

Day  
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31

Month  
1 2 3 4 5 6 7 8 9 10 11 12

Subject

Sensitivity, recall, hit rate (TPR)

$$TPR = \frac{TP}{P} = \frac{TP}{TP+FN}$$

Precision (PPV)

$$PPV = \frac{TN}{N} = \frac{TN}{TN+FP}$$

Accuracy (ACC)

$$ACC = \frac{TP + TN}{P + N} = \frac{TP + TN}{TP + TN + FP + FN}$$

F1 score

$$F1 = 2 \cdot \frac{PPV \cdot TPR}{PPV + TPR} = \frac{2TP}{2TP + FP + FN}$$

Nuttige Funktionen numpy

np.concatenate

.vstack/hstack

.r\_

.where() returns indices

select ( $\Rightarrow$  pandas select)

pandas pivot  
crosstab

margin = total

agg = np.[mean,

sum,

median]

np.array [for row, col]

from numpy import doc

np.lookfor('query')

np.tile

np.reshape

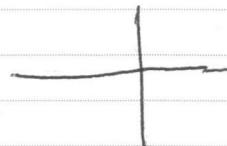
np.repeat

~~TPR = 4/100~~

$$\frac{TN}{TN+FP}$$

$$\frac{0.90}{0.98}$$

h



## NP. array

X	

NP  
· .shape  $[a[0, 2]]$

· .size  $a[0:2]$

X	X	
---	---	--

· .add(a, b)

· .arange(0, 100, 10)  $a[a < 2]$

$a + * / b$

· .reshape (keep)

· .resize (delete) · .repeat()

0	X	1	X	2	3
---	---	---	---	---	---

From numpy import doc

## Pandas

df = pd.DataFrame([ [1, 2, 3],  
[4, 5, 6],

PD  
· .query('a >= 100')  $[7, 8, 9]$ , index = [1, 2, 3],  
· .rename('val', 'value')

? np.doc

np.lookfor

· .melt(df)  $\boxed{\quad}$  to  $\boxed{\quad}$

	a	b	c
1	1	2	3
2	4	5	6
3	7	8	9

~~· .pivot~~  
df.pivot(col=, val=)  $\boxed{\quad}$  to  $\boxed{\quad}$

A

· .concat([df1, df2]) axis=0 rows  
axis=1 columns  $a+b = \text{error?}$   
 $\downarrow$  broadcasting apply

· .dropna (columns=)

df[df.length > 7]

· .dropna

df[df['length'] > 7]

· .sort\_values

.value\_counts()

· .sort\_index

.unique()

· .head

.Count

· .tail

sum, mean, min, max, (med), std

· .fillna(x)

.plot, hist

· .nlargest(n, 'value')

.read\_html(link)

· .nsmallest(n, 'value')

.groupby(by="col")

# CHEAT SHEET #1



**NP** ( $y, x$ ) ,  $\text{np.zeros} - \text{np.ones} - \text{np.full}()$  -> can shape  
 ↳  $y$  ↳ column  
 ↳  $x$  ↳ np.arange() -> cannot shape, is linear & np.linspace() (evenly spaced between given values).  
 else: np.random.random() - np.random.normal() - np.random.randflat()  
 => creating arrays  
 $x[0, 0] = 12$  -> changing value of that index to 12 -  $x[2].astype('float')$  ->  
 got to assign to new variable -  $x[start:stop:step] = x[::-1]$  -> all elements reversed -  $x[3, ::2]$  -> all rows, every other column

accessing column:  $x[:, 0]$  -> first column of  $x$  - accessing row:  $x[0, :]$  -> first row of  $x$ . -> if you modify a subarray, the original is changed -> use .copy combination of np.arange(). reshape -> can get a multidimensional something

np.concatenate([array1, array2, axis=1]): joining two or more arrays  
 np.hstack() & np.vstack() -> add the array according to columns  
 np.mathematics:  $x_1 - x_2, x_1 + x_2, x_1 / x_2, x_1 * x_2, \text{np.sqrt}()$ ,  
 np.log(), -> must be the same shape, np.floor\_divide(), %

Python built-in absolute value abs() -> np.absolutel()  
 np.reduce() -> applies a given operation to elem of an array until only a single result remains., note: np.cumulative() stores all intermediate results of the operation - np.multiply.outer(x, x) (x = np.arange(1, 6)) -> create multiplication table.

$x[3].min(axis=0)$  -> min value across each column. - np.sum / prod / mean / std max/min/avg/argmin/argmax / median/any/all have nan-safe versions, just put nan in front of x now  
 ↳ finds index

Broadcasting: array - array is element by element basis if arrays are same size  
 $(m, n) + (n, 1)$  -> copies m times to  $(m, n)$  use [:, np.newaxis]  
 $\star (m, 1)$  -> copies n times to  $(m, n)$  np.sum([index > 1 & index < 1])

np.sum(x < 6) -> counts how many are bigger than 6  
 np.sum(x <= 6) -> counts values in x, if smaller than 6 [ :, np.newaxis ]  
 fancy indexing: shape of result reflects shape of indexing array i.e. shape of array being indexed - np.argsort() & np.partition(x, 3) -> smaller = 3 values to the left, let's do it! & 2 [ ] also works in mask merging

**PD** df.loc[index, df.value = df.column = list(series.items()) - d][new column] = ..., string by explicit index series["a": "c"] -> final index = included as implicit index - data.loc[1] -> finds explicit index -> data.loc[1] finds implicit index - add new column: data['density'] = data['pop'] / data['area']  
 Series & columns: df.T - passing a single index to off. values gives you a row  
 passing a single index to df passes a column. Use loc to slice a df. -> the ix indexer allows you to do both! data.loc[data['density'] > 100, ['area', 'density']] -> columns you want in resulting dataframe. Indexing refers to columns & slicing to rows  
 A.add(B, fill\_value=0) -> makes sure adding doesn't result in NaN's use A.stack.mean() to get the overall mean, (column wise); A+B column wise A.add(B, axis=0)

isnull(), dropna(), fillna(), notnull()  
 ↳ drops all columns containing null values, how = all -> drops where whole row or column is nan, unstack() multi-index -> regular and vice versa

data.set\_index() - pd.concat(axis=..., ignore\_index=True/False, join='inner'), append(), pd.merge(on..., left\_on="employee", right\_on="name"), df.join() -> merge that defaults to joining on indices.

pd.merge(df1, df2, left\_index=True, right\_index=True), merging df's with not of set\_index(), pd.merge(df1, df2, how="outer") also left and right, merge with overlapping column names? pd.merge(df1, df2, how="outer", suffixes=[ "L", "R" ]), pd.read\_csv - df.value\_counts, lambda x: x + 10, isnull().any() - check if only value is missing, np.argmax() / sort df.sort\_index(by="F") -> if pct (kind="bar") -> df.groupby("oplossing")['aantal/konditien'].sum(), pd.to\_datetime()

df[df['plaats'].str.contains('amsterdam')]

precision: how many of pred of class were correct? TP / (TP+FP) - recall: how many instances of class (predicted to be C) TP / (TP+FN), accuracy: how many of all values are correct TP + TN / (TP + TN + FP + FN)

sort\_values(by="F")

reflex: str.replace(['a-20-g'], 'A') / [ 'TN' ]

str.replace("allen/borden", "A")

Multi-index: pop[0, 200]

from numpy import  
 np.lookfor()

neg. Pos neg

Pos TP FP

Neg FN TN

xx(x, y % k)

## TP/TP+FP

Precision: how many of the predictions for class c were correct (hoeveel van het totaal van wel/niet zieke heeft een pos/neg uitkomst)

Recall: how many instances of class c predicted to be c ( $TP/TP+FN$ )

getest

daadwerkelijk positieve zielte

hoeveel van de True class zijn 'opgepikt' door de test.

Accuracy: how many of all predictions were correct.

Hoeveel van het totaal van zieke/niet zieke heeft een pos/neg uitkomst

zicht:  $P = 10\%$   $Acc = 90\%$

		had nodis			
		TP	FP	$\frac{90}{90+90} = 0.5$	
DT	TP	90	90	$\frac{10}{10+820} = 0.9878$	
	FN	10	820	$10\% \text{ (ziekte)}$	
		100	900		

Precision (kans, accuracy):

$$TP = \text{kans} * \text{accuracy}$$

$$FP = (1 - \text{kans}) * (1 - \text{accuracy}) \quad FN = (1 - \text{acc}) * \text{kans}$$

$$\text{return } TP / (TP + FP)$$

--- --- --- --- --- axis=0 (rowwise)  $\downarrow$  axis=1 (kolomwise)  $\rightarrow$

	A	B	
0	↑		axis=1 $\rightarrow$
1		↓	
2	axis=0		

$df = df[df != 1] \rightarrow$  filter alle 1 eruit en zet die op NaN |  $\text{fillna}(0, inplace=True) \rightarrow$  NaN vervangen met 0's en opslaan

$df.max() \rightarrow$  geeft hoogste waarde per kolom in df

Pd. Value\_counts  $\rightarrow$  counts unique values

$df.idxmax() \rightarrow$  geeft index & van hoogste waarde in kolom (ut zwr v. axis)

V.b.  $\downarrow$  van kolom over dat

$df[["col1", "col2", etc]] \rightarrow$  selectie op kolommen  $\xrightarrow{\text{Kies uit}} df.loc[:, ["col1", "col2", etc]].mean(axis=1)$

$\dots = df[(df["col1"] < 5) \mid (df["col2"] > 6)] \rightarrow$  selectie op kolommen met voorwaarde

from numpy import doc  
Np. Lookfor("")  
Lookfor

matrix normalisieren:  $P / P.sum(axis=0)$

$$P = \sum_{i=1}^I$$

diagonale matrix:  $P[[0, 1, 2, \dots], [0, 1, 2, \dots]]$  de diag grote met reshape

Loop door kolommen (matrix): for  $x$  in range( $\text{len}(\text{matrix}), \text{shape}[1])$ :

$$L[:, x][\text{getal}] = \dots \quad \text{Np.isnan}(\text{check input})$$

flatten: P.flatten() maakt een 1D array

Np.array[...]. $\tilde{}$  2 of "..., etc. geeft verdelen langs (hier x later gedaan etc.)

L[Np.newaxis(L)]  $\rightarrow$  Selectie op de Numpy array (rechts panden)

$$a = \text{Np.arange}(1, 4)$$

$$b = \text{Np.arange}(0, 4)[:, \text{Np.newaxis}]$$

$$a+b = [[1, 2, 3, 4],$$

$$[3, 4, 5]$$

$$[4, 5, 6]]$$

$$L = [[1, 3, 4, 5],$$

$$[2, 6, 1, 8]]$$

$\downarrow$  tweede

$L[:, 1] \rightarrow$  eerste kolom

$L[:, 0] \rightarrow$  tweede kolom

%als de directie. misleidt

%fn ... Vervangen ...

%time %timeit %prun

Pd (Series / Dataframe)

NP  $\times [\text{start}, \text{stop}, \text{step}]$  (-1 for reverse)

Calc.

mean / mode / median

odd / substract / mod / power

Sum / min / max (arg / new)

Any / all / percentile

Sort / partitioning (arg for k-means)

reorder

random / copy

orange

recharge / concatenate

shape

(V / H) (Stack / split)

transpose

roll

linspace

full

test

unique

reduce / accumulate

Outer

Pd (Series / Dataframe)

read\_csv

Doc import doc  
Loop for ..

crossval / pivotable

ix

(index int or bytes mix)

dropna

value\_count

sort\_index / sort\_values

mean / mode / median

corr

groupby . fume -  
nlargest

replace

lower

replace

lower

replace

lower

replace

lower

count

len

P =  $\frac{T_P}{T_P + T_D}$

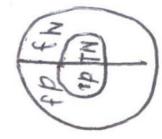
test

reduce

accumulate

Outer

Pd . Series . str . IETS ?



P =  $\frac{T_P}{T_P + T_D}$

np.zeros ((3,4))

np.ones ((2,2))

np.full ((2,2), 7)

a.shape = array of dimensions

len(a) = length of array

b.ndim = num array dimensions

c.size = num array elements

a - b → np.subtract (a,b)

b + a → np.add (b,a)

a / b → np.divide (a,b)

a \* b → np.multiply (a,b)

a.sort () → sort an array

c.sort (axis=0) → sort the elements of an array's axis

np.sort (c, axis=0) → sort each column of c

np.mean (x) / np.mean (x[0])

np.mean (x, axis=1)

↳ gem.

P is een multidimensionale array

P.flatten () → nu 1 dim

L [np.nonzero (L)] → alle elementen die niet nul zijn uit L

x.namen.value\_counts ()

naam1 370 ↗  
naam2 198

b = np.arange (14)[:, np.newaxis]

array ([1, 2, 3], [1, 2, 3], [1, 2, 3])

a[:, :, 0, 2]  
kolommen nieuwe volgorde

np.sum (P < n)  
↳ hoeveel waarden kleiner dan P

ap.isnan  
↳ bool, returned True / False

df.iloc [2,1]

↳ selecting specific data

df.loc [df['a'] > 10]

↳ selecting rows meeting logical condition

df['W'].value\_counts ()

↳ count number of rows

df["New-Column"]

= (a.4 \* 2) + (0.6 \* 1)

data [data > 0.3] & (data < 0.7)]

data [~ data.isin (data2)]

↳ data die niet ook in data2 staat

df.groupby(['Animal']).mean()

KVR.kvrs.Party.str.contains ("vragen").

↳ replace

KVR = KVR [~ KVR.Party.str.contains ("vragen", na=False)]

KVR.Party.value\_counts().sort\_index ()

a.shape = array of dimensions

len(a) = length of array

b.ndim = num array dimensions

c.size = num array elements

np.exp (b)

np.sqrt (b)

a.sum ()

a.min ()

b.max (

axis = 0)

↳ max of row

a.mean ()

np.std (b)

a.sum ()

• mean (1)

= gerelateerd over kolommen

np.all → all

elements along a given axis must be true

np.any → any

↳ horizontaal = columns : axis = 1  
↳ verticaal = rijen : axis = 0

a[::-1] → reversed array a

a[a < 2] → select elements from a less than 2

i.transpose () → rijen en kolommen verswisselen

g.reshape ((3,3)) ↳ geen negatieve

np.append (h,g)

np.delete (a, [1])

np.concatenate ((x,y), axis=0)

↳ samenvoegen arrays

np.intersect1d (a,b)

↳ get common values

np.setdiff1d (a,b)

↳ from a remove all b

np.where (a == b)

↳ positions where a & b match

df["Ass-mean"] = df[df.columns[0:6]].mean (axis=1)

x.min (axis=0)

↳ min / column

x.plot (kind=

"Bar") ;

all = df[df.columns[0:6]]

greater-than-g = all [all > g].count ().sum ()

df.sort\_values ([ 'Name', 'Age' ], ascending = [False, True])

df.sort\_values ([ 'Name', 'Age' ])

df.sort\_index ()

df.drop (columns = [ 'Length' ])

pd.concatenate ([df1, df2], axis=1)

↳ reshaping

df3 = pd.merge (df1, df2)

↳ combine datasets

df.fillna (0)

df.dropna (subset = ["Age"], inplace = True)

df.values ()

↳ raw data

df.transpose ()

↳ rijen x kolommen omdraaien

len(df) → num of rows

str.lower() / str.replace ()

x = x.drop (6, axis=1) ↳ kolom 6 weghalen

np.array([1, 2, 3])

a.ndim → Dimensions

np.info

a[2] = 

1	2	3
---	---	---

 2e item

a[1, 2] = 

1	2	3
4	5	6

  
↑  
Rj Kolom

B[0:2, 1] = 

1	2	3
4	5	6

  
↑  
Rj Kolom

a[0:2] = 

1	2	3
---	---	---

  
OTR

a[:, :-1] = Reverse

np.arange(3) = np.array[0, 1, 2]  
Start, Stop, step

np.reshape([3, 2]) = 

1	2
3	4

  
↑  
Rj Kolom

a[:, -1] = 

1	2
3	1

  
↑  
Rj Kolom

np.ndarr = Num array

np.mean

np.transpose = 

1	2	3
4	5	6

 = 

1	3
2	4

np.concatenate = 

1	2
3	4

 [3, 4]  
① ② ③ ④  
↑  
Pos NR

Pd. read\_csv(file, sep= , row)

Knr. ValueCounts = 2.0 → 2x etc

Nan = knr. loc[knr["nan"].min]

Pd. count = 

Rij	Kolon
-----	-------

Knr. dropna (subset = F["nan"] )

Knr[F["nan"]]= knr["nan"].str.lstrip()

.str.replace("W", "I").str.replace("I")

R = 

F	F	F
I	I	I
I	I	I

 .str.contains("p")

knr.loc[knr["p">>9]

Von = knr["p"].str.com("3")

Von2 = Von. where(Von <= 50)

mean{median, mode}[0]

Von - Von. com["p"] str.lstrip()

.pd.Pivot\_table(df, values, index, np.sum, margin)

not - values [A, P] orc = [T, F]

.between(-1, 1)

idmax() = id largest kolom

## Pandas

- df.sort\_values(['col1', 'col2'], ascending=[F,T])
- df.fillna('waarde?')  
inplace=True)
- pd.read\_csv('naam', index\_col='naam', names=[])
- df[> 0] < 'waarde'
- pd.concat([df1, df2]) → append rows
- pd.concat([df1, df2], axis=1) → append columns
- df.rename(columns={'y': 'year'})
- df.dropna() → drop rows where at least 1 element is missing
- df.dropna(how='all') → where all elements are missing
- df.loc[:, 'X2': 'X4'] → select all columns between X2 and X4 incl
- df.drop(columns=['a', 'b'])
- pd.isnull()
- pd.merge(left=df, right=df, how='left'/right/inner/outer), on='X1')
- df['w'].nunique() → # of distinct values in column
- pd.notnull() / notna()
- .str.lower()
- .str.contains('string')
- series.str.startswith('string')
- series.str.replace('iets', 'door-dit')
- series.value\_counts(sort=True, ascending=False, dropna=False)
- df.drop\_duplicates(columns[], inplace=False)
- df.iloc[:, [1, 2, 5]] → select cols in positions 1, 2, 5

$$\text{accuracy} = \frac{TP + TN}{\text{TOTAL}}$$

$$\text{precision} = \frac{TP}{TP + FP} = \frac{(P \cdot acc) / ((P \cdot acc) + (1 - P) \cdot (1 - acc))}{C(1 - P) \cdot acc} = \text{precision}_\text{ZIEK}$$

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

## Numpy

- np.intersect1d(a, b)
- np.setdiff1d(a, b) verwijder uit arrA wat in B zit.
- np.arange(start, stop, step)
- np.isnan(array)
- np.nanmedian(array, axis=)
- np.nanmean(array, axis=)
- arr.flatten() → maakt van MA 2D array
- np.nonzero(arr)
- np.sort(arr, axis=)
- np.sqrt(x)
- np.multiply(x1, x2)
- np.vstack([a, b])
- np.hstack([a, b])
- np.argwhere(X>1) → geeft indices van values die voldoen
- np.random.choice(5, 3) → kiest 3 dingen uit np.arange(5)
- np.where(condition)
- np.unique(arr)
- np.linspace(0, 2, 9) → return\_index=False, return\_counts=True
- arr = arr[~np.isnan(arr)]

		Predicted	
		N	P
Actual	Negative	TN	FP
	Positive	FN	TP

plt.bar(x, height, color, label)  
 plt.xlabel('string', fontweight, fontsize)  
 plt.xticks(np.arange(8), rotation=90)  
 plt.bach(x='ministerie')

## RegEx

- \d → matches any decimal digit
- \D → matches any char that isn't dec digit
- \w → matches any word char
- \W → matches any non-word char
- \s → matches whitespace chars
- \Length\$ → matches strings ending with length
- '^Sepal' → matches strings beginning with 'Sepal'
- ? → 0 or 1 with 'Sepal'
- + → 1 or more reps
- \* → 0 or more reps

Numpy

np.setdiff1d  
 np.equal  
 np.reshape  
 np.transpose  
 np.argwhere  
 np.unique  
 np.equal  
 np.logical  
 np.where  
 np.shape  
 np.ndim  
 np.size  
 np.idxmax  
 np.nonzero  
 np.absolute  
 df.columns  
 df.value\_counts  
 df.index  
 df.values  
 df.crosstab(df.X, df.Y)  
 df.copy  
 df.dropna(subset=['x'])  
 df.astype(str)  
 df.contains  
 df.nunique  
 df.sort\_index  
 df.loc  
 df.count  
 df.describe  
 df.pivot\_table  
 df.concat

PandasRegex

\=match any  
 a-z = range  
 \d = digit  
 \w = not  
 \w = word  
 | = or  
 \* = 0 or more  
 . = any char  
 \D = non digit  
 \r = regeo  
 \n = newline

pd.MultiIndex.from\_tuples([(a; 2000), (b; 3000)])  
 rgen kolum  
 ↓  
 L[:, 0]

[:, newaxis]

$P / P.sum(axis=0) = \text{normalise}$

$((1-z)*a) / (((1-z)*a) + (z*(1-a))) = nP$

$(a*z) / ((a*z) + ((1-z)*(1-a))) = P$   $a = \text{accurz}$   
 $z = \text{zick}$

$$\text{Precision} = \frac{TP}{TP+FP} \quad \text{Recall} = \frac{TP}{TP+FN}$$

round(,1)  
 %IS  
 %CP ..  
 %time  
 %timeit  
 assert  
 aTab

recall =  $\frac{TP}{TP + FN}$   
 precision =  $\frac{TP}{TP + FP}$   
 acc = ~%. reliable

		has dis		no dis		
		post pos	TP	FP	TN	
		neg	FN			
%IS	0.01	0.95	0.01	0.95	0.05	0.95
%CP	0.01	0.95	0.01	0.95	0.05	0.95
%time	0.01	0.95	0.01	0.95	0.05	0.95
%timeit	0.01	0.95	0.01	0.95	0.05	0.95
assert	0.01	0.95	0.01	0.95	0.05	0.95
aTab	0.01	0.95	0.01	0.95	0.05	0.95

$0.01 \times 0.95 / (0.01 \times 0.95 + 0.99 \times 0.05) = 0.16$

- `isna()` from collections imp Counter  
% matPbtl/b inheritance
- `ValueCounts()`
- `drop()` = `dropna()`  $\Rightarrow$  how='all' = drops rows/cols = all nans thresh=3  
subset=[...]  
Walker min met 3)
- `unique()`
- `sort_index()` - `sort_values()` (by=[...], axis=[0, 1])
- `groupby = kvi[kvi >= 1]`
- `for i in range(L.shape[0]):`  
 $L[:, i] \rightarrow np.isnan[L[:, i]] = np.$   
num  
, to\_datetime  $L[i]$  ?
- `inplace=True`
- `idxmax`
- `copy`
- `print(df1)`
- `print("%s - %s" % (per, mis))`
- `merge`
- `melt`
- `query()`
- `unique()`
- `clip()` (bv. orderset characteristic)

- plt.plot (kind='bar', 'barh')
- plt.hist scatter
- plt.xlabel('...')
- plt.legend('...')
- plt.title('...')
- plt.show()
- Str
  - len()
  - contains('...')
  - replace
  - Count
  - Strip() lower strip
  - upper
  - Space
- pd.MultiIndex?
  - on='...'
- pd.merge()
  - how='inner'/'outer'
- loc [::3]
- iloc [::3, ::2]
- ix [::3, 'pop'] aggfunc='sum' mean
- ascending
- nlargest
- nsmallest(n, value)
- filters
- Stack()
- unstack() level=0/1
- of/flatten()

Set((...)) & |^ ((...))

- concat (2 df's samenvoegen), ignore\_index=True
- join='inner'/'outer'
- append()
- index (sort) → no sort\_val. => eerste
- groupby('year', 'sex')[['count']] → by=, level=
- agg/aggregate ('first').unstack()
- index\_col = ...
- reset\_index() method='ffill' → vorige val
- fillna(0 or 1) 'bfill' → vorige val // 2 = 2 stappen
- mean() etc. maar 0 = col 1 = ry
- apply()
- isnull() notnull() cumsum with next prod
- nieuwe col = df[['...']] = ...
- sort\_values([!...], ascending=[..., -1])
- corr()
- any() head().tail() \* pow() / div()
- index dtype all() shape() % mod()
- describe
- items() T transpose() add() fill\_value=0
- mark() level='year'

- np.array([...])
- np.arange(10) → start, stop, step.
- zeros .. ones
- full((3,5), 3.14) empty
- einsum ( ) → start, stop, howvel  
dim 1, 2, 3 = 1, 1.5, 2
- linspace ( ) → start, stop, howvel  
dim 1, 2, 3 = 0, 1, 2
- eye(3) = 1.00 . eye(3, k=1) = 0.01  
0.00
- add. at (x, i, 1)
- np.random.randint (low=, high=, size=) → 2d array
- random((3,3)) → 2d array
- normal
- rand
- ndim
- shape
- size
- dtype
- setdiff1d
- count\_nonzero
- reshape( )
- asarray . astype
- np.newaxis
- row/col
- & | ^ ~ → nicht => np.bitwise\_not
- hstack
- vstack
- concatenate (axis = 0/1)
- (stack = 3rd axis)?
- vsplit(..., [2])
- hsplit
- absolute/abs
- n/cos/tan
- log  
2  
(3, x).log10
- in np.add
- np.divide
- np.power
- np.mod
- etc.
- zeros .. ones
- full((3,5), 3.14) empty
- einsum ( ) → start, stop, howvel  
dim 1, 2, 3 = 1, 1.5, 2
- linspace ( ) → start, stop, howvel  
dim 1, 2, 3 = 0, 1, 2
- eye(3) = 1.00 . eye(3, k=1) = 0.01  
0.00
- add. at (x, i, 1)
- np.random.randint (low=, high=, size=) → 2d array
- random((3,3)) → 2d array
- normal
- rand
- array  
row = [0, 1, 2]  
col = [2, 1, 3]  
row[i, newaxis] \* col  
000  
213  
426
- logical\_and
- unique
- isnan
- reverse 2d:  
x[:: -1, :: -1]
- first row/col  
x2[:, 0] col  
x2[0, :] row
- np.newaxis
- np.newaxis
- Broadcasting: → moet zelfde laag hebben  
np.arange(3) + 5  
0 1 2 → 5 6 7  
np.ones((3,3)) + np.arange(3)  
1 1 1 + 0 1 2 → 1 2 3  
1 1 1 + 0 1 2 → 1 2 3  
np.arange(3).reshape((3, 1)) + np.arange(3)  
0 1 2 → 0 1 2 + 0 1 2  
2 2 2 + 1 2 3 → 1 2 3  
2 3 4
- sum(), min(), max()
- mean . prod
- nanmean . std . var
- argmin, max
- median
- percentile
- copy
- insert
- delete
- from scipy import stats  
modus = mode

Precision & Recall *Reliability: does not account for errors* Good

Evaluating

- Classification: Accuracy, Precision, Recall, F1
- Regression: RMSE

Accuracy: How many of all predictions correct

Recall: How many instances of class C predicted to be C  $\frac{TP}{TP+FN}$

Precision: How many predictions of class C are correct  $\frac{TP}{TP+FP}$

Axes:

Numpy



Axes 0

constant

NP.ARRAY([[[ ]]])

Create:

- NP.ARRANGE
- NP.LINSPACE
- NP.FULL((2,2), 7)
- NP.RANDOM.RANDOM((2,2))

Inspect:

- A.SHAPe
- LBN(A)
- a.NDIM
- a.SIZE

SORT LOG

SIN

COS

Methods:

- A.SUM()
- A.MIN
- A.MAX(axis=1)
- A.MEAN()
- A.CORRCOFF
- NP.STD(A)

Slices:

- A[1,2]
- A[[1,0],[0,1,2]]  $\rightarrow (1,0), (0,1), (1,2)$

Methods:

- A.T
- A.RAVEL
- A.RESHAPE
- A.SORT(axis=0)
- NP.ARGSUM

\*matplotlib inline

% MAGIC  $\rightarrow$  explains major commands.

w = !ls  $\rightarrow$  LBN(w) as level separator. \* + ? or more, for more, or /  
a{z, } z of more a's

PANDAS

DF = PD.DataFrame([[[ ]]], index=[1, 2, 3], columns=['A', 'B', 'C'])

DF.SORT\_VALUES

is now order mat columns

DF.RENAME

DF.RESET\_INDEX

DF.ILoc[10:20]

BU

now order mat columns

DF.LOC[:, 'x2': 'x4']

DF.LOC[df.a > 10, ['a', 'c']]

DF.W.VALUE\_COUNTS

BU

DF.W.NUNIQUE

DF.DROPNA

DF.FILLNA(onlas)

DF.GROUPBY(by='cd').SUM()

is with max val

PD.PIVOT\_TABLE

DF.IDKHDX

MERGE

PD.MERGE(a, b, How='left', on='x1')

PD.plot(title="")

is now order mat columns

DF.plot.HIST()

DF.plot.BARH()

DF = DF[DF.Notnull()]

is now order mat columns

DF.A.STR.REPLACE(pattern, "")

$$\text{Precision} : \frac{tp}{tp + fp}$$

$$\text{Recall} : \frac{tp}{tp + fn}$$

$$\text{Accuracy} : \frac{t}{tf}$$

Numpy array indexing.

$X[0, 0]$   
↑ ↑  
Row column

[start: stop : step]

Slicing

One-dimensional

$X[:5]$  eerste vijf  
 $X[5:]$  alles na vijf  
 $X[::2]$  om de twee  
 $X[::-1]$  omgedraaid.

Multi-dimensional

$X[:, :3]$  stop bij de tweede rij  
stop bij de derde kolom

$X[:, ::2]$  stop bij de derde rij  
om de udam.

$X[0, :]$  rij 0, van alle kolommen

Filtering pandas

a	0,25
b	0,50
c	0,75

= data[(data > 0,3) & (data < 0,8)]  
data.values

`pd.read_csv('file', compression='gzip', sep='\t', delimiter='\\t', header=None, index_col=0)`  
 - `fillna(0)` → start niet op  
`df.ioc[:, -4:-1].sum(axis=1)`  
`(df[[ASS1, ASS2, ASS3]] > 9).sum().sum()`  
`df.sort_values(['deeltoets-mean', 'names'], ascending=[False, True])`  
`c = df[[ASS1, ...]].corr()`  
`c = c[c!=1]`  
`r_ij = c.max().idxmax()`  
`wolom = c.loc[r_ij].idxmax()`  
`V10 = {r_ij, wolom}`

`df.groupby('columns')[I].mean()`  
`df.value_counts()`  
`.index`  
`.iloc[0] implicit`  
`.loc['a'] explicit`  
`.idxmax()`  
`.isinc()`

`pd.pivot_table(df, columns='sod', index='Name', aggfunc={'count': np.sum}, values='count')`

### [numpy]:

`df.normalize(p):`  
 - `return (p / (p.sum(axis=0)))`  
`d = np.arange(S.shape[0])`  
`S[d, d], np.diagonal(S)`

$$\text{ACC} = \frac{\text{TP}}{\text{TP} + \text{TN} + \text{FP} + \text{FN}}$$

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

$$\text{RECALL} = \frac{\text{TP}}{\text{TP} + \text{FN}}$$

`cito.str.lower().str.strip().str.replace(' ', '') .str.replace`

`regex: .str.replace(r'^[a-zA-Z\d]', '')`

`pd.cutsc.tab(kvr.party, kvr.JAAR) [most common].index.array`

- `contains('vragen')`
- `unique`
- `values`

### [for loop]

`maxline = ''`

`for line in fin:`

`if len(line) > len(maxline):`

`maxline = line`

`np.reshape(L, g, f)`  
`L.shape[0]`  
`L.shape[1]`  
`np.transpose()`  
`np.where()`  
`np.flatten() → 1d`  
`np.sum(p < n)`  
↳ van p, waarvan waardestrict kleiner zijn dan n  
`np.sqrt(np.mean((L[:, i] - L[:, 0]) ** 2))`  
`np.argsort(L, axis=1)` → sorteert per reg., van laag → hoog  
`np.all(np.abs(L) > 10)`

`L[i, c]`

`L[np.nonzero(L)]`

`i: np.arange(1, 4)`  
`i: np.arange(1, 4)[:, newaxis]`

`L[<]`

`np.random.randint(0, n)`

`.append`

`np.arange(n)`

`np.reshape(L, i, c)`  
`L.size`

`last column = L[:, -1]`

`n = len(L[0])`

`L[int(n/2)-2:int(n/2)+2, int(n/2)-2:int(n/2)+2]`

**buren:**  
 $y\text{-lim} : x\text{-lim} = \text{grid}.shape$   
 $\text{grid}[(y:y+yy)\%y\text{-lim}, (x+xx)\%x\text{-lim}]$   
 for  $xx \in (-1, 0, 1)$  for  $yy \in (-1, 0, 1)$   
 if  $xx = 0$  and  $yy = 0$

`(::2)::2`

np. arange(0,0,step) - np.size  
 Array.reshape (n,n) np.ndim  
 ~np.array(L'.N, dtype=bool)  
np.transpose(array) [1,2] → [0,3],  
[3,4] → [2,4])  
Aggregate Functions  
 a.sum() sum array  
 a.min()  
 b.max(axis=0) max value per row  
NP.Sort(axis) manipulation

NP.concatenate()  
 NP.vstack() vertical  
 NP.r\_[e,f] stack vertical rowwise  
 np.hstack  
 np.column\_stack  
 np.hsplit(9,3) split at index  
 np.vsplit(9,3)

For y,x in zip(Array, Array)  
 np.linspace(0,2,9) → [0,0.25,0.5,...]  
 np.full((2,2),7) → [[7,7][7,7]]  
 np.eye(2,2) → identity matrix  
 np.random.random((2,2)) → random values  
 np.ones()  
np.zeros() slicing  
 L[0:3,1] row 0 to 3, 1st value 7  
 L[:,:] print alles  
 L[:,1] 1st row  
 L[L<2] alle values < 2  
 L[[0,1],[1,0]] print L[0,1] en L[1,0]

broadcasting

$$\begin{array}{r}
 \text{Arange}(3) + 5 \\
 \boxed{0} \boxed{1} \boxed{2} + 5 \boxed{5} \boxed{6} \boxed{7} = \boxed{5} \boxed{6} \boxed{7} \\
 \hline
 \text{np.ones}(3) + \text{Arange}(3) \\
 \begin{array}{ccc|c}
 1 & 1 & 1 & 3 \\
 1 & 1 & 1 & 3 \\
 1 & 1 & 1 & 3 \\
 \hline
 0 & 1 & 2 & 6 \\
 1 & 1 & 2 & 6 \\
 2 & 1 & 2 & 6
 \end{array} \rightarrow \begin{array}{ccc|c}
 1 & 1 & 1 & 6 \\
 1 & 1 & 1 & 6 \\
 1 & 1 & 1 & 6 \\
 \hline
 0 & 1 & 2 & 6 \\
 1 & 1 & 2 & 6 \\
 2 & 1 & 2 & 6
 \end{array}
 \end{array}$$

## Syntax

Index = pd.MultiIndex.from\_tuples([(d1,1..3)])

reshape pd.melt(df) → columns to rows  
 df.pivot\_table() spread rows to columns  
 df.concat([df1,df2]) append rows, axis 1 for column

## Observations:

df[df.length>7] df.drop\_duplicates()  
 df.head(n) df.nlargest df.nsmallest()  
 df.filter(regex) df.loc[df['a']>5, ['a','b']]

summarize  
 df['w'].value\_counts()  
 df['w'].nunique()

min()  
 max()  
 quantile()  
 var()  
 apply(np.sqrt)

df.plot.hist(), plot.bar() / plot.barch  
 pd.crosstab()  
 df.Pivot\_table()

df.div(sum,..)

$$ACC = \frac{TP+TN}{Total}$$

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

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

.sort\_values(by=[]) altijd str. lower  
 kwarg, str.replace  
 str.contains  
 str.len  
 str.count

## Pandas

df[[Kolommen]] projectie v.d. Kolommen

df[[Kolommen]]>x) booleanmask op de projectie

df.column.mean() gemiddelde over Kolommen

df.column.mean(1) gemiddelde over r̄gen

df.column>x).mean()\*100 berekenen percentage m.b.v booleanmask

Variantie = ((df-df.mean())<sup>2</sup>).mean()

Std = np.sqrt(variantie)

corr\_matrix = df[[Kolommen]].corr() correlatie matrix van df[[Kolommen]]

corr\_matrix = corr\_matrix[corr\_matrix != 1] corr\_matrix zonder items die met zichzelf correleren

r̄j = corr\_matrix.max().idxmax() key die hoort bij de hoogste corr waarde in de r̄gen

Kolom = corr\_matrix.loc[r̄j].idxmax() " " " " " " " Kolommen

What are the max sepal\_lengths for each species? df.groupby('species')[['sepal-length']].max()

## Numpy

axis=0 → Kolom axis=1 → r̄j

Fancy indexing: array[[r̄j, r̄j, r̄j], [Kolom, Kolom, Kolom]] 1<sup>e</sup> l̄st plaats r̄j, 2<sup>e</sup> l̄st plaats Kolom

Normaliseren: (array / (array.sum(axis=0))) elke waarde delen door gem. van de Kolom.

np.arange(S.shape[0]) maakt een l̄st van waardes gebasseerd op aantal r̄gen in een array

array.shape → (aantal r̄gen, aantal Kolommen)

2D-array → L[0] = eerster̄j

L[:, 0] = eerste Kolom

L[:, x][np.isnan(L[:, x])] → checkt voor alle waardes in Kolom x of deze nan zijn

array.reshape([array.size, ])

np.arange(1, 4) | 1 r̄j, 4 Kolommen

np.arange(1, 4) | 1, np.newaxis | 1 Kolom, 4 r̄gen

(top 10 ziekt)

Precision = TP / (TP + FP)

Recall = TP / (TP + FN)

	Z	NZ	
P	1.09	9.01	1.8
N	1.91	9.99	8.2
	1	9	10

accuracy = 90%

$$\begin{aligned} \text{precision} &= \frac{\text{TP}}{\text{TP} + \text{FP}} \\ \text{recall} &= \frac{\text{TP}}{\text{TP} + \text{FN}} \\ \text{F1} &= \frac{2 \cdot \text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}} \\ \text{accuracy} &= \frac{\text{TP} + \text{TN}}{\text{TP} + \text{FN} + \text{TN} + \text{FP}} \end{aligned}$$

Magic  
%ls -lh: Files in dir + size  
%rm file x? istime for delete  
%time  
%timeit  
%prun

a.shape  
a.ndim  
a.size  
a.dtype.name  
a.astype = cast to type  
np.zeros(len, dtype=int)  
np.multiply.outer(a, b)  
np.add.reduce(a, b) reduce dim  
np.add.accumulate  
np.power  
np.abs  
np.cumsum(axis=1)  
• transpose

np.random.seed(0)  
np.random.rand(10, size=amt)  
np.reshape(array, (4, 3))  $\xrightarrow{x}$   
eye(amt, dtype=int) =  $\begin{bmatrix} 1, 0 \\ 0, 1 \end{bmatrix}$ ,  
resize (= affects orig)  $\begin{bmatrix} 0, 1 \end{bmatrix}$

append  
insert  
delete  
concatenate  
vstack = stack vertically

read\_csv(file, skipinitialspace, sep)  
• values  
• index  
• loc / iloc  
• series([1, 2, 3], index=[a, b, c])  
• head()  
• drop(index or column)  
• size  
• copy (copying)  
• merge

a = np.array([5, 10, 15])  
sepal > 5  
• df.query("sepal > 5 and width < 0.9")  
sort\_values([col1], ascending=[False, True])  
argmax vs max  
sns.heatmap

Regex

\$ start string  
\$ end  
\w whitespace | \S  
\d | \D  
\* 0 or more  
+ 1 or more  
? 0 or 1  
\{3\} 3  
\{3, 3 or more

[abc]  
[a-q]

pd.read\_csv(link\_to\_file(), compression='gzip', header=None, names=['Jaar', 'etc.'], sep='\t', quotechar='', error\_bad\_lines=False, skipinitialspace=True)

top\_0 = kur['Partij'].value\_counts()[1].index\_values

Partij\_0 = kur[(kur['Partij'] == 'PVDA') | etc]

cross = pd.crosstab(Partij\_0, Jaar, Partij\_0, Partij)

cross.plot()

parties\_left = len(set(kur['Partij']))  
of str.lower of (' ', '')

kur['Partij'].str.replace(' ', '')

Kur = kur[pd.notnull(kur['Partij'])]

Kur = kur[kur['Partij'].str.contains('vragen')]

Kur['Partij'].value\_counts().sort\_index()

Kur['Partij'].value\_counts().sort\_index(inplace=True)

index\_col of sort\_index (inplace=True)  
pd.pivot\_table(df, values=['Count'], index=['Name'], columns=['Sex'], aggfunc=np.sum, margins=True)

True, dropna=False).sort\_values(by='All', ascending=False)

piut['ratio'] = np.log2(pivot['Count']['M'] / pivot['Count']['F'])

set(pivot[(pivot['ratio'] > -0.1) & (pivot['ratio'] < 0.1)])

df = df.fillna(0)

df['ass\_mean'] = df[['ass\_week1', 'etc']].mean(1)

greater = (df[['ass1', 'etc']] > 9).sum().sum() sum groter dan 9

df['weighted\_mean'] = 0.4 \* df['ass\_mean'] + 0.6 \* df['deel\_mean'] \* 100

perc = ((df['ass\_mean'] < 5.5) | (df['deel\_mean'] < 5.5)).mean()

Z\_DF = (df - df.mean()) / df.std()

df.sort\_values(['deeltoets\_mean'], ascending=[False, True])

c = ((c != 1)

rij = c.max().idxmax()

Kolom = c.loc[rij].idxmax()

vraagID = {rij, Kolom}



## np.setdiff1d

axis=0, gaat over de columns ↓  
axis=1, gaat over de rows →

top = kur. Partij.value\_counts().head(10)

cross = pd.crosstab(kur.Jaar, kur.Partij)[top.keys()]

remove = [^a-zA-Z0-9], \s, allen, beiden, groep, fr, lid]

kur.Partij = kur.Partij.str.lower().replace(regex=remove, value="")

kur = kur[~kur.Partij.str.contains("vragen", na=False)]

kur.dropna(subset=[Partij], inplace=True)

parties\_left = kur.Partij.unique().size

kur["deelvragen"] = kur.Vraag.str.count("\\"?)

## Pearson correlation

numpy = np.corrcoef(x, y)[0,1]

pandas = x.corr(y, method="pearson")

import scipy.stats

skip = scipy.stats.pearson(x, y)[0]

df[["assignment\_mean"]] = df[df.columns[0:6].mean(axis=1)]

all\_assignments = df[df.columns[0:6]]

greater\_than\_g = all\_assignments[all\_assignments > g.count().sum()]

df = df.sort\_values(["deeltjes\_mean", "nauw"], ascending=[False, True])

# vraag\_10 = set()

c = df[df.columns[0:6]].corr()

c = c[c[:] = 1]

row = c.max().idxmax()

col = c[r\_max].idxmax()

vraag\_10 = (row, col)

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

vraagho - genetisch

hp.where

np.argwhere

$$\text{normalize} = P / P_{\text{sum}}(\text{axis}=0)$$

$$\text{accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

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

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

## Pandas

open files: → pd.read\_csv ('file.csv', sep="\t", encoding='utf-8')  
→ with open ('file') as f:  
from web → f = requests.get(url)

$$\begin{aligned} \text{precision accuracy} &= \frac{\text{True positive}}{\text{True Positive} + \text{False Positive}} \\ \text{recall} &= \frac{\text{True positive}}{\text{True positive} + \text{False Negative}} \end{aligned}$$

Reshaping: pd.concat, pd.melt

select col: df.loc

merge: pd.merge (df1, df2, how='inner', on='x1')  
df.groupby()

Numpy:

inspecting: a.shape, len(a), a.ndim, a.size, b.astype(int)

Sort: a.sort (axis=0)

Slicing: a [row] [col]

indexing: a [a<2] → boolean

Fancy Indexing: [..., ...] [..., ...]

- null NaN → df.fillna(0)

- pd neue kolom: df ["new-column"] = ...

- boolean indexing: df ["column"] > 4

- Z-df = (df - df.mean()) / df.std() ← Z-normalisieren

Precision

P = .1

$$\text{Precision} = \frac{\text{Font} \times \text{TP}}{\text{TP} + \text{FP}}$$
$$= \frac{0.9}{0.9 + 0.1}$$

$\text{Prec.} = \frac{\text{TP}}{\text{TP} + \text{FP}}$     $\text{Rec.} = \frac{\text{TP}}{\text{TP} + \text{FN}}$     $\text{Acc.} = \frac{\text{TP} + \text{TN}}{\text{Total}}$ 
  
`np.int0(np.array.dtype) → b.astype(int) ? → np.int64, float32`
  
`np.: add, subtract, divide, multiply, sqrt, log, log2`
  
`np.: sum, min, max, cumsum, corrcoef, corr, std`
  
`np.: zeros((3,4)), ones((2,3,4)), arange(10, 25, 5), linspace(0, 2, 9)`
  
`np.: any, all .ravel() == .flatten(), reshape vs. resize`
  
`np.: sort(), sort(axis=0), concatenate([a,b], axis=0)`
  
`diag.: return L[range(L.shape[0]), range(L.shape[0])]`
  
 $L[:, i][\text{np.isnan}(L[:, i])] = \text{np.nanmedian}(L[:, i]), \text{np.nonzero}()$ 
  
`np.arange(1, 4)[:, np.newaxis], np.split()`
  
`%ls -lh, %rm <file>, %time, %time it, <TAB>, ?`

`df[x == 2] vs. df.loc[x == 2] vs. df.iloc[1:3]`  
`pd.isnull() vs pd.notnull(), df.sort_values, df.sort_index`  
`df.dropna(subset=["x"]), df.nlargest, nsmallest, df[x].value_counts`  
`df.plot(kind="bar/barch"), df.plot.hist, df.plot.scatter, df.fillna()`  
`pd.merge(a, b, how="inner/outer/left/right", on="x")`  
`df.drop_duplicates(), pd.Series(data=values, index=labels)`  
`(df.assignments > 9).sum().sum(), data[-1], data[::2]`  
`df.groupby().apply() → .unstack() → .head()`  
`df[~df.x1.isin(df2.x1)], .index, .values`  
~~df[~df.x1.str.replace("\d", ".join(["\s", "\W"], "\u00a0))]~~  
`df[~df.x1.str.contains("string")], pivot[[F, M]].min(axis=1)`  
`c = df.assignments.corr(), x = c[c != 1].max().idxmax()`  
`y = c[c != 1].loc[x].idxmax(), np.transpose == np.T, np.abs()`  
`pd.merge(), pd.concat(), df.reset_index(), pd.melt`  
`np.copy(a), a.copy(), a.view(), random- <TAB>`  
`divisible: L[L % n == 0], round(float, 1)`  
`from numpy import doc → np.doc? → np.lookfor`

<sup>^ starts, ends \$, (\.) alle ., ((\(? Species\)).\*)<sup>\*</sup></sup>  
<sup>alles behält species</sup>  
`\d digit, \W not \w, \S whitespace, {3,4} 3, 4, 3+ x`

## Numpy

Maken

`np.random.randint(0, 100, (shape))`  
`np.arange(start, stop, steps)`

Informatie

`a.shape` - dimensions

`a.shape[0]` - rij

`a.shape[1]` - kolom

`a.size` - elementen

`a.dtype` - elementen

`a.astype(int)` - convert

Manipuleren

`np.transpose(a)`

`np.insert(a, 1, 5)`

`np.append()`

`np.delete()`

`np.reshape()`

Math

`a.sum()`

`a.min()`

`a.max()`

`a.cumsum()`

`a.mean()`

`a.median()`

`np.std(a)`

over hele array

Index Slicing

`a[0, 1]`

`a[1:3, 0:2]`

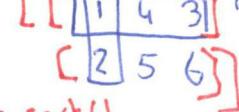
`a[:, 3]`

`a[:-2, 0]`

`a[3, :]`

`a[2, 2:]`

`np.sum(axis=0)`

 `np.sum(axis=1)`

`a.argsort()`

`a.argsort(axis=0)` kolommen

`(axis=1) rijen`

went over

op deze

0	[1 8 2 3 2]
1	[5 6 5 3 6]
2	[1 3 7 4 5]
3	[7 4 4 9 9]

## Jupyter handig

function? - documentatie

help(function) - documentatie

function?? - source code

Regular expressions

1 of \* `clf.str.replace(re, nf)`

\w words

\W not-words

[0-9] numbers

[a-z] lowercase

[A-Z] uppercase

\hoi - begint met hoi:

\s space

\? alles met ?

@[...] niet.

\Is magic

`np.var(a)`

`np.std(a)`

Rmse =  $\sqrt{\text{np.mean}((L[:, 0] - L[:, 1])^2 / 2)}$

waardes kolommen

## Pandas

Interen

`pd.read_csv('naambest', sep=',', index_col=1, skipinitialspace=True)`

`df.sort_index()`

`df.sort_values('kolomnaam', ascending=False)`

`df.rename(columns={'old': 'new'})`

Displays b boolean!

`df[df.length > 7]` - rows that meet criteria

`df.drop_duplicates()`

`df.drop(columns=['x', 'y'])`

`df.sample(frac=0.5)`

n=10

`df.nlargest(n, 'values')`

Subsets

`df[['naam', 'maai']]`

`df.filter(regex)`

`df.loc[:, [1, 2, 5]]`

Summarizing

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

`df.sum()`

`df.count()`

`df.describe()`

`df.agg(['count', 'min', 'max'])`

foreachin

)

what is total

`df.groupby('kolom').kolom.sum()`

voor totalen bij elkaar

`df.pivot_table('values' = , index=, columns=, aggfunc='mean', fill_value, margins=False)`

`df.dropna(subset=['kolom1', 'kolom2'])`

Naam moet in beide sets voorkomen

`pd.merge(df1, df2, how='left' on='kolom1')`

intersection

optional

`pd.crosstab(df['kolom1'], df['kolom2'])`

callij count

## R Base

`df['kolom1'], ['kolom2], cross()`

`pd.set_option('display.max_columns', 10)`

`df.isnull(), isin(), contains() -> all boolean`

lambda x, y: x\*y -> shortcut for function



Arrays are created with `np.array([ ])`  
`np.arange(10, 25, 5)` creates an array of evenly spaced values (5 is the step value)

Inspecting your array is done by

a.shape	array dimensions
len(a)	length of array
b.ndim	number of array dimensions
e.size	number of array elements
b.dtype	datatype of array elements.

`a[2]` subsetting, select element on 2nd index  
`b[1,2]` equivalent to `b[1][2]` or [row][column]

$$\text{precision} = \frac{(1-p) \times 0.9}{(p \times 0.1) + ((1-p) \times 0.9)}$$

`a[0:2]` slicing for a in a row

`a[0:2, 1]` slicing for a/ in a column

`a[:, -1]` get the last column

$$\text{precision} = \frac{p \times \text{precision\_left}}{(p \times 0.9) + ((1-p) \times 0.1)}$$

`a[a < 2]` boolean indexing

`a[[ "fill in columns, row, etc"]]` select a subset of the matrix's rows and columns. = fancy indexing

`i = np.transpose(b)` } permute array  
`i.T` } dimensions

`np.intersect1d()` or `np.union1d()` or `np.setdiff1d()`

Flatten the array `np.flatten()` or `b.ravel()`

Reshape the array `g.reshape(3, 2)`

concatenate arrays `np.concatenate((a,d), axis=0)`

axis has to be specified.

`np.arange(start, stop, step)` optional `[:, np.newaxis]`

make a dataframe, a multiIndex dataframe is created  
`pd.DataFrame([E[4, 7, 10],`  
`[5, 8, 11],`  
`[6, g, 12]],`  
`index = [2, 2, 3],`  
`columns = ['a', 'b', 'c']`

mse =

$$\frac{\sum (\text{Predicted} - \text{Actual})^2}{N}$$

Reshaping data is done with the following

`pd.melt(df)` (columns → to rows)

`pd.concat([df1, df2])` (`df1+df2`)

`pd.concat([df1, df2], axis=1)` (`df1 + df2`)

`df.pivot(columns = 'var', values = 'val')`  
`(rows → columns)`

`df['w'].value_counts()` Count number of rows with each unique value as variable  
`len(df)` # of rows in df  
`df['w']` # of unique values in a column.  
`df.describe()` describes the df.

`df.groupby(by = "col")` return a groupby object grouped by values in a column.

additional groupby function =  
`size()` \* `agg(function)`

`dF.groupby(by = "")["col"]`

new column? `df['volume'] = 'value'`

`g = a - b`

subtraction

`np.subtract(a, b)`

addition

`np.add(a, b)`

division

`g = a / b`

`np.divide(a, b)`

multiplication

`g = a * b`

`np.multiply(a, b)`

$\text{noog begaafd} = \frac{0,01}{0,01 \times \text{precision\_beginnig}} \times \text{precision\_eind}$

`np.exp(b)`

exponentiation

`np.sqrt(b)`

square root

`np.log(b)`

logarithm

`e.dot(f)`

dot product

`a == b` returns True or False

`a < 2` returns True or False

`a != 2` returns True or False

`np.array_equal(a, b)`

`np.all()` or `np.any()`

`a.sum()` sum of values

`a.min()` b. `max()` returns max of min

a. `cumsum()` cumulative sum

a. `mean()` mean

b. `median()` median

a. `corr` correlations

`np.std()` standard deviation

`np.where()` replace values in array or to see indices

a. `sort()` sort an array

b. `sort(axis=0)` sort an array on axis

`df.head()`

`df.tail()`

`df.idxmax()`

subset variables (columns)

`df[['width', 'length', 'species']]`

`df.width` or `df['width']`

`df.filter(regex = 'regrex')`

select multiple columns w specific names.

→

`df.sort_values('mpg')` order rows by values in a column (low optional arguments apply)

`df.rename()` rename columns.

`df.sort_index()` sort the index of a dataframe

subset observations (rows)

`df[df.Length > 7]` extract rows that meet logic

`df.iloc[10 : 20]` select rows by position

`df.nlargest(n, 'value')` select and order top n entries

`df.nsmallest(n, 'value')` select and order bottom n entries

`df.loc[:, 'x2': 'x4']` select all columns between x2 an x4 (in

`df.iloc[:, [1, 2, 5]]` select columns in positions 1, 2 & 5

`df.loc[df['a'] > 10, ['a', 'c']]` select rows meeting

a logical condition, and only specific to the column.

`.sum()` . `mean()` . `min()` . `max()` . `count()` . `median()`

`.var()` . `std()` . `apply(function)` . `df.dropna()`

`df.fillna(value)` pd. `merge()` df. `intersect1d()` df. `union1d()`

`df.difference()`



$$acc = \frac{TP + TN}{Total} \quad prec = \frac{TP}{TP + FP} \quad rec = \frac{TP}{TP + FN}$$

// = floor div

np.array  $\Rightarrow$  np.functie: np.sum, np.max, etc..

np.zeros([length/dims], dtype)	ar.reshape([dims])	np.concatenate([arr1, arr2])..	np.split(grid, [index])
[S:] all after S	np.count_nonzero(statement) counts books	np.ravel(arr)	↳ hsplit, vsplit
[4:7] subarray	np.sqrt(arr), np.argsort(arr)	np.arange(n)	
[:, 2] step 2	arr.transpose()	ar.flatten() / np.ravel(arr)	
X [start:stop:step]			

arith: np.add/subtract/negative/multiply/divide/floor\_divide/power/mod/abs/sqrt

agg: np.sum/prod/mean/std/var/min/max/argmin/argmax/median/percentile/any/all (let op axis)

↳ nan error on nan's te handelen.

np.cumsum(axis),

data[1:3] = implicit, data.loc[1:3] = explicit, data.iloc[1:3] explicit index

df.add/subtract/multiply/divide/floorDiv/mod/pow (hoeft? voor args)

dropna (axis, how, thresh) fillna(method (ffill, bfill))

pd.concat() (check? voor args) ~~join~~

df.append(df2)

pd.merge() (check? voor args)

~~pd.read\_csv~~ values\_count, is\_in, cross-tab

str.lower/upper/contains/strip/replace

mean/median/mode/std

.T, .unstack, .stack

idxmax, idxmin, nunique, nlargest

groupby.apply.unstack.

\d digit

\w word char

\s whitespace char

\D not \d

\W not \w

\S not \s

+ one or more

{3} 3+times

{2,4} 2-4 times

{3,5} 3 or more

\* zero or more

? once or none

\* any except line break

[] select group

^ start string

[^...] not

\$ end string

$$\text{Accuracy} = \frac{\text{TP} + \text{TN}}{(\text{TP} + \text{TN} + \text{FP} + \text{FN})}$$

$$\text{recall} = \frac{\text{TP}}{\text{TP} + \text{FN}}$$

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

`dropna(axis=1)`

`fillna()`

`isnull()` → boolean

`isnotnull` → boolean

Numpy

`np.arange(start, stop, step)`

`np.reshape(..., ...)`

`[L % n == 0] → deelbaar door`

`add_per_column = np.arange(Lshape[1]) + 1`

`add_per_row = np.arange(0, add_per_column(L.T)).T`

`random.choice(i for i in range(5))`

bent niet

→ df[~df['Party'].str.contains('Wagen')]

→ len(df['Party'].unique())

replace tokens + lowercase

→ df['Party'].str.lower().str.replace("W", "")

with open bestand als f:

for j in f:

for j in i:

print(j)

remove specific list of words

→ df['Party'].str.replace("|".join(listofwords), "")

→ drop rows with non

→ df.dropna(subset=['Party'])

→ count = len } agrfunction

→ sum = sum }

→ df[column].str.count('X') = tel aantal dertallen

j. crosstab(df['column'], df['column']).filter(...)

j.read\_csv(skipinitialspace=True, sep='|', names='list of columns')

j.merge(df1, df2, how='left/right/inner/outer', on='X1')

df1[~df1.X1.isin(df2.X1)]