Big Data Architectures

Big Data Management





Knowledge objectives

- 1. Explain the problema of a spaghetti architecture
- 2. Explain the need of the Lambda architecture
- 3. Explain the difference between the Kappa and Lambda architectures
- 4. Justify the need of a Data Lake
- 5. Identify the difficulties of a Data Lake
- 6. Explain the need of each component in the Bolster architecture
- 7. Map the components of Bolster to a RDBMS architecture





Application Objectives

1. Given a use case, define its software architecture



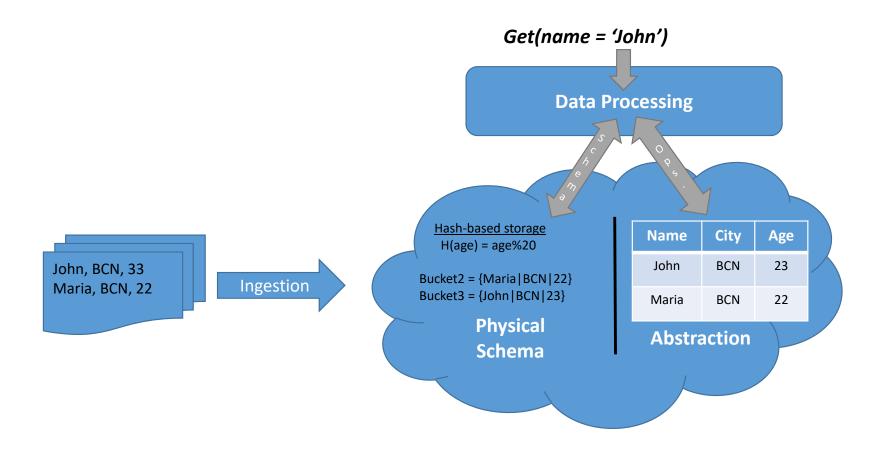


Problem definition





Data Management (I)







Data Management (II)

Data management refers to the features a DBMS must provide:

- Ingestion: means provided to insert /upload data
 - E.g., ORACLE SQL*Loader
- Storage: format/structures used to persist data
 - E.g., hash, B-tree, heap file
- Modelling: arrangement of data within the available structures
 - E.g., normalization, partitioning
- Processing: means provided to manipulate data
 - E.g., PL/SQL
- Querying/fetching: means provided to allow users to retrieve data
 - E.g., SQL, relational algebra

In Big Data settings, they are the same concepts but assuming NOSQL underneath

- Typically, a distributed system
- Possibly with an alternative data model to the Relational one
- Implementing ad-hoc architectural solutions





Big Data Architectures

- Question the main principles of traditional DB architectures
- Implement from scratch the whole stack
 - Ingestion, Storage, Modeling, Processing, and Querying
- Use new trendy technological features
 - Primary indexes to implement the global catalog
 - Distributed Tree
 - Dynamic Hashing
 - In-memory processing
 - Columnar block iteration: vertical fragmentation + fixed-size values + RLE compression
 - Heavily exploited by column-oriented databases
 - Good for read-only workloads
 - Sequential reads for large workloads
 - Take the most out of databases by boosting sequential reads
 - Enables pre-fetching
 - Option to maximize the effective read ratio (by a good DB design)
 - Key design





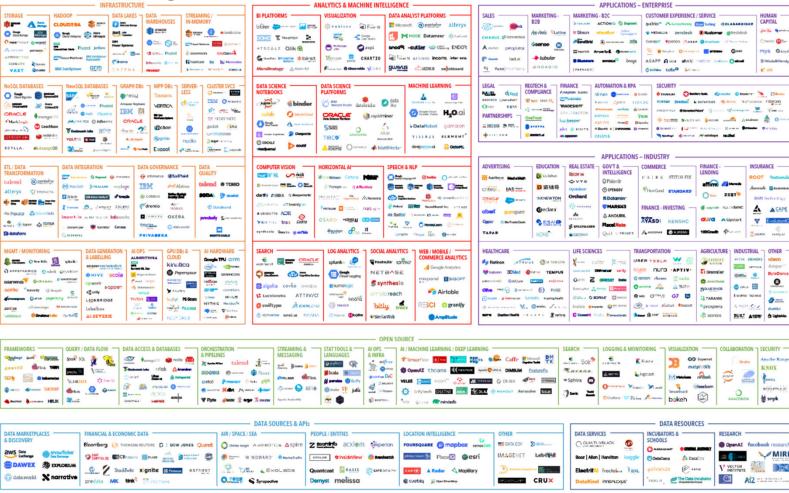
The Multi-Project Approach

- The DBMS tasks can be spread over different systems
 - Independent
 - Heterogeneous
- Hadoop as paradigmatic case:
 - Storage: HDFS + Hbase
 - Modeling: HCatalog
 - Ingestion: Sqoop
 - Processing: Spark
 - Querying: Spark SQL





Big Data Landscape

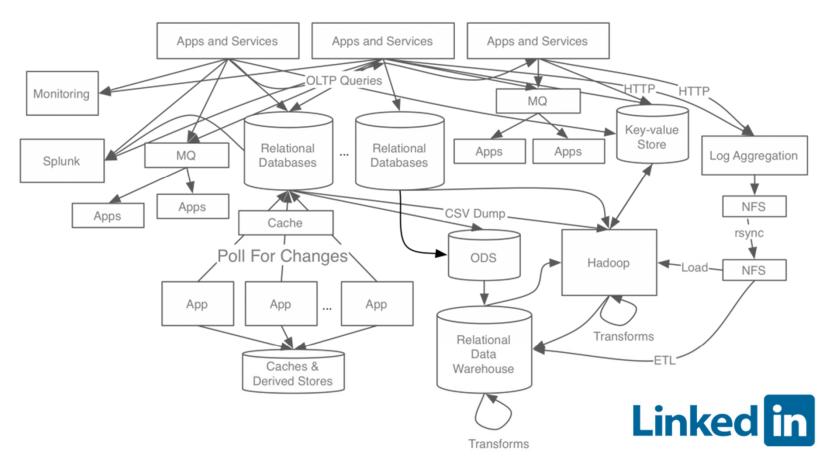


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Spaghetti architecture





https://www.confluent.io/blog/event-streaming-platform-1

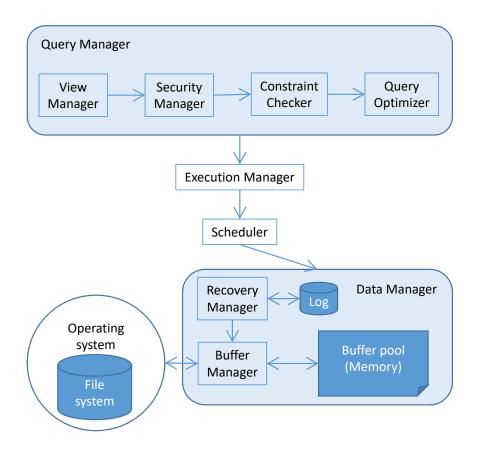


Database Management System view





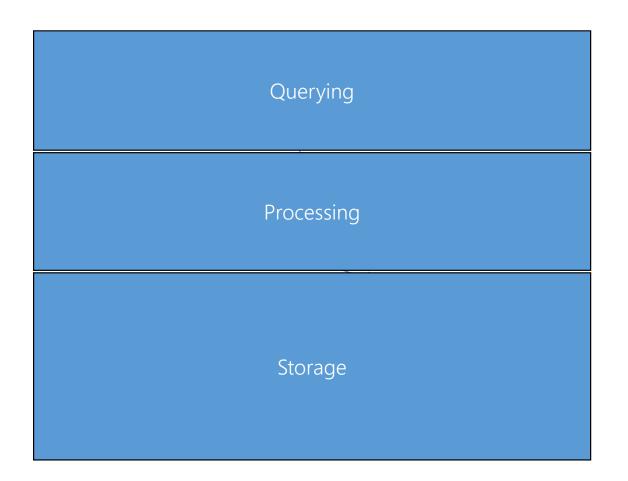
Centralized DBMS Architecture







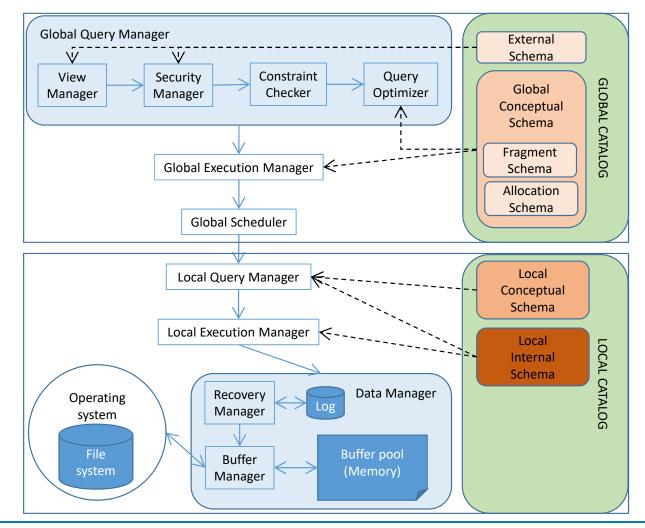
Centralized DBMS Architecture







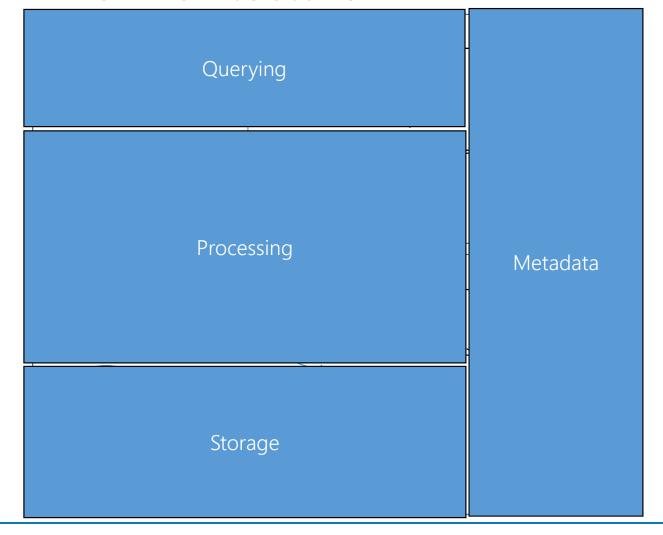
Distributed DBMS Architecture







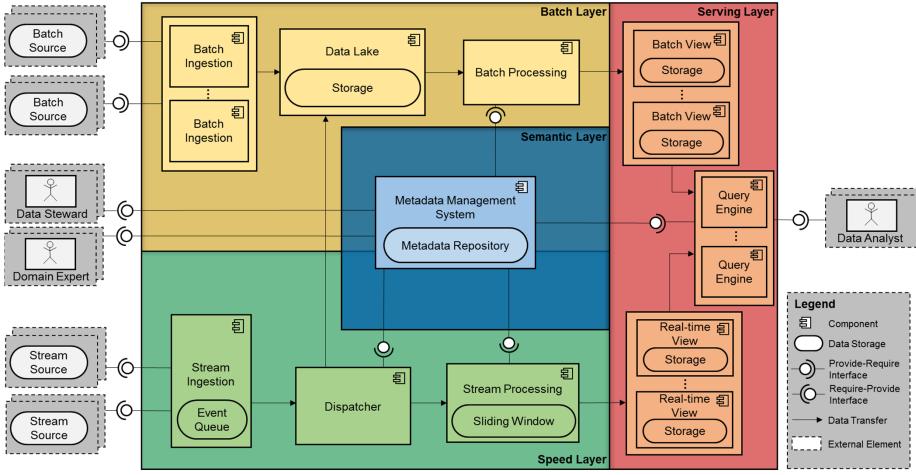
Distributed DBMS Architecture







Bolster

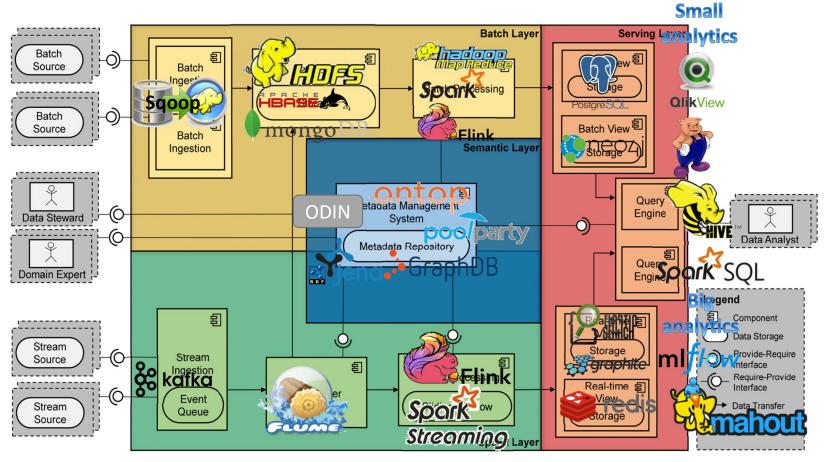




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Bolster Instantiation

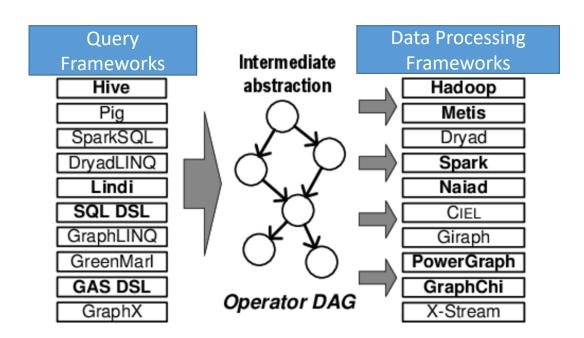






Workflow Orchestrators

- Current workflow orchestrators are rather poor: Oozie
- But there are attempts for smarter approaches: the ideas behind Musketeer deserves special attention



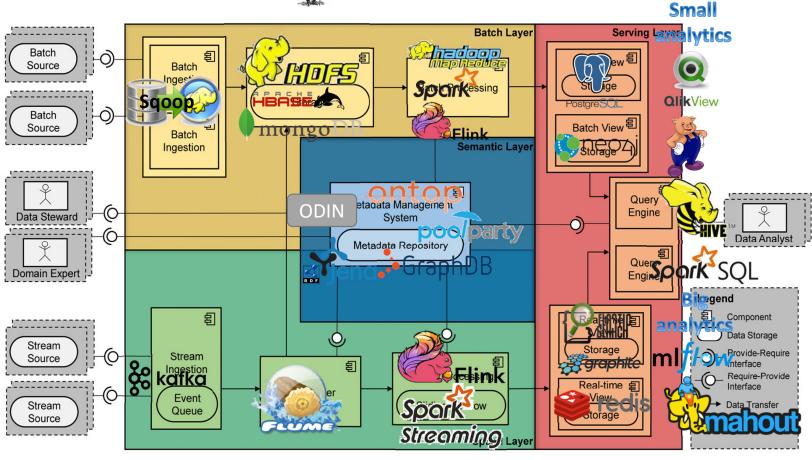
Shortly, does a similar job to global query optimizers of traditional distributed RDBMS





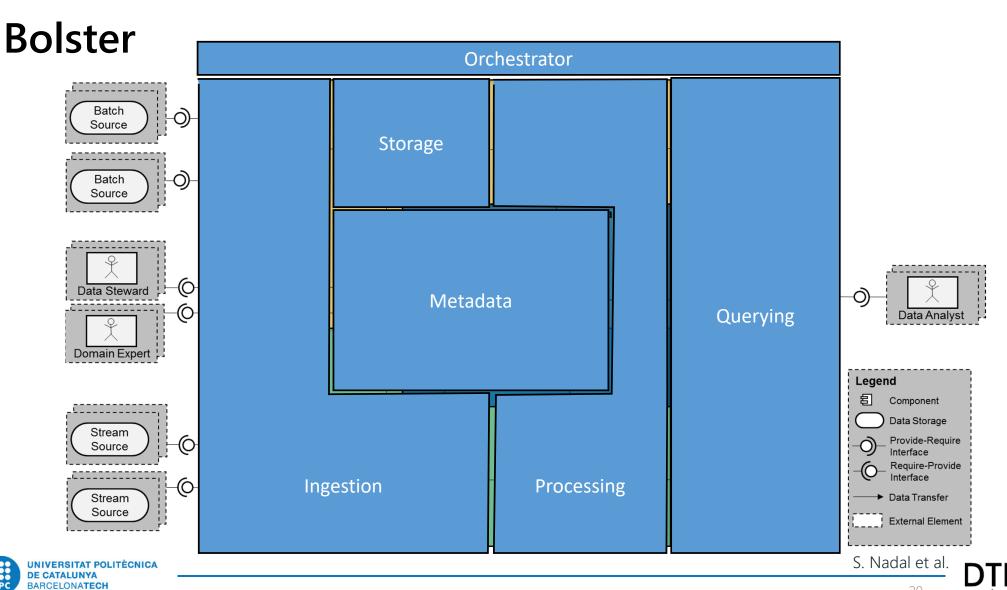
Bolster Instantiation











New Storage Architectural Pattern

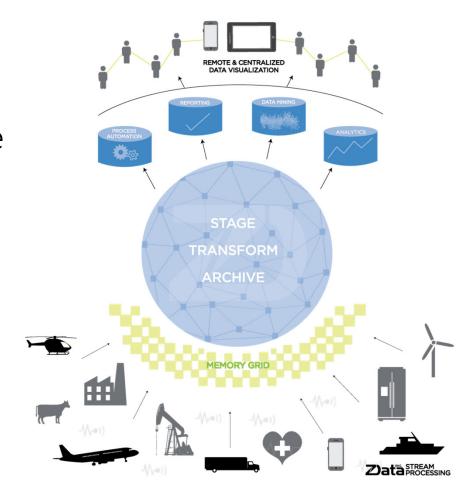
From data warehousing to data lakes





The Data Lake

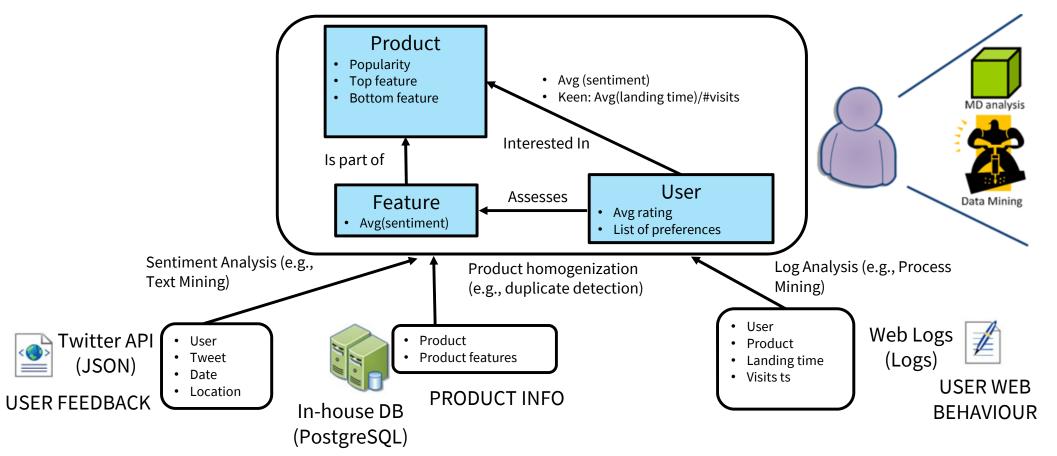
- IDEA: Load-first, Model-Later
- Modeling at load time restricts the potential analysis that can be done later (Big Analytics)
- Store raw data and create ondemand views to handle with precise analysis needs







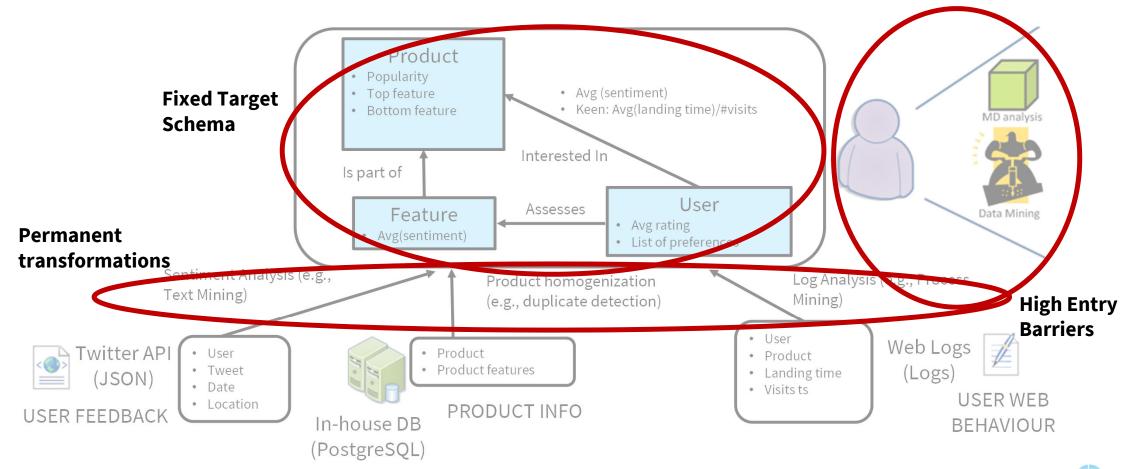
Model-First (Load-Later)





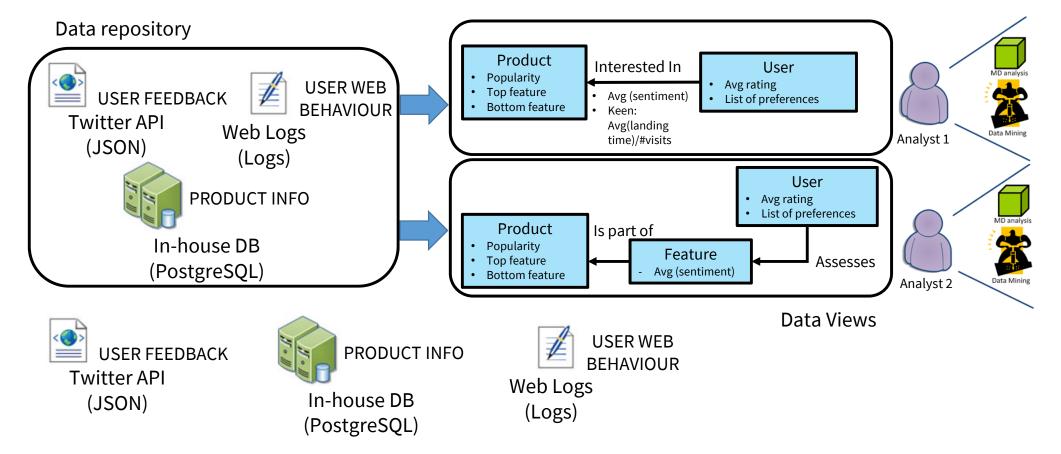


Drawbacks of Model-First (Load-Later)





Load-First (Model-Later)

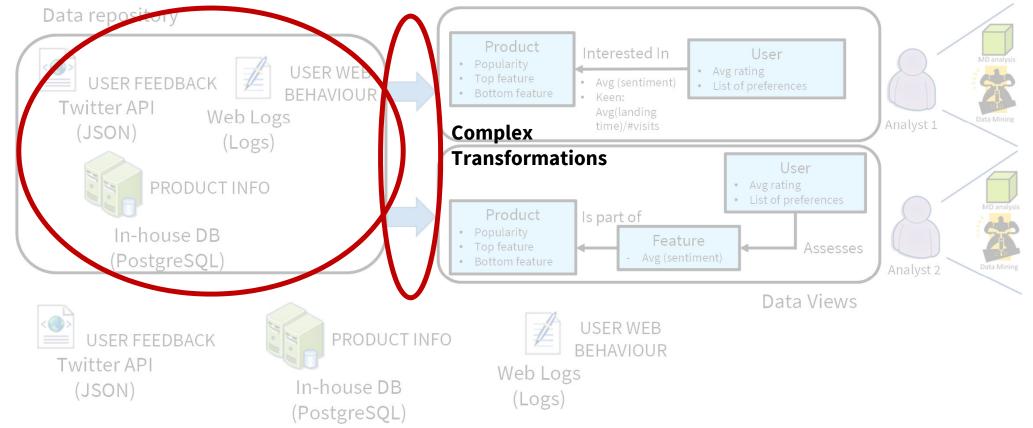






Drawbacks of Load-First (Model-Later)

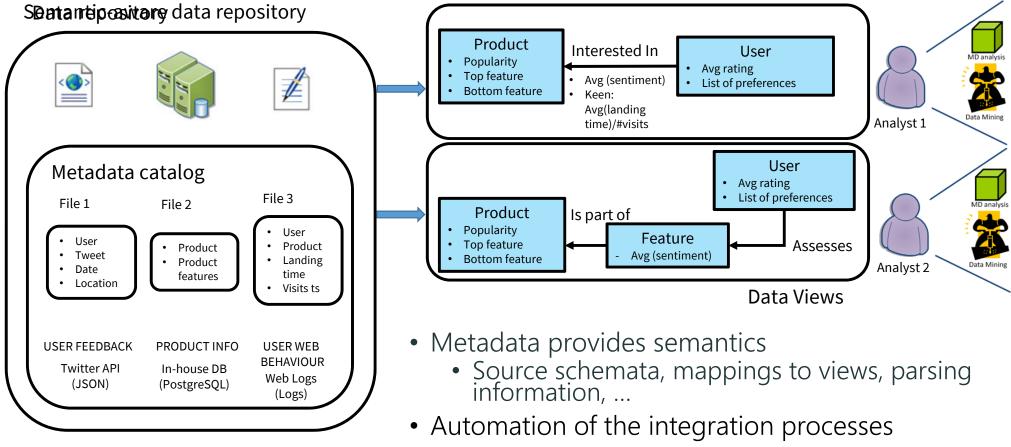








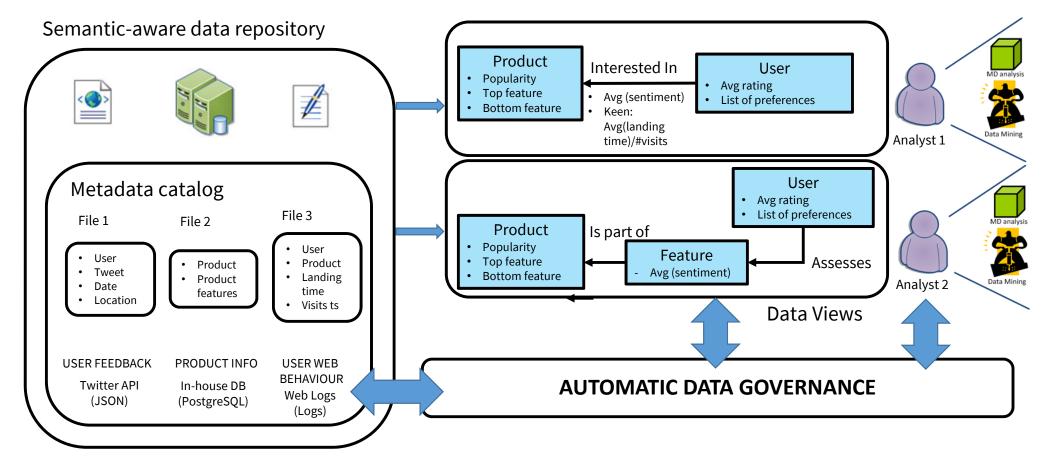
Towards semantic-awareness







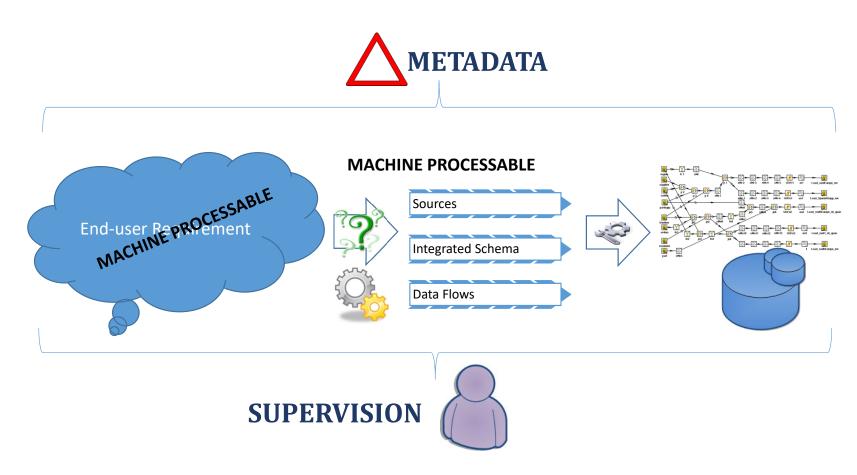
From IT-Centered to User-Centered







The Missing Link: Metadata





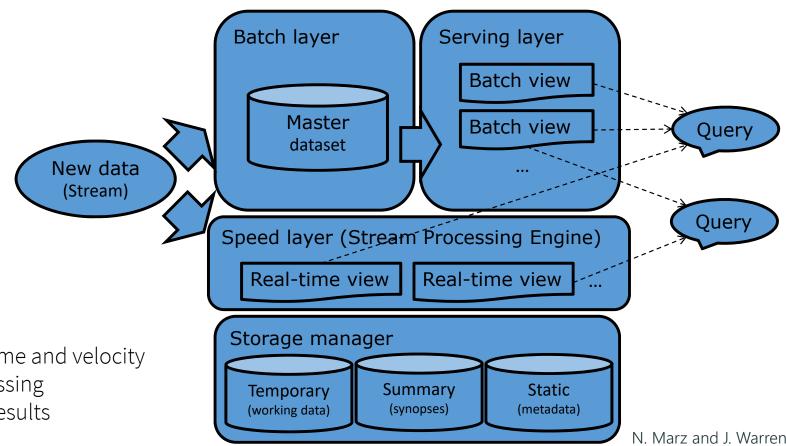


New Processing Architectural Patterns





λ-Architecture

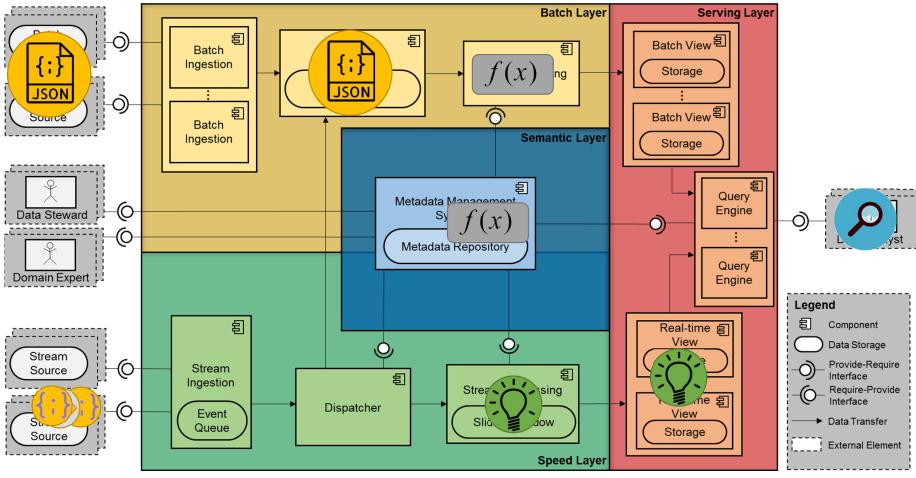


IDEA: Accommodate volume and velocity Real time Vs. Batch processing Precise Vs. Approximate results





Bolster

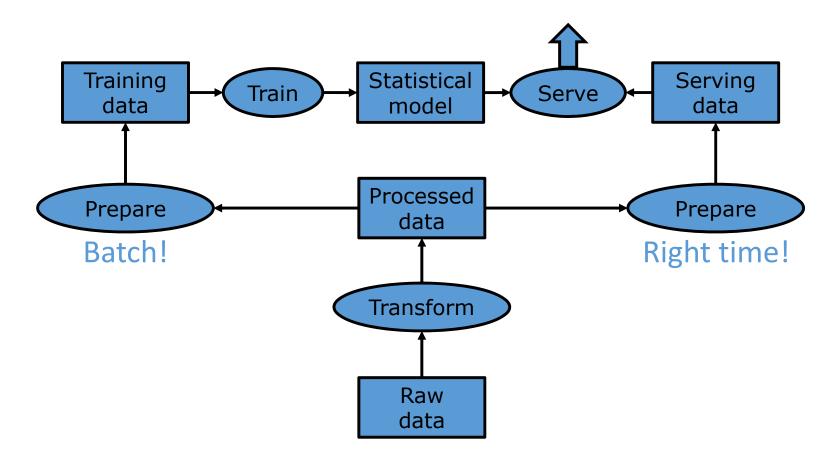




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Data-centered architecture

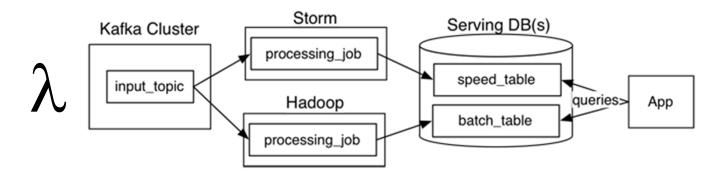


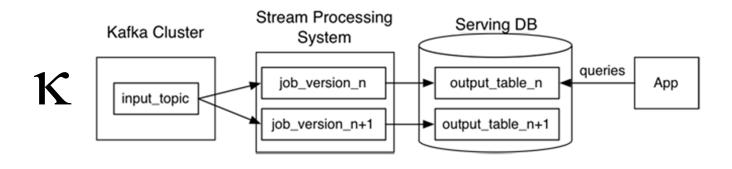




к-architecture

• Data is considered to be a never-ending stream









Closing





Summary

- New architectural solutions
 - Lambda
 - Kappa
 - Polyglot systems
- Data Lakes
 - The need of metadata
- Reference architectures
 - Bolster
 - Quarry





References

- D. McCreary and A. Kelly. *Making Sense of NoSQL*. Manning, 2014
- M. Grover et al. Hadoop Application Architectures. O'Reilly, 2015
- N. Marz and J. Warren. *Big Data: Principles and best practices of scalable realtime data systems*. Manning Publications Co., 2015
- S. Nadal et al. *A Software Reference Architecture for Semantic-Aware Big Data Systems*. Information and Software Technology 90. Elsevier, 2017
- S. Nadal et al. *ODIN: A Dataspace Management System*. International Semantic Web Conference 2019
- S. Nadal. Metadata-Driven Data Integration (PhD Thesis). 2019
- P. Jovanovic et al. *Quarry: A User-centered Big Data Integration Platform*. Information Systems Frontier, 2021





Resources

- http://hadoop.apache.org
- http://www.cloudera.com
- http://hortonworks.com



