Report

Assignment 2 - MySQL

Group: 50
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Introduction

The assignment was to insert and interact with the large Geolife GPS Trajectory dataset using a MySQL database and Python. The dataset contains data for user, activities, and track points, which the group inserted into separate tables. The user can have many activities (many to one), and an activity may have many track points representing location data (many to one).

The group chose to run the inserts locally using a MySQL-container in Docker. We did additionally use a PHPMyAdmin-container to simplify viewing the tables and testing SQL-queries. Development was done with the Visual Studio Code feature "Live Share", giving the opportunity for all three group members to work concurrently on the code. After each work session the progress was committed to the Github repository found here: https://github.com/Her0elt/TDT4225-Very-Large-Distributed-Data-Volumes

For the tasks in part two, the group's approach was to try to do as much as possible with SQL-queries due to the performance/speed compared to Python. Even though efficiency was not a criteria, it should be considered when working with large amounts of data. It is also a fun challenge and good practice. For certain more complicated tasks, Python operations were needed to process the returned data in order to answer properly.

Results

Part 1

1. Connects to the MySQL server on your Ubuntu virtual machine.

Since we chose to do the task locally, rather than connecting to the VM, we connected to the MySQL-container in Docker. Docker Compose was used to easily run MySQL and PHPMyAdmin together. The Makefile defines the commands used for setup and development.

2. Creates and defines the tables User, Activity and TrackPoint

The types are as defined in the text document. Has_labels is defaulted to false as it cannot be null. The activity-id was created by combining the trajectory-filename and user_id. The integer id in track point is auto incremented. For the track_points, we decided to ignore the days-since-column and combine date and time into a datetime-column as this was enough to complete the tasks.

```
def create_user_table(self):
    query = """CREATE TABLE IF NOT EXISTS users (
                id VARCHAR(225) NOT NULL PRIMARY KEY,
                has_labels BOOL NOT NULL DEFAULT false
                )
            """
    return query

def create_activity_table(self):
    query = """CREATE TABLE IF NOT EXISTS activities (
                id VARCHAR(225) NOT NULL PRIMARY KEY,
                transportation_mode VARCHAR(120),
                start_date_time DATETIME NOT NULL,
                end_date_time DATETIME NOT NULL,
                user_id VARCHAR(225) NOT NULL,
                CONSTRAINT FK_UserActivity FOREIGN KEY (user_id) REFERENCES users(id)
            )
            """
    return query

def create_track_point(self):
    query = """CREATE TABLE IF NOT EXISTS track_points (
                id INT AUTO_INCREMENT NOT NULL PRIMARY KEY,
                lat DOUBLE(20, 10),
                lon DOUBLE(20, 10),
                altitude INT(30),
                date_time DATETIME NOT NULL,
                activity_id VARCHAR(225) NOT NULL,
                CONSTRAINT FK_ActivityTrackPoint FOREIGN KEY (activity_id) REFERENCES activities(id)
                )
            """
    return query
```

3. Inserts the data from the Geolife dataset into your MySQL database

For the insertion of the data, we chose to firstly insert the users, as these exist independent of any other table. To insert users, we first need to find the users to insert, and their has_labels field. As there are not too much data to insert yet, we can do this in one executemany operation:

```python
def _get_users_with_labeled_ids(self):
    """Get a list of users with labeled activities from the `labeled_ids`.txt-file"""
    path = "./dataset/dataset/labeled_ids.txt"
    with open(path, "r") as labeled_ids:
        return list(map(lambda x: x.rstrip("\n"), labeled_ids.readlines()))

def _get_users(self):
    """Get a list of users and if they have created labels"""
    path = "./dataset/dataset/Data/"
    data = []
    users_with_labeled_ids = self._get_users_with_labeled_ids()
    for user_id in os.listdir(path):
        has_labels = user_id in users_with_labeled_ids
        data.append((user_id, has_labels))
    return data

def insert_users(self):
    users = self._get_users()

    query = "INSERT INTO users (id, has_labels) VALUES (%s, %s)"
    self.cursor.executemany(query, users)
    self.db_connection.commit()
```

Activities and their corresponding trajectories / track points were inserted together, and executed in batches. As these are large operations with a lot of data, this was considered to be the most efficient option. To limit memory usage, the execution is limited to one activity at a time. Meaning that the program first inserts an activity for a user and executes, then finds and inserts all the corresponding track points with executemany. This process is repeated for all user activities, before it moves on to the next user.

```python
# 1. For each user:
# 2. Loop through each activity-file: `trajectory/<yyyyMMddHHmmss>.plt`.
#    1. Check if there exists a activity in the users `labels.txt` where start time corresponds with the name of the trajectory-file.
#       1. If there is a match, use the `labels.txt`-listing to create an activity before all track-points are added to the activity
#       2. If there is no match, use the first and last point in the trajectory to create an activity before all track-points are added
def insert_user_activities(self, user):
    path = f"./dataset/dataset/Data/{user[0]}/Trajectory/"
    for user_file in os.listdir(f"{path}"):
        self.create_activity(user, user_file)

def get_track_points_from_file(self, trajectory_lines, activity_id):
    return list(
        map(
            lambda line: Trajectory(line).to_tuple(activity_id),
            trajectory_lines[6:],
        )
    )

def insert_trajectories(self):
    count = 0
    users = self._get_users()
    for user in users:
        print(f"Start {user[0]}, count: {count}")
        self.insert_user_activities(user)
        self.db_connection.commit()
        count += 1
        print(f"Finished {user[0]}, count: {count}")
```

```python
def create_activity(self, user, file_path):
    user_id, has_labels = user
    activityId = f"{file_path}_{user_id}"
    with open(
        f"./dataset/dataset/Data/{user_id}/Trajectory/{file_path}", "r"
    ) as trajectory_file:
        trajectory_lines = trajectory_file.readlines()

        # Don't add activities with more then 2500 trajections
        # (2506 because the first 6 lines contains other information)
        if len(trajectory_lines) > 2506:
            return

        activityQuery = None
        # The first 6 lines in each trajectory file contains useless information,
        # we therefore start a index: 6
        first_trajectory = Trajectory(trajectory_lines[6])
        last_trajectory = Trajectory(trajectory_lines[-1])
        # If the user has_labels
        if has_labels:
            with open(
                f"./dataset/dataset/Data/{user_id}/labels.txt", "r"
            ) as labels_file:
                # For all lines in label file, compare to trajectory date, insert if match
                for label_line in labels_file.readlines()[1:]:
                    labels_activity = LabelsActivity(label_line)
                    if (
                        labels_activity.start_date == first_trajectory.date
                        and labels_activity.end_date == last_trajectory.date
                    ):
                        activityQuery = self._insert_activity_query(
                            activityId,
                            labels_activity.transportation_mode,
                            labels_activity.start_date,
                            labels_activity.end_date,
                            user_id,
                        )
                        continue

        if not activityQuery:
            activityQuery = self._insert_activity_query(
                activityId,
                None,
                first_trajectory.date,
                last_trajectory.date,
                user_id,
            )

        self.cursor.execute(activityQuery)

        track_point_data = self.get_track_points_from_file(
            trajectory_lines, activityId
        )
        self.cursor.executemany(self._insert_track_point_query(), track_point_data)
```

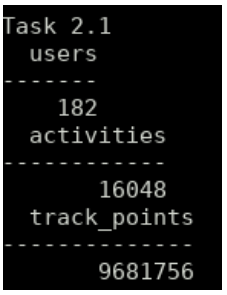
It is **important to note** that for our specific implementation, we chose to interpret the task to ask for an exact match for activity and track points on **either the start and end time of the transportation mode activity.** We are aware that one way to interpret the task is to consider all datetimes of the track points in the file. This will result in more relations between activities and track points, however we chose the more strict way of handling the relation based on our understanding of the task.

Part 2

For these tasks, most were done solely with SQL-queries, others were done partly in Python. Some are pretty straight forward, while for some tasks a short explanation is necessary.

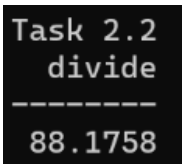
1. How many users, activities and trackpoints are there in the dataset (after it is inserted into the database).

Answer: 182 users, 16048 activities, 9681756 trackpoints



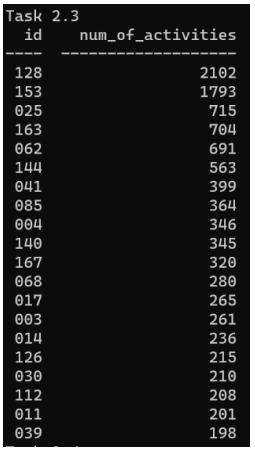
2. Find the average number of activities per user.

Answer: 88.1758 activities on average per user



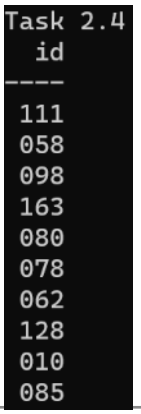
3. Find the top 20 users with the highest number of activities.

Answer: To solve this, the count of entries is selected as num_of_activites. We group the users by their id, and order by the count. The table represent the top 20 most active users.



4. Find all users who have taken a taxi.

Answer; By selecting distinct users, we find that the following users have taken a taxi: { 111, 058, 098, 163, 080, 078, 062, 128, 010, 085 }.



5. Find all types of transportation modes and count how many activities that are tagged with these transportation mode labels. Do not count the rows where the mode is null.

Answer: The table represents the most used transportation modes. Note that we check if we've inserted 'None' when no transportation mode is registered. This is not optimal, and we should have used a Null value instead, and would change this if we were to do the assignment again.

```
Task 2.5
transportation_mode      transportation_count
---------------------  ----------------------
walk                                      480
car                                       419
bike                                      263
bus                                       199
subway                                    133
taxi                                       37
airplane                                    3
train                                       2
run                                         1
boat                                        1
```

6. Task 6
   a. Find the year with the most activities.

```
Task 2.6 a)
  start_year    year_count
------------  ------------
        2008          5895
        2009          5879
        2010          1487
        2011          1204
        2007           994
        2012           588
        2000             1
```

Answer: To find the year with the most activities, we group by start year and order by year count (we could limit 1, but out of interest we wanted to see all years). The most activities were registered in 2008.
Note: A task for a year is based on the start_date. This means that a task started in 2008 and ended in 2009, would be registered as an activity from 2008. This is to prevent cases where the sum of activities from each year exceeds the total amount of activities due to overlap.

   b. Is this also the year with the most recorded hours?

```
Task 2.6 b)
  start_year     sum_time
------------  ----------
        2009         11612
        2008          9201
        2007          2315
        2010          1389
        2011          1132
        2012           711
        2000             0
```

Answer: The group understands the task asks for the total time spent on activities in total (mort recorded hours). By finding the sum of time difference in seconds and dividing by 3600, we can see that the year with the most recorded hours was in fact 2009, with 2008 moving down to second place.

7. Find the total distance (in km) walked in 2008, by user with id=112

```
Task 2.7
Total distance by user 112 in 2008: 115.47465961508007 km
```

Answer: We retrieve all track_points in activities that belong to the user: 112, has transportation_mode: 'walk' and is in 2008. Then we group the track_points into their activity and calculate the total distance walked in each activity with a function that finds the distance in kilometers for a list of geo-coordinates. Finally the distance for each activity is summed up.

8. Find the top 20 users who have gained the most altitude meters.

*Answer: Loops through all trackpoints and creates a count of altitude gained for each user. Only increases it if the compared trackpoint is from the same user **and** activity. Finally the users are sorted by altitude gained and limited to top 20.*

Task 2.8

UserId	Altitude meters gained
128	650890
153	554942
004	332036
041	240758
003	233664
085	217642
163	205264
062	181688
144	179456
030	175680
039	146704
084	131161
000	121505
002	115063
167	112973
025	109148
037	99220.9
140	94838.8
126	83024.2
017	62566.3

9. Find all users who have invalid activities, and the number of invalid activities per user

Task 2.9

UserId	Invalid activities
128	720
153	557
025	263
062	249
163	233
004	219
041	201
085	184
003	179
144	157
039	147
068	139
167	134
017	129
014	118
030	112
126	105
000	101
092	101
037	100
084	99
002	98
104	97
034	88
140	86
112	67
091	63
038	58
115	58
022	55
042	55
174	54
142	52
010	50
015	46
101	46
001	45
005	45
052	44
012	43
089	40
028	36
051	36
096	35
036	34
067	33
011	32
044	32
009	31
019	31
134	31
007	30
147	30
155	30
013	29
071	29
179	28
018	27
024	27
082	27
065	26
111	26
029	25
125	25
103	24
035	23
119	22
043	21
016	20
020	20
074	19
078	19
168	19
026	18
073	18
006	17
040	17
110	17
008	16
057	16
081	16
094	16
150	16
055	15
083	15
097	14
154	14
181	14
046	13
058	13
102	13
032	12
061	12
139	12
023	11
088	11
099	11
131	10
138	10
105	9
157	9
158	9
162	9
169	9
172	9
050	8
063	8
076	8
130	8
176	8
021	7
045	7
053	7
056	7
064	7
146	7
161	7
047	6
066	6
069	6
075	6
080	6
122	6
129	6
136	6
164	6
059	5
070	5
086	5
098	5
108	5
135	5
145	5
159	5
173	5
093	4
095	4
121	4
124	4
127	4
133	4
175	4
031	3
077	3
087	3
090	3
100	3
106	3
109	3
114	3
117	3
118	3
123	3
132	3
171	3
027	2
033	2
054	2
072	2
079	2
152	2
165	2
166	2
170	2
180	2
048	1
060	1
107	1
113	1
141	1
151	1

Answer: Loops through all trackpoints and creates a count of invalid activities per user. Only increases it if the compared trackpoint is more than 5 minutes apart and the activity hasn't been marked as invalid already. Finally the users are sorted by the amount of invalid activities and limited to top 20.

10. Find the users who have tracked an activity in the Forbidden City of Beijing

Answer: We use the BETWEEN method to find the latitude and longitude values that are within the definition of the task (since the track point data is more precise and has more decimals we cannot do an exact match). The distinct select gives us user 018 and 019.

```
Task 2.10
  id
----
 018
 019
```

11. Find all users who have registered transportation_mode and their most used transportation_mode.

Answer: Using a sub-query which returns transportation_mode and user_id grouped, it is easy to find the most used transportation_mode for each user with the MAX-function. We had to use "transportation_mode != 'None'" here as well since activities without a transportation_mode was registered with None and not NULL. Only 59 of the 69 users in labeled_ids.txt do have activities which are labeled. The 10 users without labeled activities are not included in this list.

```
Task 2.11
  id  most_used_transportation_mode
----  ------------------------------
 010  taxi
 020  walk
 021  walk
 052  bus
 056  bike
 058  walk
 060  walk
 062  walk
 064  bike
 065  walk
 067  walk
 069  bike
 073  walk
 075  walk
 076  car
 078  walk
 080  taxi
 081  walk
 082  walk
 084  walk
 085  walk
 086  walk
 087  walk
 089  walk
 091  walk
 092  walk
 097  bike
 098  taxi
 101  car
 102  walk
 107  walk
 108  walk
 111  taxi
 112  walk
 115  walk
 117  walk
 125  bus
 126  walk
 128  walk
 136  walk
 138  bike
 139  walk
 144  walk
 153  walk
 161  walk
 163  walk
 167  walk
 175  bus
```

Discussion

We found the overall assignment to be instructive and fun. It was a little challenging to navigate the file to properly insert the track_points to the activities, but conceptually it was fairly straight forward. Some parts of the task were either a bit hard to grasp, or left to interpretation. For example whether to cut down track points to 2500, or leave them out entirely if more than 2500 was registered (this was the solution choice based on an answer in piazza). In these cases, the answers from the assistants or lectures were quite helpful.

Since we decided to only match activities with transportation mode, on trackpoints with exact matches on start and end date only, it is possible that some tasks gave fewer results than the alternative, regardless we feel it still answered the task description. Regardless of the specifics in the data cleaning, the queries and methods designed for task 2 should be scalable regardless of how one chooses to insert the data.

Some of the things we have learned through this assignment is that it's important to read the tasks and assignment-description carefully. If we'd done that better from the start, we would probably have used less time on the assignment, not having to edit the insert script and run the inserts multiple times, something that took quite a long time on a couple of our machines. Additionally we've also become even more familiar with MySQL and it's built-in functions and query-options. Executing queries in MySQL really is much quicker than to manually do them in Python.