Mathematisch-Naturwissenschaftliche Fakultät

Wilhelm-Schickard-Institut für Informatik

Datenbanksysteme · Prof. Dr. Grust





Datenbanksysteme I

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Assignment #3

Submission Deadline: November 17, 2021 - 10:00

Exercise 1: Declarativity and Query Execution Strategies

(15 Points)

Database queries are written in a *declarative manner*: The user specifies "what" the desired result is, but not "how" to compute it. But before we delve into this declarative world of SQL and let the query processor of PostgreSQL do the job, we are going to explore the effects that different – but semantically equivalent – execution strategies can have on query runtime.

We will query the LEGO dataset (contains.csv, bricks.csv, minifigs.csv and sets.csv) which should be familiar from the lecture and are provided to you in the legosets.zip archive.

1. The following naive PyQL query returns all minifigs of category '67' that are part of sets '671-1' or '377-1'. However the query does not perform well:

Implement an alternative version that is as efficient as possible by exploiting properties of the data and the query itself, such as:

Order of rows, early filtering of data, uniqueness or disjointness of values in columns, query strategy, etc.

(Keep in mind that not all of them may apply here).

Note: As in the lecture, you may use the loop control keywords break (and continue) or for...else.

Exercise 2: Query Optimization through Data Pre-Processing

(15 Points)

In the lecture, we discussed multiple variations of pre-processing the data to further improve the efficiency of the "weight of LEGO set 5610" query. Another such strategy was proposed by a fellow student in a previous course.

The idea is simple: A piece referenced in contains.csv may be found in either bricks.csv or minifigs.csv. To look up a piece's details, we have to search both files since we can't tell from a piece identifier whether it refers to a brick or a minifig. However, this knowledge could be explicitly represented in the data by extending contains.csv with an additional column type: For each entry in contains.csv, type indicates whether the piece is a brick ('b') or minifig ('m').

- 1. Construct a new file contains_type.csv¹, which is a copy of contains.csv but with an additional column type as described above. Approach this problem in two steps:
 - (a) Write a PyQL query that constructs the rows to be placed in **contains_type.csv**. In this first step, the query will generate a Python list 1 of dictionaries.
 - Hint: A naive query would probably take far too long (\sim several hours). Instead, try to optimize your PyQL query. Consider that minifigs.csv is a small table compared to bricks.csv (\sim 17% of all pieces). We also know that if a piece is not a minifigure, it is a brick. Further, pieces are unique and minifigs.csv is sorted by its column piece.

Note: Even optimized your query may take some minutes to process. So think carefully about possible improvements and be patient while running your query.

(b) Use code from DB1.py to construct a Table object from the list 1 and dump it to a file in the usual CSV format.

Consider the following example:

```
1 = [{'a': 23, 'b': 'foo'}, {'a': 42, 'b': 'bar'}]
Table(1).dump('foo.csv')
```

- 2. Implement a PyQL query leveraging the new type column. Compare the query execution times for the lecture's baseline query (slide 12 in slide set db1-03.pdf) and your new optimized variant.
- 3. Provide a formula that estimates the work performed by the optimized PyQL query. (For reference, see slide 13 and 15 in slide set db1-03.pdf about *Declarativity*.)

¹This file has also been provided to you in the legosets.zip archive