Assignment 1: Single-threaded, In-memory L-Store
CS 451/551 - Fall 2024
Deadline: October 29, 2024 at 11:59pm

In the first assignment, you will gain a broad understanding of the basics of building a relational database management system.

There are three key objectives of this assignment:
(S1) Data Model: to store the schema and instance for each table in columnar form.
(S2) Bufferpool Management: to maintain data in memory.
(S3) Query Interface: to offer data manipulation and querying capabilities such as select, insert, update, and delete of a single key along with a simple aggregation query, namely, to return the summation of a single column for a range of keys.

The overall goal of this assignment is to create a single-threaded, in-memory database based on L-Store, capable of performing simple SQL-like operations.

To improve performance, you may consider adding indexing support by employing hash tables or trees. To receive an outstanding grade (A+) besides basic requirements, you are encouraged to add indexing functionality (including both the primary and/or secondary indexes), experimental analysis with graphs (see the L-Store paper), and/or extended query capabilities.

Bonus: Kindly note that the fastest L-Store implementations (the top three groups) will be rewarded. You may also earn bonus points for creative design by improving upon L-Store.

Overall, each group may receive up to a **10% bonus**.

Introduction
To derive real-time actionable insights from the data, it is important to bridge the gap between managing the data that is being updated (write) at a high velocity and analyzing a large volume of data (read). However, there has been a divide where specialized solutions were often deployed to support either read-intensive or write-intensive workloads but not both, thus, limiting the analysis to stale and possibly irrelevant data.

Lineage-based Data Store (L-Store) is a solution that combines the real-time processing of transactional and analytical workloads within a single unified engine by introducing a novel update-friendly lineage-based storage architecture. By exploiting the lineage, we will develop a contention-free and lazy staging of columnar data from a write-optimized (tail data) form into a read-optimized (base data) form in a transactionally consistent approach that supports querying and retaining current and historical data. During this course, we will develop a stripped-down version of L-Store through three assignments. For the first assignment, we will focus on a simplified in-memory (volatile) implementation that provides basic relational data storage and querying capabilities. In the second assignment, we will focus on data durability by persisting data

on a disk (non-volatile) and merging the base and tail data. The third assignment will focus on concurrency and multi-threaded transaction processing.

L-Store Fundamentals

L-Store is a relational database. Simply put, data is stored in a table form consisting of rows and columns. Each row of a table is a record (also called a tuple), and the columns hold the attributes of each record. Each record is identified by a unique primary key that can be referenced by other records to form relationships through foreign keys.

(S1) Data Model:

The key idea of L-Store is to separate the original version of a record inserted into the database (a base record) and the subsequent updates to it (tail records). Records are stored in physical pages where a page is basically a fixed-size contiguous memory chunk, say 4 KB (you may experiment with larger page sizes and observe its effects on the performance). The base records are stored in read-only pages called base pages. Each base page is associated with a set of append-only pages called tail pages that will hold the corresponding tail records, namely, any updates to a record will be added to the tail pages and will be maintained as a tail record. We will generalize how we associate base and tail pages when we discuss page ranges.

Data storage in L-Store is columnar, meaning that instead of storing all fields of a record contiguously, data from different records for the same column are stored together. Each page is dedicated to a certain column. The idea behind this layout is that most update and read operations will only affect a small set of columns; hence, separating the storage for each column would reduce the amount of contention and I/O. Also, the data in each column tends to be homogeneous, and the data on each page can be compressed more effectively. As a result, the base page (or tail page) is a logical concept because, physically, each base page (or tail page) consists of a set of physical pages (4K each, for example), one for each column.

In the database, each record is assigned a unique identifier called a RID, which is often the physical location of where the record is actually stored. In L-Store, this identifier will never change during a record's lifecycle. Each record also includes an indirection column that points to the latest tail record holding the latest update to the record. When updating a record, a new tail record is inserted in the corresponding tail pages, and the indirection column of the base record is set to point to the RID of this new tail record. The tail record's own indirection is set to point to the RID of the previous tail record (the previous update) for the same base record if available.

Tail records can be either cumulative or non-cumulative. A cumulative tail record will contain the latest updated values for all columns while a non-cumulative one only includes values of the updated columns and sets the values of other columns to a special NULL value. The choice between cumulative or non-cumulative updates offers a trade-off between update and read performance. For non-cumulative updates, the whole lineage needs (past updates) to be traversed to get the latest values for all columns. This design might seem inefficient in the sense that it needs to read multiple records to yield all columns; however, in practice, the entire lineage may rarely be traversed as most queries need specific columns. In your implementation, you may choose either option, or you may experiment with both options and quantify the difference. In order to see a difference, you need to insert/update many records, perhaps up to a few million records.

Each base record also contains a schema encoding column. This is a bit vector with one bit per column that stores information about the updated state of each column. In the base records, the schema encoding will include a 0 bit for all the columns that have not yet been updated and a 1 bit for those that have been updated. This helps optimize queries by determining whether we need to follow the lineage or not. In non-cumulative tail records, the schema encoding serves to distinguish between columns with updated values and those with NULL values.

L-Store also supports deleting records. When a record is deleted, the base record will be invalidated by setting the RID of itself and all its tail records to a special value. Other possible implementation would be to add a special delete flag to the indirection field of the base record, indicating a logical delete. These invalidated records will be removed during the next merge cycle for the corresponding page range. The invalidation part needs to be implemented during this milestone. The removal of invalidated records will be implemented in the merge routine of the next milestone. Records are virtually partitioned into disjoint page ranges. Each page range consists of a set of base pages. For example, each page range may contain 64K records while each base page holds 4K records; thus, 16 base pages. Each base page (or tail page) itself consists of a set of physical pages (4 KB each), one for each column. Each page range also consists of a set of tail pages. Thus, tail pages are the granularity of the page range, not base pages. Suppose the page range consists of 16 base pages. We initially started with a single tail page. Any updates to these 16 pages are appended to this tail page. Once the tail page is filled, we allocate a new tail page. The tail pages and base pages within each range are periodically merged to yield a set of new base pages that contain all the latest values for each column. This prevents the need for consulting tail pages when answering queries. The merge will be discussed in Milestone 2.

(S2) Bufferpool Management:
In this assignment, we have a simplified bufferpool because data resides only in memory and is not backed by disk. To keep track of data, whether, in memory (or disk), we require to have a page directory that maps RIDs to pages in memory (or disk) to allow fast retrieval of records. Recall records are stored in pages, and records are partitioned across page ranges. Given a RID, the page directory returns the location of a certain record inside the page within the page range. The efficiency of this data structure is a key factor in performance.

(S3) Query Interface:
We will require simple query capabilities in this milestone that provides standard SQL-like functionalities, which are also similar to Key-Value Stores (NoSQL). For this assignment, you need to provide select, insert, update, and delete of a single key along with a simple aggregation query, namely, to return the summation of a single column for a range of keys.

Implementation
We have provided a **code skeleton** that can be used as a baseline for developing your version. Our implementation followed the figures described above. This skeleton is merely a suggestion, and you are free and **even encouraged** to come up with your own design.

You will find three main classes in the provided skeleton. Some of the needed methods in each class are provided as stubs. But you must implement the APIs listed in **db.py, query.py, table.py,**

and **index.py**; you also need to ensure that you can run **main.py** to allow auto-grading as well (which accounts for **50% of the overall grade**). We have provided several such methods to guide you through the implementation.

The Database class is a general interface to the database and handles high-level operations such as starting and shutting down the database instance and loading the database from stored disk files. This class also handles the creation and deletion of tables via the create and drop function. The create function will create a new table in the database. The Table constructor takes as input the name of the table, the number of columns, and the index of the key column. The drop function drops the specified table.

The Query class provides standard SQL operations such as insert, select, update, delete, and sum. The select function returns all the records matching the search key (if any), and only the projected columns of the matching records are returned. The insert function will insert a new record in the table. All columns should be passed a non-NULL value when inserting. The update function updates values for the specified set of columns. The delete function will delete the record with the specified key from the table. The sum function will sum over the values of the selected column for a range of records specified by their key values. We query tables by direct function calls rather than parsing SQL queries.

The Table class provides the core of our relational storage functionality. All columns are 64-bit integers in this implementation. Users mainly interact with tables through queries. Tables provide a logical view of the actual physically stored data and mostly manage the storage and retrieval of data. Each table is responsible for managing its pages and requires an internal page directory that, given a RID, returns the actual physical location of the record. The table class should also manage the periodical merge of its corresponding page ranges.

The Index class provides a data structure that allows fast processing of queries (e.g., select or update) by indexing columns of tables over their values. Given a certain value for a column, the index should efficiently locate all records having that value. The key column of all tables is required to be indexed by default for performance reasons. However, supporting secondary indexes is optional for this milestone. The API for this class exposes the two functions create_index and drop_index (optional for this milestone).

The Page class provides low-level physical storage capabilities. In the provided skeleton, each page has a fixed size of 4096 KB. This should provide optimal performance when persisting to disk, as most hard drives have blocks of the same size. You can experiment with different sizes. This class is mostly used internally by the Table class to store and retrieve records. While working with this class, keep in mind that tail and base pages should be identical from the hardware's point of view. The config.py file is meant to act as centralized storage for all the configuration options and the constant values used in the code. It is good practice to organize such information into a Singleton object accessible from every file in the project. This class will find more use when implementing persistence in the next milestone.

Self-Testing Scripts:

You will also observe that there are several scripts in the repository, namely, m1/m2/m3_tester.py and exam_tester_m1/m2/m3.py. Those scripts are for your self-debugging and testing. Here, m1 refers to the assignment 1, m2 is for assignment 2, and m3 is for assignment 3.

Assignment Deliverables/Grading Scheme: What to submit?

At the end of this assignment, each team needs to prepare a presentation that concisely summarizes the entire progress, including the data model, bufferpool, and query API, followed by a live demo of your L-Store implementation. Your submission should have working create, insert, select, update, delete, and sum methods and correct implementation of L-Store fundamentals such as base and tail pages and records. Further, your submission should successfully run and pass main.py; otherwise, a grade of zero will be received on the auto-grading component of the assignment. You will need to submit the presentation slides in .pptx, .key, or .pdf format by the due date. The submission is made through Canvas, and only one group member must submit the package on behalf of the entire group.

The actual presentation and evaluation will be scheduled after the assignment due date.