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1 TDT4225 - Report

1.1 Assignment 2 - MySQL

• Group: 89

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1.1.1 Introduction

This project focuses on analyzing the Geolife GPS Trajectory dataset using MySQL and Python. The dataset, which tracks the outdoor movements of 182 users, requires setting up a database on a MySQL server and performing data cleaning before insertion. The tasks include creating and defining tables for users, activities, and trackpoints, integrating and cleaning data from the dataset, and ensuring only activities with fewer than 2,500 trackpoints are included. Additionally, we develop SQL queries to extract insights about user activities, transportation modes, distances covered, and altitude changes. The project aims to uncover patterns in user behavior by efficiently managing and querying large amounts of GPS data.

1.1.2 Results

• DATA INSERTION

As we can see from the results, the time taken to enter data into the database is very high when there are large volumes of data, such as the **TrackPoint table** which contains more than **nine million data points**. The time was greatly reduced for the entry of track points by using a *batch entry* of points instead of a *single point by point entry*, in fact it went from about **1700 seconds taken to enter all points** (*about half an hour*) to just over **180 seconds** (*3 minutes*).

```
First 10 rows of User table:
  id | has_labels |
   000 I
                     0
   001
                     0
  002
                     0
  003
                     0
  004
                     0
  005 I
                     0
  006
                     0
  007
                     0
  008
                     0
   009
                     0
```

Fί	rst 10 m	rows of Activity table:		
	id	user_id transportation_mod	T	end_date_time
	1 2 3 4 5 6 7 8 9	135 135 135 135 135 135 135 135 135 132	2009-01-02 04:31:27 2009-01-27 03:00:04 2009-01-10 01:19:47 2009-01-14 12:17:57 2009-01-12 01:41:22 2008-12-24 14:42:07 2008-12-28 10:36:05 2010-02-15 10:56:35	2009-01-03 05:40:31 2009-01-02 04:41:05 2009-01-27 04:50:32 2009-01-10 04:42:47 2009-01-14 12:30:53 2009-01-12 02:14:01 2008-12-24 15:26:45 2008-12-28 12:19:32 2010-02-15 12:22:33 2010-05-01 00:35:31
+-	+		+	++

	id I	activity_id	I la	t I	lon	altitude	I	date_days		date_time	
i											
	1 I	1	1 39.974	3 I	116.4	l 492		39816.1		2009-01-03	01:21:34
	2	1	1 39.974	3 I	116.4	492		39816.1	П	2009-01-03	01:21:35
	3 I	1	1 39.974	3 I	116.4	l 492	1	39816.1		2009-01-03	01:21:36
	4 I	1	1 39.974	3 I	116.4	492	1	39816.1		2009-01-03	01:21:38
	5 I	1	1 39.974	4 1	116.4	491	1	39816.1		2009-01-03	01:21:39
	6 I	1	1 39.974	4 I	116.4	491	1	39816.1		2009-01-03	01:21:42
	7 I	1	1 39.974	4 1	116.4	491	1	39816.1		2009-01-03	01:21:46
	8 I	1	1 39.974	5 I	116.4	491	1	39816.1		2009-01-03	01:21:51
	9 I	1	1 39.974	5 I	116.4	490	1	39816.1		2009-01-03	01:21:56
	10 I	1	1 39.974	5 1	116.4	489		39816.1	ī	2009-01-03	01:22:01

• QUERIES

As can be seen from the resulting times of the first seven queries, the relational database is very convenient when executing very simple queries involving a few joins of different tables or counting or displaying data organised in different ways. In fact, the times are practically less than a few seconds, with queries executed almost instantaneously. The relational model is very powerful in this respect, since the DBMS allows a high degree of optimisation of the search for data in the tables.

```
Task 2.1: How many users, activities and trackpoints are there in the dataset? (Executed in 1.33 seconds)
Number of users: 182
Number of activities: 16046
Number of trackpoints: 9676756
Task 2.2: Find the average number of activities per user. (Executed in 0.00 seconds)
Average number of activities per user: 92.75
Task 2.3: Find the top 20 users with the highest number of activities. (Executed in 0.00 seconds)
                Activity Count |
    User ID |
                           2102
        128 I
        153 I
                           1793 I
        025 I
                            715 I
                            704
691
        163 I
        062 I
        144
                            563
        041 I
                            399
        085
                            364
                            346 I
        004
        140
                            345 I
        167 I
                            320 I
        068
                            280
        017
                            265 I
        003
                            261
        014 I
                            236
        126
                            215
        030 I
                            210
        112
                            208
                            201
        011 I
        039 I
                            198
```

Task 2.7: Find the total distance (in km) walked in 2008, by user with id=112. (Executed in 0.06 seconds) Total distance walked in 2008 by user 112: 115.47 km

As for more complex queries that, for example, require **several joins** between different tables (*including the TrackPoint table that contains more than nine million data*), we can see from the times that the **optimisation of the DBMS** manages to cut the time down to a certain point. In fact, some queries take up to **one minute** to execute.

Task 2.8: Find the top 20 users who have gained the most altitude meters. (Executed in 37.88 seconds) User ID | Total Altitude Gain (kilometers) | 128 I 2135.03 | 1820.58 | 1089.36 | 789.622 | 766.613 | 714.041 | 673.296 | 595.932 | 587.632 | 576.377 | 481.311 | 430.319 | 153 I 004 | 041 | 003 I 085 | 163 | 062 | 144 | 030 I 039 I 084 I 000 I 430.319 | 398.638 | 377.503 | 370.646 I 357.995 | 325.561 | 037 I 311.073 | 272.374 | 140 I 126 I 017 I 205.265 I

Task 2.9: Find all users who have invalid activities. (Executed in 68.07 seconds) User ID | Invalid Activity Count | 128 I 153 I 557 I 025 I 263 I 249 I 062 I 163 I 233 004 I 219 I 041 I 201 I 085 I 184 I 003 I 179 I 144 I 157 I 039 147 I 068 I 139 I 167 I 134 I 129 I 017 I 014 I 118 I 030 112 I 105 I 126 I 000 I 101 I 092 I 101 I 100 I 037 l 084 I 99 I 002 I 98 I 104 I 97 I 034 I 88 I 86 | 140 I

The **last query** is a good way to cross-check the **correctness** of the data in the relational schema, since the result of the query is the data contained in the file "*labeled_ids.txt*". In addition, the query is executed via a **sub-query**, demonstrating the **efficiency** of the relational schema and the DBMS to execute several **optimised queries** and merge the data.

1.1.3 Discussion

As far as the **execution** and **writing of the code** is concerned, the **text of the assignment** is very clear and helps a lot by guiding step by step through the various points to be executed. I did not adopt any methods other than those suggested, as I was comfortable with them due to my experience with

both Python and relational databases, particularly MySQL, even with very complex queries. Some critical points of this code are certainly, as already mentioned, the complex queries, especially those using window functions such as ROW_NUMBER(), which can be slow on large datasets. Likewise, the insertion of more than **nine million records** into the TrackPoint table is very slow and memory-intensive. Although batch insertion helps, it may not be sufficient for datasets of this size. A non-relational schema might help with some of these critical points, but it also presents challenges. There would certainly be more data flexibility: formats such as XML and **JSON** allow for more flexibility in data structure, which can be useful for unstructured or semi-structured data. In addition, inserting large amounts of data into a **NoSQL** database (such as MongoDB) can be faster than a relational database, especially if the data is already in JSON format. On the other hand, complex queries, such as those using window functions, can be more difficult to execute in a NoSQL database. Aggregation and sorting operations may be less efficient, and above all, many NoSQL databases do not support ACID transactions, which can complicate the handling of operations that need to be **atomic**. In my academic experience, I had never worked with a dataset of this size in an SQL database (I have worked on large datasets, but always imported them into memory as dataframes in Python). This assignment certainly helped me to better appreciate the ability of such widely used systems, like SQL DBMSs, to perform complex queries and operations in a short time thanks to various optimisations. However, I also better understood the limitations of these DBMSs, and perhaps found alternative methods of addressing these problems (such as batch data entry into the DBMS).