

Benefits of mining for a brand?

analysis to discover customer's sentiment for a brand

You can do sentimental

popularity using the actively engaged tweeters

You can measure brand

It is used to identify the pain points of customers i.e. customer relationship management

It is widely used for predictions and forecasting



The Business Problem Let's say, we want to find the features of an Apple iPhone which are most popular amongst the fans on Twitter.

What to do next?

We've extracted all the tweets related to consumer opinions of iPhone.

Here's a sample tweet on which we'll perform data cleaning TWEET "I luv my <3 iphone & you're awsm apple. DisplayIsAwesome, sooo happppppy:) http://www.apple.com"

steps for Data eleaning



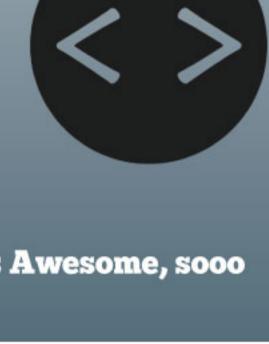
Decoding data



import HTMLParser html_parser = HTMLParser.HTMLParser()

tweet = html_parser.unescape(original_tweet) Output >> "I luv my <3 iphone & you're awsm apple. Display Is Awesome, sooo

happppppy http://www.apple.com"



STEP

02

Code tweet = original_tweet.decode("utf8").encode('ascii','ignore')

Output

sooo happppppy:) http://www.apple.com"

» "I luv my <3 iphone & you're awsm apple. DisplayIsAwesome,

STEP

words = tweet.split()

03 Code

Apostrophe Lookup

APPOSTOPHES = {"'s" : " is", "'re" : " are", ...} ## Need a huge dictionary

reformed = [APPOSTOPHES[word] if word in APPOSTOPHES else word for word in words]



>> "I luv my <3 iphone & you are awsm apple. DisplayIsAwesome, sooo happppppy:) http://www.apple.com"

Removal of Stop-Words 04

When data analysis needs to be data driven at the word level, the

predefined language specific libraries.

commonly occurring words (stop-words) should be removed. One can either create a long list of stop-words or one can use

STEP

Removal of Punctuations STEP 05 All the punctuation marks according to the priorities should be dealt with. For example: ".", ",","?" are important punctuations

that should be retained while others need to be removed.

Textual data (usually speech transcripts) may contain human expressions like [laughing], [Crying], [Audience paused]. These expressions are usually non relevant to content of the speech and

Split Attached Words

Removal of Expressions

STEP

08

STEP

10

STEP

06

cleaned = " ".join(re.findall('[A-Z][^A-Z]*', original_tweet)) Outcome >> "I luv my <3 iphone & you are awsm apple. Display Is Awesome, sooo happppppy:) http://www.apple.com"

STEP

09

hence need to be removed.

STEP

07

Code

tweet = _slang_loopup(tweet)

Slangs lookup

Outcome

Code

Outcome

Code tweet = ".join(".join(s)[:2] for _, s in itertools.groupby(tweet))

"I love my <3 iphone & you are awesome apple. Display Is</p>

Awesome, so happy:) http://www.apple.com

Standardizing word

Removal of URLs

"I love my <3 iphone & you are awesome apple. Display Is</p>

Awesome, sooo happppppy:) http://www.apple.com"

URLs and hyperlinks in text data like comments, reviews, and tweets should be removed.

Final cleaned tweet:

Grammar checking

Spelling correction

http://bit.ly/1JjslYe

http://bit.ly/1LDPF6c

"I love my iphone & you are awesome apple. Display Is Awesome, so happy!", <3,:)

Grammar checking is majorly learning based, huge amount of proper text data is learned and models are created. Many online tools are available for grammar correction purposes.

Advanced Data Cleaning

In natural language, misspelled errors are

following techniques (in no order) of Text Mining-

Dictionary Lookup etc. other modules and packages to fix these errors.

encountered. One can use algorithms like the Levenshtein Distances,

Your Next Steps...

1. Framework to build a niche dictionary for text mining http://bit.lu/1eetMw6

Now that the data (tweet) is cleaned, you are ready to practice and learn the

3. 2014 FIFA World Cup Prediction using Twitter Mining http://bit.ly/1kLeYSk

2 Step by Step guide to extract insights from free text

- 4. Text Mining Hack using Google API

For more resources on analytics/data science, visit www.analyticsvidhya.com

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