data.dens > 100, [x, y]]															
a.add(b, fill_value=0)															
→ sub(), mul(), div(), mod(), pow()															
• notnull() → all non nan values.															
• dropna / fillna()															
MultIndex.unstack() → back to df.															
df.set_index([arr]) → multi index															
df.append(df2)		<p>df.col.str.contains().sum()</p> <p>(Counter.most_common)</p> <p>index-col sep.</p> <p>filteren met booleans:</p> <p>df [filters.values]</p>	<p>df.col.str.contains().sum()</p> <p>(Counter.most_common)</p> <p>index-col sep.</p> <p>filteren met booleans:</p> <p>df [filters.values]</p>												
pd.merge															
df.col.quantile															
df. pd.DataFrame.from_dict(dict, orient=index, columns=[])															

<p>(df.col1 == 'test').sum() \rightarrow 0</p> <p>df[(col1 == 'test') & (df.col2 > 5)] \rightarrow DF</p> <p>df[['col1', 'col2']] \rightarrow DF OF 2 cols</p> <p>df.values \rightarrow DF in list without index</p> <p>list.items \rightarrow error 'too many values to unpack'</p> <p>df.set_values('col1')</p> <p>Pd.Series(df.col1, index=df.index)</p> <p>Pd.deop(columns=['col1', 'col2'], axis=1)</p> <p>Pd.DataFrame.from_dict(dict, orient='index', columns=['col1'])</p> <p>Pd.concat([dict1, dict2], axis=1, join='inner')</p> <p>df1.join(df2)</p> <p>((abs(df[col1] - df[col2])) / df[col1]).mean() * 100 mismatch %</p>	<p>df.mean()</p> <p>df.median()</p> <p>df.mode()</p> <p>df.count()</p> <p>df.first()</p> <p>last()</p> <p>min()</p> <p>max()</p> <p>std()</p> <p>var()</p> <p>mad()</p> <p>prod()</p> <p>sum()</p> <p>titanic.sex.value_counts() SEXES IN TITANIC</p> <p>- titanic.groupby(['sex']).survived().count()</p> <p>- {s: titanic[titanic.CSex == s].Sex, count} for S in titanic.sex.unique</p> <p>- titanic.pivot_table(index='sex', values='survived', aggfunc='sum')</p> <p>2) titanic.groupby(['Sex', 'class'])['survived'].sum(axis=1)</p> <p>titanic.pivot_table(index='sex', values='survived', columns='class', aggfunc='mean')</p> <p>3) titanic.groupby(['sex', 'survived'])['age'].agg(average)</p> <p>4) Pd.crosstab(titanic.Pclass, titanic.class)</p> <p>5) test = titanic.deptron(subset=['age'])</p> <p>test = test[test.sex == 'male']</p> <p>Pd.crosstab(test.age, test.adult_male)</p>
<p>df.loc?</p> <p>df.col1.quantile(q=0.99)</p> <p>df.nona = df.deopna()</p>	<p>b) bestand = 'Prince_lyrics.csv'</p> <p>Pd.read_csv('bestand'):</p> <p>From collections import defaultdict</p> <p>telling = defaultdict(int)</p> <p>With open(bestand) as f:</p>
<p>df.fillna(0, inplace=True)</p> <p>df.col1.astype(int)</p> <p>df['Percentagecol'] = round((df.col1 / (df.col1 + df.col2) * 100), 0)</p> <p>df['mannen'] = df.Zeitens - df.Vrouwen</p>	<p>for l in f:</p> <p>for c in l[1:]:</p> <p>tellings[c] += 1</p> <p>tellings = Pd.Series(tellings).sort_values(ascending=False)</p>
<p>1) df = Pd.read_csv('xxx.csv', index_col='names') TENTAMEN 1</p> <p>df.sort_index(inplace=True)</p> <p>2) df = df.Rfillna(0)</p> <p>3) df['assmean'] = df[['ass1', 'ass2']].mean(axis=1)</p> <p>4) (df[['ass1', 'ass2']] > 9).sum().sum()</p> <p>5) df['deelmen'] = 0,4 * df[['assmean']] + 0,6 * df[['deelmen']]</p> <p>6) z_df = (df - df.mean()) / df.std()</p> <p>7) df.sort_values(['col1', 'col2'], ascending=(False, True))</p>	<p>A: hoeveel nummers bewerken "sex"</p> <p>Prince.text.str.contains(r'\b[ss]ex\b').sum()</p> <p>B: hoeveel nummers beginnen met 'sex'</p>
<p>8) core = df[['col1', 'col2']].corr()</p> <p>core = core[core != 1]</p> <p>rij = core.max().idxmax()</p> <p>kolom = core.loc[rij].idxmax()</p> <p>vercap = (rij, kolom)</p>	<p>9) Prince.text.str.startswith('Sex').sum + lowercase</p> <p>Sex = Prince.text.str.findall(r'[ss]ex\w+')</p> <p>print(Sex[Sex.str.len() > 0].head(10))</p> <p>C: hoeveel nummers bewerken sexy etc?</p>
<p>core = df[['col1', 'col2']].corr()</p> <p>core = core[core != 1]</p> <p>rij = core.max().idxmax()</p> <p>kolom = core.loc[rij].idxmax()</p> <p>vercap = (rij, kolom)</p>	<p>D: print all die woorden uit</p> <p>From collections import Counter</p> <p>print(Counter([s for s in sex.values for s in 1]))</p>
<p>E: u2 hu etc</p> <p>ef = Prince.text.str.findall(r'\b[A-Za-z0-9]+\w+\b')</p> <p>print(ef[ef.str.len() > 0].head(10))</p> <p>F: % links per track</p>	<p>E: u2 hu etc</p> <p>ef = Prince.text.str.findall(r'\b[A-Za-z0-9]+\w+\b')</p> <p>print(ef[ef.str.len() > 0].head(10))</p> <p>F: % links per track</p>
<p>Hoofdstuk 3</p> <p>df.loc[:, :2]</p> <p>df.isnull()</p> <p>Pd.merge(df1, df2)</p> <p>df.unstack()</p> <p>df.groupby('col1').aggregate([min, np.median, max])</p>	<p>(100 * Prince.text.str.findall(r'\b[AB]louder\b'))/str.len()</p> <p>Prince.text.str.len().describe()</p>
<p>titanic.groupby(['Sex', 'class'])['survived'].agg(average)</p> <p>.unstack() \rightarrow titanic.pivot_table('survived', index='Sex', 'age', columns='class')</p> <p>titanic.pivot_table('survived', index='Sex', columns='class')</p> <p>age = pd.cut(titanic['age'], [0, 18, 80])</p> <p>titanic.pivot_table('survived', index=['Sex', 'age'], columns='class', margins=True) \rightarrow margins give totals</p> <p>births.pivot_table('births', [births.index.month, births.index.day])</p>	<p>[abc] anything abc [^abc] anything not abc a-z A-Z 0-9</p> <p>\s whitespace \d digit \w word \b word boundary </p> <p>() capture ^ start of string \$ end of string RegEx</p> <p>a? zero or one a* zero or more a+ one or more</p>
<p>See = Pd.Series(['Jan Smith', 'John Kaas'])</p> <p>See.str.lower().str.strip().str.capitalize()</p>	<p>A = alcoholist B = man</p> <p>P(A ∩ B) = 0,0225</p> <p>P(A ∪ B) = P(A ∩ B)</p> <p>Conditioner P(B)</p> <p>Kans = 0,0225</p> <p>0,5</p> <p>= 0,045</p>
<p>P.Percentage.plot('bar', yerr = (df.Percentage - df.Percentag).tolist(), **extra)</p> <p>=(Q3-Q1)/1,5</p> <p>Q1 - x en Q3 + x = data</p>	<p>X - min(x)</p> <p>max(x) - min(x)</p> <p>min-max normalization</p>
<p>df.col1.fillna((df.col1.mean()), inplace=True)</p> <p>for b in freqsec:</p> <p>chancedict[b] = dict()</p> <p>valuedict = {}</p> <p>for a in freqsec:</p> <p>numdict[a] = len(dummydf[dummydf[b] == 1] & (dummydf[a] == 1))</p> <p>valuedict[a] = numdict[a] / len(dummydf[dummydf[b] == 1])</p> <p>chancedict[b] = valuedict</p> <p>condchance = Pd.DataFrame(chancedict).sort_index(axis=1)</p>	<p>X - Mean(x)</p> <p>STD(X)</p> <p>z-score normalization</p>
<p>freqsec = freqsec[:df.nlargest(20).index]</p> <p>dummymdf = df.set_index('id').apply(Pd.Series).stack()</p> <p>freqsec = [x for x in freqsec.values]</p> <p>dummymcols = Pd.DataFrame(column=freqsec)</p> <p>for col in freqsec:</p> <p>dummymdf[[col]] = dummymdf[df]</p>	<p>Re. findall</p> <p>Re. sub</p> <p>Re. match if beginning with (+F)</p>
<p>dummymdf = dummymdf[[col]]</p> <p>dummymdf.groupby(level=0).sum() == 1</p> <p>chancedict = dict()</p> <p>Re. findall</p> <p>Re. sub</p> <p>Re. match if beginning with (+F)</p>	<p>Re. split</p>

- NaNs:
 - Replace with constant
 - Replace with mean/median with random imputed value
 - changes center and spread
- Misclassifications: value counts
 outliers: histogram, boxplot, scatterplot
 center measures: mean, mode, median
 Variability: Range (max-min) STD, MAD, IQR
 Normalization (reduce variation):
 • Min-Max: $x^* = \frac{x - \text{min}(x)}{\text{range}(x)}$
 alle 0-1.
 • Z-score: $z = \frac{(x - \text{mean}(x))}{\text{SD}(x)}$
 mean \rightarrow mean +
 Kan leiden tot overval, if not skewed.
 • Decimal: $x^* = \frac{x_{\text{old}}}{\text{max}(x)} \rightarrow d = \text{decim.}$
- Skewness: $\frac{3 \cdot (\text{mean} - \text{median})}{\text{SD}}$
 right mean > median
- Transformation (reduce skewness)
 • log, \sqrt{x} , $1/\sqrt{x}$
- outliers: if $z_{\text{sc}} > 3 \text{ or } < -3$.
 • Z-score: ~~if $z_{\text{sc}} > 3 \text{ or } < -3$~~ .
- IQR: Q1: 25% Q2: 50 (med)
 Q3: 75%
 IQR = Q3 - Q1 $\rightarrow Q_1 - 1.5 \cdot IQR$
 IQR = $Q_3 + 1.5 \cdot IQR$
 outlier if x located
 1.5 IQR below Q1, above Q3
- num \rightarrow cat: binning (K-means e.g.)
 cat \rightarrow num: dummy var.
- unsupv.: clustering/grouping
 reduces complexit.
- supv: regressie, classificatie
 3: reinforcement learning
- evaluatie: verg. met gouden std.
 class.: Acc., Pre, Rec., MRR ...
 Regn.: (R)MSE, AVE ...
- Acc: how many predicted were correct?
 Prec: how many pos. pred were com
 Rec: how m. pos. pred. as pos? $\frac{TP}{TP+FN}$
- | | | |
|------|-------------|-------------|
| ziek | + | - |
| Test | + TP FP | - FN TN |
| | alle ziek | gezond |
- Train, Val, test
 weinig data: k-fold.
- men/women titanic:
 t.sex.value_counts()
 t.groupby('sex')[survived].count()
 t.pivot_table(i='sex', v=survived, aggfunc='len')
 t.crosstab(t.sex, t.survived).sum(axis=1)
 per geslacht / per klasse / daal surv.
 t.pct(i='sex', v=survived, c='class')
 t.pivot(i='sex', c='class')[survived].agg('mean')
- gen / men. vergen sun =
 t.pct(i='sex', aggfunc='sum')
 c = surv. val = age, aggf (np.mean)
- tel de letters:
 from col.. import defaultdict
 with open(file, encoding='utf8') as f:
 for line:
 for c in line:
 tel[c] += 1
 tel = pd.Series(tel).sort_values(...)
 return tel
- Regex:
 • str.match \rightarrow bool
 • extract \rightarrow strings
 • findall \rightarrow replace
 • contains \rightarrow returns bool
 • count \rightarrow count occ. of part.
 [abc] inv: [^ABC]
 \\$: whitespace \d: digit
 \w word (alb) a m b
 \A a?: a / a
 a*: 0+ a at: 1+ a
 a{3}: 3 ^ start of str
 \\$ end of str \b wordbound.
 *? non greedy
- n numbers - sex -
 Prince.txt.str.contains('b[es]or[b]').sum()
- n numbers starts with - sex -
 .contains('b|^es|or[b]')
- n numbers before - sex -
 .contains('([es]|or[b])w+').sum()
- Print alle woorden die - sex --
 p(camer(x for y in P for x in y))
- Perct. klinkers:
 findall('aeiou'), str.len() / pt.txt.str.len()
- (Rij, kolom) . Axis 0: within each col. temp_dict[a] = len(df[[a]] for x in f-s):
 df2[x] = df - df
- Ptot = df2 = freq - sec
 dum_d = bcdem.glob(level=0) si
 chance_dict = dict() =
 for b in frequent_sections:
- lamste sta: 1. lang = str()
 2. pak regel, meet len() vergl. len(lang)
 3. if >, lang = 2 \rightarrow next
- Age = pd.cut(titanic.age, [0, 18, 80])
 t.pct(survived['Sex', 'Age'], 'class')
- com: df1.com(df2) -> $A \cup B = \frac{A \cap B}{A \cup B}$
- Jaccard $\frac{A \cap B}{A \cup B}$
- Conditionele kans $P(A|B)$
 = $P(A \cap B) \rightarrow$ aantal A en B is
 P(B) \rightarrow totaal aantal B's
- ice flows dat N plaats vindt, gev.
 dat Book ...
- P(choog sal).H2P(youth) \rightarrow P(meuh)
- Regressie: sns.lmplot(x, y, data, fitreg=True)
 df \rightarrow multi.index \rightarrow set_index((cde1, cde2))
- Concat: row-wise default
 short: df1.append(df2)
- Merge: join autom. on key on cols.
 merge(df1, df2, on='coll')
- left on, right on ...
 default on indexes \rightarrow df1.join(df2)
- Pivot-tables (values, index, cols, agg)
- groupby. aggregate(col): mean, col2: max
 pd.qcut \rightarrow quantiles
- Pivot-tables can be multi leveled -
 Rgex: re.findall(regex, text) \rightarrow lis
- match \rightarrow determine if begin string
 • split \rightarrow list
 • sub \rightarrow replaces
- f-s = setitemlist(sf).nlargest(n)
- dummy_d = df.sections.apply(pd.Series).stack
 cols = [x for x in frequent_sections.value]
- df2 = pd.DataFrame(columns=cols)
- for x in f-s:
 df2[x] = df - df
- Ptot = df2 = freq - sec
 dum_d = bcdem.glob(level=0) si
 chance_dict = dict() =
 for b in frequent_sections:
- lamste sta: 1. lang = str()
 2. pak regel, meet len() vergl. len(lang)
 3. if >, lang = 2 \rightarrow next
- Age = pd.cut(titanic.age, [0, 18, 80])
 t.pct(survived['Sex', 'Age'], 'class')
- com: df1.com(df2) -> $A \cup B = \frac{A \cap B}{A \cup B}$
- Jaccard $\frac{A \cap B}{A \cup B}$
- Conditionele kans $P(A|B)$
 = $P(A \cap B) \rightarrow$ aantal A en B is
 P(B) \rightarrow totaal aantal B's