



Fastest Table Sort in the West - Redesigning DuckDB's Sort

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Sorting

- One of the most well-studied problems in CS
 - Cache-efficiency
 - Worst-case analysis
 - Parallelism









Sorting Relational Data

- Use cases:
 - ORDER BY
 - WINDOW
 - Sort-Merge Join
 - Inequality Join







Sorting Relational Data

- Performance challenges
 - Multiple order clauses
 - Different types
 - Columns vs. Rows

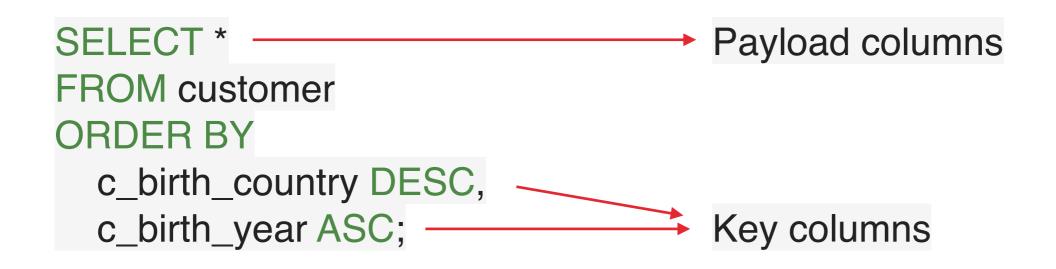






Sorting Relational Data

We distinguish two column types:



Both key and payload columns must be ordered!





The Cost of Sorting

- Dominated by
 - Comparing values
 - Moving data

This presentation: Comparing key column values
Focus: Columnar representation



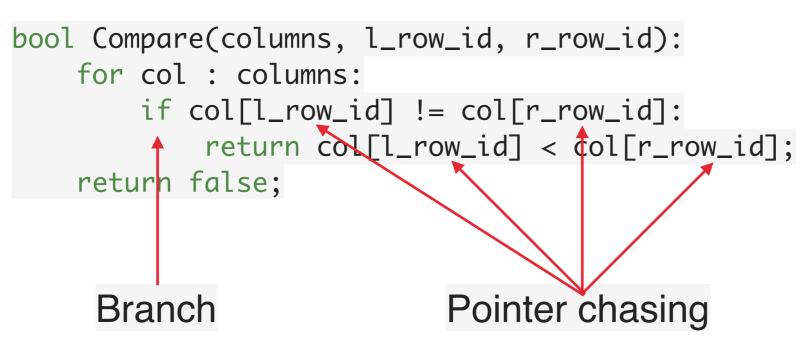
Comparing Values

DuckDB

Sorting relational data: row-wise operation

How to implement comparison for columns?

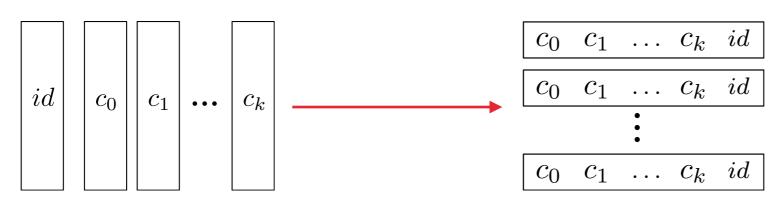
Need to sort indices:





Pointer Chasing

Solution: pack key columns into a row



No chasing pointers by index

Sort data directly

Better locality

Values in row are co-located with each other





Pointer Chasing

- Simulated experiment:
 - 2¹⁰ to 2²⁴ tuples
 - 1 to 8 key columns (uint32_t)
 - Data distribution: many ties

Measure relative runtime difference

Hardware: 2020 MacBook Pro (M1)



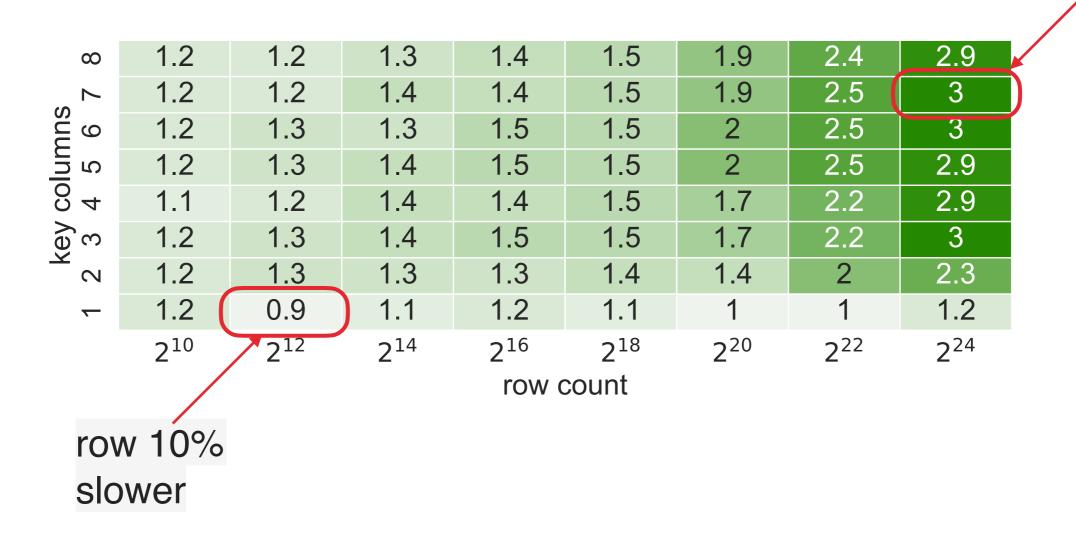


row 3x

faster

Pointer Chasing

Sorting row data vs. column data



Time includes column to row transformation!





Branch Prediction

SELECT * FROM customer ORDER BY c_birth_country DESC, c_birth_year ASC;

Comparison has many branches:

If c_birth_country equal, compare c_birth_year

ASC/DESC

NULLS FIRST/LAST



Key Normalization

Encode keys as binary string

SELECT * FROM customer ORDER BY c_birth_country DESC, c_birth_year ASC;

	birth_country	birth_year		
(a)	NETHERLANDS	1992		
	GERMANY	1924		
	birth_country		birth_year	
(b)	78 69 84 72 69 82 7	76 65 78 68 83 0	$200\ 7\ 0\ 0$	
	71 69 82 77 65 78 8	39 0	132 7 0 0	
	binary string			
(c)	177 186 171 183 18	$6\ 173\ 179\ 190\ 1$	77 187 172 25	5 128 0 7 200
	$184\ 186\ 173\ 178\ 19$	$0\ 177\ 166\ 255\ 2$	55 255 255 258	$5\ 128\ 0\ 7\ 132$

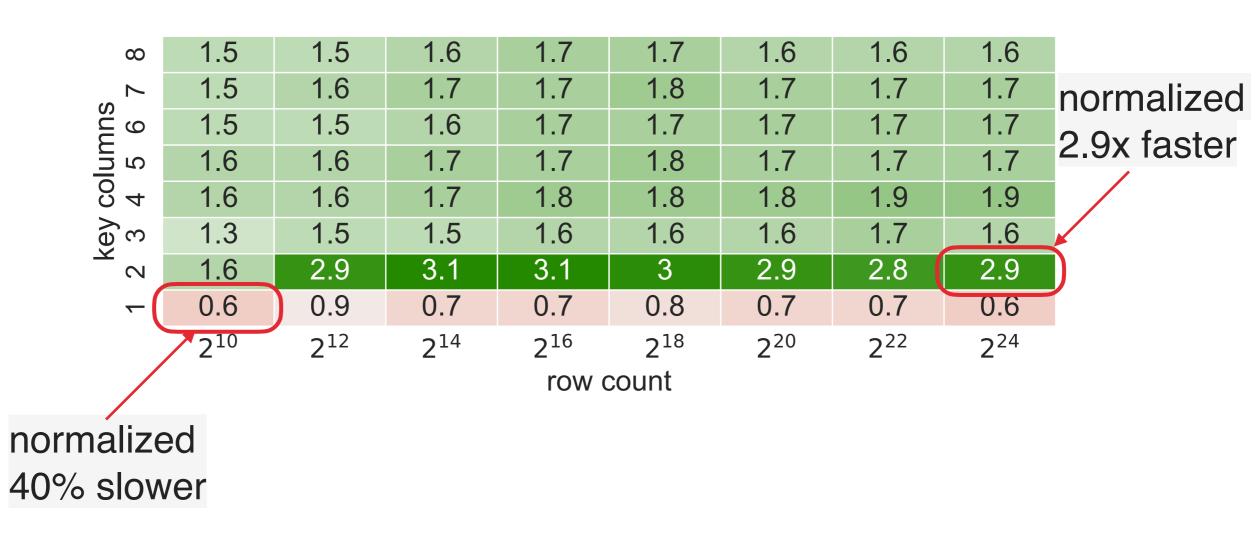
Eliminates branch predictions from comparison





Key Normalization

Sorting normalized row data vs. row data



Time includes creating the normalized keys!



Comparing Values: Summary

DuckDB

Comparing rows: Eliminates pointer chasing

Key normalization: Eliminates branch predictions

Both optimizations are almost always worth it



Sorting Algorithm

DuckDB

Key normalization enables byte-by-byte Radix Sort

► Key size ≤ 4 bytes: LSD

Key size > 4 bytes: MSD

Insertion sort





Looks pretty bad for Radix Sort ...

ω	0.5	0.5	0.6	0.5	0.6	0.7	0.6	0.7
	0.6	0.5	0.6	0.6	0.7	0.8	0.7	0.8
columns 4 5 6	0.5	0.5	0.6	0.5	0.6	0.7	0.6	0.7
5 5	0.6	0.5	0.7	0.6	0.7	0.8	0.7	0.8
	0.5	0.5	0.6	0.5	0.6	0.7	0.8	0.7
key 3	0.7	0.5	0.8	0.9	0.9	1	1	1
N ×	0.5	0.3	0.4	0.4	0.4	0.4	0.4	0.4
~	0.4	0.3	0.3	0.3	0.3	0.3	0.3	0.3
	2 ¹⁰	2 ¹²	2 ¹⁴	2 ¹⁶	2 ¹⁸	2 ²⁰	2 ²²	2 ²⁴
	row count							





- Simulation knows key size at compile time
 - In practice we don't

DuckDB

How much does pdqsort benefit from this?



DuckDB

Now with dynamic comparison:

∞	0.6	0.5	0.7	0.6	0.7	0.8	0.8	0.9
	0.7	0.5	0.7	0.6	0.7	0.9	0.8	1
columns 4 5 6	0.6	0.5	0.6	0.5	0.7	0.8	0.7	0.9
5 5	0.7	0.5	0.7	0.6	0.8	1	0.9	1
4 C	0.6	0.5	0.7	0.5	0.7	0.9	0.9	0.8
key 3	1.7	1.3	2.2	2.5	2.4	2.8	2.9	2.8
\sim	1.7	1.5	2	2.2	2.2	2.2	2.2	2.1
~	1.5	1.5	1.4	1.5	1.4	1.3	1.4	1.4
	2 ¹⁰	2 ¹²	2 ¹⁴	2 ¹⁶	2 ¹⁸	2 ²⁰	2 ²²	2 ²⁴
	row count							

Still cumbersome: many struct definitions



DuckDB

- To make quicksort efficient we could:
 - Create a lot of templated structs/functions
 - In and blow up our binary size :(
 - At this point it becomes an arms race





Sorting Algorithm: Summary

Performance depends on key size / distribution

Quicksort needs compile-time optimization
Radix Sort does not

- Efficient quicksort is possible for relational data
 - Comes at a cost

DuckDB

In or for free for JIT systems



End to end performance?

Relational sorting benchmark using TPC-DS

catalog_sales

DuckDB

- 34 columns
- SF10: 14.4M rows
- SF100: 144M rows

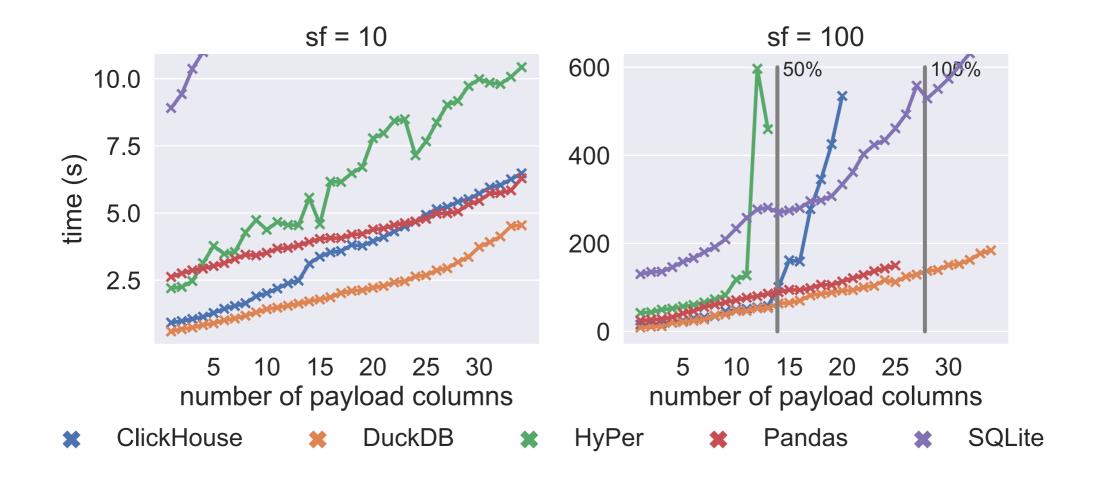
Ordered by cs_quantity, cs_item_sk





TPC-DS catalog_sales

Increasing number of payload columns



Note: 3GB/s SSD write speed



Wrapping Up

DuckDB

- Sorting relational data efficiently is challenging
- Performance is impacted by:
 - Random access
 - Branch predictions
- We can mitigate these problems with:
 - Row layout in memory
 - Key normalization
- Trade-off between Radix Sort and Quicksort





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Key Normalization

Strings:

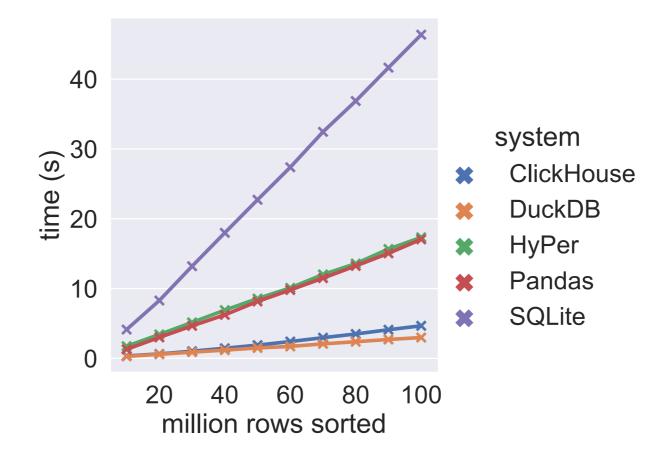
- Encode a prefix
- Collation can be encoded

- NULL values:
 - Encode using additional byte



Random Integers

10-100M random integers



In Actual relational data in the next experiment!

CWI

Parallelism

DuckDB

DuckDB uses Morsel-Driven Parallelism

Threads collect data locally

Each thread sorts its own data

Merge Sort needed for final result!





Merge Sort

Parallelized using Merge Path

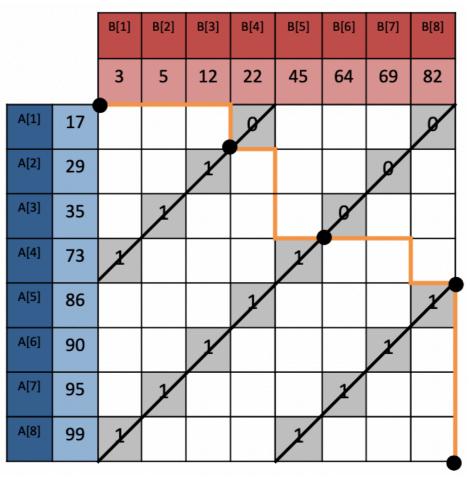


Image by Oded Green et al.





External Sorting

Serialize payload in row format

Buffer-managed blocks

	now Layout							
	pointer	intA	str	ingA	intB		str	ingB
	_•0x0001	37	/0x0(001	42		0x0	002
í	0x0003	37	/ /0x0(003	66		0x0	004 \
, /	0x0005	42	"/0x0(005	66		0x0	006 \
۸ ، ۱۸					Row He	ap		1¥ • •
۸ ۱ /	``		+\~	radi	x poin	ter		- ' //
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	``		`- `+	Duck	DBLabs	go	ose	■ -′

Row Layout





External Sorting

- Heap also uses rows
- Swizzle pointer -> offset

		<u> </u>		
offset	intA	stringA	intB	stringB
0	37	0	42	5
12	37	0	66	3
24	42	0	66	10

Row I	Heap
-------	------

radix		poin	ter
CWI	SI	wizzli	ing
DuckDBLabs			goose





External Sorting

- External sorting is made possible because of
 - merge sort
 - buffer manager
 - pointer swizzling

Modern hardware helps too!