



# Data Warehouses, Data Lakes, and Lakehouses (v1.01)

Week 2: September 9

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# Key Questions

What are the main differences between operational and analytical infrastructure?

What are data warehouses?

What problems did they evolve to solve?

What are data lakes and lakehouses?

What problems did they evolve to solve?

What are the components of modern data platforms?

How do operational and analytical data models differ?

What goes on in ETL/ELT?

How do different physical representations of data affect storage, compute, and other tradeoffs within data platforms?

# This Week

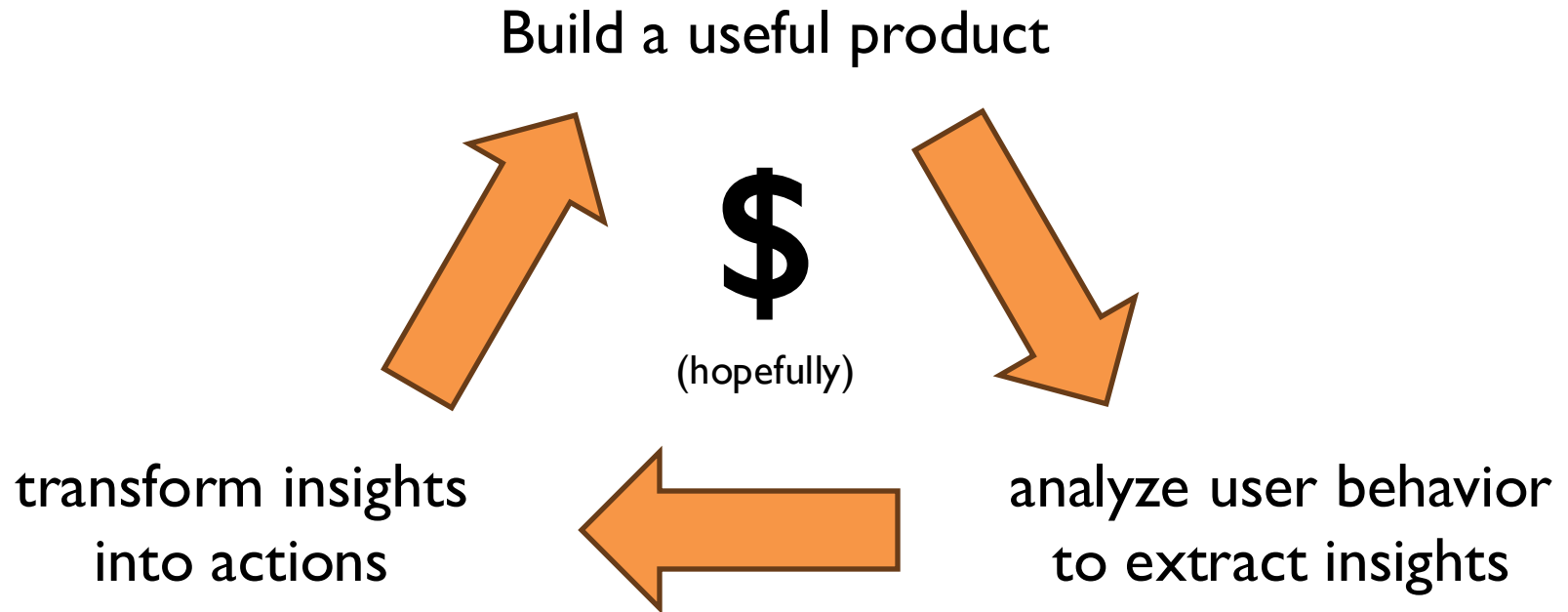
**Now:** Evolution of Data Platforms  
Data Warehouses, Data Lakes, and Lakehouses

Next: Three Deep Dives  
Data Modeling, ELT, Physical Representations

Context...

# The Data Flywheel

(a virtuous cycle)



Google. Facebook. Twitter. Amazon. Uber.

# Context...

## What's this course about?

The *infrastructure* that supports the data flywheel.

data platforms + data engineering

# Context...

## What problems do data platforms solve?

Ingesting, storing, manipulating, maintaining, serving...  
the data that supports the data flywheel.

# Evolution of Data Platforms

## Data Warehouses, Data Lakes, and Lakehouses

In the beginning...

users



```
graph TD; users --> MonolithicApplication[Monolithic Application];
```

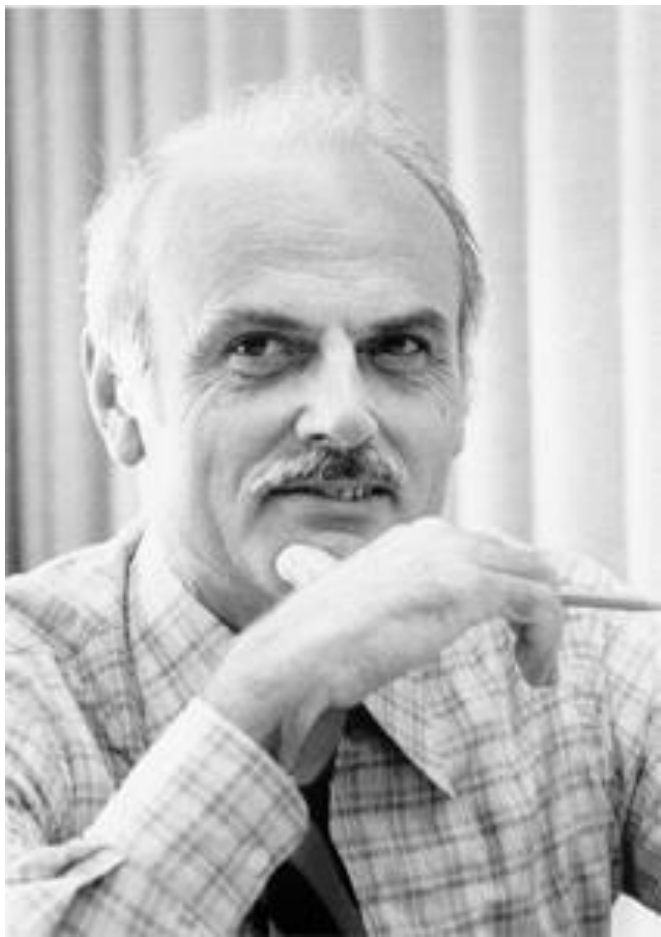
Monolithic  
Application

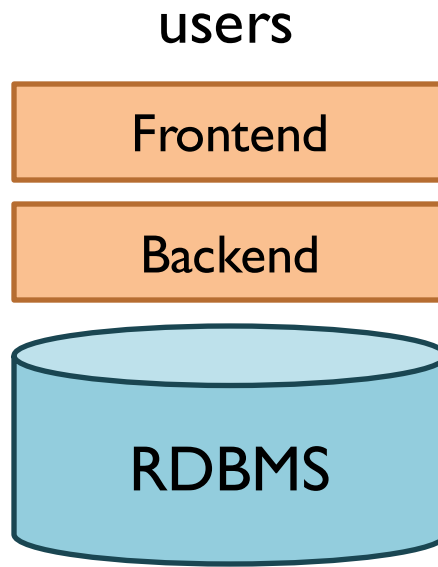


users

Frontend

Backend

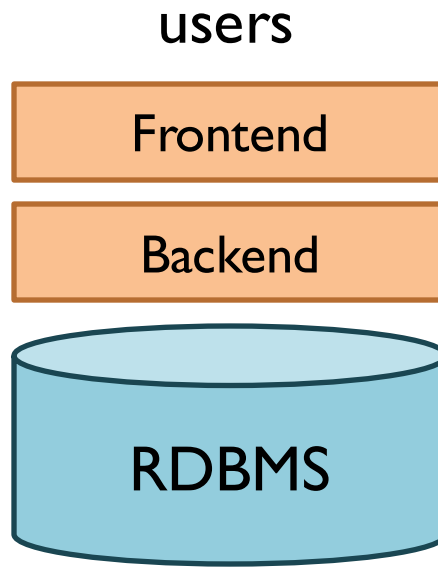




RDBMS = Relational Database Management System

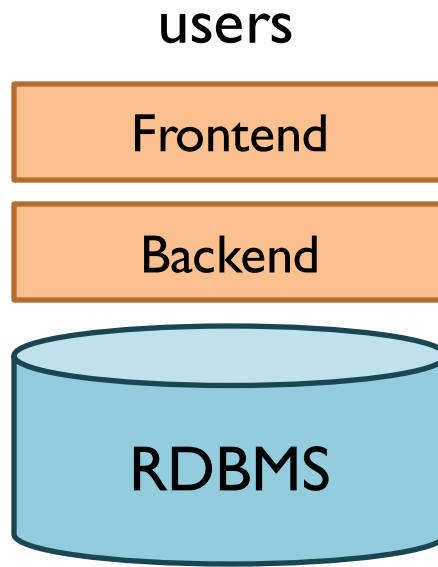
Imposes a relational view of data: tables, rows, columns

Provides a set of relational operators to manipulate data: SQL



## Why is this a good idea?

- Offload physical data design
- Standardize query processing
- Ensure data integrity, manage concurrency
- Handle backup and recovery



Okay, but  
*why relational?*

RDBMS = Relational Database Management System

Imposes a relational view of data: tables, rows, columns

Provides a set of relational operators to manipulate data: SQL

But we're constrained by the relational model?

(1) The relational model is surprisingly general.

(2) More later...

Remember from last time?

## Business Intelligence: Case Study

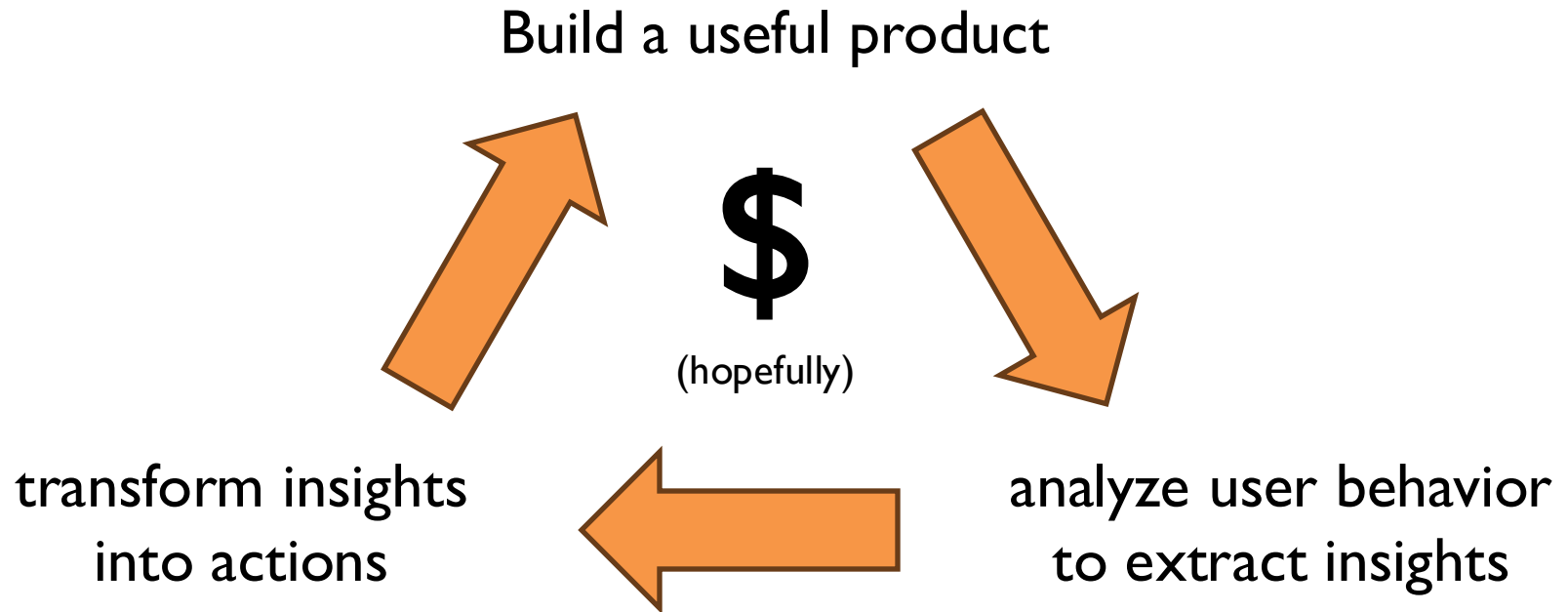
In the 1990s, Wal-Mart found that customers tended to buy diapers and beer together. So they put them next to each other and increased sales of both.\*

\* BTW, this is completely apocryphal. (But it makes a nice story.)

Remember from last time?

# The Data Flywheel

(a virtuous cycle)



Google. Facebook. Twitter. Amazon. Uber.

users

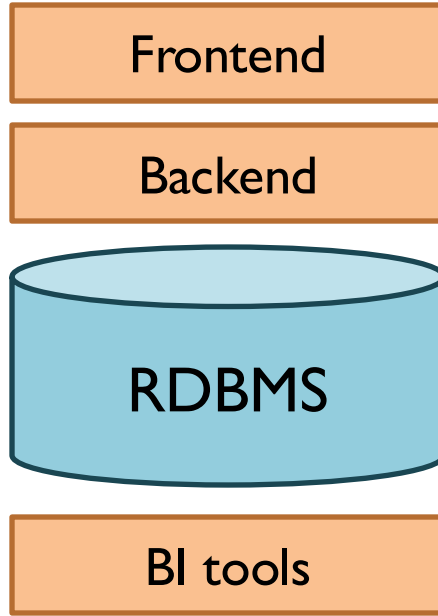
Frontend

Backend

RDBMS

BI tools

analysts





users

Frontend

Backend

RDBMS

BI tools

analysts



Why is my application so slow?

Why does my analysis take so long?



# RDBMS Workloads

## OLTP (online transaction processing)

Typical applications: e-commerce, banking, airline reservations

Customer-facing: real-time, low latency, highly-concurrent

Tasks: relatively small set of transactional queries; CRUD

Data access pattern: random reads, updates, writes (small amounts of data)

## OLAP (online analytical processing)

Typical applications: business intelligence, data mining

Back-end processing: batch workloads, less concurrency

Tasks: complex analytical queries, often ad hoc

Data access pattern: table scans, large amounts of data per query

# OLTP and OLAP Together?

Downsides of co-existing OLTP and OLAP workloads

Conflicting queries and data access patterns

Poor memory management

Variable latency



users and analysts

**Solution? Separate the two!**



Build a data warehouse!

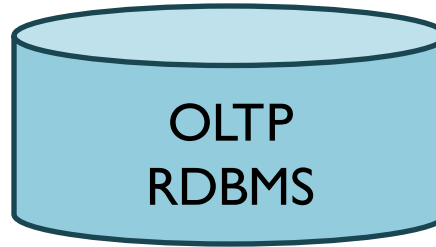


users

Frontend

Backend

OLTP database for  
user-facing transactions



ETL

(Extract, Transform, and Load)

*Where and when does  
this actually happen?*

*Deep Dive later...*

OLAP database for  
data warehousing

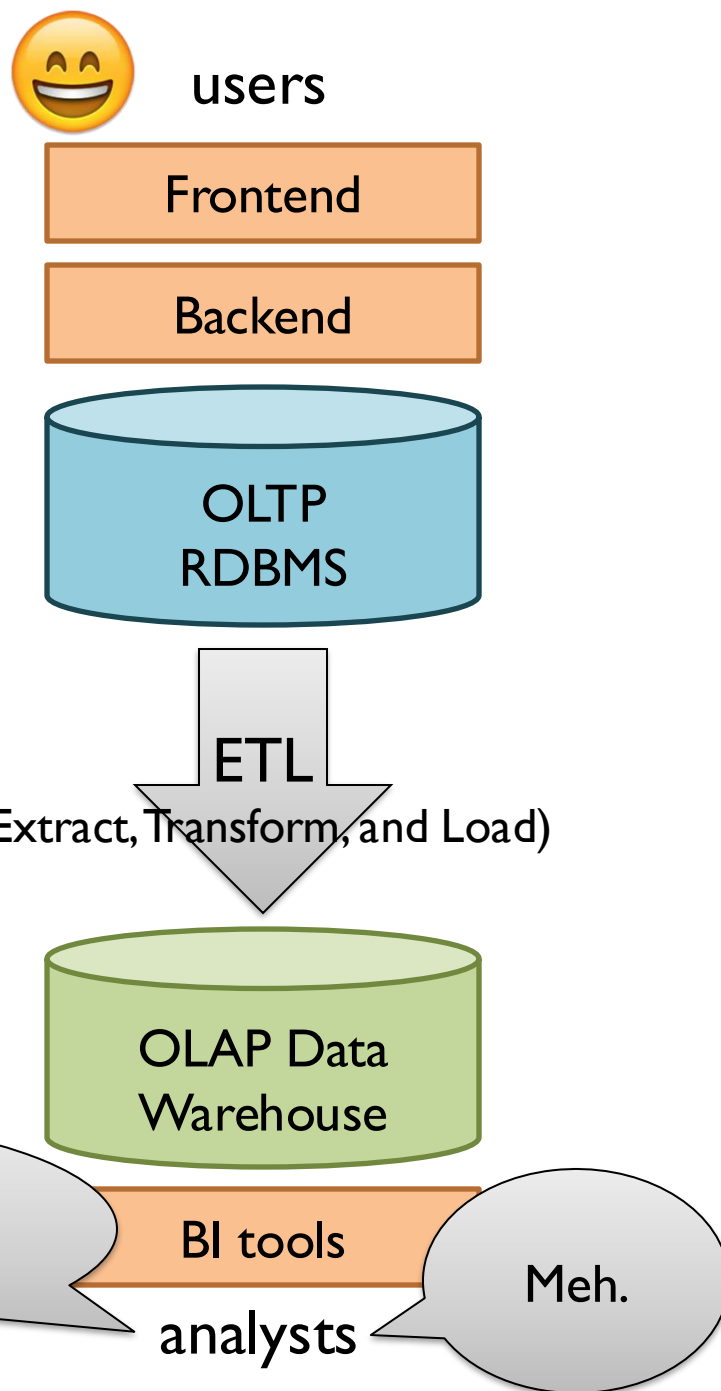


BI tools

analysts



Aside...

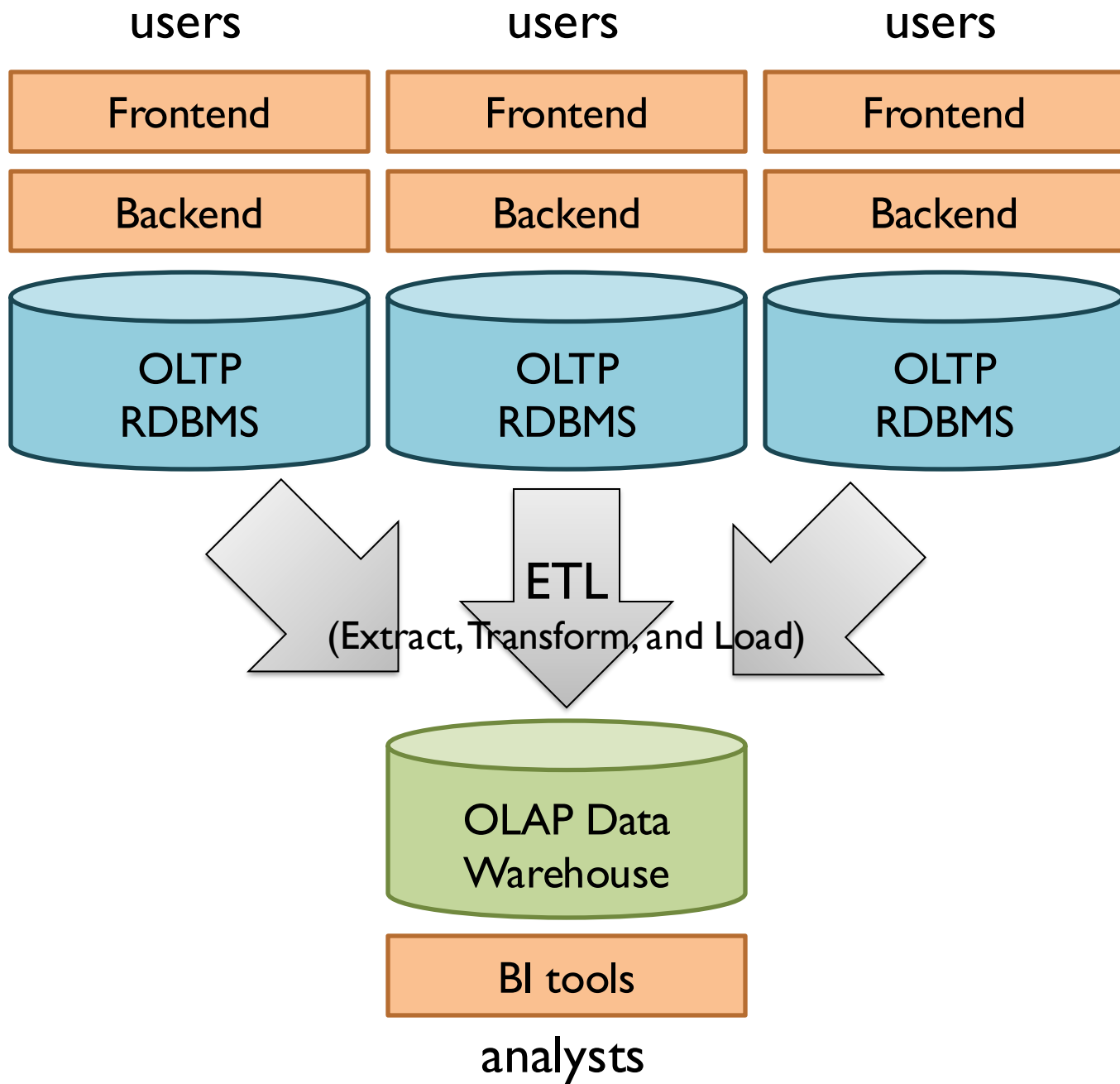


# Aside...

## Why is this a difficult problem?

Characteristics of the data: Volume, Velocity, Variety\* + Veracity

\* Coined by Gartner analyst Doug Laney in 2001.





# facebook®

Jeff Hammerbacher, Information Platforms and the Rise of the Data Scientist.  
In, *Beautiful Data*, O'Reilly, 2009.

“On the first day of logging the Facebook clickstream, more than 400 gigabytes of data was collected. The load, index, and aggregation processes for this data set really taxed the Oracle data warehouse. Even after significant tuning, we were unable to aggregate a day of clickstream data in less than 24 hours.”

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users

Frontend

Backend

OLTP  
RDBMS

Facebook context?

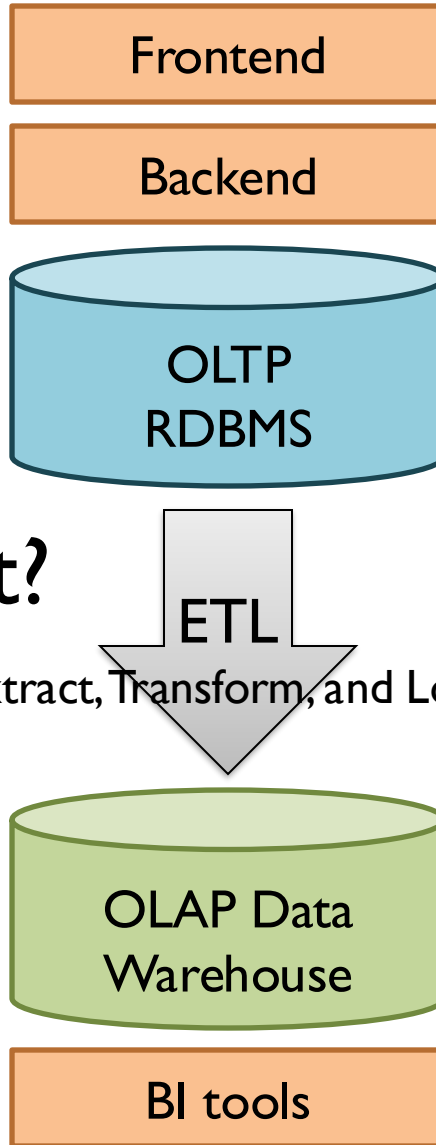
ETL

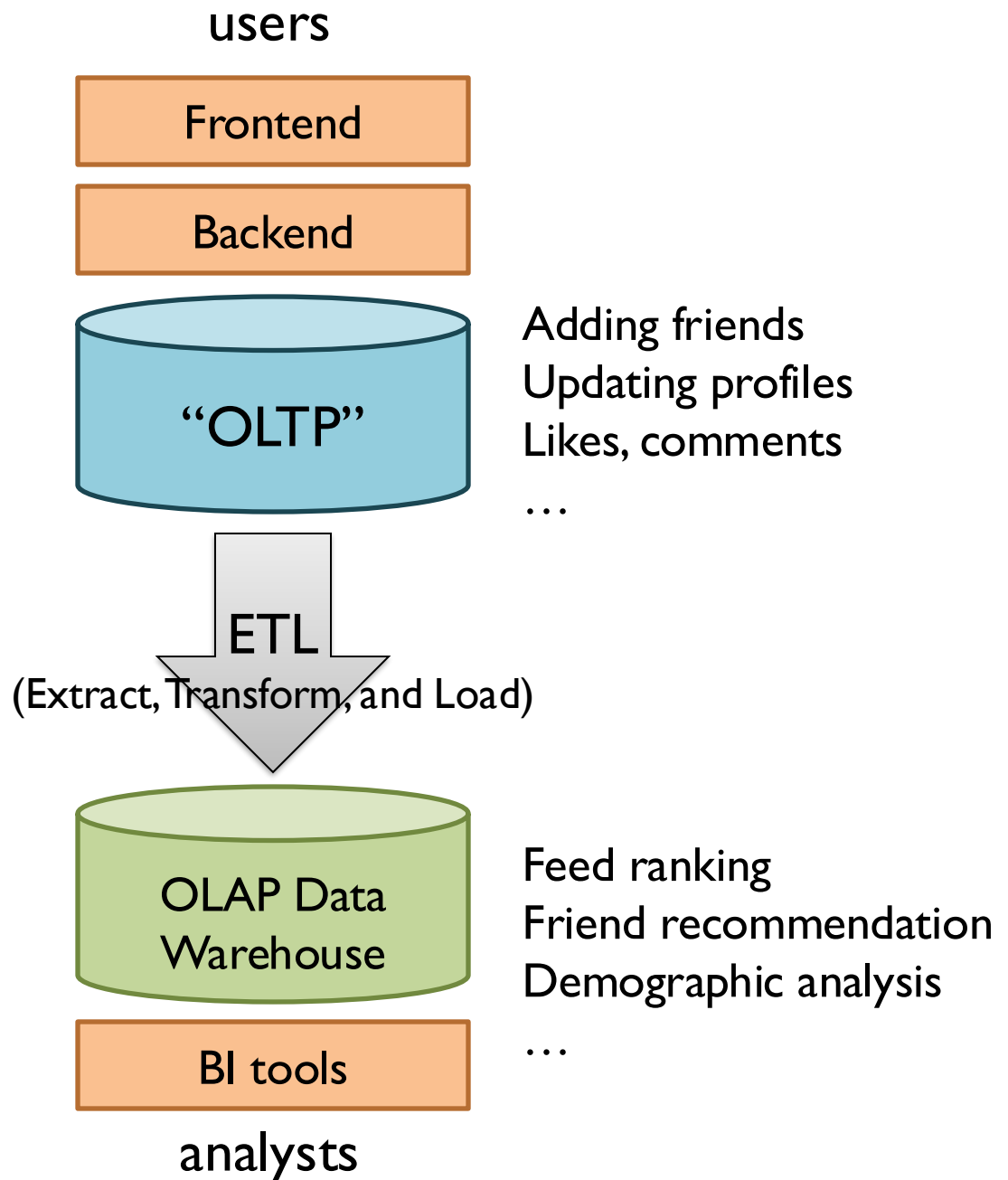
(Extract, Transform, and Load)

OLAP Data  
Warehouse

BI tools

analysts





users

Frontend

Backend

“OLTP”

PHP/MySQL

ETL or ELT?

(Extract, Transform, and Load)

Hadoop



~~analysts~~

data scientists

users

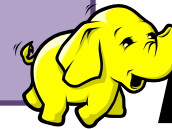
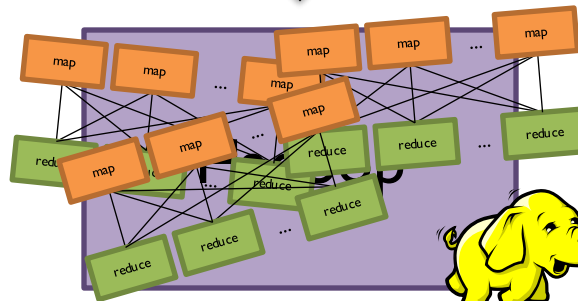
Frontend

Backend

“OLTP”

ELT

(Extract, Load, and Transform)

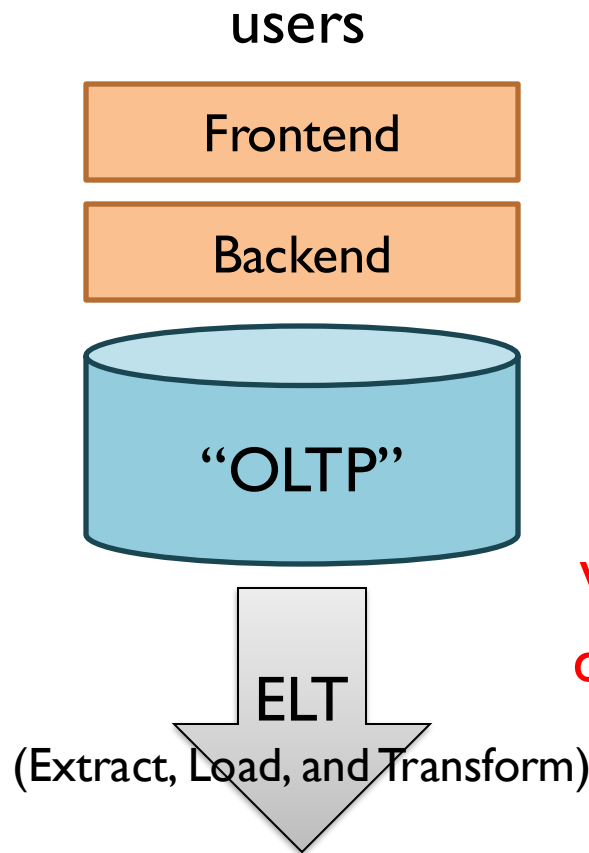


**hadoop**

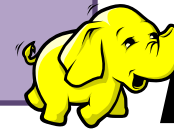
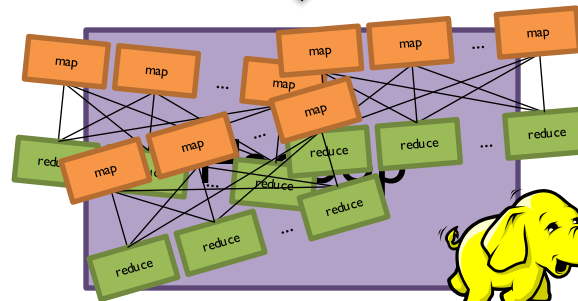
data scientists

Data storage on HDFS  
Processing with MapReduce

# The Irony...



Wait, so why not use a database to begin with?



data scientists

Data storage on HDFS  
Processing with MapReduce

# Why not just use an existing analytical database?

**Scalability. Cost. Flexibility.**

Jeff Hammerbacher, Information Platforms and the Rise of the Data Scientist.  
In, *Beautiful Data*, O'Reilly, 2009.

“On the first day of logging the Facebook clickstream, more than 400 gigabytes of data was collected. The load, index, and aggregation processes for this data set really taxed the Oracle data warehouse. Even after significant tuning, we were unable to aggregate a day of clickstream data in less than 24 hours.”



## Databases are great...

If your data has structure (and you know what the structure is)

If your data is reasonably clean

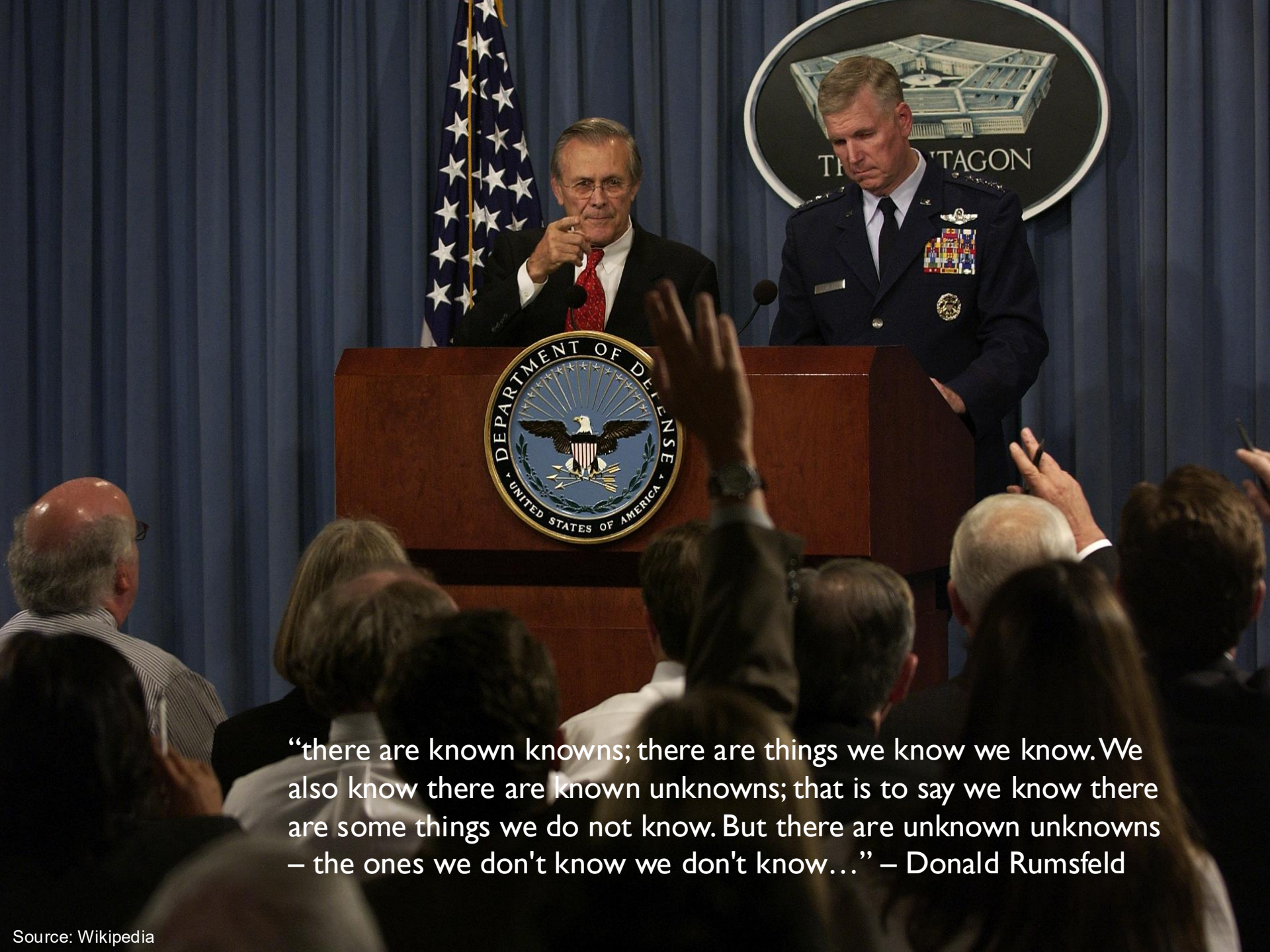
If you know what queries you're going to run ahead of time

## Databases are not so great...

If your data has little structure (or you don't know the structure)

If your data is messy and noisy

If you don't know what you're looking for



“there are known knowns; there are things we know we know. We also know there are known unknowns; that is to say we know there are some things we do not know. But there are unknown unknowns – the ones we don't know we don't know...” – Donald Rumsfeld

## Databases are great...

If your data has structure (and you know what the structure is)

If your data is reasonably clean

If you know what queries you're going to run ahead of time

**Known unknowns!**

## Databases are not so great...

If your data has little structure (or you don't know the structure)

If your data is messy and noisy

If you don't know what you're looking for

**Unknown unknowns!**

# It'd be great...

If I could just ingest all my data  
(relational, semi-structured, unstructured, graph, multimodal, etc.)  
in a multitude of formats  
(text, csv, json, etc.)  
and figure out what to do with it later.



Data storage on HDFS

(later, Amazon S3, Google Cloud Storage, Azure Blob Storage, etc.)

If I could then process / manipulate / transform the data...  
(I want to write *ad hoc* scripts **and** SQL queries)



Processing with MapReduce

(later, Spark, Presto, Trino, Flink, etc.)

tl;dr – I want *flexibility*.

# What is a data lake?

A data lake is a central location that holds a large amount of data in its native, raw format... a data lake uses a flat architecture and object storage to store the data... By leveraging inexpensive object storage and open formats, data lakes enable many applications to take advantage of the data.

<https://www.databricks.com/discover/data-lakes>

A data lake is a centralized repository that ingests and stores large volumes of data in its original form. The data can then be processed and used as a basis for a variety of analytic needs. Due to its open, scalable architecture, a data lake can accommodate all types of data from any source, from structured... to semi-structured... to unstructured... The data files are typically stored in staged zones—raw, cleansed, and curated—so that different types of users may use the data in its various forms to meet their needs. Data lakes provide core data consistency across a variety of applications, powering big data analytics, machine learning, predictive analytics, and other forms of intelligent action.

<https://azure.microsoft.com/en-us/resources/cloud-computing-dictionary/what-is-a-data-lake>

# What is a data lake?

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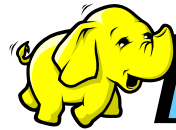
<https://azure.microsoft.com/en-us/resources/cloud-computing-dictionary/what-is-a-data-lake>



# Check?

## It'd be great...

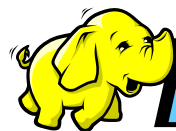
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**hadoop** Data storage on HDFS

(later, Amazon S3, Google Cloud Storage, Azure Blob Storage, etc.)

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**hadoop** Processing with MapReduce

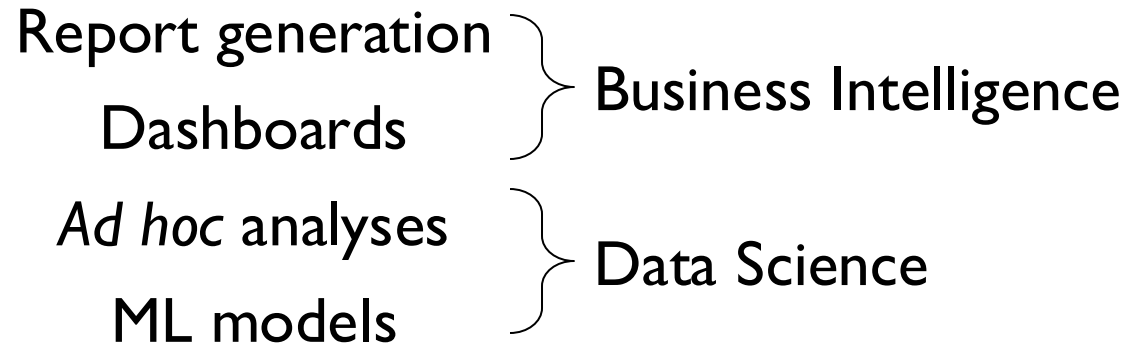
(later, Spark, Presto, Trino, Flink, etc.)



Reminder...

# Transform Insights into Actions

What does that really mean?



known unknowns and unknown unknowns?

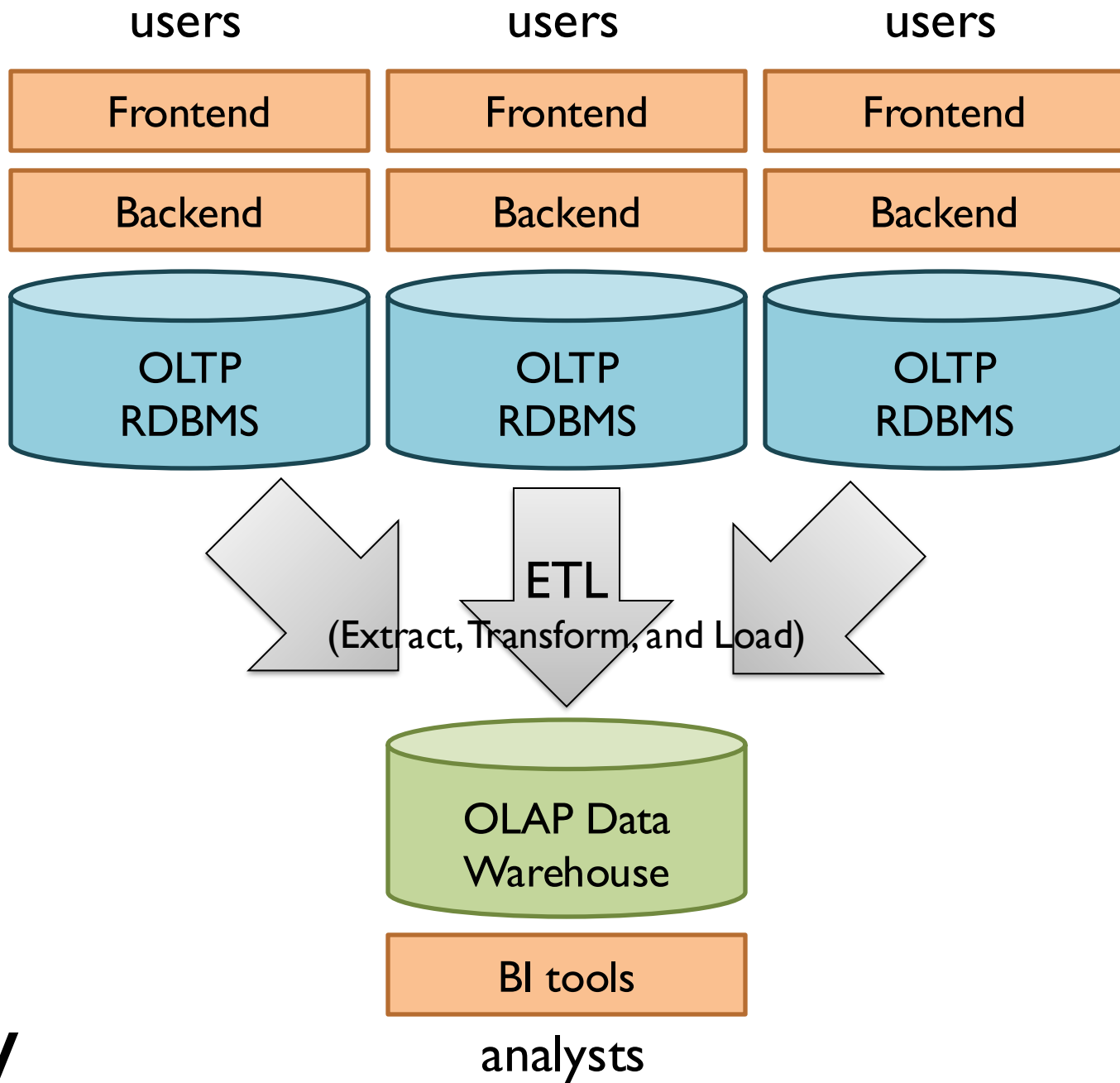
# EDL vs. EDW

All types of data  
vs. only relational data

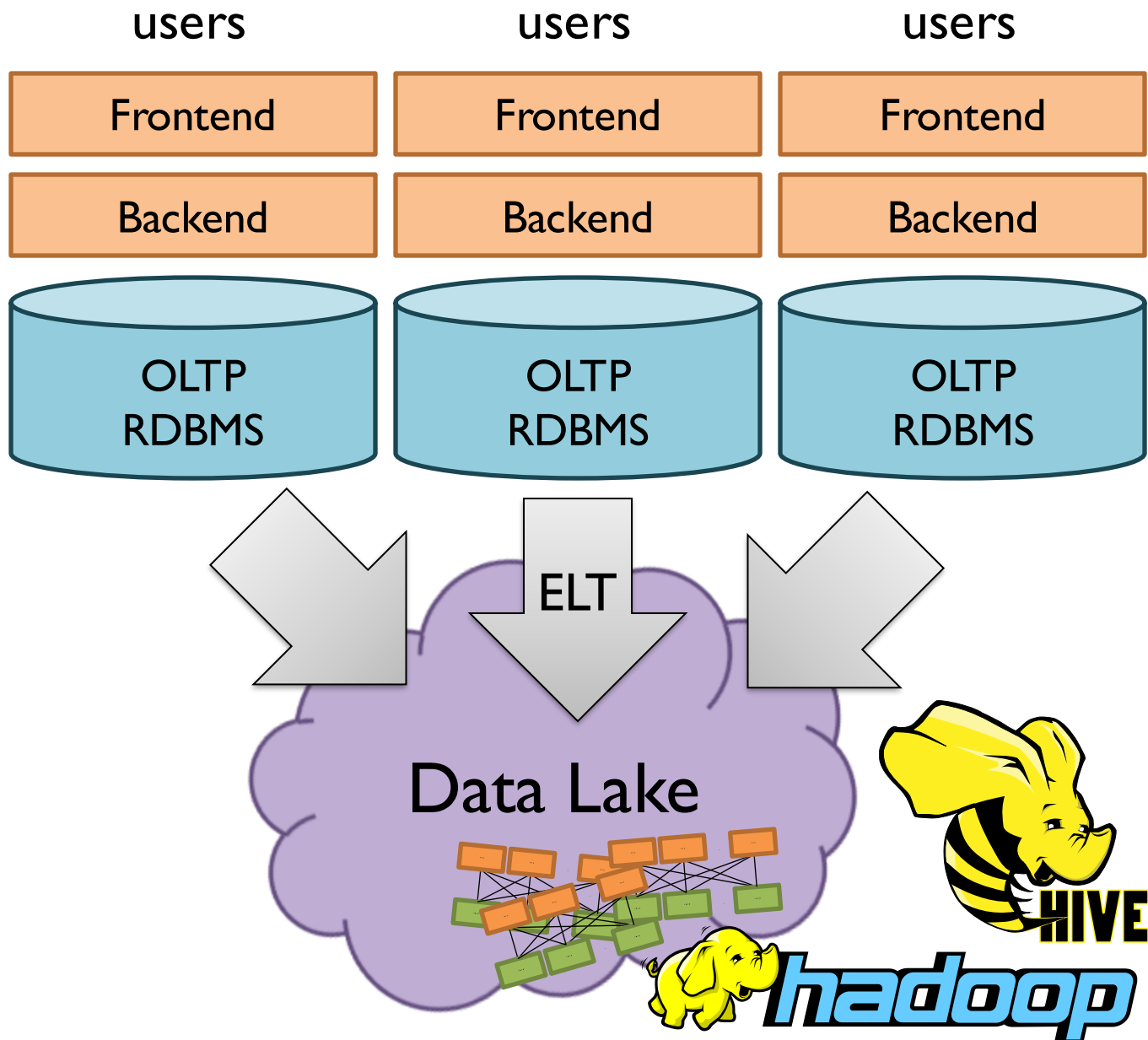
Open data formats  
vs. proprietary formats

Cheap (object) storage  
vs. expensive file systems

Flexible query processing  
vs. only SQL



EDW



EDL (1<sup>st</sup> gen)

data scientists

# Data Lakes to Data Swamps

tl;dr – I want *flexibility*.

Be careful what you wish for...

Why can't I find anything?

Where's the schema?

Why are there all these multiple near-duplicate copies?

Where did this come from?

Did this job actually finish?

Am I supposed to have access to this?

“jank”

# The “Lakehouse” Vision

A unified metadata and governance layer over data lakes

Transaction support

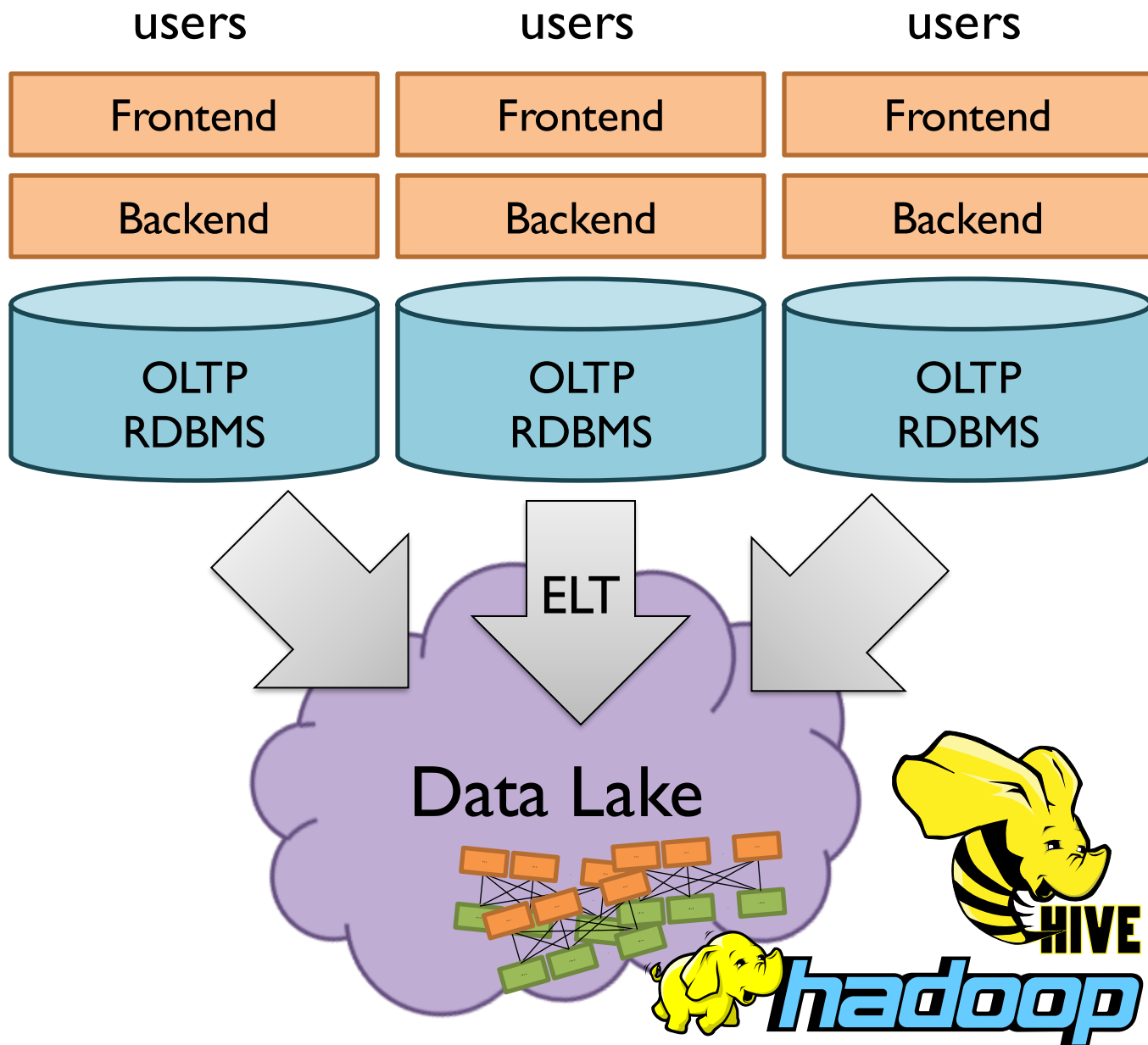
Schema enforcement and governance

Storage is decoupled from compute

Openness

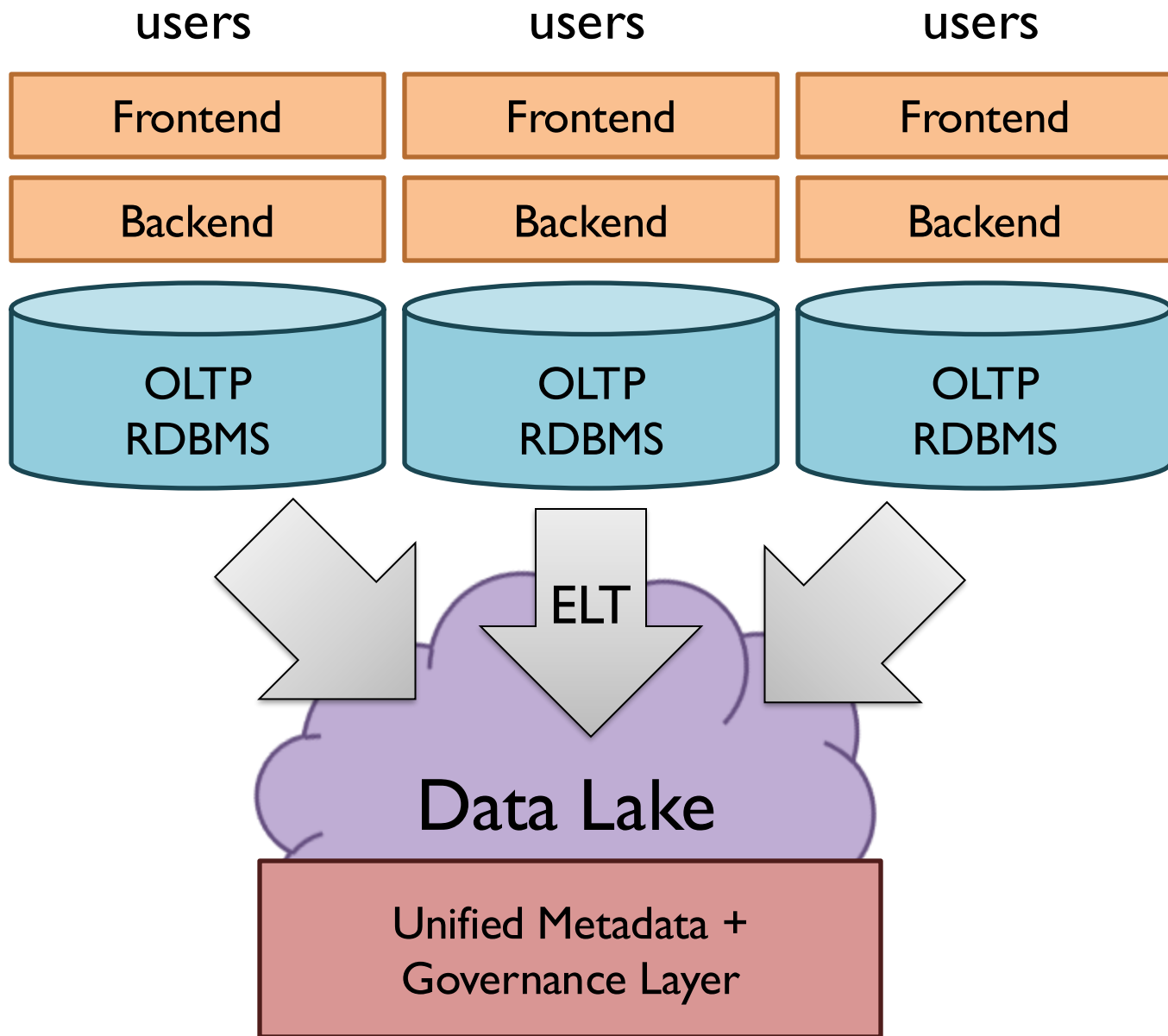
Support for diverse data types

Support for diverse workloads



EDL (1<sup>st</sup> gen)

data scientists



# Lakehouse





# EDL vs. EDW

## Revenge of the DBs

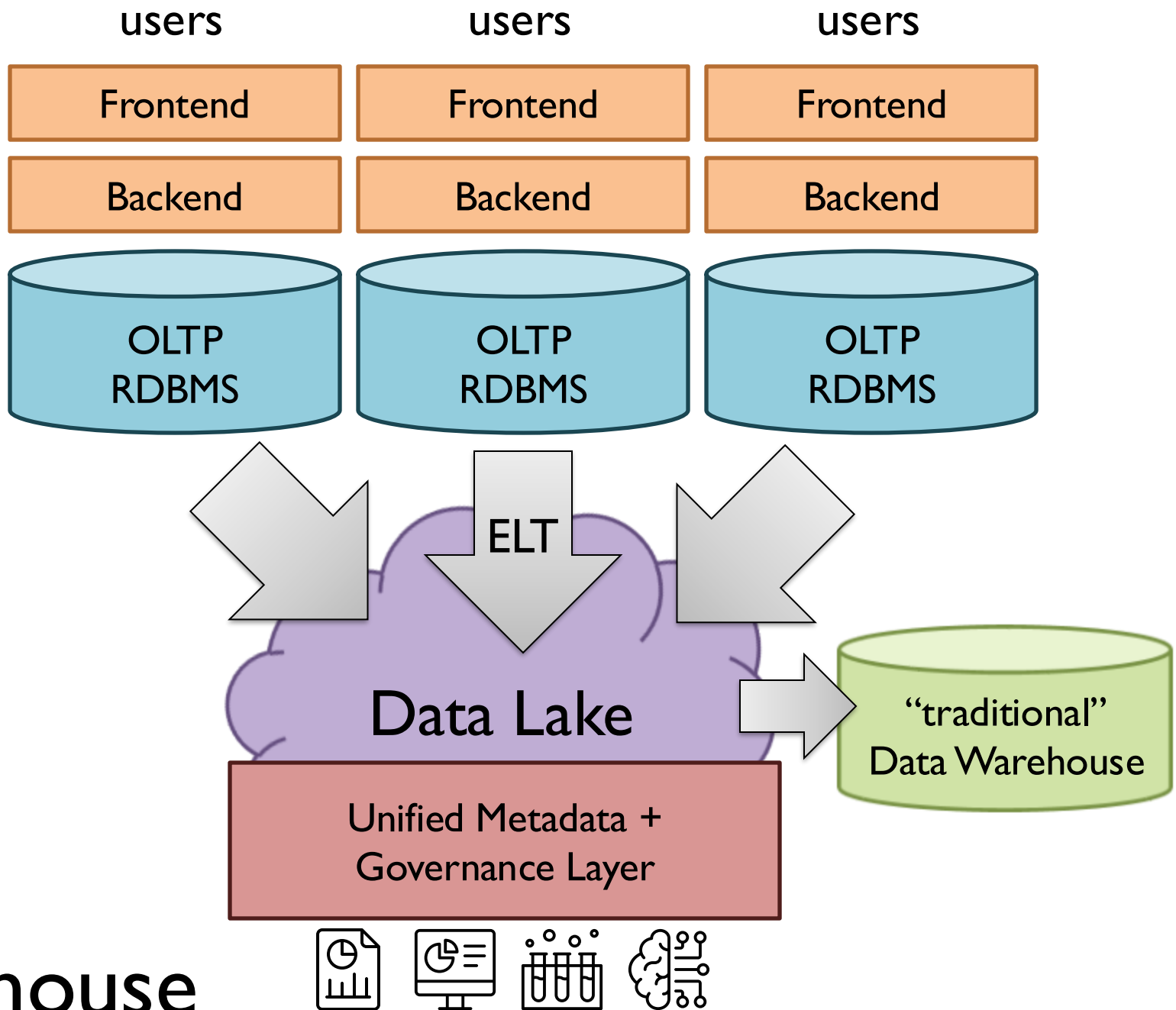
All types of data ✓  
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Open data formats ✓  
vs. proprietary formats

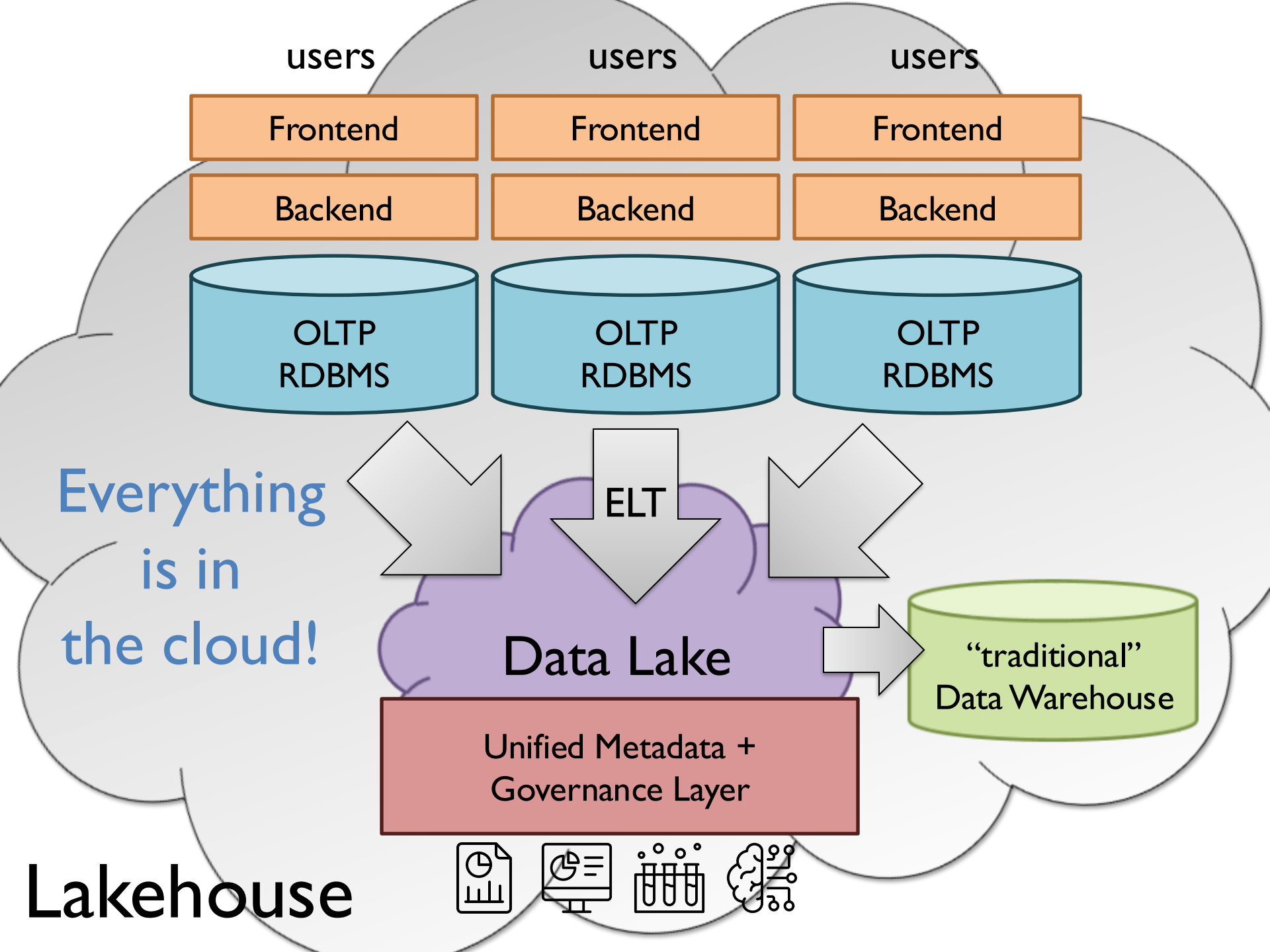
Cheap (object) storage ✓  
vs. expensive file systems

Flexible query processing ✓  
vs. only SQL

Sometimes, I really just want a RDBMS!



# Lakehouse



users

users

users

Frontend

Frontend

Frontend

Backend

Backend

Backend

OLTP  
RDBMS

OLTP  
RDBMS

OLTP  
RDBMS

Everything  
is in  
the cloud!

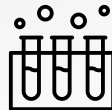
ELT

Data Lake

Unified Metadata +  
Governance Layer

"traditional"  
Data Warehouse

Lakehouse

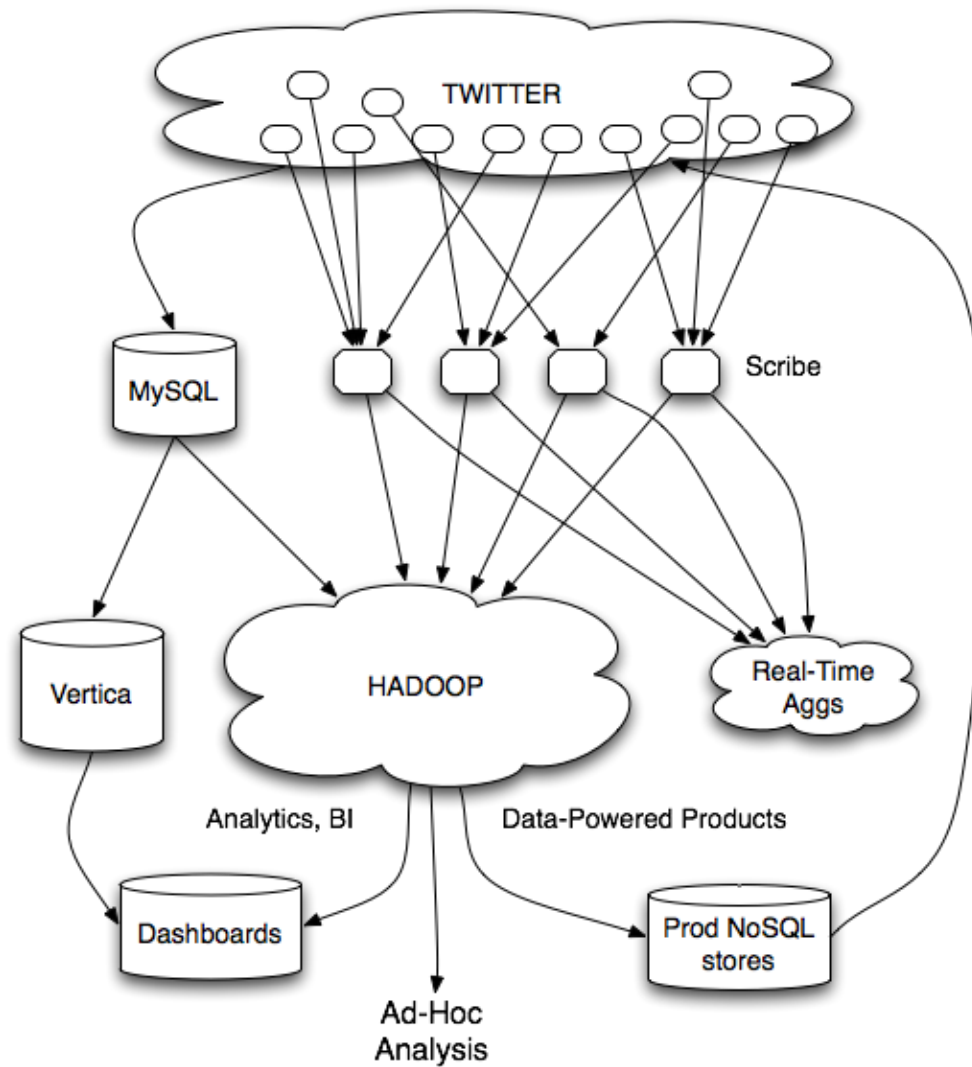


# The Cloud

✓ Pro – you don't have to worry about it.

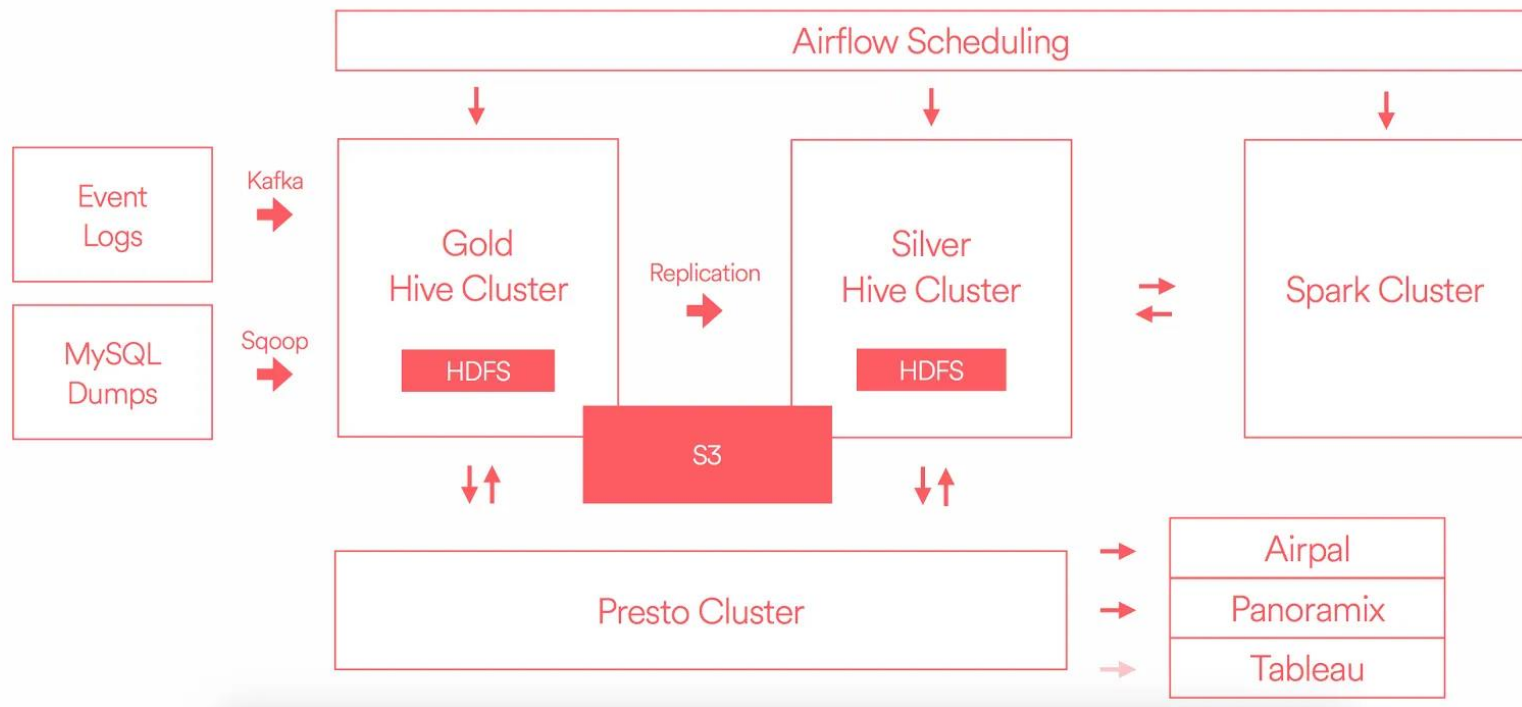
Con – you don't have to worry about it. ~~X~~

(More on this the coming weeks...)



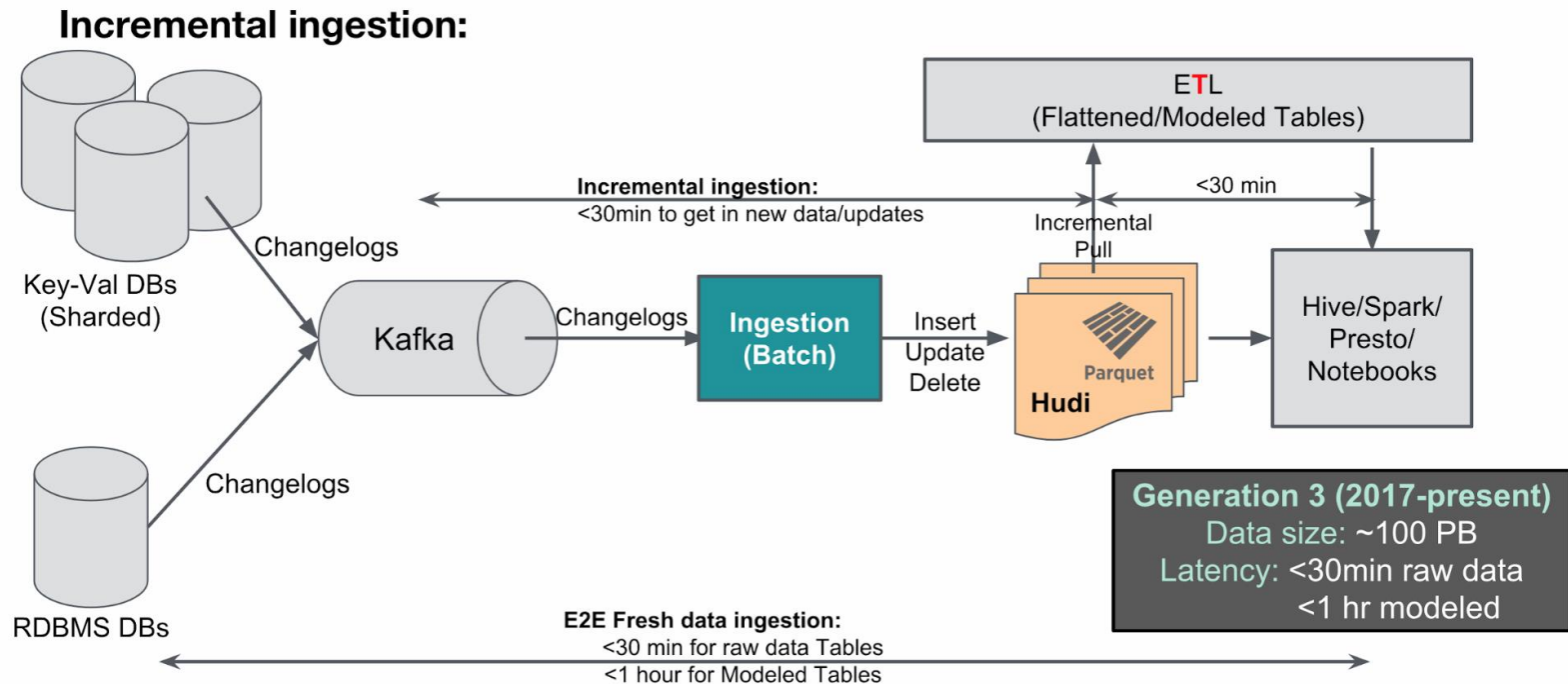
Twitter's data platform (circa 2012)

# AIRBNB DATA INFRA



AirBnB's data platform (circa 2016)

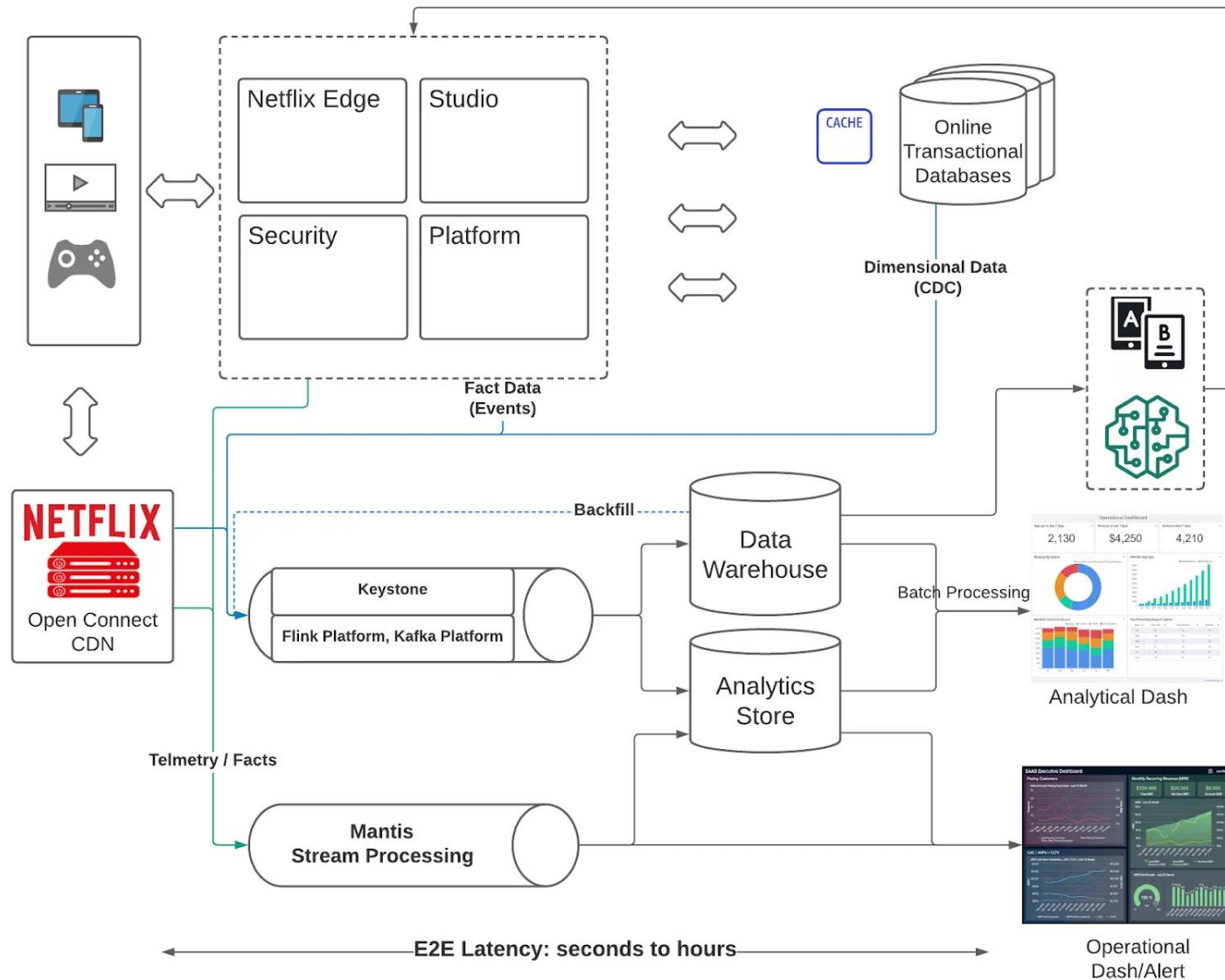
## Generation 3 (2017-present) - Let's rebuild for long term



## Uber's data platform (circa 2018)

<https://www.uber.com/en-CA/blog/uber-big-data-platform/>

## How Stream Processing fit in Netflix (2021)



## Netflix's data platform (circa 2021)





## Spotify's data platform (circa 2024)

<https://engineering.atspotify.com/2024/5/data-platform-explained-part-ii>



富嶽三十六景 神奈川  
浪裏

