DataComp2 - Project Spark & Elasticsearch

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1 Introduction and Motivations

The goal of our project is to make a short program to process and analyze real-time stream data (in this case, tweets) using Sparklab Streaming[3], index some of this data with ElasticSearch[1] and evaluate some queries with it. Our project will be devided in three main parts that we will detail in this report.



Figure 1: Schema

2 Processing Twitter 2.2 Stream Listener Class Data In order to get the tweets in rea

2.1 Settings: Defining Twitter developer credentials and Authenticating the connection

The first step was to create a Twitter Application[2] so as we can get both the Consumer Keys and the Authentication Tokens. Using the tweepy library[4], we can retrieve Tweets from the Twitter API and store them into sockets before making any analysis on them as illustrated in figure1.

For this project, we focused on tweets containing the keyword **netflix**.

In order to get the tweets in real-time, we have to implement a new class that inherits from the Stream module of Tweepy. The constructor takes a Socket object and calls the super constructor, pretty straightforward. Then, we overwrite the on_data[5] method to adapt it to our needs.

Figure 2:

The on_data method will format the received JSON file (a tweet) into a string. We only keep some features such as the name, the location, hashtags (if any) and the date. These different fields are are delimited by the tag "###field### field_name:".

Note that the \n is replaced by a simple space "". So we make sure that each tweet corresponds to just one line and it's correctly read when creating the DStream in spark with the function socketTextStream().

2.3 Establish connexion with client

3 Spark Analysis

3.1 Process data

We process the data we receive by first, spliting the incoming string so each row in our datastream 'tweets' be a tuple where each element will correspond to a field of the tweet received. And using textblob, we perform sentiment analysis on the text of the tweet and we add this sentiment to the previous tuple. So each tweet will be represented by a tuple of the form (user, text, date, location, hashtags, sentiment).

Where sentiment equals -1 if the tweet is more likely negative. It will equal 1 if it's more positive, and 0 otherwise.

Remark: We chosed to take into account the polarity returned by the TextBlob text processing library and to ignore the subjectivity.

4 ElasticSearch Indexing

References

- [1] E. Elastichsearch. Official documentation. https://www.elastic.co/guide/en/elasticsearch/reference/current/index.html.
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- [3] S. Streaming. Official documentation. https://spark.apache.org/docs/latest/streaming-programming-guide.html.
- [4] Tweepy. Index official documentation. https://docs.tweepy.org/en/stable/index.html.
- [5] Tweepy. Streaming. https://docs.tweepy.org/en/stable/streaming.html.