Prepare a report detailing the HQL queries for creating Hive tables and the results obtained for the HQL tasks

Note: We will also leverage the Hive External Tables (non_event_data and event_data) here that were created for generating the S3 datasets as documented in DataPreparation.pdf

Hive connection using beeline:

```
[hadoop@ip-172-31-82-238 ~]$ beeline -u jdbc:hive2://localhost:10000/default -n hadoop
Connecting to jdbc:hive2://localhost:10000/default
Connected to: Apache Hive (version 2.3.4-amzn-2)
Driver: Hive JDBC (version 2.3.4-amzn-2)
Transaction isolation: TRANSACTION REPEATABLE READ
Beeline version 2.3.4-amzn-2 by Apache Hive
Hive database creation:
0: jdbc:hive2://localhost:10000/default> create database if not exists telco db;
No rows affected (1.378 seconds)
0: jdbc:hive2://localhost:10000/default> show databases;
| database_name |
+----+
| default
| telco_db
+----+
2 rows selected (0.318 seconds)
0: jdbc:hive2://localhost:10000/default> describe database telco db;
----+
| db_name | comment |
                               location
                                                        owner_name | owner_type |
parameters
| hdfs://ip-172-31-82-238.ec2.internal:8020/user/hive/warehouse/telco_db.db | hadoop |
telco_db
USER
1 row selected (0.058 seconds)
0: jdbc:hive2://localhost:10000/default> use telco_db;
No rows affected (0.042 seconds)
Hive configuration:
0: jdbc:hive2://localhost:10000/default> set hive.cli.print.header=true;
No rows affected (0.011 seconds)
0: jdbc:hive2://localhost:10000/default> set hive.resultset.use.unique.column.names=false;
No rows affected (0.015 seconds)
```

```
0: jdbc:hive2://localhost:10000/default> set hive.exec.dynamic.partition=true;
No rows affected (0.011 seconds)
0: jdbc:hive2://localhost:10000/default> set hive.exec.dynamic.partition.mode=nonstrict;
No rows affected (0.006 seconds)
0: jdbc:hive2://localhost:10000/default> set hive.strict.checks.cartesian.product=false;
No rows affected (0.008 seconds)
0: jdbc:hive2://localhost:10000/default> set hive.mapred.mode=nonstrict;
No rows affected (0.011 seconds)
Hive external table creation and data loading for 'train' table:
0: jdbc:hive2://localhost:10000/default> create external table if not exists train (
 . . . . . . . . . . . . . . . . gender string,
  . . . . . . . . . . . . . . . . . . age bigint,
  . . . . . . . . . . . . . . . . . . group_train_string
    .....| location '/user/hadoop/telco/train/';
No rows affected (0.457 seconds)
0: jdbc:hive2://localhost:10000/default> select count(*) from train;
| _c0 |
+----+
74645
1 row selected (20.985 seconds)
0: jdbc:hive2://localhost:10000/default> select * from train limit 5;
device_id | gender | age | group_train |
5 rows selected (2.622 seconds)
Hive external table creation and data loading for 'brand_device' table:
```

```
No rows affected (0.091 seconds)
0: jdbc:hive2://localhost:10000/default> select count(*) from brand_device;
| _c0 |
187245
1 row selected (7.478 seconds)
0: jdbc:hive2://localhost:10000/default> select * from brand_device limit 5;
 device_id | phone_brand | device_model |
| 1845358998536310000 | meitu | 2
| 3126957642374570000 | meitu | 2
| -3051457881268070000 | meitu | 2
| 4608241502940040000 | meitu | 2
| 6005031767544890000 | meitu | 2
5 rows selected (0.269 seconds)
Hive external table creation and data loading for 'events' table:
0: jdbc:hive2://localhost:10000/default> create external table if not exists events (
. . . . . . . . . . . . . . . . . device_id string,
 ..... event_time string,
  ..... | location '/user/hadoop/telco/events/';
No rows affected (0.09 seconds)
0: jdbc:hive2://localhost:10000/default> select count(*) from events;
| _c0 |
3252950
1 row selected (19.082 seconds)
0: jdbc:hive2://localhost:10000/default> select * from events limit 5;
event_id | device_id | event_time | longitude | latitude |
1 | 29182687948017100 | 2016-05-01 00:55:25.0 | 121.38 | 31.24
       | -6401643145415150000 | 2016-05-01 00:54:12.0 | 103.65 | 30.97
       | -4833982096941400000 | 2016-05-01 00:08:05.0 | 106.6
| 3
                                                  29.7
       | -6815121365017310000 | 2016-05-01 00:06:40.0 | 104.27 | 23.28
5 | -5373797595892510000 | 2016-05-01 00:07:18.0 | 115.88 | 28.66
```

5 rows selected (0.196 seconds)

Hive external table creation and data loading for 'app_events' table:

```
0: jdbc:hive2://localhost:10000/default> create external table if not exists app events (
No rows affected (0.059 seconds)
0: jdbc:hive2://localhost:10000/default> select count(*) from app events;
| _c0 |
1 row selected (13.979 seconds)
0: jdbc:hive2://localhost:10000/default> select * from app events limit 5;
+----+
event_id app_id is_installed is_active |
+----+
2 | 5927333115845830913 | 1 | 1
    | -5720078949152207372 | 1
    -1633887856876571208 1
| 2
    -653184325010919369 1
                     | 1
    8693964245073640147 | 1 | 1
+-----+--------
5 rows selected (0.18 seconds)
```

Hive external table creation and data loading for 'app_labels' table:

```
0: jdbc:hive2://localhost:10000/default> create external table if not exists app labels (
No rows affected (0.065 seconds)
0: jdbc:hive2://localhost:10000/default> select count(*) from app labels;
| _c0 |
1 row selected (7.131 seconds)
0: jdbc:hive2://localhost:10000/default> select * from app_labels limit 5;
```

,

Hive external table creation and data loading for 'label_categories' table:

```
0: jdbc:hive2://localhost:10000/default> create external table if not exists label categories (
 ..... | location '/user/hadoop/telco/label_categories/'
No rows affected (0.131 seconds)
0: jdbc:hive2://localhost:10000/default> select count(*) from label_categories;
| _c0 |
+----+
930
1 row selected (6.642 seconds)
0: jdbc:hive2://localhost:10000/default> select * from label_categories limit 5;
| label_id | category |
| 5 | game-Leisure time |
5 rows selected (0.207 seconds)
```

Hive queries and analysis:

1. The 10 most popular brands and the percentage of the respective Male and Female owners of these brands [Handle the device id duplicates from brand_device table.]

We will first de-duplicate the brand_device table with inline OVERWRITE by grouping over the device_id column and selecting the first row in each group to eliminate the duplicates using this Hive Query –

We can see that the brand_device row count **reduced from 187245 to 186713** after eliminating the duplicate rows. This matches the **Total Device IDs Count** and **Unique Device IDs Count** numbers as seen in **SQLTasks.pdf**

We will leverage the non_event_data table to check the 10 most popular brands overall irrespective of gender -

+	+			
phone_brand	bcount			
+	++			
Xiaomi	17299			
samsung	13669			
Huawei	12960			
OPPO	5783			
vivo	5637			
Meizu	4699			
Coolpad	3339			
lenovo	2691			
Gionee	1123			
HTC	1013			
++				

10 rows selected (7.553 seconds)

Finally we will extract the 10 most popular brands from the **non_event_data** table with the respective gender split counts and percentages using above query as a Common Table Expression (CTE) -

brand	brand_count	female_count	male_count	female_percentage	male_percentage
Xiaomi	17299	5918	11381	34.21	65.79
samsung	13669	5431	8238	39.73	60.27
Huawei	12960	4244	8716	32.75	67.25
OPPO	5783	2571	3212	44.46	55.54
vivo	5637	2651	2986	47.03	52.97

ı	Meizu	4699	1302	3397	27.71	72.29	
İ	Coolpad	3339	1079	2260	32.32	67.68	
ĺ	lenovo	2691	893	1798	33.18	66.82	
	Gionee	1123	402	721	35.8	64.2	
	HTC	1013	320	693	31.59	68.41	
		_					

¹⁰ rows selected (17.21 seconds)

2. The 10 most popular brands for Male and Female?

We will leverage the **non_event_data** table to extract and concatenate (UNION operation) the 10 most popular brands for Female and Male genders -

gender	brand	brand_count	
F	Xiaomi	5918	
į F	samsung	5431	
į F	Huawei	4244	
F	vivo	2651	
F	OPPO	2571	
F	Meizu	1302	
F	Coolpad	1079	
F	lenovo	893	
F	Gionee	402	
F	HTC	320	

M	Xiaomi	11381
M	Huawei	8716
M	samsung	8238
M	Meizu	3397
M	OPPO	3212
M	vivo	2986
M	Coolpad	2260
M	lenovo	1798
M	Gionee	721
j M	HTC	693
_		

20 rows selected (13.784 seconds)

We can see that the 10 most popular brands are overall the same for Female and Male genders but with some interim ranking changes. For example, Samsung and Huawei are seen to swap ranks 2 and 3 across the two genders.

3. The count and percentage analysis of the Gender in the train data set

We will extract the respective gender counts and percentages using group by gender on the 'train' table along with a CTE to self-join with the 'train' table -

4. The top mobile phone brands offering the highest number of models [Provide details about the top three brands.]

We will extract the top phone brands using group by phone_brand on the 'brand_device' table and then count the distinct device_model's within each brand in descending order to yield the top 3 brands -

5. The average number of events per device id [Applicable to the device_id column from the train table, which has at least one associated event in the event table]

We will leverage the **event_data** table to check the device_events_count for a few device_id rows that have one or more associated events -

Finally we will use above query as a CTE to aggregate and compute the overall avg_events_per_device number -

6. Whether the count and percentage of the device_id column in the train table have corresponding events data available

We will first check the count of device_id's from the 'train' table that have corresponding associated events from the 'events' table -

Finally we will extract the percentage value using a CTE to self-join with the 'train' table -