

Data-Intensive Distributed Computing

CS 431/631 451/651 (Fall 2019)

Part 5: Analyzing Relational Data (1/3) October 10, 2019

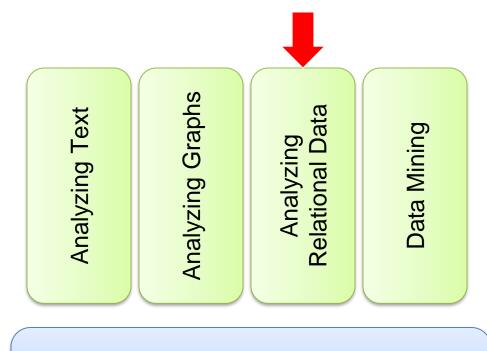
Ali Abedi

These slides are available at https://www.student.cs.uwaterloo.ca/~cs451



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Structure of the Course



"Core" framework features and algorithm design

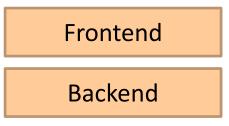
Evolution of Enterprise Architectures

Next two sessions: techniques, algorithms, and optimizations for relational processing

users

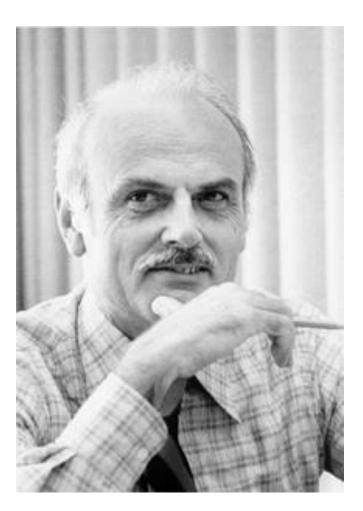
Monolithic Application

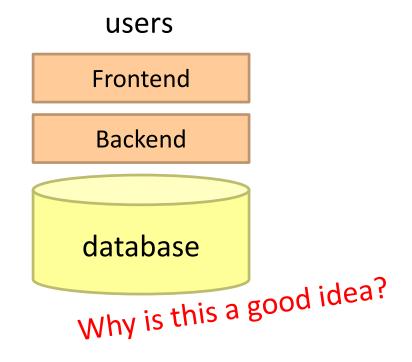
users



Edgar F. Codd

- Inventor of the relational model for DBs
- SQL was created based on his work
- Turing award winner in 1981

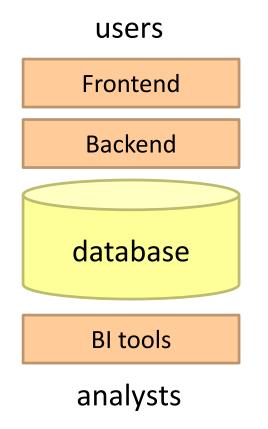


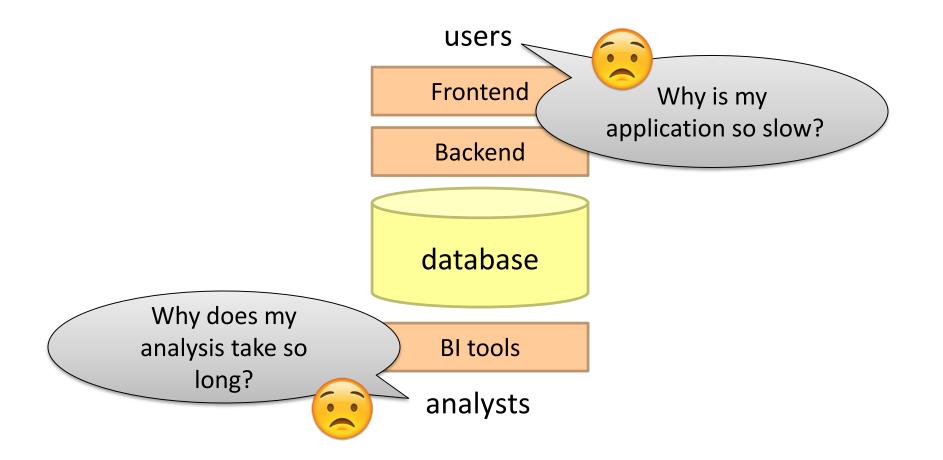


Business Intelligence

An organization should retain data that result from carrying out its mission and exploit those data to generate insights that benefit the organization, for example, market analysis, strategic planning, decision making, etc.







Database Workloads

OLTP (online transaction processing)

Typical applications: e-commerce, banking, airline reservations User facing: real-time, low latency, highly-concurrent Tasks: relatively small set of "standard" transactional queries 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

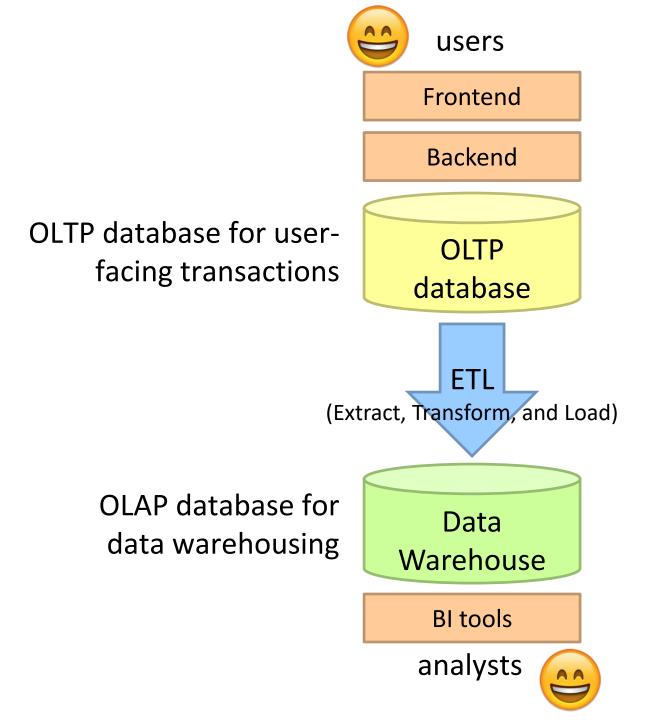
Poor memory management Conflicting data access patterns Variable latency



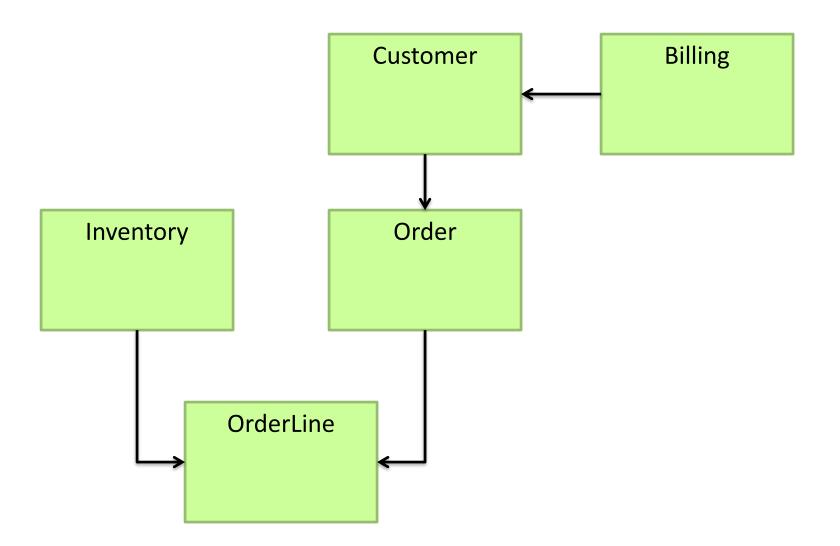
users and analysts

Solution?

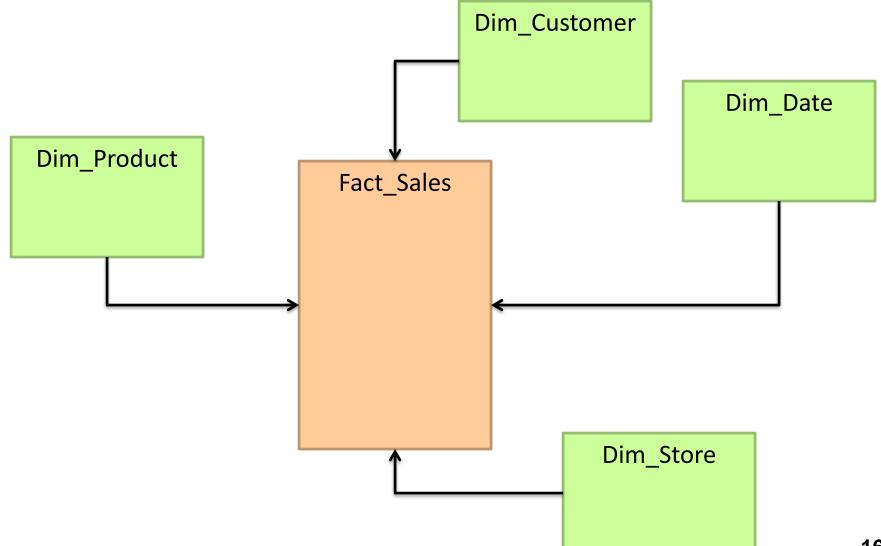
Build a data warehouse!



A Simple OLTP Schema



A Simple OLAP Schema



ETL

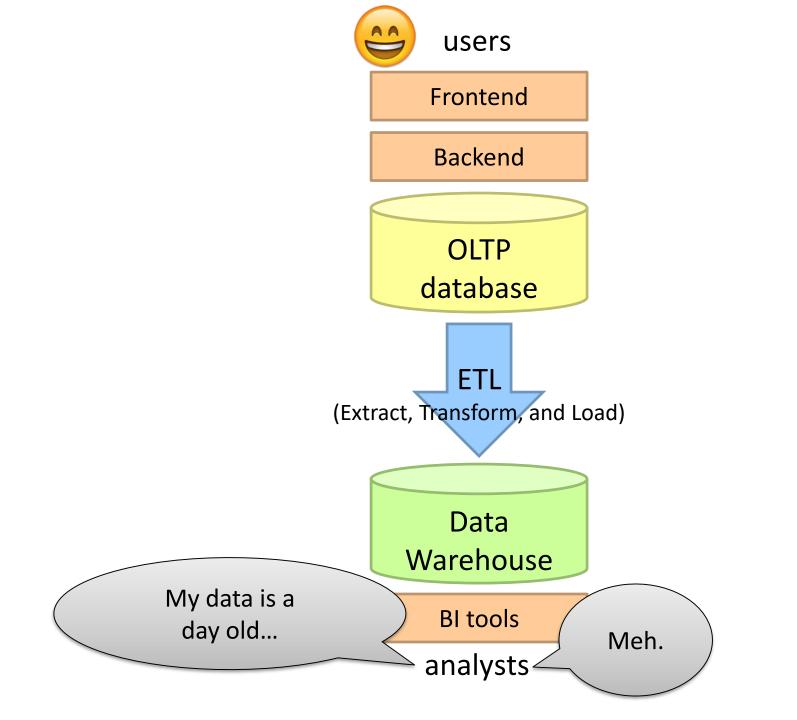
Extract

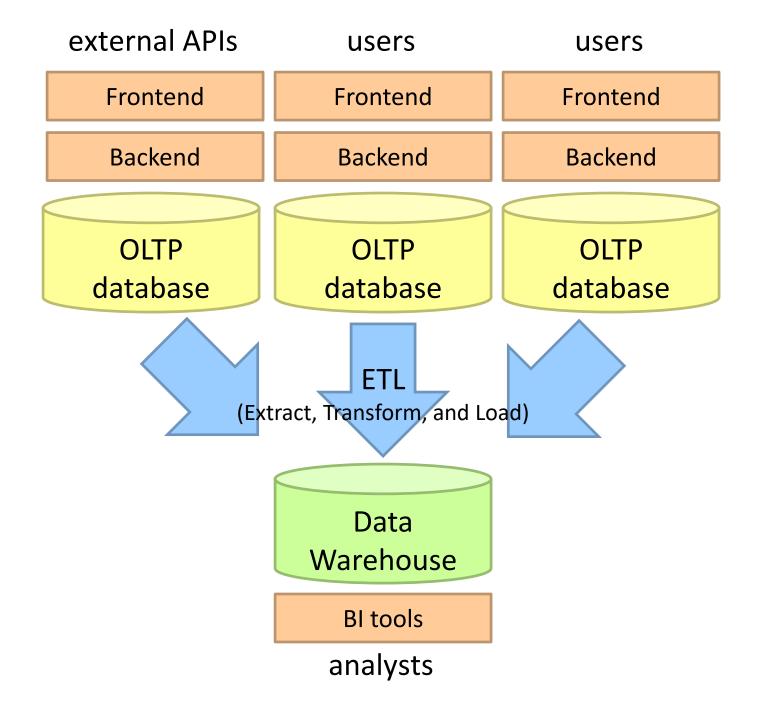
Transform

Data cleaning and integrity checking Schema conversion Field transformations

Load

When does ETL happen?

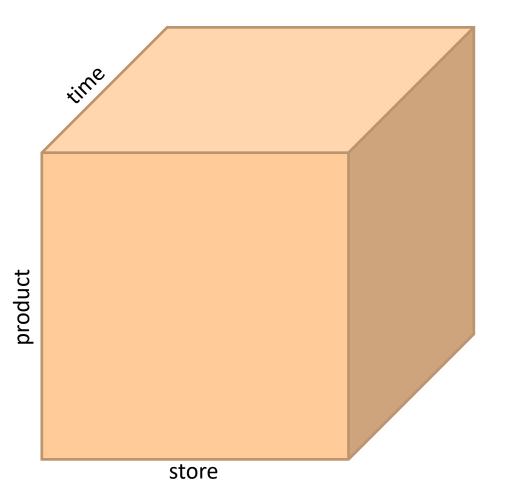




What do you actually do?

Report generation Dashboards Ad hoc analyses

OLAP Cubes

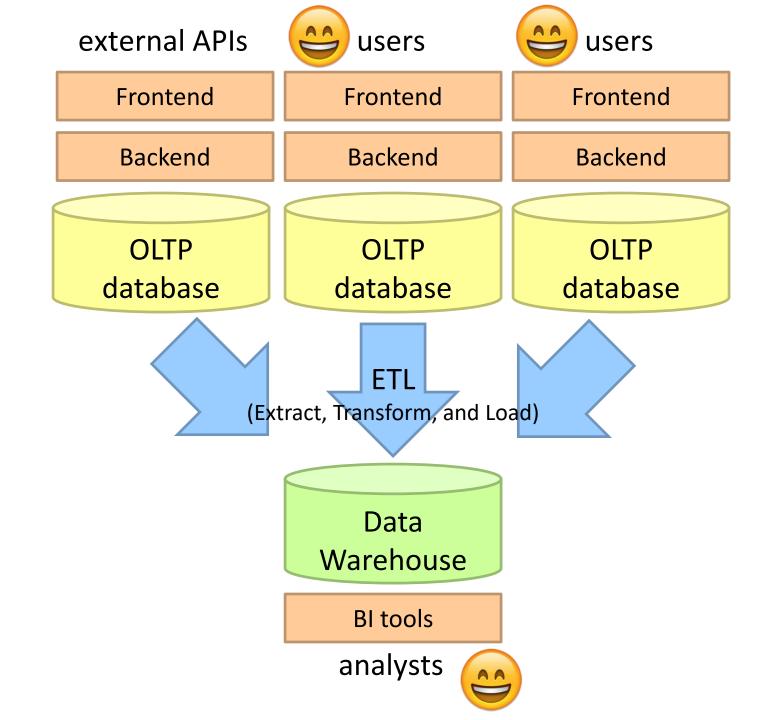


Common operations slice and dice roll up/drill down pivot

OLAP Cubes: Challenges

Fundamentally, lots of joins, group-bys and aggregations How to take advantage of schema structure to avoid repeated work?

> Cube materialization Realistic to materialize the entire cube? If not, how/when/what to materialize?

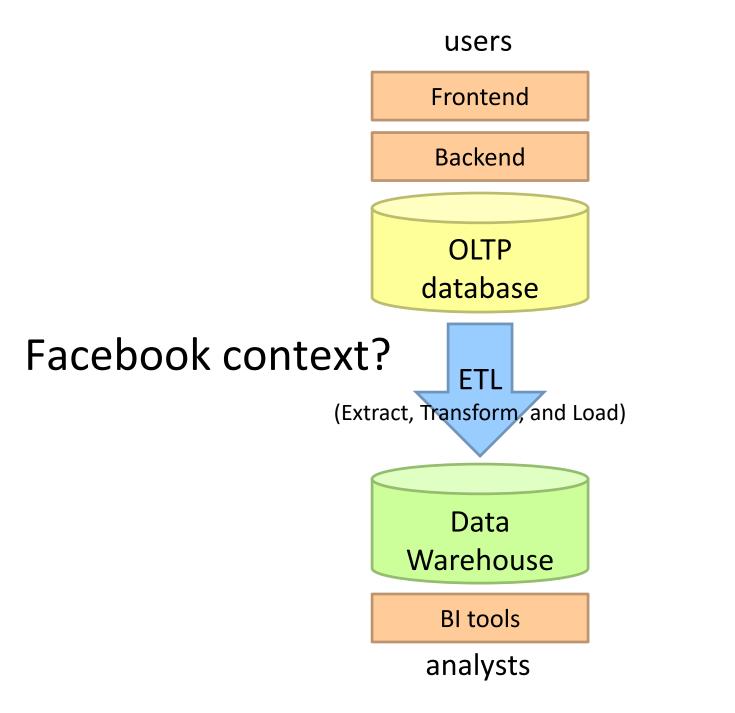


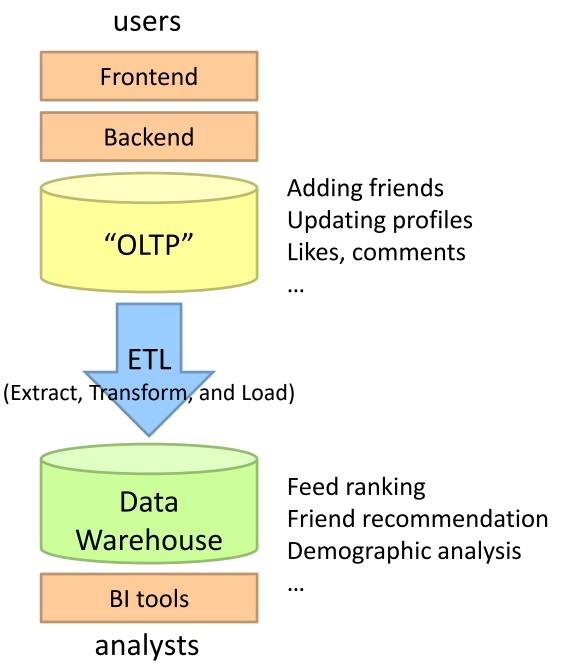
Fast forward...

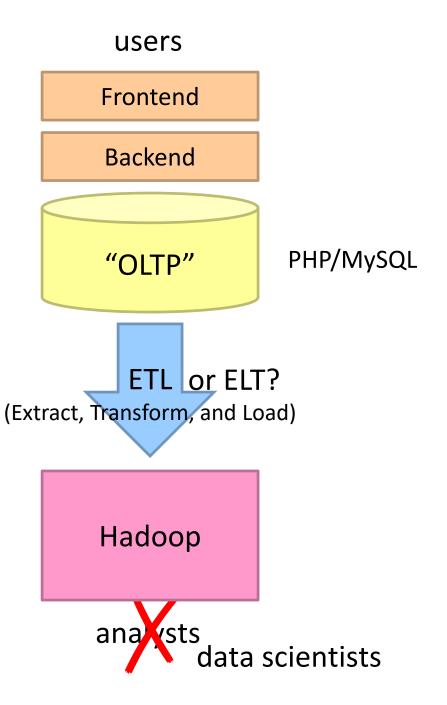
facebook.

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."









Cheaper to store everything

What's changed?

Dropping cost of disks Cheaper to store everything than to figure out what to throw away

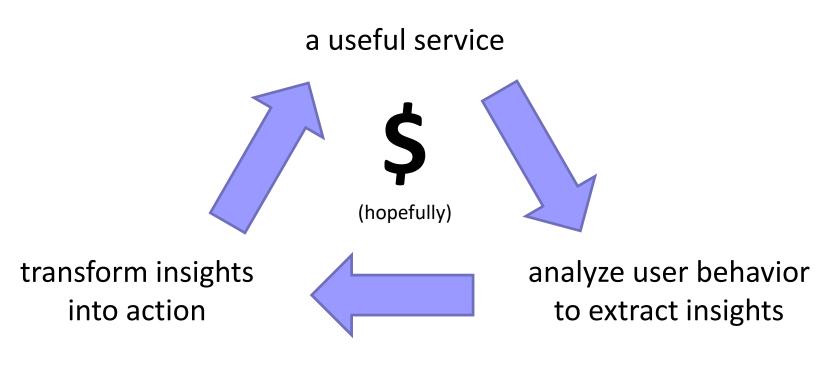
Types of data collected

From data that's obviously valuable to data whose value is less apparent

Rise of social media and user-generated content Large increase in data volume

Growing maturity of data mining techniques Demonstrates value of data analytics

Virtuous Product Cycle



Google. Facebook. Twitter. Amazon. Uber.

What do you actually do?

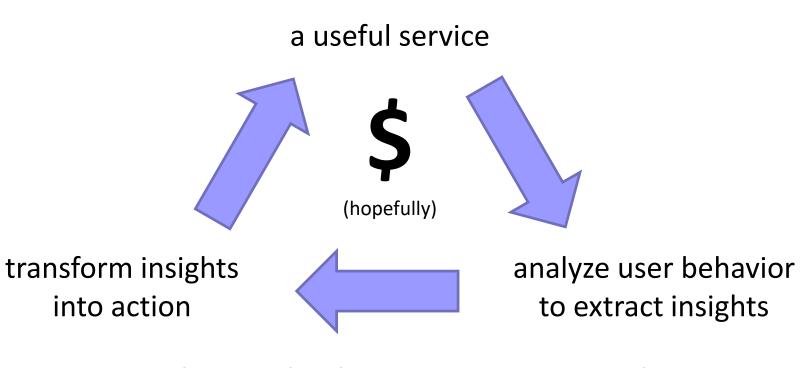
Report generation

Dashboards

Ad hoc analyses "Descriptive" "Predictive"

Data products

Virtuous Product Cycle



Google. Facebook. Twitter. Amazon. Uber.

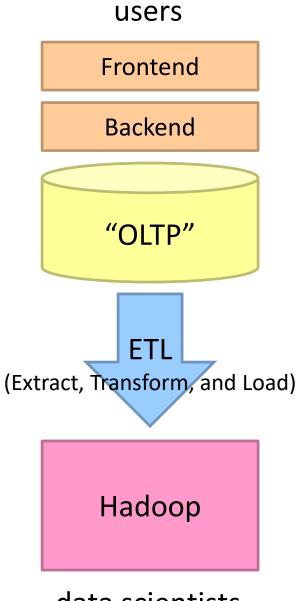
data products

data science

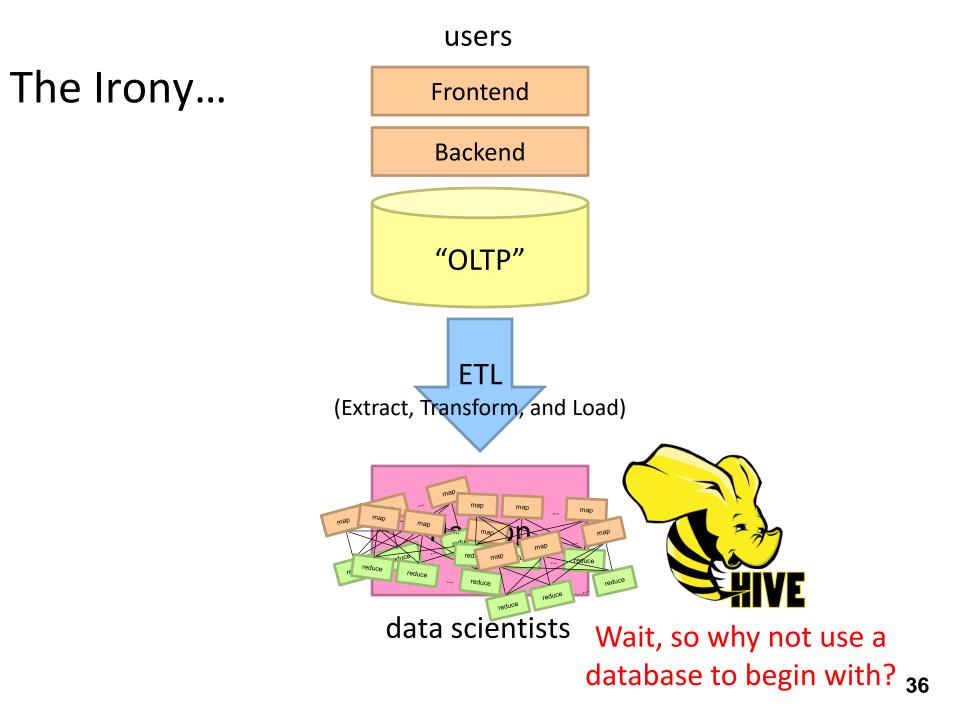
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data scientists



Why not just use a database? SQL is awesome

Scalability. Cost.

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

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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!

Advantages of Hadoop dataflow languages

Don't need to know the schema ahead of time Raw scans are the most common operations Many analyses are better formulated imperatively Much faster data ingest rate

What do you actually do?

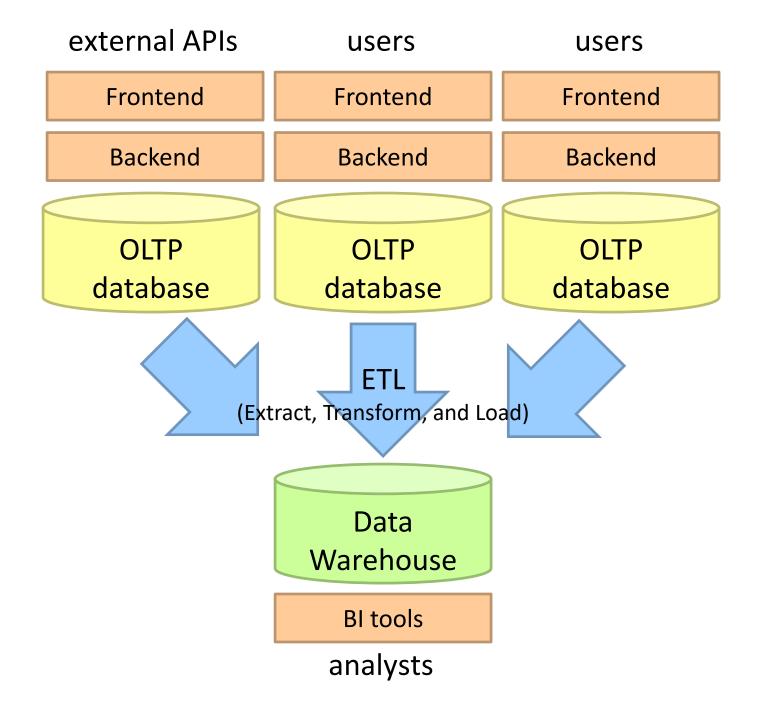
Report generation

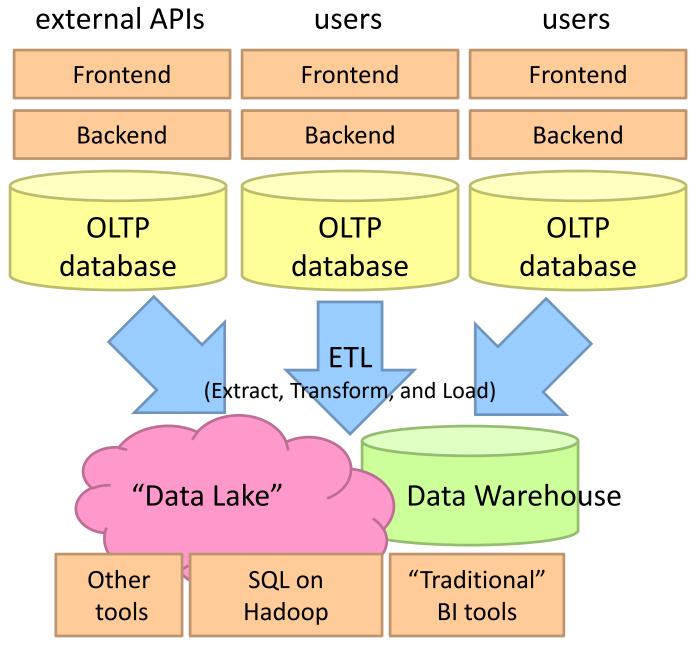
Dashboards

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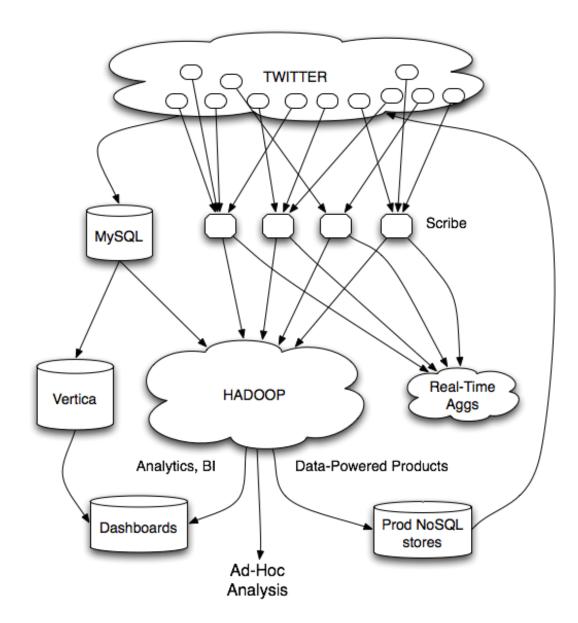
Data products

Which are known unknowns and unknown unknowns?





data scientists



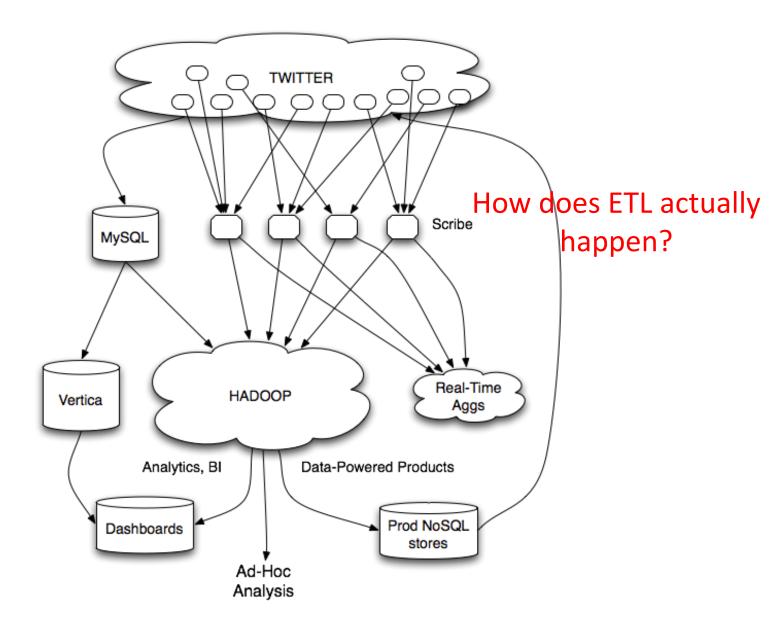
Twitter's data warehousing architecture (2012)

~2010

~150 people total ~60 Hadoop nodes ~6 people use analytics stack daily

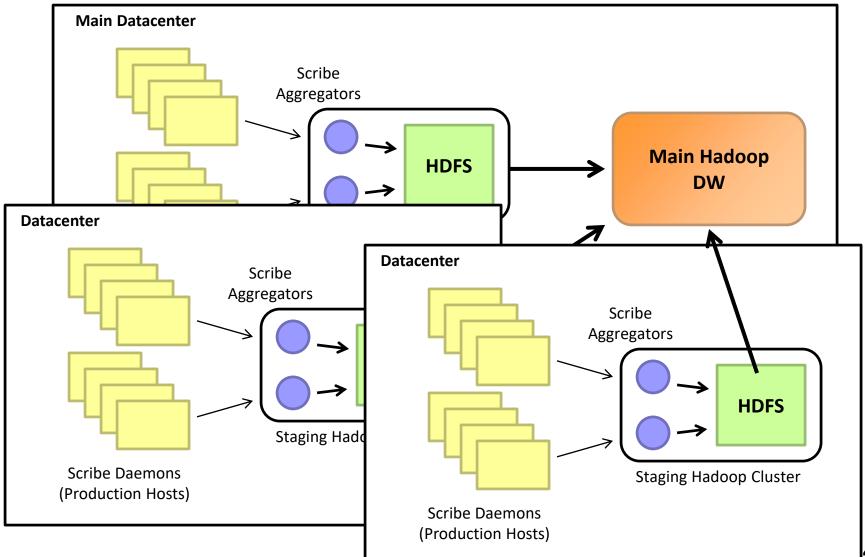
~2012

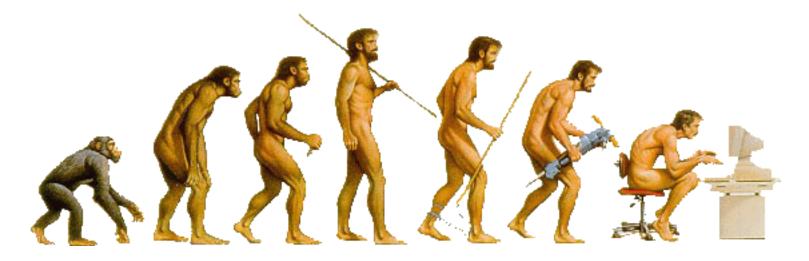
~1400 people total 10s of Ks of Hadoop nodes, multiple DCs 10s of PBs total Hadoop DW capacity ~100 TB ingest daily dozens of teams use Hadoop daily 10s of Ks of Hadoop jobs daily



Twitter's data warehousing architecture (2012)

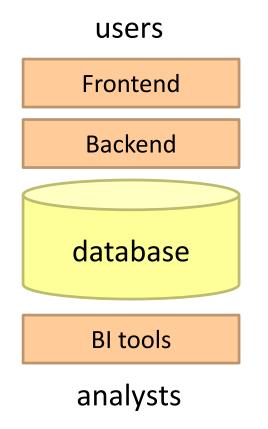
Importing Log Data

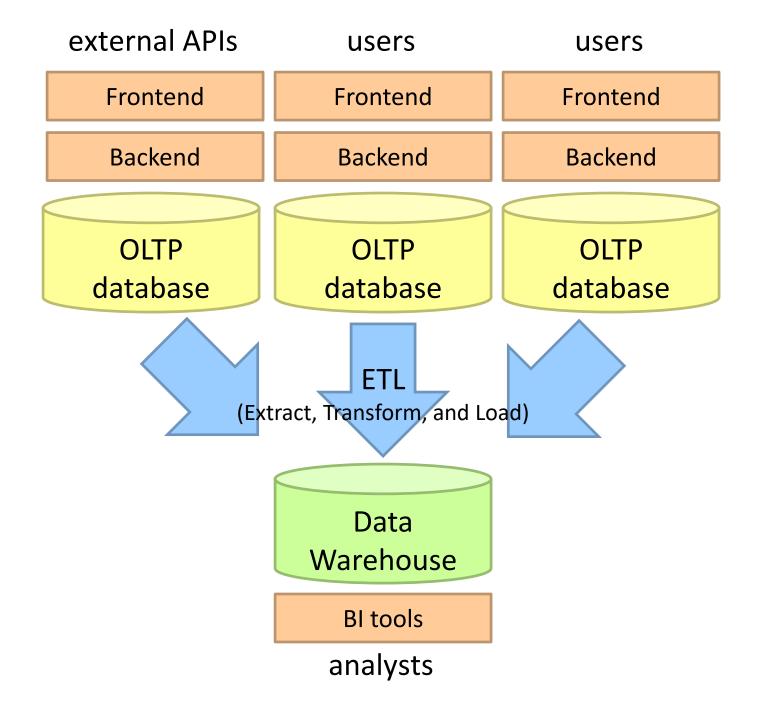


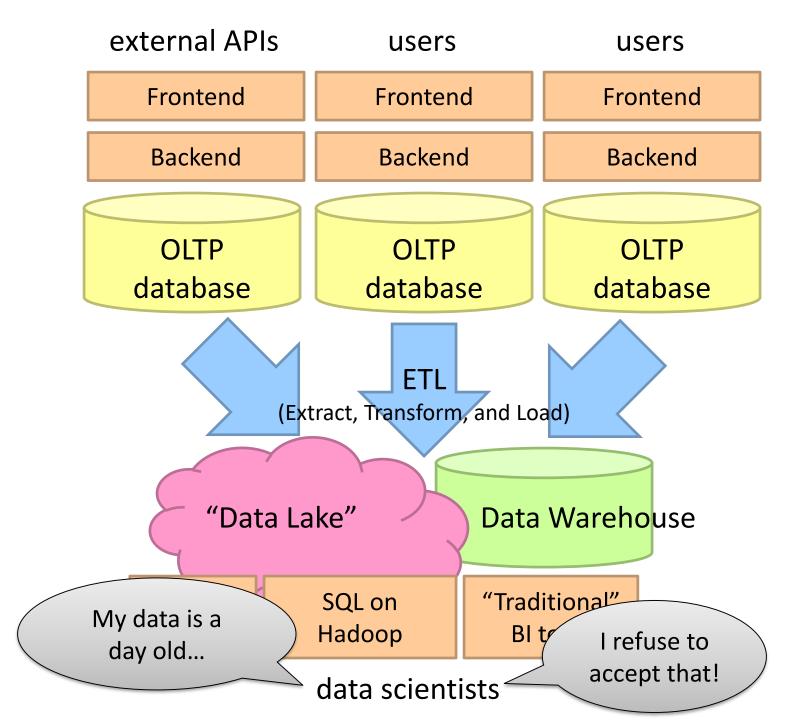


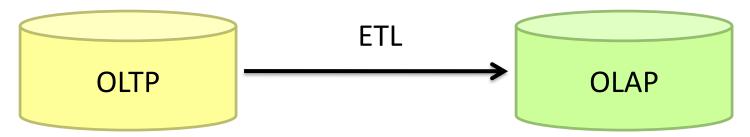
What's Next?

Two developing trends...

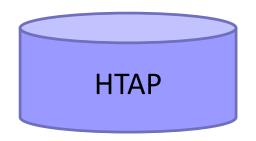






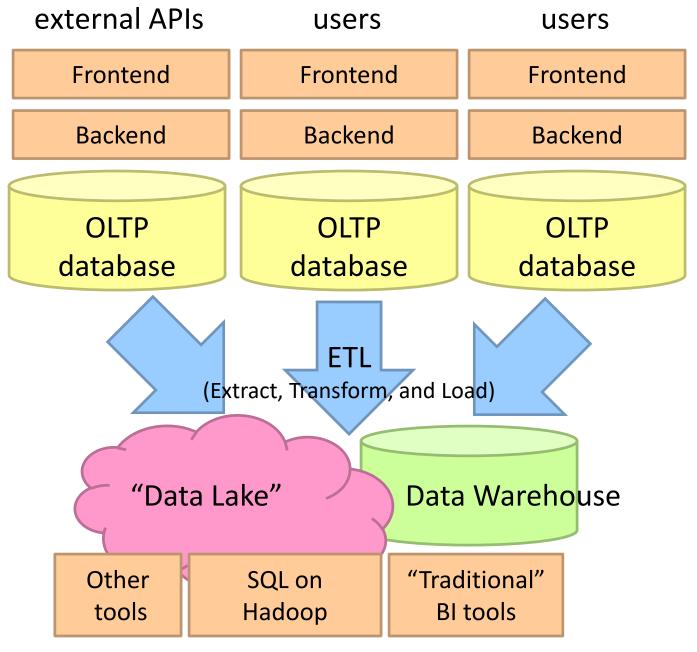


What if you didn't have to do this?



Hybrid Transactional/Analytical Processing (HTAP)

Coming back full circle?



data scientists

