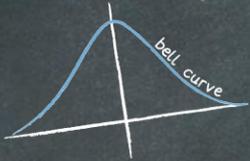


FORECASTING

PREDICTING
FUTURE
CUSTOMER
BEHAVIOUR



$$\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\}$$



IMPROVE
STRATEGY

USEFUL

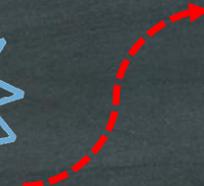


ANALYTICS

Spot



OPPORTUNITIES



Mit Text Mining zum Wettbewerbsvorteil

Marco Nätlitz | Data Scientist

areto

CONSULTING. IT WORKS.

Agenda

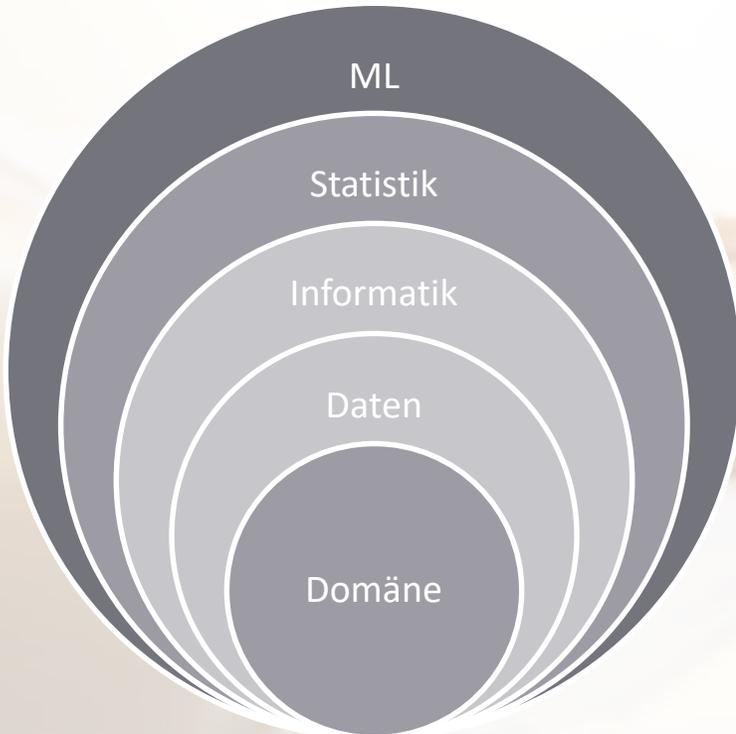
Mit Text Mining zum Wettbewerbsvorteil



- Was ist eigentlich Data Science?
- Was ist unter Text Mining zu verstehen?
- Wie kann man aus Text Wissen extrahieren?
- Und zwar mit der Statistikumgebung R ...
- ... am Beispiel von Twitter

Was ist eigentlich Data Science?

Data Science beschreibt die Extraktion von Wissen aus Daten, Begriff geprägt durch Naur (1974)



Warum Data Science?



- knappe Ressourcen
- schnelle Märkte
- Wettbewerbsvorteile

Enabler: Text Mining
Extrahiere Wissen aus Texten

80%



Enabler: Data Science

Warum werden meine Produkte gekauft?

Welcher meiner PKW-Schäden sind Betrugsfälle?

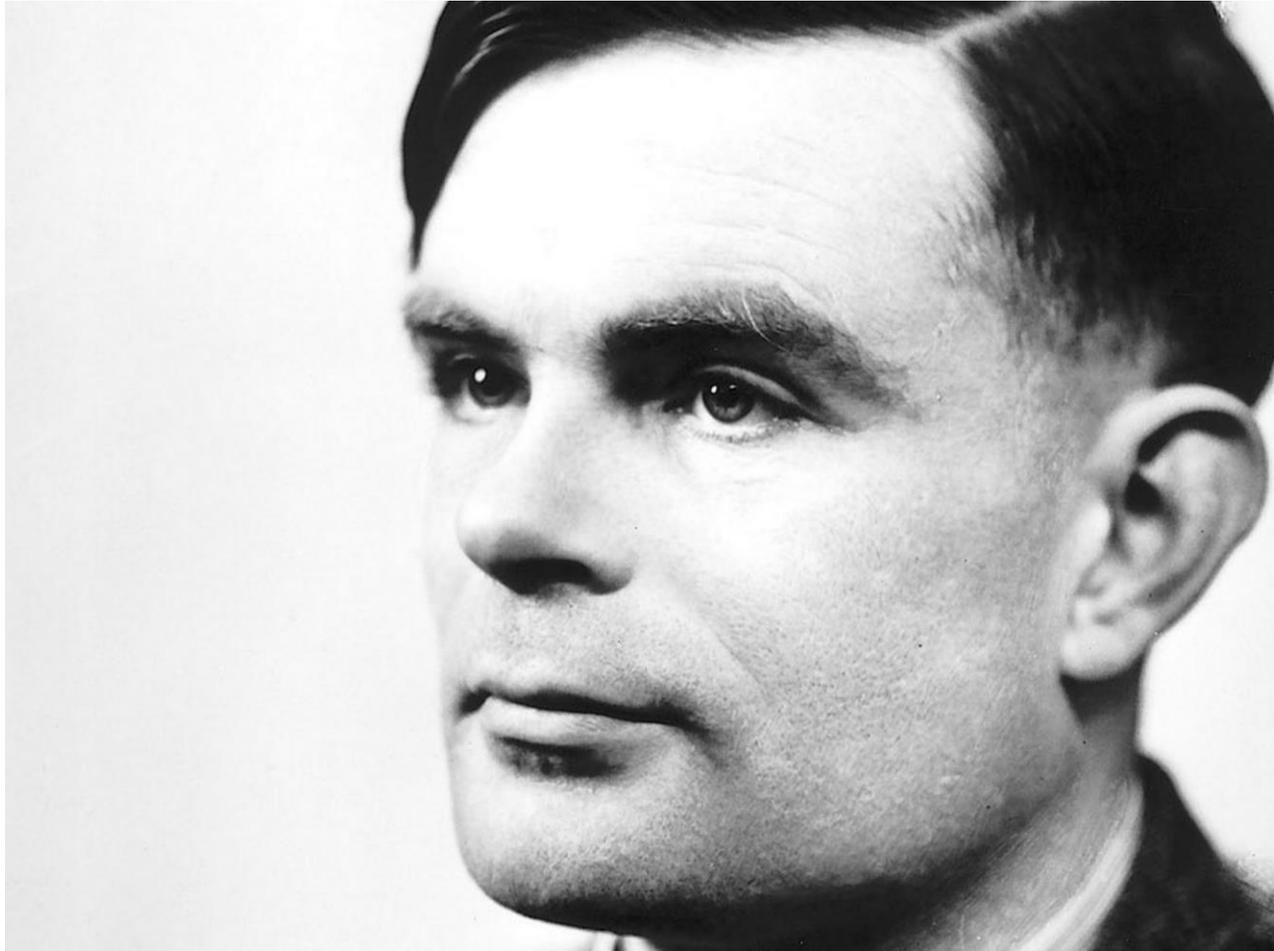
Welche Proteine interagieren miteinander?



Natural Language Processing (NLP)

Was ist unter Text Mining zu verstehen?

- Natural Language Processing (NLP)
- Extrahiere Wissen aus Daten
- Alan Turing (Mathematiker, 1912 – 1954)
- „Computer is able to impersonate a human in a real-time written conversation with a human judge“





Herausforderungen

- Homonyme, Synonyme
- Metaphern, Ironie
- Polystrukturiert
- „Petra geht zur Bank.“
- „Was du erzählst, ist doch Schnee von gestern.“

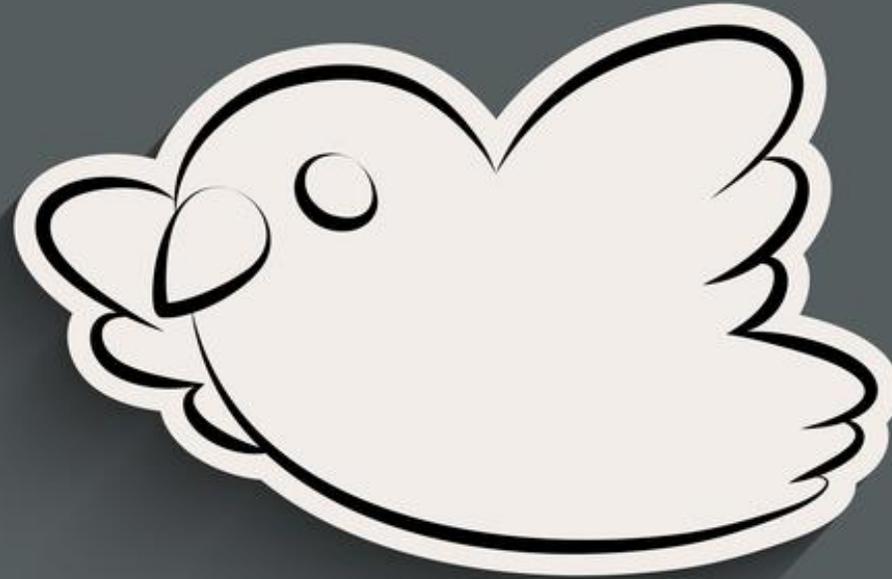
Mit Twitter zu Wettbewerbsvorteilen



Was denken
meine Kunden?

Einnahmen / Jahr \$ 665 Mio.

@katyperry hat 84.950.745 Follower



Hash Tag

320 Mio. Nutzer

2006 gegründet

140 Zeichen

twitter

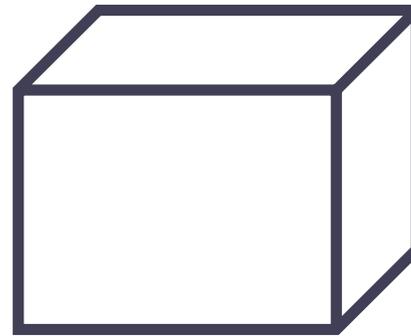
Der Weg zum Sentiment Mining-Modell

Textanalyse mit Twitter-Daten

**Basis:
Tweets**



**Klassifikations-
modell**



**Ist der Tweet
positiv oder
negativ?**



TWEET

75 @Apple Is Testing #Technology w/ 'M7' Chip To Help You Figure Out Where You Pa...
76 I have to say, Apple has by far the best customer care service I have ever rec...
77 iOS 7 is so fricking smooth & beautiful!! #ThanxApple @Apple
78 LOVE U @APPLE
79 Thank you @apple, loving my new iPhone 5S!!!! #apple #iphone5S pic.twitter.c...
80 .@apple has the best customer service. In and out with a new phone in under 10...
81 @apple ear pods are AMAZING! Best sound from in-ear headphones I've ever had!
82 Omg the iPhone 5S is so cool it can read your finger print to unlock your iPho...
83 the iPhone 5c is so beautiful <3 @Apple
84 #AttributeOwnership is exactly why @apple will always be #one! #apple #marketi...
85 Just checked out the specs on the new iOS 7...wow is all I have to say! I can'...
86 I love the new iOS so much!!!! Thnx @apple @phillydvibing
87 Can't wait to get my #Iphone5S!!! @apple
88 @V2vista Fingerprint scanner: The killer feature of iPhone 5S. This is so bloo...
89 Interesting how so many people seem to be almost willing the demise of @Appl...
90 I LOVE @BNBuzz @NOOKstudy @nookBN and @apple made my life so much easier this ...
91 Just watched the keynote of @apple latest iPhones. I just love the #iPhone5S a...
92 My iPhone wasn't calling correctly so I went to an @apple store (first time) t...
93 Great job @apple on providing best users experience #thinkauto
94 Swapped my #galaxys2 for an #iPhone4S. After one day I'd say I'm an @apple con...
95 Can't wait for my #orange phone upgrade in November :-) #apple iPhone 5s here ...
96 #colored #iphone! The new 5C iphone comes in colors. I love this!! Can't wait ...
97 Whether you're an @apple fan or not, this iPhone 5C video is worth watching, f...
98 @apple Impressive features on the Iphone 5s - fingerprint recognition, now tha...
99 @KuqoGroup @jimmykimmel @BlackBerry @Apple that is unbelievably awesome!!!! #...
100 EarPods are amazing, thanks @Apple
101 In luv with the iPhone 5s luv the champagne colour and the fingerprint reader ...
102 Used @Apple TV and @explaineverythng to demonstrate student understanding of fa...
103 @brody_knibbs haha mint arent they! top company @apple
104 I'm back yaw, Thanks to @Apple !!!!!

1181 *Tweets*
Extraktion einer Datenbasis

Der Weg zum Sentiment Mining-Modell

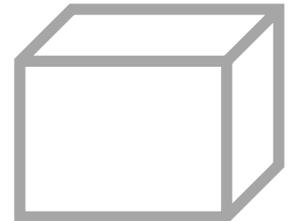
Textanalyse mit Twitter-Daten

Basis:
Tweets

Zielvariable:
Sentiment

Unabhängige
Variablen

Klassifikations-
modell



Amazon Mechanical Turk

Manche Aufgaben lassen sich nur mit menschlicher Intelligenz lösen

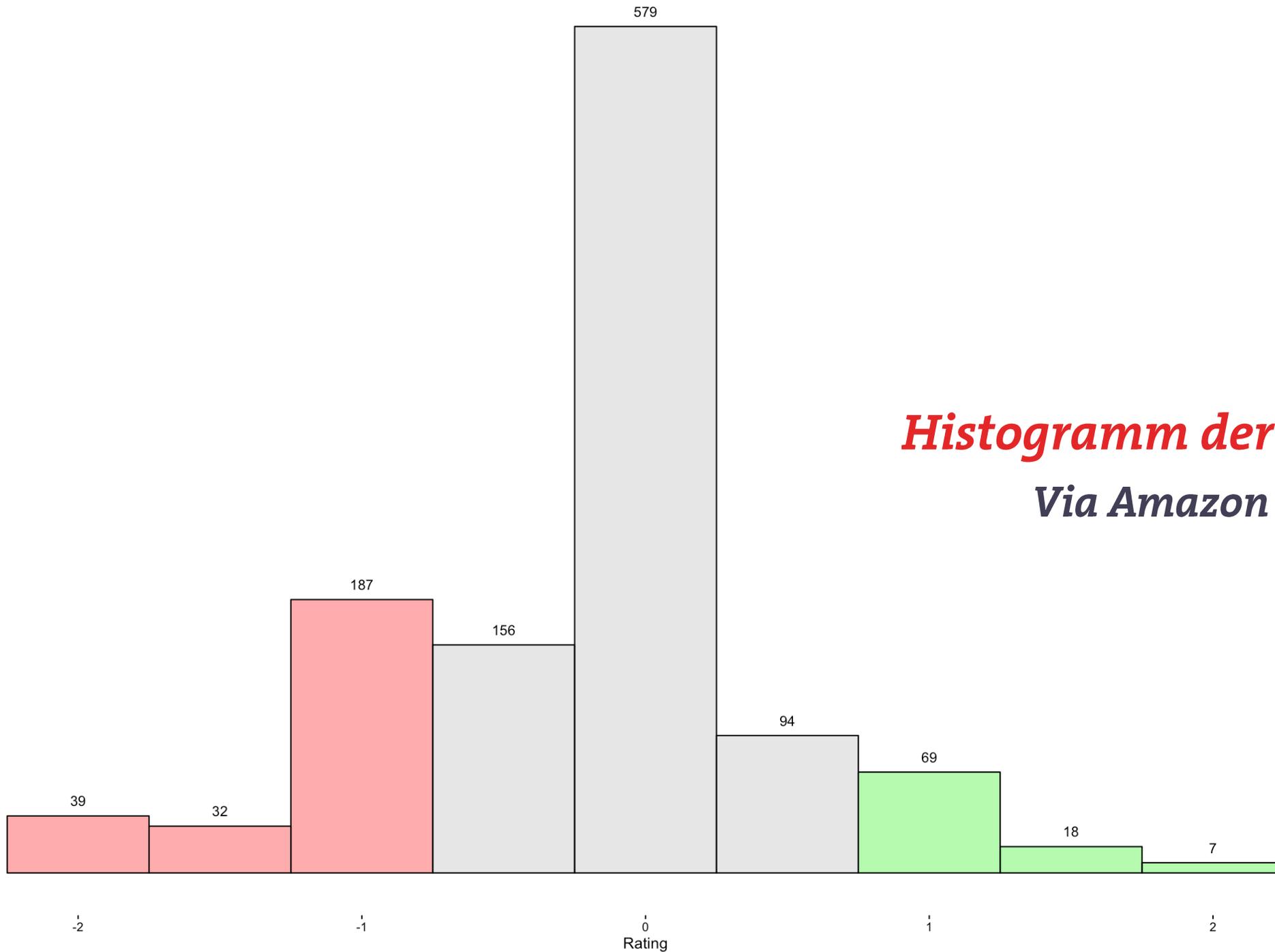
- Crowdsourcing Internet Marketplace
- Tätigkeiten auf Zeit
- Ohne feste Anstellung
- Tätigkeit nicht durch Maschinen machbar
- Sog. Human Intelligence Tasks (HIT)
- 500.000 Auftragssuchende
- In 19 Ländern





Beurteilen Sie die Einstellung
des Verfassers des folgenden
Tweets gegenüber Apple Inc.

Histogramm der Beurteilungen
Via Amazon Mechanical Turk



TWEET	AVG
1 I have to say, Apple has by far the best customer care service I have ever receiv...	2
2 iOS 7 is so fricking smooth & beautiful!! #ThanxApple @Apple	2
3 LOVE U @APPLE	1,8
4 Thank you @apple, loving my new iPhone 5S!!!! #apple #iphone5S pic.twitter.com/...	1,8
5 .@apple has the best customer service. In and out with a new phone in under 10min!	1,8
6 @apple ear pods are AMAZING! Best sound from in-ear headphones I've ever had!	1,8
7 Omg the iPhone 5S is so cool it can read your finger print to unlock your iPhone ...	1,8
8 Can't wait to get my #Iphone5S!!! @apple	1,6
9 I love the new iOS so much!!!! Thnx @apple @phillydvibing	1,6
10 Just checked out the specs on the new iOS 7...wow is all I have to say! I can't w...	1,6
11 #AttributeOwnership is exactly why @apple will always be #one! #apple #marketing ...	1,6
12 the iPhone 5c is so beautiful <3 @Apple	1,6
13 @V2vista Fingerprint scanner: The killer feature of iPhone 5S. This is so bloody ...	1,6
14 Just watched the keynote of @apple latest iPhones. I just love the #iPhone5S and ...	1,6
15 I LOVE @BNBuzz @NOOKstudy @nookBN and @apple made my life so much easier this mor...	1,6
16 Interesting how so many people seem to be almost willing the demise of @Apple. ...	1,6
17 My iPhone wasn't calling correctly so I went to an @apple store (first time) told...	1,4
18 Great job @apple on providing best users experience #thinkauto	1,4
19 Swapped my #galaxys2 for an #iPhone4S. After one day I'd say I'm an @apple convert!	1,4
20 Can't wait for my #orange phone upgrade in November :-) #apple iPhone 5s here I c...	1,4
21 EarPods are amazing, thanks @Apple	1,4
22 @KuqoGroup @jimmykimmel @BlackBerry @Apple that is unbelievably awesome!!!! #cli...	1,4
23 @apple Impressive features on the Iphone 5s - fingerprint recognition, now thats ...	1,4
24 Whether you're an @apple fan or not, this iPhone 5C video is worth watching, for ...	1,4
25 #colored #iphone! The new 5C iphone comes in colors. I love this!! Can't wait til...	1,4

Ø-Rating pro Tweet

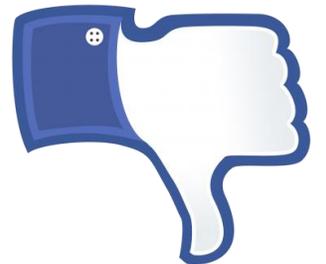
Extraktion einer Datenbasis



TWEET	AVG
1 iphone 5c is ugly as heck what the freak @apple @iphonecompanies	-2
2 @aGounalakis that's nasty! @apple is a nasty brat	-2
3 freak you @apple	-2
4 @APPLE YOU RUINED MY LIFE	-2
5 @apple I hate apple!!!!	-2
6 @apple worst customer service ever.	-2
7 @Apple YOU COW	-2
8 you guys are cheep little freaks and I hope u get testicular cancer and loose a nut @apple	-2
9 We should boycott @Apple or freakin flame them or something like how they have all this money but can't ...	-2
10 wtf @telstra @apple why would you have pre-order for the 5c the crap phone no one wants and not the 5s i...	-2
11 Hate you @apple	-2
12 freak u @apple	-2
13 freak @apple	-2
14 WHY CANT I freakING SEE PICTURES ON MY TL IM ANNOYED freak YOU @TWITTER @APPLE	-2
15 @APPLE YOU freakING COWS freak YOU	-2
16 @apple I hate you why is my phone not working I'm going to freak out	-2
17 freak YOU @APPLE	-2
18 @verizon @apple I HATE y'all	-1,8
19 I freaking hate you @apple	-1,8
20 freak you @apple my battery is stuff, you should be ashamed, that is all	-1,8
21 They beasted RT @Apple: go freak yourself "@RespectRoyalty5: Apple needs to get that bum butt iPhone 5c ...	-1,8
22 @Apple is trying to force me to update the iOS by crashing every time I try editing a photo and acting l...	-1,8
23 freak you for not having a 5S preorder, @Apple.	-1,8
24 pictures on here won't load freak you @twitter @apple	-1,8
25 freak you @apple @siri	-1,8
26 And this is a reason why I despise @Apple and their "Innovations". We didn't like them then OR NOW pic.t...	-1,8
27 freak you @Apple. Hoping your company goes bankrupt. Yours sincerely, an envious yet loyal customer of e...	-1,8
28 I hate how my phone won't focus when I take a picture with flash wow frick off @apple @iphone	-1,8

Ø-Rating pro Tweet

Extraktion einer Datenbasis



Der Weg zum *Sentiment Mining-Modell*

Textanalyse mit Twitter-Daten

Basis:
Tweets



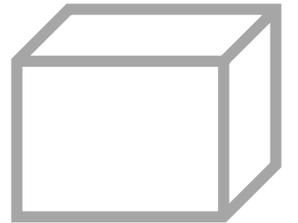
Zielvariable:
Sentiment



Unabhängige
Variablen

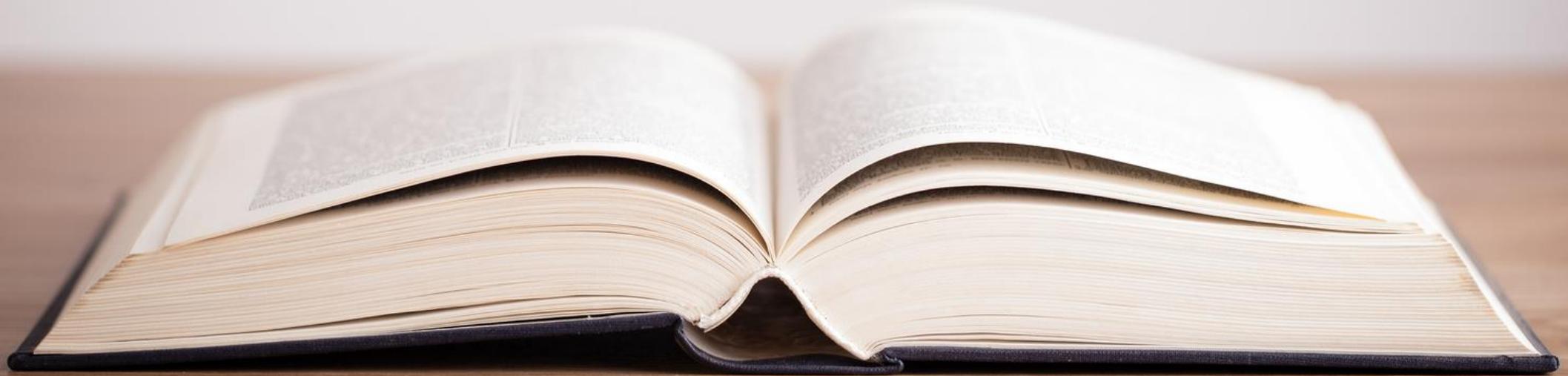


Klassifikations-
modell



Der „Bag Of Words“-Ansatz

- Unstrukturierter Text wird zu strukturierten Informationen transformiert
- Bereinigung der Groß- und Kleinschreibung, Zeichen, Füllwörter
- Stammformreduktion
- Zähle die Anzahl der Stammformen



Bereinigung von Groß- und Kleinschreibung

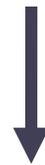
Der „Bag Of Words“-Ansatz

1

Omg! the iPhone 5S is so COOL it can read your finger print to unlock your iPhone 5S and to make purchases without a passcode. #Apple @Apple



Kleinbuchstaben



2

omg! the iphone 5s is so cool it can read your finger print to unlock your iphone 5s and to make purchases without a passcode. #apple @apple

Bereinigung der Zeichensetzung

Der „Bag Of Words“-Ansatz

2

omg! the iphone 5s is so cool it can read your finger print to unlock your iphone 5s and to make purchases without a passcode. #apple @apple



Zeichensetzung entfernen



3

omg the iphone 5s is so cool it can read your finger print to unlock your iphone 5s and to make purchases without a passcode apple apple

Entfernung von sogenannten Stoppworten

Der „Bag Of Words“-Ansatz

3

omg **the** iphone 5s **is so** cool **it** can read **your** finger print **to** unlock **your** iphone 5s **and to** make purchases without **a** passcode **apple apple**



Entfernung von Artikeln, Konjunktionen, Präpositionen, Füllwörtern, etc.



4

omg iphone 5s cool can read finger print unlock iphone 5s make purchases without passcode

Stammformreduktion mit dem Porter-Stemmer-Algorithmus

Der „Bag Of Words“-Ansatz

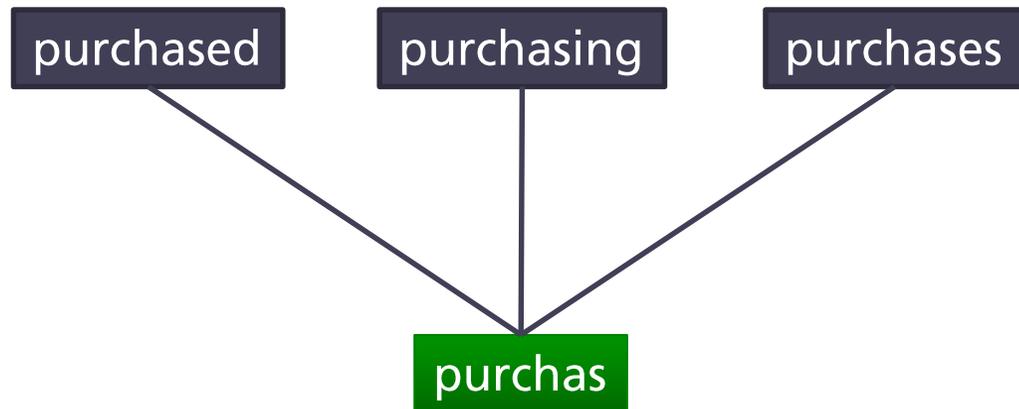


Foto: © Martin Porter | tartarus.org

Porter-Stemmer-Algorithmus:

- Anwendung von Verkürzungsregeln
- Solange Minimalanzahl von Silben erreicht ist

„Stemming“ → Stammformreduktion

Der „Bag Of Words“-Ansatz

4

omg **iphone** 5s cool can read finger print unlock **iphone** 5s make
purchases without **passcode**

↓
Stammformreduktion
↓

5

omg **iphon** 5s cool can read finger print unlock iphon 5s make **purchas**
without **passcod**

Textanalyse mit R

Der „Bag Of Words“-Ansatz

```
library(tm)
library(SnowballC)

tweets = read.csv("tweets_and_rating.csv")

corpus = Corpus(VectorSource(tweets$Tweet))

corpus = tm_map(corpus, tolower)
corpus = tm_map(corpus, PlainTextDocument)
corpus = tm_map(corpus, removePunctuation)
corpus = tm_map(corpus, removeWords, c("apple", stopwords("english")))
corpus = tm_map(corpus, stemDocument)
```

```
corpus[[834]]$content → "omg iphon 5s cool can read finger print unlock
                        iphon 5s make purchas without passcod"
```



Zähle die Stammformen je Tweet

Der „Bag Of Words“-Ansatz

1

omg iphon 5s cool can read
finger print unlock iphon 5s
make purchas without passcod

2

want finger print nsa

	iphon	finger	print	omg	want	...
Summe	2	2	2	1	1	...
Tweet 1	2	1	1	1	0	...
Tweet 2	0	1	1	0	1	...
...

	TWEET_ID	IPHON	LIKE	LOVE	HATE	PHONE	ITUN	WANT
1	16	1	0	1	0	0	0	0
2	32	1	0	0	0	0	0	0
3	1	0	0	0	0	0	0	0
4	64	0	0	1	0	0	0	0
5	109	0	0	0	0	0	0	0
6	113	0	0	1	0	0	0	0
7	94	0	0	0	0	0	0	0
8	105	1	0	0	0	0	0	0
9	120	0	0	0	0	0	0	0
10	147	0	0	0	0	0	0	0
11	168	0	0	0	0	0	0	0
12	172	0	1	0	0	0	0	0
13	203	1	1	0	0	0	0	0
14	215	0	0	0	0	0	0	0
15	226	0	0	0	0	0	0	0
16	309	0	0	0	0	0	0	0
17	313	0	0	0	0	0	2	0
18	255	0	0	0	0	0	3	0
19	275	0	0	0	0	0	0	0
20	278	0	1	0	0	0	0	0
21	308	0	0	0	0	0	0	0
22	371	0	0	0	0	0	0	0
23	333	0	0	0	0	0	0	0
24	410	1	0	0	0	0	0	0
25	425	1	0	0	0	0	0	0
26	427	1	0	0	0	0	0	0
27	432	0	0	0	0	0	0	0

Stammformen → Variablen
Die Anzahl zum Wert der Variablen

```

sparseDTM = removeSparseTerms(frequencies, 0.995)
sparse = as.data.frame(as.matrix(sparseDTM))
colnames(sparse) = make.names(colnames(sparse))

```



Der Weg zum Sentiment Mining-Modell

Textanalyse mit Twitter-Daten

Basis:
Tweets



Zielvariable:
Sentiment

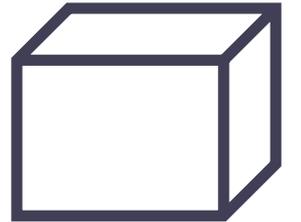


Unabhängige
Variablen

Bag Of
Words



Klassifikations-
modell



Vereinfachung der Zielvariablen

Aus Zahlen wird eine binäre Entscheidung



	TWEET_ID	RATING	IPHON	LIKE	LOVE	HATE	PHONE	ITUN	WANT
1	777	Positive	2	0	0	0	0	0	
2	1163	Negative	1	0	0	1	0	0	
3	1140	Negative	1	0	0	0	0	0	
4	1110	Negative	1	0	0	0	0	0	
5	1077	Negative	1	0	0	0	0	0	
6	1035	Negative	1	0	0	0	0	0	
7	795	Positive	1	0	0	0	0	0	
8	557	Positive	1	0	0	0	0	0	
9	292	Positive	1	0	0	0	0	0	
10	169	Positive	1	0	0	0	0	0	
11	439	Positive	0	0	0	0	0	0	
12	107	Positive	0	0	0	0	0	0	
13	544	Positive	0	0	0	0	0	0	
14	1157	Negative	0	1	0	0	0	0	
15	569	Positive	0	1	0	0	0	0	
16	622	Positive	0	0	0	0	0	0	
17	672	Positive	0	0	0	0	0	0	
18	700	Positive	0	0	0	0	0	0	
19	703	Positive	0	0	0	0	0	0	
20	1123	Negative	0	0	0	1	0	0	
21	115	Positive	0	0	0	0	0	0	
22	886	Positive	0	0	0	0	0	0	
23	928	Positive	0	0	0	0	0	0	
24	899	Positive	0	0	0	0	0	0	
25	901	Positive	0	0	0	0	1	0	
26	946	Positive	0	0	0	0	0	0	
27	965	Positive	0	0	0	0	0	0	
28	976	Positive	0	0	0	0	0	1	
29	1017	Negative	0	0	0	0	0	0	
30	1063	Negative	0	0	0	0	0	0	
31	1174	Negative	0	0	0	1	1	0	

Binäre Zielvariable

Bildung des Klassifikationsmodells

```
sparse$Rating = 'Negative'
sparse[tweets$Avg >= -1,]$Rating = 'Positive'
```

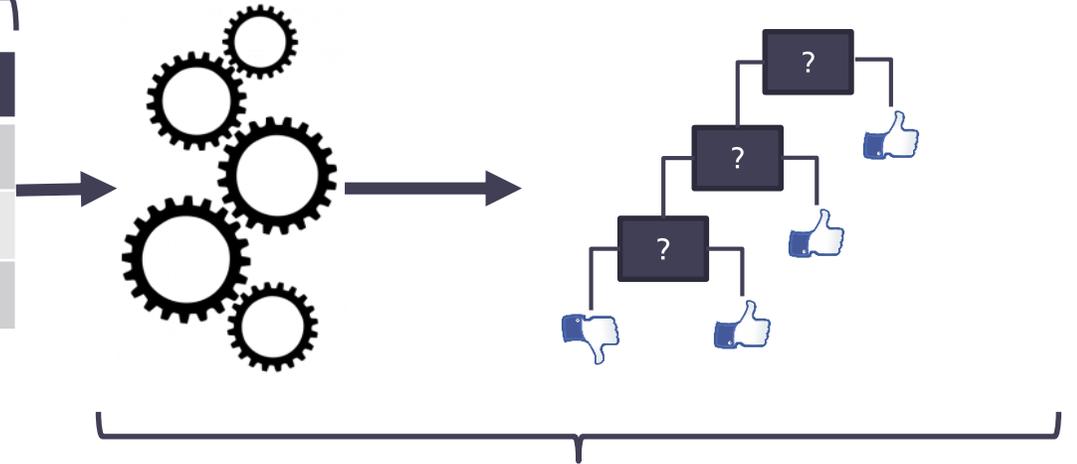


Klassifikationsmodell

Ist ein Tweet negativ oder positiv?

Anzahl der Stammformen je Tweet erklären die Zielvariable

	iphon	finger	print	omg	want	...	Rating
Tweet 1	1	1	1	1	0	...	
Tweet 2	0	1	1	0	1	...	
...

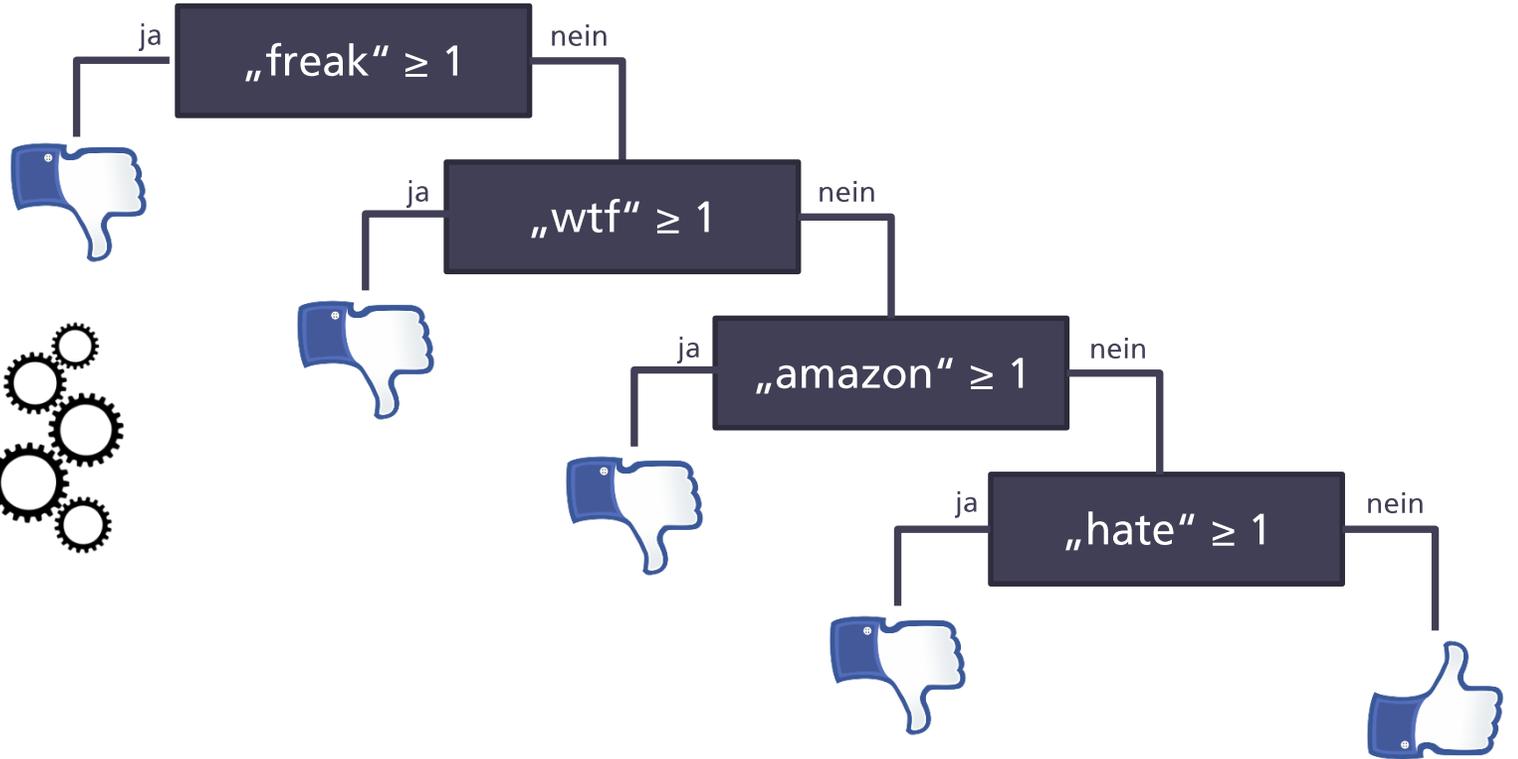


Die Klassifikation erfolgt anhand eines Entscheidungsbaumes.

1181 Tweets

Training
80%

Test
20%



Klassifikationsmodell

Ist ein Tweet negativ oder positiv?

Berechnung des Klassifikationsmodells mit R

Ist ein Tweet negativ oder positiv?

```
library(caTools)  
library(rpart)
```

```
set.seed(123)
```

```
split = sample.split(sparse$Rating, SplitRatio = 0.8)
```

```
trainSparse = subset(sparse, split == TRUE)  
testSparse = subset(sparse, split == FALSE)
```

```
decTree = rpart( Rating ~ . -TWEET_ID, data=trainSparse, method="class")
```

```
pred = predict(decTree, newdata=testSparse, type="class")
```



Evaluation des Modells

Ist ein Tweet negativ oder positiv?

(...)

```
predictCART = predict(tweetCART, newdata=testSparse, type="class")  
table(testSparse$Negative, pred)
```



	Positiv	Negativ
Positiv (real)	18	37
Negativ (real)	6	294

→ $\frac{294 + 18}{294 + 6 + 37 + 18}$ → 84,5%



Fazit & Ausblick



Referenzmodell in Textanalyse-Projekten

„Bag Of Words“

Social Media als Datenbasis

Die Häufigkeit eines jeden
Wortes wird gezählt

80% Text

Extraktion von Wissen aus Texten

Text Mining

Menschliche
Intelligenz nicht
ersetzbar

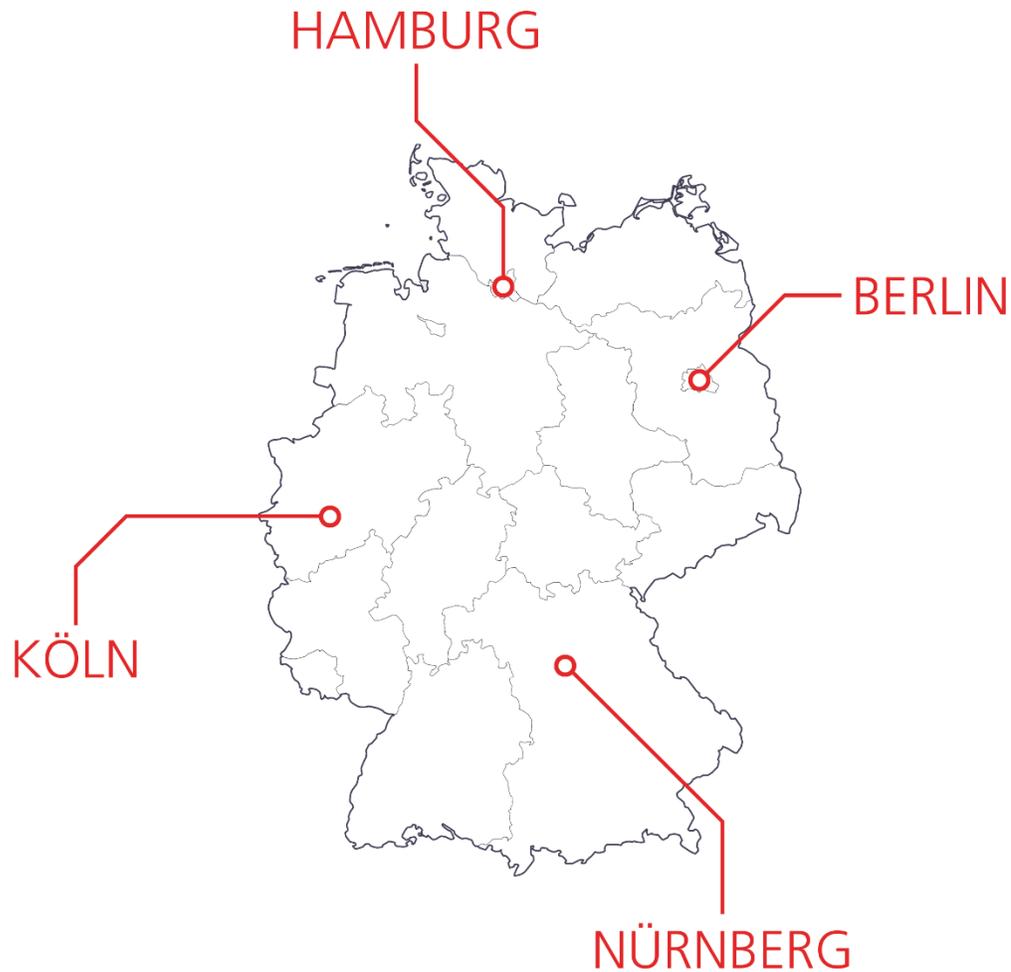
Hohe Genauigkeit mit einfachen Mitteln

Welche Fragen darf ich Ihnen noch beantworten?

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Data Warehouse
Business Intelligence

areto
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