# Big Data Project

# Covid-19 Tweets Sentimental Analysis

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## INTRODUCTION

In the view of the pandemic many have tweeted regarding the Covid 19 pandemic. It becomes essential to analyse the impact of the onset of the covid-19 pandemic. One best way is to assign a polarity to the tweet and determine which tweet is positive, negative and neutral.

## ABOUT OUR DATASET

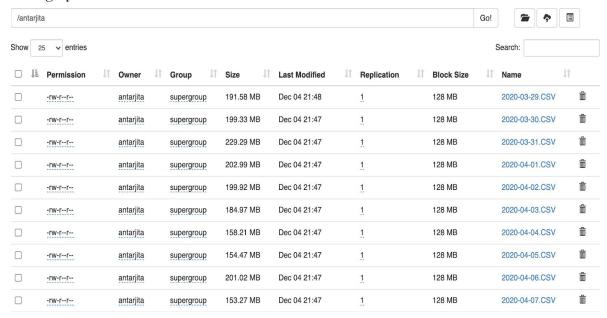
Our data set is <u>Coronavirus (covid19) Tweets - early April</u> from kaggle. It consists of 18 files each containing 22 columns. It contains various columns like friends\_count, followers\_count, country\_code etc and we can check the validity of the account through columns like verified etc.

## TOOLS USED:-

We have used Hdfs as the underlying storage. Hive is used for basic query retrieval and Pyspark is used to analyse the data. The below procedure is carried out in a pseudo distributed mode locally on our system (Mac/Ubuntu).

# APPROACH:-

All the files are first stored in hdfs which is connected to our hive metastore and pyspark through varn.



- 1. First we cleaned our data in pyspark and retained the columns necessary for further analysis.
  - Initial Set-up in pyspark shell:-

```
1 import sparknlp
   import os
 4 os.environ['PYSPARK_SUBMIT_ARGS'] = 'pyspark-shell'
   from pyspark.sql import SparkSession, functions as F
 2 from pyspark.sql.types import *
1 #from pyspark.sql import SparkSession
            SparkSession.builder.appName("Covid-SentimentAnalysis").getOrCreate()
 3 df = spark.read.option("header" , "true" ).option("inferSchema", "true" ).csv('/antarjita/*.CSV')
1 df.printSchema()
 -- status_id: string (nullable = true)
 -- user id: string (nullable = true)
 -- created_at: string (nullable = true)
 -- screen_name: string (nullable = true)
  -- text: string (nullable = true)
 -- source: string (nullable = true)
-- reply_to_status_id: string (nullable = true)
-- reply_to_user_id: string (nullable = true)
  -- reply_to_screen_name: string (nullable = true)
  -- is_quote: string (nullable = true)
-- is_retweet: string (nullable = true)
-- favourites_count: string (nullable = true)
  -- retweet_count: string (nullable = true)
  -- country_code: string (nullable = true)
  -- place_full_name: string (nullable = true)
  -- place_type: string (nullable = true)
  -- followers_count: string (nullable = true)
  -- friends_count: string (nullable = true)
  -- account_lang: string (nullable = true)
 -- account_created_at: string (nullable = true)
-- verified: string (nullable = true)
 -- lang: string (nullable = true)
 1 type(df)
 2 df.count()
```

• Dropping unnecessary columns :-

• Removing Nan values and splitting created\_at into valid time:-

```
1 #find NAN values
    2 from pyspark.sql.functions import isnan, when, count, col
    4
               df.select([count(when(isnan(c), c)).alias(c) for c in df.columns]).show()
|created_at|text|favourites_count|retweet_count|
            _____+
                                                                                                                             0 |
                                                                                                                                                                                        0
 split_col = F.split(df['created_at'],'T')
 2 df = df.withColumn('Date', split_col.getItem(0))
3 df = df.withColumn('Time', split_col.getItem(1))
  4 df=df.drop('created_at')
  5 df.show()
  6 #df.select("created_at") =df.select("created_at").apply(lambda i:(int(i.split("T")[1].split(":"))))
                                                  text | favourites count | retweet count |
                                                                                                                                                        0 | 2020-03-29 | 00:04:062 | 1 | 2020-03-29 | 00:41:052 | 1 | 2020-03-29 | 00:41:092 | 0 | 2020-03-29 | 00:41:092 | 0 | 2020-03-29 | 00:42:402 | 0 | 2020-03-29 | 01:04:162 | 0 | 2020-03-29 | 01:04:162 | 0 | 2020-03-29 | 01:24:352 | 0 | 2020-03-29 | 01:24:352 | 0 | 2020-03-29 | 01:23:482 | 0 | 2020-03-29 | 01:33:482 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 | 01:36:532 | 0 | 2020-03-29 |
@ayush4bharat Ver...
 I have made a sma...
                                                                                                  48715
#COVID19: Madurai...
 #Coronavirus | Po...
                                                                                                          339
Because killers a...
                                                                                                   3798
                                                                                                  65863
     ""Coronavirus Tw...
For every article...
                                                                                                  32080
                                                                                                   32080
We need to unders...
Having said all t...
                                                                                                   32080
Is it the onset o...
                                                                                                   22105
```

Removing special characters from the tweets:-

2. Initial insights from the data is obtained through basic querying in hive and pyspark.

select retweer count from one where (lang='en' and verified='TRUE');

This query shows the number of retweets that all the verified accounts that tweeted in english got.

This query was used to check how popular the tweets about coronavirus by famous people were.

```
hive> select retweet_count from one where (lang='en' and verified='TRUE');
Query ID = nayana_20201206101419_759ebf77-658a-49d8-8dc7-0bede033e6da
Total jobs = 1
Launching Job 1 out of 1
Number of reduce tasks is set to 0 since there's no reduce operator
Starting Job = job_1607229408127_0001, Tracking URL = http://nayana-VirtualBox:8088/proxy/application_1607229408127_0001/
Kill Command = /home/nayana/hadoop-3.2.1/bin/mapred job -kill job_1607229408127_0001
Hadoop job information for Stage-1 number of mappers: 6; number of reducers: 0
2020-12-06 10:18:34,050 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:19:35,929 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:20:38,123 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:22:38,123 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:22:40,436 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:22:40,436 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:23:40,551 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:23:40,551 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:25:44,067 Stage-1 map = 0%, reduce = 0%
2020-12-06 10:26:44,971 Stage-1 map = 0%, reduce = 0%, Cumulative CPU 145.56 sec
2020-12-06 10:26:44,971 Stage-1 map = 0%, reduce = 0%, Cumulative CPU 386.65 sec
2020-12-06 10:28:57,649 Stage-1 map = 0%, reduce = 0%, Cumulative CPU 386.65 sec
2020-12-06 10:28:57,649 Stage-1 map = 3%, reduce = 0%, Cumulative CPU 384.68 sec
2020-12-06 10:28:57,649 Stage-1 map = 3%, reduce = 0%, Cumulative CPU 496.48 sec
2020-12-06 10:28:57,649 Stage-1 map = 3%, reduce = 0%, Cumulative CPU 490.57 sec
2020-12-06 10:30:27,207 Stage-1 map = 9%, reduce = 0%, Cumulative CPU 495.03 sec
2020-12-06 10:30:27,207 Stage-1 map = 17%, reduce = 0%, Cumulative CPU 495.03 sec
```

```
26
17
10
4
0
0
0
11
12
4
0
11
14
3
2
0
8
9
6
6
2
1
7
0
1
1
7
0
1
1
3
Time taken: 1356.35 seconds, Fetched: 122970 row(s)
```

select sum(retweer count) from one where(verified='TRUE')

This query finds the sum of the number of retweets of all tweets about coronavirus By verified accounts.

This shows the importance of the topic and the engagement of the verified accounts.

```
Notes select sunfretwest_coint) from one where(verifiede: TRUE):
Ouery 10 = nayen_2020120014758_1c78808-af7f-4045-bbc-9f1b36f2098e

Total jabs = 1
Launching Job | out of 1
Number of reduce tasks determined at compile time: 1
In order to change the average load for a reducer (in bytes):
    set hive.exec.reducers.bytes.per.reducer=cnumbers
    norder to thirt the naxinum number of reducers:
    set tive.exec.reducers.bytes.per.reducer=cnumbers
    set tive.exec.reducers.naxcumbers
    set intreducers.naxcumbers
    set intreducers.naxcumbers
    set intreducers.naxcumbers
    set intreducers.bytes.per.reducers.naxcumbers
    set intreducers.naxcumbers
    set intreducers.nax
```

select sum(retweet count) from one where(lang='en')

This query gives the number of retweet counts of all tweets on the coronavirus topic tweeted in English.

```
2020-12-06 12:25:39,974 Stage-1 map = 19%, reduce = 0%, Cumulative CPU 428.05 sec
2020-12-06 12:25:54,874 Stage-1 map = 25%, reduce = 0%, Cumulative CPU 474.39 sec
2020-12-06 12:26:13,691 Stage-1 map = 31%, reduce = 0%, Cumulative CPU 474.39 sec
2020-12-06 12:26:27,257 Stage-1 map = 37%, reduce = 0%, Cumulative CPU 493.1 sec
2020-12-06 12:26:38,122 Stage-1 map = 42%, reduce = 0%, Cumulative CPU 505.89 sec
2020-12-06 12:26:41,504 Stage-1 map = 51%, reduce = 0%, Cumulative CPU 511.65 sec
2020-12-06 12:26:54,353 Stage-1 map = 54%, reduce = 0%, Cumulative CPU 511.65 sec
2020-12-06 12:26:58,416 Stage-1 map = 54%, reduce = 0%, Cumulative CPU 531.21 sec
2020-12-06 12:27:35,104 Stage-1 map = 72%, reduce = 0%, Cumulative CPU 531.21 sec
2020-12-06 12:27:43,611 Stage-1 map = 76%, reduce = 0%, Cumulative CPU 582.32 sec
2020-12-06 12:27:47,527 Stage-1 map = 91%, reduce = 0%, Cumulative CPU 587.76 sec
2020-12-06 12:27:49,521 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 590.85 sec
2020-12-06 12:28:52,508 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 604.55 sec
2020-12-06 12:29:06,390 Stage-1 map = 100%, reduce = 67%, Cumulative CPU 604.55 sec
2020-12-06 12:29:06,390 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 623.87 sec
MapReduce Total cumulative CPU time: 10 minutes 23 seconds 870 msec
Ended Job = job_1607229408127_0003
MapReduce Jobs Launched:
Stage-Stage-1: Map: 6 Reduce: 1 Cumulative CPU: 623.87 sec HDFS Read: 1777492136 HDFS Write: 107 SUCCESS
Total MapReduce CPU Time Spent: 10 minutes 23 seconds 870 msec

OK
3670548
Time taken: 864.684 seconds, Fetched: 1 row(s)
```

select avg(retweet count) from one where(lang='en);

This query gives the average number of retweets per tweet written in english.

```
select avg(retweet_count) from one where(lang='en')
Query ID = nayana_20201206123832_10c26c3d-6543-4130-b62d-0259ea37c8ac
Total jobs = 1
Launching Job 1 out of 1
Number of reduce tasks determined at compile time: 1
In order to change the average load for a reducer (in bytes):
   set hive.exec.reducers.bytes.per.reducer=<number:
In order to limit the maximum number of reducers:
   set hive.exec.reducers.max=<number>
In order to set a constant number of reducers: set mapreduce.job.reduces=<number>
Starting Job = job_1607229408127_0004, Tracking URL = http://nayana-VirtualBox:8088/proxy/application_1607229408127_0004/Kill Command = /home/nayana/hadoop-3.2.1/bin/mapred job -kill job_1607229408127_0004
Hadoop job information for Stage-1: number of mappers: 6; number of reducers: 1
2020-12-06 12:40:25,607 Stage-1 map = 0%, reduce = 0%
2020-12-06 12:41:27,096 Stage-1 map = 0%, reduce = 0%
2020-12-06 12:42:30,553 Stage-1 map = 0%, reduce = 0%
2020-12-06 12:43:33,596 Stage-1 map = 0%, reduce = 0%
2020-12-06 12:44:35,140 Stage-1 map = 0%, reduce = 0%
2020-12-06 12:45:36,224 Stage-1 map = 0%, reduce = 0%
2020-12-06 12:46:37,041 Stage-1 map = 0%, reduce = 0%, Cumulative CPU 150.13 sec
2020-12-06 12:46:37,041 Stage-1 map = 0%, reduce = 0%, Cumulative CPU 150:13 sec 2020-12-06 12:47:38,124 Stage-1 map = 0%, reduce = 0%, Cumulative CPU 285.9 sec 2020-12-06 12:48:16,214 Stage-1 map = 2%, reduce = 0%, Cumulative CPU 338.2 sec 2020-12-06 12:49:02,399 Stage-1 map = 6%, reduce = 0%, Cumulative CPU 406.28 sec 2020-12-06 12:49:08,715 Stage-1 map = 12%, reduce = 0%, Cumulative CPU 411.22 sec 2020-12-06 12:49:19,640 Stage-1 map = 15%, reduce = 0%, Cumulative CPU 430.51 sec 2020-12-06 12:49:48,307 Stage-1 map = 21%, reduce = 0%, Cumulative CPU 465.18 sec 2020-12-06 12:50:02,163 Stage-1 map = 27%, reduce = 0%, Cumulative CPU 489.41 sec 2020-12-06 12:49:42.64 Stage-1 map = 27%, reduce = 0%, Cumulative CPU 489.43 sec
2020-12-06 12:50:24,261 Stage-1 map = 33%,
2020-12-06 12:50:28,897 Stage-1 map = 45%,
2020-12-06 12:50:33,066 Stage-1 map = 54%,
                                                                         reduce = 0%, Cumulative CPU 519.23 sec
reduce = 0%, Cumulative CPU 525.87 sec
reduce = 0%, Cumulative CPU 530.62 sec
2020-12-06 12:50:33,066 Stage-1 map = 54%, reduce = 0%, Cumulative CPU 530.62 sec
2020-12-06 12:50:43,008 Stage-1 map = 58%, reduce = 0%, Cumulative CPU 543.12 sec
2020-12-06 12:51:40,776 Stage-1 map = 72%,
                                                                                    reduce = 0%, Cumulative CPU 602.4 sec
2020-12-06 12:51:48,270 Stage-1 map = 78%,
                                                                                     reduce = 0%, Cumulative CPU 608.55 sec
2020-12-06 12:51:50,836 Stage-1 map = 82%,
                                                                                     reduce = 0%, Cumulative CPU 610.63 sec
2020-12-06 12:51:53,241 Stage-1 map = 87%,
                                                                                    reduce = 0%, Cumulative CPU 613.83 sec
2020-12-06 12:52:00,193 Stage-1 map = 91%,
                                                                                    reduce = 0%, Cumulative CPU 613.83 sec
2020-12-06 12:52:30,133 Stage-1 map = 94%,
                                                                                    reduce = 0%, Cumulative CPU 630.53 sec
2020-12-06 12:52:34,397 Stage-1 map = 100%, reduce = 0%, Cumulative CPU 632.08 sec
2020-12-06 12:53:17,826 Stage-1 map = 100%, reduce = 67%, Cumulative CPU 645.52 sec
2020-12-06 12:53:47,411 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 679.02 sec
MapReduce Total cumulative CPU time: 11 minutes 19 seconds 20 msec
Ended Job = job_1607229408127_0004
MapReduce Jobs Launched:
Stage-Stage-1: Map: 6 Reduce: 1 Cumulative CPU: 679.02 sec
                                                                                                                           HDES Read: 1777496383 HDES Write: 118 SUCCESS
Total MapReduce CPU Time Spent: 11 minutes 19 seconds 20 msec
2.3999027105698296
Time taken: 923.418 seconds, Fetched: 1 row(s)
```

select avg(retweet\_count) from one where(lang='es')

This gives the average number of retweets of tweets written in español

```
hive> select avg(retweet_count) from one where(lang='es');
Query ID = nayana_20201206163457_9eecbbce-249d-4d02-a4f5-bc459538e415
Total jobs = 1
Launching Job 1 out of 1
Number of reduce tasks determined at compile time:
In order to change the average load for a reducer (in bytes):
set hive.exec.reducers.bytes.per.reducer=<number>
In order to limit the maximum number of reducers:
  set hive.exec.reducers.max=<number:
In order to set a constant number of reducers
  set mapreduce.job.reduces=<number>
Starting Job = job_1607229408127_0005, Tracking URL = http://nayana-VirtualBox:8088/proxy/application_1607229408127_0005/
Kill Command = /home/nayana/hadoop-3.2.1/bin/mapred job -kill job_1607229408127_0005
Hadoop job information for Stage-1: number of mappers: 6; number of reducers: 1
2020-12-06 16:36:35,060 Stage-1 map = 0%, reduce = 0%
2020-12-06 16:37:36,211 Stage-1 map = 0%,
                                                    reduce = 0%
2020-12-06 16:38:36,350 Stage-1 map = 0%,
                                                    reduce = 0%
2020-12-06 16:39:37,794 Stage-1 map = 0%,
2020-12-06 16:40:38,629 Stage-1 map = 0%,
                                                    reduce =
                                                    reduce = 0%
2020-12-06 16:41:40,654 Stage-1 map = 0%,
                                                    reduce = 0%,
                                                                    Cumulative CPU 23.47 sec
2020-12-06 16:42:41,275 Stage-1 Map = 0%,
2020-12-06 16:42:41,275 Stage-1 map = 0%,
2020-12-06 16:44:17,305 Stage-1 map = 6%,
2020-12-06 16:44:27,952 Stage-1 map = 11%
                                                    reduce = 0%,
                                                                    Cumulative CPU 174.04 sec
                                                                    Cumulative CPU 309.67 sec
                                                    reduce = 0%,
                                                    reduce = 0%, Cumulative CPU 361.09 sec
                                                      reduce = 0%, Cumulative CPU 379.35 sec
                                            = 11%.
                                                      reduce = 0%,
2020-12-06 16:44:37,923 Stage-1 map = 15%,
                                                                     Cumulative CPU 392.34 sec
2020-12-06 16:44:47,852 Stage-1 map
                                           = 19%,
                                                      reduce = 0%,
                                                                     Cumulative CPU 406.08 sec
2020-12-06 16:44:49,860 Stage-1 map = 25%,
                                                      reduce = 0%,
                                                                     Cumulative CPU 410.66 sec
2020-12-06 16:45:06,416 Stage-1 map
                                                      reduce = 0%, Cumulative CPU 433.6 sec
                                            = 31%,
2020-12-06 16:45:32,678 Stage-1 map
                                                      reduce = 0%, Cumulative CPU 471.63 sec
                                            = 44%,
                                                     reduce = 0%, Cumulative CPU 481.16 sec
reduce = 0%, Cumulative CPU 501.29 sec
2020-12-06 16:45:41,236 Stage-1 map
2020-12-06 16:45:57,338 Stage-1 map = 54%,
2020-12-06 16:45:59,348 Stage-1 map = 62%,
                                                                     Cumulative CPU 505.72 sec
                                                      reduce = 0%,
2020-12-06 16:46:16,266 Stage-1 map
                                                                     Cumulative CPU 527.56 sec
                                                      reduce
                                                              = 0%.
                                              72%
                                                             reduce = 0%,
2020-12-06 16:44:49,860 Stage-1
2020-12-06 16:45:06,416 Stage-1 map = 31%,
                                                            reduce = 0%, Cumulative CPU 433.6 sec
2020-12-06 16:45:32,678 Stage-1 map = 40%,
2020-12-06 16:45:41,236 Stage-1 map = 44%,
                                                             reduce = 0%, Cumulative CPU 471.63 sec
                                                             reduce = 0%, Cumulative CPU 481.16 sec
                                                    54%,
2020-12-06 16:45:57,338 Stage-1 map
                                                             reduce = 0%, Cumulative CPU 501.29 sec
2020-12-06 16:45:59,348 Stage-1 map
                                                    62%,
                                                             reduce = 0%, Cumulative CPU 505.72 sec
                                                             reduce = 0%, Cumulative CPU 527.56 sec
2020-12-06 16:46:16,266 Stage-1 map =
                                                    72%,
2020-12-06 16:46:24,993 Stage-1 map = 2020-12-06 16:46:34,633 Stage-1 map =
                                                    82%,
                                                            reduce = 0%, Cumulative CPU 536.44 sec
                                                             reduce = 0%, Cumulative CPU 544.25 sec
                                                    91%,
2020-12-06 16:46:38,140 Stage-1 map =
2020-12-06 16:47:38,193 Stage-1 map =
                                                    100%,
                                                              reduce = 0%, Cumulative CPU 546.85 sec
                                                    100%,
                                                              reduce = 0%, Cumulative CPU 547.53 sec
                                                    100%,
2020-12-06 16:47:39,371 Stage-1 map
                                                              reduce = 67%, Cumulative CPU 560.6 sec
2020-12-06 16:48:03,054 Stage-1 map = 100%, reduce = 100%, Cumulative CPU 591.83 sec
MapReduce Total cumulative CPU time: 9 minutes 51 seconds 830 msec
Ended Job = job_1607229408127_0005
MapReduce Jobs Launched:
Stage-Stage-1: Map: 6 Reduce: 1 Cumulative CPU: 591.83 sec
                                                                                        HDFS Read: 1777496383 HDFS Write: 118 SUCCESS
Total MapReduce CPU Time Spent: 9 minutes 51 seconds 830 msec
Time taken: 795.778 seconds, Fetched: 1 row(s)
```

Surprisingly, according to the last two queries, the average number of retweets in español (3.8269) is greater than the average number of retweets in english (2.399). This could probably be because English is more popular than español. Most people use English to tweet (i.e, even people who don't have many followers, and hence lesser retweets). This might have brought down the average retweet\_count.

#### Insights from the dataset

1) Top five retweets replied back

```
retweet=df['retweet_count','text'].sort(col('retweet_count').desc()).rdd.flatMap(lambda x: x).collect()

#retweet.show(5,False)
for i in range(5):
    retweet[2*i+1]=retweet[2*i+1].lower()
    print(i,']', retweet[2*i+1],'\n')
```

- 0 ] dear ji matching steps with your relentless efforts to save our people from coronavirus we have made this pra yer to appreciate the spirit of coronawarriors amp make the cry of humanity reach out to the almighty sambhaall
- 1 ] ten 10 new ventilators installed at rims on 07 04 2020 ready to overcome covid 19 thanks to dr shanta sing h director rims and other officials for your tireless efforts
- 2 ] big rajasthan govt is going all out to take over 84 private hospitals a circular has been sent to all hospital s to give over the control within two hours of when the order will be released they have been informed today covid 19india coronavirus
- 3 ] 12 staffers of jaslok hospital including nurses technicians helpers and a cook turn positive for covid 19 swab s of 300 staffers and patients taken hospital declared a containment zone covid 19
- 4 1 i love my india stavhomestavsafe

#### Top 5 FavouriteTweets

2)top five most favourited tweets

```
favourite_tweets = df['favourites_count','text'].sort(col('favourites_count').desc()).rdd.flatMap(lambda x: x).col
for i in range(5):
    favourite_tweets[2*i+1]=favourite_tweets[2*i+1].lower()
    print(i,']', favourite_tweets[2*i+1],'\n')
```

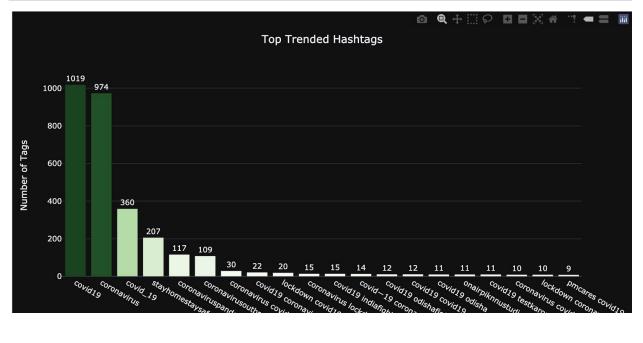
- 0 ] stop the stigma covid 19 a south mumbai coronavirus survivor speaks to me about the harassment he faced i pr ay this doesn t repeat let's show some empathy love and compassion he battled and survived this and is back to nor mal
- 1 ] a south mumbai resident a coronavirus survivor speaks to me harassed named shamed affluent educated neighb ourhood misbehaves are we covidiots to stigmatise this society needs to show empathy get real accept and coopera te interview will be on air soon
- 2 ] news reports coming in taken into icu condition worsens coronavirus
- 3 ] ji is suffering from multiple ailments including hypertension chronic kidney disease and hypertrophy of prosta te in the wake of the apex court s directive on covid 19 an ailing and ageing chief must be given parole releas elaluprasadyadav
- 4 ] a tiger tests positive for covid19

#### Viral Hashtags

```
import re
def find_hash(text):
    line=re.findall(r'(?<=#)\w+',text)
    return " ".join(line)
data['hash']=data['text'].apply(lambda x:find_hash(x))

hastags=list(data[(data['hash'].notnull())&(data['hash']!="")]['hash'])
hastags = [each_string.lower() for each_string in hastags]
hash_df=dict(Counter(hastags))
top_hash_df=pd.DataFrame(list(hash_df.items()),columns = ['word','count']).sort_values('count',ascending=False)[:20]
top_hash_df.head(10)</pre>
```

ord		count
119	1	1019
rus	5	974
_19	7	360
afe	1479	207
coronaviruspandemic		
eak	2	109
119	47	30
rus	260	22
119	57	20
wn	251	15



# 3. The sentimental analysis of the data is done in Pyspark.

• Libraries imported for NLP

```
from textblob import TextBlob
from pyspark.sql.functions import udf
from pyspark.sql.types import DoubleType
import seaborn as sns
```

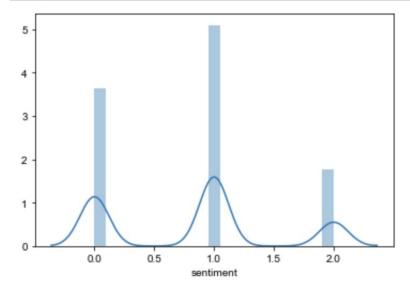
• Checking the value of positivity, negativity or neutrality of the sentiments

```
def apply_blob(sentence):
        temp = TextBlob(sentence).sentiment[0]
        if temp == 0.0:
    return 0.0 # Neutral
        elif temp >= 0.0:
           return 1.0 # Positive
            return 2.0 # Negative
 1 sentiment = udf(apply blob, DoubleType())
   result=df.withColumn("sentiment", sentiment(df['text']))
 2 result.show()
                 text | favourites count | retweet count |
                                                                       Date
                                                                                 Time|sentiment|
     Covid19 Soc...
                                   10374
                                                                 2020-03-29 | 00:04:06Z |
  Very unfortunat...
                                    236
                                                      οi
                                                                2020-03-29 | 00:17:57Z
                                                                                             2.0
I have made a sma...
COVID19 Madurai...
                                   48715
                                                                2020-03-29 | 00:35:142 |
                                                                                             0.0
                                   339
                                                                2020-03-29 | 00:41:092 |
 Coronavirus Po...
                                     339
                                                                2020-03-29 | 00:42:40Z
Because killers a...
                                   3798
                                                                2020-03-29 00:53:50z
                                                                                             1.0
                                                                2020-03-29 | 01:04:16Z |
  Coronavirus Tw...
                                   65863
                                                                                             0.0
                                                                2020-03-29 | 01:24:35Z
For every article...
                                   32080
                                                                                             2.0
We need to unders...
                                                                2020-03-29 | 01:27:172
Having said all t...
                                   32080
                                                                2020-03-29 | 01:32:482
Is it the onset o...
                                  22105
                                                                2020-03-29 01:36:53Z
                                                                                             0.0
                                                                2020-03-29 01:40:05Z
                                   4189
```

Where o.o=neutral 1.o=positive 2.o=negative

#### • Tweets Sentiment distribution

```
df_res_pandas = result.toPandas()
sns.distplot(df_res_pandas['sentiment'])
sns.set(rc={'figure.figsize':(11.7,8.27)})
```



• To determine the polarity of the tweets:-

```
from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator
import matplotlib.pyplot as plt
stopwords = set(STOPWORDS)
```

```
1 stopwords
'to',
 'too',
 'under',
 'until',
 'up',
 'very',
 'was',
 "wasn't",
 'we',
 "we'd",
 "we'll",
 "we're",
 "we've",
 'were',
 "weren't",
 'what',
 "what's",
 'when',
 "when's",
'where'.
```

```
df_res_pandas['sentiment'] = ' '
   df_res_pandas['polarity'] = None
for i,tweets in enumerate(df_res_pandas.text):
       blob = TextBlob(tweets)
       df_res_pandas['polarity'][i] = blob.sentiment.polarity
6
       if blob.sentiment.polarity > 0 :
7
           df_res_pandas['sentiment'][i] = 'positive'
8
       elif blob.sentiment.polarity < 0 :</pre>
          df_res_pandas['sentiment'][i] = 'negative'
9
10
       else :
11
           df_res_pandas['sentiment'][i] = 'neutral'
12 df_res_pandas.head(10)
```

	text	favourites_count	retweet_count	Date	Time	sentiment	polarity
0	Covid19 SocialDistancing the Indian way	10374	0	2020-03-29	00:04:06Z	neutral	0
1	Very unfortunate to see this unplanned ill	236	0	2020-03-29	00:17:57Z	negative	-0.575
2	I have made a small contribution towards my co	48715	1	2020-03-29	00:35:14Z	negative	-0.25
3	COVID19 Madurai Kavalan app uses GPS trackin	339	0	2020-03-29	00:41:09Z	neutral	0
4	Coronavirus Police deploy Drones to monito	339	0	2020-03-29	00:42:40Z	neutral	0
5	Because killers are having a team of activists	3798	0	2020-03-29	00:53:50Z	positive	0.09375
6	Coronavirus Tweets from Indian authorities	65863	0	2020-03-29	01:04:16Z	neutral	0
7	For every article and every opinion attacking	32080	0	2020-03-29	01:24:35Z	negative	-0.25
8	We need to understand most of India has a join	32080	0	2020-03-29	01:27:17Z	positive	0.142424
9	Having said all this migrant workers should ha	32080	1	2020-03-29	01:32:48Z	positive	0.275

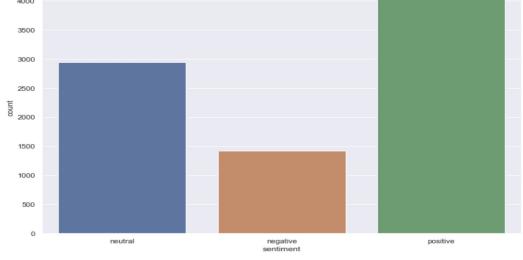
This piece of code determines how much positive, negative or neutral the tweets are. 0.09375 polarity means it is slightly positive, more towards the neutral side.

# • Number of positive, negative and neutral tweets

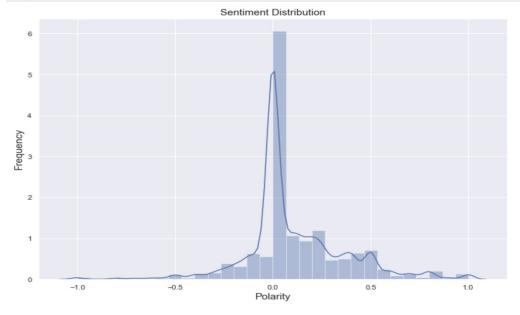
```
print(df_res_pandas.sentiment.value_counts())
sns.countplot(x='sentiment', data = df_res_pandas);

positive 4129
neutral 2945
negative 1424
Name: sentiment, dtype: int64

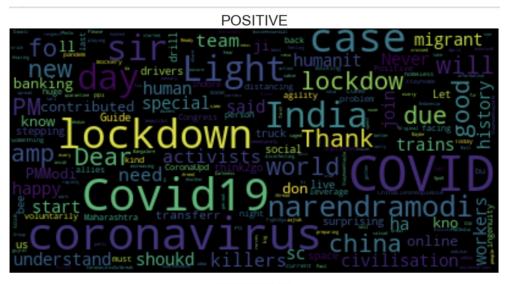
4000
3500
3000
```



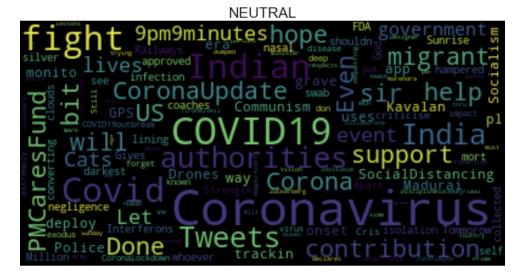
```
sns.distplot(df_res_pandas['polarity'], bins=30)
plt.title('Sentiment Distribution', size = 15)
plt.xlabel('Polarity', size = 15)
plt.ylabel('Frequency', size = 15)
plt.show();
```



## • Word Clouds:-







# Conclusion

Most of the tweets in the early april are toward and positive or neutral side during the early stages of the pandemic.