

# State of the Art in Twitter Sentiment Analysis

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## ABSTRACT

The extensive growth of user-generated content in social networks and the common usage of emoticons and hashtags has introduced new possibilities to classify these information. In this paper, an overview of the state-of-the-art regarding sentiment analysis of Twitter messages is provided. The usefulness of existing lexical resources and special expressions like smileys or hashtags is evaluated and an general introduction to sentiment analysis is given as part of the introduction.

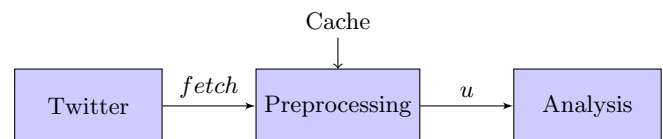
## 1. INTRODUCTION

Within the last years Twitter and similar social media platforms grew considerably in terms of user numbers and mainstream fame. Twitter has about 284 Mio. active users, generating approximately 500 Mio. posts<sup>1</sup> on a single day<sup>2</sup>. These impressive numbers show the amount of information generated by the crowd. Furthermore Twitter is able to reach a vast number of potential customers especially for consumer businesses. As a result the named services gain more and more acceptance as opinion platforms and powerful tools for both, opinion leader as well as pollster. This trend influenced various marketing and sales strategies and created a new market for companies specialized on sentiment analysis of tweets, like tweetfeel<sup>3</sup>, Social Mention<sup>4</sup> and Twitratr<sup>5</sup>.

The aim of sentiment analysis is to determine and try to measure positive and negative feelings, emotions and opinions written in a text. English as language allows to express

the same intent in different ways. The main challenge therefore consists in abstracting the intention of the writer from the grammatic and language specific rules.

In general sentiment analysis can be splitted into 2 steps: a preprocessing and a constitutive sentiment analysis phase. In Twitter sentiment analysis there is another step right bevor preprocessing: fetching the data from the application programming interface (API). This task is not trivial, because the official API is limited regarding the datasets fetched within certain intervals. This leads to the requirement of a caching mechanism to avoid fetching the same tweets multiple times and allow the usage of more datasets.



While sentiment analysis of conventional resources like news papers or articles is quite well investigated the special research area of sentiment analysis in terms of social network posts is relatively unexplored. Especially the possibility of prioritizing keywords<sup>6</sup> and the common usage of emoticons lead to advanced possibilities to categorize feelings of posts.

## 2. RELATED WORK

Sentiment analysis a growing and well explored part of Natural Language Processing (NLP). There are several papers from Pang and Lee regarding sentiment analysis in general and related topics like the effects of various machine learning approaches[9][8]. Especially the last topic is a well studied field [7]. Some related research areas are document level classification, sentence-level classification and machine learning. Due the limited number of characters within a single twitter post, this topic is quite similar to sentence-level sentiment analysis. There are several recent papers regarding sentiment analysis in the context of twitter posts. Agarwal et al. published within "Sentiment analysis of Twitter data" general information and techniques which cover this area[1]. In "Twitter Sentiment Analysis : The Good the Bad and the OMG !" the utility of linguistic features for de-

<sup>1</sup>posts are also known as tweets

<sup>2</sup><https://about.twitter.com/company>,  
12.11.2014

<sup>3</sup>[www.tweetfeel.com](http://www.tweetfeel.com)

<sup>4</sup>[www.socialmention.com](http://www.socialmention.com)

<sup>5</sup>[www.twitratr.com](http://www.twitratr.com)

Effective

<sup>6</sup>known as hashtags

tecting the sentiment of tweets are investigated. Saif, He, and Alani introduce a novel approach of adding semantics as additional features into the training set. For each extracted entity (e.g. Galaxy S) from Twitter Messages, they add a semantic concept (e.g. Samsung product) as an additional feature, and measure the correlation of the representative concept with negative/positive sentiment[10]. In *Twitter Sentiment Analysis* a algorithm is presented, which accurately classifies tweets as positive or negative, in respect to a query term[3].

### 3. DATA

Twitter is a social network and a microblogging service that allows their users to compose and read so called *tweets*. The user registration it is not limited to professional writers, authors or similar expert groups and there are no quality safeguards or writing guidelines.

Each tweet is a text message and can contain up to 140 characters including links, pictures<sup>7</sup>, and special tagged words. These special tagged words start with a well defined prefix and add semantic information to the prefixed string. Table 1 shows an overview of common prefixes supported by Twitter and their purpose. The hashtag # labels a word, also known as hashtag. It is used to mark important keywords within tweets. There is also a at-sign-prefix @, which allows to mention and link other Twitter users in a post. Tweet composers can insert URLs as part of their messages. Regarding the hard length limitation of a tweet, url-shortener are often used to insert longer internet addresses.

prefix	name of character	meaning
#	hashtag	keyword, index
@	at-sign	twitter username

**Table 1: Common Twitter word prefixes**

The combination of the 140 character restriction and the missing writing guidelines leads to another problem: Slang, emoticons and punctuation that are uncommon in longer text messages are heavily used.[2] In terms of sentiment analysis, tweet processing is more comparable to sentence level analysis as paragraph analysis.

### 4. PREPROCESSING

The data preprocessing step is an essential part of sentiment analysis. Its goal is to prepare data for the sentiment analysis and remove noisy, unnecessary, inconsistent or incomplete parts.[4] In a first step a tokenizer splits the tweet into conjugate parts, called words. The resulting list of words is filtered regarding the word class, the word itself and the intention of the sentiment analysis. In case of sentiment analysis of twitter posts it is common to remove the parts described in the following paragraphs.[4]

**URLs.** Uniform Resource Locators (URLs) refer to another site, which in fact can contain very important information, but the internet adress on its own is not of interest, because

<sup>7</sup>pictures are replaced by a url targeting a public available image resource

in the most cases it contains only auto-generated information, which can not be influenced by the poster.

**Unnecesarry words.** It is not a obvious task to classify if a word is unnecesarry or not. But some word classes, like pronouns, articles or interrogatives do not contain any relevant information and can be skipped. (TODO:Referenz)

**Retweets.** Retweets reflect the postion of another person and are not always created to express the same position (e.g. statement of sarcasm).

**Repeated characters.** Tweets contain often deformed words containing repeated characters like 'heeeey' or 'whuhuuu'. In this cases it is intended to reconstruct the correct english term and only if it is not possible or to sophisticated, the word is removed by the preprocessor.

#### 4.1 Tokenizer

There are already several open source tokenizer implementations available. Some popular programming libraries for the english language include the Stanford NLP PTBTokenizer<sup>8</sup> and the Apache OpenNLP Tokenizer<sup>9</sup>. These examples are quite versatile and can be even used for different tasks and languages. To overcome the special needs of sentiment analysis in terms of Twitter posts, highly specialized Tokenizer were created (e.g. Carnegie Mellon Twokenizer<sup>10</sup>). Some tokens like emoticons are difficult to handle. Although they are a good and popular ressource for sentiment analysis. [5] It is a sophisticated task to recognize them correctly and some Tokenizers like the Stanford NLP PTBTokenizer even skips them per default. The Carnegie Mellon's Twokenizer recognizes the related parts and forwards them to the Tagger.

#### 4.2 Tagger

### 5. METHODS

### 6. CONCLUSIONS

### APPENDIX

#### A. HEADINGS IN APPENDICES

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<sup>8</sup><http://nlp.stanford.edu/software/tokenizer.shtml>

<sup>9</sup><http://opennlp.apache.org/documentation/manual/opennlp.html>

<sup>10</sup><http://www.ark.cs.cmu.edu/TweetNLP/>

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