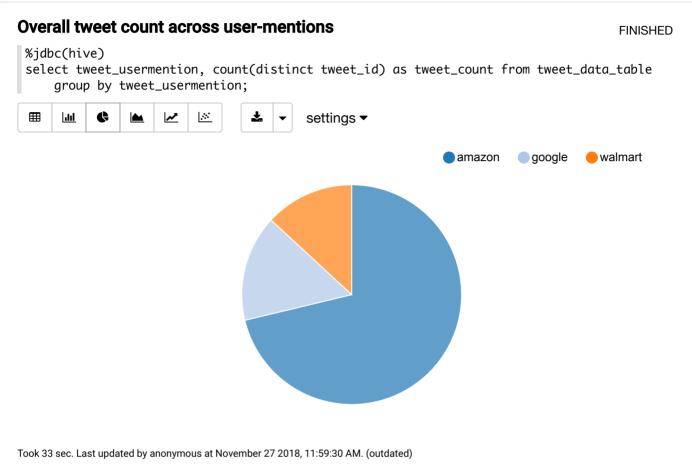
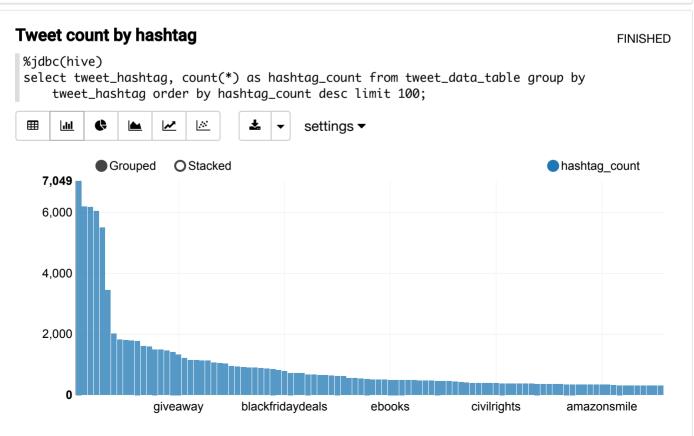
### **Tweets & Hashtags**





Took 44 sec. Last updated by anonymous at November 27 2018, 4:12:44 AM. (outdated)

#### Hashtag graph for Google **FINISHED** %jdbc(hive) select tweet\_hashtag, count(\*) as hashtag\_count from tweet\_data\_table where tweet\_usermention = 'google' group by tweet\_hashtag order by hashtag\_count desc limit 25; ▦ ılıl <u>₩</u> <u>.::</u> settings ▼ google dropdragonfly iot ux pixel3 teampixel ui wearabletech sensors gesture recognition ai machinelearning tech seo espy4all espy4senate innovation espyformississippi nohydesmith bestbuy unforumbhr privacy dragonfly Took 42 sec. Last updated by anonymous at November 27 2018, 12:05:32 PM. (outdated)

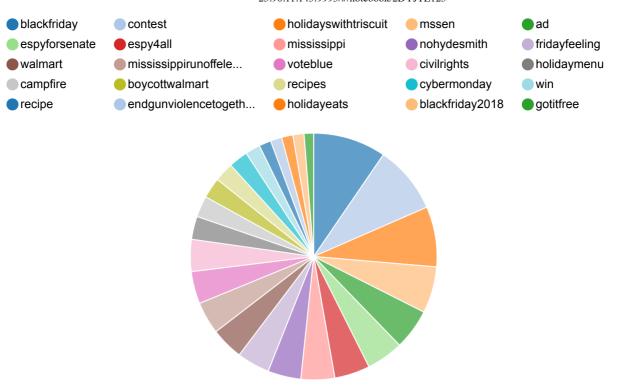
#### **Hashtag graph for Walmart**

**FINISHED** 

%jdbc(hive)

select tweet\_hashtag, count(\*) as hashtag\_count from tweet\_data\_table where
 tweet\_usermention = 'walmart' group by tweet\_hashtag order by hashtag\_count desc limit
25;



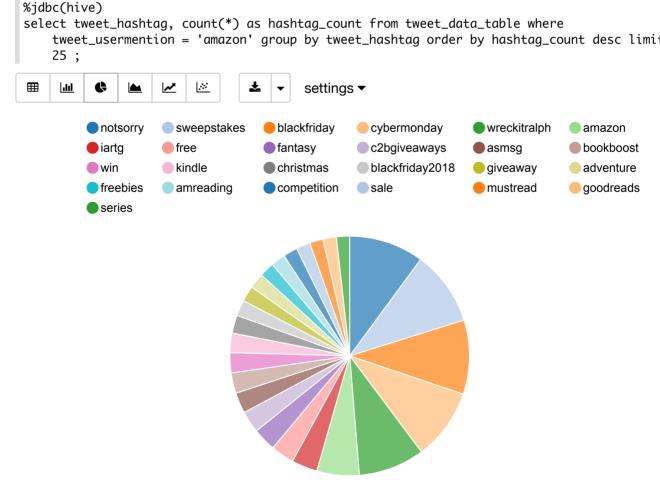


Took 44 sec. Last updated by anonymous at November 27 2018, 12:07:30 PM. (outdated)

#### **Hashtag graph for Amazon**

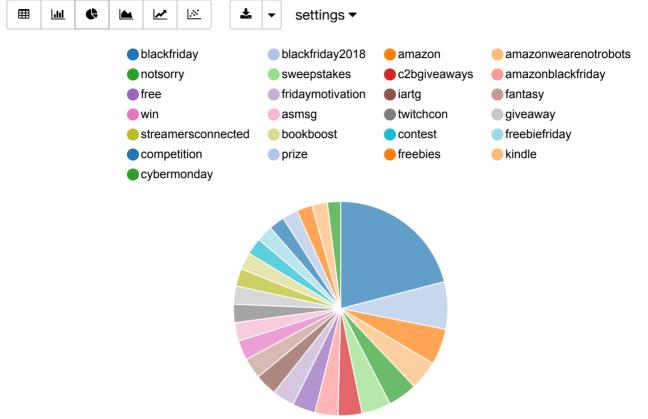
**FINISHED** 

tweet\_usermention = 'amazon' group by tweet\_hashtag order by hashtag\_count desc limit



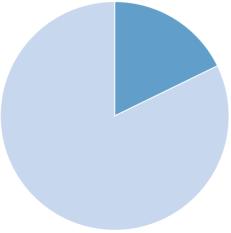
Took 43 sec. Last updated by anonymous at November 27 2018, 12:16:27 PM. (outdated)

# Hashtag graph for Amazon with timestamp query %jdbc(hive) select tweet\_hashtag, count(\*) as hashtag\_count from tweet\_data\_table where tweet\_usermention = 'amazon' and tweet\_timestamp >= '2018-11-22' and tweet\_timestamp <= '2018-11-24' group by tweet\_hashtag order by hashtag\_count desc limit 25;</pre>

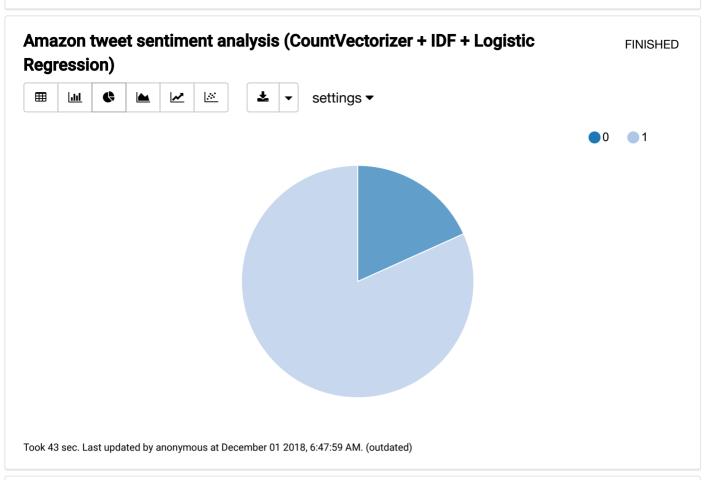


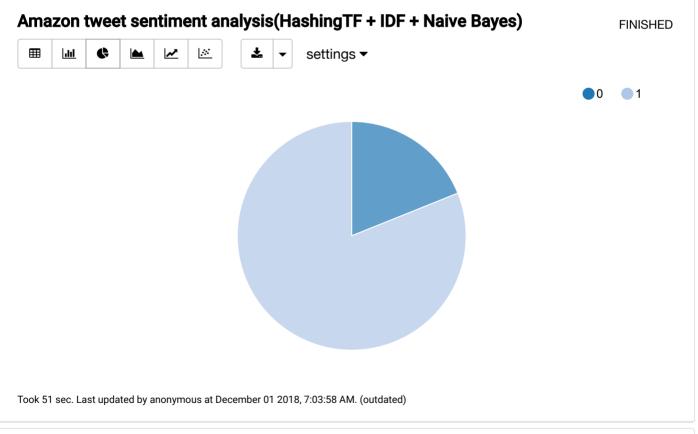
Took 43 sec. Last updated by anonymous at December 01 2018, 5:54:59 AM. (outdated)



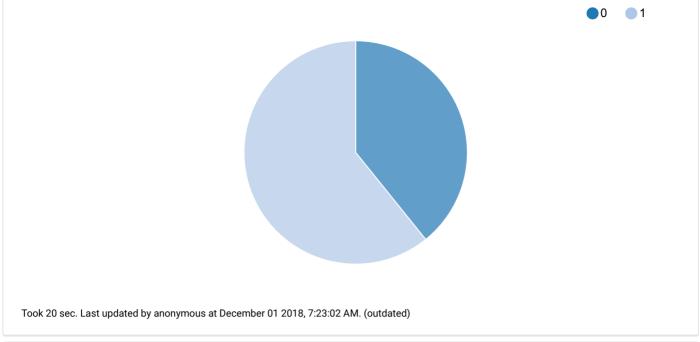


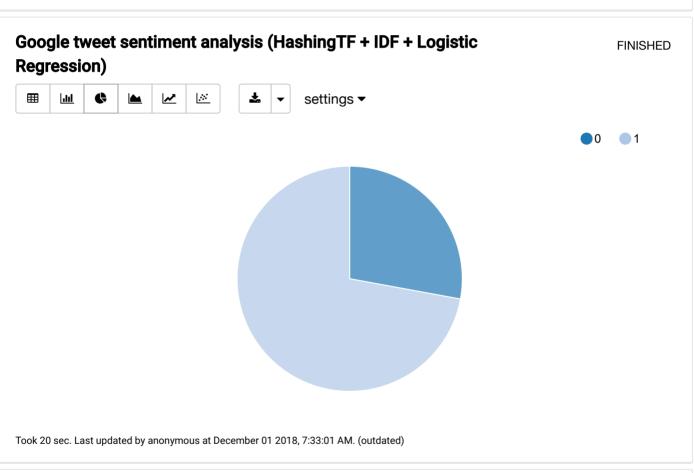
Took 23 sec. Last updated by anonymous at December 01 2018, 6:58:53 AM. (outdated)

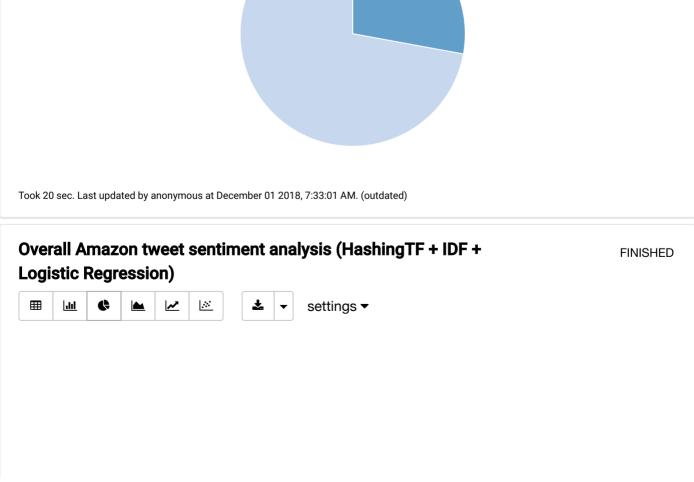


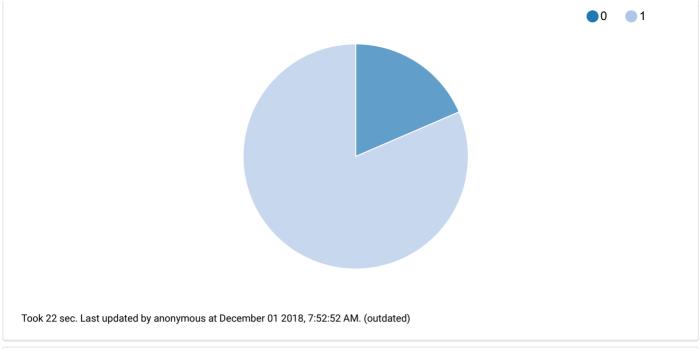


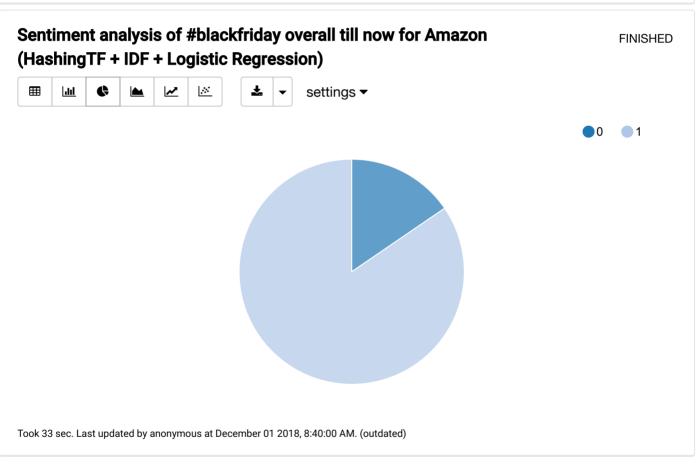








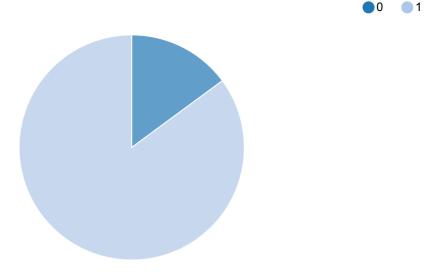




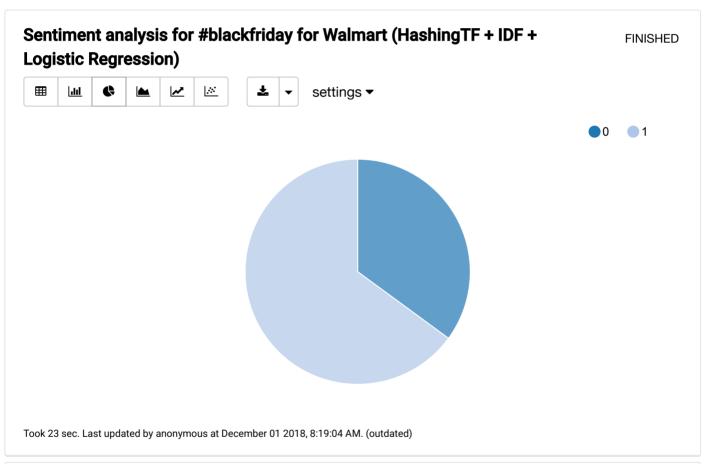
## Sentiment analysis for #blackfriday for Amazon(HashingTF + IDF + Logistic Regression)

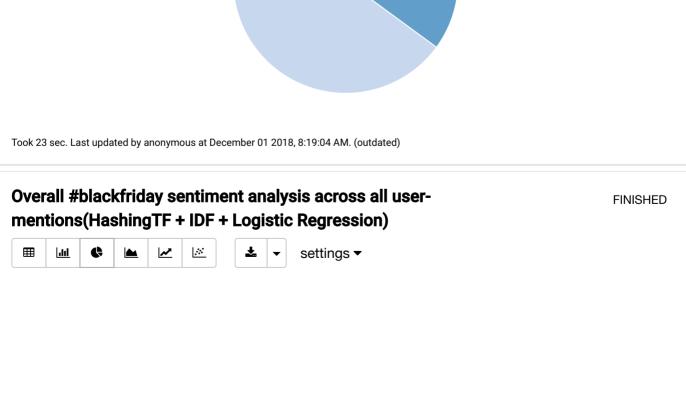
**FINISHED** 

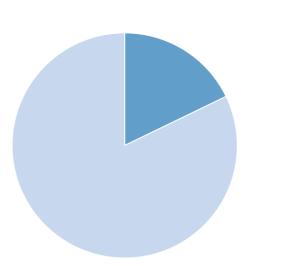




Took 20 sec. Last updated by anonymous at December 01 2018, 8:11:27 AM. (outdated)







Took 22 sec. Last updated by anonymous at December 01 2018, 8:32:58 AM. (outdated)

```
%livy2.pyspark
                                                                                   FINISHED
import re
from pyspark import SparkContext
from pyspark.ml import Pipeline
from pyspark.ml.classification import LogisticRegression, NaiveBayes
from pyspark.ml.feature import Tokenizer, StopWordsRemover, CountVectorizer, IDF, HashingTI
from pyspark.sql import SparkSession
def train_filter_func(row):
    row_dict = row.asDict()
    val = row_dict['sentiment']
    if val in ['0', '4']:
        return True
    else:
        return False
def int_cast(row):
    row_dict = row.asDict()
    val = int(row_dict['sentiment'])
    if val == 0:
        row_dict['sentiment'] = 0
    else:
        row_dict['sentiment'] = 1
    return row_dict
def _clean_tweet_text(text):
    return ' '.join(re.sub("(@[A-Za-z0-9]+)|([^A-Za-z \t])|(\w+:\/\\S+)", " ", text).spli+
def clean_train_tweet(row):
    Function to clean tweet text by removing links, special characters using regex .
    :param tweet_text: text of tweet recieved from database
    :return: cleaned text of tweet
    row['tweet_text'] = _clean_tweet_text(row['tweet_text'])
    return row
```

```
def clean_test_tweet(row):
  Function to clean tweet text by removing links, special characters using regex .
  :param tweet_text: text of tweet recieved from database
  :return: cleaned text of tweet
  row = row.asDict()
  row['tweet_text'] = _clean_tweet_text(row['tweet_text'])
  return row
train_df = spark.read.csv("hdfs:///user/maria_dev/sent_140.csv")
train_df = train_df.selectExpr("_c0 as sentiment", "_c5 as tweet_text").select('sentiment'
train_rdd = train_df.rdd
train_rdd = train_rdd.filter(train_filter_func).map(int_cast)
train_rdd = train_rdd.map(clean_train_tweet)
clean_train_df = spark.createDataFrame(train_rdd, schema="tweet_text: string, sentiment: ir
tweet df = spark.read.orc('hdfs:///user/maria dev/tweet usermention')
tweet_df = tweet_df.filter(tweet_df.tweet_usermention=='amazon').filter(tweet_df.tweet_hasl
#.filter("tweet_timestamp >= '2018-11-25'")
tweet_df = tweet_df.select('tweet_text')
tweet_rdd = tweet_df.rdd
clean_tweet_rdd = tweet_rdd.map(clean_test_tweet)
clean_tweet_df = spark.createDataFrame(clean_tweet_rdd, "tweet_text: string")
clean_tweet_df = clean_tweet_df.dropDuplicates()
tokenizer = Tokenizer(inputCol="tweet_text", outputCol="words")
remover = StopWordsRemover(inputCol="words", outputCol="filtered")
cv = CountVectorizer(inputCol="filtered", outputCol="cvfeatures", minDF=2.0)
hashtf = HashingTF(numFeatures=2 ** 16, inputCol="words", outputCol='tffeatures')
idf = IDF(inputCol='cvfeatures', outputCol="features",
      minDocFreq=5) # minDocFreq: remove sparse terms # it down-weights
# columns which appear frequently in a corpus.
idf2 = IDF(inputCol='tffeatures', outputCol="features", minDocFreq=5)
lr = LogisticRegression(labelCol="sentiment")
nb = NaiveBayes(labelCol="sentiment")
pipeline2 = Pipeline(stages=[tokenizer, remover, hashtf, idf2, lr]) # HashingTF + IDF + Lc
pipelineFit2 = pipeline2.fit(clean_train_df)
predictions2 = pipelineFit2.transform(clean_tweet_df)
pp = predictions2.select('prediction')
pp.write.csv("hdfs:///user/maria_dev/tweet_sentiment", mode="overwrite")
```

Previous livy session is expired, new livy session is created. Paragraphs that depend on this paragraph need to be re-executed!

Spark Application Id: application\_1543341205840\_0109
Spark WebUI: http://sandbox-hdp.hortonworks.com:8088/proxy/application\_1543341205840\_0109/
(http://sandbox-hdp.hortonworks.com:8088/proxy/application\_1543341205840\_0109/)

Took 2 min 48 sec. Last updated by anonymous at December 01 2018, 12:09:23 PM.

**READY**