# Efficient Sorting, Duplicate Removal, Grouping, and Aggregation (2022)

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# What is this paper about?

Database query processing requires algorithms for duplicate removal, grouping, and aggregation.

In SQL you can chain operations together in a single query. Based on earlier operations, later operations may be faster based on groupings/sorts from the output of the previous operation.

This paper focuses on an algorithm that performs well on run-time efficiency in both cases.

#### Applications of these ideas

SQL duplicate removal are used in applications of all kinds:

- The example given in the paper is removing duplicate users from a website log. This can reduce billions of rows to millions of rows.
- In my UROP, different parts of the genome may have same effects on the upstream and downstream regulators

#### **Duplicate Removal**

- SELECT DISTINCT Color FROM Foods
- Returns unique values

bod	Color
wberry	Red
nach	Green
spberry	Red
ettuce	Green

### Grouping and Aggregation

- SELECT Color, Count(\*) FROM Foods GROUP BY Color
- Creates group objects where data is sorted into contiguous chunks, and then applies aggregation function that calculates value based off group

Food	Color		
Strawberry	Red	Color	Count
Spinach	Green	 Red	2
Raspberry	Red	Green	2
Lettuce	Green	 	I

# **Existing Implementations**

- <u>In-stream aggregation</u> - This is the most efficient implementation, but it requires sorted input

If the input is not sorted:

- <u>Sort-based aggregation</u> uses an external memory merge sort. Relatively slower and uses more memory, but good if sorted order can help with subsequent operations
- <u>Hash-based aggregation</u> uses in-memory hash table + hash partitioning to temporary storage. Aggregates with hash buckets, this is good when the hash table (the output of the aggregation), can fit into memory.
- Uses estimation in compile time to determine whether to use sort or hash based aggregation in this step

#### What is the problem with this?

- The estimations for input and output size isn't reliable and is often inaccurate during query compilation
- Because of this, the wrong algorithm choice is often made, causing unpredictable performance and slower runtimes
- Having unpredictable performance allows developers to better account for performance delays, which can help make better user interfaces

#### **Contribution of Paper**

- This paper aims to create a new in-sort aggregation algorithm to replace the sort-based and hash-based aggregation algorithms with a single algorithm.
- The goal is to make it competitive with the hash based algorithm, while creating a sorted order that will help with previous runs.
- Now, instead of having to rely on erroneous estimates, database engines can rely on a single algorithm



Fig. 3. Motivating performance example from Google's F1 Query.

#### **Benefits**

- Less production code means less maintenance and more readability and understandability
- Reduces unpredictable performance
- Easier to understand how to optimize queries
- Easier to analyze runtime and bottlenecks in specific queries

#### Table 1. Traditional Decision Procedure

Condition	Query Optimization Choice	
Sorted Input?	In-stream aggregation	
Output < Memory	- Hash aggregation	
Unsorted output ok?		
Input/Output < fan-in	Traditional in-sort aggregation	
Otherwise	Hash aggregation + sort	

Table 2. Decision Procedure with theNew Algorithm

Condition	Query Optimization Choice		
Sorted input?	In-stream aggregation		
Otherwise	New in-sort aggregation		

#### How can we optimize these operations?

- We can't change the input or output size of the data, but we can control how much temporary storage we use in the grouping and aggregation of data to achieve the output.
- This is what the paper focuses on optimizing



Fig. 1. Optimization opportunity in sorting and grouping.

#### Introducing the Implementation

- This paper introduces two new techniques. They give sorting the property of early aggregation by using in memory binary tree indexes, and using wide merging to reduce the amount of temporary storage needed.
- Reducing memory is the name of the game.

# **Technique 1: Early Aggregation**

- It allows for aggregation in sorting data structures by using a binary tree (rather than priority queues) that can eliminate duplicate values in the primary runs to temporary storage (during sorting) rather than after the merge. This uses less temporary memory.



In-stream aggregation

Fig. 6. Run generation using an ordered in-memory index.

Fig. 2. In-stream duplicate removal after sort (red) vs. in-sort duplicate removal within runs (blue).

# Technique 2: Wide Merging

- The second new technique, wide merging, improves the final merge step of external merge sort by having a larger fan-in.
- Rather than using a memory page for each input, wide merging uses a single page for all inputs. It absorbs inputs from others into a single page.
- Reduces the amount of memory necessary for merge.



#### Performance - CPU Time

- CPU Time - Speed of queries



Fig. 15. Performance of in-memory indexes.

#### Performance - Spillage

- Spillage - amount of external temporary storage used. We want to reduce the blow-up factor of this if output is greater than the size of memory. Sometimes it's unavoidable



## Conclusions

- This paper introduces a new that is competitive with the hash based algorithm, while creating a sorted order that will help with subsequent runs.
- Database engines can rely on a single more reliable algorithm
- Less production code means less maintenance and more readability, and understandability

#### Pros

- Very thorough paper with many theoretical and experimental justifications
- Competitive with both previous approaches with sorting benefit

### Cons

- More figures to explain how it works would be useful
- Some number dense paragraphs in paper are not as readable

#### Thank you! Questions