

CS167 Twitter Analysis Report

Project Group D1

Minsoo Kim 862238343 - Task 1

Rajeev Thundiyil 862077977 - Task 2

Christian Boroff 862181900 - Task 3

Introduction

In our project we performed analysis on twitter data, and used Apache Spark (Beast) to do so. We used this big-data system because it is useful for working with dataframes and performing analysis on them, as it has sql and machine learning support, both of which we used throughout the project.

Within the project, we cleaned the data, assigned a topic to each tweet, and then built a machine learning model that predicted the topic of a tweet.

Task 1 - Minsoo Kim

I was tasked with the data preparation portion of the project. Specifically, I had to prepare the data for analysis later in the project. To give a brief summary, I first took the given data set and kept only the some of the attributes and created a new output file with it. Then I extracted the top 20 keywords from the newly created dataset.

In order to first load the data I did the following:

```
// Load the given input file using the json format.  
  
val tweetsDF = sparkSession.read.format("json")  
  .option("sep", "\t")  
  .option("inferSchema", "true")  
  .option("header", "true")  
  .load(inputFile)
```

Then, in order to keep only the relevant attributes, I ran the following:

```
// Keep only the following attributes {id, text, entities.hashtags.txt,  
user.description, retweet_count, reply_count, and quoted_status_id}  
  
val clean_tweets_df: DataFrame = tweetsDF.selectExpr("id", "text",  
"transform(entities.hashtags, x -> x.text) AS hashtags", "user.description AS  
user_description", "retweet_count", "reply_count", "quoted_status_id")
```

The relevant attributes mentioned above included the following,

- {id, text, entities.hashtags.txt, user.description, retweet_count, reply_count, and quoted_status_id}

Tweets_1k.json

Running the `printSchema()` function, I got the following schema

```
root
|-- id: long (nullable = true)
|-- text: string (nullable = true)
|-- hashtags: array (nullable = true)
|   |-- element: string (containsNull = true)
|-- user_description: string (nullable = true)
|-- retweet_count: long (nullable = true)
|-- reply_count: long (nullable = true)
|-- quoted_status_id: long (nullable = true)
```

Then running the `show()` on the `tweets_clean` dataset, I got the following

```
-----+-----+-----+-----+-----+-----+-----+-----+
|          id|          text|          hashtags|          user_description|retweet_count|reply_count| quoted_status_id|
|-----+-----+-----+-----+-----+-----+-----+-----+
|921633443934433280| saya tahu dia ter...| []| I'm not here anymore| 0| 0| null| | |
|921633444219596800| ふみとおおお 🍌🍌🍌🍌🍌🍌...| []| 野球ガールの服やってます🍌 甲子園...| 0| 0| 921338657436459008|
|null| 0| 0| ... اللهم إجعلني خير | []| اللهم إني أعوذ بك...| 0| 0| 921633444131680256|
|921633443976568838| find a boo &mp; ...| []| TxSU • 🍌 HTX...| 0| 0| null|
|921633443708096512| He referred to hi...| []| constantly grumpy| 0| 0| null|
|921633444400115713| @BarRefaeli my fi...| []| transduce| 0| 0| null|
|92163344443770241| Idk bout y'all bu...| []| Sierra High Schoo...| 0| 0| null|
|921633444223954944| 🍌🍌🍌🍌🍌🍌gwym mane...| []| David killed Goli...| 0| 0| 921633138190815232|
|921633444773400576| I'm at 北京餃子 in 仙台...| []| 五橋→元茶畑68→東京某学校/元I...| 0| 0| null|
|921633444953772032| I'm at G Tower in...| []| Zpace for share w...| 0| 0| null|
|921633445326942211| beach https://t.c...| []| they hurt u and t...| 0| 0| 899837211901820928|
|921633445360623617| @christophelicat ...| [Taquet, TeamSilv...| 10 Années de dans...| 0| 0| null|
|921633444693663745| #NaturalCure #Hea...| [NaturalCure, Hea...| #HealthTips News ...| 0| 0| null|
|921633445788258304| @ShirazHassan my ...| []| Cheif operating o...| 0| 0| null|
|921633445499097088| So true! https://...| []| Area Manager for ...| 0| 0| 921491726585487360|
|921633445729538048| @Avisirilak1 waw!...| []| null| 0| 0| 0| null|
|921633444915904512| เน้นพวงหรีดว่า...| []| รักสันโดษ เจ็บแต่...| 0| 0| null|
|921633445301964800| 5mentarios https://...| []| 🍌:michellerosello| 0| 0| 921501191401623552|
|921633446518079488| Mas okay talaga k...| []| hopeless romantic| 0| 0| null|
|921633444496576512| Paesaggiando http://...| []| tremeran le fogli...| 0| 0| null|
-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows
```

This is how I wrote the new file for the clean dataset

```
// Store the output in a new JSON file named tweets_clean
clean_tweets_df.write.mode(SaveMode.Overwrite).json("tweets_clean")
```

Then, I ran a top-k SQL query to select the top 20 most frequent hashtags. The below shows the SQL and how I collected the result in an array of keywords.

```
// run a top-k SQL query to select the top 20 most frequent hashtags as follows.
clean_tweets_df.createOrReplaceTempView("clean_tweets")
val frequent_hashtags = sparkSession.sql(
  """
  SELECT hashtag, count(*) as count
  FROM (
    SELECT explode(hashtags) as hashtag
    FROM clean_tweets
  ) t
  GROUP BY hashtag
```

```

ORDER BY count DESC

LIMIT 20

""")

val keywords: Array[String] = frequent_hashtags.select("hashtag").rdd.map(row =>
row.getString(0)).collect()

```

The result of the above code snippet was the following:

```

+-----+-----+
|      hashtag|count|
+-----+-----+
|    ALDUBxEBLoveis| 7| | |
|    FurkanPalalı| 5|
|      no309| 5|
|     La10n| 5|
|    sbhawks| 3|
|DoktorlarDenkliği...| 3|
|    Benimisteğim| 3|
|3| |احتاج-بالوقت-هذا| |
|      art| 2|
| CNITextravaganza2017| 2|
|      love| 2|
|     happy| 2|
|2| |السعودية| |
|    nowplaying| 2|
|    beautiful| 2|
|     türkiye| 1|
|    vegalta| 1|
|    KittyLive| 1|
|      鯛| 1|
|    tossademar| 1|
+-----+-----+

```

Tweets_10k.json

Following the same steps but for `Tweets_10k.json` gave me the following results

Running the `printSchema()` function on `tweets_clean` for 10k dataset, I got the following schema

```

root
 |-- id: long (nullable = true)
 |-- text: string (nullable = true)
 |-- hashtags: array (nullable = true)
 |    |-- element: string (containsNull = true)
 |-- user_description: string (nullable = true)
 |-- retweet_count: long (nullable = true)
 |-- reply_count: long (nullable = true)
 |-- quoted_status_id: long (nullable = true)

```

Then running the `show()` on the `tweets_clean` for the 10k dataset, I got the following

id	text	hashtags	user_description	retweet_count	reply_count	quoted_status_id
921633443934433280	saya tahu dia ter...	[]	I'm not here anymore	0	0	null
921633444219596800	ふみとおおおお👉👉👉👉👉👉...	[]	野球ガールの服やってます👉 甲子園...	0	0	921338657436459008
null	0	0	0	0	0	921633444131680256
921633443976568038	find a boo ∓ ...	[]	* TxSU * 🍷 HTX...	0	0	null
921633443708096512	He referred to hi...	[]	constantly grumpy	0	0	null
921633444400115713	@BarRefaeli my fi...	[]	transduce	0	0	null
921633444437770241	Idk bout y'all bu...	[]	Sierra High Schoo...	0	0	null
921633444223954944	👉👉👉👉👉👉gym mane...	[]	David killed Goli...	0	0	921633138190815232
921633444773400576	I'm at 北京餃子 in 仙台...	[]	五橋>元菜類68>東京某学校/元I...	0	0	null
921633444953772032	I'm at 6 Tower in...	[]	Zpace for share w...	0	0	null
921633445326942211	beach https://t.c...	[]	they hurt u and t...	0	0	899837211901820928
921633445360623617	@christophelicat ...	[Taquet, TeamSilv...]	10 Années de dans...	0	0	null
921633444693663745	#NaturalCure #Hea...	[NaturalCure, Hea...]	#HealthTips News ...	0	0	null
921633445788258304	@ShirazHassan my ...	[]	Cheif operating o...	0	0	null
921633445499097088	So true! https://...	[]	Area Manager for ...	0	0	921491726585487360
921633445729538048	@Auisirilaki mau...	[]	null	0	0	null
921633444915904512	ເປັນເປັນວິວັດ...	[]	วิวัฒนาการ ความเป็น...	0	0	null
921633445301964800	5mentarios https://...	[]	✳️:michellerosello	0	0	921501191401623552
921633446518079488	Mas okay talaga K...	[]	hopeless romantic	0	0	null
921633444496576512	Paesaggiando http://...	[]	tremoran le fogli...	0	0	null

only showing top 20 rows

Finally, the top 20 hashtags for the 10k dataset was the following

hashtag	count
ALDUBxEBLoveis	84
FurkanPalalı	51
no309	51
La10n	51
chien	30
job	28
Hiring	22
sbhawks	16
Top3Apps	16
perdu	15
trouv�	15
CareerArc	14
Job	12
trumpRussia	12
trndn	12
Jobs	11
ShowtimeLetsCelebr8	9
hiring	9
impeachtrump Pence	9
music	8

Task 2 - Rajeev Thundiyl

My task was to show the topic of the tweet by doing an array intersection of the most frequent hashtags from task 1 with the hashtags within the tweet itself. If it's a match, a new column is created with the topic shown, found through the intersection.

```
clean_tweets_df.createOrReplaceTempView( viewName = "tweets_clean")
//convert keywords to an array separated with , so it can be used for array intersect in a query
val topics: String = "" + keywords.mkString(", ") + ""

//dataframe
```

Taking the tweets_clean dataset from task 1, I use createOrReplaceTempView in order to create a local temporary view. This will be used later in the query.

Taking the keywords string array from task 1, I parse through it and put in commas between each space. This is necessary so it can be parsed and properly used for the array_intersect function within the query below.

```
val topics_df: DataFrame = sparkSession.sql(
  sqlText = s"""
    SELECT id, text, element_at(t1.tweet_topic, 1), user_description, retweet_count, reply_count, quoted_status_id
    FROM ( SELECT *, array_intersect(hashtags, array($topics)) AS tweet_topic FROM tweets_clean) AS t1 WHERE size(tweet_topic) > 0;
  """)

//write to json
topics_df.write.json( path = "tweets_topic.json")
topics_df.show()
val t4 = System.nanoTime()
```

The query picks up all of the data types in order, as given in the project specifications.

root

```
|-- id: long (nullable = true)
|-- text: string (nullable = true)
|-- topic: string (nullable = true)
|-- user_description: string (nullable = true)
|-- retweet_count: long (nullable = true)
|-- reply_count: long (nullable = true)
|-- quoted_status_id: long (nullable = true)
```

I create a table called t1, and within it a new column named tweet_topic is created, in which the hashtags on the line get intersected with the frequent hashtags from task 1. If it gets matched, then it gets included in the datatype, tweet_topic. I use element_at in order to get only the first frequent hashtag matched with the current tweet being checked.

Results shown below

Operations on file 'Tweets_1k.json' took 7.5779234 seconds

id	text	element_at(tweet_topic, 1)	user_description	retweet_count	reply_count	quoted_status_id
921633446644080641	#negramaroofficia...	love	Negramanteinside...	0	0	null
921633445845864497	#CNIextravaganza2...	CNIextravaganza2017	Hebat Produknya H...	0	0	null
921633449882128384	#DoktorlarDenklik...	DoktorlarDenkliği...	emin ben	0	0	null
921633451773648896	Na miss ko mag tw...	ALDUBxEBLoveis	Resilient. Object...	0	0	null
921633452289642497	#FurkanPalalı Değ...	FurkanPalalı	null	0	0	null
921633451714920448	#KittyLive penuh ...	KittyLive	Semangat	0	0	null
null	0	0	اجتاحت بالوقت هذا ...	0	0	921633453715738624
921633455766728704	Don't run from yo...	ALDUBxEBLoveis	BEHIND THE THICK ...	0	0	null
921633464276996097	Künefe Ocağı Rezi...	türkiye	Bym İsı Rezistans...	0	0	null
921633465921028096	ブルーマウンテンさんやん！\n追加...	sbhawks	野球場です！ホークスファンです！	0	0	null
921633468634632194	Pastinya dong, ka...	CNIextravaganza2017	BLOGGER BUZZER ...	0	0	921627555261702144
921633470413262848	#FurkanPalalı Değ...	FurkanPalalı	null	0	0	null
921633476964646912	Start by doing wh...	ALDUBxEBLoveis	BEHIND THE THICK ...	0	0	null
921633490482946048	Stay by Rihanna F...	nowPlaying	A live stream of ...	0	0	null
92163349356998848	#FurkanPalalı Değ...	FurkanPalalı	null	0	0	null
921633494530392064	Saturday let's go...	nowPlaying	Student of Life!	0	0	null
null	0	0	التخفيق مع رجال أ...	0	0	921633495511896065
921633502453518336	There are many th...	ALDUBxEBLoveis	BEHIND THE THICK ...	0	0	null
921633507558002688	#happy #Saturday ...	happy	Pastry chef, pers...	0	0	null
921633511752257536	#FurkanPalalı Değ...	FurkanPalalı	null	0	0	null

After running it for the tweets_10k.json file, there were **269 results** that were generated and shown to have the topic of the tweet shown.

Task3 - Christian Boroff

My task was to build a machine learning model that assigns a topic for each tweet using the classified tweets from the previous part. I used Apache Spark for this part, as it is very useful for machine learning, and building machine learning models. I made a pipeline for the model that includes a tokenizer, a HashingTF, a StringIndexer, and finally we used a LogisticRegression classifier to determine the topic of each tweet. My code for doing this can be seen below.

```
val tokenizer = new Tokenizer().setInputCol("text").setOutputCol("words")

val hashingTF = new HashingTF().setInputCol("words").setOutputCol("features")

val stringIndexer = new StringIndexer().setInputCol("element_at(tweet_topic, 1)").setOutputCol("label").setHandleInvalid("skip")

val logisticRegression = new LogisticRegression()

val pipeline = new Pipeline().setStages(Array(tokenizer, hashingTF, stringIndexer, logisticRegression))

val Array(trainingData, testData) = topics_df.randomSplit(Array(0.7, 0.3))

val logisticModel = pipeline.fit(trainingData)

val predictions = logisticModel.transform(testData)

predictions.select( col="id", cols="text", "element_at(tweet_topic, 1)", "user_description", "label", "prediction").show( numRows= 50)
```

As you can see above I used a 70/30 split for the trainingData and testData respectively. I then trained the model on the first 70% of the data and tested it on the remaining testData, which resulted in the following output.

id	text	element_at(tweet_topic, 1)	user_description	label	prediction
921633452289642497	#FurkanPalalı Değ...	FurkanPalalı	null	1.0	1.0
921633470413262848	#FurkanPalalı Değ...	FurkanPalalı	null	1.0	1.0
921633493569998848	#FurkanPalalı Değ...	FurkanPalalı	null	1.0	1.0
921633502453518336	There are many th...	ALDUBxEBLoveis	BEHIND THE THICK ...	0.0	0.0
921633525505310722	あー落ちない #sbhawks	sbhawks	野球垢です！ホークスファンです！！	6.0	6.0
921633536628551680	Everyone's deserv...	ShowtimeLetsCelebr8	love + trust	8.0	0.0
921633595747213312	If you're looking...	job	Staffing and Recr...	3.0	3.0
921633595801792512	Can you recommend...	job	Quintiles and IMS...	3.0	3.0
92163359960658688	Sometimes people ...	ALDUBxEBLoveis	BEHIND THE THICK ...	0.0	0.0
921633653129486336	@kuriring01 @Bern...	ALDUBxEBLoveis	Businesswoman, lo...	0.0	0.0
92163366656284672	#FurkanPalalı Değ...	FurkanPalalı	null	1.0	1.0
92163366656284672	#FurkanPalalı Değ...	FurkanPalalı	null	1.0	1.0
921633683789856768	Can you recommend...	job	Follow us for job...	3.0	3.0
921633684708454400	See our latest #R...	job	Join a team that ...	3.0	3.0
921633691469787136	#trumprussia #imp...	trumprussia	#NYC Registered N...	11.0	11.0
921633698423955456	🤔\nla ako masa...	ALDUBxEBLoveis	CMSRN	0.0	0.0
921633705260548096	Ganda nila ☺ #Sh...	ShowtimeLetsCelebr8	Hotelier and a ☺...	8.0	8.0

I then needed to compute the precision and recall of my result on the 10k dataset, which I did with the following code.

```
// Compute the number of true positives, false positives, and false negatives for each class
val tp = (0 to 10).map(c => predictions.filter(col( colName = "label") === c && col( colName = "prediction") === c).count()).sum
val fp = (0 to 10).map(c => predictions.filter(col( colName = "label") != c && col( colName = "prediction") === c).count()).sum
val fn = (0 to 10).map(c => predictions.filter(col( colName = "label") === c && col( colName = "prediction") != c).count()).sum

// Compute overall precision and recall
val overallPrecision = tp.toDouble / (tp + fp)
val overallRecall = tp.toDouble / (tp + fn)

println(s"Overall Precision: $overallPrecision, Overall Recall: $overallRecall")
```

This code found the total number of true positives, false positives, and false negatives for all of our classes, and then used them to compute the overall precision and recall. The precision and recall we obtained for the 10k dataset were the same, both coming out to **0.94382**. This means that 94.38% of the time, my model correctly identified true positive cases, while about 5.62% of the time, it incorrectly identified positive cases as negative.