

Data Integration and Analysis

10 Distributed Data Storage

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Last update: Dec 11, 2020

Announcements/Org

■ #1 Video Recording

- Link in **TeachCenter** & **TUbe** (lectures will be public)
- Optional attendance (independent of COVID)



■ #2 COVID-19 Restrictions (HS i5)

- Corona Traffic Light: **RED** → **ORANGE** (Dec 07)
- Webex lectures reevaluated over holidays



■ #3 Projects and Exercises

- **34x** SystemDS projects, **11x** exercise projects
- Final kickoff discussions last week

■ #4 Course Evaluation

- Evaluation period: **Dec 15 – Jan 31**
- Exam date: **TBD**, but likely end of January

Course Outline Part B: Large-Scale Data Management and Analysis

**12 Distributed Stream
Processing [Jan 15]**

**13 Distributed Machine
Learning Systems [Jan 22]**

Compute/
Storage

11 Distributed Data-Parallel Computation [Jan 08]

10 Distributed Data Storage [Dec 11]

Infra

09 Cloud Resource Management and Scheduling [Dec 04]

08 Cloud Computing Fundamentals [Nov 27]

Agenda

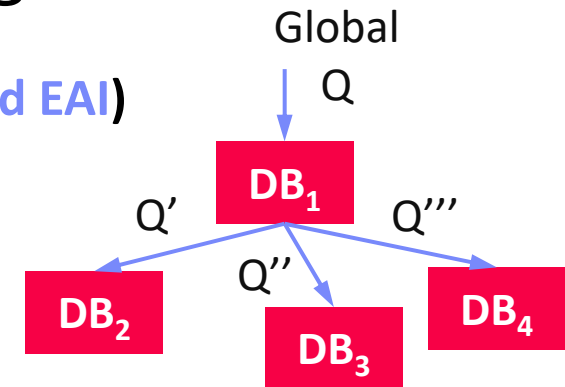
- **Motivation and Terminology**
- **Object Stores and Distributed File Systems**
- **Key-Value Stores and Cloud DBMS**

Motivation and Terminology

Overview Distributed Data Storage

■ Recap: Distributed DBS (03 Replication, MoM, and EAI)

- **Distributed DB**: Virtual (logical) DB, appears like a local DB but consists of multiple physical DBs
- Components for global query processing
- **Virtual DBS** (homo.) vs **federated DBS** (hetero.)



■ Cloud and Distributed Data Storage

- **Motivation**: **size** (large-scale), **semi-structured**/nested , **fault tolerance**
- #1 **Cloud and Distributed Storage**
 - **Block storage**: files split into blocks, read/write (e.g., SAN, AWS EBS)
 - **Object storage**: objects of limited size, read/replace (e.g., AWS S3)
 - **Distributed file systems**: file system on block/object stores (NFS, HDFS)
- #2 **Database as a Service**
 - **NoSQL stores**: Key-value stores, document stores
 - **Cloud DBMSs** (SQL, for OLTP and OLAP workloads)

Central Data Abstractions

■ #1 Files and Objects

- **File:** Arbitrarily large sequential data in specific file format (CSV, binary, etc)
- **Object:** binary large object, with certain meta data

■ #2 Distributed Collections

- Logical multi-set (**bag**) of **key-value pairs** (**unsorted collection**)
- Different physical representations
- **Easy distribution** of pairs via horizontal partitioning (aka shards, partitions)
- Can be created from single file, or directory of files (unsorted)

Key	Value
4	Delta
2	Bravo
1	Alfa
3	Charlie
5	Echo
6	Foxtrot
7	Golf
1	Alfa

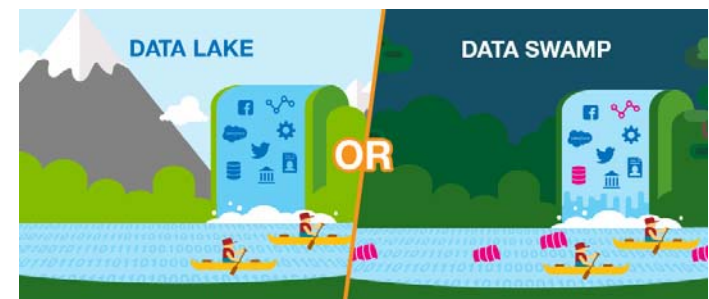
Data Lakes

■ Concept “Data Lake”

- **Store massive amounts of un/semi-structured, and structured data** (append only, no update in place)
- **No need for architected schema** or upfront costs (unknown analysis)
- Typically: file storage in open, raw formats (inputs and intermediates)
- ➔ **Distributed storage and analytics** for scalability and agility

■ Criticism: Data Swamp

- Low data quality (lack of schema, integrity constraints, validation)
 - Missing meta data (context) and data catalog for search
- ➔ **Requires proper data curation / tools**
According to priorities (data governance)



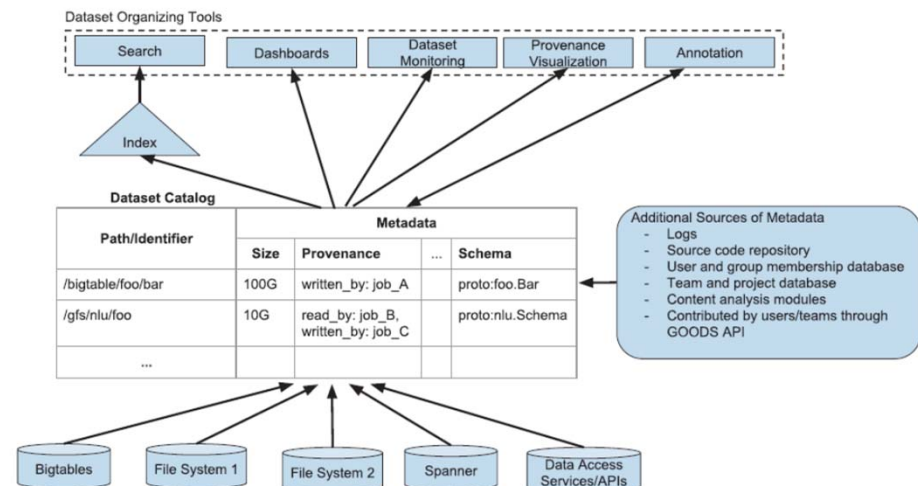
[Credit: www.collibra.com]

- Data curation in repositories for finding relevant datasets in data lakes
- Augment data with open and linked data sources

[Alon Y. Halevy et al: Goods: Organizing Google's Datasets. **SIGMOD 2016**]



Google Data Search



Excursus: Research Data Management (RDM)

■ Overview

- Ensure reproducibility of research results and conclusions
- **Common problem:** “All code and data was on the student’s laptop and the student left / the laptop crashed.”
- **Create value for others** (compare, reuse, understand, extend)
- EU Projects: Mandatory proposal section & deliverable on RDM plan

■ RDM @ TU Graz: <https://www.tugraz.at/sites/rdm/home/>

- Toni Ross-Hellauer and team (ISDS):
Open and Reproducible Research Group (ORRG)
- TU Graz RDM Policy since 12/2019, towards faculty-specific RDM policies

“Ensure that research data, code and any other materials needed to reproduce research findings are appropriately documented, stored and shared in a research data repository in **accordance with the FAIR principles** (Findable, Accessible, Interoperable and Reusable) **for at least 10 years from the end of the research project**, unless there are valid reasons not to do so. [...]

Develop a **written data management strategy** for managing research outputs within the first 12 months of the PhD study as part of their supervision agreements.”

Object Stores and Distributed File Systems

Object Storage

Recap: Key-Value Stores

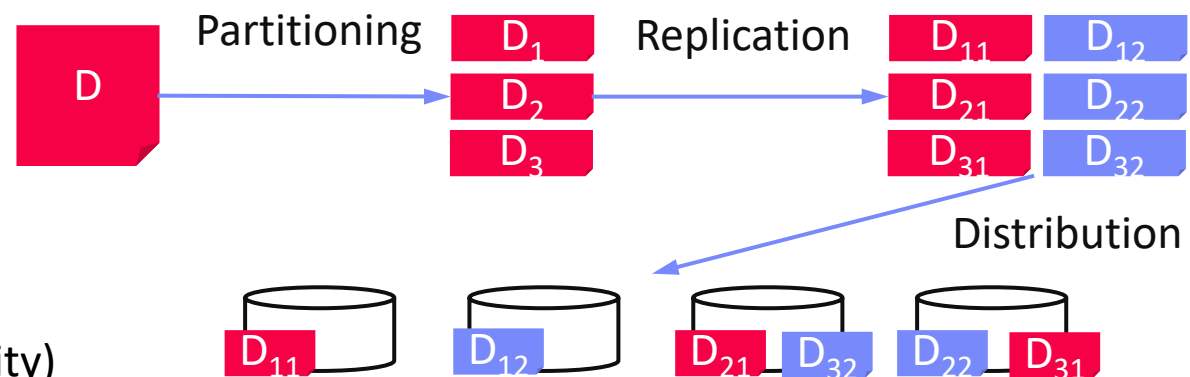
- **Key-value** mapping, where values can be of a variety of data types
- APIs for CRUD operations; scalability via sharding (**objects** or object segments)

Object Store

- Similar to key-value stores, but: **optimized for large objects in GBs and TBs**
- Object identifier (**key**), **meta data**, and object as binary large object (**BLOB**)
- APIs: often REST APIs, SDKs, sometimes implementation of DFS APIs

Key Techniques

- Partitioning
- Replication & Distribution
- Erasure Coding (partitioning + parity)



Object Storage, cont.

■ Example Object Stores / Protocols

- Amazon Simple Storage Service (S3)
- OpenStack Object Storage (Swift)
- IBM Object Storage
- Microsoft Azure Blob Storage



■ Example Amazon S3

- Reliable object store for photos, videos, documents or any binary data
- **Bucket:** Uniquely named, static data container
`http://s3.amazonaws.com/eu-central-1.amazonaws.com/mboehm7datab`
- **Object:** key, version ID, value, metadata, access control
- Single (5GB)/multi-part (5TB) upload and direct/BitTorrent download
- **Storage classes:** STANDARD, STANDARD_IA, GLACIER, DEEP_ARCHIVE
- **Operations:** GET/PUT/LIST/DEL, and SQL over CSV/JSON objects
- Eventual consistency → **Dec 1 2020:** **read-after-write and list consistency**

Hadoop Distributed File System (HDFS)

■ Brief Hadoop History

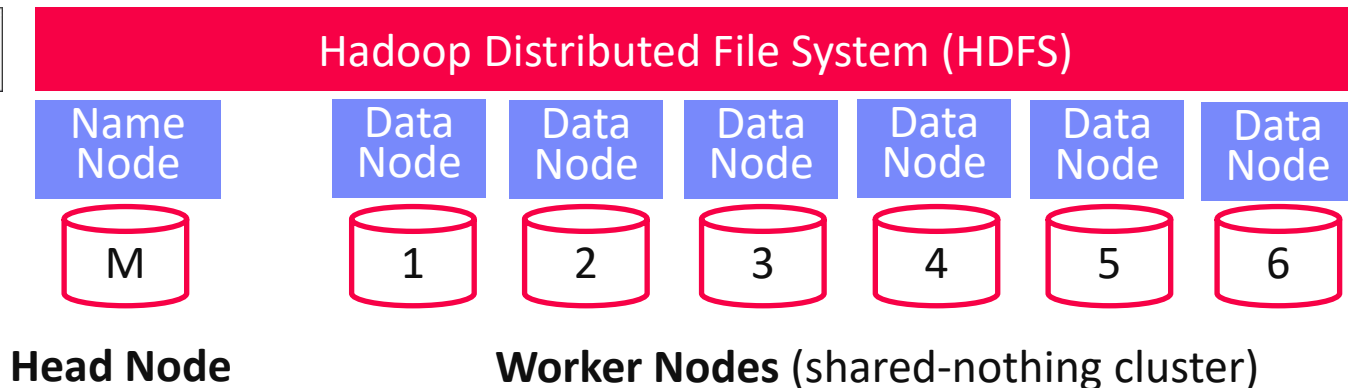
- Google's GFS + MapReduce [ODSI'04]
→ **Apache Hadoop** (2006)
- Apache Hive (SQL), Pig (ETL), Mahout/SystemML (ML), Giraph (Graph)

[Sanjay Ghemawat, Howard Gobioff, Shun-Tak Leung: **The Google file system. SOSP 2003**]



■ HDFS Overview

- Hadoop's distributed file system, for large clusters and datasets
- Implemented in Java, w/ native libraries for compression, I/O, CRC32
- Files split into 128MB blocks, replicated (3x), and distributed



HDFS Daemon Processes

■ HDFS NameNode

- Master daemon that manages file system namespace and access by clients
- Metadata for all files (e.g., replication, permissions, sizes, block ids, etc)
- FSImage**: checkpoint of FS namespace
- EditLog**: **write-ahead-log (WAL)** of file write operations (merged on startup)

```
hadoop fs -ls ./data/mnist1m.bin
```

```
rw-r--r-- 3 mboehm hdfs 104910159 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00000
-rw-r--r-- 3 mboehm hdfs 137887319 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00001
-rw-r--r-- 3 mboehm hdfs 139012247 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00002
-rw-r--r-- 3 mboehm hdfs 139123247 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00003
-rw-r--r-- 3 mboehm hdfs 139053743 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00004
-rw-r--r-- 3 mboehm hdfs 138928955 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00005
-rw-r--r-- 3 mboehm hdfs 139016375 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00006
-rw-r--r-- 3 mboehm hdfs 139047923 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00007
-rw-r--r-- 3 mboehm hdfs 139042307 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00008
-rw-r--r-- 3 mboehm hdfs 139068143 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00009
-rw-r--r-- 3 mboehm hdfs 139029875 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00010
-rw-r--r-- 3 mboehm hdfs 138901043 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00011
-rw-r--r-- 3 mboehm hdfs 139042763 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00012
-rw-r--r-- 3 mboehm hdfs 139030751 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00013
-rw-r--r-- 3 mboehm hdfs 139172051 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00014
-rw-r--r-- 3 mboehm hdfs 138962735 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00015
-rw-r--r-- 3 mboehm hdfs 139075495 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00016
-rw-r--r-- 3 mboehm hdfs 63417008 2018-10-20 22:59 /user/mboehm/data/mnist1m.bin/0-m-00017
```

■ HDFS DataNode

- Worker daemon per cluster node that manages block storage (list of disks)
- Block creation, deletion, replication as individual files in local FS
- On startup: scan local blocks and send **block report** to name node
- Serving block read and write requests
- Send heartbeats to NameNode (capacity, current transfers) and receives replies (replication, removal of block replicas)

HDFS InputFormats and RecordReaders

■ Overview InputFormats

- **InputFormat**: implements access to distributed collections in files
- **Split**: record-aligned block of file (aligned with HDFS block size)
- **RecordReader**: API for reading key-value pairs from file splits
- Examples: FileInputFormat, TextInputFormat, SequenceFileInputFormat

■ Example Text Read

```
FileInputFormat.addInputPath(job, path); # path: dir/file
TextInputFormat informat = new TextInputFormat();
InputSplit[] splits = informat.getSplits(job, numSplits);
```

```
LongWritable key = new LongWritable();
Text value = new Text();
for(InputSplit split : splits) {
    RecordReader<LongWritable,Text> reader = informat
        .getRecordReader(split, job, Reporter.NULL);
    while( reader.next(key, value) )
        ... //process individual text lines
}
```

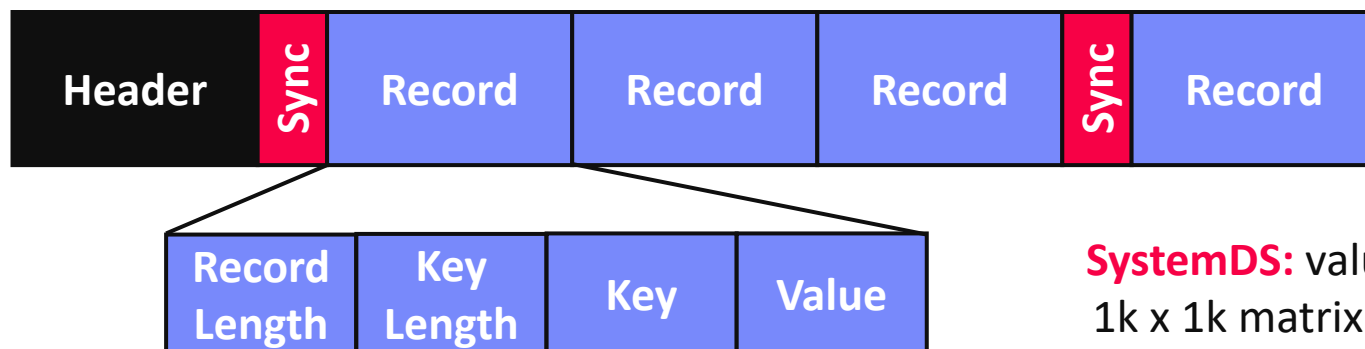

HDFS InputFormats and RecordReaders, cont.

■ Sequence Files

- **Binary files for key/value pairs**, w/ optional compression (MapReduce/Spark inputs/outputs, MapReduce intermediates)
- InputFormat with readers, writers, and sorters

■ Example Uncompressed SequenceFile

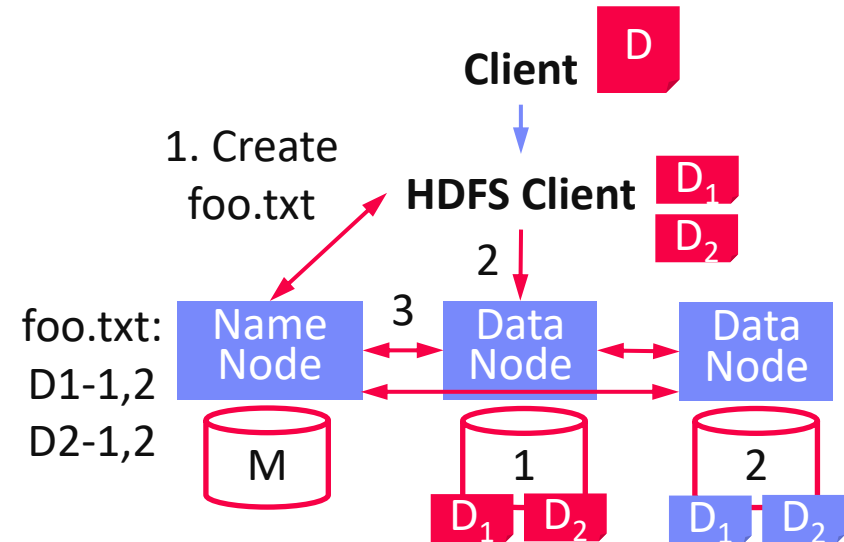
- **Header:** SEQ+version (4 bytes), keyClassName, valueClassName, compression, blockCompression, compressor class (codec), meta data
- Splittable binary representation of key-value pair collection



SystemDS: values are
1k x 1k matrix blocks

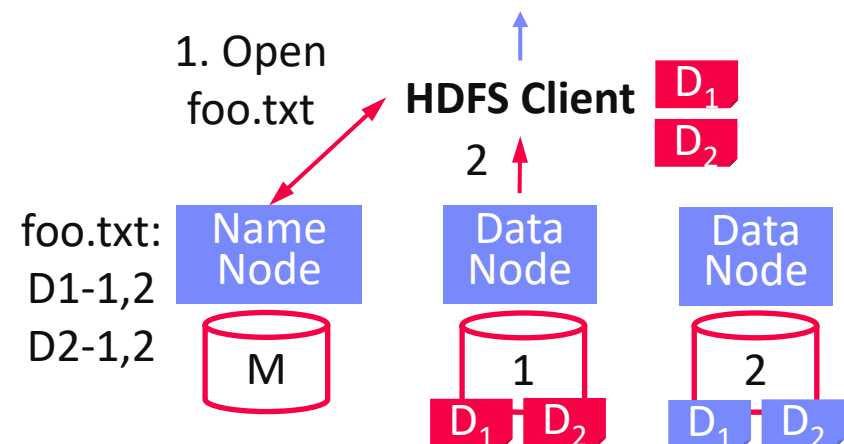
■ HDFS Write

- #1 Client RPC to NameNode to create file → lease/replica DNs
- #2 Write blocks to DNs, pipelined replication to other DNs
- #3 DNs report to NN via heartbeat



■ HDFS Read

- #1 Client RPC to NameNode to open file → DNs for blocks
- #2 Read blocks sequentially from closest DN w/ block
- InputFormats and RecordReaders as abstraction for multi-part files (incl. compression/encryption)



HDFS Data Locality

■ Data Locality

- **HDFS is generally rack-aware** (node-local, rack-local, other)
- Schedule reads from closest data node
- **Replica placement** (rep 3): local DN, other-rack DN, same-rack DN
- MapReduce/Spark: locality-aware execution (**function vs data shipping**)

■ Custom Locality Information

- Custom **InputFormat** and **FileSplit** implementations
- Return customized mapping of locations on `getLocations()`
- Can use block locations of arbitrary files

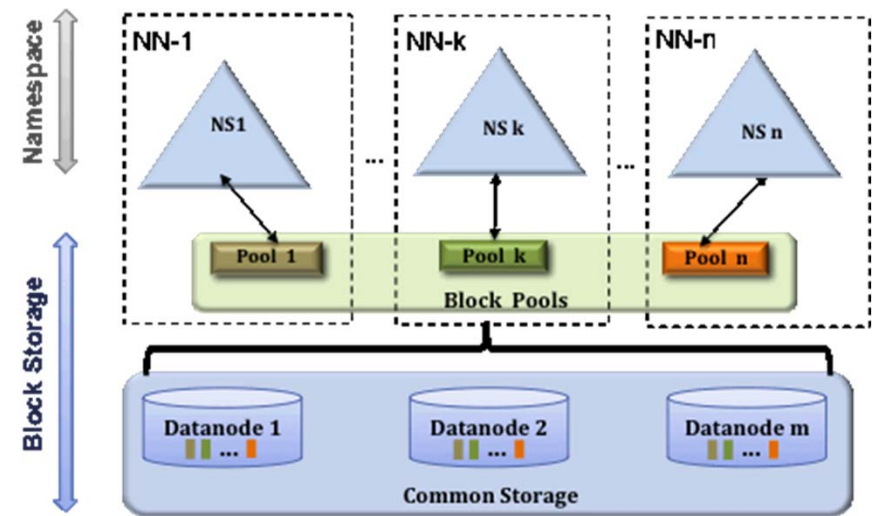
```
public class MyFileSplit extends FileSplit
{
    public MyFileSplit(FileSplit x, ...) {}
    @Override
    public String[] getLocations() {
        return new String[]{"node1", "node7"};
    }
}
```

```
FileStatus st = fs.getFileStatus(new Path(fname));
BlockLocation[] tmp1 = fs.getFileBlockLocations(st, 0, st.getLen());
```

HDFS Federated NameNodes

■ HDFS Federation

- Eliminate NameNode as namespace scalability bottleneck
- Independent NameNodes, responsible for name spaces
- DataNodes store blocks of all NameNodes
- Client-side mount tables



[Credit: <https://hadoop.apache.org/docs/current/hadoop-project-dist/hadoop-hdfs/Federation.html>]

■ GFS Multiple Cells

- *"We also ended up doing what we call a "multi-cell" approach, which basically made it possible to put multiple GFS masters on top of a pool of chunkservers."*
-- Sean Quinlan

[Kirk McKusick, Sean Quinlan:
GFS: evolution on fast-forward.
Commun. **ACM 53(3)** 2010]



Other DFS

■ HDFS FileSystem Implementations (subset)

- LocalFileSystem (**file**), DistributedFileSystem (**hdfs**)
- FTPFileSystem, HttpFileSystem, ViewFilesystem (ViewFs – mount table)
- NativeS3FileSystem (**s3**, **s3a**), NativeSwiftFileSystem, NativeAzureFileSystem
- Other proprietary: IBM **GPFS**, Databricks FS (DBFS)

■ Google Colossus

- More fine-grained accesses, Google Cloud Storage

[WIRED: Google Remakes
Online Empire With 'Colossus',
<https://www.wired.com/2012/07/google-colossus/>]

■ High-Performance Computing

- **Scope**: Focus on high IOPs (instead of bandwidth) with block write
- IBM **GPFS** (General Parallel File System) / Spectrum Scale
- **BeeGFS** (Fraunhofer GFS) – focus on usability, storage/metadata servers
- **Lustre** (Linux + Cluster) – GPL license, LNET protocol / metadata / object storage
- RedHat **GFS2** (Global File System) – Linux cluster file system, close to local
- **NAS** (Network Attached Storage), **SAN** (Storage Area Network)
- **GekkoFS** (Uni Mainz / Barcelona SC) – data-intensive HPC applications

Key-Value Stores and Cloud DBMS

Motivation and Terminology

■ Motivation

- **Basic key-value mapping via simple API** (more complex data models can be mapped to key-value representations)
- **Reliability at massive scale on commodity HW** (cloud computing)

■ System Architecture

- **Key**-value maps, where values can be of a variety of data types
- APIs for CRUD operations (create, read, update, delete)
- Scalability via sharding (horizontal partitioning)

users:1:a

“Inffeldgasse 13, Graz”

users:1:b

“[12, 34, 45, 67, 89]”

users:2:a

“Mandellstraße 12, Graz”

users:2:b

“[12, 212, 3212, 43212]”

■ Example Systems

- **Dynamo** (2007, AP) → **Amazon DynamoDB** (2012)
- **Redis** (2009, CP/AP)



redis



[Giuseppe DeCandia et al: Dynamo: amazon's highly available **key-value store**. SOSP 2007]



Example Systems: Dynamo

[Giuseppe DeCandia et al:
Dynamo: amazon's highly available
key-value store. SOSP 2007]



■ Motivation

- **Simple, highly-available** data storage for small objects in ~1MB range
- Aim for **good load balance** (99.9th percentile SLAs)

■ #1 System Interface

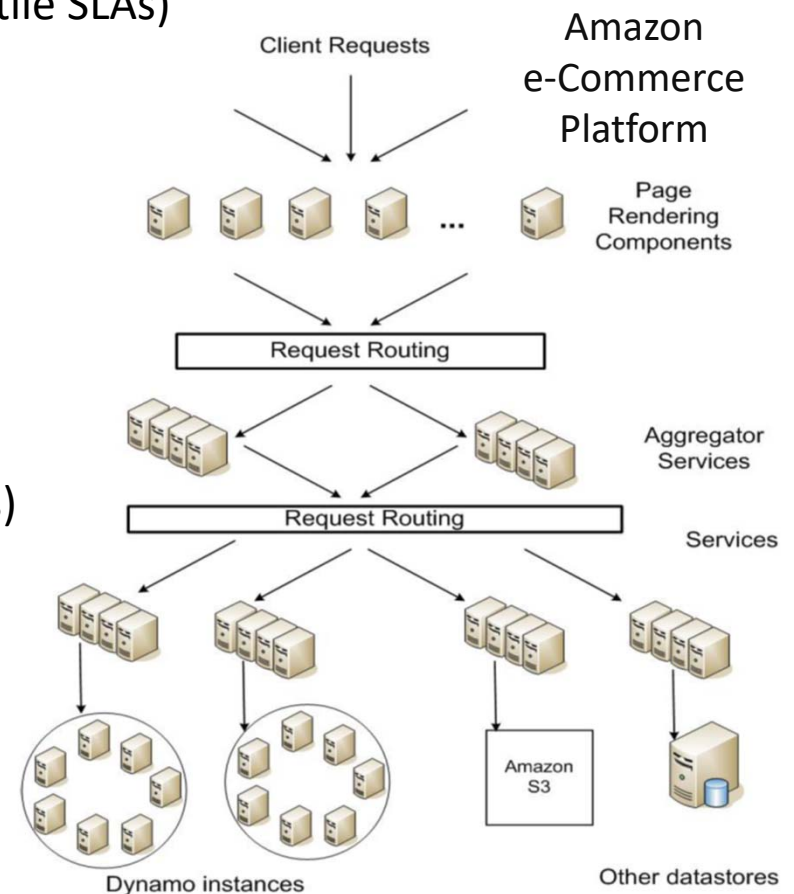
- Simple get(k, ctx) and put(k, ctx) ops

■ #2 Partitioning

- **Consistent hashing** of nodes and keys on circular ring for **incremental scaling**
- Nodes hold **multiple virtual nodes** for **load balance** (add/rm, heterogeneous)

■ #3 Replication

- Each data item **replicated N times** (at coord node and N-1 successors)
- Eventual consistency with async update propagation based on **vector clocks**
- Replica synchronization via **Merkle trees**



Example Systems, cont.

■ Redis Data Types



- Redis is not a plain KV-store, but “data structure server” with persistent log (**appendfsync no/everysec/always**)
- **Key:** ASCII string (max 512MB, common key schemes: comment:1234:reply.to)
- **Values:** strings, lists, sets, sorted sets, hashes (map of string-string), etc

■ Redis APIs

- **SET/GET/DEL:** insert a key-value pair, lookup value by key, or delete by key
- **MSET/MGET:** insert or lookup multiple keys at once
- **INCRBY/DECBY:** increment/decrement counters
- Others: EXISTS, LPUSH, LPOP, LRANGE, LTRIM, LLEN, etc

■ Other systems

- Classic KV stores (AP): **Riak**, **Aerospike**, **Voldemort**, **LevelDB**, **RocksDB**, **FoundationDB**, **Memcached**
- Wide-column stores: **Google BigTable** (CP), **Apache HBase** (CP), **Apache Cassandra** (AP)



LEVELDB

APACHE
HBASE

Log-structured Merge Trees

[Patrick E. O'Neil, Edward Cheng,
Dieter Gawlick, Elizabeth J. O'Neil:
The Log-Structured Merge-Tree
(LSM-Tree). *Acta Inf.* 1996]

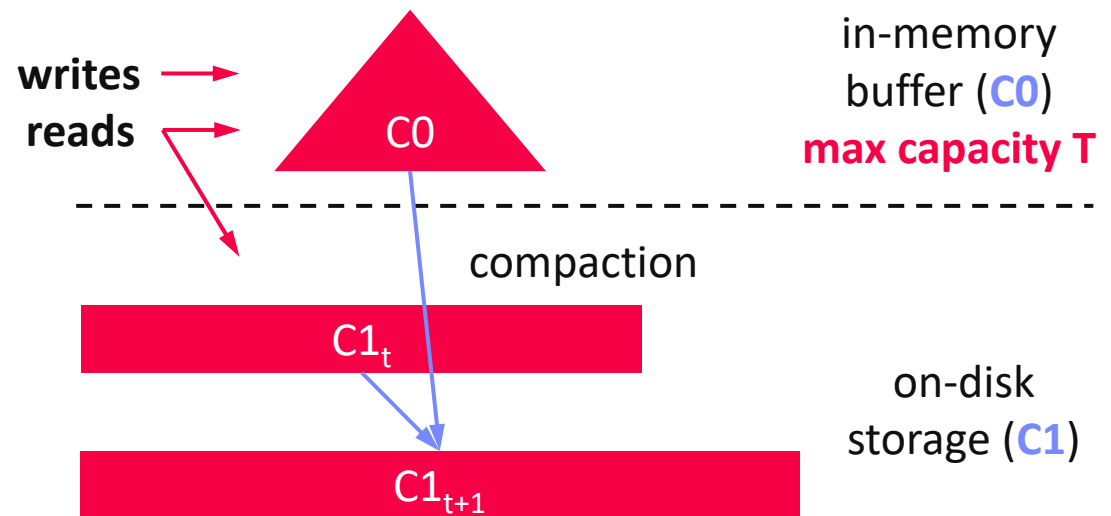


■ LSM Overview

- Many KV-stores rely on LSM-trees as their storage engine (e.g., **BigTable**, **DynamoDB**, **LevelDB**, **Riak**, **RocksDB**, **Cassandra**, **HBase**)
- Approach:** Buffers writes in memory, flushes data as sorted runs to storage, merges runs into larger runs of next level (compaction)

■ System Architecture

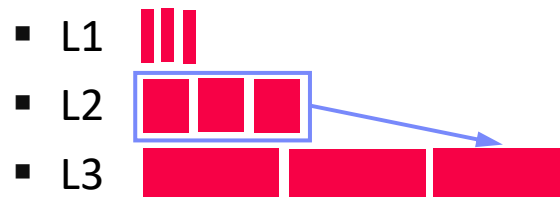
- Writes in C0
- Reads against C0 and C1 (w/ buffer for C1)
- Compaction (rolling merge): sort, merge, including **deduplication**



Log-structured Merge Trees, cont.

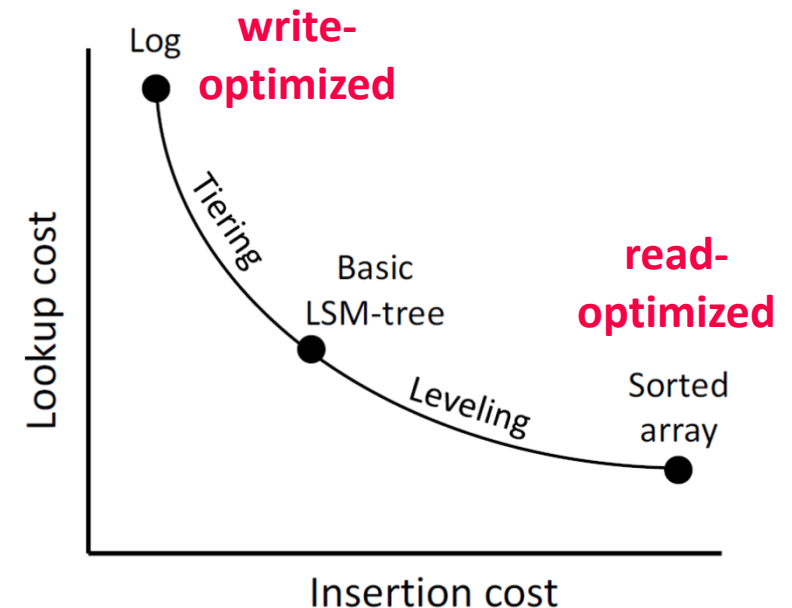
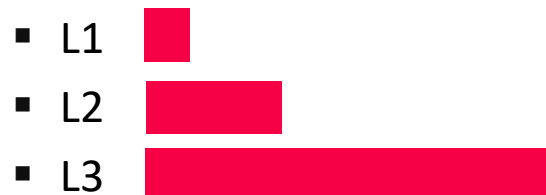
LSM Tiering

- Keep up to $T-1$ runs per level L
- Merge all runs of L_i into 1 run of L_{i+1}



LSM Leveling

- Keep 1 run per level L
- Merge run of L_i with L_{i+1}



[Niv Dayan: Log-Structured-Merge Trees, **Comp115** guest lecture, 2017]



[Stratos Idreos, Mark Callaghan: Key-Value Storage Engines (Tutorial), **SIGMOD 2020**]



Cloud Databases (DBaaS)

■ Motivation DBaaS

- Simplified setup, maintenance, tuning and auto scaling
- Multi-tenant systems (scalability, learning opportunities)
- Different types based on workload (OLTP vs OLAP)



■ Elastic Data Warehouses

- Motivation: Intersection of data warehousing (**02 DWH, ETL, SQL/OLAP**), cloud computing (**08/09 Cloud Computing**), Distributed Storage (**10 today**)
- Example Systems
 - **#1** Snowflake
 - **#2** Google BigQuery (Dremel)
 - **#3** Amazon Redshift
 - Azure SQL Data Warehouse

Commonalities:
SQL, **column stores**,
data on **object store / DFS**,
elastic cloud scaling

Example Snowflake

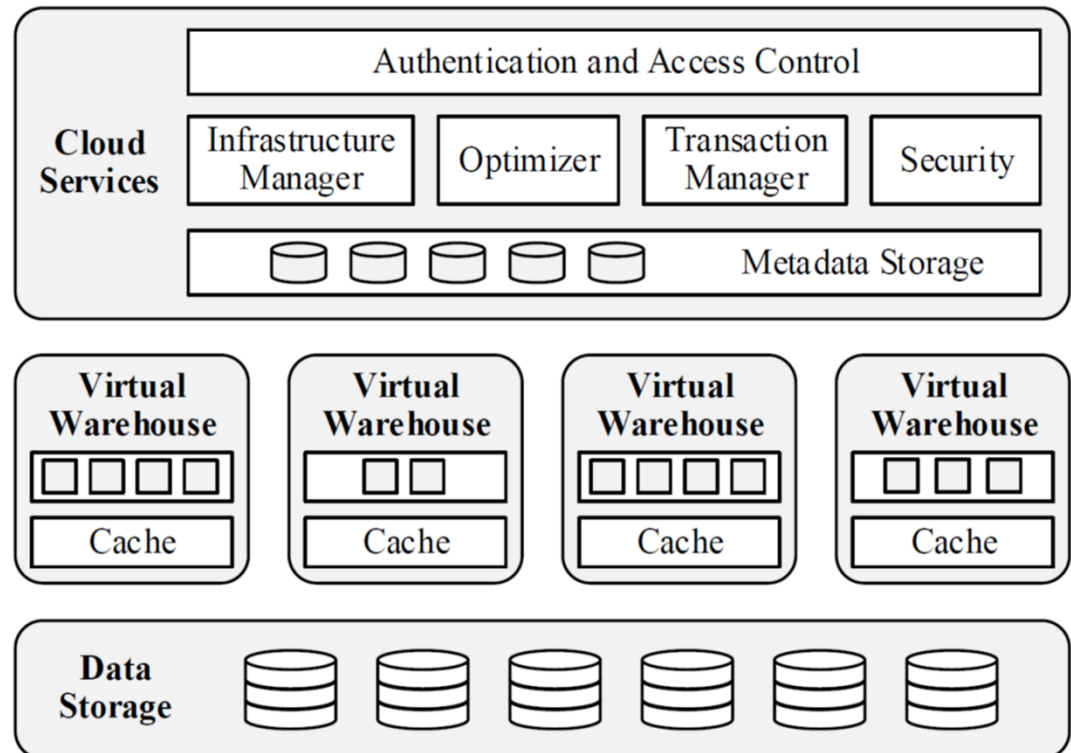
[Benoît Dageville et al.: The Snowflake Elastic Data Warehouse. **SIGMOD 2016**]



- **Motivation** (impl started late 2012)
 - Enterprise-ready DWH solution for the cloud (elasticity, semi-structured)
 - Pure SaaS experience, high availability, cost efficient



- **Cloud Services**
 - Manage virtual DHWs, TXs, and queries
 - Meta data and catalogs
- **Virtual Warehouses**
 - Query execution in EC2
 - Caching/intermediates
- **Data Storage**
 - Storage in AWS S3
 - PAX / hybrid columnar
 - Min-max pruning



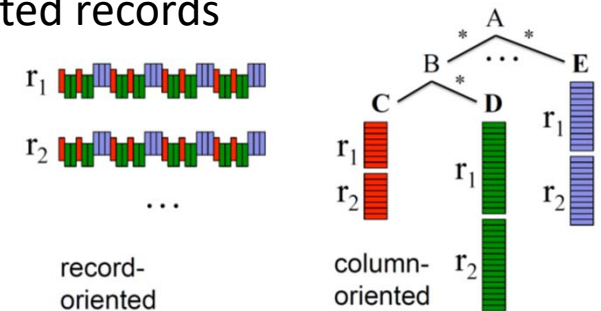
Example Google BigQuery

[Sergey Melnik et al.: Dremel: Interactive Analysis of Web-Scale Datasets. **PVLDB 3(1) 2010**]



■ Background Dremel

- Scalable and fast **in-situ analysis of read-only nested data** (DFS, BigTable)
- **Data model:** protocol buffers - strongly-typed nested records
- **Storage model:** **columnar storage of nested data** (efficient splitting and assembly records)
- Query execution via **multi-level serving tree**

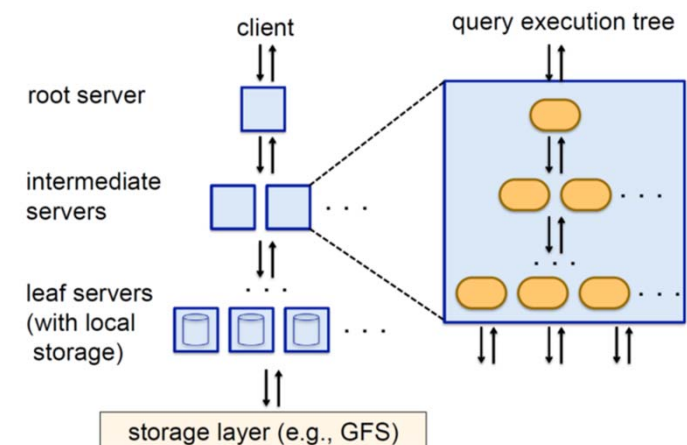


■ BigQuery System Architecture

- Public impl of internal Dremel system (2012)
- SQL over structured, nested data (OLAP, BI)
- **Extensions:** web Uis, REST APIs and ML
- **Data storage:** Colossus (**NextGen GFS**)



[Kazunori Sato: An Inside Look at Google BigQuery, Google BigQuery White Paper 2012.]



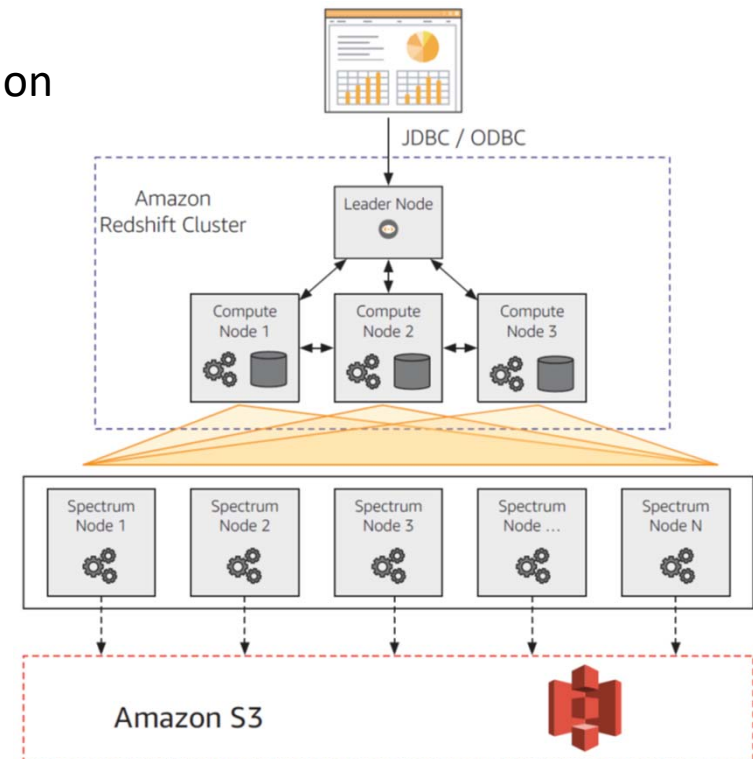
Example Amazon Redshift

- **Motivation** (release 02/2013)
 - **Simplicity and cost-effectiveness**
(fully-managed DWH at petabyte scale)
- **System Architecture**
 - **Data plane:** data storage and SQL execution
 - **Control plane:** workflows for monitoring, and managing databases, AWS services
- **Data Plane**
 - Initial engine licensed from ParAccel
 - Leader node + sliced compute nodes in **EC2** (with **local storage**)
 - Replication across nodes + **S3 backup**
 - **Query compilation** in C++ code
 - Support for **flat and nested files**

[Anurag Gupta et al.: Amazon Redshift and the Case for Simpler Data Warehouses. **SIGMOD 2015**]



[Mengchu Cai et al.: Integrated Querying of SQL database data and S3 data in Amazon Redshift. **IEEE Data Eng. Bull.** 41(2) 2018]



Summary and Q&A

- Motivation and Terminology
- Object Stores and Distributed File Systems
- Key-Value Stores and Cloud DBMS

- Next Lectures
 - [11 Distributed, Data-Parallel Computation](#) [Jan 08]
 - [12 Distributed Stream Processing](#) [Jan 15]
 - [13 Distributed Machine Learning Systems](#) [Jan 22]

**Happy
Holidays**