Name: Manikandan Mohan

#### Case Study 1

Amazon Books Catalog Search, Data Cleansing, Analytics, Pipeline & Reporting using Apache Beam, Python pySpark on a Google Data Proc Cluster [ubuntu 20.04 VMs].

#### **PROBLEM STATEMENT:**

- 1. Find the average user rating for Fiction books, Non-fiction books and for all books together and write the result in separate files. python (analytics), PySpark
- 2. Sort the books in alphabetical order and the last 5 books and print the output Pyspark 3. Kaggle Dataset https://www.kaggle.com/sootersaalu/amazon-top-50-bestselling-books-2009-2019

Notes - This is a dataset on Amazon's Top 50 bestselling books from 2009 to 2019., it contains 550+ books, and has been categorized into fiction and non-fiction using Goodreads.

[Note, one can add/multiple the data set to about 5000+] books to utilize power of GCP Cluster, Analytics, Spark Job, BigQuery, BigTable capabilities.

#### **Implementation Steps -**

· Pre-requisite – One should download this csv and do data cleaning (replace the commas in the Name column with semicolons). [Upload this csv in the GCP bucket] – Pyspark, Hive [10% credits]

One has to build a DataProc Cluster [at least 2 nodes] to run Python pySpark batch jobs [Kaggle dataset, from GCP bucket] on Google Cloud. – DataProc [15% credits]

- · Advisable to connect to the nodes using CLI and provision the VMs with required configurations using [15% credits] o GCP Cloud SDK o gsutil tool
- · Create/Configure a pipeline [10% credits] using Command Line

Pipeline = Reading input + Transforming Data + Writing Output

Output results should be stored as raw data in Hive[BigQuery] before stored finally into HBase[BigTable] for further processing & reporting to Analytics Engines. – Hbase, Hive [15% Credits]

· Run the pipeline – Apache Beam [10% credit]

· Print the results to console and store for further analytics – Hbase [10% credits]

Extract of the pipeline execution and output is here - (good to use HBase) - [10% credits]

Engineering Best Practices to be followed [30% credits]

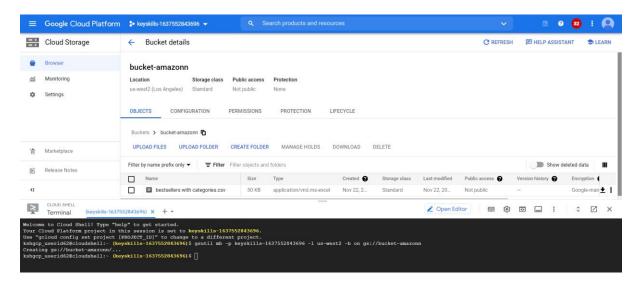
- 1. Logging has to be implemented in all the Python Components
- 2. Exception handling framework need to be implemented
- 3. Unit testing of each Python component/service using Pytest/UnitTest/Mock Objects
- 4. Unit Testing of Spark Cluster, Data Pipeline [is desirable]
- 5. Usage of 4 to 5 Big Data Design Patterns using Python [is desirable]

#### **Answer**

Project Name: keyskills-1637220444729

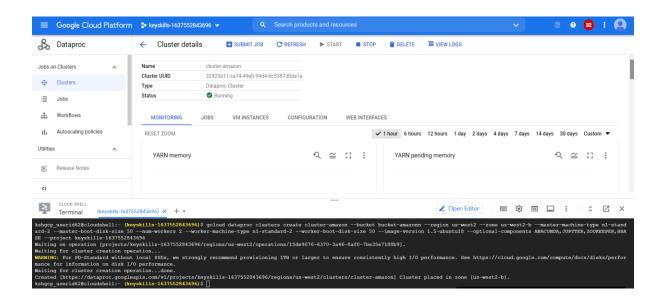
## Creating a bucket storage named "bucket-amazonn" using CLI and uploading the kaggle dataset.

gsutil mb -p keyskills-1637552843696 -l us-west2 -b on gs://bucket-amazonn



#### **Creating a cluster using Command Line Interface (CLI):**

gcloud dataproc clusters create cluster-amazon --bucket bucket-amazonn --region us-west2 --zone us-west2-b --master-machine-type n1-standard-2 --master-boot-disk-size 50 --num-workers 2 --worker-machine-type n1-standard-2 --worker-boot-disk-size 50 --image-version 1.5-ubuntu18 --optional-components ANACONDA,JUPYTER,ZOOKEEPER,HBASE --project keyskills-1637552843696



## Creating a python (spark) file with the required queries and uploading it into the bucket.

Pyspark code:

amazon\_book\_analysis.py

import pyspark as sc

from pyspark import SparkConf, SparkContext

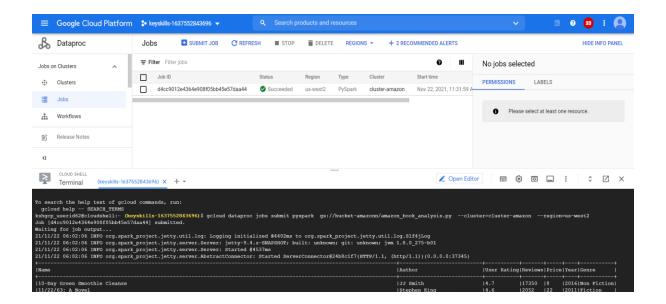
from pyspark.sql import SparkSession

```
#Reading Data using SparkSession
spark = SparkSession.builder.master("local[*]").getOrCreate()
data =
spark.read.option ("header", True).options (inferSchema="True', delimiter=',').csv ("gsark.read.option") and the second option ("header", True) and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") and the second option ("header") are second option ("header") are second option ("header") and the second option ("header") are second option ("header") 
://bucket-amazonn/bestsellers with categories.csv")
data.show(truncate=False)
data.printSchema()
#Replacing ',' with ';' in the name column
data = data.withColumn('Name', translate('Name', ',', ';'))
data.show(truncate=False)
#Find averages of fiction, non-fiction and overall
data.filter(data['Genre']=='Fiction').agg({'User Rating': 'avg'}).show()
data.filter(data['Genre']=='Non Fiction').agg({'User Rating': 'avg'}).show()
data.agg({'User Rating': 'avg'}).show()
avg_fict = data.filter(data['Genre']=='Fiction').agg({'User Rating': 'avg'})
avg_non_fict = data.filter(data['Genre']=='Non Fiction').agg({'User Rating': 'avg'})
avg_all = data.agg({'User Rating' : 'avg'})
avg_fict.write.csv('gs://bucket-amazonn/Avg_of_Fiction')
avg_non_fict.write.csv('gs://bucket-amazonn/Avg_of_Non_Fiction')
avg_all.write.csv('gs://bucket-amazonn/Avg_of_Allbooks')
data.orderBy(data.Name.desc()).limit(5).show(truncate=False)
```

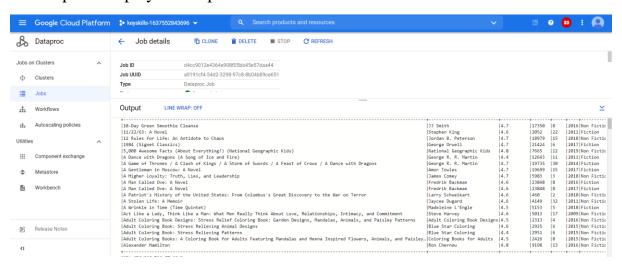
data.write.csv('gs://bucket-amazonn/clean\_books\_amazon.csv')

### Executing the python file by submitting a job in dataproc using CLI.

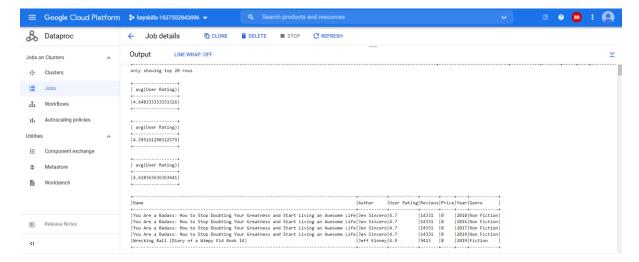
gcloud dataproc jobs submit pyspark gs://bucket-amazonn/amazon\_book\_analysis.py --cluster=cluster-amazon --region=us-west2



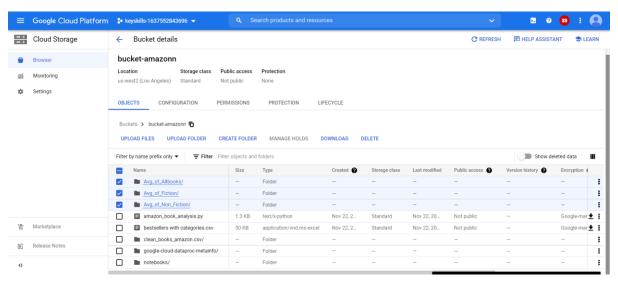
#### Job output: Displays the top 20 rows.



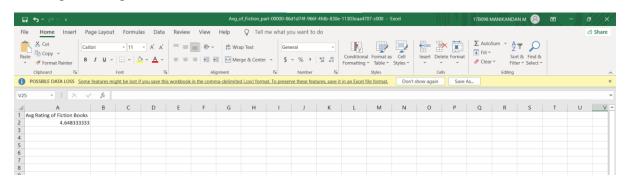
Displays the average rating of books and the last 5 rows of the dataset.



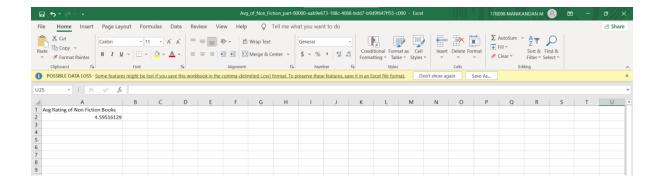
## Output CSV files written to the bucket.



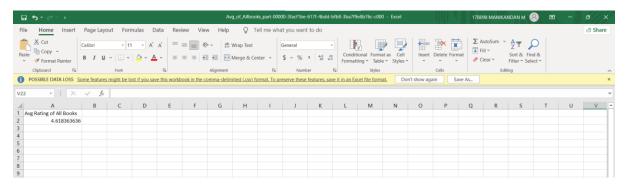
#### Average Rating of Fiction Books



Average Rating of Non Fiction Books



### Average Rating of all the Books.

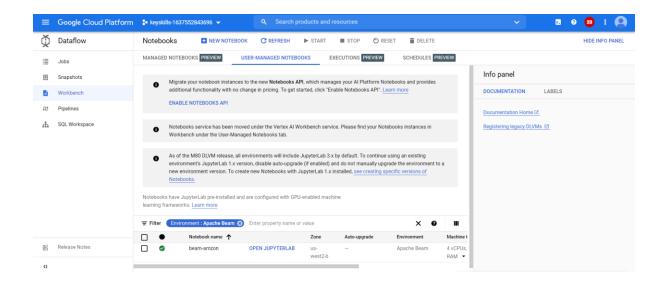


#### Last 5 rows while the dataset in alphabetical order



# Creating a dataflow pipeline job using apache beam (python) for executing all the queries and to write the table in the bigquery.

Creating a juputer lab notebook with apache beam on dataflow workbench.



#### Pipeline code:

Dataflowpipeline.ipynb [Using Jupyter Lab on Google Cloud]

import apache\_beam as beam

from apache\_beam.runners.interactive.interactive\_runner import InteractiveRunner from apache\_beam.options.pipeline\_options import PipelineOptions import google.auth

from apache\_beam.options import pipeline\_options
from apache\_beam.options.pipeline\_options import GoogleCloudOptions
from apache\_beam.runners import DataflowRunner
from google.cloud import bigquery
from apache\_beam.runners.runner import PipelineState

def Split(element):
 return element.split(",")

# Finding the average of given list of values class avgRatingFn(beam.CombineFn):

from IPython.core.display import display, HTML

```
def create accumulator(self):
    return (0.0, 0) # initialize (sum, count)
  def add_input(self, sum_count, input1):
     (sum1, count) = sum\_count
    (sum2, c2) = input1
    return sum1 + float(sum2) * c2, count + c2
  def merge_accumulators(self, accumulators):
    ind_sums, ind_counts = zip(*accumulators)
    return sum(ind_sums), sum(ind_counts)
  def extract_output(self, sum_count):
    (sum1, count) = sum_count
    return sum1 / count if count else float('NaN')
pipe = beam.Pipeline(InteractiveRunner())
# Setting up the Apache Beam pipeline options.
options = pipeline options.PipelineOptions(flags=[])
# Sets the project to the default project in the current Google Cloud environment.
_, options.view_as(GoogleCloudOptions).project = google.auth.default()
options.view_as(GoogleCloudOptions).region = 'us-west2'
dataflow_gcs_location = 'gs://bucket-amazonn/dataflowAmz'
options.view_as(GoogleCloudOptions).staging_location = '%s/staging' %
dataflow_gcs_location
```

```
# The Dataflow Temp Location location is used to store intermediate results before
outputting the final result.
options.view_as(GoogleCloudOptions).temp_location = '%s/temp' %
dataflow_gcs_location
clientBQ = bigquery.Client()
dataset_id = "keyskills-1637552843696.dataflowpipelineAmz"
dataset = bigquery.Dataset(dataset_id)
dataset.location = "us-west2"
dataset.description = "Amazon dataset books"
clientBQ.create_dataset(dataset, timeout = 30)
#Converts csv to json
def to_json(csv_file):
  fields = csv_file.split(',')
  json_file = {"Name": fields[0],
         "Author": fields[1],
         "User_Rating": fields[2],
         "Reviews": fields[3],
         "Price": fields[4],
         "Year": fields[5],
         "Genre": fields[6]
          }
  return json_file
table_schema =
'Name:STRING,Author:STRING,User_Rating:FLOAT,Reviews:INTEGER,Price:
INTEGER, Year: INTEGER, Genre: STRING'
```

```
bookAmz = (pipe | beam.io.ReadFromText("gs://bucket-
amazonn/clean_books_amazon.csv/part-00000-44653796-5293-4a2a-b34e-
225c74788cd8-c000.csv"))
(bookAmz | 'cleaned_data to json' >> beam.Map(to_json)
| 'write to bigquery' >> beam.io.WriteToBigQuery(
       "keyskills-1637552843696:dataflowpipelineAmz.tableAmz",
       schema=table_schema,
       create_disposition=beam.io.BigQueryDisposition.CREATE_IF_NEEDED,
       write_disposition=beam.io.BigQueryDisposition.WRITE_APPEND,
      custom_gcs_temp_location="gs://bucket-amazonn/dataflowAmz/temp"
    )
)
pipelineAmz = pipe.run()
if pipelineAmz.state == PipelineState.DONE:
  print('The pipeline is running successfully \n The table is sent to big query
successfully !!!')
else:
  print('Error running the pipeline')
# Reads data and split based on ','
pipe2 = beam.Pipeline(InteractiveRunner())
AmzBooks = (pipe2 | beam.io.ReadFromText("gs://bucket-
amazonn/clean books amazon.csv/part-00000-44653796-5293-4a2a-b34e-
225c74788cd8-c000.csv") | beam.Map(Split))
# Filter records having fiction. Map each rating as a set (rating,1). Using
combineperkey to count the number of each rating. Running the average function
and write the result to Fiction_result1
```

```
avg_fict = (
    AmzBooks
    | beam.Filter(lambda rec: rec[6] == "Fiction")
    | beam.Map(lambda rec: (rec[2], 1))
    | "Fict Combine keys1" >> beam.CombinePerKey(sum)
    | "Fict Combine Global keys1" >> beam.CombineGlobally(avgRatingFn())
    | "Fict Write to bucket1" >> beam.io.WriteToText("gs://bucket-
amazonn/dataflowAmz/FictionAvg_Result")
)
#Same for "non fiction"
avg_non_fict = (
    AmzBooks
    | beam.Filter(lambda rec: rec[6] == "Non Fiction")
    | beam.Map(lambda rec: (rec[2], 1))
    | "N_Fict Combine keys" >> beam.CombinePerKey(sum)
    | "N_Fict Combine Global keys" >> beam.CombineGlobally(avgRatingFn())
    | "N_Fict Write to bucket" >> beam.io.WriteToText("gs://bucket-
amazonn/dataflowAmz/NonFictionAvg Result")
)
#Same for "all genre"
avg all = (
    AmzBooks
    | beam.Map(lambda rec: (rec[2], 1))
    | "All Combine keys" >> beam.CombinePerKey(sum)
    | "All Combine Global keys" >> beam.CombineGlobally(avgRatingFn())
    | "All Write to bucket" >> beam.io.WriteToText("gs://bucket-
amazonn/dataflowAmz/AllBookAvg_Result")
```

)

# Map each record's 0th column that is name with value 1.

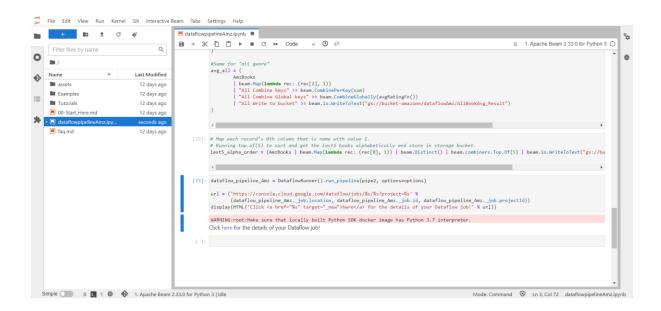
# Running top.of(5) to sort and get the last5 books alphabetically and store it.

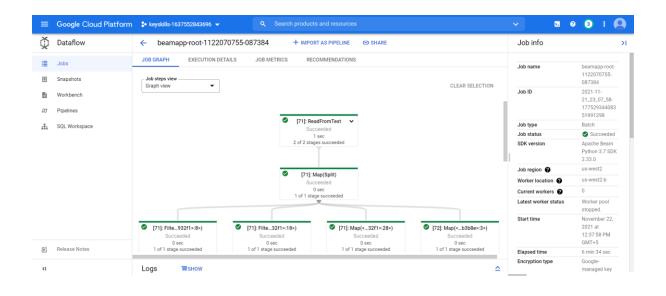
 $last5\_alpha\_order = (AmzBooks \mid beam.Map(lambda\ rec:\ (rec[0],\ 1)) \mid beam.Distinct() \mid beam.combiners.Top.Of(5) \mid beam.io.WriteToText("gs://bucket-amazonn/dataflowAmz/Last5\_Result"))$ 

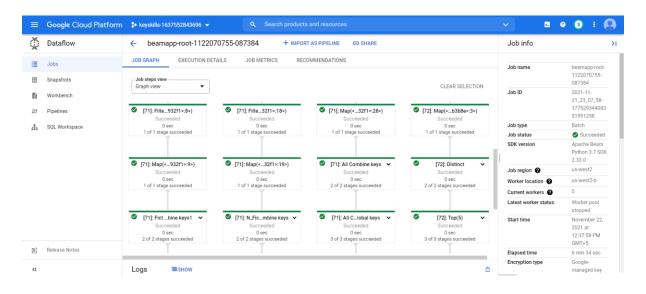
dataflow\_pipeline\_Amz = DataflowRunner().run\_pipeline(pipe2, options=options)
url = ('https://console.cloud.google.com/dataflow/jobs/%s/%s?project=%s' %

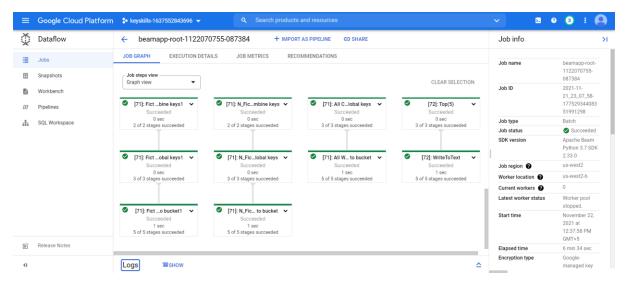
(dataflow\_pipeline\_Amz.\_job.location, dataflow\_pipeline\_Amz.\_job.id, dataflow\_pipeline\_Amz.\_job.projectId))

display(HTML('Click <a href="%s" target="\_new">here</a> for the details of your Dataflow job!' % url))

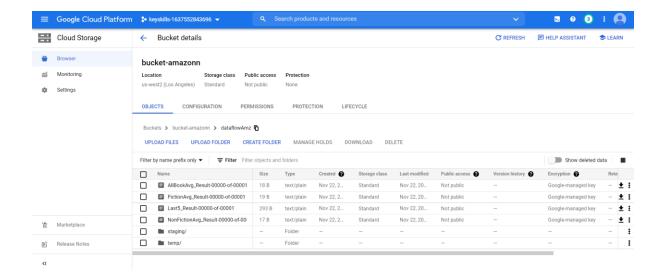








All the necessary dataflow files written to the bucket.



The dataflow pipeline read the csv file from the bucket and wrote it as a table to the dataset in BigQuery.

