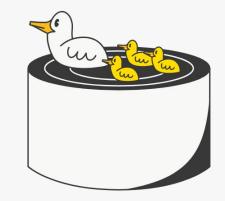


# WHAT CAN POSTGRES LEARN FROM DUCKDB?

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Maintainer of PgBouncer & pg\_duckdb
2025-05-15





Lightweight in-process SQL Analytics Engine



DuckDB is a new category of database





### DuckDB is a new category of database







created at:



created by:



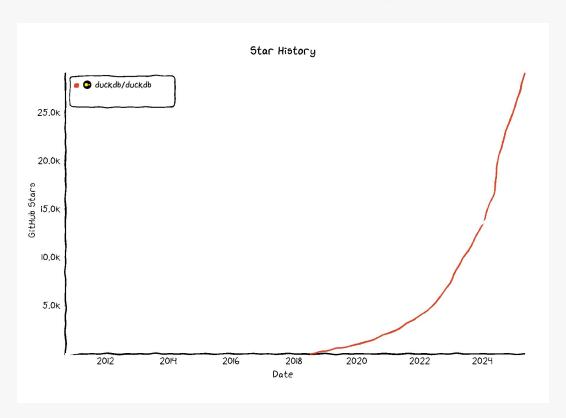
maintained by:





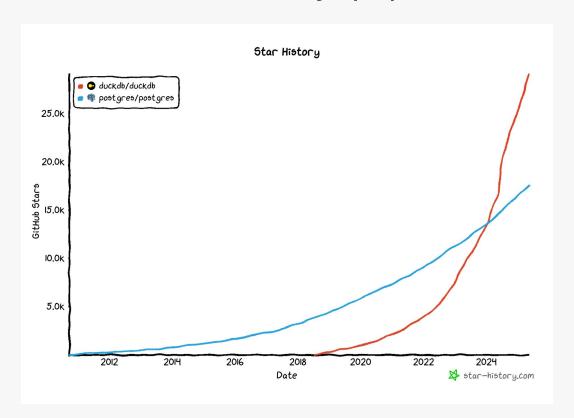
Community & Foundation

### And it's very popular





### And it's very popular





# Swiss army-knife for data



### World's best CSV parser

- An absurd amount of the world runs on CSV files
- An absurd amount of the world has broken / wonky CSV files
- An absurd amount of data engineering time is spent dealing with CSV file peculiarities
- Wouldn't it be nice if they just ... worked?

#### **Multi-Hypothesis CSV Parsing**

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### | Incorductions Group.5: 780,89.5: 90,50 | Hatchworks Ltd. 5: 2500 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000300 | 200.5: 1000

ABSTRACT

Comma Separated Value (CSV) files are commonly used to represent data. CSV is a very simple format, yet we show that it gives rise to a surprisingly large amount of ambiguities in its parsing and interpretation. We summarize the state-of-the-art in CSV parsers, which typically make a linear series of parsing and interpretation decisions, such that any wrong decision at an earlier stage can negatively affect all downstream decisions. Since computation time is much less scarce than human time, we propose to turn CSV parsing into a ranking problem. Our quality-oriented multi-hypothesis CSV parsing approach generates several concurrent hypotheses about dialect, table structure, etc. and ranks these hypotheses based on quality features of the resulting table. This approach makes it possible to create an advanced CSV parser that makes many different decisions, vet keeps the overall parser code a simple plug-in infrastructure. The complex interactions between these decisions are taken care of by searching the hypothesis space rather than by having to program these many interactions in code. We show that our approach leads to better parsing results than the state of the art and facilitates the parsing of large corpora of heterogeneous CSV

#### CCS CONCEPTS

Information systems → Inconsistent data;

#### ACM Reference format:

Till Döhmen, Hannes Mühleisen, and Peter Boncz. 2017. Multi-Hypothesis CSV Parsing. In Proceedings of SSDBM '17, Chicago, IL, USA, June 27-29, 2017, 12 pages.

https://doi.org/http://dx.doi.org/10.1145/3085504.3085520

#### 1 INTRODUCTION

Data scientists typically lose much time in importing and cleaning data, and large data repositories such as open government collections with tens of thousands of datasets remain under-exploited due to the high human cost of discovering, accessing and cleaning this data. CSV is the most commonly used data format in such repositories. The lack of explicit information on the CSV dialect.

Figure 1: Ambiguous CSV file which is at risk to be parsed incorrectly, because the number of commas and the number of semi-colons per row are the same.

the table structure, and data types makes proper parsing tedious and error-prone.

Tools currently popular among data scientists, such as R and Python offer robust CSV parsing libraries, which try to address parsing of messy CSV files with a number of practical heuristics. These libraries makes a linear sequence of parsing and interpretation decisions, such that any wrong decision at an earlier stage (e.g. determining the separator character) will negatively affect all downstream decisions. Interlinking different parsing steps (backtracking on prior decisions) is not done, because if all parsing decisions affect each other, the parsing code becomes very complex (code size would need to grow quadratically in the amount of decisions or even worse).

Since CPU-cycles are currently plentiful but human time is not, this research pursues an approach where CSV parsing becomes an computerized search problem. Our quality-oriented CSV parsing approach generates several concurrent hypotheses about dialect, table structure, etc. and in the end ranks these hypotheses based on quality features of the resulting table, such that the top-1 would be the automatic parsing result, or a top-K of parsed tables could be presented to the user. A high absolute score from the quality function can also be used to automatically parse large amounts of files. Only ambigous cases would be presented to a user. This can strongly reduce human data interpretation effort.

This very practical problem touches on various areas of related work. In the extended version of this paper [6], we survey the state-of-the art on this topic, which covers areas such as computer-assisted data-cleaning (data-wrangling), table-interpretation (e.g. on the web), automatic list extraction and even automated semantic (web) enrichment; covered more briefly in the related work Section 5.

Outline. In Section 2 we explaine CSV parsing problem by example, and introduce our multi-hypothesis parsing framework in Section 3. We demonstrate the improved parsing quality of our approach with computed quality metrics on the full data govale dataset collection, as well as on a sample of this collection using human ground truth in Section 4. We summarize related work in Section 5 and describe next steps in Section 5 afort concluding in Section 7.

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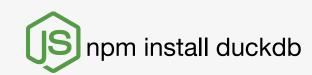
https://doi.org/http://dx.doi.org/10.1145/3085504.3085520

### **DuckDB** runs anywhere





curl https://install.duckdb.org | sh



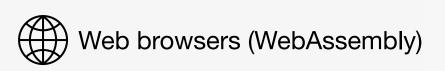


pip install duckdb





install.packages('duckdb')



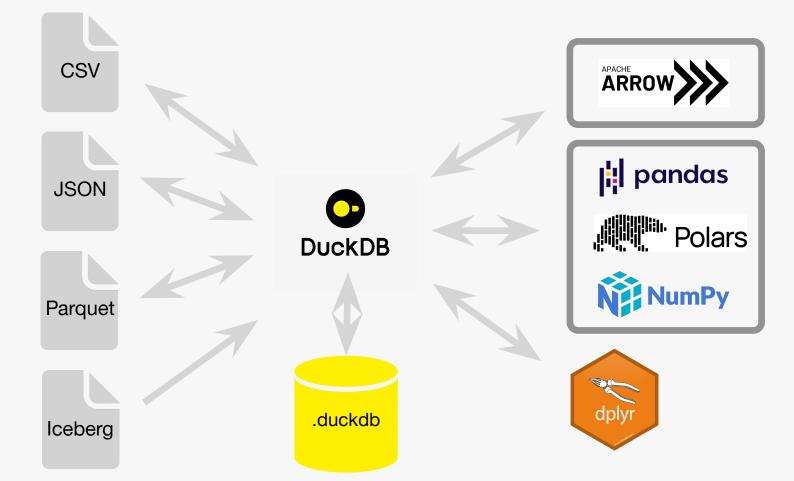


cargo add duckdb

ODBC, ADBC, & ~10 more languages!

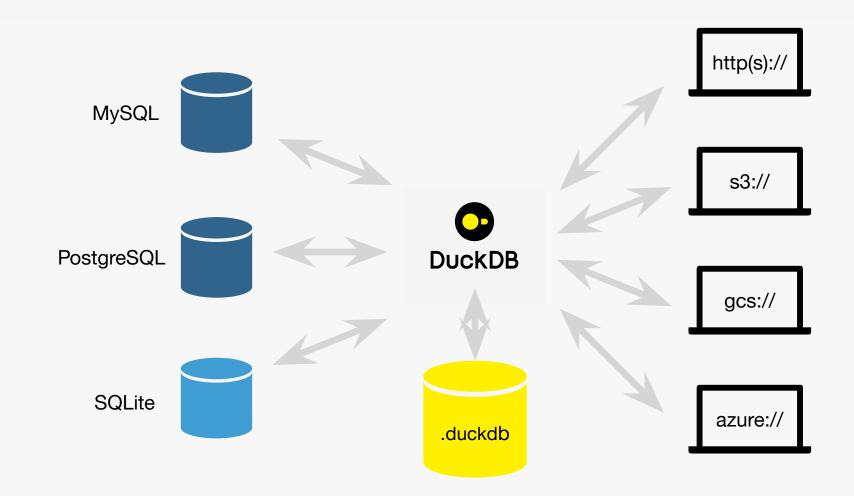
### Input and output formats





### Data sources and destinations





### But is it fast???



### YES!

System & Machine	Relative time (lower is better)	
DuckDB (c6a.4xlarge, 500gb gp2):		×1.27
ClickHouse (c6a.4xlarge, 500gb gp2):		×1.90
Snowflake (128×4XL):		×2.88
Spark (c6a.4xlarge, 500gb gp2):		×90.94



### YES!

System & Machine	Relative time (lower is better)	
DuckDB (c6a.4xlarge, 500gb gp2):		×1.27
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Snowflake (128×4XL):		×2.88
Spark (c6a.4xlarge, 500gb gp2):		×90.94
PostgreSQL (c6a.4xlarge, 500gb gp2):		×2435.13



# What can Postgres learn?



# What can Postgres learn?

- 1. Ease of use
- 2. Performance
- 3. Extensibility



### Ease of use



# Demo time

# \$ duckdb -ui



## Generic types

- 1. MAP(TEXT, INT)
- 2. UNION(i INT, f FLOAT)
- 3. STRUCT(t TEXT, i INTEGER)
- 4. Sensible arrays: TEXT[2][6] and TEXT[][]

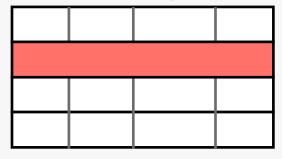


# Performance



# Postgres storage format

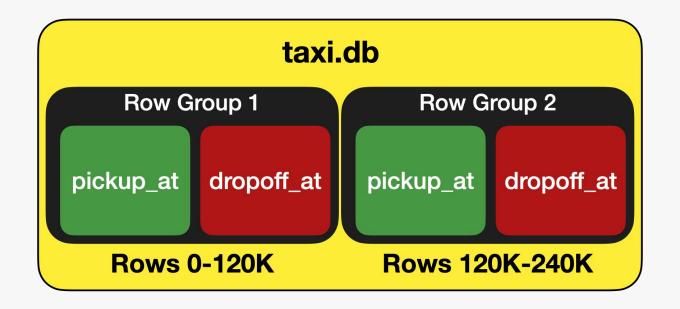
#### Row-based (tuples)



#### Optimized for:

- \* low memory footprint
- \* transactional workloads

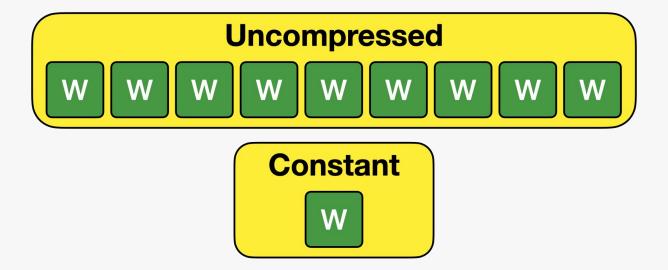
# DuckDB storage format



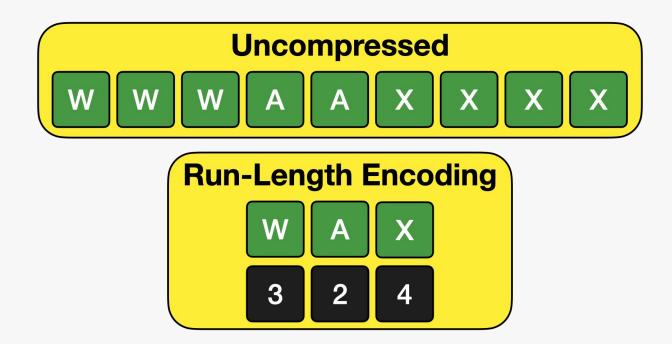
# ... with lightweight compression



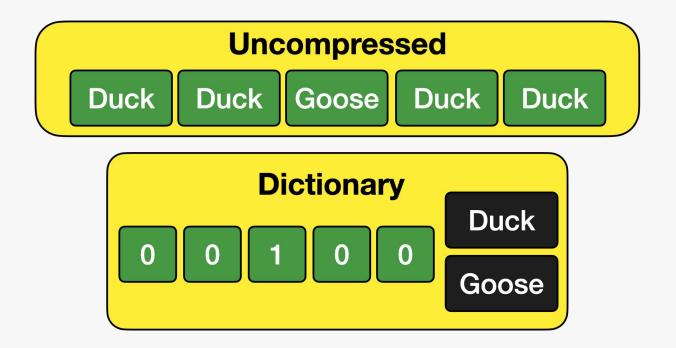
### Constant vectors



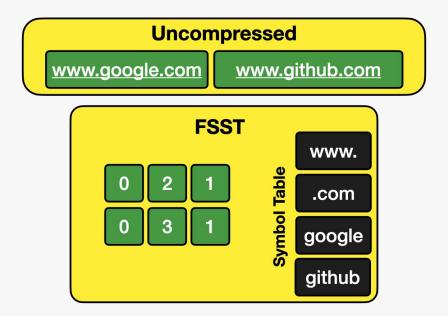
# Run-Length Encoding (RLE)



# Dictionary Encoding



# Fast Static Symbol Table



### How about execution?



### Morsel-Driven Parallelism

# Morsel-Driven Parallelism: A NUMA-Aware Query Evaluation Framework for the Many-Core Age

```
Viktor Leis* Peter Boncz<sup>†</sup> Alfons Kemper* Thomas Neumann*

* Technische Universität München <sup>†</sup> CWI

* {leis,kemper,neumann}@in.tum.de <sup>†</sup> p.boncz@cwi.nl
```

Source: https://db.in.tum.de/~leis/papers/morsels.pdf

### Morsel-Driven Parallelism

• Morsel-driven query execution is a new parallel query evaluation framework that fundamentally differs from the traditional Volcano model in that it distributes work between threads dynamically using work-stealing. This prevents unused CPU resources due to load imbalances, and allows for elasticity, i.e., CPU resources can be reassigned between different queries at any time.

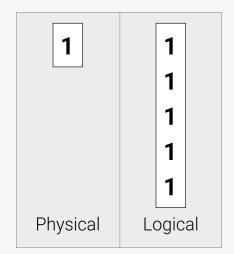
# Execution on compressed data

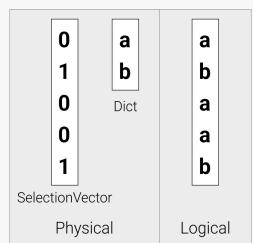
Flat

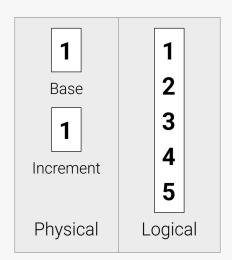
Constant Dictionary

Sequence

3 5 Physical & Logical







# Factorized joins

# Extensibility



# Built-in package manager



# INSTALL spatial



### FDWs on steroids

ATTACH 'host=my-postgres.com' AS postgres (TYPE postgres);

FROM postgres.myschema.users WHERE age > 21;



# Custom filesystems

FROM 's3://my-bucket/myfile.parquet';

FROM iceberg\_scan('az://metadata/v1.metadata.json');

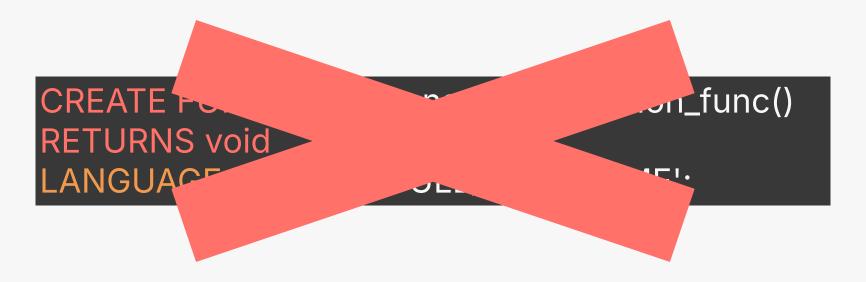


### Runtime added native functions

CREATE FUNCTION my\_native\_extension\_func()
RETURNS void
LANGUAGE C AS 'MODULE\_PATHNAME';



### Runtime added native functions







### There's more!

- duckdb.org/news
- github.com/duckdb/duckdb
- github.com/duckdb/pg\_duckdb
- MIT licensed

Questions?

