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



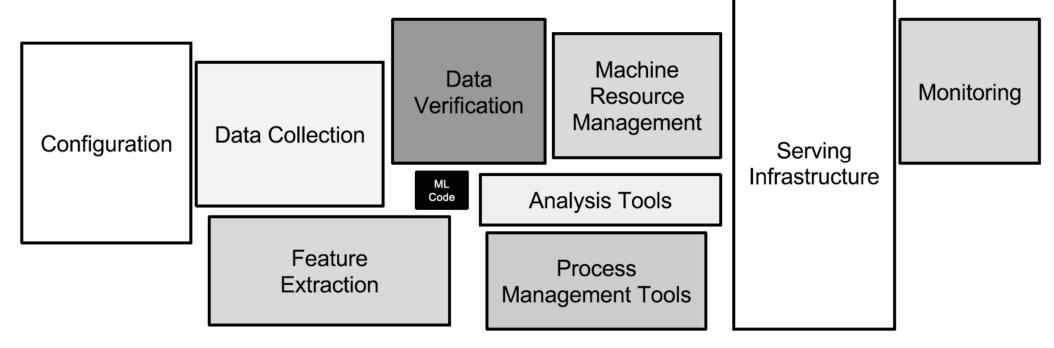


Problem definition





Vast and Complex surrounding infrastructure



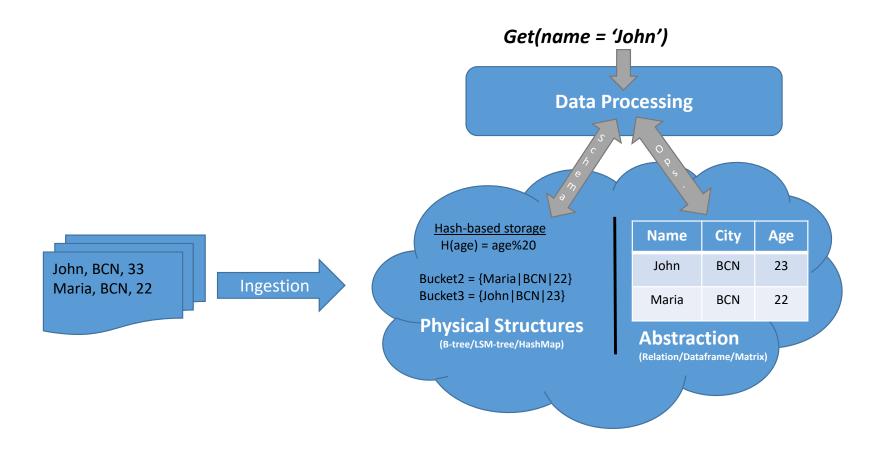
- ML code is just a small box in the middle
 - ... with a lot of plumbing around it

D. Sculley te al.





Data Management (I)







Data Management (II)

Data management refers to the functionalities 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

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





Big Data Architectures

- Question the main principles of traditional DB architectures
 - Data can grow beyond limits requiring scale out (a.k.a. Volume)
 - Data is not necessarily persisted (a.k.a. Velocity)
 - Data structure is neither known a priori, nor fixed (a.k.a. Variety and Variability)
- 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
- Implement from scratch the whole stack
 - Ingestion, Storage, Modeling, Processing, and Querying





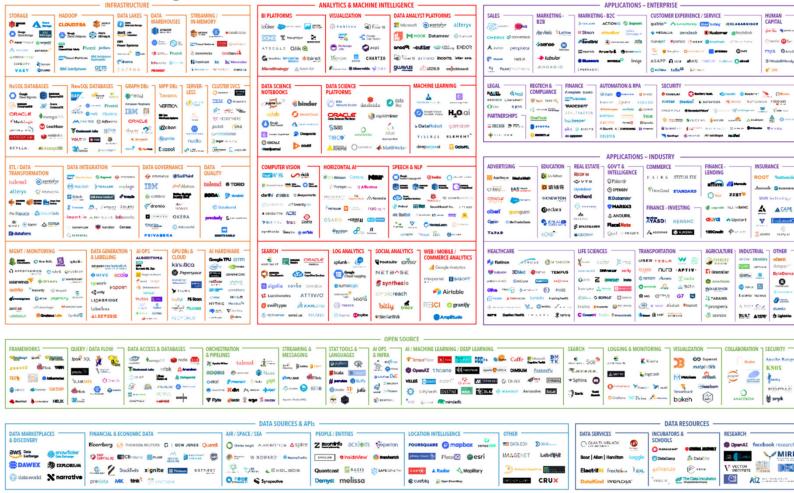
The Multi-Project Approach

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





Big Data Landscape

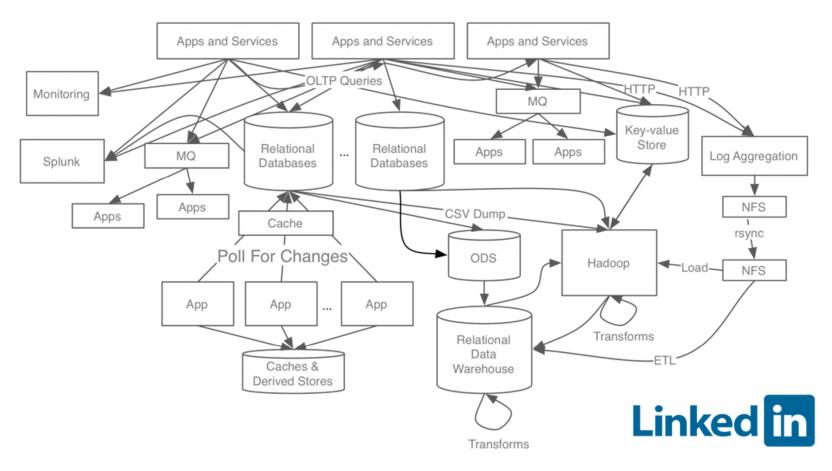


DATA & ALLANDSCAPE 2020





Spaghetti architecture





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



New Storage Architectural Pattern

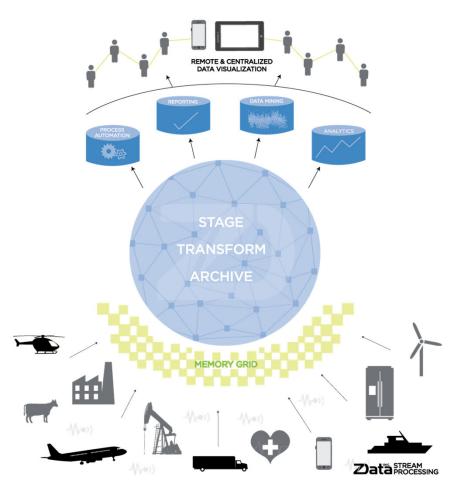
From data warehousing to data lakes





The Data Lake

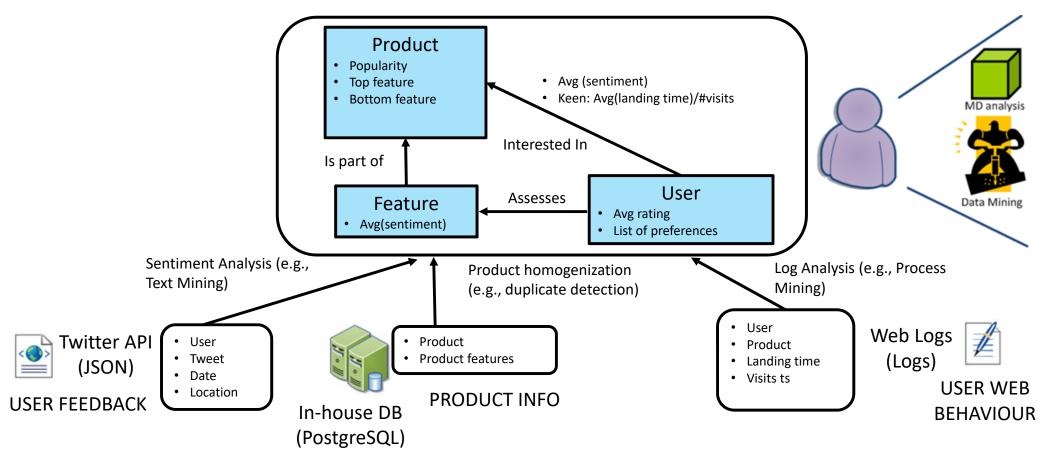
- Idea: Load-First, Model-Later
- Modelling at load time restricts the potential analysis that can be done later (Big Analytics)
- Characteristics:
 - a) Store raw data
 - b) Provide governing functionalities
 - c) Create on-demand views to handle precise analysis needs







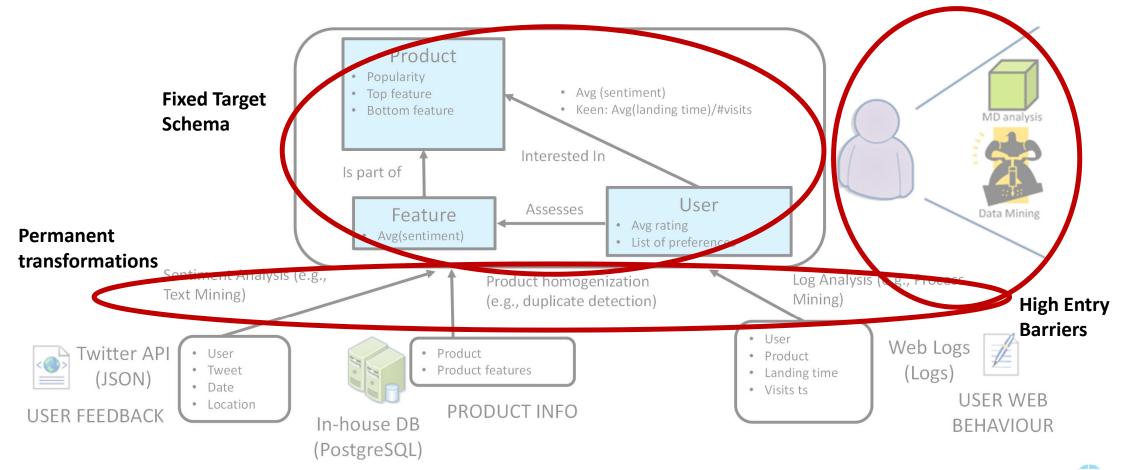
Model-First (Load-Later)







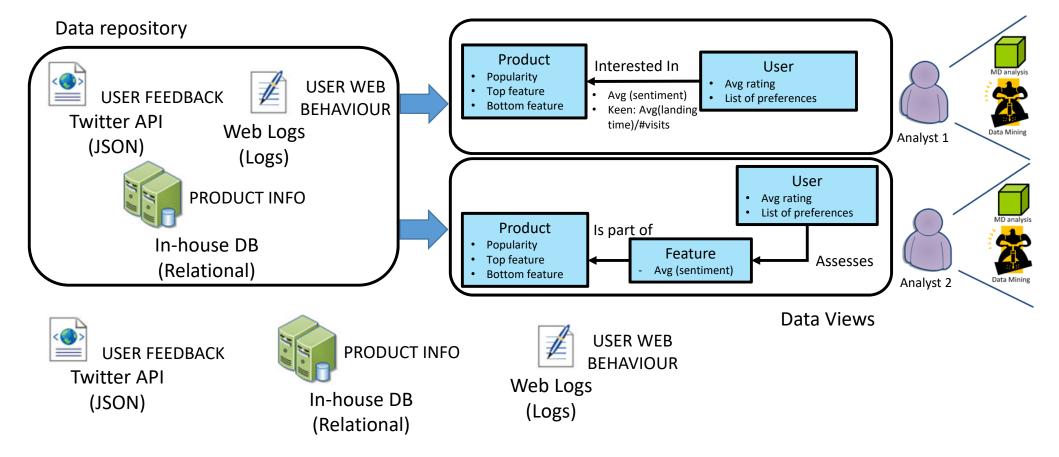
Drawbacks of Model-First (Load-Later)







Load-First (Model-Later)

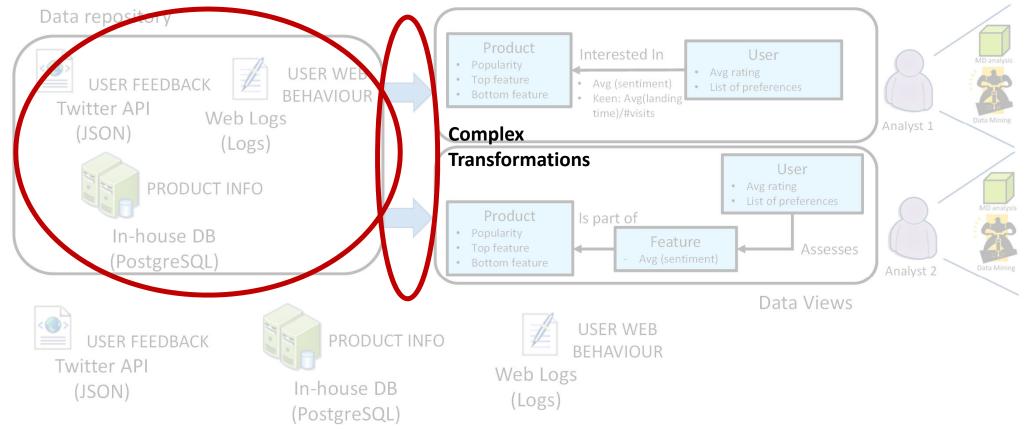






Drawbacks of Load-First (Model-Later)

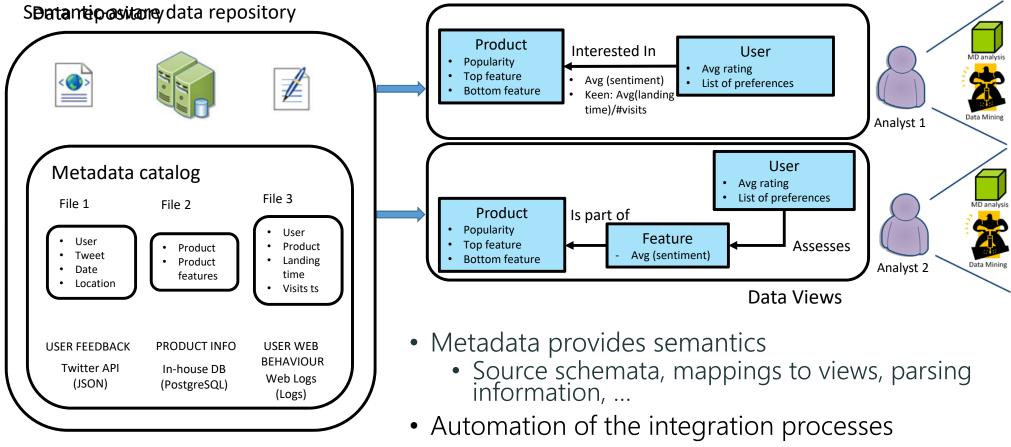








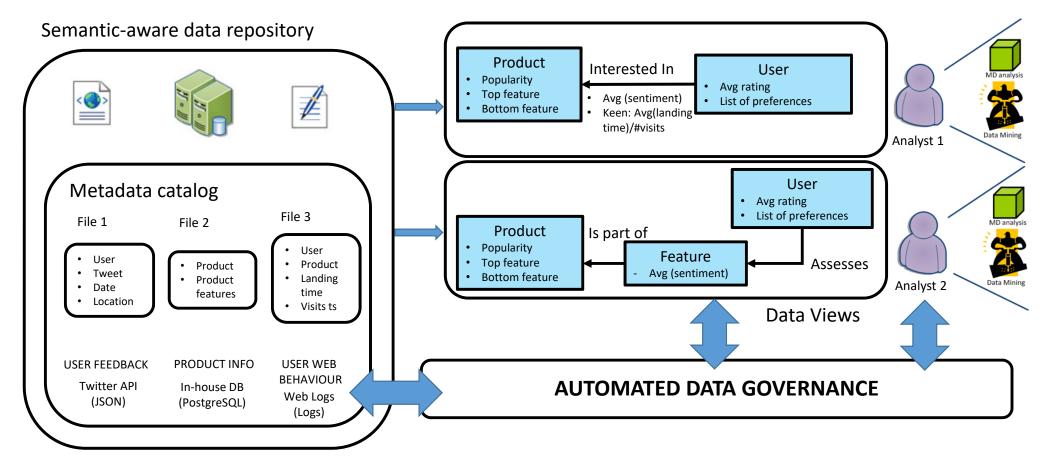
Towards semantic-awareness







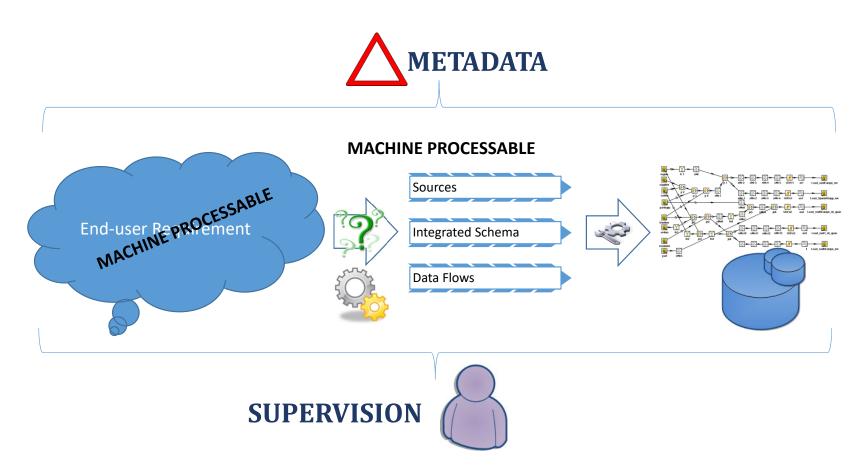
From IT-Centered to User-Centered







The Missing Link: Metadata





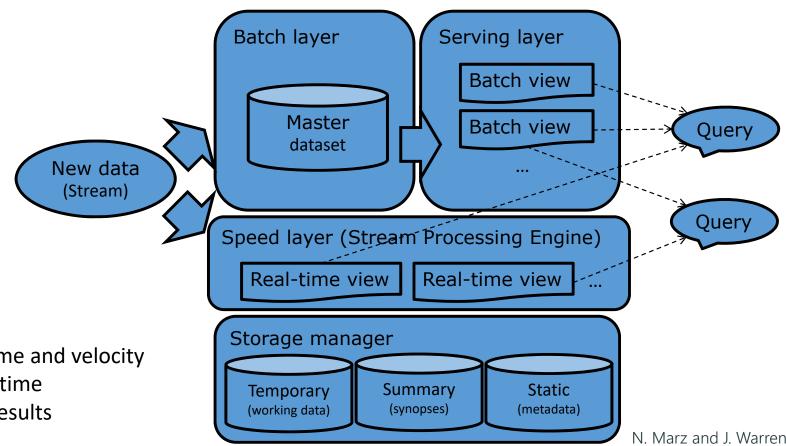


New Processing Architectural Patterns





λ-Architecture

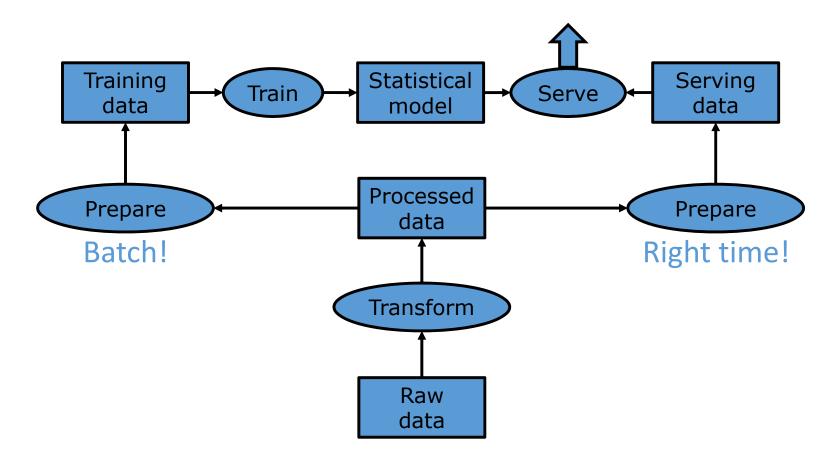


Idea: Accommodate volume and velocity Batch processing vs. Real time Precise vs. Approximate results





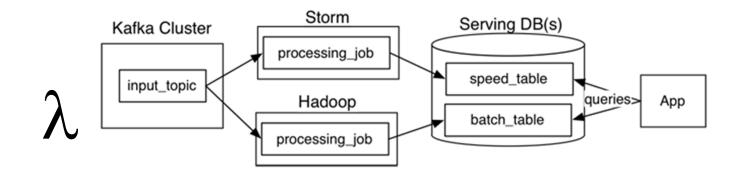
Data-centered architecture



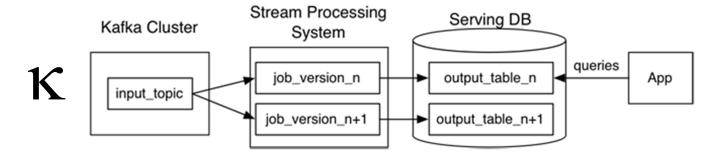




к-architecture



• Data is considered to be a never-ending stream





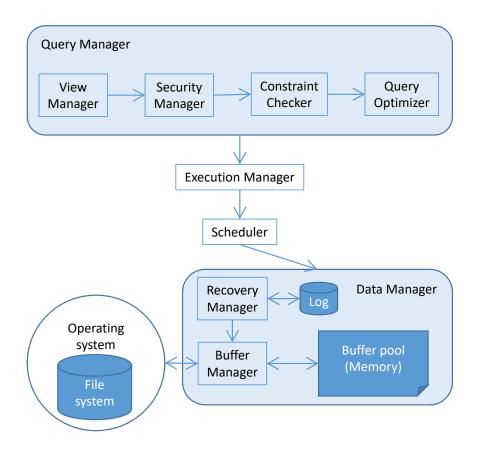


Database Management System view





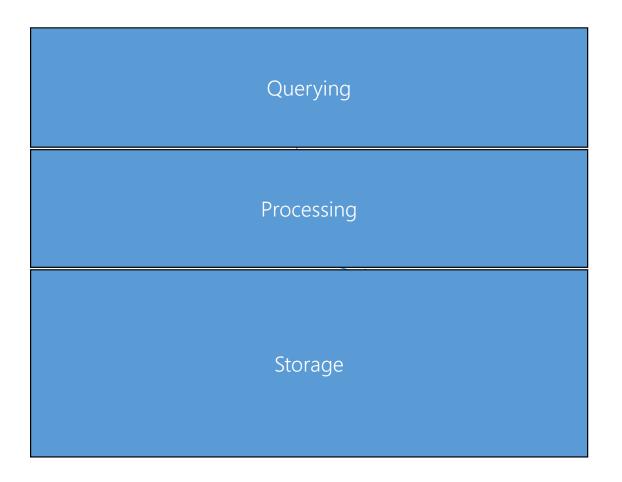
Centralized DBMS Architecture







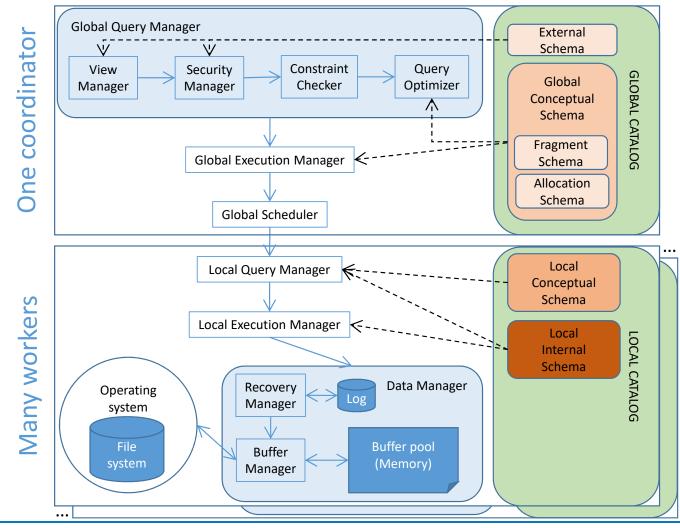
Centralized DBMS Architecture







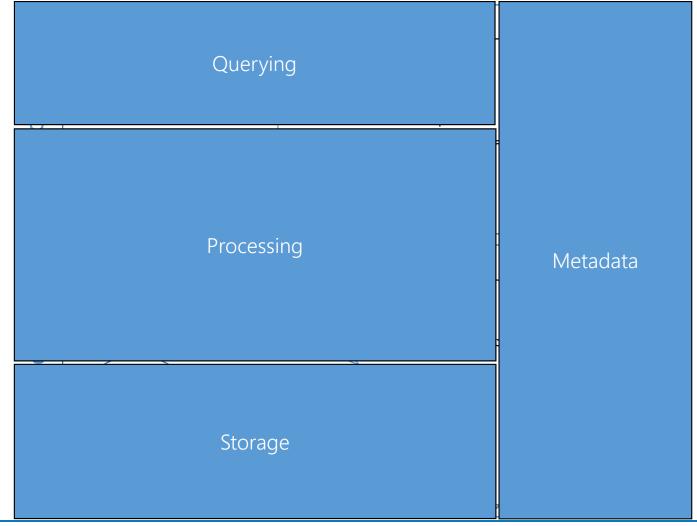
Distributed DBMS Architecture







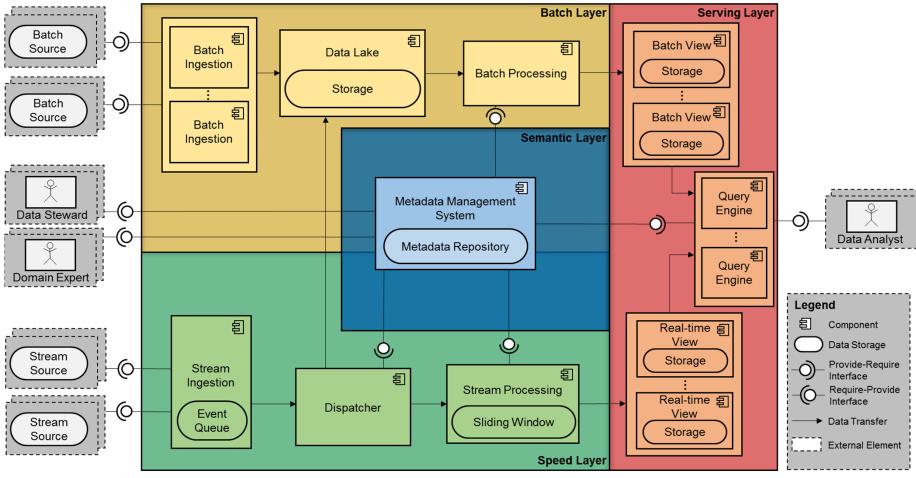
Distributed DBMS Architecture







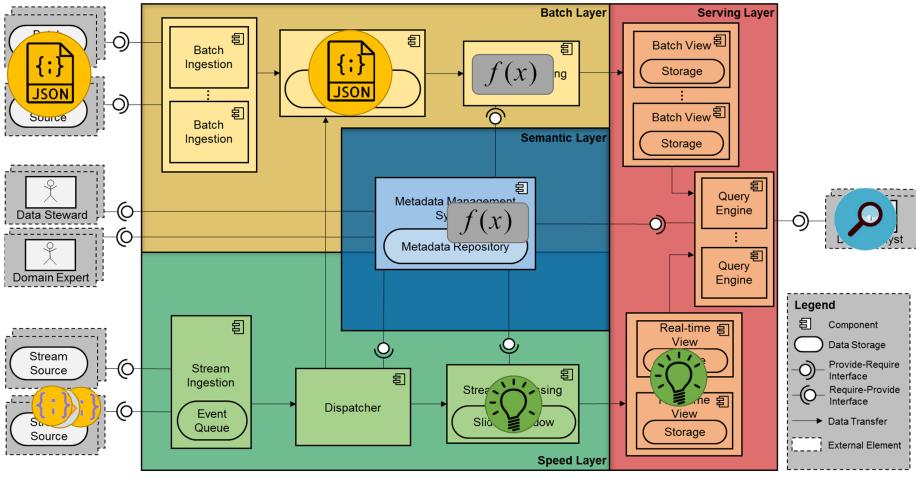
Bolster





S. Nadal et al.

Bolster

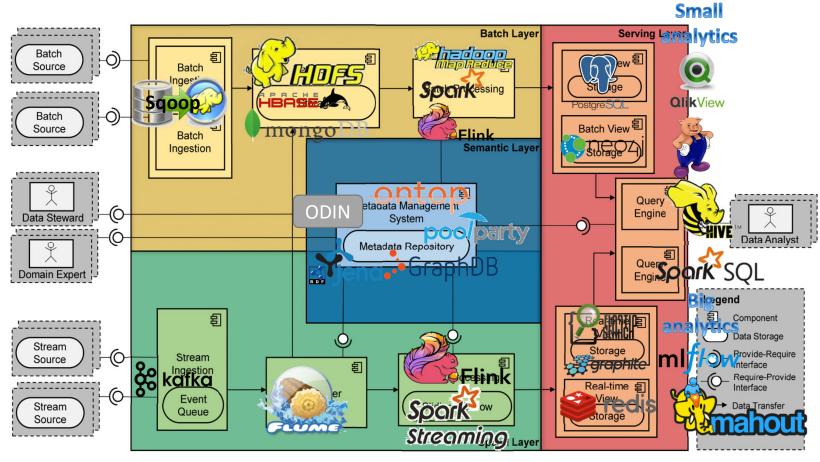




S. Nadal et al.



Bolster Instantiation

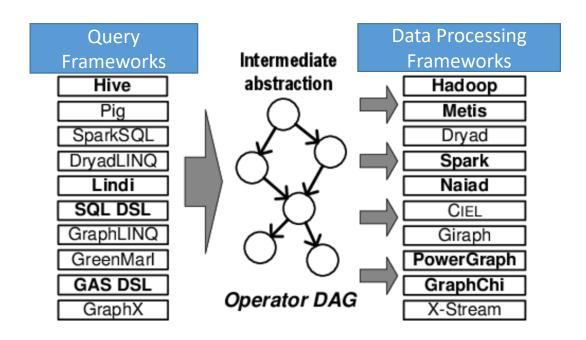






Workflow Orchestrators

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



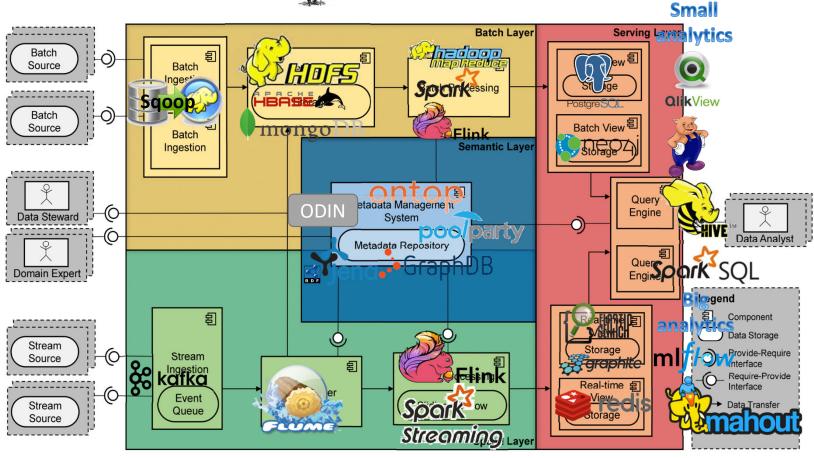
Does a similar job to global execution manager of traditional distributed RDBMS





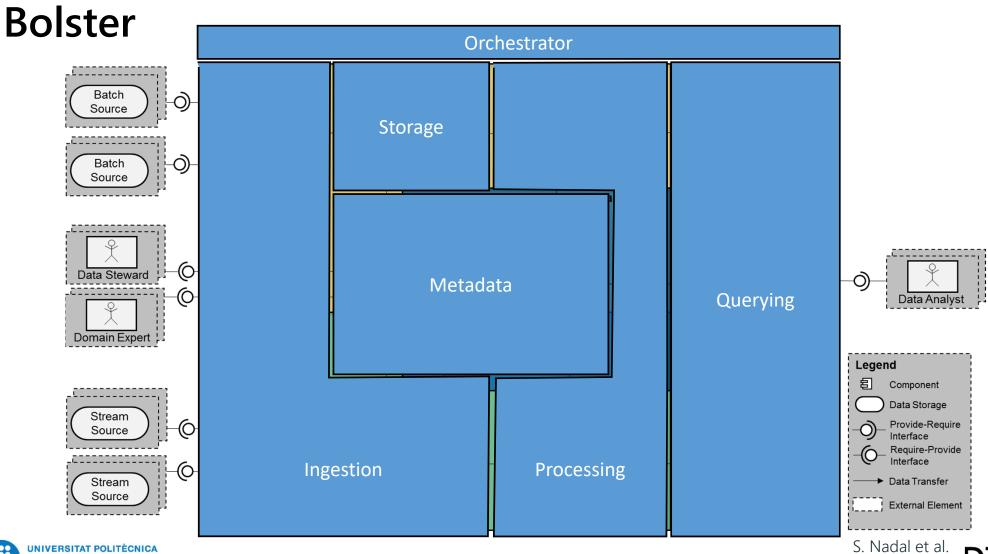
Bolster Instantiation











Closing





Summary

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





References

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Resources

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- http://www.cloudera.com
- http://hortonworks.com



