



**Pyramid**

# **How Can the Evolution of Data Management Systems Help for Big Data Applications**

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**\* Query Processing & Optimization in Parallel & Large-scale Distributed Environments**

# I. Introduction: Evolution of Data Management Systems [ Gra 96, Ham 13]

➔ “The present without past has not future” Fernand Braudel

▶ <Concept → Systems: *Objective* >

■ **File Management Systems:** *Storage Device Independence*

■ **Uni-processor Rel. DB Systems DBMS [Codd 70]:** *Data Independence*

■ **Parallel DBMS [Dew 92, Val 93]:** *High Performance & Data Availability*

■ **Distributed DBMS [Ozs 11]:** *Location/Frag./Replication Transparency*

■ **Data Integration Systems [Wie 92]:** *Uniform Access to Data Sources*

Characteristics = <Distribution, *Heterogeneity*, *Autonomy*>

■ **Data Grid Systems [Fos 04, Pac 07]:** *Sharing of Available Resources*

■ **Cloud Data Manag. Systems:** *<Pay-Per-Use> → Economic Models*

[Aba 09, Sto 10/13, Agr10/12, Cha 12, Col12, Kald 12, Zho 12, Sul 12, Gre 13, Li 14, Unt 14, Mes 16, Ham 16, ...]

Characteristics = <*Elasticity*, *Fault-Tolerant*, *Performance Isolation*>

## II. **S7** Towards Cloud Data Management Systems CDMS

[Aba 09, Sto 10/13, Agr 10-12, Chaud 12, Zhou 12, Kald 12, Gra 13, LI 14, Unt 14, Norvag 14, Akba 15, Bon 15, Akba 16, Ham 16, ...]

### Outline

- **Big Data, Cloud Computing & MapReduce: Motivations?**
- **Main Characteristics of Cloud Systems [D. Agrawal et al. 2011]**
- **“Hot Debate” on: MapReduce Versus Parallel DBMS: friends or foes?**  
[M. Stonebraker et al., 2010], [D. Agrawal et al. 2010, S. Chaudhuri 2012 ]
- **“ Reconciling Debate”** [Zhou et al. 2012, Kaldewey et al. 2012/EDBT]  
“SCOPE : Parallel Databases **Meet** MapReduce” [Zhou et al. 2012, VLDB Jo.]
- **Classification of CDMS**
- **Comparison between // DBMS and MR**
- **Conclusion**

## II.1 Big Data, Cloud Computing & MapReduce: **Motivations(1/3)**

- “Big (Very Large?) Data” : Generated from
  - Computer Simulations, Astronomy, Earth Observation, Biology....
  - Specific Requirements of **Web Applications** : Log Processing, Analysis of Streaming Sensor Data, Social Network, Document Indexing,.....

Remarks: **43<sup>rd</sup>** Intl. Conf. **on Very Large Databases**; **36<sup>th</sup>** Intl . Conf. On **Data Management**  
// **DBMS**: <**Teradata**, Since 1984; DB Size: **11 Terabytes** , Since 1996>

➔ Big Data ➔ **“Moving Target”**

- Big Data Characteristics [Val 14, Sto 13]: **The 4 V's**
  - **Volume**: Refers to very large amounts of Data
  - **Variety**: Heterogeneity of Data Formats and Semantics
  - **Velocity**: Data Streaming (Producer-Consumer Dataflow)
  - **Veracity/Value**: Meaningful of the results? (Data Mining)
  - **Other V's**:
    - **Validity**: Correction and accuracy of data?
    - **Volatility**: Necessary period to store this data?

## II.1 Big Data, Cloud Computing & MapReduce: **Motivations(2/3)**

- **Big Data Characteristics: a Solution for “the 3 V’s” [Val 14] ?**
  - **Volume:** Refers to very large amounts of Data
    - ➔ **Parallel Database Systems [Dew 92]**
  - **Variety:** Heterogeneity of Data Formats and Semantics
    - ➔ **Data Integration Systems [Wied 92]**
  - **Velocity:** Streaming Data
    - ➔ **Data Stream Management Systems [Ozu 11, Chapter 18]**

**However, why these systems are not naturally used?**

## II.1 Big Data, Cloud Computing & MapReduce: **Motivations(3/3)**

- **Current Solutions (Infrastructures & Software) are:**
  - Proprietary & Expansive** (Oracle, DB2/IBM, SQL Server/MS, ....)
  - ➔ **Open Source Alternatives, Simple Programming Model ! (e.g. MapReduce), Low Costs (Commodity Hardware)**
- **Ability to scale resources (up, down, out) dynamically on-demand :** ➔ **Elasticity (≠ Scalability) (➔ Pay-Per-Use PPU)**
- **How the systems should react to Failures?**
  - ➔ **Fault-Tolerance (Commodity Hardware , HDFS (Hadoop Env.), Data Replication)**
- **Cloud Environments do not to be owned nor managed (PPU Approach) by a Customer:**
  - From "Investment" to "Renting" ➔ Users ➔ Multi-tenant (➔ Performance Isolation)**

## II.2 Main Characteristics of Cloud Systems [Agra. et al. 2011]

- **Scalability (Infrastructure: Shared-nothing Architecture)**
- **Elasticity [Ozu 11]**  
«The ability to scale resources out, up, and down dynamically to accommodate changing conditions»
- **Performance Isolation [Nara 13]: Users → Multi-tenant & SLA (Service Level Agreement) Meeting**
- **Strong Fault-Tolerance: (CH, HDFS (Hado. Env.), Data Replication)**
- **Ability to run on Commodity Hardware CH (Low Cost!)**
- ➔ **New Context = < // & Dist., Large-scale, Stable, Service on-demand, Multi-tenant, Commodity Hardware >**
- ➔ **Introduction of Economic Models in the Resource Management**

## II.3 Hot Debate (“Storm”): MapReduce versus // DBMS

### ■ “MapReduce and Parallel DBMSs: Friends or Foes?”

[Stonebraker et al. 2010 Com. of the ACM, Jan. 2010, Vol 3. No. 1]

◆ The performance results (between Hadoop (MR) and 2 // DBMSs ) show that the **DBMSs are substantially faster** than the **MR systems** once the **data is loaded**.

➔ Conclusion: “**MR complements DBMSs** since DB are not designed for ETL (Extraction-Transform –Load) tasks, a MR specialty “

### ■ “Big Data and Cloud Computing: New Wine or Just New Bottles?”

[Agrawal 2010 et al. , Univ of California/Santa Barbara] VLDB’2010 Tutorial

### ■ “An Interview with S. Chaudhuri”, [Sept. 2012, XRD, Vol.19, No. 1]

“If I were to look at recent research publications, a disproportionately large fraction of them **are focused on solving for MapReduce platforms the same problems we addressed for parallel database systems. We can and should do much more.**”



# Petasky – Mastodons Project (CNRS, LIMOS/LIRIS, FR)

“Benchmarking SQL on MapReduce systems using large astronomy databases”; A. Mesmoudi et al.; In: Intl journal PDBD, 34(3), 2016

- **Objectives:** “They report on the capability of 2 MR systems (Hive and HadoopDB) to accommodate LSST data management requirements” in terms of loading & execution times : < Data Loading & Indexing and Queries (Selection, Group By, Join) >
- **Conclusions [Mes 2016] :**
  - ➔ “We believe that the **model is efficient** for queries that need **one pass** on the data (e.g. Selection and Group By)”
  - ➔ “ We believe that MR model **is not suitable** for handling **Join queries** ”

## II.4 Reconciling Debate (1/2) [Zhou 2012, VLDB Journal, Kald 2012]

“SCOPE: Parallel Databases Meet MapReduce” ; **MicroSoft**

### ■ Objective : **combines benefits** from execution engines

- **Parallel DB Systems**

&

→ for Large-scale Data Analysis

- **MapReduce**

→ <Easy Programmability, Massive Scalability, HP >

### ■ Advantages of // DB Systems [Dew 92]

- Relational Schema (→ Annotations)
- Declarative Query Language (→ Automatic Optimization)
- Sophisticated Query Optimizers {Partitioned, Independent, Pipelined //}
- +/- Communication Costs : Avoid the **Data Redistribution** (in some cases)

### ■ Weakness of // DB Systems (in Massive Large Scale):

- Run Only on Expensive Servers
- Fault - Tolerance (in the case of massive // DB)
- Web Data Sets are not structured
- Communication Costs : Data Redistribution (= **Reshuffling in MR**)

## II.4 Reconciling Debate (2/2) [Zhou 2012, Kalde 2012]

“SCOPE\*: Parallel Databases **Meet** MapReduce”; **MicroSoft**

### ■ Advantages of MR

- **Scaling very well (to manage massive data sets)**
- **Strong Fault -Tolerance (HDFS, Data Replication, Commodity Hardawre)**
- **Mechanism to achieve Load-Balancing**
- **Support the Intra-operation & Independent Parallelisms**

### ■ Weakness of MR: Side Applications

**Developers:**

- **Are forced to translate their business logic to MR model**
- **Have to provide implementation for the M & R functions**
- **Have to give the best scheduling of M & R operations**
- ➔ **More Hot Problems!**
- **+ Data Dependence (Data Independence in DB !)**
- **+ Extensive Materialization (I/O)**
- **+ Data Reshuffling (Repartitioning) ➔ Plague of Parallelism**

\*: **SCOPE Proposals (Structured Computations Optimized for Parallel Execution)**

## II.5 Classification of Cloud Data Manag. Systems (1/3)

- **Early Generation of Big Data Manag. Systems BDMS:**
  - **NoSQL Databases/MapReduce Systems based on Type of Data Store**
- **Next Generation of BDMS (Evolution of NoSQL Systems):**
  - **New SQL = Scalable Power of NoSQL Systems + ACID Properties (of Rel. DBMS!)**
- **Latest Generation of BDMS: Data Integration Approach based on Mediator –Wrapper Architecture [Wied 92]**
  - ➔ **Insure a Uniform Access to Heterogeneous, Autonomous, and Distributed Data Sources**
    - **Multistore Systems :**  
**Polybase [Dew 13], SCOPE [Zho 12] , CoherentPaas Proj. [Bon 15]**

## II.5 Classification of Cloud Data Manag. Systems (2/3)

### ■ Classification of NoSQL Systems: Type of Data Store (Approx. 130 Systems!)

- **Key-value Store: <Azure Table Storage, DynamoDB, Redis, Riak, Voldemort, ...>**
- **Document Store (XML, JSON): <MongoDB, CouchDB, RavenDB>**
- **Column-family (similar to rel. DB, Data is structured in columns): <Hbase, Cassandra, Hypertable>**
- **Graph Databases (Social Networks): <Neo4j, Infinity Graph, InfoGrid, ...>**

## II.6 Comparison between // Rel. DBMS & MapReduce

Systems Parameters	DB & // Rel. DBMS	MapReduce (Hadoop Env.)/ <u>Cloud. Systems</u>
Applications	OLAP & OLTP (ACID)	OLAP: Yes; <b>OLTP: Not suitable (Initially!)</b> → New SQL
Data Models	Data Structured (Data Schema)	Unstructured or semi-Structured , ...(more Flexible!)
Data Independence	Yes	No (Initially)
Query Languages	Declaratives	Procedurals (initially)
Optimization & Parallelization	Automatic Optim. & // Annotations: Easy	Explicit Optim. (initially) Annotations: Very difficult
Scalability & Elasticity	Scalable & <b>Dynamic</b>	Scalable & <b>Elastic</b>
Fault-Tolerance	Weak	Strong
Location	Known in advance	SLA Negotiation
----- Maturity	----- Strong	----- Weak (at this moment!)

# III. Conclusion (1/2)

## Maturity Degree of Cloud/Big Data Management Systems ?

- **Query Languages**

- **Declarative Languages**
- **Standardization**

- **More Experimentation & Benchmarking**

- **TPC – H & TPC - DS**

- **Administration & Tuning/Supervision Tools**

- **Let time do its work!**

# III. Conclusion (2/2)

## ■ Scientific Aspects

- **New “Concept” introduced by the Cloud Computing CC?**  
New dimensions of CC = **<Elasticity , Perf. Isolation, ?...>**
  - ▶ **Introduc. of Economic Models** (Rationalization & Cost effectiveness)
    - ➔ **Objective Function: Find the best trade-off between**
      - **Multi-tenant Satisfaction (QoS (e.g. Response Time))**
      - **Cost-effectiveness of Provider Services (IaaS & SaaS)**
- **Risk of a Gradual Shift of Fundamental Research Activities towards only Engineering Activities :**
  - ➔ **Best trade-off between: < Fund. Research & R&D>**

**<Concepts, Approaches/Methods, Techniques> & <Applications>**



*Thank you for your attention*



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