Twitter is a great source for sentiment data and social media mining furthermore it is quite easy to get significant amounts of data to be able to scrape data from Twitter you need a standard Twitter account and you need to update it to a developer account

- note that Twitter limits the amount of searches you can perform (15min: 15 scrapes)
- package twitteR

library("twitteR")

- all this info is obtained for the Twitter developer account:
 - o key = "your key"
 - o secret = "your secret"
- set a working directory for the whole process you need to download a few files and R needs to know where to look for that stuff

setwd("C:/Users/HCBM/Desktop/Mining Course")



- this is a crucial step at least for windows users
- if you are on Linux or Mac you might skip this step, Win needs the certificate collection
- Cacert.pem is a collection of certificates

download.file(url="http://curl.haxx.se/ca/cacert.pem",

destfile="C:/Users/HCBM/Desktop/Mining Course/cacert.pem",
method="auto")

- hint: download.file is really handy when it comes to downloading material from the web
- the dest file is the location on your computer, here it is my working directory
- and url points to the place where you want to get the file from
- we are entering the whole Twitter API info and call the whole object authenticate

authenticate <- OAuthFactory\$new(consumerKey=key,

consumerSecret=secret,

requestURL='https://api.twitter.com/oauth/request_token',

accessURL='https://api.twitter.com/oauth/access_token',

authURL='https://api.twitter.com/oauth/authorize')



- this will get you to a Twitter Site obtain the PIN
- the whole process is meant to provide the signature for your Twitter usage

authenticate\$handshake(cainfo="C:/Users/HCBM/Desktop/Mining Course/cacert.pem")

- insert the PIN from Twitter

4264654

save(authenticate, file="twitter authentication.Rdata")

registerTwitterOAuth(authenticate)



Let's start with the Twitter scraping!

library("twitteR")

 we need to specify the cainfo to avoid a SSL cert error - this is for Windows machines

Let's check the latest tweets of Udemy userTimeline("Udemy", cainfo="cacert.pem")

- searchTwitter is the main function of the package

?searchTwitter

- arguments: since and until are for time specifications
- lang: for languge specification
- geocode: for location specification
- we are now scraping 1k tweets for Udemy, and we als specify our certificate

udemytweets = searchTwitter("#Udemy", n=1000, cainfo="cacert.pem")



- as you can see, scraping that data is quite time consuming your machine limits the efficiency and speed of your mining
- if you are plan to scrape a lot in the future 64bit systems and high RAM is desirable

class(udemytweets)

length(udemytweets)

head(udemytweets)

library("tm")

udemylist <- sapply(udemytweets, function(x) x\$getText()) # initiating a
function</pre>

 in depth info about the apply family and functions in the course "R Level 1"

udemycorpus <- Corpus(VectorSource(udemylist)) # use the corpus
function</pre>

 a corpus is the text body consisting of all the text including the meta info



udemycorpus <- tm_map(udemycorpus, tolower) # putting text to lower
case</pre>

udemycorpus <- tm_map(udemycorpus, removePunctuation) # remove
punct.</pre>

udemycorpus <- tm_map(udemycorpus,</pre>

function(x)removeWords(x,stopwords())) # remove
stopwords (meaningless words)

- there is a link to a stop word list in the link lecture

Let's see which other transformations tm offers ?getTransformations

to trasform to plain text which wordcloud can use
 udemycorpus <- tm_map(udemycorpus, PlainTextDocument)

library("wordcloud")

? wordcloud



wordcloud(udemycorpus, min.freq=4, scale=c(5,1),
 random.color=F, max.word=45, random.order=F)

- changing to a tdm

udemytdm <- TermDocumentMatrix(udemycorpus)</pre>

A DocumentTermMatrix is a very useful tool when it comes to text mining. It structures the text in a matrix where each term is organized in a column. Each row is a document and the number represents the counts of that term.

udemytdm

- frequent terms

findFreqTerms(udemytdm, lowfreq=11)

?findFreqTerms

associations

findAssocs(udemytdm, 'android', 0.60)



Let's get a dendrogram to see related terms

- remove sparse (infrequently used) terms from the term-document matrix

udemy2tdm <-removeSparseTerms(udemytdm, sparse=0.9)</pre>

Let's scale the data
udemy2tdmscale <- scale(udemy2tdm)

- distance matrixudemydist <- dist(udemy2tdmscale, method = "euclidean")
- hierarchical clustering
 udemyfit <- hclust(udemydist)
- Visualize the result plot(udemyfit)
- to calculate a certain number of groups
 cutree(udemyfit, k=6)



we can even color the 6 groups and plot them
 rect.hclust(udemyfit, k=6, border="red")

SENTIMENT ANALYSIS

Sentiment analysis is used to see if a text is neutral, positive or negative emotion analysis is used to see which emotion a text has (happy, fear, anger) both are using similar codes but the comparison lexicon is different.

Example: What is the sentiment towards my company?

Twitter data is useful for that type of analysis because:

- high volumes (500 mill/day)
- short messages like sms 140 words
- special strings (hashtags)
- but creative word usage makes it hard for analysis, spelling mistakes

There is TONS of sentiment in it!



Example of text sentiments:

- #1. Udemy provides great opportunity for life long learning both words would hint for pos
- #2. *Udemy is too expensive and slow* both words would hint towards neg sentiment
- #3. Udemy is a learning platform neutral sentiment
- #4. Problem: what? Udemy is a fast platform? sarcasm is hard to analyse

Sentiment Lexicon: a list of words which you are using to compare your scraped txt with

Hu Liu Lexicon got the standard of sentiment analysis lately list of pos and negative words - manually created - approx. 6800

- download the txt files to your wd
- import positive and negative words

pos = readLines("positive_words.txt")

neg = readLines("negative_words.txt")



Let's run a test to see how this works!

mytest= c("great you re here", "awesome experience",

"You had a bad night", "She loves ugly candy")

the score.sentiment function is self writtentestsentiment = score.sentiment(mytest, pos, neg)

class (testsentiment)

testsentiment\$score

output corresponds to the 4 test sentences - sentences can be manipulated

Let's do the whole process: writing the function and scraping - approach after J. Breen

library("stringr")

library("plyr")



function score.sentiment - this is how the whole function is written
 score.sentiment = function(sentences, pos.words, neg.words,
 .progress='none')

Parameters

- sentences: vector of text to score
- pos.words: vector of words of postive sentiment
- neg.words: vector of words of negative sentiment
- .progress: passed to laply() to control of progress bar
- create simple array of scores with laply

- remove punctuation using global substitute
 sentence = gsub("[[:punct:]]", "", sentence)
- remove control characters

```
sentence = gsub("[[:cntrl:]]", "", sentence)
```

- remove digits

sentence = gsub('\\d+', ", sentence)



define error handling function when trying tolowertryTolower = function(x)

{

create missing value

$$y = NA$$

tryCatch error

```
try_error = tryCatch(tolower(x), error=function(e) e)
```

- if not an error

```
if (!inherits(try_error, "error"))
y = tolower(x)
```

- result

```
return(y)
}
```

- use tryTolower with sapply

```
sentence = sapply(sentence, tryTolower)
```

split sentence into words with str_split (stringr package)

```
word.list = str_split(sentence, "\\s+")
words = unlist(word.list)
```



compare words to the dictionaries of positive & negative terms
 pos.matches = match(words, pos.words)
 neg.matches = match(words, neg.words)

- get the position of the matched term or NA
- we just want a TRUE/FALSE

```
pos.matches = !is.na(pos.matches)
neg.matches = !is.na(neg.matches)
```

final score

}

```
score = sum(pos.matches) - sum(neg.matches)
return(score)
}, pos.words, neg.words, .progress=.progress)
```

data frame with scores for each sentence
 scores.df = data.frame(text=sentences, score=scores)
 return(scores.df)



- tweets for country

```
usatweets = searchTwitter("usa", n=900, lang="en", cainfo="cacert.pem")
indiatweets = searchTwitter("india", n=900, lang="en",
cainfo="cacert.pem")
russiatweets = searchTwitter("russia", n=900, lang="en",
cainfo="cacert.pem")
chinatweets = searchTwitter("china", n=900, lang="en",
cainfo="cacert.pem")
```

get text

```
usa_txt = sapply(usatweets, function(x) x$getText())
india_txt = sapply(indiatweets, function(x) x$getText())
russia_txt = sapply(russiatweets, function(x) x$getText())
china_txt = sapply(chinatweets, function(x) x$getText())
```

how many tweets of each country

```
nd = c(length(usa_txt), length(india_txt), length(russia_txt),
length(china_txt))
```

- join texts

country = c(usa_txt, india_txt, russia_txt, china_txt)



- apply function score.sentimentscores = score.sentiment(country, pos, neg, .progress='text')
- scores\$country = factor(rep(c("usa", "india", "russia", "china"), nd))
 scores\$very.pos = as.numeric(scores\$score >= 2)
 scores\$very.neg = as.numeric(scores\$score <= -2)</pre>
- how many very positives and very negativesnumpos = sum(scores\$very.pos)numneg = sum(scores\$very.neg)

- add variables to data frame

- global score
global_score = round(100 * numpos / (numpos + numneg))

head(scores)

boxplot(score~country, data=scores)



library("lattice")

histogram(data=scores, ~score|country, main="Sentiment Analysis of 4 Countries", xlab="", sub="Sentiment Score")

