

LECTURE #11 >> 15-445/645 FALL 2025 >> PROF. ANDY PAVLO

ADMINISTRIVIA



Homework #3 is due Sunday Oct 5th @ 11:59pm

Mid-Term Exam is on Wednesday Oct 8th

- \rightarrow Lectures #01–11 (inclusive)
- \rightarrow Study guide is now available (<u>@126</u>)
- → Bring your CMU ID!

Project #2 is due Sunday Oct 26th @ 11:59pm

COURSE OUTLINE



We are now going to talk about how to execute queries using the DBMS components we have discussed so far.

Next four lectures:

- → Operator Algorithms
- → Query Processing Models
- → Runtime Architectures

Query Planning

Operator Execution

Access Methods

Buffer Pool Manager

Disk Manager

QUERY PLAN



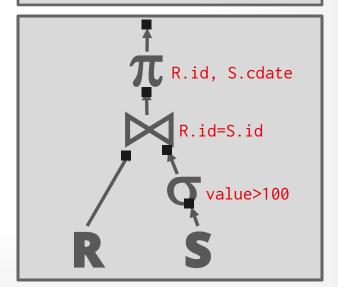
The operators are arranged in a tree.

Data flows from the leaves of the tree up towards the root.

→ We will discuss the granularity of the data movement next week.

The output of the root node is the result of the query.

```
SELECT R.id, S.cdate
FROM R JOIN S
ON R.id = S.id
WHERE S.value > 100
```



DISK-ORIENTED DBMS



Just like it cannot assume that a table fits entirely in memory, a disk-oriented DBMS cannot assume that query results fit in memory.

We will use the buffer pool to implement algorithms that need to spill to disk.

We are also going to prefer algorithms that maximize the amount of sequential I/O.

WHY DO WE NEED TO SORT?



Relational model/SQL is <u>unsorted</u>.

Queries may request that tuples are sorted in a specific way (ORDER BY).

But even if a query does not specify an order, we may still want to sort to do other things:

- \rightarrow Trivial to support duplicate elimination (**DISTINCT**).
- \rightarrow Bulk loading sorted tuples into a B+Tree index is faster.
- → Aggregations (**GROUP BY**).

IN-MEMORY SORTING



If data fits in memory, then the DBMS can use a standard sorting algorithm.

- \rightarrow Optimized algorithms if data is mostly sorted (<u>VergeSort</u>).
- → Otherwise use your favorite (QuickSort, TimSort, RadixSort).
- → Many online <u>visualization tools</u>.

If data does not fit in memory, then we need to use a technique that is aware of the cost of reading and writing disk pages ...

SORTING ALGORITHMS



For a given input <u>run</u> (i.e., list of key/value pairs), sort it based on a comparison function and sorting parameters.

Key: The attribute(s) to compare to compute the sort order.

Value: Two choices

- \rightarrow Tuple (<u>early materialization</u>).
- → Record ID (**late materialization**).

Early Materialization

K ₁	<tuple data=""></tuple>
K ₂	<tuple data=""></tuple>

•

Late Materialization



SORTING ALGORITHMS

S 8

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Value: Two choices

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- → Record ID (**late materialization**).



TODAY'S AGENDA

Sall Sall

Top-N Heap Sort

External Merge Sort

Aggregations

5DB Flash Talk: MotherDuck

TODAY'S A DuckDB

Top-N Heap Sort

External Merge Sort

Aggregations

5DB Flash Talk: Mother Dug

Redesigning DuckDB's Sort, Again



Laurens Kuiper 2025-09-24 · 16 min

TL;DR: After four years, we've decided to redesign DuckDB's sort implementation, again. In this post, we present and evaluate the new design.

DuckDB v1.4.0 was just released, which includes a complete redesign of DuckDB's sort implementation. We redesigned DuckDB's sort just four years ago, which allowed DuckDB to sort more data than fits in main memory, in parallel, with highly efficient comparisons. This implementation served us well, but since then we've implemented larger-than-memory query processing for more operators, such as the hash join \(\mathbb{Z} \) and hash aggregation \(\mathbb{Z} \), which both use a new and improved spillable page layout \(\mathbb{Z} \). We presented this layout in an earlier blog post. We decided to integrate this layout in DuckDB's sort, and completely redesigned the implementation \(\mathbb{Z} \).

Not interested in the implementation? Jump straight to the benchmark!

Two-Phase Sorting

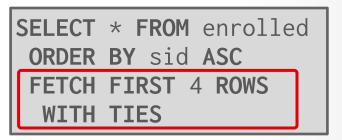
DuckDB implements parallel query execution using Morsel-Driven Parallelism 2.



If a query contains an **ORDER BY** with a **LIMIT**, then the DBMS only needs to scan the data once to find the top-*N* elements.

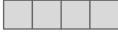
Ideal scenario for <u>HeapSort</u> if the top-N elements fit in memory.

→ Scan data once, maintain an in-memory sorted priority queue.



Original Data



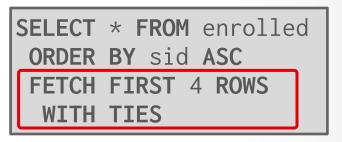




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Original Data





Sorted Heap

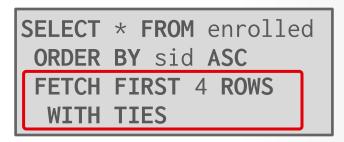
3



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Sorted Heap

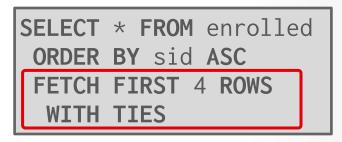
3 4



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Sorted Heap

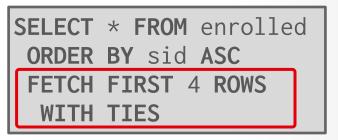
3 4 6



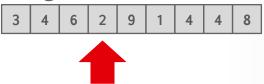
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Sorted Heap

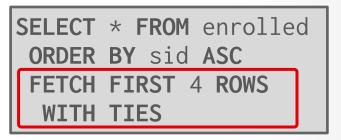
3 4 6

TOP-N HEAP SORT

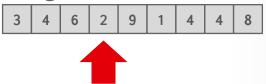
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Original Data



Sorted Heap

2 3 4 6

TOP-N HEAP SORT

If a query contains an **ORDER BY** with a **LIMIT**, then the DBMS only needs to scan the data once to find the top-*N* elements.

Ideal scenario for <u>HeapSort</u> if the top-N elements fit in memory.

→ Scan data once, maintain an in-memory sorted priority queue.

SELECT * FROM enrolled
ORDER BY sid ASC
FETCH FIRST 4 ROWS
WITH TIES

Original Data





Skip!

Sorted Heap

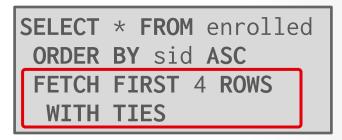
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Original Data



Sorted Heap

1 2 3 4

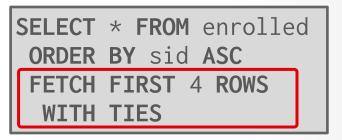
10 3.4

TOP-N HEAP SORT

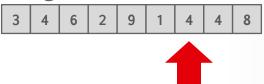
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Original Data



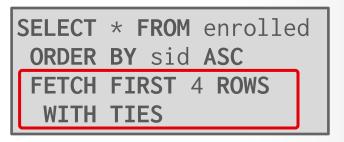
1 2 3 4 4 4	



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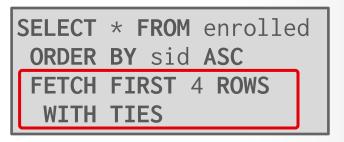
1 2 3 4 4 4		4	4	4	3	2	1		



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Ideal scenario for <u>HeapSort</u> if the top-*N* elements fit in memory.

→ Scan data once, maintain an in-memory sorted priority queue.



Original Data





Skip and done!



EXTERNAL MERGE SORT



Divide-and-conquer algorithm that splits data into separate <u>runs</u>, sorts them individually, and then combines them into longer sorted runs.

Phase #1 – Sorting

- → Sort chunks of data that fit in memory and then write back the sorted chunks to a file on disk.
- → Pick your favorite in-memory sorting algorithm.

Phase #2 – Merging

→ Combine sorted runs into larger chunks.

2-WAY EXTERNAL MERGE SORT



We will start with a simple example of a 2-way external merge sort.

 \rightarrow "2" is the number of runs to merge into a new run for each pass.

Data is broken up into *N* pages.

The DBMS has a finite number of **B** buffer pool pages to hold input and output data.

SIMPLIFIED 2-WAY EXTERNAL MERGE SORT

Pass #0

- → Read one page of the table into memory
- → Sort page into a "run" and write it back to disk
- → Repeat until the whole table has been sorted into runs

Pass #1,2,3,...

- → Recursively merge pairs of runs into runs twice as long
- → Need at least 3 buffer pages (2 for input, 1 for output)



In each pass, the DBMS reads and writes every page in the file.

Pass #0

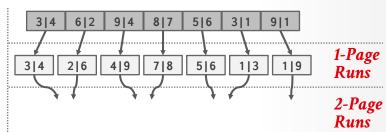
Pass #1

Number of passes

$$=1+\lceil \log_2 N \rceil$$

Total I/O cost

 $= 2N \cdot (\# \text{ of passes})$





In each pass, the DBMS reads and writes every page in the file.

Pass #0 Pass #1

8|7

3|1

5|6

911

1-Page Runs 2-Page Runs

Number of passes $=1+\lceil \log_2 N \rceil$ Total I/O cost

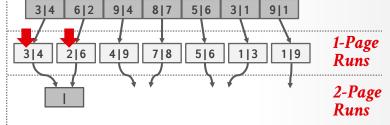
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DATABASE SYSTEMS (FALL 2025)



In each pass, the DBMS reads and writes every page in the file.

Pass #0
Pass #1



Number of passes

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Total I/O cost

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In each pass, the DBMS reads and writes every page in the file.

Pass #0
Pass #1

3|4 6|2 9|4 8|7 5|6 3|1 9|1 3|4 2|6 4|9 7|8 5|6 1|3 1|9 1-Page Runs 2-Page Runs

Number of passes

$$=1+\lceil \log_2 N \rceil$$

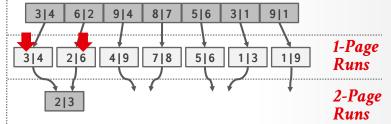
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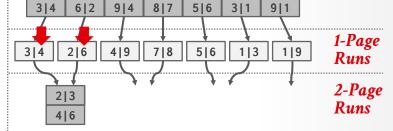
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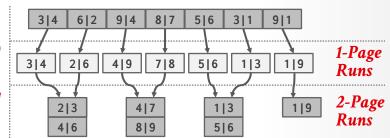
Pass #0
Pass #1

Number of passes

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Total I/O cost

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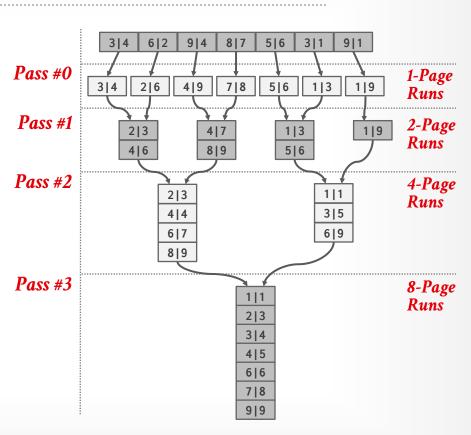
In each pass, the DBMS reads and writes every page in the file.

Number of passes

$$=1+\lceil \log_2 N \rceil$$

Total I/O cost

 $= 2N \cdot (\# \text{ of passes})$



SIMPLIFIED 2-WAY EXTERNAL MERGE SORT

This algorithm only requires three buffer pool pages to perform the sorting (B=3).

→ Two input pages, one output page

But even if we have more buffer space available (B>3), it does not effectively utilize them if the worker must block on disk I/O...

GENERAL EXTERNAL MERGE SORT

Pass #0

- \rightarrow Use **B** buffer pages
- \rightarrow Produce [N/B] sorted runs of size B

Pass #1,2,3,...

 \rightarrow Merge **B-1** runs (i.e., M-way merge)

Number of passes =
$$1 + \lceil \log_{B-1} \lceil N/B \rceil \rceil$$

Total I/O Cost = $2N \cdot (\# \text{ of passes})$

EXAMPLE

Determine how many passes it takes to sort 108 pages with 5 buffer pool pages: N=108, B=5

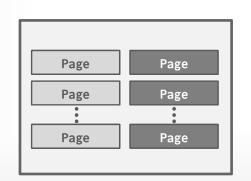
- → Pass #0: [N/B] = [108/5] = 22 sorted runs of 5 pages each (last run is only 3 pages).
- \rightarrow **Pass #1:** [N'/B-1] = [22/4] = 6 sorted runs of 20 pages each (last run is only 8 pages).
- \rightarrow **Pass #2:** [N''/B-1] = [6/4] = 2 sorted runs, first one has 80 pages and second one has 28 pages.
- \rightarrow **Pass #3:** Sorted file of 108 pages.

1+
$$\lceil \log_{B-1}[N/B] \rceil$$
 = 1+ $\lceil \log_4 22 \rceil$ = 1+ $\lceil 2.229... \rceil$ = 4 passes

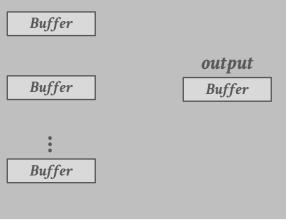
DOUBLE BUFFERING

- \rightarrow Overlap CPU and I/O operations
- \rightarrow Reduces effective buffers available by half





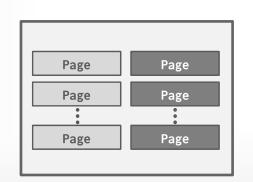




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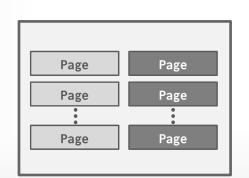


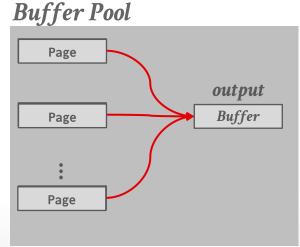


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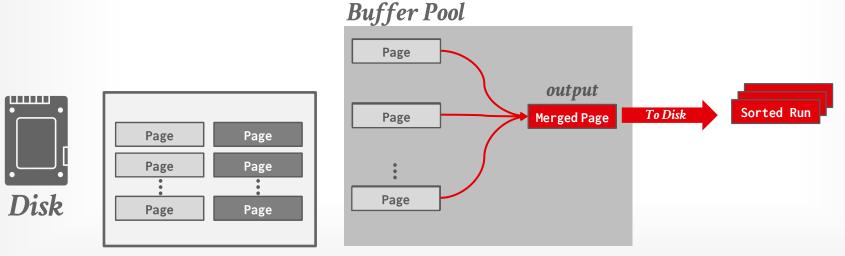






DOUBLE BUFFERING

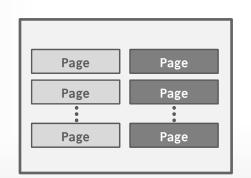
- → Overlap CPU and I/O operations
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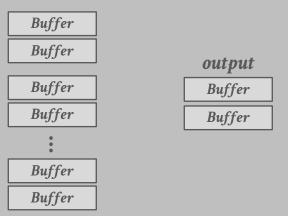
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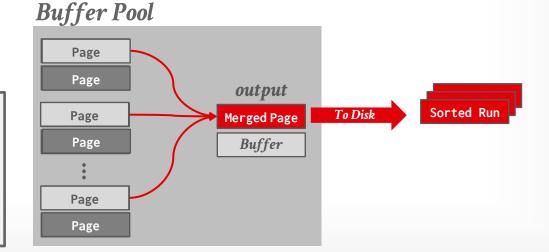




DOUBLE BUFFERING

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- → Reduces effective buffers available by half

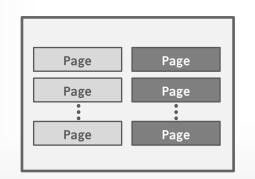


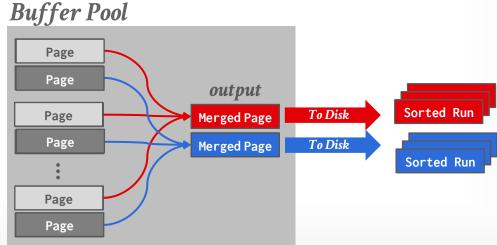


DOUBLE BUFFERING

- → Overlap CPU and I/O operations
- \rightarrow Reduces effective buffers available by half







COMPARISON OPTIMIZATIONS

Approach #1: Code Specialization

→ Instead of providing a comparison function as a pointer to sorting algorithm, create a hardcoded version of sort that is specific to a key type.

Approach #2: Suffix Truncation

→ First compare a binary prefix of long **VARCHAR** keys instead of slower string comparison. Fallback to slower version if prefixes are equal.

Approach #3: Key Normalization

→ Transform variable-length attribute(s) into a single encoded/padded string that preserves sort order.

USING B+TREES FOR SORTING

If the table that must be sorted already has a B+Tree index on the sort attribute(s), then we can use that to accelerate sorting.

→ Some DBMSs support prefix key scans for sorting.

Retrieve tuples in desired sort order by simply traversing the leaf pages of the tree.

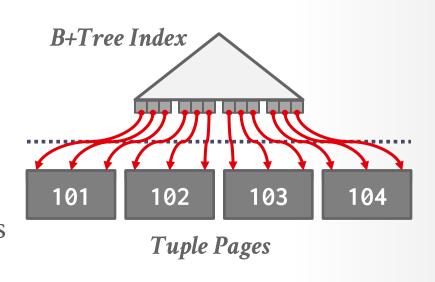
Cases to consider:

- → Clustered B+Tree
- → Unclustered B+Tree

CASE 1: CLUSTERED B+TREE

Traverse to the left-most leaf page, and then retrieve tuples from all leaf pages.

This is always better than external sorting because there is no computational cost, and all disk access is sequential.

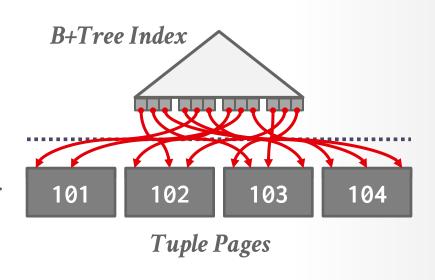


CASE #2: UNCLUSTERED B+TREE

Chase each pointer to the page that contains the data.

This is almost always a bad idea except for Top-N queries where N is small enough relative to total number of tuples in table.

→ In general, one I/O per data record.



AGGREGATIONS

Collapse values for a single attribute from multiple tuples into a single scalar value.

The DBMS needs a way to quickly find tuples with the same distinguishing attributes for grouping.

Two implementation choices:

- → Sorting
- → Hashing

SORTING AGGREGATION



SELECT DISTINCT cid
 FROM enrolled
WHERE grade IN ('B','C')
ORDER BY cid

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С



sid	cid	grade
53666	15-445	С
53688	15-826	В
53666	15-721	С
53655	15-445	С



cid
15-445
15-826
15-721
15-445



cid
15-445
15-445
15-721
15-826

SORTING AGGREGATION



SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')
ORDER BY cid

enrolled (sid, cid, grade)

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С



sid	cid	grade
53666	15-445	С
53688	15-826	В
53666	15-721	С
53655	15-445	С



cid
15-445
15-826
15-721
15-445



cid	4
15-445	
15-445	
15-721	
15-826	

Eliminate Duplicates

SORTING AGGREGATION



SELECT DISTINCT cid
 FROM enrolled
WHERE grade IN ('B','C')
ORDER BY cid

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С



sid	cid	grade
53666	15-445	С
53688	15-826	В
53666	15-721	С
53655	15-445	С



cid
15-445
15-826
15-721
15-445



	cid		
	15-445		
	15-45		
	15-721		
	15-826		
El	imina	tè	
Duplicates			

```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
 WHERE s.sid = e.sid
 GROUP BY cid
```

Running Totals

AVG(col)	>	(COUNT,	SUM)
MIN(col)	>	(MIN)	
MAX(col)	>	(MAX)	
SUM(col)	>	(SUM)	
<pre>COUNT(col)</pre>	>	(COUNT)	



cid	s.gpa
15-445	3.62
15-826	2.89
15-721	3.33
15-445	3.7



cid	s.gpa	
15-445	3.62	
15-445	3.7	
15-721	3.33	
15-826	2.89	

Previous: null (1, 3.62)

AGGREGATION SUMMARIZATION

```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

AVG(col) → (COUNT, SUM)
MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



cid	s.gpa
15-44	5 3.62
15-82	6 2.89
15-72	1 3.33
15-44	5 3.7



cid	s.gpa	
15-445	3.62	
15-445	3.7	,
15-721	3.33	
15-826	2.89	

Previous: 15-445

(1, 3.62)

AGGREGATION SUMMARIZATION

```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
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GROUP BY cid
```

Running Totals

AVG(col) → (COUNT, SUM)
MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



cid	s.gpa
15-445	3.62
15-826	2.89
15-721	3.33
15-445	3.7



cid	s.gpa	
15-445	3.62	
15-445	3.7	
15-721	3.33	
15-826	2.89	

Previous: 15-445

(2, 7.32)



```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

AVG(col) → (COUNT, SUM)

MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



cid	s.gpa
15-445	3.62
15-826	2.89
15-721	3.33
15-445	3.7



cid	s.gpa	
15-445	3.62	
15-445	3.7	
15-721	3.33	
15-826	2.89	

Previous: 15-721

(1, 3.33)

cid	AVG(gpa)
15-445	3.66



```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

AVG(col) → (COUNT, SUM)

MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



cid	s.gpa
15-445	3.62
15-826	2.89
15-721	3.33
15-445	3.7



cid	s.gpa	
15-445	3.62	
15-445	3.7	
15-721	3.33	
15-826	2.89	

Previous: 15-721

(1, 3.33)

cid	AVG(gpa)
15-445	3.66



```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

AVG(col) → (COUNT, SUM)

MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



cid	s.gpa
15-445	3.62
15-826	2.89
15-721	3.33
15-445	3.7



	cid	s.gpa	
	15-445	3.62	
	15-445	3.7	
[15-721	3.33	4
Ŀ	15-826	2.89	

Previous: 15-826

(1, 2.89)

cid	AVG(gpa)
15-445	3.66
15-721	3.33

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```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

AVG(col) → (COUNT, SUM)

MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



cid	s.gpa
15-445	3.62
15-826	2.89
15-721	3.33
15-445	3.7



cid	s.gpa
15-445	3.62
15-445	3.7
15-721	3.33
15-826	2.89



cid	AVG(gpa)
15-445	3.66
15-721	3.33
15-826	2.89



ALTERNATIVES TO SORTING

What if we do not need the data to be ordered?

- → Forming groups in **GROUP BY** (no ordering)
- → Removing duplicates in **DISTINCT** (no ordering)

Hashing is a better alternative in this scenario.

- \rightarrow Only need to remove duplicates, no need for ordering.
- \rightarrow Can be computationally cheaper than sorting.

HASHING AGGREGATE

Populate an ephemeral hash table as the DBMS scans the table. For each record, check whether there is already an entry in the hash table:

- → **DISTINCT**: Discard duplicate
- → **GROUP BY**: Perform aggregate computation

If everything fits in memory, then this is easy.

If the DBMS must spill data to disk, then we need to be smarter...

EXTERNAL HASHING AGGREGATE

Divide-and-conquer approach to computing an aggregation when data does not fit in memory.

Phase #1 – Partition

- → Split tuples into buckets based on hash key
- → Write them out to disk when they get full

Phase #2 – ReHash

→ Build in-memory hash table for each partition and compute the aggregation



PHASE #A: PARTITION

Use a hash function h_1 to split tuples into <u>partitions</u> on disk.

- → A partition is one or more pages that contain the set of keys with the same hash value.
- → Partitions are "spilled" to disk via output buffers.

Assume that we have **B** buffers.

We will use **B-1** buffers for the partitions and **1** buffer for the input data.

PHASE #1: PARTITION



SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

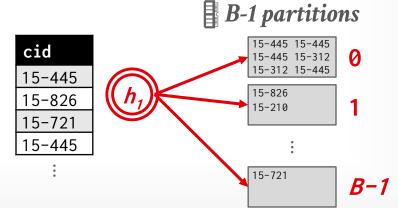
enrolled (sid, cid, grade)

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С



sid	cid	grade
53666	15-445	С
53688	15-826	В
53666	15-721	С
53655	15-445	С

Remove Columns



PHASE #1: PARTITION



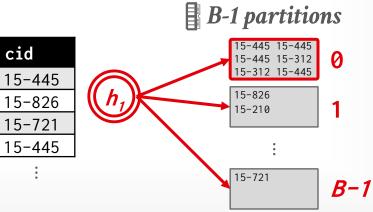
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С



sid	cid	grade
53666	15-445	С
53688	15-826	В
53666	15-721	С
53655	15-445	С







PHASE #1: PARTITION



SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

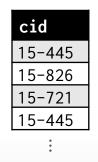
enrolled (sid, cid, grade)

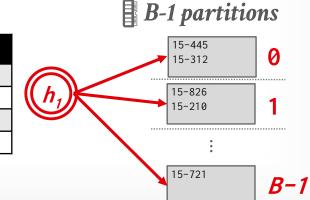
sid	cid	grade
53666	15-445	С
53688	15-721	A
53688	15-826	В
53666	15-721	С
53655	15-445	С



sid	cid	grade
53666	15-445	С
53688	15-826	В
53666	15-721	С
53655	15-445	С

Remove Columns







PHASE #2: REHASH

For each partition on disk:

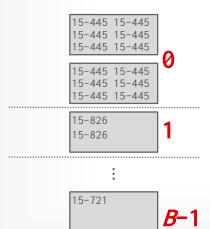
- \rightarrow Read it into memory and build an in-memory hash table based on a second hash function h_2 .
- → Then go through each bucket of this hash table to bring together matching tuples.

This assumes that each partition fits in memory.



```
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')
```

Phase #1 Buckets

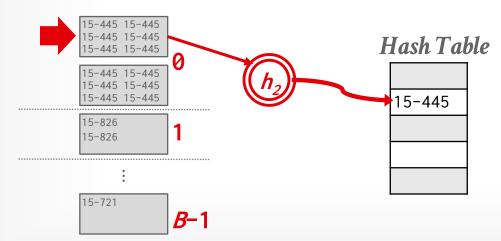


sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С
	53666 53688 53688 53666	53666 15-445 53688 15-721 53688 15-826 53666 15-721



```
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')
```

Phase #1 Buckets

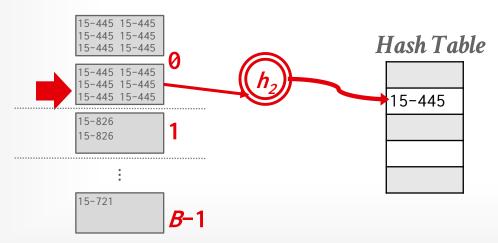


	· · · · · · · · · · · · · · · · · · ·	
sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С
	53666 53688 53688 53666	53666 15-445 53688 15-721 53688 15-826 53666 15-721



```
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')
```

Phase #1 Buckets



sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

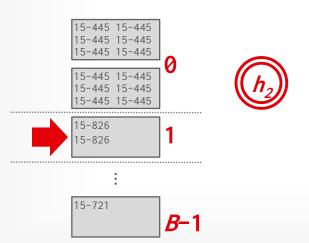


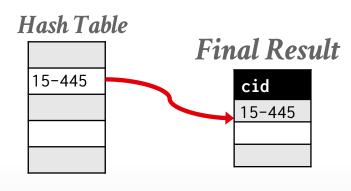
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

enrolled (sid, cid, grade)

	·	
sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

Phase #1 Buckets





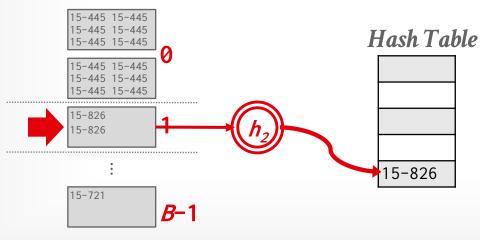


SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

enrolled (sid, cid, grade)

	· · · · · · · · · · · · · · · · · · ·	
sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

Phase #1 Buckets



cid
15-445

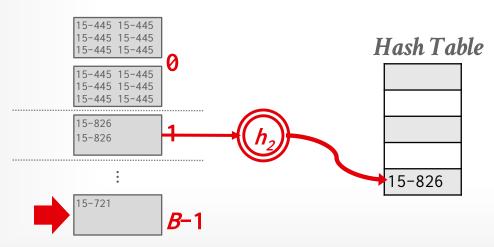


```
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')
```

enrolled (sid, cid, grade)

	· · · · · · · · · · · · · · · · · · ·	
sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

Phase #1 Buckets



cid
15-445

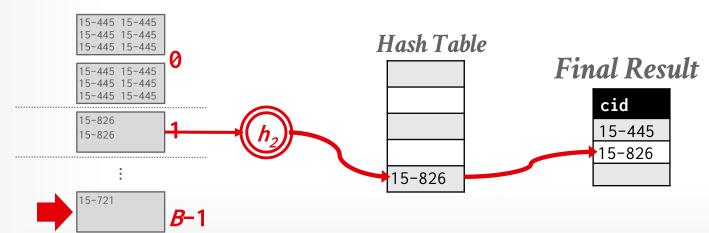


SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

enrolled (sid, cid, grade)

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

Phase #1 Buckets



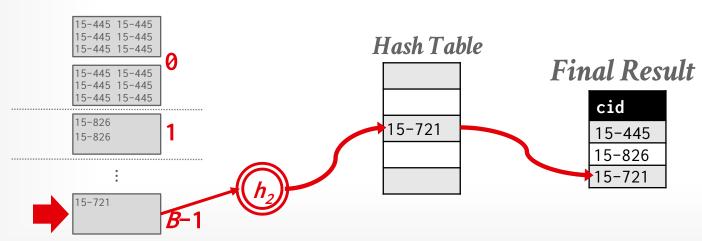


SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')

enrolled (sid, cid, grade)

sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

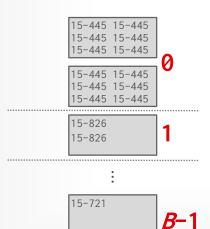
Phase #1 Buckets





```
SELECT DISTINCT cid
FROM enrolled
WHERE grade IN ('B','C')
```

Phase #1 Buckets



enrolled (sid, cid, grade)

	•	
sid	cid	grade
53666	15-445	С
53688	15-721	Α
53688	15-826	В
53666	15-721	С
53655	15-445	С

cid		
15-445		
15-826		
15-721		

HASHING SUMMARIZATION

During the rehash phase, store pairs of the form (GroupKey>RunningVal)

When we want to insert a new tuple into the hash table as we compute the aggregate:

- → If we find a matching **GroupKey**, just update the **RunningVal** appropriately
- → Else insert a new **GroupKey→RunningVal**

HASHING SUMMARIZATION



```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

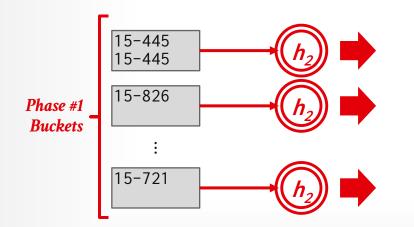
Running Totals

AVG(col) → (COUNT, SUM)
MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



Hash Table

key	value
15-445	(2, 7.32)
15-826	(1, 3.33)
15-721	(1, 2.89)

HASHING SUMMARIZATION



```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

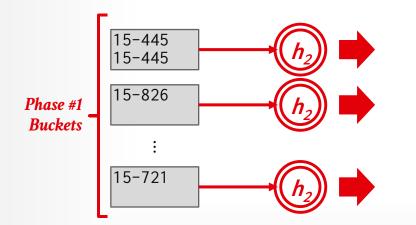
AVG(col) → (COUNT, SUM)

MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



Hash Table

key	value
15-445	(2, 7.32)
15-826	(1, 3.33)
15-721	(1, 2.89)

HASHING SUMMARIZATION



```
SELECT cid, AVG(s.gpa)
  FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
GROUP BY cid
```

Running Totals

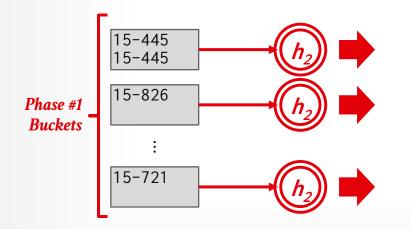
AVG(col) → (COUNT, SUM)

MIN(col) → (MIN)

MAX(col) → (MAX)

SUM(col) → (SUM)

COUNT(col) → (COUNT)



Hash Table

key	value
15-445	(2, 7.32)
15-826	(1, 3.33)
15-721	(1, 2.89)

cid	AVG(gpa)
15-445	3.66
15-826	3.33
15-721	2.89

CONCLUSION

Choice of sorting vs. hashing is subtle and depends on optimizations done in each case.

We already discussed the optimizations for sorting:

- → Chunk I/O into large blocks to amortize costs
- → Double-buffering to overlap CPU and I/O

NEXT CLASS

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Nested Loop Join Sort-Merge Join Hash Join