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DuckDB A Modern Modular & Extensible Database System

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Who Am I?

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What is DuckDB?

- DuckDB
- In-Process OLAP DBMS
 - "The SQLite for Analytics"

- Free and Open Source (MIT)
- <u>duckdb.org</u>





What is DuckDB?

- SQLite inspired us in many ways:
 - Easy installation
 - Ease of use
 - Robustness

DuckDB aims to be the "SQLite for Analytics"





Analytics has unique challenges

Transactional workloads = simple queries SQLite can be feature complete with a very small footprint

Analytical workloads = **complex queries** DuckDB needs many more operations, functions, optimizers,....





Analytics requires a giant diversity of operations **Collaboration is required!**

A single entity cannot hope to implement: Data wrangling tools Classification algorithms, ML toolkits Data cleaning tools





Research other important collaboration area

DuckDB originates from research world At CWI and elsewhere, people do their research in DuckDB

Important to allow extending/modifying system e.g. add new join operator, new optimizer, ...





System builders want to use components of DuckDB

For example: Use only the front-end Use only the back-end





How do we enable this collaboration?

Three aspects:

Flexible data import and export **Extensibility** of the system Hooks in different locations in the system





Flexible Data Import & Export



Data import and export

Crucial for communication **between libraries**

Many use cases

Load data exported from other systems

Pre-process in DuckDB \rightarrow use plotting/statistics/ML libraries

Mix usage DuckDB + other data wrangling libraries

Export data to persistent storage (e.g. Parquet files on S3)





What makes DuckDB different? All database systems can import/export data ... but very slowly!

Ancient protocols, designed to transfer kilobytes of data Unsuitable for modern workloads! Import/export through text-based formats (CSV files)

Don't Hold My Data Hostage – A Case For Client Protocol Redesign

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ABSTRACT

ADSTRAC1 Transferring a large amount of data from a database to a client program is a surprisingly expensive operation. The time this requires can easily dominate the query execution time for large result sets. This represents a significant hurdle for external data analysis, for example when using statistical software. In this paper, we explore and analyse the result set serialization design space. We present experimental results from a large chunk of the database market and show the inefficiencies of current approaches. We then propose a columnar serialization method that improves transmission mnar serialization method that improves tra formance by an order of magnitude

Keywords Databases, Client Protocols, Data Export

1. INTRODUCTION

Transferring a large amount of data out of a database sy n to a client program is a common task. Examples inclu nplex statistical analysis or machine learning application mplex statistical analysis or machine learning applications at need access to large samples for model construction verification. However, that operation is expensive. It is en more costly if the data is transported over a network

rall system performance. Figure 1 shows the time taken in the SQL query "SELECT * FROM lineitem" using an BC connector and then fetching the results for various ta management systems. We see large differences betwee ms and disappointing performance overall. Modern data gement systems need a significant amount of time t fer a modest amount of data from the server to the ransier a modest amount of data from the server to the lient, even when they are located on the same machine. From a survey of popular machine learning tools, we have bund that none of them integrate into database systems evond the bulk transfer of data sets. RapidMiner [18] and Veka [16] allow analyzing data from a database connection,

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Figure 1: Wall clock time for retrieving the lineitem table (SF10) over a loopback connection. The table (SF10) over a loopback connection. The dashed line is the wall clock time for netcat to trans-

Connection Query Execution RSS + Transfer

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fer a CSV of the data.

but this is strictly limited to loading. The users must issue their own queries to load the data from the database. This will likely be a query such as the one above. None of R's machine learning packages [19] use a database interface or import a package that does. Instead, they again rely on the user to precide them with data in a table that is already user to provide them with data in a table that is alre loaded into memory. The same is true for Python-based toolkits like SciKit-Learn [26] or TensorFlow [2].

Because of the large cost of data export, analysts settl for exporting small samples from the database. This way data export is not a bottleneck in their analysis pipeline However, this generally reduces the accuracy of machin earning and classifi The issue of slow result export has been identified befor

quire large, system-specific, overnauis of existing pipelin and are difficult to implement and optimize. There is little o scientific work done on improving result set serializati

lesign of existing serialization formats they perform when transferring various data sets in different etwork scenarios, and examine why they show this perfo mance. We explore the design space of serialization for with experiments, and look at the different trade-offs that are made when designing a client protocol.

Don't Hold My Data Hostage -A Case For Client Protocol Redesign



DuckDB is designed for **bulk data export/import** In-process = zero-copy data sharing Versatile input & output APIs

Same API used for both internal tables & external sources Allows for streaming data, parallel scans, projection & filter

pushdown, index usage, ...





DuckDB can efficiently consume and output many formats

i pandas julia vere









Extensibility of the system DuckDB supports extensions that add new functionality

Allows users to integrate new functionality in the system



We create extensions ourselves liberally Eat your own duck food

Allows us to add functionality...

...without bloating the core system

...that is only included in certain distributions (e.g. Python client)

...without introducing external dependencies to the core





ICU Extension

Adds support for time-zone awareness

Extension is several megabytes Includes ICU localization data Roughly same size as DuckDB core! As an extension we keep this optional

Adds support for collations (language-based sorting, comparisons)







HTTP-FS Extension

Adds support for reading/writing data over HTTP(S) Adds support for reading/writing data with S3

Adds dependency on OpenSSL As an extension, we keep the core dependency-free









Extensibility of the System Extensions are **powerful** Goal: Allow every component of the system to be extended

Currently extensions can:

- Add functions (scalar, aggregate, window)
- Add new types
- Add data sources and sinks
- Add collations, time zones
- Add parser functionality, optimizers





No need to talk to us No need for us to maintain their code **Extension Repository** Extensions can be installed using SQL **INSTALL** httpfs;

LOAD httpfs;

By default from our repository

Custom repository can be used

Goal: allow users to create and maintain their own extensions







DuckDB uses a typical pipeline for query processing



Standard workflow: use all components



Standard pipeline: SQL input → Data output

System builders/researchers often want to use parts of DuckDB Different hooks in/out of the system





Velox: to use the parser Parser pipeline: SQL Input → Parse Tokens Output







Ibis: provide a new front-end to DuckDB Ibis front-end pipeline: Substrait plan input → Data Output







Ibis: consume SQL and use a different back-end Ibis back-end pipeline: SQL input → Substrait plan output













- Goal: more extensibility options
 - New index types
 - Alternate versions of existing operators (e.g. new joins)
 - Define new casts between types







Goal: support fully custom storage back-end Currently support data-input and copy output

Missing support for routing insert, update, delete to custom storage





Goal: extended custom catalog support

Currently support **replacement scans** This allows for views to be stored in a different catalog

Needs to be extended to fully support custom catalogs Including for all **different catalog types**







Thanks for having me!



Any questions?