

A Short Summary of the Last Decades of Data Management



Researcher

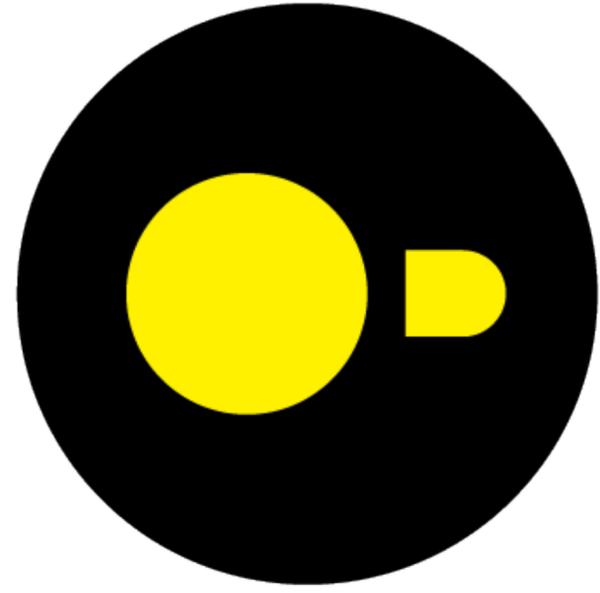


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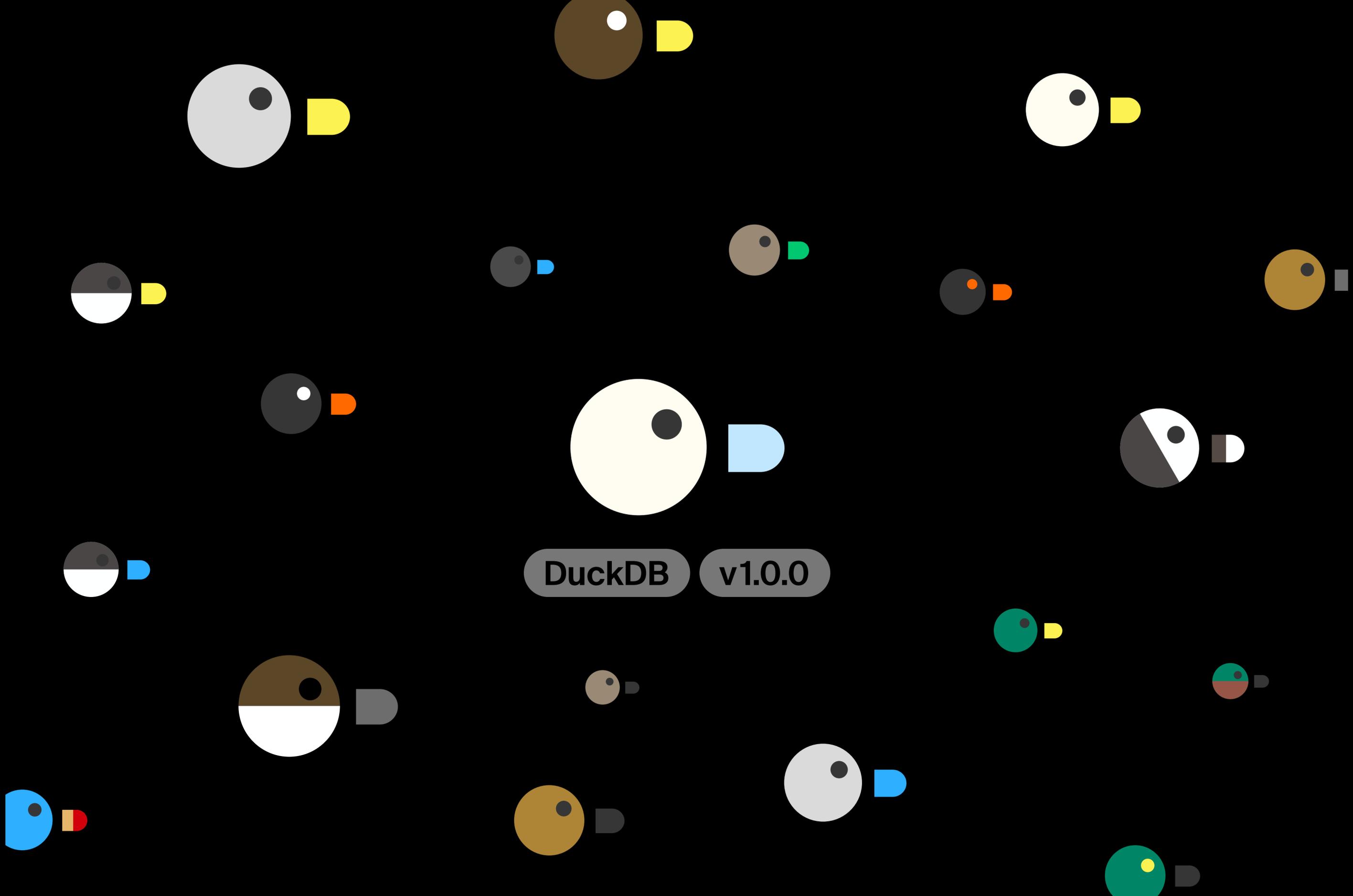


Professor





DuckDB



DuckDB

v1.0.0

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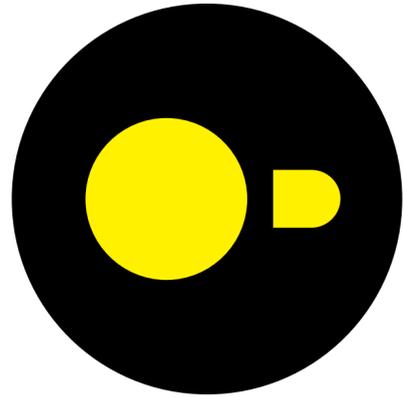
📅 Wednesday Jun 12 🕒 14:30 – 15:20
📍 Veilingzaal

DuckDB: Crunching data anywhere, from laptops to servers



Gábor Szárnyas

Technical Writer at DuckDB



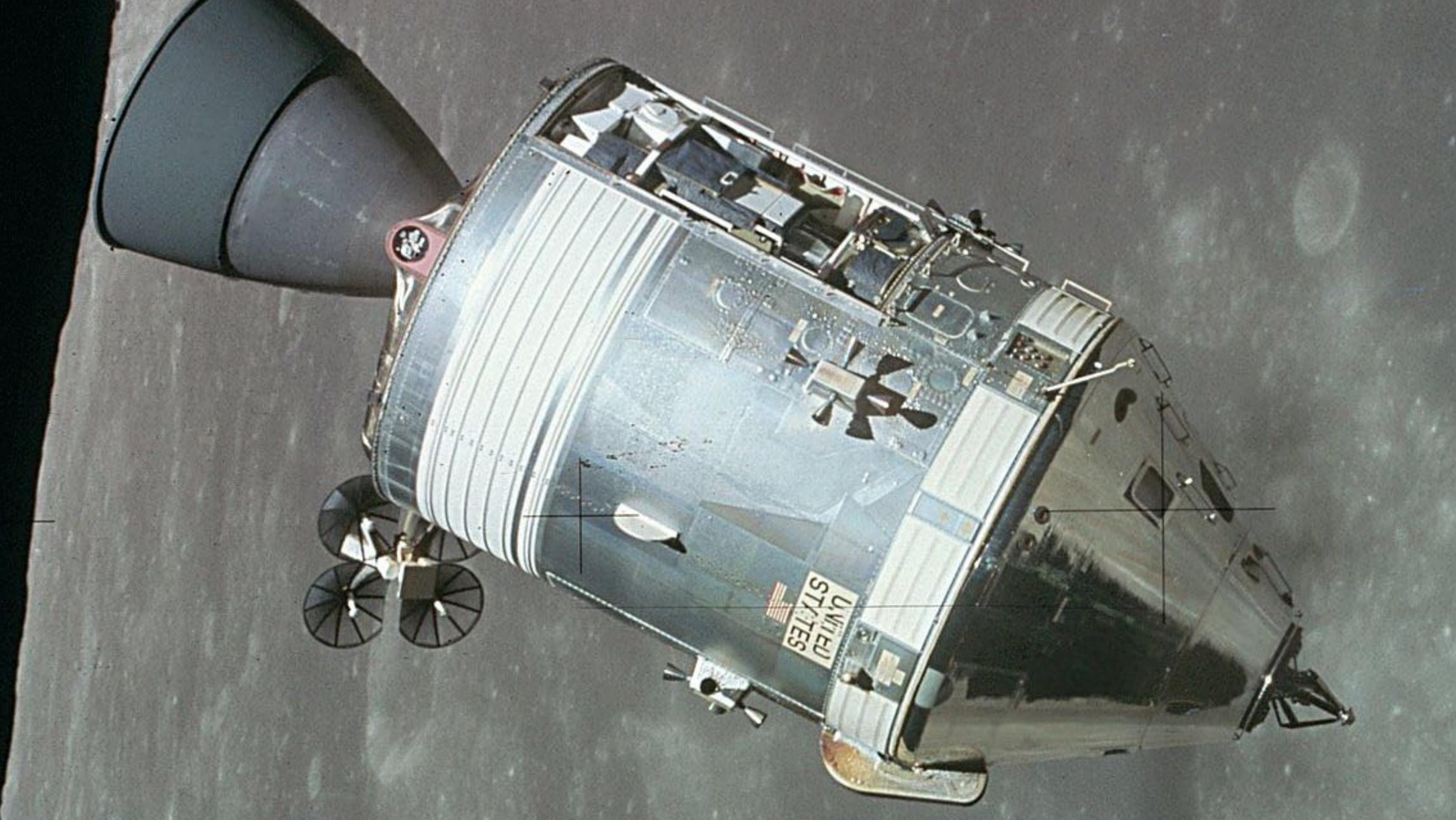
Ancient History





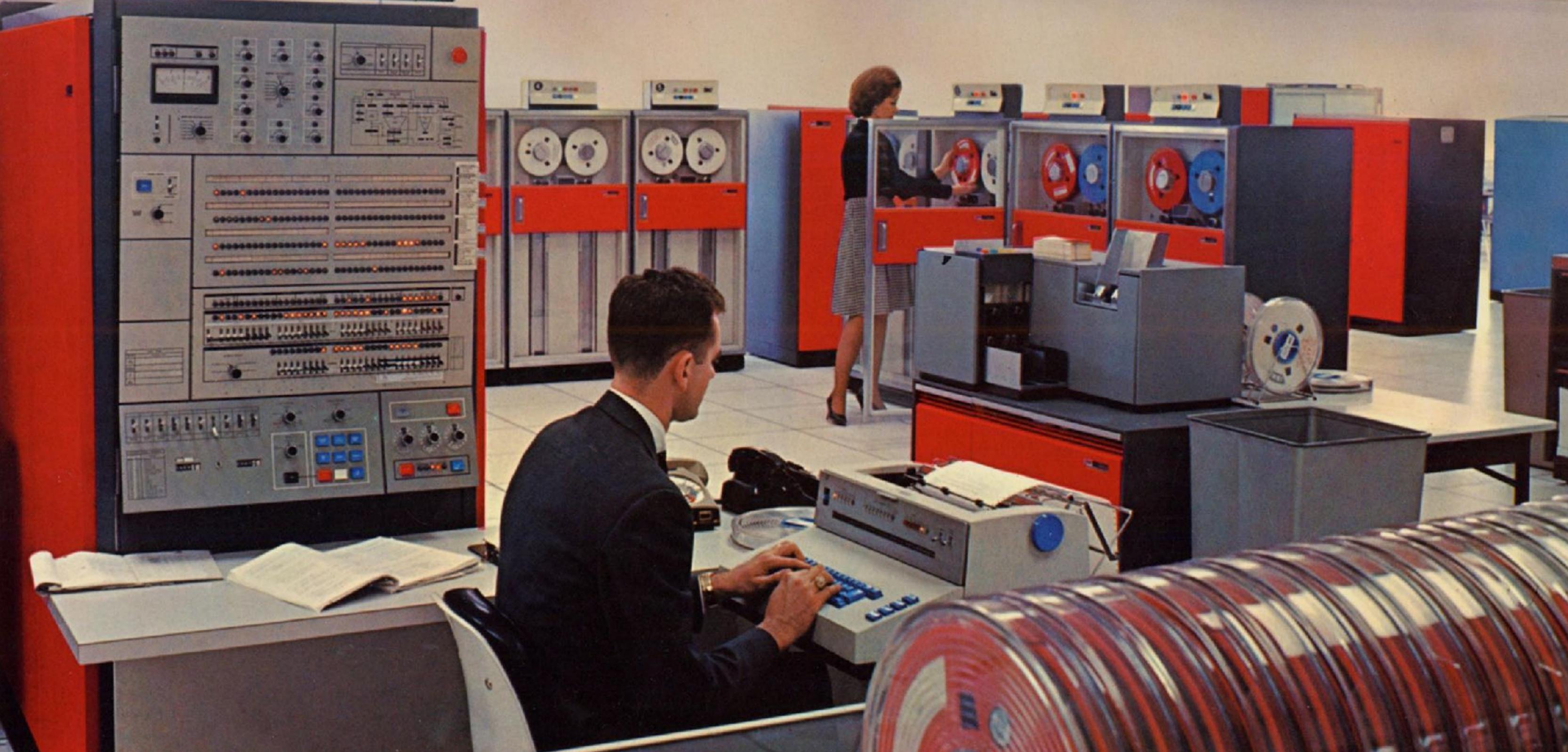
Item	Amount
Hop	42
Barley	84



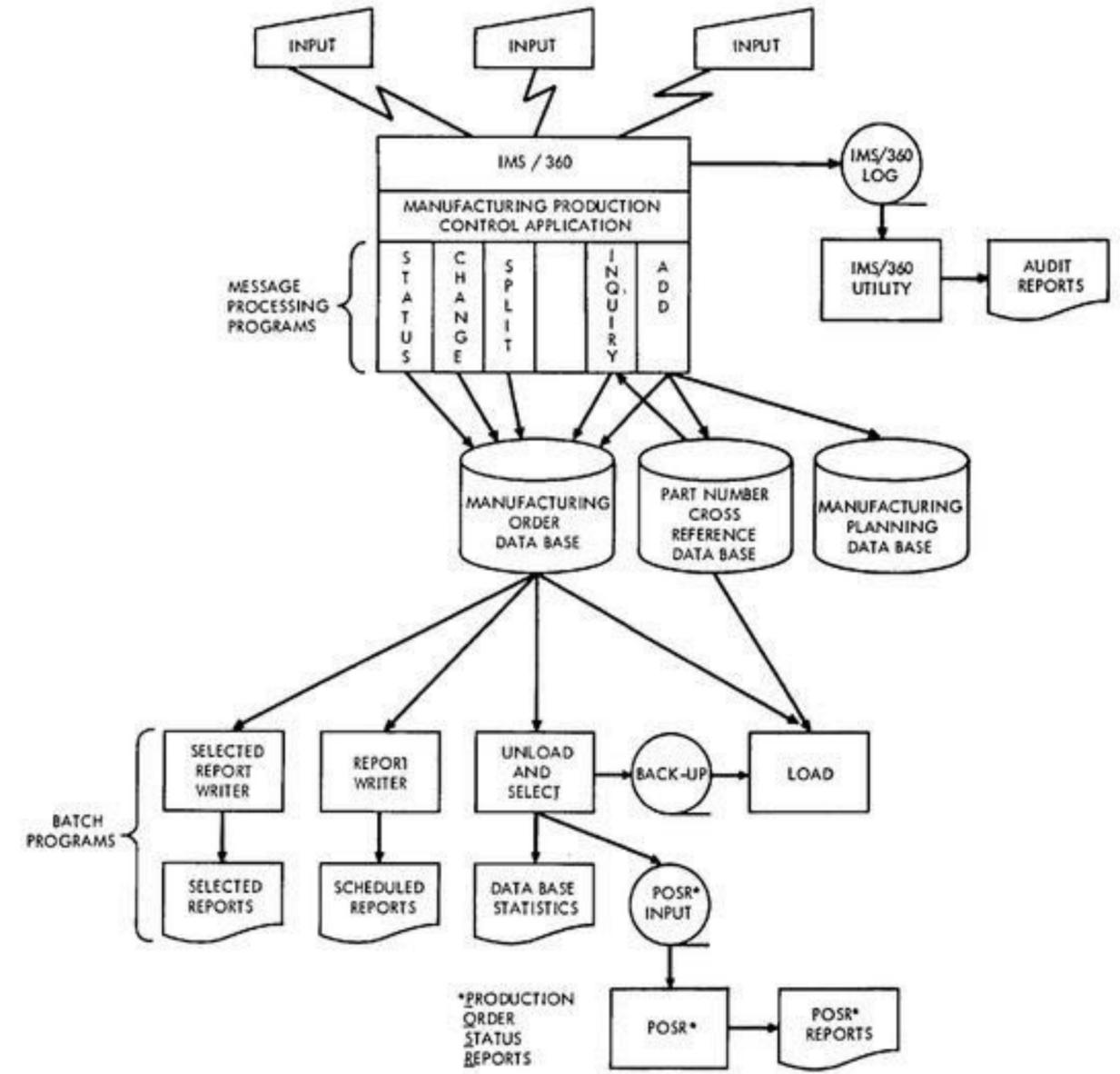


UNITED STATES

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A Relational Model of Data for Large Shared Data Banks

E. F. CODD

IBM Research Laboratory, San Jose, California

Future users of large data banks must be protected from having to know how the data is organized in the machine (the internal representation). A prompting service which supplies such information is not a satisfactory solution. Activities of users at terminals and most application programs should remain unaffected when the internal representation of data is changed and even when some aspects of the external representation are changed. Changes in data representation will often be needed as a result of changes in query, update, and report traffic and natural growth in the types of stored information.

The relational view (or model) of data described in Section 1 appears to be superior in several respects to the graph or network model [3, 4] presently in vogue for non-inferential systems. It provides a means of describing data with its natural structure only—that is, without superimposing any additional structure for machine representation purposes. Accordingly, it provides a basis for a high level data language which will yield maximal independence between programs on the one hand and machine representation and organization of data on the other.

A further advantage of the relational view is that it forms a sound basis for treating derivability, redundancy, and consistency of relations—these are discussed in Section 2. The network model, on the other hand, has spawned a number of confusions, not the least of which is mistaking the derivation of connections for the derivation of relations (see remarks in Section 2 on the “connection trap”).

Finally, the relational view permits a clearer evaluation of the scope and logical limitations of present formatted

1973 IBM System R
1979 Oracle
1983 IBM DB/2
1996 PostgreSQL

**One Size
Fits All?**



Transactional

row-based

date

id

type

station

A diagram illustrating row-based data storage. It shows a 4x4 grid of colored rectangles. Each row represents a record, and each column represents a field. The columns are labeled 'date', 'id', 'type', and 'station'. The rows are colored as follows: Row 1 (blue), Row 2 (green), Row 3 (yellow), and Row 4 (red). This indicates that all data for a single record is stored together in a single row.

date	id	type	station

Analytical

column-based

date

id

type

station

A diagram illustrating column-based data storage. It shows a 4x4 grid of colored rectangles. Each column represents a field, and each row represents a record. The columns are labeled 'date', 'id', 'type', and 'station'. The columns are colored as follows: Column 1 (cyan), Column 2 (pink), Column 3 (blue), and Column 4 (red). This indicates that data for a single field is stored together in a single column.

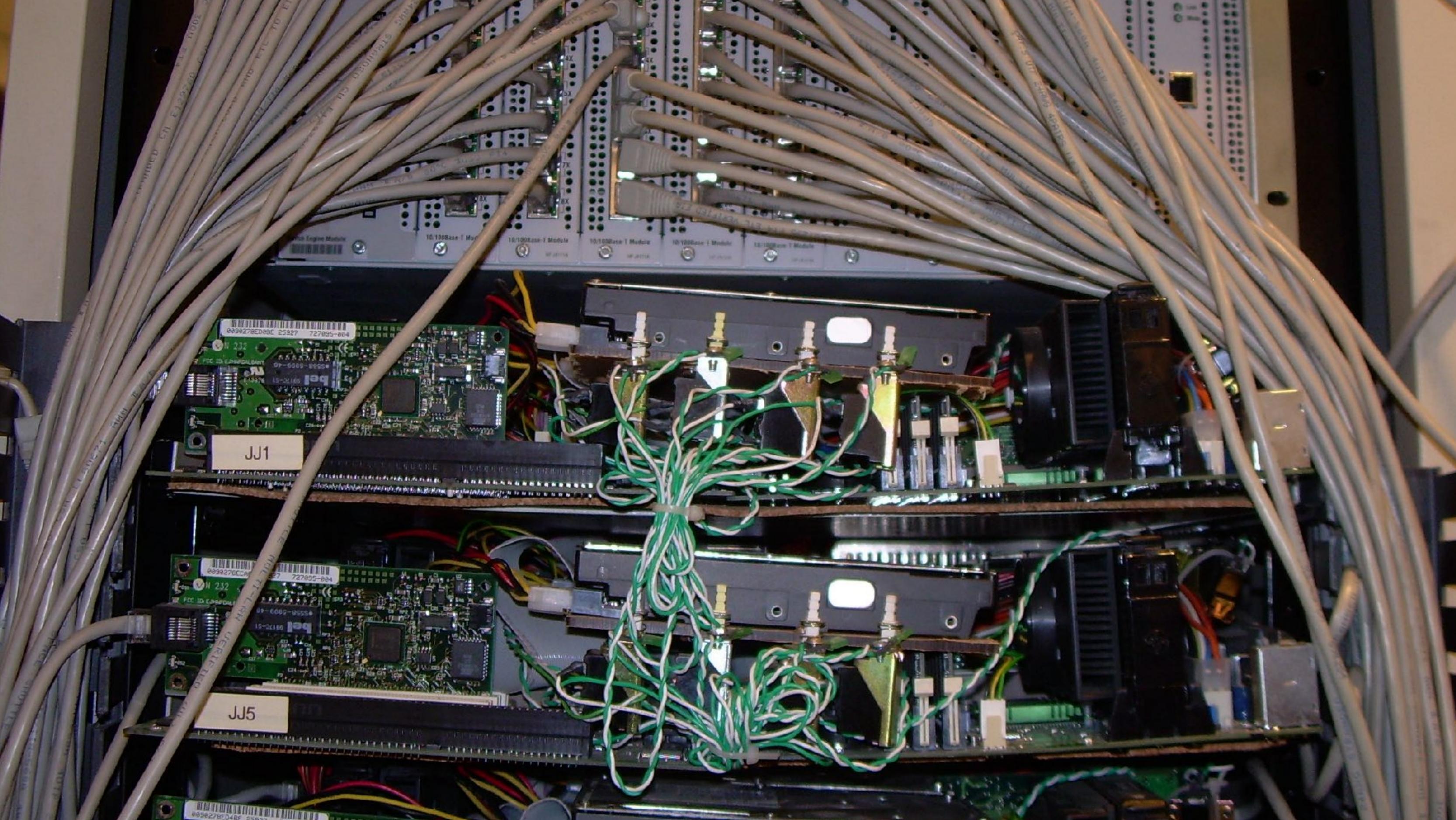
date	id	type	station

Transactional



Analytical





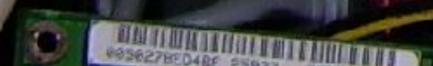
10/100Base-T Module 10/100Base-T Module 10/100Base-T Module 10/100Base-T Module 10/100Base-T Module



JJ1



JJ5



2006

No declarative query language.

For scale!



~~2006~~

~~No declarative query language.~~

~~For scale!~~

2010

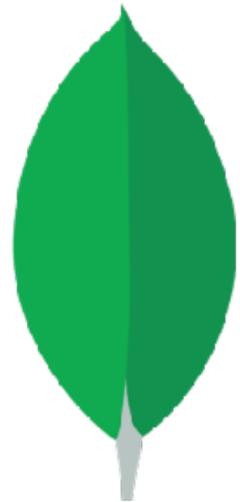
Let's have SQL again



2009

No schema.

For scale!



mongoDB®

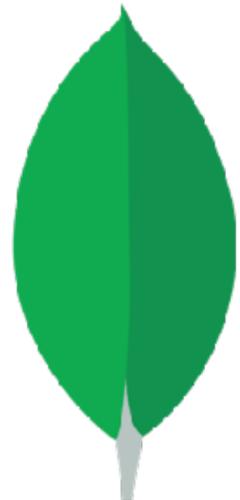
~~2009~~

~~No schema.~~

~~For scale!~~

2017

Schema validation



mongoDB®

2008

No ACID transactions.

For scale!



cassandra

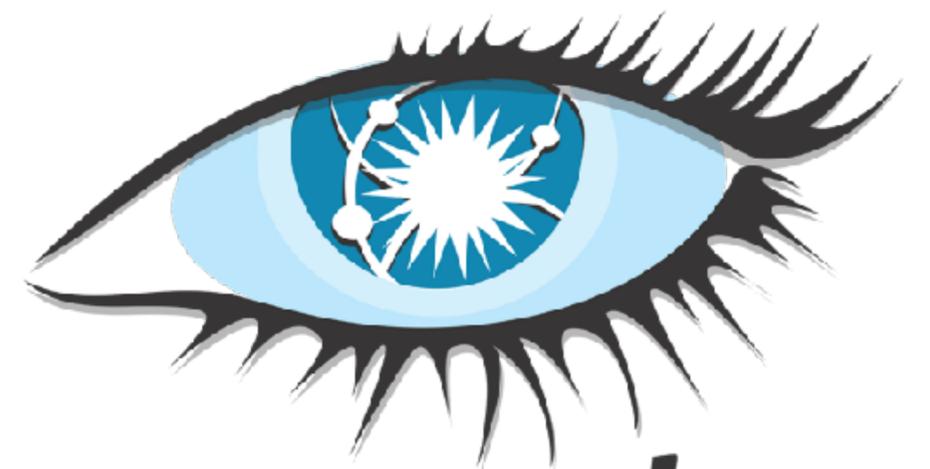
~~2008~~

~~No ACID transactions.~~

~~For scale!~~

2023

Support ACID after all



cassandra

2014

No Internal Storage

For freedom!



~~2014~~

~~No Internal Storage~~

~~For freedom!~~

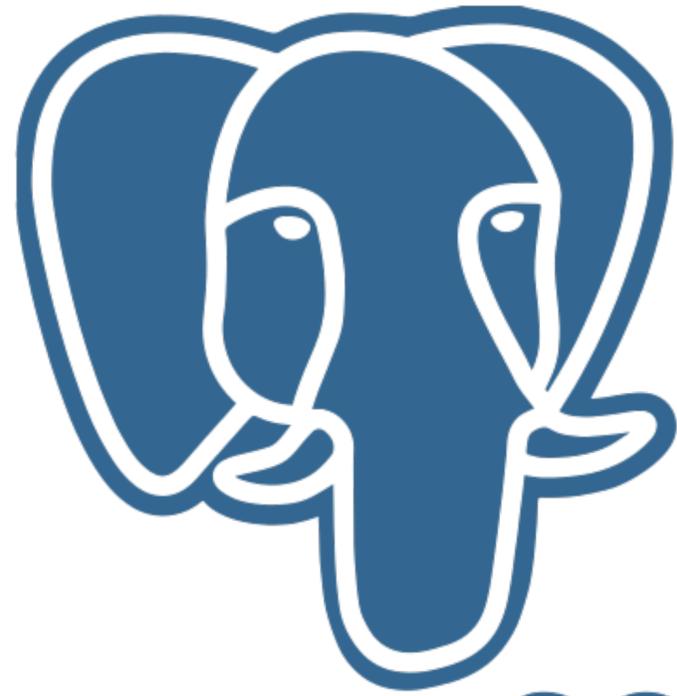
2024

Internal Storage





Tables
SQL
ACID



PostgreSQL



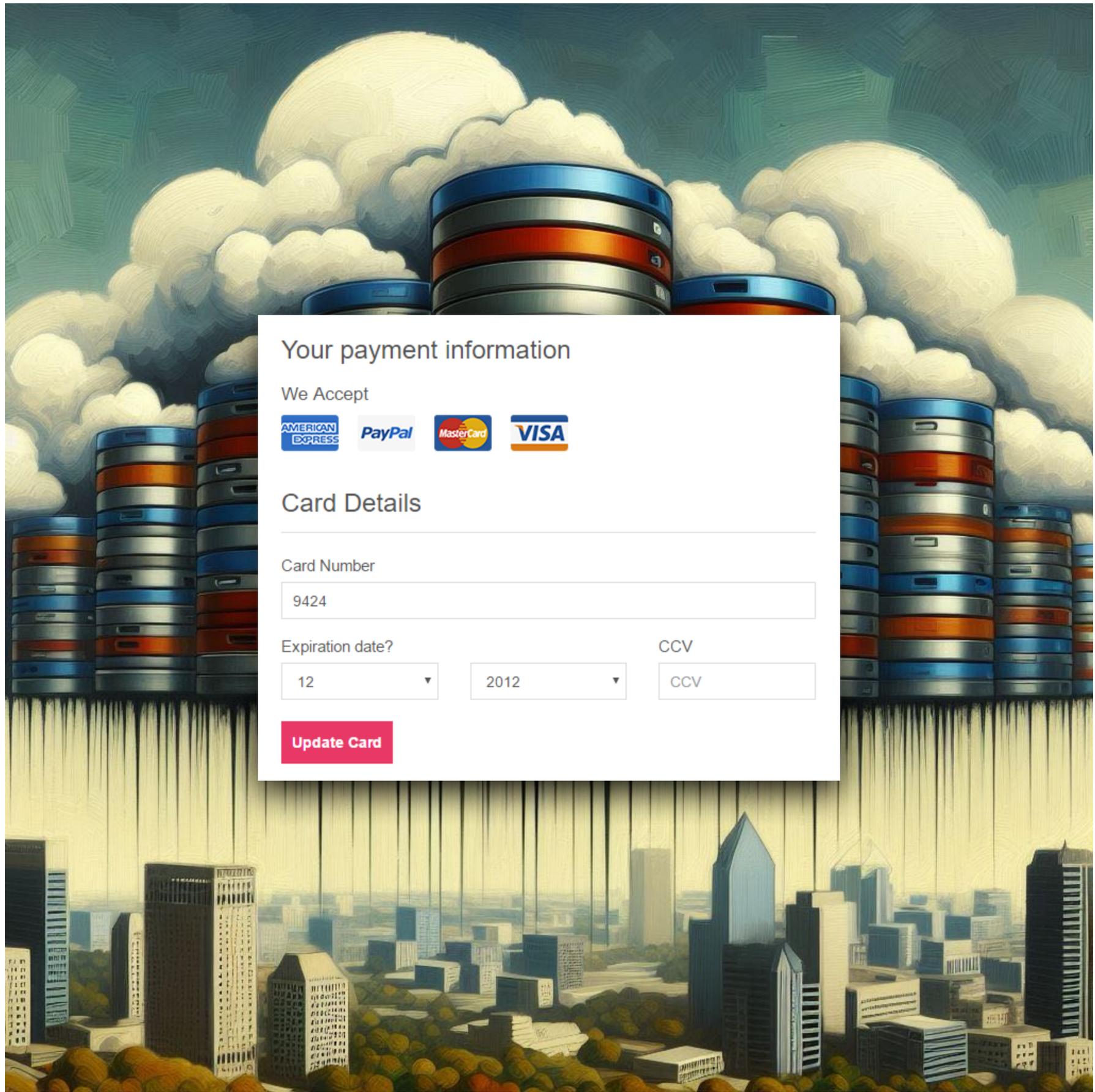
New SQL!



~~ETL~~

ELT





Your payment information

We Accept



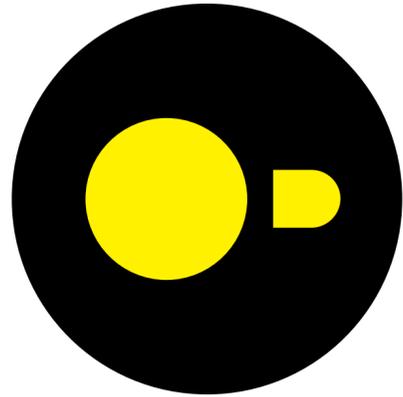
Card Details

Card Number

Expiration date?

CCV

Update Card



**Relational eats
Everything***





XML
RDF
Blockchain
ORM
...

Key/Value

Key VARCHAR	Value VARCHAR

Document

Key VARCHAR	Value JSON

Time Series

Time TIMESTAMP	Measurement DOUBLE

Vectors

Key VARCHAR	Embedding FLOAT[256]



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[Doug Hood](#)

Oracle AI Vector Search Product Manager



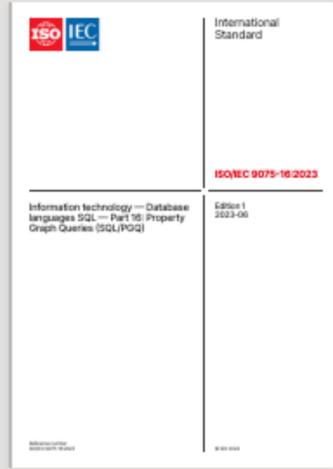
[Ranjan Priyadarshi](#)

Senior Director, Product Management(Mission-Critical Data and AI Engines)

Graphs

ID BIGINT	Properties KEY/VALUE

Source BIGINT	Target BIGINT	Properties KEY/VALUE



Read sample

ISO/IEC 9075-16:2023

Information technology — Database languages SQL

Part 16: Property Graph Queries (SQL/PGQ)

Published (Edition 1, 2023)

ISO/IEC 9075-16:2023

CHF 216

Format

PDF

Language

English

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Data Frames



Full Text Search

Old Dogs Are Great at New Tricks: Column Stores for IR Prototyping

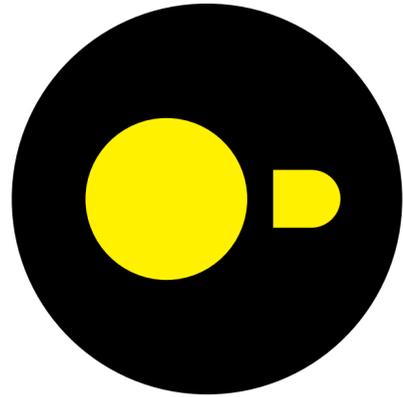
Hannes Mühleisen,¹ Thaer Samar,¹ Jimmy Lin,² and Arjen de Vries¹

TermID	Term
1	put
2	robe
3	wizard
4	hat

TermID	DocID	Pos
1	1	2
2	1	5
3	1	7
4	1	8

**What refuses
to be eaten?**





Big Data Is Dead?



Human typing speed					200	Characters/s
NL working population			10 000 000			
NL work year				100 000		Minutes
Total data produced	200 000 000 000 000					Bytes
					200	Terabytes
After compression					60	Terabytes



Scalability! But at what COST?

Frank McSherry Michael Isard Derek G. Murray
Unaffiliated Unaffiliated* Unaffiliated†

Abstract

We offer a new metric for big data platforms, COST, or the Configuration that Outperforms a Single Thread. The COST of a given platform for a given problem is the hardware configuration required before the platform outperforms a competent single-threaded implementation. COST weighs a system’s scalability against the overheads introduced by the system, and indicates the actual performance gains of the system, without rewarding systems that bring substantial but parallelizable overheads.

We survey measurements of data-parallel systems recently reported in SOSP and OSDI, and find that many systems have either a surprisingly large COST, often

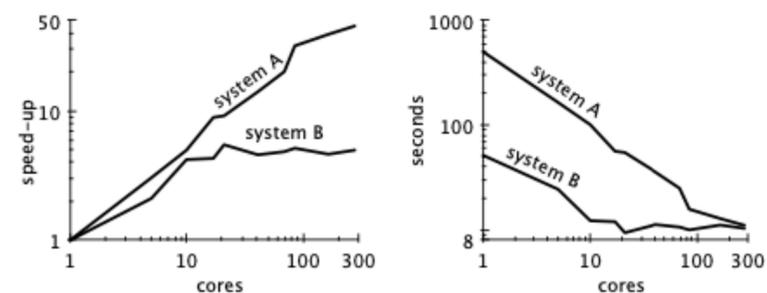
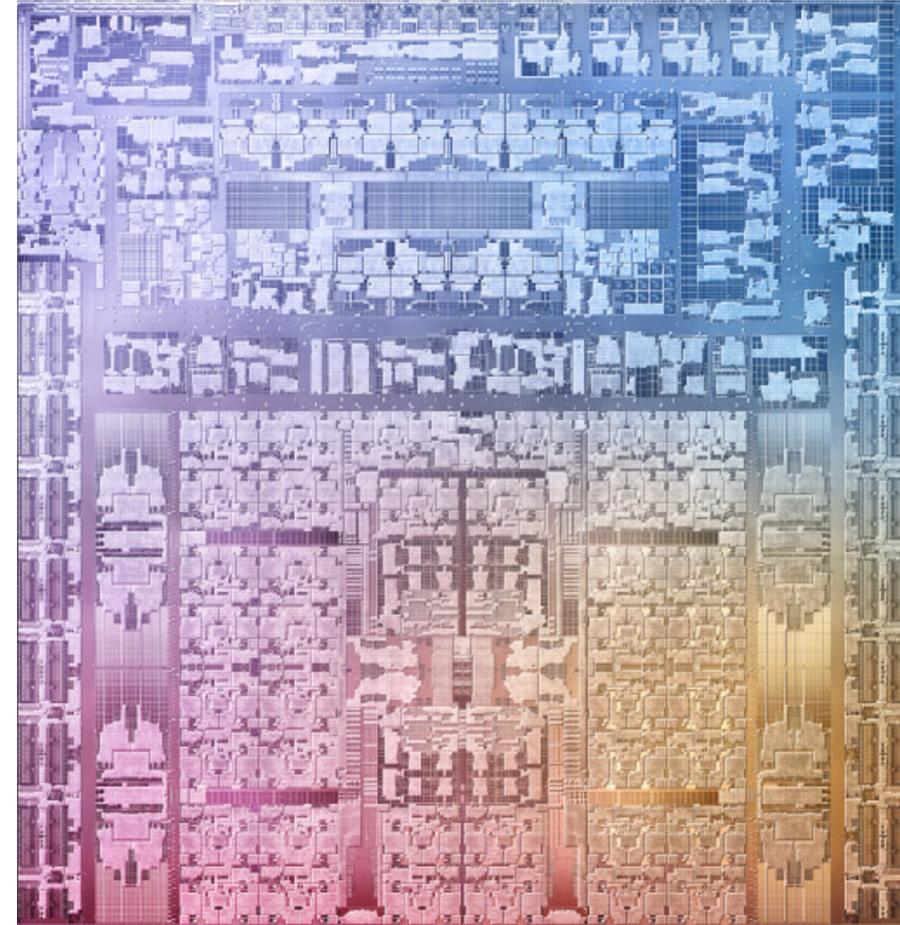


Figure 1: Scaling and performance measurements for a data-parallel algorithm, before (system A) and after (system B) a simple performance optimization. The unoptimized implementation “scales” far better, despite (or rather, because of) its poor performance.

CPU



SSDs





min	max
May 30	May 31

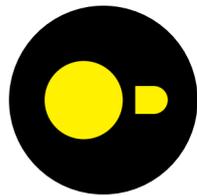
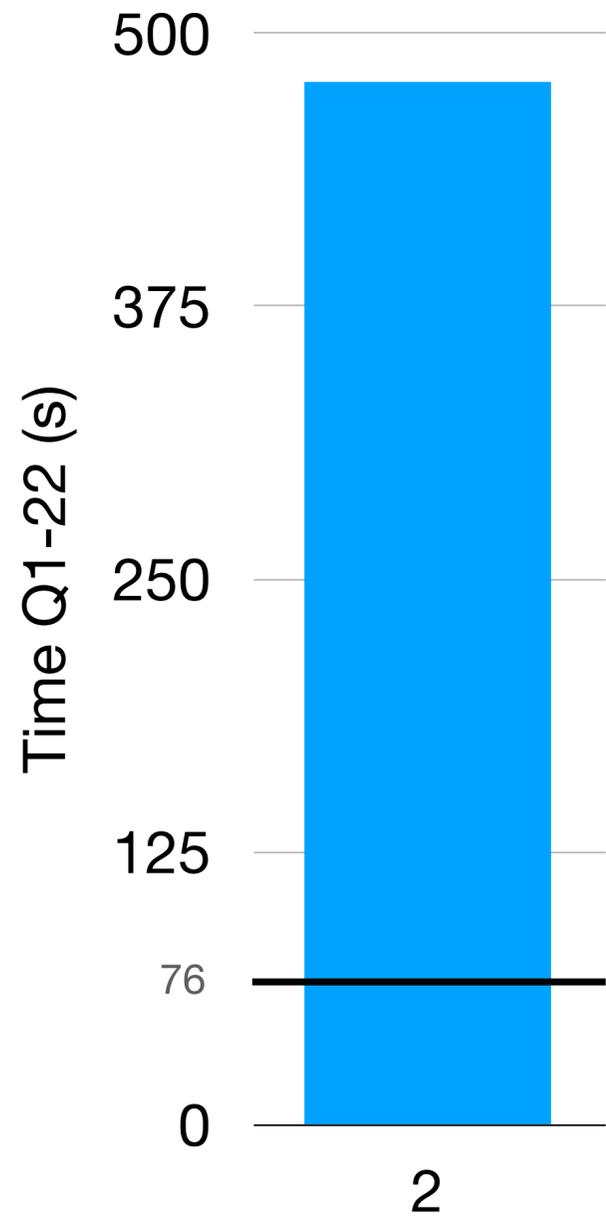
min	max
June 1	June 2

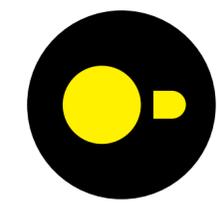
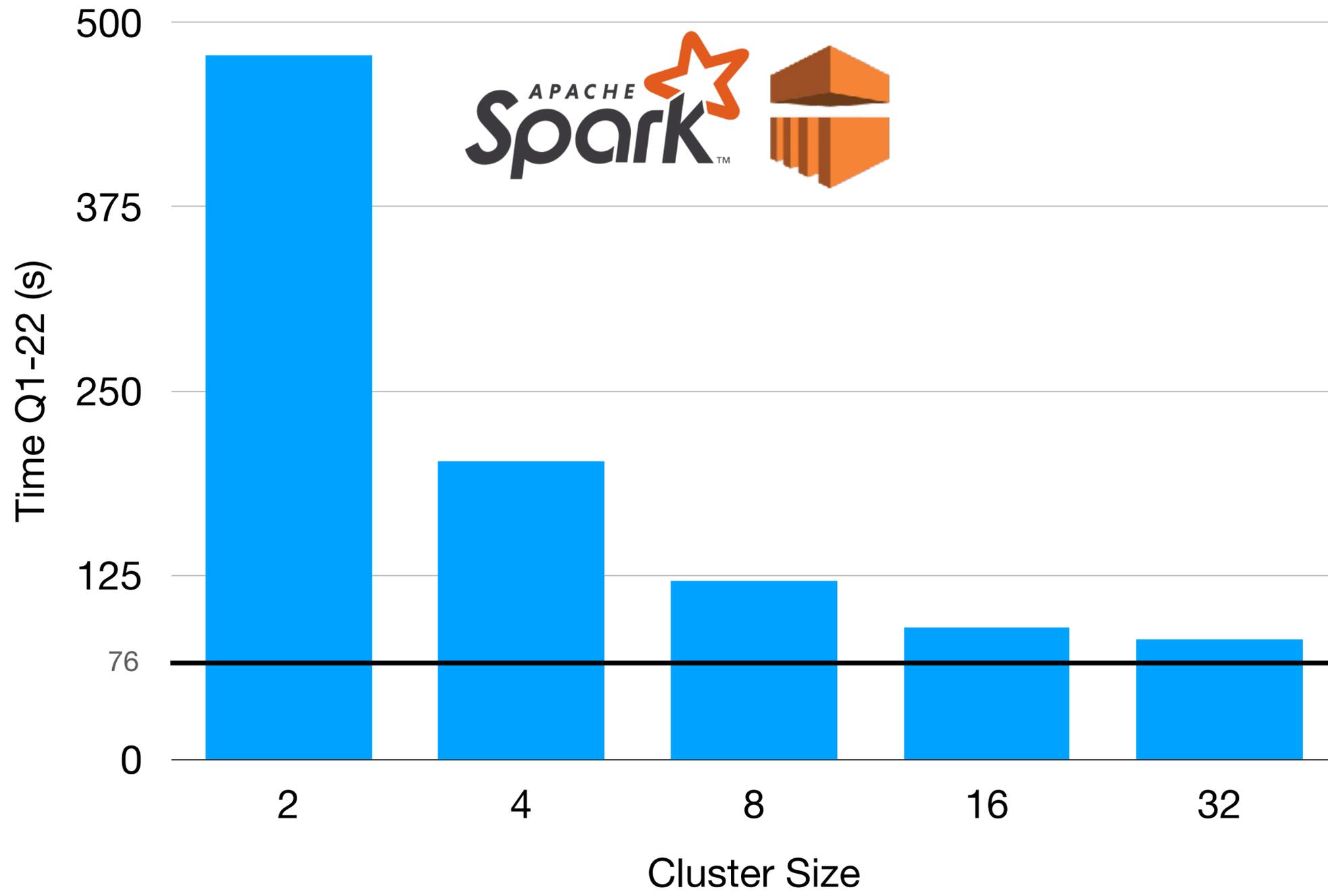
date	id	type	station
May 30			Ams C
May 30			Utrecht
May 31			Ams C
May 31			Schiphol
June 1			Schiphol
June 1			Utrecht
June 1			Utrecht
June 2			Ams C



Experiments!

TPC[®]





A Short Summary of the Last Decades of Data Management

- Tables are eternal
- NoSQL was a bad idea
- Relational systems eat most things
- Big data is dead
- DuckDB



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