

Large Scale Data Management

MSc Data Science (Part-Time)

Programming Project 2

(Kafka - Spark - Cassandra)

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Part I

Below you can find images depicting the developed script for the first part and a screenshot of the messages produced into kafka. Monitoring the messages produced into kafka was done via the excellent library provided by Magnus Edenhill, namely `kafkacat`.

You can read more about `kafkacat` at <https://github.com/edenhill/kcat>

Kakfa producer script

```
home > konstantinos > LSDM-project2 > project-2 1 > code >  
1   import csv  
2   import json  
3   import asyncio  
4   import random  
5   import time  
6  
7   from datetime import datetime  
8   from itertools import cycle  
9  
10  from aiokafka import AIOKafkaProducer  
11  
12  from faker import Faker  
13  
14  # Create a Faker instance  
15  fake = Faker()  
16  
17  # Kafka topic  
18  topic = 'spotify'
```

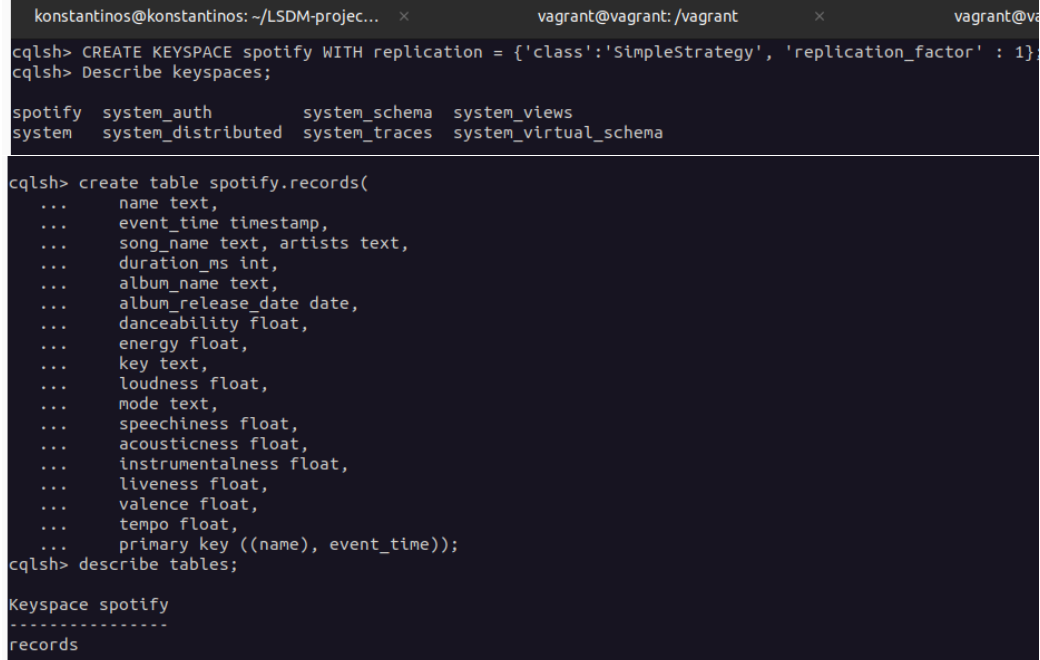
```

part1.py 2 X
part1.py > DataGenerator
20 class DataGenerator:
21
22     def load_spotify_songs(self):
23         """Load spotify songs"""
24         with open(self.spotify_songs_handle) as f:
25             reader = csv.DictReader(f)
26             data = []
27             for row in reader:
28                 song = dict(name=row['name'], danceability=float(row['danceability']))
29                 data.append(song)
30
31         self.songs = data
32
33     def create_fake_names(self):
34         """Persist a list of fake names"""
35         self.fake_names = []
36         for _ in range(0, self.fake_names_num):
37             self.fake_names.append(self.fake_instance.name())
38
39         self.name_iterator = cycle(self.fake_names)
40
41     def current_time_millis(self):
42         date = datetime.utcnow() - datetime(1970, 1, 1)
43         seconds = (date.total_seconds())
44         milliseconds = round(seconds*1000)
45
46         return milliseconds
47
48     def generate_sample(self, fake):
49         """ Create a sample payload (user, song, timestamp)
50
51         Args:
52             fake: (Boolean) Indicate if a fake or not fake person should be fetched
53         """
54         # get a user (either from fake list or student name)
55         user = None
56
57         if fake:
58             user = next(self.name_iterator)
59         else:
60             user = self.STUDENT_NAME
61
62         # randomly pick a spotify song
63         idx = random.randint(0, len(self.songs)-1)
64         selected_song = self.songs[idx]
65
66         # assemble the payload (user, song, current time)
67         time_millis = self.current_time_millis()
68         payload=dict(name=user, song=selected_song['name'],event_timestamp=str(time_millis))
69
70         return payload
71
72
73
74
75
76
77
78
79
80
81

```


Part II

Lets start with cassandra's data modeling. The following screenshots depict the commands used in order to create the data model.



```
konstantinos@konstantinos: ~/LSDM-projec... x vagrant@vagrant: /vagrant x vagrant@va
cqlsh> CREATE KEYSPACE spotify WITH replication = {'class':'SimpleStrategy', 'replication_factor' : 1};
cqlsh> Describe keyspaces;

spotify  system_auth          system_schema  system_views
system   system_distributed  system_traces  system_virtual_schema

cqlsh> create table spotify.records(
...     name text,
...     event_time timestamp,
...     song_name text, artists text,
...     duration_ms int,
...     album_name text,
...     album_release_date date,
...     danceability float,
...     energy float,
...     key text,
...     loudness float,
...     mode text,
...     speechiness float,
...     acousticness float,
...     instrumentalness float,
...     liveness float,
...     valence float,
...     tempo float,
...     primary key ((name), event_time));
cqlsh> describe tables;

Keyspace spotify
-----
records
```

You can also find these commands in the file "code/scripts-commands-and-cassandra-schema.txt"

Briefly, the spotify.records table includes the name of the user (fake, and myself), the time of the event, and all the metadata of the song the user listened to.

The most important part of this model is its primary key. In order to be able to perform aggregations for a particular person and hour the following key is employed: **primary key ((name), event_time)**

The name of the user is the partition key, and the rest of the primary key which is the event_time forms the clustering key. With this primary key in place, performing queries on a specific person in a given range of time is done in an optimal way.

With this setup we achieve fast access on the person (fast access on the partition) along with uniqueness on the pair (name, event_time). In general the partition key (name) is responsible for data distribution accross nodes and the clustering key (event_time) is responsible for data sorting within the partition.

Needless to say that event_time is configured to be of type timestamp in order to provide millisecond query precision if needed.

Spark job script

The script for the second part can be found in the images below, and of course at the "code/part2.py".

```
part2.py - Visual Studio Code

Selection View Go Run Terminal Help

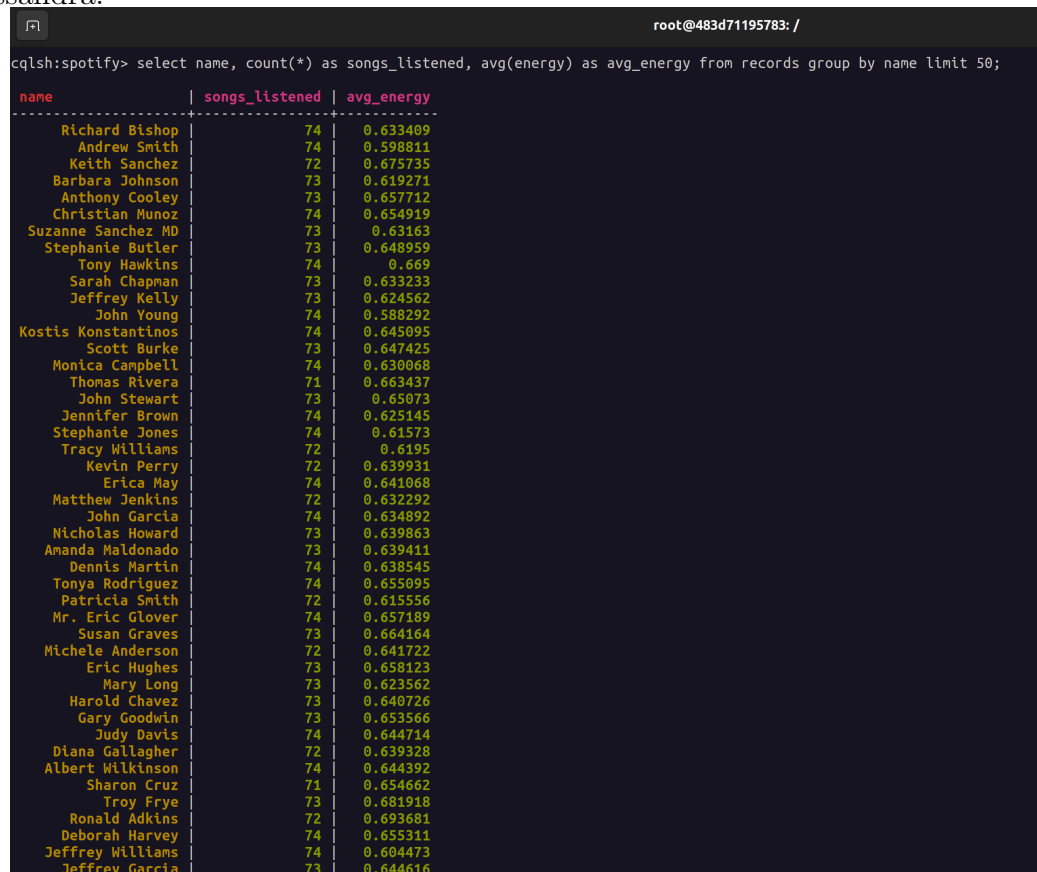
% lsdm-project-2-kostis-konstantinos.tex  part2.py 3 X

home > konstantinos > LSDM-project2 > project-2 1 > code > part2.py > ...

1 from pyspark.sql import SparkSession
2 from pyspark.sql.types import StructType, StructField, LongType, FloatType, StringType, IntegerType, DateType
3 from pyspark.sql.functions import from_json, col
4
5 # Define the write interval to sink (cassandra)
6 SINK_WRITE_INTERVAL = "30 seconds"
7
8 # The data schema to be received from kafka
9 eventSchema = StructType([
10     StructField("name", StringType(), False),
11     StructField("song", StringType(), False),
12     StructField("event_timestamp", StringType(), False)
13 ])
14
15 # The data schema of the spotify-songs csv
16 songSchema = StructType([
17     StructField("name", StringType(), False),
18     StructField("artists", StringType(), False),
19     StructField("duration_ms", IntegerType(), False),
20     StructField("album_name", StringType(), False),
21     StructField("album_release_date", DateType(), False),
22     StructField("danceability", FloatType(), False),
23     StructField("energy", FloatType(), False),
24     StructField("key", StringType(), False),
25     StructField("loudness", FloatType(), False),
26     StructField("mode", StringType(), False),
27     StructField("speechiness", FloatType(), False),
28     StructField("acousticness", FloatType(), False),
29     StructField("instrumentalness", FloatType(), False),
30     StructField("liveness", FloatType(), False),
31     StructField("valence", FloatType(), False),
32     StructField("tempo", FloatType(), False)
33 ])
34
35 # The spark session
36 spark = SparkSession.builder.appName("Spark streaming using Kafka and Cassandra")\
37     .config("spark.jars.packages", "org.apache.spark:spark-sql-kafka-0-10_2.12:3.5.0,com.datastax.spark:spark-cassandra-connector_2.12:3.0.0")\
38     .getOrCreate()
39 spark.sparkContext.setLogLevel("ERROR")
40
41 # Read (and cache) the spotify-songs csv as a dataframe
42 spotify_songs_df = spark.read.format("csv").option("header", True).schema(songSchema).load("spotify-songs.csv").\
43     withColumnRenamed("name", "song_name").cache()
44
45 # The streaming dataframe from kafka
46 streaming_df = spark.readStream.format("kafka").option("kafka.bootstrap.servers", "localhost:29092").option("subscribe", "spotify").option("startingOffsets", "latest").load()
47
48 # Read (and cache) the spotify-songs csv as a dataframe
49 spotify_songs_df = spark.read.format("csv").option("header", True).schema(songSchema).load("spotify-songs.csv").\
50     withColumnRenamed("name", "song_name").cache()
51
52 # The streaming dataframe from kafka
53 streaming_df = spark.readStream.format("kafka").option("kafka.bootstrap.servers", "localhost:29092").option("subscribe", "spotify").option("startingOffsets", "latest").load()
54
55 # The structured dataframe
56 sdf = streaming_df.selectExpr("CAST(value AS STRING)").select(from_json(col("value"), eventSchema).alias("data")).select("data.*")
57
58 # Transform the event timestamp into a Long (due to milliseconds precision)
59 transformed_df = sdf.withColumn("event_time", sdf.event_timestamp.cast(LongType)).drop("event_timestamp")
60
61 # Join the stream with the spotify-songs dataframe
62 with_join_df = transformed_df.join(spotify_songs_df, transformed_df.song == spotify_songs_df.song_name, "inner")
63
64 # The dataframe to be sent to Cassandra
65 records_df = with_join_df.drop("song")
66
67 # Write the dataframe to cassandra
68 def writeToCassandra(writeDF, _):
69     writeDF.write.format("org.apache.spark.sql.cassandra").mode('append').options(table="records", keyspace="spotify").save()
70
71 result = None
72
73 while result is None:
74     try:
75         # connect
76         result = records_df.writeStream.option("spark.cassandra.connection.host", "localhost:9042").\
77             option("failOnDataLoss", "false").foreachBatch(writeToCassandra).\
78             trigger(processingTime=SINK_WRITE_INTERVAL).\
79             outputMode("update").start().awaitTermination()
80     except:
81         pass
```

Regarding samples of about 50 records

The following screenshots verify presence of data for different people, in cassandra.



```
root@483d71195783: /
cqlsh:spotify> select name, count(*) as songs_listened, avg(energy) as avg_energy from records group by name limit 50;
```

name	songs_listened	avg_energy
Richard Bishop	74	0.633409
Andrew Smith	74	0.598811
Keith Sanchez	72	0.675735
Barbara Johnson	73	0.619271
Anthony Cooley	73	0.657712
Christian Munoz	74	0.654919
Suzanne Sanchez MD	73	0.63163
Stephanie Butler	73	0.648959
Tony Hawkins	74	0.669
Sarah Chapman	73	0.633233
Jeffrey Kelly	73	0.624562
John Young	74	0.588292
Kostis Konstantinos	74	0.645095
Scott Burke	73	0.647425
Monica Campbell	74	0.630068
Thomas Rivera	71	0.663437
John Stewart	73	0.65073
Jennifer Brown	74	0.625145
Stephanie Jones	74	0.61573
Tracy Williams	72	0.6195
Kevin Perry	72	0.639931
Erica May	74	0.641068
Matthew Jenkins	72	0.632292
John Garcia	74	0.634892
Nicholas Howard	73	0.639863
Amanda Maldonado	73	0.639411
Dennis Martin	74	0.638545
Tonya Rodriguez	74	0.655095
Patricia Smith	72	0.615556
Mr. Eric Glover	74	0.657189
Susan Graves	73	0.664164
Michele Anderson	72	0.641722
Eric Hughes	73	0.658123
Mary Long	73	0.623562
Harold Chavez	73	0.640726
Gary Goodwin	73	0.653566
Judy Davis	74	0.644714
Diana Gallagher	72	0.639328
Albert Wilkinson	74	0.644392
Sharon Cruz	71	0.654662
Troy Frye	73	0.681918
Ronald Adkins	72	0.693681
Deborah Harvey	74	0.655311
Jeffrey Williams	74	0.604473
Jeffrey Garcia	73	0.644616

```
cqlsh:spotify> select name, song_name, album_name, energy, danceability from records where name='Stephanie Butler' limit 50;
```

name	song_name	album_name	energy	danceability
Stephanie Butler	Sans Coeur	Pourvu qu'il pleuve	0.764	0.765
Stephanie Butler	Rahasia Hati	Paradoks	0.526	0.542
Stephanie Butler	AMBIENTE ERRADO - Ao Vivo	LUAN CITY 2.0 - FASE 2	0.845	0.484
Stephanie Butler	Γολροφα	NARRATIVE	0.542	0.666
Stephanie Butler	A Rainy Night in Soho	Rum Sodomy & The Lash (Expanded Edition)	0.449	0.375
Stephanie Butler	RITUAL	LVEU: VIVE LA TUYA...NO LA MIA	0.706	0.832
Stephanie Butler	Сher пастаял на плечах	Сher пастаял на плечах	0.952	0.72
Stephanie Butler	Hartslag Van De Stad	Hartslag Van De Stad	0.737	0.808
Stephanie Butler	Stal	Fabula	0.433	0.562
Stephanie Butler	Fact Check	Fact Check - The 5th Album	0.988	0.758
Stephanie Butler	Sena Kalpu Dziesma	Tautas Laiks (Latviesu Patriotisko Dziesmu Izlase)	0.759	0.642
Stephanie Butler	Cariceps	Cariceps	0.809	0.963
Stephanie Butler	Widok	Widok	0.739	0.845
Stephanie Butler	Tāl Vi Sao	99%	0.682	0.76
Stephanie Butler	Soy Feo Pero Rico	De Amor y Vacilón	0.847	0.738
Stephanie Butler	GIPSY TRAP	GIPSY TRAP	0.621	0.82
Stephanie Butler	Mucki Bar	Når sjælen kaster op	0.739	0.785
Stephanie Butler	3D (Justin Timberlake Remix)	3D (Justin Timberlake Remix)	0.854	0.817
Stephanie Butler	CHUPON	CHUPON	0.748	0.73
Stephanie Butler	Longbiča / Prospekti	Melnežera grāmatā	0.701	0.763
Stephanie Butler	Stjernestøv	Stjernestøv	0.346	0.38
Stephanie Butler	käpy	käpy	0.631	0.705
Stephanie Butler	Halloween Night in the Forest	Halloween Night in the Forest	0.186	0.187
Stephanie Butler	Never Ever	Presido La Pluto	0.484	0.597
Stephanie Butler	Mang Compromisa (feat. Focalistic & Makhekhe Jr)	Boroko Keng	0.644	0.879
Stephanie Butler	Longbiča / Prospekti	Melnežera grāmatā	0.701	0.763
Stephanie Butler	Nichtjo	Nichtjo	0.708	0.686
Stephanie Butler	STAY (with Justin Bieber)	F*CK LOVE 3+: OVER YOU	0.764	0.591
Stephanie Butler	Ik Maak Mee	Ik Maak Mee	0.405	0.694
Stephanie Butler	Badman	Badman	0.497	0.841
Stephanie Butler	Ya No Vuelvas (Versión Cuarteto)	Ya No Vuelvas (Versión Cuarteto)	0.9	0.736
Stephanie Butler	Blonde Chaya	Blonde Chaya	0.81	0.815
Stephanie Butler	Saiyan	Chef D'orchestre	0.808	0.775
Stephanie Butler	จังหวัดขอนแก่น - Magic Moment	จังหวัดขอนแก่น (Magic Moment)	0.424	0.563
Stephanie Butler	Neman mood	FLOWERHORT	0.797	0.833
Stephanie Butler	1 2 3 4	SÜR ET CERTAIN	0.668	0.825
Stephanie Butler	El Diario De Um Borracho	El Condor Legendario	0.703	0.754
Stephanie Butler	Нічченька	Нічченька	0.533	0.575
Stephanie Butler	Clean (Taylor's Version)	1989 (Taylor's Version)	0.386	0.771
Stephanie Butler	Arbolito de Navidad	El Mejor Disco de Diciembre, Vol. 2	0.631	0.741
Stephanie Butler	Lucas Tenues	Lucas Tenues	0.802	0.76
Stephanie Butler	Rockin' Around The Christmas Tree	Merry Christmas From Brenda Lee	0.472	0.589
Stephanie Butler	Aşk Olsun	Aşk Olsun	0.849	0.754
Stephanie Butler	Dyn	Dyn	0.718	0.718

Queries on my name

Below you can find screenshots of the queries (and results) for my name, as requested in the exercise.

```
vagrant@vagrant: /vagrant
```

```
cqlsh:spotify> select avg(danceability) from records where name='Kostis Konstantinos' and event_time >= '2024-02-25 22:00' and event_time <= '2024-02-25 23:00';
```

system.avg(danceability)
0.666568

```
(1 rows)
```

```
cqlsh:spotify> select count(*) from records where name='Kostis Konstantinos' and event_time >= '2024-02-25 22:00' and event_time <= '2024-02-25 23:00';
```

count
74

```
(1 rows)
```

```
cqlsh:spotify>
```



```
root@483d71195783:/
qlsh:spotify> select song_name, danceability from records where name='Kostis Konstantinos' and event_time >= '2024-02-25 22:00' and event_time <= '2024-02-25 23:00';

song_name | danceability
-----|-----
          | 0.658
          | 0.771
          | 0.823
          | 0.833
          | 0.813
          | 0.811
          | 0.681
          | 0.315
0.583 | عراقي سريع - هبة جويي وياه | مطلوب انترنت |
          | 0.722
          | 0.852
          | 0.861
          | 0.896
          | 0.557
          | 0.327
          | 0.903
          | 0.587
          | 0.549
          | 0.725
          | 0.877
          | 0.419
          | 0.833
          | 0.727
          | 0.642
          | 0.584
          | 0.683
          | 0.252
          | 0.875
          | 0.825
          | 0.571
          | 0.65
          | 0.728
          | 0.804
          | 0.639
          | 0.694
          | 0.54
          | 0.586
          | 0.266
          | 0.315
          | 0.772
          | 0.694
          | 0.784
          | 0.701
          | 0.825
          | 0.571
          | 0.65
          | 0.728
          | 0.804
          | 0.639
          | 0.694
          | 0.54
          | 0.586
          | 0.266
          | 0.315
          | 0.772
          | 0.694
          | 0.784
          | 0.701
          | 0.58
          | 0.814
          | 0.928
          | 0.707
          | 0.696
          | 0.49
          | 0.528
          | 0.56
          | 0.709
          | 0.732
          | 0.639
          | 0.691
          | 0.77
          | 0.173
          | 0.623
          | 0.763
          | 0.39
          | 0.716
          | 0.679
          | 0.804
          | 0.846
          | 0.759
          | 0.794
          | 0.5
          | 0.701
          | 0.685
          | 0.696
          | 0.678
          | 0.429
          | 0.547
          | 0.751

(74 rows)
qlsh:spotify>
```