CloudDB: A Data Store for all Sizes in the Cloud

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Historical perspective and motivation

(Preliminary) Technical Approach

Current Status

Food for Thought



Why Data Management Research?



- Many Data Management **Technologies and Products** have been around
- Data Centers have evolved over the time
- Data Center hosting became a business
- Database Community was successful in creating technologies and business



Why Data Management (Again)?







Cloud Computing

- A paradigm shift in how and where a workload is generated and it gets executed
 - Cloud service provider Cloud service consumer



Market Size

- Data Management Market ~\$20B
- IT Cloud Service ~\$42B (by 2012) (IDC)



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Animoto on Amazon EC2



- Rapid growth in three days, the number of users increased from 25k to 250k
- Number of servers from 50 to 3500
- Assume \$500 per machine, \$1.75M!
- Instead, they used Amazon EC2

Problem: It is not trivial to distribute users' accesses to the data by just scaling out cloud computing nodes



Database-as-a-Service?



Data Management in Cloud

Cloud computing model may provide a platform to address new challenges

But the problem is:

Data Management Systems <u>were not</u> designed and implemented with cloud computing model in mind

So the question is:

What are the data management challenges we need to address before the full potential of cloud computing can be realized?



Need for New Solutions

Massive scalability to handle

- Very large amount of data
- Very large number of diverse users/requests
- Elasticity to
 - handle varying demand
 - optimize operating costs
- Flexibility to handle different data and processing models
- Massively multi-tenanted to achieve economies of scale
- More intelligent system monitoring and management



Cloud Data Management Challenges





Buy All Sizes?



"One Size Fits All": An Idea Whose Time Has Come and Gone

Michael Stonebraker

Uğur Çetintemel Empowered by Innovation

NE



Buy One Size?





Let Someone Else Do All That





Let Someone Else Do All That



Wish Lists

Clients

- Standard language API (e.g., SQL)

- Identifiable and verifiable Service Level Agreements

- Common DBMS maintenance tasks, (e.g. backup, versioning, patching etc.)

- Availability of value-add services, such as business analytics, information sharing, collaboration etc.

Service Provider

- Satisfying clients' SLAs to sustain revenue

- Great cost efficiency via high level of automation and resource sharing to ensure profitability

- Maintaining an extendable platform for value-add services

(Some) Storage Models

Store Type	Main Purpose	Pro	Con
Relational	- Transaction processing	 Standardization Higher performance on Online Transaction Processing (OLTP) ACID properties 	- Scalability
Key/Value	- Scalable data storage - Read/Write intensive workload	-Scalability	 Standardization Performance issues Complex query capability ACID properties(?)
Column-Oriented	 Analytics processing Read optimized, throughput oriented 	-Higher performance on Online Analytical Processing (OLAP) - More flexible schema evolution (?)	 Standardization Complex query capability



Application Scenario





Data Model Decisions

- Problem: Users are forced to make a decision on the data model based on the current needs of the applications
 - Is it possible to make the "right" decision all the time?
- Problem: The developer (client) has to re-architect their application in order to take advantage of different data models
 - How easy is it to change the architecture and the implementation?





Remember Data Independence?

DESCRIPTION OF A SET-THEORETIC DATA STRUCTURE

David L. Childs

FEASIBILITY OF A SET-THEORETIC DATA STRUCTURE

A General Structure Based on a Reconstituted Definition of Relation

David L. Childs

1968

A Relational Model of Data for Large Shared Data Banks

E. F. CODD IBM Research Laboratory, San Jose, California

1970



Data Independence

FILE DEFINITION AND LOGICAL DATA INDEPENDENCE C J Date & P Hopewell IBM UK LABORATORIES LTD., HURSLEY PARK, ENGLAND MAY-1971.

- Decouple application logic from data processing
- Let them be optimized and managed independently
- Enabled decades of innovation and improvement in databases





Data Independence

- The application should not have to be aware of the physical organization of the data (and how it can be accessed)
- All it needs is a logical (declarative) specification
- CloudDB makes decisions based on application context, workload characteristics, etc.





Language?

New Breed Databases

- CouchDB, Project Voldemort (Dynamo), Cassandra, BigTable, Tokyo Cabinet, MangoDB, SimpleDB,
- MapReduce/Hadoop



By far the most widely used data access <u>language</u>

- It has nothing to do with
 - How the data is stored
 - How the queries are executed
 - How the transactions are handled

Very large number of skilled programmers

Huge amount of existing applications and tools



HIVE: <u>SQL API</u> op top of MapReduce

Google BigQuery: <u>SQL</u> over data stored in non-relational databases

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CloudDB - Guiding Principals

- Embrace heterogeneity
 - One size does not fit all
 - Leverage specialized technologies

Maintain and restore "declarative" nature of data processing

Understand and Define dimensions of scalability



CloudDB Middleware – Opaque vs. Transparent



- System Independence?
- The middleware would be responsible for making all the decisions regarding the choice of data stores, processing the queries, and end-to-end system optimization
- While the middleware can abstract away the underlying storage systems, it should explicitly express certain essential aspects of the system, such as consistency levels and scalability of transactions



CloudDB Platform



CloudDB Platform – Key Points



Our Data Management Platform Key Research Areas





CloudDB System Architecture --Microsharding is a *part* of CloudDB



SQL over Key-Value Stores

<u>Microsharding</u> to enable SQL over key-value stores



- Key-Value stores are good at scaling write intensive workloads
- But, they don't leverage a large body of technologies developed in databases over the decades such as:
 - Relationships
 - Transactions
 - Advanced query functions etc.
- □ These are *hand-coded* by *developers*
- Microsharding aims at bringing those capabilities into keyvalue stores in a principled way



Key Technical Questions Addressed

- How can we map relational schemas to key-value store data models?
- How can we map relational tuples to key-value objects?
- Once we have those mappings, how can we define transaction classes that can be supported in a scalable way in key-value stores?

What are the system implementation issues with such a middleware?



Query and Data Transformation

Physical design: mapping between relational data and K/V data



Microsharding

- A *micro*shard is
 - a logical unit of data
 - a principled way to shard a database into small fragments
 - a unit of transactional data access
 - is accessed by its key, key of root relation





Isolation Levels

No consistency guarantee on read/write outside of a microshard





Scale Independence

- **Experiment Setup**
 - RUBiS benchmark (eBay type auction application)
 - Read/Write workload (transition matrix)
 - Short think time to saturate the system
 - Voldemort (Dynamo) key-value store





Support for Specifying Relaxed Consistency

Tooling to relax consistency just to the degree that there exists a feasible solution (physical design and query plans) for the specification

Scalable Data Organization over heterogeneous data stores

- Physical design over heterogeneous stores such that the service level specifications are met
- Scalability vs. Consistency



The Cast

- NEC Labs Researchers
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 - Wang-Pin Hsiung
 - Hojjat Jafarpour
 - Hyun J. Moon
 - Oliver Po
 - Junichi Tatemura
 - Jagan Sankaranarayanan
- Advisors/Collaborators
 - Michael Carey (U. of California, Irvine)
 - Hector Garcia-Molina (Stanford)
 - Jeff Naughton (U. of Wisconsin, Madison)



A <u>unified data management platform</u> that provides capabilities to <u>transparently</u> and <u>efficiently</u> support <u>heterogeneous workloads</u> by leveraging <u>specialized</u> <u>storage models</u> with <u>SLA-conscious profit optimization</u> in the <u>cloud.</u>



Thank You!

