An overview of Google F1 (with an emphasis on schema change)

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Talk outline

Introduction and background



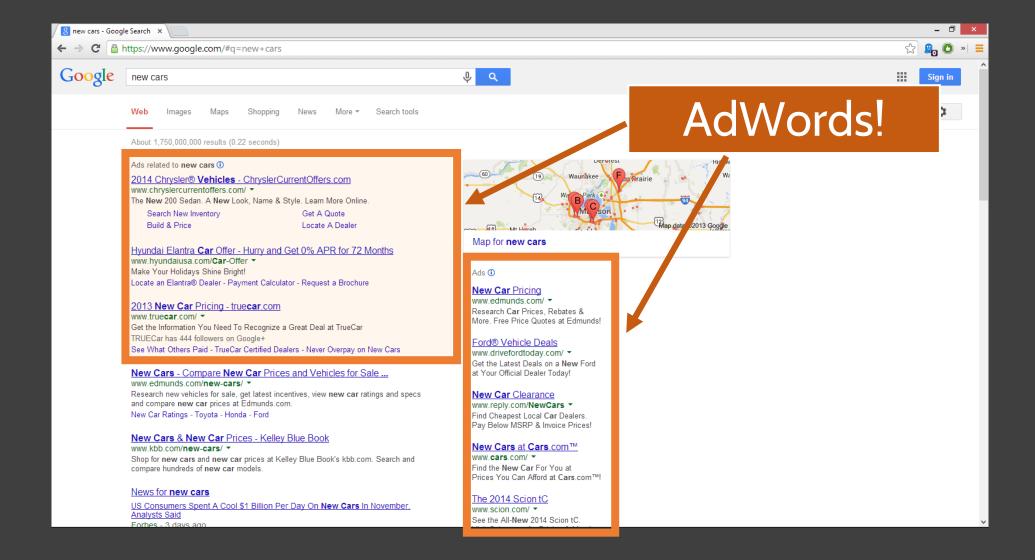
Design overview

Schema changes

Conclusion

Introduction and background

Where Google makes its cash



AdWords overview

~97% of Google's revenue is from advertising!

Need to track lots of info:

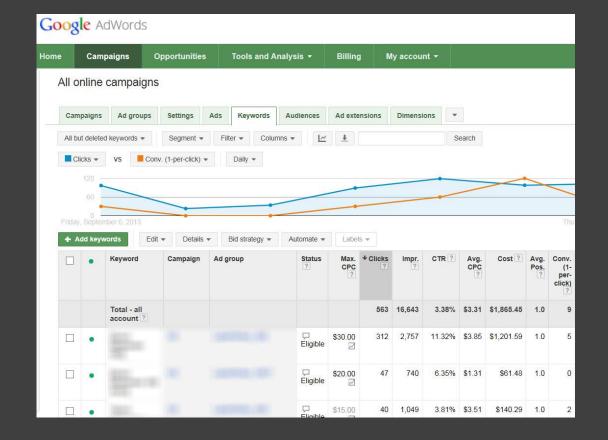
Customer information

Ad campaign preferences

Displayed ads

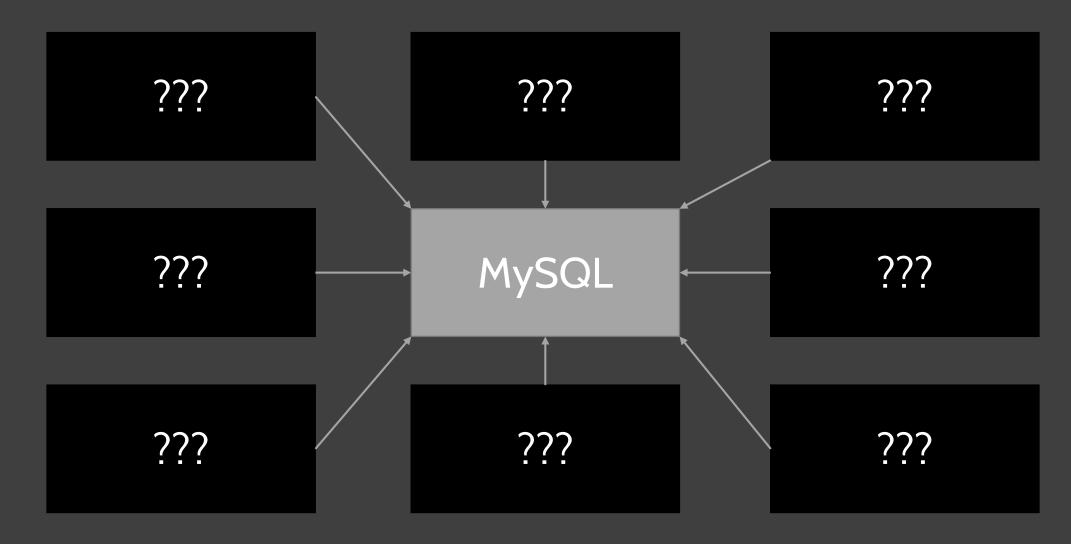
Clicked ads

Follow-through purchases



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AdWords technology ecosystem



The homegrown parallel RDBMS blues

Data partitioned across dozens of MySQL instances.

Have to manually repartition to add servers.

Developers make assumptions about where data lives.

Limited cross-machine transactions.



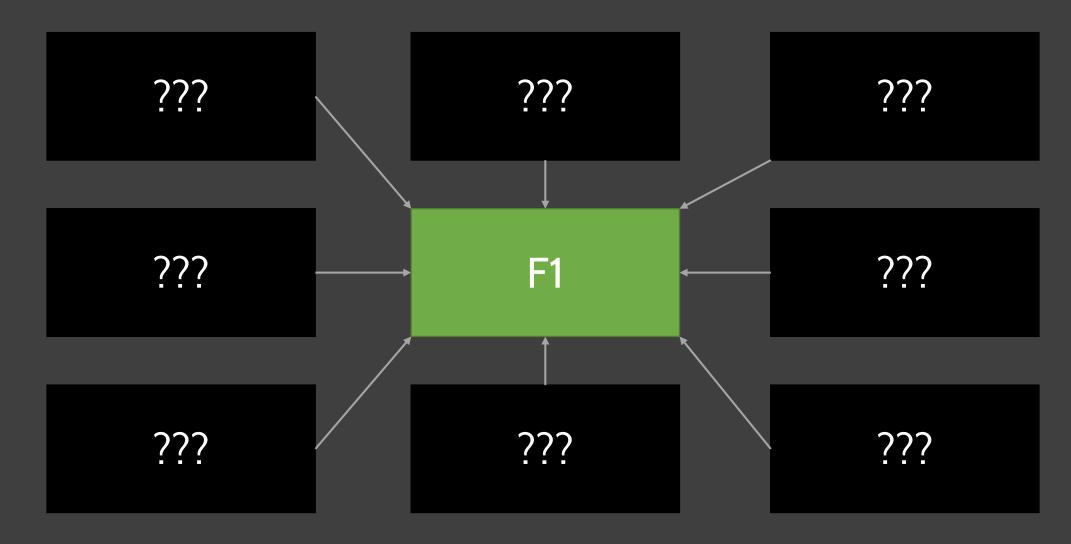
Revenue paranoia



Data synchronously replicated across multiple machines.

Can handle machine failure; what about datacenter failure?

AdWords technology ecosystem



F1 design overview

What is F1?



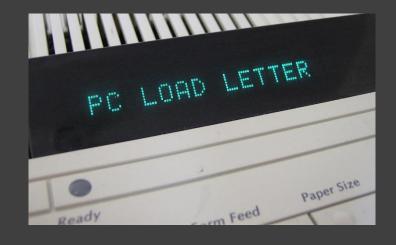
F1 is a distributed, relational database designed for both OLAP and OLTP.

Full SQL support with ACID semantics for transactions.

Shute, J., Vingralek, R., Samwel, B., et al. (2013). "F1: A Distributed Database That Scales," *VLDB*, 6(11).

Two main design goals

Fault tolerance

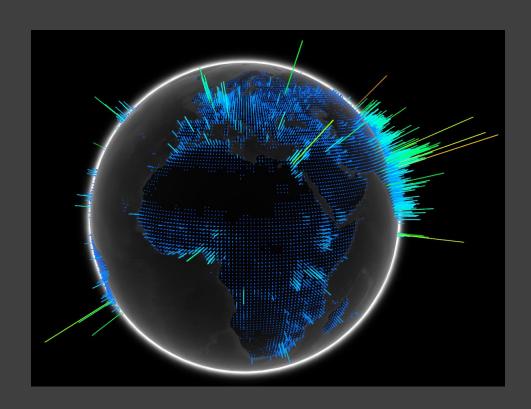




Scalability

Fault tolerance

F1 is globally distributed



A single F1 instance consists of thousands of servers in datacenters across the globe.

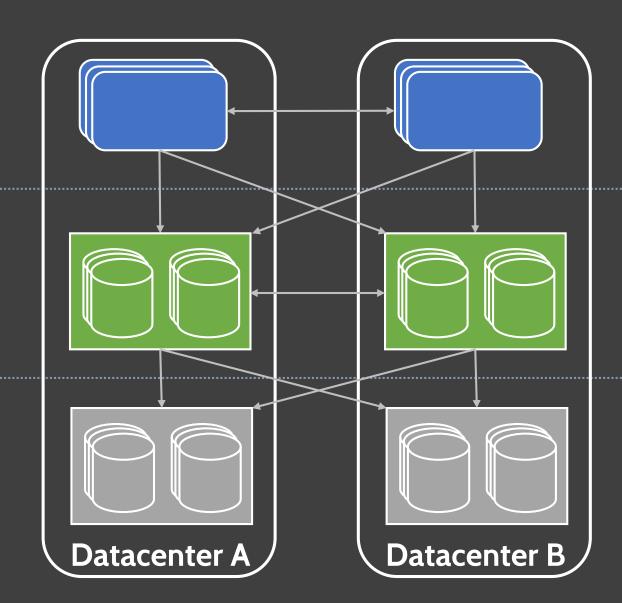
Data is synchronously replicated across datacenters.

F1 architecture

F1 servers (query processing)

Spanner (cross-datacenter storage)

Colossus (distributed filesystem)



Spanner: next-generation BigTable



Spanner does the storage-related "heavy lifting" for F1.

Spanner uses Paxos and 2PC to synchronously replicate data across datacenters.

Corbett, J. C., Dean, J., Epstein, M., et al. (2012). "Spanner: Google's Globally-Distributed Database," *OSDI*.

More Spanner features



Spanner supports strict twophase locking for pessimistic transactions.

Spanner provides guaranteed unique commit timestamps for transactions.

F1 and Spanner

F1 uses Spanner mostly as a key-value store:

Get(key prefix)

Put(key, value)

Delete(key)

Spanner pessimistic transactions are used to enable atomic test-and-set of multiple values.

Scalability

Stateless servers

All data is shared among all servers.

Servers can be added or removed with no data movement.

Clients can send a request to **any server**, even different requests that are **part of the same transaction**.

Transactions in F1

Use a form of optimistic concurrency control, with all state stored on the client (not F1 server).

Limited to one atomic write operation (implicitly commits).

Spanner **pessimistic** transactions also supported, but **not stateless**.

Optimistic lock columns



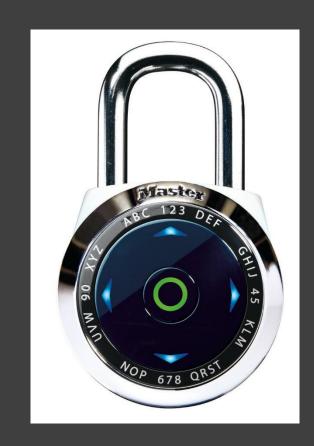
Every column is **covered by** a hidden **optimistic lock column** containing a last-modified timestamp.

When a column is updated, the **commit timestamp** of the updating transaction is **stored in its covering lock column**.

Configurable locking granularity

Users can specify which lock covers a column.

By default, all columns in a row are covered by a **default lock**.



Optimistic transactions: reads

Lock	Buffered
name	timestamp
Lock1	ts ₁
Lock7	ts ₂
Lock3	ts ₃

When an optimistic transaction reads a column value, it also reads the corresponding lock timestamp.

Lock timestamps for all reads are buffered on the client for the duration of the transaction.

Optimistic transactions: write + commit

Lock	Buffered	Current
name	timestamp	timestamp
Lock1	ts ₁	ts ₁
Lock7	ts ₂	ts ₄
Lock3	ts ₃	ts ₃

At commit, all buffered timestamps are validated against the lock timestamps currently in the database.

If there is a mismatch, the transaction aborts.

Optimistic transaction example

T1: read (Age) -> get value 26, read lock1 and get *ts*₁

Name	Age	Lock1
John Doe	27←	ts ₂

T2: write (Age) -> set value = 27, lock1 is updated to ts_2

T1: commit -> validate lock1 (*ts₁!= ts₂*), abort

Schema changes

Schemas in F1

F1 servers use a schema to interpret key-value pairs as rows and to translate relational operations into key-value operations.

Why is schema change in F1 important?



Data in F1 is **critical** to Google's business.

Any downtime or corruption is measured in dollars!

The AdWords F1 instance is **shared** by many teams with **hundreds of developers**.

Schema changes requested daily.

Why is schema change in F1 hard?

Every F1 server has a local cached copy of the schema.

To change the schema, we need to update all the caches, but synchronizing across all F1 servers is slow.

Until the change finishes, no operations can execute -> no money!

The goal for schema changes in F1

Enable changes to the logical and physical schema of an F1 instance in a way that is **online** and **asynchronous**.

Online

All data accessible, no downtime, and without large delays for transactions.

Asynchronous

Different servers transition to a new schema at different times.

A paper is available

A protocol for online, asynchronous schema change that permits no database corruption.

A formal model for reasoning about and proving the correctness of our protocol.

Rae, I., Rollins, E., Shute, J., et al. (2013). "Online, Asynchronous Schema Change in F1," *VLDB*, 6(11).

Some terminology

Schema elements

Any part of the schema, e.g., tables, columns, constraints, etc.

Structural elements

Tables

Columns

Indexes

Locks

Schema elements

Structural elements

User-visible states of schema elements

Absent

Doesn't exist!

Public

Available for **all** operations.



Ensuring correctness



Use intermediate states that restrict allowed operations on an element.

Decompose incompatible schema changes into a series of changes that are pair-wise compatible.

An illustration

All servers on All servers on old schema. new schema. **Both** schemas

in use!

Supported schema changes

Add/drop structural elements

Table add + drop Column add + drop Index add + drop Lock add + drop

Concurrency control

Change lock coverage

Add/drop constraints

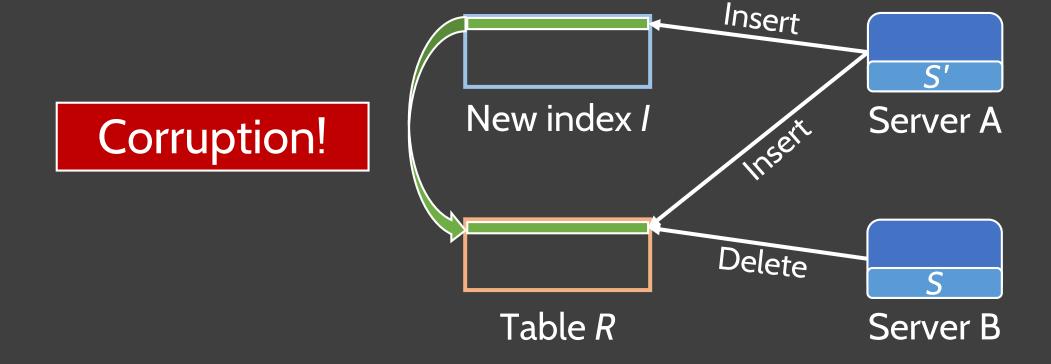
Change column type
Make column unique/non-unique
Foreign key add + drop
Make column required/optional
Change protocol buffer definition

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Adding and dropping structural elements

Index add corruption

Change from schema S to S', adding index I on table R.



Intermediate states for structural elements

Delete only

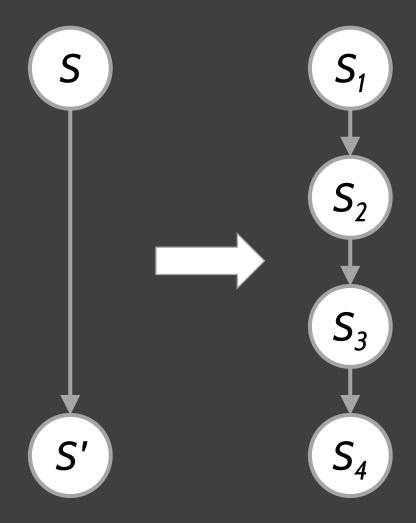
Updated by **delete** operations; cannot be read.

Write only

Updated by delete and insert operations; cannot be read.

Index add revisited

Change from schema S_1 to S_4 , adding index I on table R.



Index add: absent to delete only

S₁ Index I doesn't exist.

Index *I* exists, updated only by deletes.

Index *I* is **not used** for reads.

Index is always empty, but unused.



Index add: delete only to write only

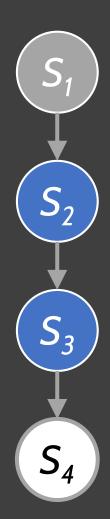
Index *I* exists, updated only by deletes.

Index *I* is **not used** for reads.

Index *I* exists, updated by deletes & inserts.

Index *I* is not used for reads.

All servers delete entries, so no dangling entries are possible.



Index add: backfill



Index *I* exists, updated by **deletes & inserts**. Index *I* is **not used** for reads.

A MapReduce starts to backfill index I.

All servers maintain index for new rows.



Index add: write only to public



Index I is completely backfilled.

Index I is public and ready to use.



Adding and dropping constraints

Constraint corruption

Change from schema S to S', making column C unique.

Problem: servers on schema *S* can insert duplicates into column *C* that servers on schema *S'* don't expect!

Bonus problem: how do we verify that column C only contains unique values?

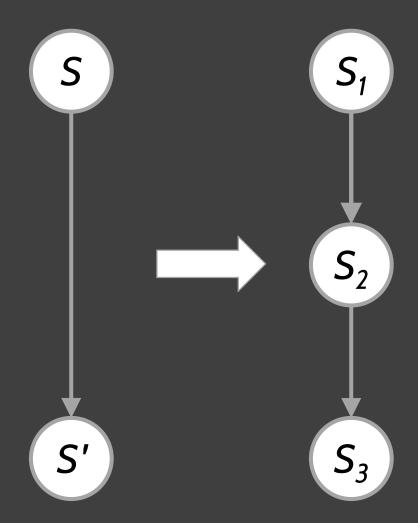
Intermediate states for constraints

Write only

Constraint applies to inserts and updates, but is not guaranteed to hold for reads.

Constraint add revisited

Change from schema S_1 to S_3 , making column C unique.



Constraint add: absent to write only

Column *C* is **not unique**.

Column *C* cannot have duplicates **inserted**.

Reads **may show duplicates**.



Constraint add: verification



Column *C* cannot have duplicates inserted. Reads may show duplicates.



A MapReduce starts to **verify** that column *C* contains only unique values.

No server allows new duplicates to be inserted.



Constraint add: write only to public



Column C is verified unique.



Column *C* is unique for reads and writes.



Concurrency control

Concurrency corruption

Change from schema S to S', changing the lock coverage of column C from L_1 to L_2 .

Problem: servers on schema *S* don't validate writes to column *C* by servers on schema *S'*!

Concurrency corruption example

T1: read(C) using $S \rightarrow \text{read } ts_1 \text{ from } L_1$

T2: write(C) using $S' \rightarrow \text{update } L_2 \text{ to } ts_2$

T1: write(C) using $S \rightarrow validate ts_1 against <math>L_1$ (works!)

Corruption!

Intermediate states for lock coverage

Dual coverage

A column is covered by two locks.

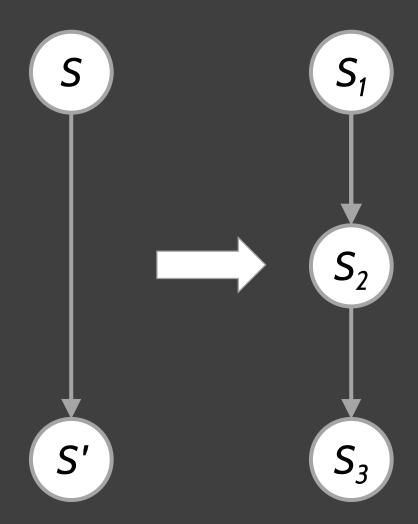
Dual coverage semantics

On a read, the timestamp returned is the maximum of both locks.

On a write, the timestamp is validated against both locks.

Lock coverage change revisited

Change from schema S_1 to S_3 , changing lock coverage of column C from L_1 to L_2 .



Coverage change: L_1 to dual coverage

Column C is covered by L_1 .

Column C is covered by L_1 and L_2 .

 L_1 handles concurrency control.



Coverage change: dual coverage to L_2

Column C is covered by L_1 and L_2 .

Column C is covered by L_2 .

 L_2 handles concurrency control.



More concurrency corruption

Suppose both L_1 and L_2 have the same timestamp.

T1: read(C) using $S_1 \rightarrow \text{read } ts_1 \text{ from } L_1$

T2: write(C) using $S_1 \rightarrow \text{update } L_1 \text{ to } ts_2$

T1: write(C) using $S_3 \rightarrow$ validate ts_1 against L_2 (works!)

Coverage change: propagation



Column C is covered by L_1 and L_2 .

A MapReduce sets $L_2 = \max(L_1, L_2)$.

Timestamps propagate from L_1 to L_2 .



Some implementation details

Schema leases

Canonical schema file is stored in Spanner.

Once per lease period, F1 servers reload the canonical schema if needed.

If a server cannot read the schema, it terminates and restarts.



Batching



Modifications to the schema are first committed to source control, not a live F1 instance.

Schema change process periodically applies modifications present in source control as a batch.

More details in the paper!

Paper has a lot of other stuff
Formal model and proofs
Concurrency control schema changes
Details on overlapping state transitions
Discussion of MapReduces needed
More implementation details

. . .



Conclusion

F1 is a globally distributed, fault-tolerant relational database that serves as the main data store for Google AdWords.

Check out the papers for more details:

Shute, J., Vingralek, R., Samwel, B., et al. (2013). "F1: A Distributed Database That Scales," *VLDB*, 6(11).

Rae, I., Rollins, E., Shute, J., et al. (2013). "Online, Asynchronous Schema Change in F1," *VLDB*, 6(11).