

(Not yet Adaptive) Compression of In-Memory Databases

Database Implementation Lab Course

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- ▶ Open Source SQL OLAP RDBMS in-process developed in Amsterdam research centre CWI (SQLite for OLAP)
<https://github.com/duckdb/duckdb>
- ▶ Columnar Storage format
- ▶ Vectorized execution engine
- ▶ Has already lots of different compression possibilities for persistent data on disk

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How do we compress the transient data while having efficient lookups without decompressing everything?

Background: Succinct Data Structures

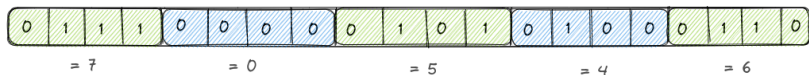
- ▶ Data structures which uses space close to the theoretic lower bound but allows efficient query operations (in-place without needing to decompress)
- ▶ Exists for e.g. (bit) vectors, trees, planar graphs, ...

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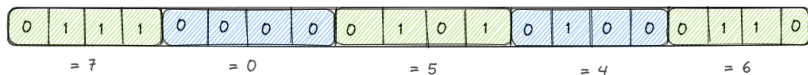
Succinct Integer Vector

Space requirement for integer x is $\ell = \lfloor \log_2(x) \rfloor + 1$ bits



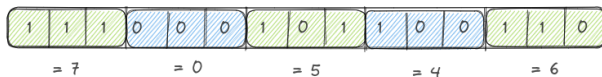
Succinct Integer Vector

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Encode integers with the minimal length of the max integer

$$3 = \lfloor \log_2(7) \rfloor + 1$$



We already reduce memory by 25%

SDSL: Succinct Data Structure Library

- ▶ C++11 library and abstraction for succinct data structures
- ▶ Open Source <https://github.com/simongog/sdsl-lite>
- ▶ Contains variety of different data structures. For now we only used the **Integer Vectors**.

SDSL: Integer Vectors

```
sdsl::int_vector<32> v(10000);  
for (size_t i = 0; i < 10000; i++) v[i] = i;  
cout << "Width: " << v.width() << ", size: "  
      << sdsl::size_in_bytes(v) << endl;  
sdsl::util::bit_compress(v);  
cout << "Width: " << v.width() << ", size: "  
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Reduces memory by 56.2% (\approx 22.5 KB)

Delta compression of sds1::int_vector

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sds1::int_vector<32> v(10000);  
for (size_t i = 0; i < 10000; i++)  
    v[i] = i + 10.000.000;  
cout << "Width: " << v.width() << ", size: "  
      << sds1::size_in_bytes(v) << endl;  
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Width: 24, size: 30008

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extractMinFromVector(v);  
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Delta compression of sds1::int_vector

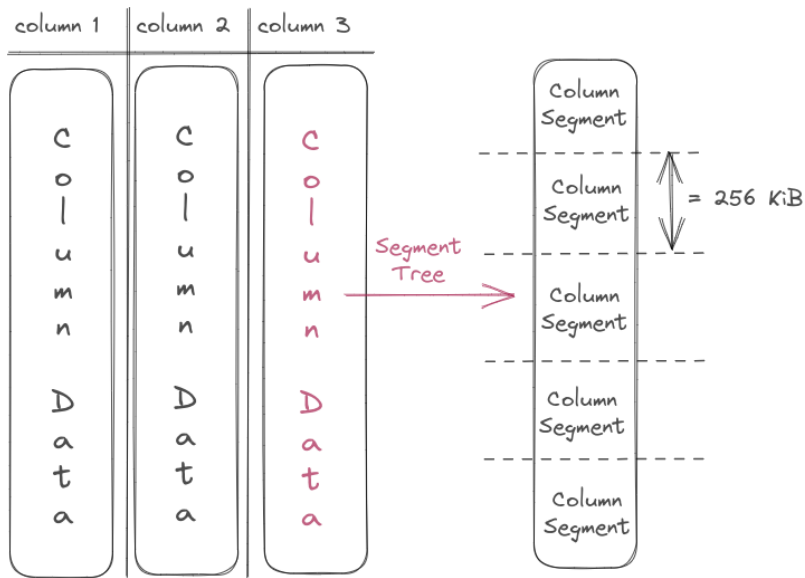
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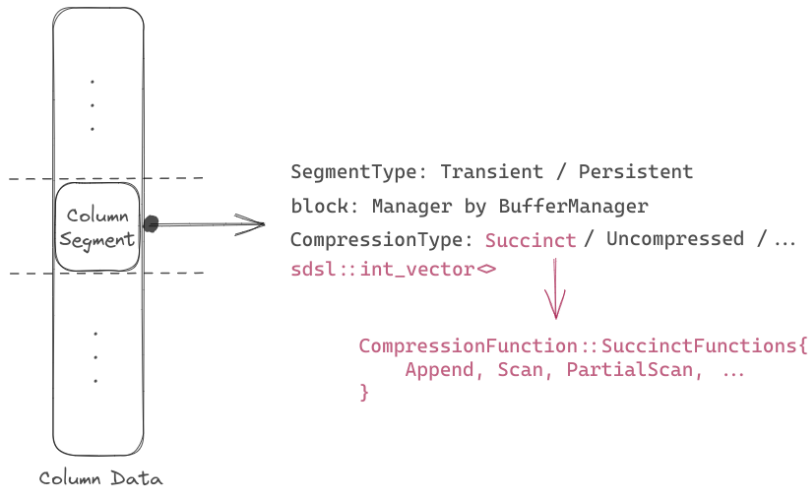
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In DuckDB we know the minimum of the vector directly without searching (column statistics)

DuckDB Storage Architecture 100 meter view



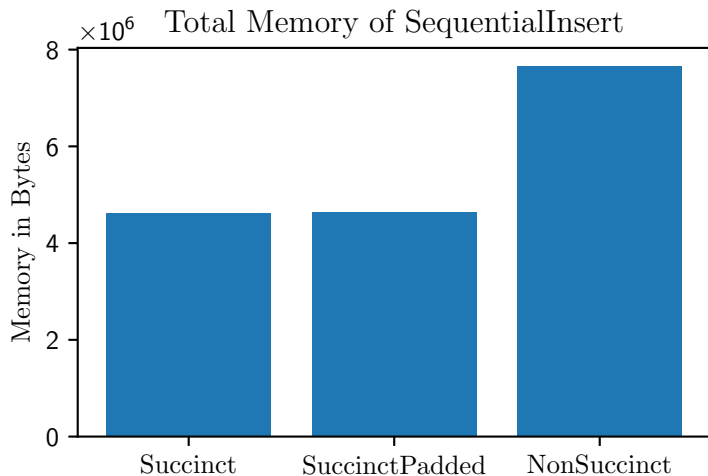
DuckDB Storage Architecture 10 meter view



Benchmarks: Sequential Insert and total Scan

Scanning 10^6 rows.

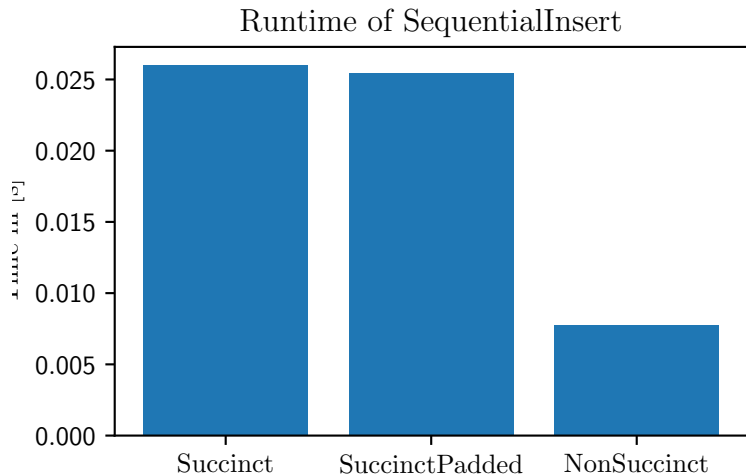
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SELECT * FROM t1;
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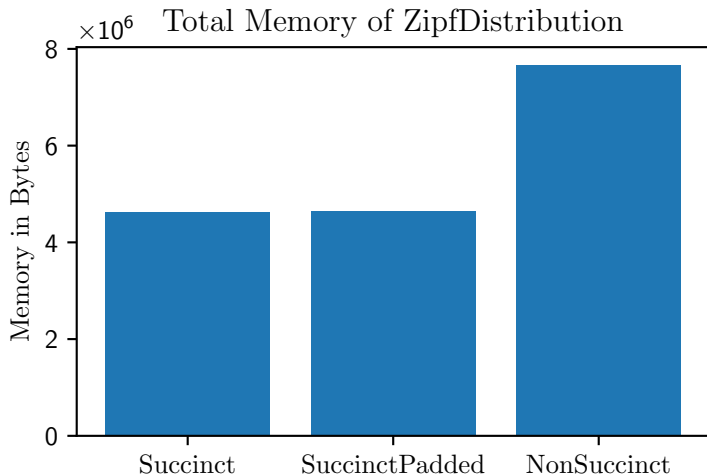
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Benchmarks: Zipf Selection

10.000 selections with Zipf Distribution of 10^6 total rows.

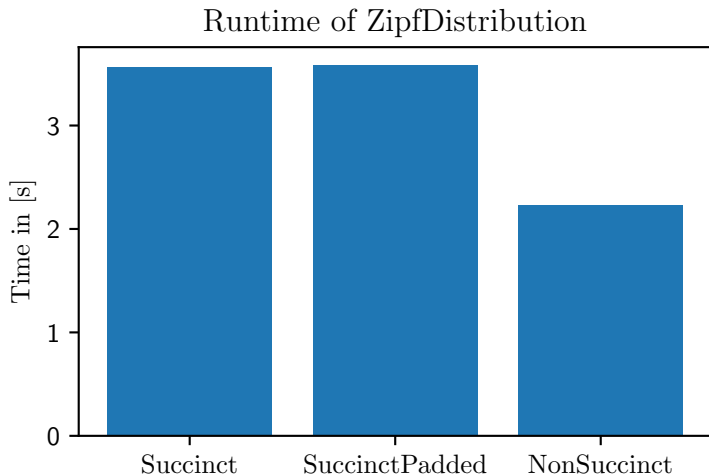
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SELECT i FROM t1  
WHERE i == {ZIPF_DISTRIBUTED_NUMBER};
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Benchmarks: Zipf Selection

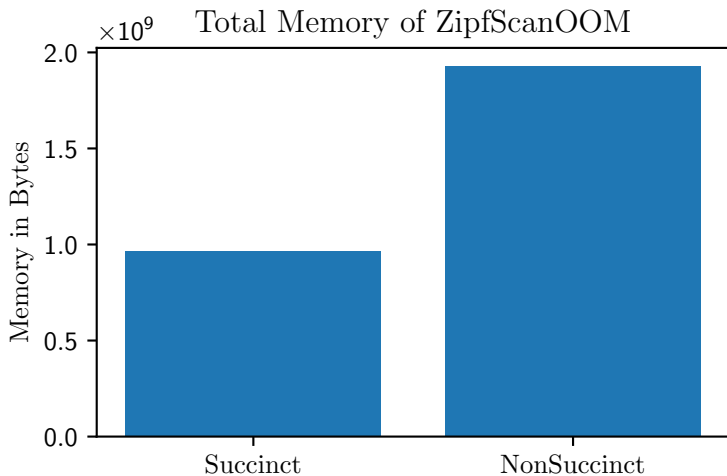
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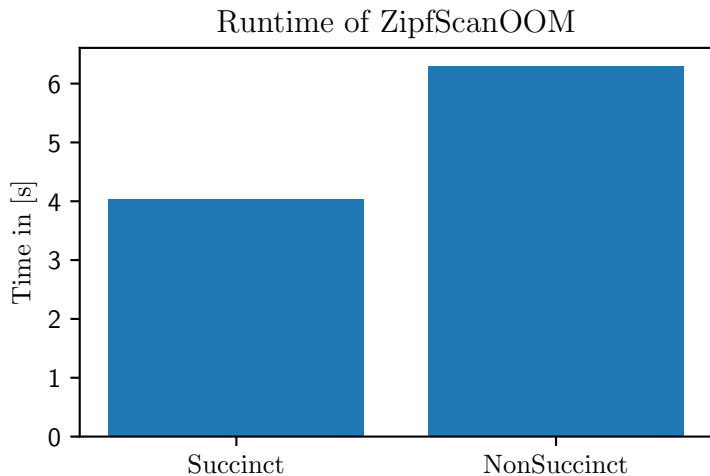
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Conclusion and Future Work

- ▶ For OLAPish queries it is not (yet?) worth it, large overhead.
- ▶ For OLTP transactions it might be worth it. Reduces memory by $\approx 40\%$ but increases runtime by $\approx 35\%$.
- ▶ Huge benefit if succinct representation fits in memory vs spilling to disk.

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Future Work:

- ▶ Need to investigate in access statistics to only compress rarely used column segments.
- ▶ Currently copying a lot of data as execution engine expects an uncompressed flat vector. Adapt execution engine to allow succinct vectors as well since they allow random access.