Project 4: Stack Overflow - askubuntu dataset

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The project involves extracting data from several smaller datasets and combining them together to do analysis.

Dataset:

The dataset contains logs for askubuntu stackexchange logs (https://askubuntu.com) in XML format.

- The total size of the dataset is 22 GB.
- It is stored in the GCS bucket gs://stackoverflow-dataset-677
- The dataset includes multiple xml files corresponding to different attributes of the dataset.

■ Badges.xml
Comments.xml
Comments_smal
PostHistory.xml
PostLinks.xml
Posts.xml
■ Tags.xml
■ Users.xml
■ Votes.xml

- The following are the relevant features for each XML file.
 - Users: Reputation, CreationDate, DisplayName, WebsiteUrl, Location, Views, UpVotes, DownVotes, AccountId
 - Posts: Id, PostTypeId, AcceptedAnswerId, CreationDate, Score, Body, OwnerUserId, Title, Tags, AnswerCount, CommentCount
 - o Comments: Rowld, Postld, Score, Text, CreationDate, Userld, ContentLicense
 - o Tags: Rowld, TagName, Count, ExcerptPostld, WikiPostld
 - o Badges: Rowld, Userld, Name, Date, Class, TagBased
 - Votes: Rowld, Postld, VoteTypeld, CreationDate

Tasks to be completed:

- 1. a) Extracting user id and username and storing it to GCS.
 - b) Extracting the comments and doing inner join with Spark SQL and display to user.

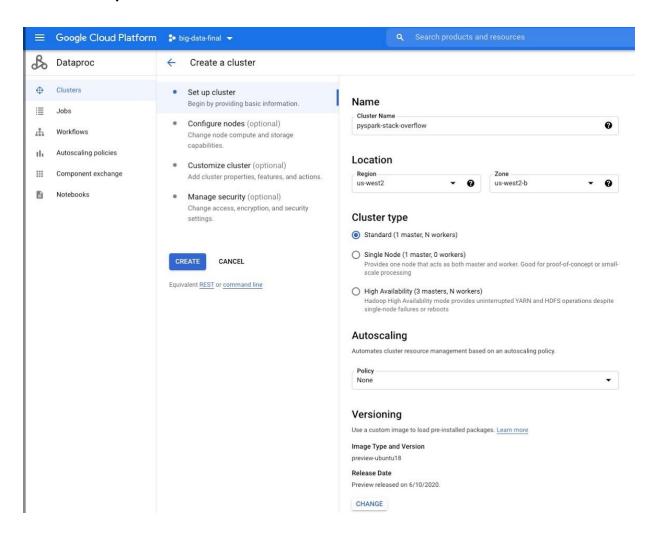
- 2. Trending Users with max comments: Finding users who posted maximum comments and visualizing it.
- 3. Trending Topics in Comments: Extracting the trending topics in comments related to ubuntu and visualizing it.
- 4. Automating task 1 with Apache Airflow i.e. in a single pipeline

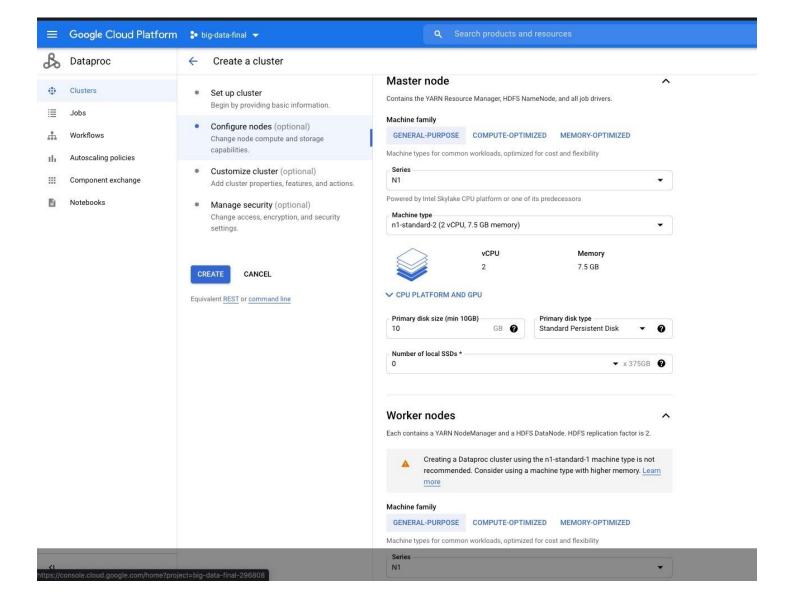
Server Setup:

We used Google Cloud DataProc to setup a pyspark cluster of 1 Master Nodes (2 vCPUs) and 6 Worker Nodes (1vCPU) each with spark 3.2, python 3 and ubuntu 18.04 operating system.

For first 3 tasks, the server setup was done manually.

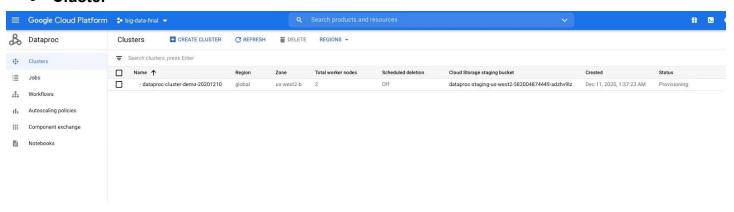
Manual Setup:



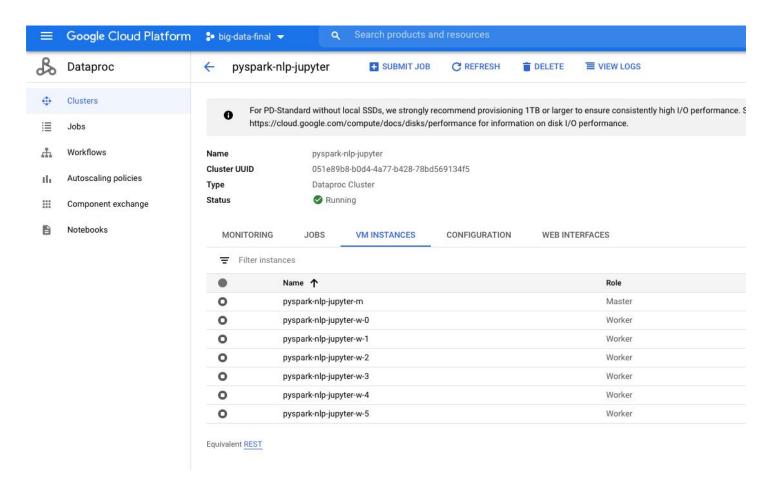


Result:

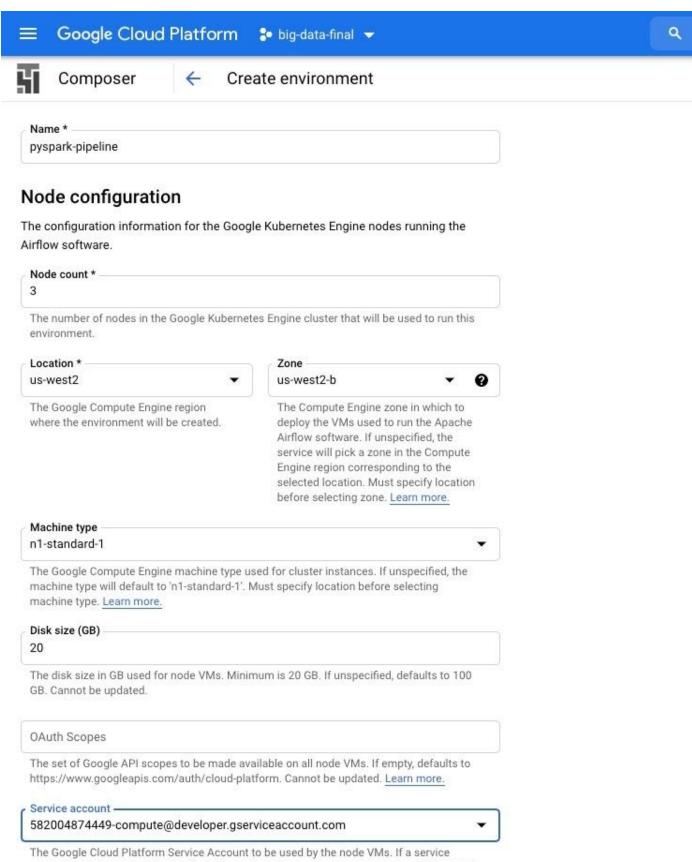
Cluster



Cluster Nodes with 1 master and 6 worker nodes

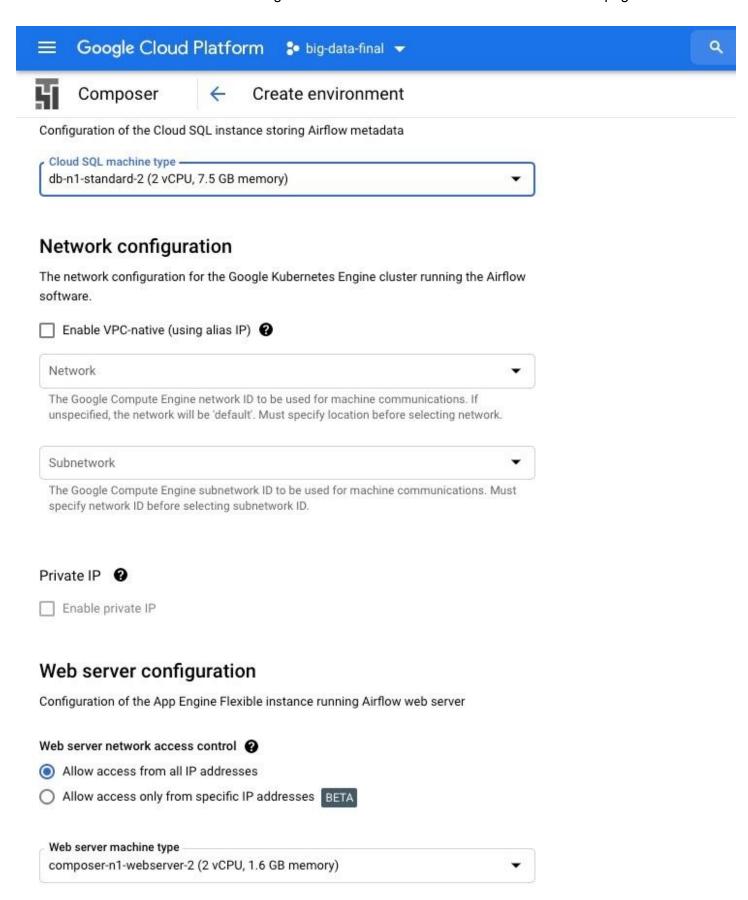


For the last task i.e. automating task 1 with airflow, an airflow cluster was set up using Google Cloud Composer that created and destroyed the dataproc (pyspark) cluster using an airflow scheduler.



The Google Cloud Platform Service Account to be used by the node VMs. If a service account is not specified, the "default" Compute Engine service account is used. Cannot be updated.

• Next, we set up a kubernetes cluster using n1-standard v2 machines for master and 3 n1-standard v2 machines with a web server using n1-standard v2 machine to host Airflow GUI web page.



Data Processing

To create a dataframe from the xml files, we are using python's xml.etree.ElementTree module.

1a) Extracting user id and username and storing to the bucket: jupyter notebook link

Raw Data:

• After cleaning, filtering and converting the data to dataframe, we get:

```
user df.show(truncate=False)
+---+
|id |username
1
    Community
2
    Geoff Dalgas
3
    Jarrod Dixon
 4
    txwikinger
 5
    Nathan Osman
 6
    Emmett
7
    Helix
 8
    mechanical meat
 9
    Andrew
10 DLH
 11 | hannes.koller
12 | Michael Terry
 13 | Keith Maurino
 14 Jweede
16 | Jeremy L
17 | tutuca
18 excid3
 20 | Paranoia Puppy
21 GeoD
22 Alan Featherston
only showing top 20 rows
```

• On completion of creating the dataframe, we store the data to GCS

```
In [82]: user_df1.count()
Out[82]: 855054
In [84]: user_df.repartition(1).write.csv("gs://stackoverflow-dataset-677/users_out", sep=',')
```

1b) Extracting the comments and joining username with Spark SQL: jupyter notebook link

Raw Data:

```
text_file.take(3)
['<?xml version="1.0" encoding="utf-8"?>',
    '<comments>',
    ' <row Id="13" PostId="23" Score="0" Text="Using /opt helps me keep track of the applications I\'ve installed myself." Creat ionDate="2010-07-28T19:36:59.773" UserId="10" ContentLicense="CC BY-SA 2.5" />']
```

• After cleaning, filtering and converting the data to dataframe, we get:

postId	score	text	creationDate	userId
	+	+		
23	0	Using /opt helps	2010-07-28T19:36:	10
18	0	but popping in a	2010-07-28T19:38:	10
27	0	That will revert	2010-07-28T19:39:	50
31	0	I think you meant	2010-07-28T19:41:	12
18	0	@DLH apparently n	2010-07-28T19:41:	63
12	2	"ssh -X <server></server>	2010-07-28T19:46:	96
12	0	@Suppressingfire:	2010-07-28T19:48:	10
50	0	Can you please re	2010-07-28T19:48:	56
27	0	It probably shoul	2010-07-28T19:49:	5
58	0	Do you mean the c	2010-07-28T19:50:	5
47	0	Have you checked	2010-07-28T19:50:	4
47	1	Might be related	2010-07-28T19:51:	104
58	0	Do you use Gnome	2010-07-28T19:51:	4
60	0	This causes data	2010-07-28T19:52:	66
18	0	no the live CD do	2010-07-28T19:53:	4
52	0	Does this let the	2010-07-28T19:55:	35
56	2	LDAP and nfs are	2010-07-28T19:56:	4
10	0	Can I use it on a	2010-07-28T19:56:	27
70	1	That's a good tip	2010-07-28T19:56:	45
70	0	That is probably	2010-07-28T19:58:	86

• Next, we read the username and user id from the csv file created in the previous step.

```
user df.show()
| id|
           username
  1
           Community
  2
       Geoff Dalgas
  3
        Jarrod Dixon
         txwikinger
   5
        Nathan Osman
   6
              Emmett
  7 |
               Helix
   8 | mechanical_meat
  9
              Andrew
  10
                 DLH
  11
       hannes.koller
       Michael Terry
 12
       Keith Maurino
 13
 14
               Jweede
 16
             Jeremy L
 17
               tutuca
 18
               excid3
        ParanoiaPuppy
  20
 21
                GeoD
| 22|Alan Featherston|
only showing top 20 rows
```

• To allow join queries, we made the columns to be integer/long type.

Converting both datasets to TempView for allowing SQL queries.

```
comments_df.createOrReplaceTempView("comments") user_df.createOrReplaceTempView("users")
```

• Inner Join Query, introducing username to comments dataframe.

comments_users_sql_df = spark.sql("SELECT * FROM users u JOIN comments c ON u.id = c.UserId")
comments_users_sql_df.show()

	creationDate	text			username		id
	2010-10-13T21:37:	I can confirm thi		0	Brummermann		964
964	2012-04-28T06:17:	They took it in d	0	118087	Brummermann	Hendrik	964
964	2015-08-03T13:26:	I have the same i	0	638027	Brummermann	Hendrik	964
1677	2011-12-03T21:56:	@fossfreedom i do	0	84949	eslambasha		1677
1697	2010-12-08T22:36:	@Marco, I know, I	0	16683	Frxstrem		1697
1697	2010-12-09T19:05:	This seems to be	0	16784	Frxstrem		1697
1697	2010-12-10T22:26:	I only want to di	1	16886	Frxstrem		1697
1697	2010-12-10T22:28:	This is not an ac	1	16892	Frxstrem		1697
1697	2010-12-11T19:22:	Have you tried bu	0	16988	Frxstrem		1697
1697	2010-12-14T23:14:	@Stefano fixed it	0	17471	Frxstrem		1697
1697	2010-12-17T13:50:	My guess is that	0	17892	Frxstrem		1697
1697	2010-12-18T17:53:	-1 It's too uncle	0	18014	Frxstrem		1697
1697	2010-12-22T17:48:	You did replace `	0	18273	Frxstrem		1697
1697	2011-10-15T22:18:	Firstly, I have a	0	67121	Frxstrem		1697
1697	2012-03-01T00:30:	You should use `t	0	108944	Frxstrem		1697
1697	2014-04-23T07:29:	Daily builds can	2	453415	Frxstrem		1697
1697	2015-08-26T16:36:	@user2662639 Simp	0	223442	Frxstrem	Ü	1697
1697	2015-08-26T16:37:	@user2662639 (I t	0	223442	Frxstrem		1697
1697	2016-03-25T12:21:	@Fiksdal I don't	2	17650	Frxstrem		1697
1697	2017-04-01T13:36:	@DavidFoerster Th	0	899129	Frxstrem		1697

2. Trending users with max comments:

• After extracting the comments data and converting to dataframe, we got:

userId	creationDate	text	score	postId
10	2010-07-28T19:36:	Using /opt helps	0	23
10	2010-07-28T19:38:	but popping in a	0	18
50	2010-07-28T19:39:	That will revert	0	27
12	2010-07-28T19:41:	I think you meant	0	31
63	2010-07-28T19:41:	@DLH apparently n	0	18
96	2010-07-28T19:46:	"ssh -X <server></server>	2	12
10	2010-07-28T19:48:	@Suppressingfire:	0	12
56	2010-07-28T19:48:	Can you please re	0	50
5	2010-07-28T19:49:	It probably shoul	0	27
5	2010-07-28T19:50:	Do you mean the c	0	58
4	2010-07-28T19:50:	Have you checked	0	47
104	2010-07-28T19:51:	Might be related	1	47
4	2010-07-28T19:51:	Do you use Gnome	0	58
66	2010-07-28T19:52:	This causes data	0	60
4	2010-07-28T19:53:	no the live CD do	0	18
35	2010-07-28T19:55:	Does this let the	0	52
4	2010-07-28T19:56:	LDAP and nfs are	2	56
27	2010-07-28T19:56:	Can I use it on a	0	10
45	2010-07-28T19:56:	That's a good tip	1	70
86	. [1] [1] [1] [1] [1] [1] [1] [1] [1] [1]	That is probably	0	70

• Next, we find the trending users by using groupBy aggregate function and sorting it with respect to count in descending order.

```
trending_users = comments_df.groupBy("userId") \
    .agg(count("text").alias("count")) \
    .where(col("userId") > 0) \
    .orderBy(col('count').desc())
```

Result

```
trending users.show()
+----+
|userId|count|
+----+
167850 | 14677 |
  4272 | 12192
158442 | 12091 |
 15811 | 10505
175814 8835
307523 7861
469152 7567
178692 | 7523
 35795 7515
126395 7398
295286 6757
19421 6679
344926 6268
 19626 6164
22949 6136
10616 6124
225694 | 5736
 94914 5692
459561 5346
 72216 | 5314 |
+----+
```

Next, we create a temp view of the trending users and merge it with usernames obtained using 1a.

Joining the usernames with the trending user.

```
trending_usernames = spark.sql("SELECT count,username FROM trending_users tu JOIN users u ON u.id = tu.userId order by count d
esc limit 10")
trending_usernames.show()
count
              username
|14677|
                Pilot6
              heynnema
12192
12091
                  muru
10505
              Rinzwind
 8835
        David Foerster
 7861 WinEunuuchs2Unix
 7567
               guiverc
 7523
            steeldriver
 7515
               Panther
 7398
               oldfred
```

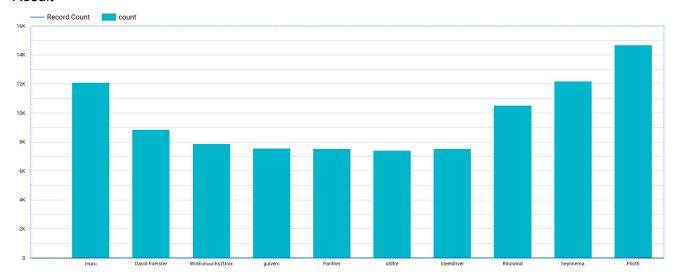
Storing it to GCS, will be imported to Google BigQuery for visualization with Google Data Studio.

```
trending_usernames.repartition(1).write.csv("gs://stackoverflow-dataset-677/trending_usernames", sep=',')
```

After importing the dataset to bigquery, and exporting the data to Google Data Studio for visualization.



Result



3: Trending Topics in Comments:

Notebook Link:

Obtaining the data from GCS and cleaning it.

After extracting the attributes (comment text) from xml,we get

```
import xml.etree.ElementTree as ET

def parse_xml(rdd):
    """

    Read the xml string from rdd, parse and extract the elements,
    then return a list of list.
    """

    root = ET.fromstring(rdd)
    rec = []

    if "Text" in root.attrib:
        rec.append(root.attrib['Text'])
    else:
        rec.append("N/A")

    return rec

records_rdd = cleanedRDD.map(lambda x : parse_xml(x))

records_rdd.take(3)

[["Using /opt helps me keep track of the applications I've installed myself."],
    ["but popping in a live cd I already have isn't going to work huh?"],
```

Next, we have to remove the stop words from this text. So, we downloaded the stop words list online.
 Importing the stop words from file

['That will revert the splash screen as well as the login? I almost did that, but grew hesitant.']]

```
stop_words_text = sc.textFile("file:///home/aarora7/P4-ayush-adarsh/03 stopwords.txt")
stop_words_text.take(3)
['a', 'about', 'above']
```

Next, we converted it into dataframe. Converting the stop words dataframe to list

•

```
stop_words_list = list(stop_words_df.select('_1').toPandas()['_1'])
  stop words list
: ['a',
    'about',
   'above',
   'across',
   'after',
   'again',
   'against',
   'all',
   'almost',
   'alone',
   'along',
   'already',
   'also',
   'although',
```

Next, removing stopwords from main dataframe and converting into (stop_word,1) tuple

```
def remove_stop_word(x):
    list = []
    s_split = x[0].split(" ")
    for i in s_split:
        i = i.lower()
        if i not in stop_words_list:
            if not i.startswith('.') and not i.endswith('.') and not i.startswith('?') and not i.
```

 Next, finding the total count of words by using reduceByKey and converting it to Dataframe for storing to GCS.

```
topic_count = cleaned_text.reduceByKey(lambda y,x: x+y)

topic_count.take(3)

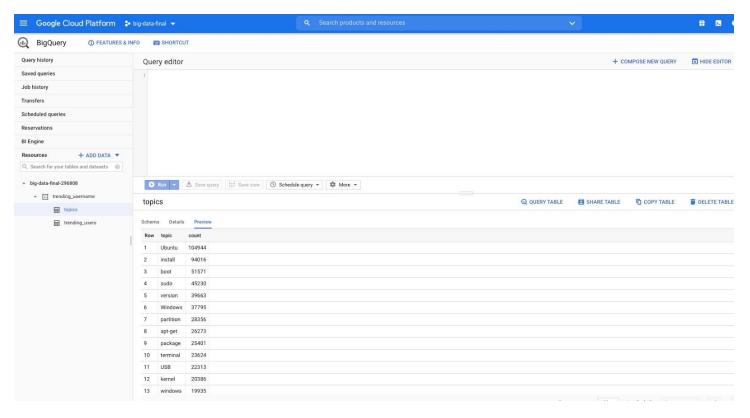
[('manpage,', 44), ('python', 6727), ("doesn't", 45536)]

topic_count_col = ["topic", "count"]
topic_count_data = topic_count.toDF(topic_count_col)
```

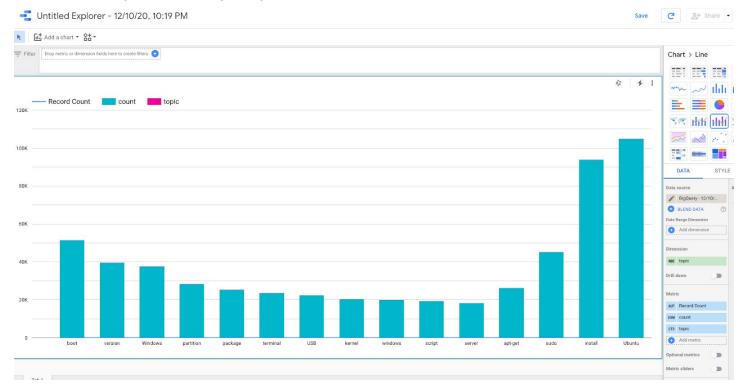
Storing to GCS and sorting the DF in descending order w.r.t count

```
final = topic_count_data.orderBy(col('count').desc())
final.repartition(1).write.csv("hdfs:///topic_count", sep=',')
```

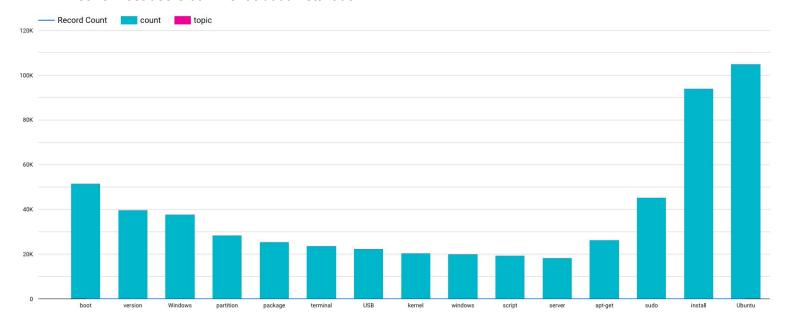
• Next, we imported the data to BigQuery from GSC to export to Google Data Studio.



Visualizing the data using Google Data Studio.



Result: Trending Keywords: Ubuntu is the most trending keyword with install as the second. This
means most users comment about installation.



- 4. Automating Part 1 with Google Cloud Composer (Apache Airflow).
 - After setting up the server (shown in the beginning), next we create python scripts for part 1 and 2 and upload them to GCS. The GCS path will be referenced during the final orchestration.

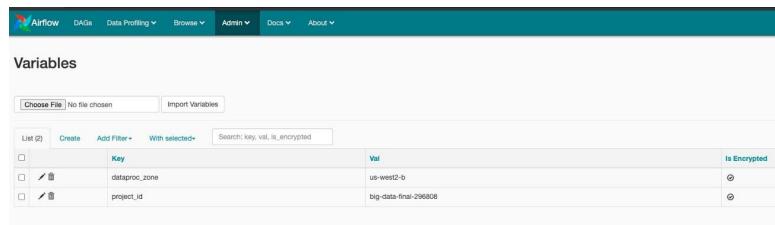
Part 1: users.py

```
from pyspark.sql.functions import *
import time
import pyspark.sql.functions as F
from pyspark.sql.types import *
import xml.etree.ElementTree as ET
from pyspark.sql import SparkSession
spark = SparkSession \
.builder \
.appName("My PySpark code") \
.getOrCreate()
text file = sc.textFile("gs://stackoverflow-dataset-677/Users.xml")
filteredRDD = text_file.filter(lambda x: x.startswith(" <row "))</pre>
cleanedRDD = filteredRDD.map(lambda x: x.lstrip(" "))
def parse_xml(rdd):
   root = ET.fromstring(rdd)
   rec = []
```

```
id = root.attrib['Id']
   if id == "-1":
       id = "1"
   rec.append(id)
   rec.append(root.attrib['DisplayName'])
   return rec
records rdd = cleanedRDD.map(lambda x : parse xml(x))
user data = ["id", "username"]
user df = records rdd.toDF(user data)
user_df.repartition(1).write.csv("gs://stackoverflow-dataset-677/users_out1", sep=',')
Part 2: comments.py
from pyspark.sql.functions import *
import time
import pyspark.sql.functions as F
from pyspark.sql.types import *
import xml.etree.ElementTree as ET
from pyspark.sql import SparkSession
spark = SparkSession \
.builder \
.appName("My PySpark code") \
.getOrCreate()
text file = sc.textFile("gs://stackoverflow-dataset-677/Comments.xml")
filteredRDD = text file.filter(lambda x: x.startswith(" <row "))</pre>
cleanedRDD = filteredRDD.map(lambda x: x.lstrip(" "))
def parse xml(rdd):
   Read the xml string from rdd, parse and extract the elements,
   then return a list of list.
   .....
   root = ET.fromstring(rdd)
   rec = []
   if "PostId" in root.attrib:
       rec.append(int(root.attrib['PostId']))
   else:
       rec.append(0)
   if "Score" in root.attrib:
       rec.append(int(root.attrib['Score']))
   else:
       rec.append(0)
```

```
if "Text" in root.attrib:
      rec.append(root.attrib['Text'])
   else:
      rec.append("N/A")
   if "CreationDate" in root.attrib:
       rec.append(root.attrib['CreationDate'])
   else:
       rec.append("N/A")
   if "UserId" in root.attrib:
       rec.append(int(root.attrib['UserId']))
   else:
      rec.append(0)
   return rec
records rdd = cleanedRDD.map(lambda x : parse xml(x))
comments data = ["postId", "score", "text", "creationDate", "userId"]
comments df = records rdd.toDF(comments data)
comments df.createOrReplaceTempView("comments")
comments sql df = spark.sql("SELECT * FROM comments")
users_data = sc.textFile("gs://stackoverflow-dataset-677/users_out1/*.csv")
def create user(rdd):
  rdd split = rdd.split(",")
  return [int(rdd split[0]),rdd split[1]]
users_rdd = users_data.map(lambda x: create_user(x))
user data = ["id", "username"]
user df = users rdd.toDF(user data)
user df.createOrReplaceTempView("users")
comments users sql df = spark.sql("SELECT * FROM users u JOIN comments c ON u.id = c.UserId")
comments_users_sql_df.repartition(1).write.csv("gs://stackoverflow-dataset-677/users_out1",
sep=',')
```

 Next, we created two variables for the final orchestration. The GCP project id and the zone where the clusters will be created.



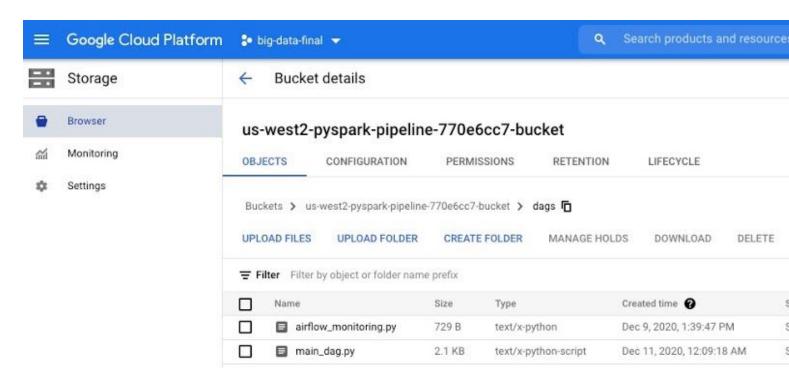
- **Final Orchestration:** To connect both the scripts together and for automatic creation and destruction of Spark Clusters, we created a DAG file.
 - This contains the path of the users.py and comments.py file and the airflow variables as well.

main_dag.py

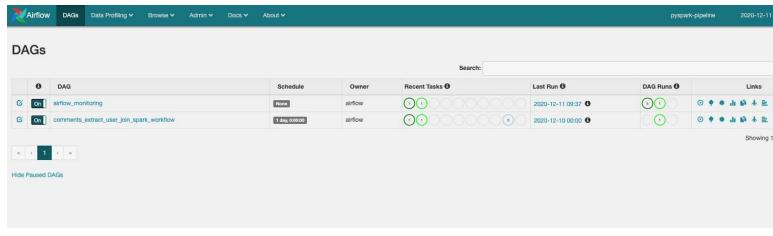
```
# Ref:
https://medium.com/analytics-vidhya/a-gentle-introduction-to-data-workflows-with-apach
e-airflow-and-apache-spark-6c2cd9aee573
from datetime import timedelta, datetime
from airflow import models
from airflow.operators.bash operator import BashOperator
from airflow.contrib.operators import dataproc operator
from airflow.utils import trigger rule
# STEP 2:Define a start date
#In this case yesterday
yesterday = datetime(2020, 12, 10)
SPARK_CODE = ('gs://stackoverflow-dataset-677/01_user.py')
SPARK CODE2 = ('gs://stackoverflow-dataset-677/02 user comments join.py')
dataproc job name = 'extract users job dataproc'
dataproc job name2 = 'extract comments join users dataproc'
# STEP 3: Set default arguments for the DAG
default dag args = {
'start date': yesterday,
'depends on past': False,
'email_on_failure': False,
'email on retry': False,
'retries': 1,
'retry delay': timedelta(minutes=5),
'project_id': models.Variable.get('project_id')
}
# STEP 4: Define DAG
# set the DAG name, add a DAG description, define the schedule interval and pass the
default arguments defined before
with models.DAG(
'comments_extract_user_join_spark_workflow',
description='DAG for extracting comments and merging with user name',
schedule interval=timedelta(days=1),
default_args=default_dag_args) as dag:
# STEP 5: Set Operators
# BashOperator
```

```
# A simple print date
  print date = BashOperator(
  task_id='print_date',
  bash command='date'
  )
# dataproc_operator
# Create small dataproc cluster
  create dataproc = dataproc operator.DataprocClusterCreateOperator(
  task_id='create_dataproc',
  cluster_name='dataproc-cluster-demo-{{ ds_nodash }}',
  num_workers=2,
   zone=models.Variable.get('dataproc zone'),
  master_machine_type='n1-standard-1',
  worker machine type='n1-standard-1')
  run spark = dataproc operator.DataProcPySparkOperator(
  task id='run spark',
  main=SPARK CODE,
  cluster name='dataproc-cluster-demo-{{ ds nodash }}',
  job_name=dataproc_job_name
  run spark2 = dataproc operator.DataProcPySparkOperator(
  task_id='run_spark2',
  main=SPARK CODE2,
  cluster name='dataproc-cluster-demo-{{ ds nodash }}',
   job name=dataproc job name2
  )
   # dataproc_operator
   # Delete Cloud Dataproc cluster.
  delete dataproc = dataproc operator.DataprocClusterDeleteOperator(
   task_id='delete_dataproc',
  cluster_name='dataproc-cluster-demo-{{ ds_nodash }}',
   trigger rule=trigger rule.TriggerRule.ALL DONE)
# STEP 6: Set DAGs dependencies
# Each task should run after have finished the task before.
print_date >> create_dataproc >> run_spark >> run_spark2 >> delete_dataproc
```

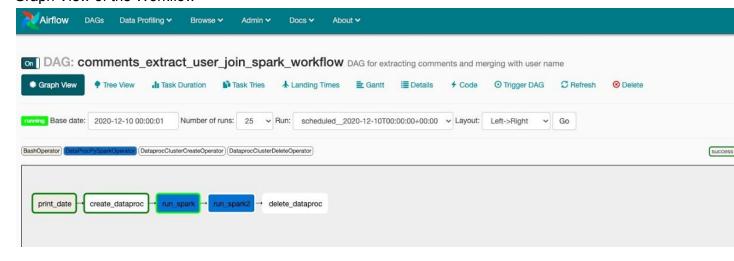
• To run the main dag file, we upload it to DAGs folder and the script ran automatically in 2 minutes.



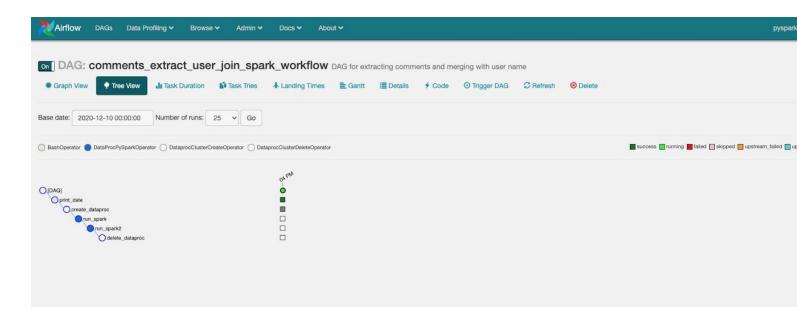
Main dag file with with workflow name "Comments_extract_user_join_spark_workflow."



Graph View of the Workflow



Tree View



• Finally, the output file was saved in GCS.