







## BigTable: Introduction

Let's explore Bigtable and its use cases.

We'll cover the following



- Goal
- What is BigTable?
- Background
- BigTable use cases

#### Goal#

Design a distributed and scalable system that can store a huge amount of structured data. The data will be indexed by a row key where each row can have an unbounded number of columns.

## What is BigTable?#

BigTable is a **distributed** and **massively scalable** wide-column store. It is designed to store huge sets of structured data. As its name suggests, BigTable provides storage for very big tables (often in the terabyte range).

In terms of the CAP theorem, BigTable is a CP system, i.e., **it has strictly consistent reads and writes**. BigTable can be used as an input source or output destination for MapReduce jobs.











BigTable was developed at Google and has been in use since 2005 in dozens of Google services. Because of the large scale of its services, Google could not use commercial databases. Also, the cost of using an external solution would have been too high. That is why Google chose to build an in-house solution. BigTable is a highly available and high-performing database that powers multiple applications across Google — where each application has different needs in terms of the size of data to be stored and latency with which results are expected.

Though BigTable is not open-source itself, its paper was crucial in inspiring powerful open-source databases like Cassandra (which borrows BigTable's data model), HBase (a distributed non-relational database) and Hypertable.

## BigTable use cases#

Google built BigTable to store large amounts of data and perform thousands of queries per second on that data. Examples of BigTable data are billions of URLs with many versions per page, petabytes of Google Earth data, and billions of users' search data.

BigTable is suitable to store large datasets that are greater than one TB where each row is less than 10MB. Since BigTable does not provide ACID (atomicity, consistency, isolation, durability) properties or transaction support, Online Transaction Processing (OLTP) applications with transaction processes should not use BigTable. For BigTable, data should be structured in the form of key-value pairs or rows-columns. Non-structured data like images or movies should not be stored in BigTable.

Here are the examples of data that Google stores in BigTable:

• URL and its related data, e.g., PageRank, page contents, crawl metadat. (e.g., when a page was crawled, what was the response code, etc.), links,







anchors (links pointing to a page). There are billions of URLs with many versions of a page.

• Per-user data, e.g., preference settings, recent queries/search results. Google has hundreds of millions of users.

#### BigTable can be used to store the following types of data:

- 1. Time series data: As the data is naturally ordered
- 2. Internet of Things (IoT) data: Constant streams of writes
- 3. Financial Data: Often represented as time-series data













## BigTable Data Model

This lesson explains how BigTable models its data.

We'll cover the following

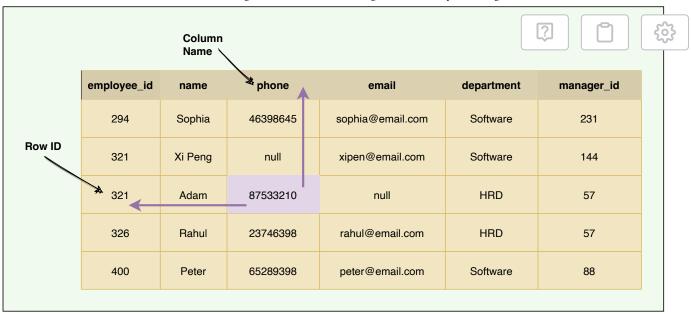


- Rows
- Column families
- Columns
- Timestamps

In simple terms, BigTable can be characterized as a sparse, distributed, persistent, multidimensional, sorted map. Let's dig deeper to understand each of these characteristics of BigTable.

Traditional DBs have a two-dimensional layout of the data, where each cell value is identified by the 'Row ID' and 'Column Name':



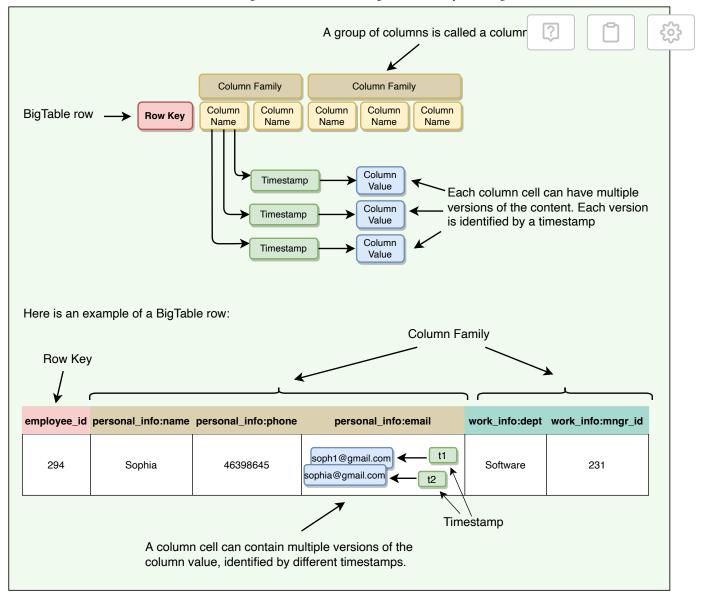


Two-dimensional layout of a traditional database

#### BigTable has a **four-dimensional data model**. The four dimensions are:

- 1. Row Key: Uniquely identifies a row
- 2. Column Family: Represents a group of columns
- 3. Column Name: Uniquely identifies a column
- 4. Timestamp: Each column cell can have different versions of a value, each identified by a timestamp





BigTable's four-dimensional data model

The data is indexed (or sorted) by row key, column key, and a timestamp. Therefore, to access a cell's contents, we need values for all of them. If no timestamp is specified, BigTable retrieves the most recent version.

( row\_key: string, column\_name: string, timestamp: int64 ) → cell
contents (string)

#### Rows#



Each row in the table has an associated row key that is an argue up to 64 kilobytes in size (although most keys are significantly smaller).

- Each row is uniquely identified by the 'row key.'
- Each 'row key' is internally represented as a string.
- Every read or write of data under a single row is atomic. This also means that atomicity across rows is not guaranteed, e.g., when updating two rows, one might succeed, and the other might fail.
- Each table's data is only indexed by row key, column key, and timestamp. There are no secondary indices.

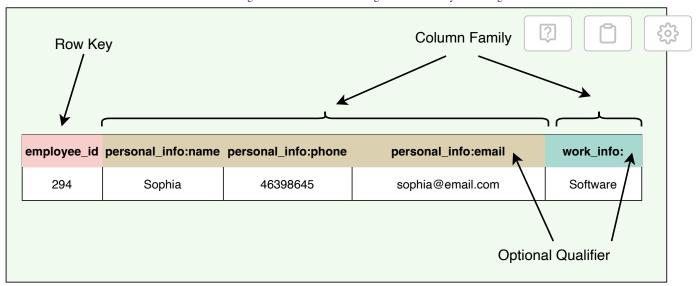
A **column** is a key-value pair where the key is represented as 'column key' and the value as 'column value.'

### Column families#

Column keys are grouped into sets called column families. All data stored in a column family is usually of the same type. The number of distinct column families in a table should be small (in the hundreds at maximum), and families should rarely change during operation. Access control as well as both disk and memory accounting are performed at the column-family level.

The following figure shows a single row from a table. The row key is 294, and there are two column families: personal\_info and work\_info, with three columns under the personal\_info column family.





Column families

- Column family format: family:optional qualifier
- All rows have the same set of column families.
- BigTable can retrieve data from the same column family efficiently.
- Short Column family names are better as names are included in the data transfer.

#### Columns#

- Columns are units within a column family.
- A BigTable may have an unbounded number of columns.
- New columns can be added on the fly.
- Short column names are better as names are passed in each data transfer, e.g., ColumnFamily:ColumnName => Work:Dept
- As mentioned above, BigTable is quite suitable for **sparse data**. This is because empty columns are not stored.

## Timestamps#

Each column cell can contain multiple versions of the content. For example is as we saw in the earlier example, we may have several timestamped

versions of an employee's email. A 64-bit timestamp identifie that either represents real time or a custom value assigned by the chem. While reading, if no timestamp is specified, BigTable returns the most recent version. If the client specifies a timestamp, the latest version that is earlier than the specified timestamp is returned.

BigTable supports two per-column-family settings to garbage-collect cell versions automatically. The client can specify that only the last 'n' versions of a cell be kept, or that only new-enough versions be kept (e.g., only keep values that were written in the previous seven days).













## System APIs

Let's explore BigTable APIs.

We'll cover the following



- Metadata operations
- Data operations

BigTable provides APIs for two types of operations:

- Metadata operations
- Data operations

## Metadata operations#

BigTable provides APIs for creating and deleting tables and column families. It also provides functions for changing cluster, table, and column family metadata, such as access control rights.

## Data operations#

Clients can insert, modify, or delete values in BigTable. Clients can also lookup values from individual rows or iterate over a subset of the data in a table.

 BigTable supports single-row transactions, which can be used to perform atomic read-modify-write sequences on data stored under a single row key.









- Bigtable does not support transactions across row keys, but provides a client interface for batch writing across row keys.
- BigTable allows cells to be used as integer counters.
- A set of wrappers allow a BigTable to be used both as an input source and as an output target for MapReduce jobs.
- Clients can also write scripts in Sawzall (a language developed at Google) to instruct server-side data processing (transform, filter, aggregate) prior to the network fetch.

Here are APIs for write operations:

- **Set()**: write cells in a row
- **DeleteCells()**: delete cells in a row
- **DeleteRow()**: delete all cells in a row

A read or scan operation can read arbitrary cells in a BigTable:

- Each row read operation is atomic.
- Can ask for data from just one row, all rows, etc.
- Can restrict returned rows to a particular range.
- Can ask for all columns, just certain columns families, or specific columns.



Next  $\rightarrow$ 

BigTable Data Model

Partitioning and High-level Architecture



Completed



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## Partitioning and High-level Architecture

This lesson gives a brief overview of BigTable's architecture and its data partitioning scheme.

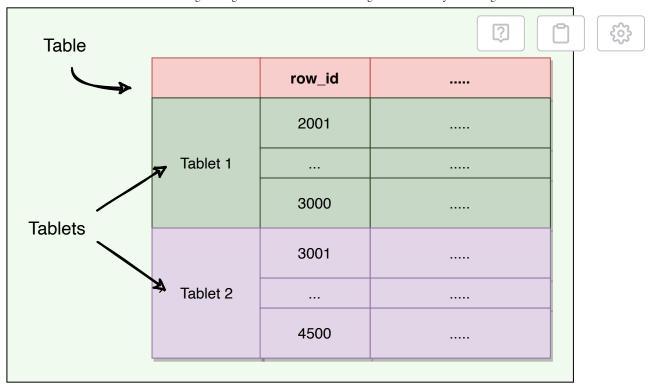
We'll cover the following

- Table partitioning
- High-level architecture

## Table partitioning#

A single instance of a BigTable implementation is known as a cluster. Each cluster can store a number of tables where each table is split into multiple **Tablets**, each around 100–200 MB in size.





- A Tablet holds a contiguous range of rows.
- The table is broken into Tablets at row boundaries.
- Initially, each table consists of only one Tablet. As the table grows, multiple Tablets are created. By default, a table is split at around 100 to 200 MB.
- Tablets are the unit of distribution and load balancing (more about this later).
- Since the table is sorted by row, reads of short ranges of rows are always efficient, that is to say, communicating with a small number of Tablets. This also means that selecting a row key with a high degree of locality is very important.
- Each Tablet is assigned to a **Tablet server** (discussed later), which manages all read/write requests of that Tablet.

## High-level architecture#

The architecture of a BigTable cluster consists of three major components: 🌣

1. **Client Library**: A library component that is linked into client talks to BigTable through this library.

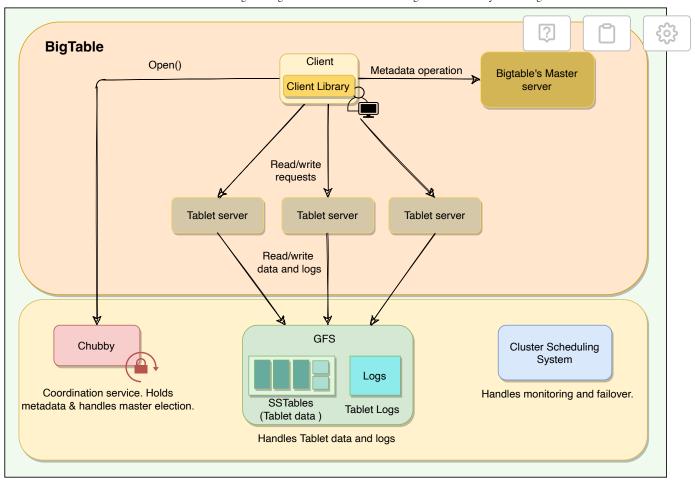
- 2. **One master server**: Responsible for performing metadata operations and assigning Tablets to Tablet servers and managing them.
- 3. **Many Tablet servers**: Each Tablet server serves read and write of the data to the Tablets it is assigned.

BigTable is built on top of several other pieces from Google infrastructure:

- 1. GFS: BigTable uses the Google File System to store its data and log files.
- 2. **SSTable**: Google's SSTable (Sorted String Table) file format is used to store BigTable data. SSTable provides a persistent, ordered, and immutable map from keys to values (more on this later). SSTable is designed in such a way that any data access requires, at most, a single disk access.
- 3. **Chubby**: BigTable uses a highly available and persistent distributed lock service called Chubby to handle synchronization issues and store configuration information.
- 4. **Cluster Scheduling System**: Google has a cluster management system that schedules, monitors, and manages the Bigtable's cluster.

Let's understand these components one by one.





High-level architecture of BigTable



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### **SSTable**

Let's learn how Tablets are stored in SSTables.

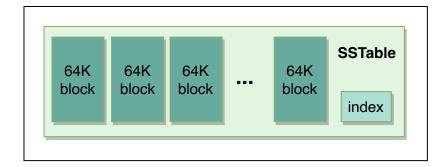


- How are Tablets stored in GFS?
  - Table vs. Tablet vs. SSTable

#### How are Tablets stored in GFS?#

BigTable uses Google File System (GFS), a persistent distributed file storage system to store data as files. The file format used by BigTable to store its files is called SSTable:

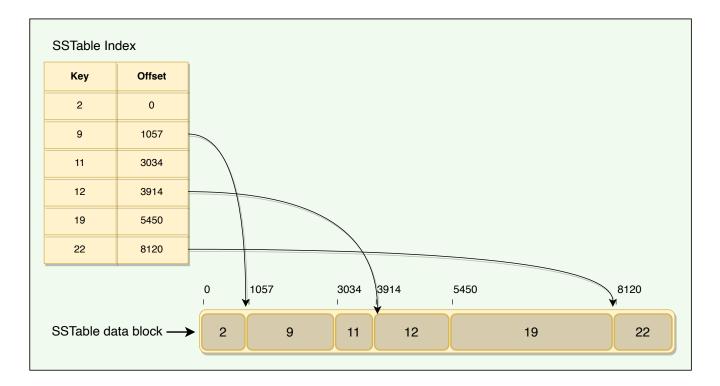
- SSTables are persisted, ordered maps of keys to values, where both keys and values are arbitrary byte strings.
- Each Tablet is stored in GFS as a sequence of files called SSTables.
- An SSTable consists of a sequence of data blocks (typically 64KB in size).



SSTable contains multiple blocks



• A block index is used to locate blocks; the index is loade when the SSTable is opened.



Reading data from SSTable

- A lookup can be performed with a single disk seek. We first find the appropriate block by performing a binary search in the in-memory index, and then reading the appropriate block from the disk.
- To read data from an SSTable, it can either be copied from disk to memory as a whole or just the index. The former approach avoids subsequent disk seeks for lookups, while the latter requires a single disk seek for each lookup.
- SSTables provide two operations:
  - o Get the value associated with a given key
  - o Iterate over a set of values in a given key range
- Each SSTable is immutable (read-only) once written to GFS. If new data is added, a new SSTable is created. Once an old SSTable is no longer needed, it is set out for garbage collection. SSTable immutability is at the

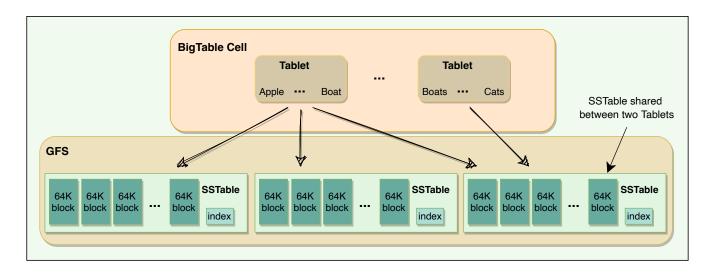
core of BigTable's data checkpointing and recovery rout immutability provides following advantages:

- No synchronization is needed during read operations.
- This also makes it easier to split Tablets.
- Garbage collector handles the permanent removal of deleted or stale data.

#### Table vs. Tablet vs. SSTable#

Here is how we can define the relationship between Table, Tablet and SStable:

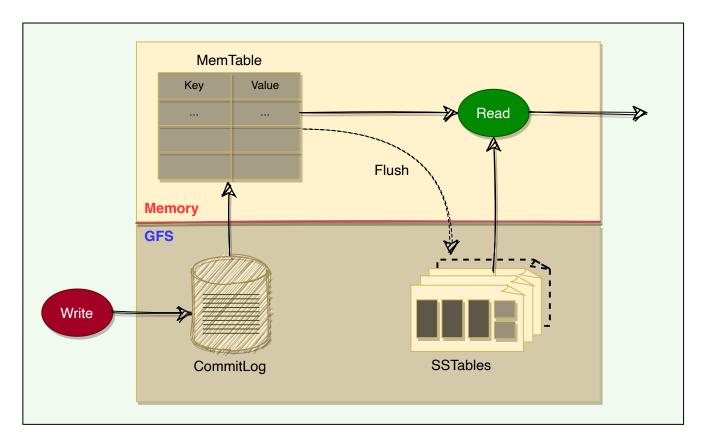
- Multiple Tablets make up a table.
- SSTables can be shared by multiple Tablets.
- Tablets do not overlap, SSTables can overlap.



Tablet vs SSTable

• To improve write performance, BigTable uses an in-memory, mutable sorted buffer called **MemTable** to store recent updates. As more writes are performed, MemTable size increases, and when it reaches a threshold, the MemTable is frozen, a new MemTable is created, and the frozen MemTable is converted to an SSTable and written to GFS.

- Each data update is also written to a commit-log which i GFS. This log contains redo records used for recovery if a raplet server fails before committing a MemTable to SSTable.
- While reading, the data can be in MemTables or SSTables. Since both these tables are sorted, it is easy to find the most recent data.



Read and write workflow



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## **GFS** and Chubby

Let's explore how BigTable interacts with GFS and Chubby.

We'll cover the following



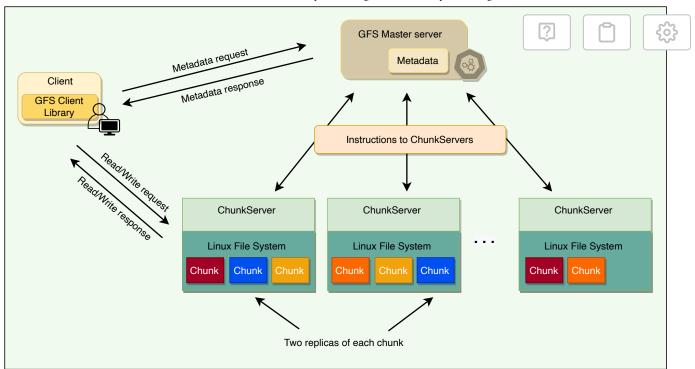
- GFS
- Chubby

#### GFS#

GFS is a scalable distributed file system developed by Google for its large data-intensive applications such as BigTable.

- GFS files are broken into fixed-size blocks, called Chunks.
- Chunks are stored on data servers called ChunkServers.
- GFS master manages the metadata.
- SSTables are divided into fixed-size, blocks and these blocks are stored on ChunkServers.
- Each chunk in GFS is replicated across multiple ChunkServers for reliability.
- Clients interact with the GFS master for metadata, but all data transfers happen directly between the client and ChunkServers.





High-level architecture of GFS

For a detailed discussion, please see GFS.

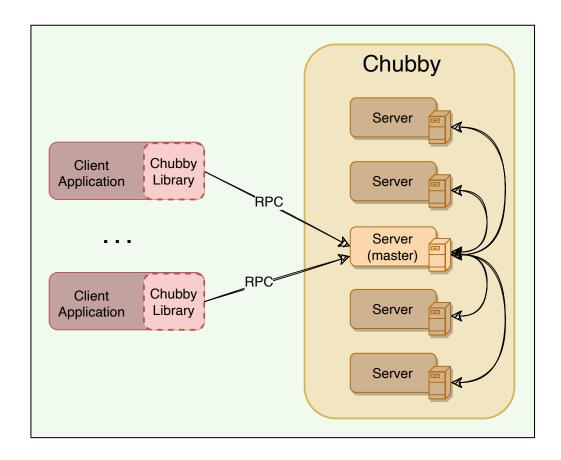
## Chubby#

Chubby is a highly available and persistent distributed locking service that allows a multi-thousand node Bigtable cluster to stay coordinated.

- Chubby usually runs with five active replicas, one of which is elected as the master to serve requests. To remain alive, a majority of Chubby replicas must be running.
- BigTable depends on Chubby so much that if Chubby is unavailable for an extended period of time, BigTable will also become unavailable.
- Chubby uses the Paxos algorithm to keep its replicas consistent in the face of failure.
- Chubby provides a namespace consisting of files and directories. Each file or directory can be used as a lock.
- Read and write access to a Chubby file is atomic.



- Each Chubby client maintains a session with a Chubby session expires if it is unable to renew its session lease within the rease expiration time. When a client's session expires, it loses any locks and open handles. Chubby clients can also register callbacks on Chubby files and directories for notification of changes or session expiration.
- In BigTable, Chubby is used to:
  - Ensure there is only one active master. The master maintains a session lease with Chubby and periodically renews it to retain the status of the master.
  - Store the bootstrap location of BigTable data (discussed later)
  - Discover new Tablet servers as well as the failure of existing ones
  - Store BigTable schema information (the column family information for each table)
  - Store Access Control Lists (ACLs)



High-level architecture of Chubby





SSTable



Bigtable Components



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## **Bigtable Components**

Let's explore the components that constitute BigTable.

We'll cover the following

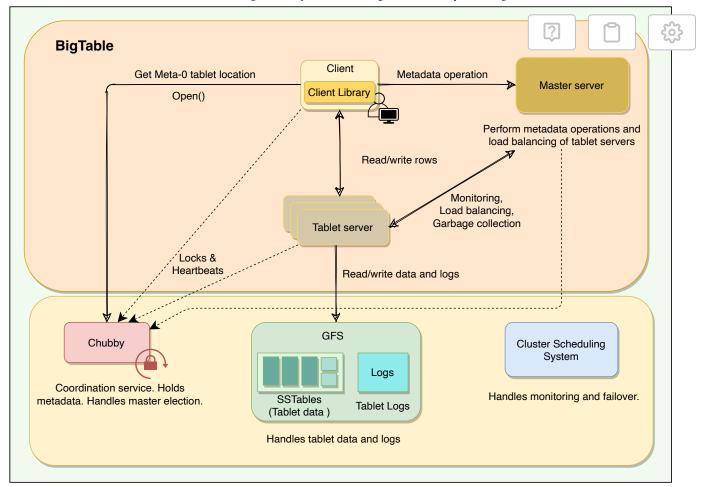


- BigTable master server
- Tablet servers

As described previously, a BigTable cluster consists of three major components:

- 1. A library component that is linked into every client
- 2. One master server
- 3. Many Tablet servers





BigTable architecture

## BigTable master server#

There is only one master server in a BigTable cluster, and it is responsible for:

- Assigning Tablets to Tablet servers and ensuring effective load balancing
- Monitoring the status of Tablet servers and managing the joining or failure of Tablet server
- Garbage collection of the underlying files stored in GFS
- Handling metadata operations such as table and column family creations



Bigtable master is not involved in the core task of mapping to underlying files in GFS (Tablet servers handle this). This means that bigtable clients do not have to communicate with the master at all. This design decision significantly reduces the load on the master and the possibility of the master becoming a bottleneck.

#### Tablet servers#

- Each Tablet server is assigned ownership of a number of Tablets (typically 10–1,000 Tablets per server) by the master.
- Each Tablet server serves read and write requests of the data of the Tablets it is assigned. The client communicates directly with the Tablet servers for reads/writes.
- Tablet servers can be added or removed dynamically from a cluster to accommodate changes in the workloads.
- Tablet creation, deletion, or merging is initiated by the master server, while Tablet partition is handled by Tablet servers who notify the master.













## Working with Tablets

We'll cover the following



- Locating Tablets
- Assigning Tablets
- Monitoring Tablet servers
- Load-balancing Tablet servers

## Locating Tablets#

Since Tablets move around from server to server (due to load balancing, Tablet server failures, etc.), given a row, how do we find the correct Tablet server? To answer this, we need to find the Tablet whose row range covers the target row. BigTable maintains a 3-level hierarchy, analogous to that of a B+ tree, to store Tablet location information.

BigTable creates a special table, called **Metadata** table, to store Tablet locations. This Metadata table contains one row per Tablet that tells us which Tablet server is serving this Tablet. Each row in the METADATA table stores a Tablet's location under a row key that is an encoding of the Tablet's table identifier and its end row.

#### METADATA:

Key: table id + end row

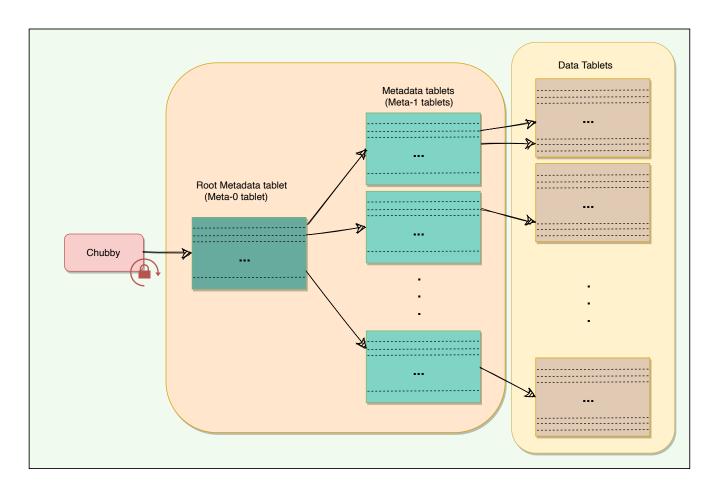
Data: tablet server location



BigTable stores the information about the Metadata table in



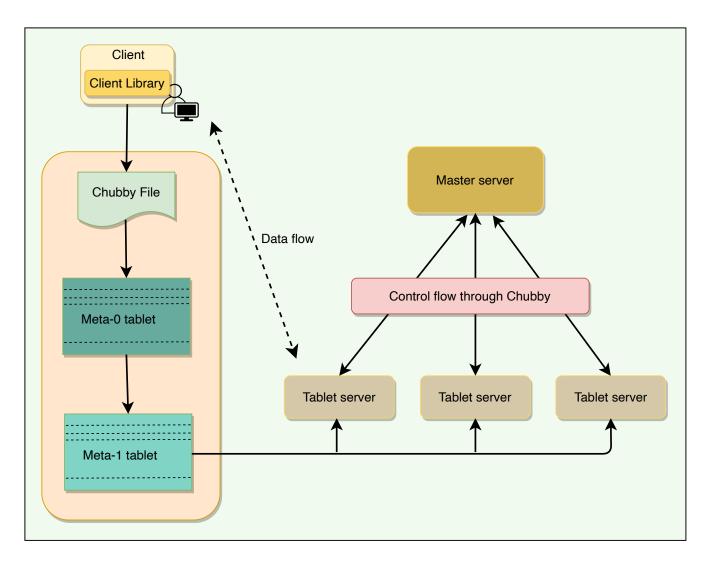
- 1. **Meta-1 Tablet** has one row for each data Tablet (or non-meta Tablet). Since Meta-1 Tablet can be big, it is split into multiple metadata Tablets and distributed to multiple Tablet servers.
- 2. **Meta-0 Tablet** has one row for each Meta-1 Tablet. Meta-0 table never gets split. BigTable stores the location of the Meta-0 Tablet in a Chubby file.



Metadata tablets

A BigTable client seeking the location of a Tablet starts the search by looking up a particular file in Chubby that is known to hold the location of the Meta-0 Tablet. This Meta-0 Tablet contains information about other metadata Tablets, which in turn contain the location of the actual data Tablets. With this scheme, the depth of the tree is limited to three. For efficiency, the client

library caches Tablet locations and also prefetch metadata a other Tablets whenever it reads the METADATA table.



Control and data flow in BigTable

## **Assigning Tablets#**

A Tablet is assigned to only one Tablet server at any time. The master keeps track of the set of live Tablet servers and the mapping of Tablets to Tablet servers. The master also keeps track of any unassigned Tablets and assigns them to Tablet servers with sufficient room.

When a Tablet server starts, it creates and acquires an exclusive lock on a uniquely named file in Chubby's "servers" directory. This mechanism is us

to tell the master that the Tablet server is alive. When the master that the Tablet server is alive. When the master that the Tablet server is alive. When the master that the Tablet server is alive. When the master that the Tablet server is alive. When the master that the Tablet server is alive. When the master that the Tablet server is alive.

- 1. The master grabs a unique master lock in Chubby to prevent multiple master instantiations.
- 2. The master scans the Chubby's "servers" directory to find the live Tablet servers.
- 3. The master communicates with every live Tablet server to discover what Tablets are assigned to each server.
- 4. The master scans the METADATA table to learn the full set of Tablets. Whenever this scan encounters a Tablet that is not already assigned, the master adds the Tablet to the set of unassigned Tablets. Similarly, the master builds a set of unassigned Tablet servers, which are eligible for Tablet assignment. The master uses this information to assign the unassigned Tablets to appropriate Tablet servers.

## Monitoring Tablet servers#

As stated above, BigTable maintains a 'Servers' directory in Chubby, which contains one file for each live Tablet server. Whenever a new Tablet server comes online, it creates a new file in this directory to signal its availability and obtains an exclusive lock on this file. As long as a Tablet server retains the lock on its Chubby file, it is considered alive.

BigTable's master keeps monitoring the 'Servers' directory, and whenever it sees a new file in this directory, it knows that a new Tablet server has become available and is ready to be assigned Tablets. In addition to that, the master regularly checks the status of the lock. If the lock is lost, the master assumes that there is a problem either with the Tablet server or the Chubby. In such a case, the master tries to acquire the lock, and if it succeeds, it concludes that Chubby is working fine, and the Tablet server is having problems. The master, in this case, deletes the file and reassigns the tablets

of the failing Tablet server. The deletion of the file works as failing Tablet server to terminate itself and stop serving the Tablets.

Whenever a Tablet server loses its lock on the file it has created in the "servers" directory, it stops serving its Tablets. It tries to acquire the lock again, and if it succeeds, it considers it a temporary network problem and starts serving the Tablets again. If the file gets deleted, then the Tablet server terminates itself to start afresh.

## Load-balancing Tablet servers#

As described above, the master is responsible for assigning Tablets to Tablet servers. The master keeps track of all available Tablet servers and maintains the list of Tablets that the cluster is supposed to serve. In addition to that, the master periodically asks Tablet servers about their current load. All this information gives the master a global view of the cluster and helps assign and load-balance Tablets.













# The Life of BigTable's Read & Write Operations

Let's explore how BigTable handles its read and write operations.

We'll cover the following

- Read request

Write request

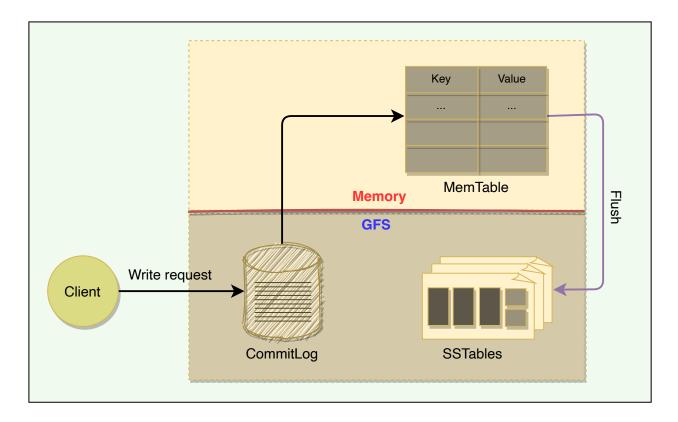
## Write request#

Upon receiving a write request, a Tablet server performs the following set of steps:

- 1. Checks that the request is well-formed.
- 2. Checks that the sender is authorized to perform the mutation. This authorization is performed based on the Access Control Lists (ACLs) that are stored in a chubby file.
- 3. If the above two conditions are met, the mutation is written to the commit-log in GFS that stores redo records.
- 4. Once the mutation is committed to the commit-log, its contents are stored in memory in a sorted buffer called MemTable.
- 5. After inserting the data into the MemTable, acknowledgment is sent to the client that the data has been successfully written.



6. Periodically, MemTables are flushed to SSTables, and SS merged during compaction (discussed later).

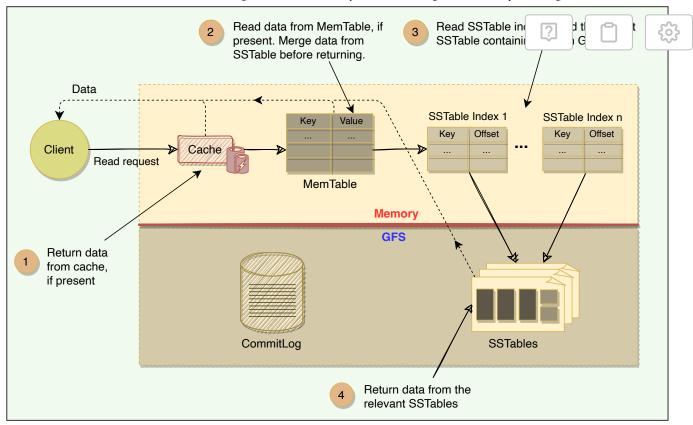


The anatomy of a write request

## Read request#

Upon receiving a read request, a Tablet server performs the following set of steps:

- 1. Checks that the request is well-formed and the sender is authorized
- 2. Returns the rows if they are available in the Cache (discussed later)
- 3. Reads MemTable first to find the required rows
- 4. Reads SSTable indexes that are loaded in memory to find SSTables that will have the required data, then reads the rows from those SSTables
- 5. Merge rows read from MemTable and SSTables to find the required version of the data. Since the SSTables and the MemTable are sorted, the merged view can be formed efficiently.



The anatomy of a read request













# Fault Tolerance and Compaction

Let's learn how BigTable handles fault tolerance and data compaction.

We'll cover the following



- Fault tolerance and replication
  - Fault tolerance in Chubby and GFS
  - Fault tolerance for Tablet server
  - Fault tolerance for the Master
- Compaction

### Fault tolerance and replication#

#### Fault tolerance in Chubby and GFS#

As discussed earlier, BigTable uses two independent systems Chubby and GFS. Both of these systems adopt a replication strategy for fault tolerance and higher availability. For example, a Chubby cell usually consists of five servers, where one server becomes the master and the remaining four work as replicas. In case the master fails, one of the replicas is elected to become the leader; thus, minimizing Chubby's downtime. Similarly, GFS stores multiple copies of data on different ChunkServers.

#### Fault tolerance for Tablet server#



BigTable's master is responsible for monitoring the Tablet semanter does this by periodically checking the status of the Chubby lock against each Tablet server. When the master finds out that a Tablet server has gone dead, it reassigns the tablets of the failing Tablet server.

#### Fault tolerance for the Master#

The master acquires a lock in a Chubby file and maintains a lease. If, at any time, the master's lease expires, it kills itself. When Google's Cluster Management System finds out that there is no active master, it starts one up. The new master has to acquire the lock on the Chubby file before acting as the master.

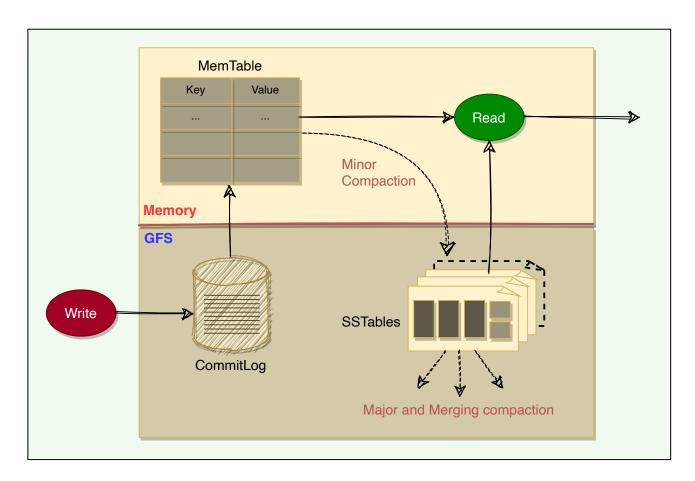
## Compaction#

Mutations in BigTable take up extra space till compaction happens. BigTable manages compaction behind the scenes. Here is the list of compactions:

- 1. **Minor Compaction**: As write operations are performed, the MemTable grows in size. When the MemTable reaches a certain threshold, it is frozen, and a new MemTable is created. The frozen MemTable is converted to an SSTable and written to GFS. This process is called minor compaction. Each minor compaction creates a new SSTable and has the following two benefits:
  - It reduces the memory usage of the Tablet server, as it flushes the MemTable to GFS. Once a MemTable is written to GFS, corresponding entries in the commit-log are also removed.
  - It reduces the amount of data that has to be read from the commit log during recovery if this server dies.
- 2. **Merging Compaction** Minor compaction keeps increasing the coun of SSTables. This means that read operations might need to merge

updates from an arbitrary number of SSTables. To reduce SSTables, a merging compaction is performed which reads the contents of a few SSTables and the MemTable and writes out a new SSTable. The input SSTables and MemTable can be discarded as soon as the compaction has finished.

3. **Major Compaction** — In Major compaction, all the SSTables are written into a single SSTable. SSTables created as a result of major compaction do not contain any deletion information or deleted data, whereas SSTables created from non-major compactions may contain deleted entries. Major compaction allows BigTable to reclaim resources used by deleted data and ensures that deleted data disappears from the system quickly, which is important for services storing sensitive data.



Major, minor, and merging compaction in BigTable





The Life of BigTable's Read & Write O...



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# BigTable Refinements

This lesson will explore different refinements that BigTable implemented.

We'll cover the following



- Locality groups
- Compression
- Caching
- Bloom filters
- Unified commit Log
- Speeding up Tablet recovery

BigTable implemented certain refinements to achieve high performance, availability, and reliability. Here are their details:

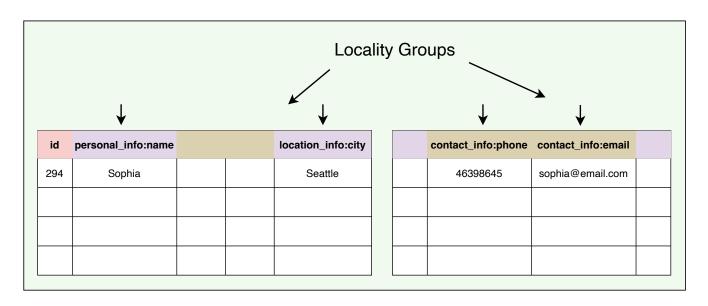
## Locality groups#

Clients can club together multiple column families into a locality group. BigTable generates separate SSTables for each locality group. This has two benefits:

- Grouping columns that are frequently accessed together in a locality group enhances the read performance.
- Clients can explicitly declare any locality group to be in memory for faster access. This way, smaller locality groups that are frequently accessed can be kept in memory.



• Scans over one locality group are  $O(\text{bytes\_in\_locality\_graph})$  a  $O(\text{bytes\_in\_table})$ .



Grouping together columns to form locality groups

## Compression#

Clients can choose to compress the SSTable for a locality group to save space. BigTable allows its clients to choose compression techniques based on their application requirements. The compression ratio gets even better when multiple versions of the same data are stored. Compression is applied to each SSTable block separately.

## Caching#

To improve read performance, Tablet servers employ two levels of caching:

- **Scan Cache** It caches (key, value) pairs returned by the SSTable and is useful for applications that read the same data multiple times.
- Block Cache It caches SSTable blocks read from GFS and is useful for the applications that tend to read the data which is close to the data the

recently read (e.g., sequential or random reads of differ the same locality group within a frequently accessed row).

#### Bloom filters#

Any read operation has to read from all SSTables that make up a Tablet. If these SSTables are not in memory, the read operation may end up doing many disk accesses. To reduce the number of disk accesses BigTable uses Bloom Filters.

Bloom Filters are created for SSTables (particularly for the locality groups). They help to reduce the number of disk accesses by predicting if an SSTable may contain data corresponding to a particular (row, column) pair. Bloom filters take a small amount of memory but can improve the read performance drastically.

## Unified commit Log#

Instead of maintaining separate commit log files for each Tablet, BigTable maintains one log file for a Tablet server. This gives better write performance. Since each write has to go to the commit log, writing to a large number of log files would be slow as it could cause a large number of disk seeks.

One disadvantage of having a single log file is that it complicates the Tablet recovery process. When a Tablet server dies, the Tablets that it served will be moved to other Tablet servers. To recover the state for a Tablet, the new Tablet server needs to reapply the mutations for that Tablet from the commit log written by the original Tablet server. However, the mutations for these Tablets were co-mingled in the same physical log file. One approach would be for each new Tablet server to read this full commit log file and apply just the entries needed for the Tablets it needs to recover. However, under such

scheme, if 100 machines were each assigned a single Tablet f Tablet server, then the log file would be read 100 times. BigTable avoius duplicating log reads by first sorting the commit log entries in order of the keys <table, row name, log sequence number>. In the sorted output, all mutations for a particular Tablet are contiguous and can therefore be read efficiently

To further improve the performance, each Tablet server maintains two log writing threads — each writing to its own and separate log file. Only one of the threads is active at a time. If one of the threads is performing poorly (say, due to network congestion), the writing switches to the other thread. Log entries have sequence numbers to allow the recovery process.

## Speeding up Tablet recovery#

As we saw above, one of the complicated and time-consuming tasks while loading Tablets is to ensure that the Tablet server loads all entries from the commit log. When the master moves a Tablet from one Tablet server to another, the source Tablet server performs compactions to ensure that the destination Tablet server does not have to read the commit log. This is done in three steps:

- In the first step, the source server performs a minor compaction. This compaction reduces the amount of data in the commit log.
- After this, the source Tablet server stops serving the Tablet.
- Finally, the source server performs another (usually very fast) minor compaction to apply any new log entries that have arrived while the first minor compaction was being performed. After this second minor compaction is complete, the Tablet can be loaded on another Tablet server without requiring any recovery of log entries.





Fault Tolerance and Compaction





! Report an Issue











# BigTable Characteristics

This lesson will explore some miscellaneous characteristics of BigTable.

We'll cover the following



- BigTable performance
- Dynamo vs. BigTable
- Datastores developed on the principles of BigTable

## BigTable performance#

Here are a few reasons behind BigTable's performance and popularity:

- **Distributed multi-level map**: BigTable can run on a large number of machines.
- **Scalable** means that BigTable can be easily scaled horizontally by adding more nodes to the cluster without any performance impact. No manual intervention or rebalancing is required. BigTable achieves linear scalability and proven fault tolerance on commodity hardware.
- **Fault-tolerant and reliable**: Since data is replicated to multiple nodes, fault tolerance is pretty high.
- Durable: BigTable stores data permanently.
- **Centralized**: BigTable adopts a single-master approach to maintain data consistency and a centralized view of the state of the system.
- **Separation between control and data**: BigTable maintains a strict separation between control and data flow. Clients talk to the Master for







all metadata operations, whereas all data access happens directly between the Clients and the Tablet servers.

## Dynamo vs. BigTable#

Here is the comparison between Dynamo and BigTable:

	Dynamo	BigTable
Architecture	Decentralized  Every node has same set of responsibilities	Centralize Master handles metadat handle read/
Data Model	Key-value	Multidimensional s
Security	X	Access rights at colun
Partitioning	Consistent Hashing Each node is assigned to a random position on the ring	Tablets Each table is broken int range of rows call
Replication	Sloppy Quorum  Each data item is replicated to 'N'  number of nodes	GFS Chunk rep Data is stored in GFS. F broken into chunks, and are replicated to diffe
CAP	АР	СР
Operations	By key	By key-rar
Storage	Plug-in	SSTables in
Memberships and failure detection	Gossip based protocol	Handshakes initiated

# Datastores developed on the principles of BigTable#









Google's BigTable has inspired many NoSQL systems. Here is a list of a few famous ones:

**HBase**: HBase is an open-source, distributed non-relational database modeled after BigTable. It is built on top of the Hadoop Distributed File System (HDFS).

Hypertable: Similar to HBase, Hypertable is an open-source implementation of BigTable and is written in C++. Unlike BigTable, which uses only one storage layer (i.e., GFS), Hypertable is capable of running on top of any file system (e.g., HDFS, GlusterFS, or the CloudStore). To achieve this, the system has abstracted the interface to the file system by sending all data requests through a Distributed File System broker process.

Cassandra: Cassandra is a distributed, decentralized, and highly available NoSQL database. Its architecture is based on Dynamo and BigTable. Cassandra can be described as a BigTable-like datastore running on a Dynamo-like infrastructure. Cassandra is also a wide-column store and utilizes the storage model of BigTable, i.e., SSTables and MemTables.





















# Summary: BigTable

Here is a quick summary of BigTable for you!

We'll cover the following



- Summary
- References & further reading

## Summary#

- BigTable is Google's **distributed storage system** designed to manage large amounts of structured data with high availability, low latency, scalability, and fault-tolerance goals.
- BigTable is a sparse, distributed, persistent, multidimensional sorted map.
- The map is indexed by a unique key made up of a row key, a column key, and a timestamp (a 64-bit integer, "real time" in millisecond).
- Each row key is an arbitrary string of up to 64 kilobytes in size, although most keys are significantly smaller.
- Unlike a traditional relational database table, BigTable is a widecolumn datastore with an unbounded number of columns.
- Columns are grouped into **column families**. Each column family stores similar types of data under a 'family:qualifier' column key.
- The row key and the column (family:qualifier) key uniquely identify a data cell. Within each cell, the data contents are further indexed by timestamps providing multiple versions of the data in time.







- Every read or write of data under a single row is atomic. Atomicity across rows is not guaranteed.
- BigTable provides APIs for metadata operations like creating and deleting tables and column families. BigTable clients can use data operation APIs for writing or deleting values, lookup values from individual rows, or iterate over a subset of the data in a table.
- A table is split into smaller ranges of rows called **Tablets**. A Tablet holds a contiguous range of rows.
- Tablets are the unit of distribution and load balancing.
- BigTable architecture consists of one master server and multiple
   Tablet servers.
- Master is responsible for assigning Tablets to Tablet servers, as well as monitoring and balancing Tablet servers' load.
- Each Tablet server serves read and write requests of the data to the Tablets it is assigned.
- BigTable clients communicate directly with the Tablet servers to read/write data.
- Each Tablet server stores the data in immutable **SSTable** files which are stored in **Google File System (GFS)**.
- New committed updates are first stored in a memory-based **MemTable**.
- BigTable performs all read operations against a combined view of SSTables and MemTable.
- Periodically, the MemTable is flushed into an SSTable, allowing for efficient memory utilization.
- To enhance read performance, BigTable makes use of caching and Bloom filters.
- BigTable relies heavily on distributed locking service **Chubby** for masserver selection and monitoring.







# References & further reading#

- BigTable
- SSTable
- Dynamo
- Cassandra
- HBase



BigTable Characteristics



Quiz: BigTable



✓ Completed



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