Percolator

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Outline

- Motivation
 - Google's Search Algorithm
 - Requirements
- 2 Design
 - Structure
 - Algorithm
 - Details and Optimizations
- Second Second



Outline

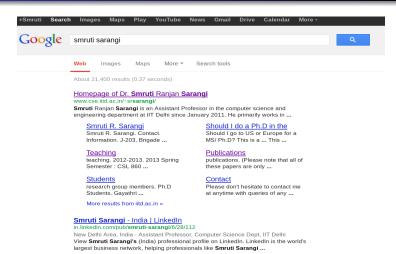
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- Updating Google's web index continually is a major challenge.
 - Tens of petabytes of data
 - Billions of updates per day
 - Thousands of machines.
 - Cascading updates.

Google's Search Algorithm

- Every page has a "page rank".
- The page rank of a popular page is supposed to be high.
- The page rank of a page is determined by the page rank of all the pages that link to it.
- For example:
 - If the New York Times website points to some link, then it has a high page rank.
 - If Hauz Khas Times points to some website, it will have a very low page rank.

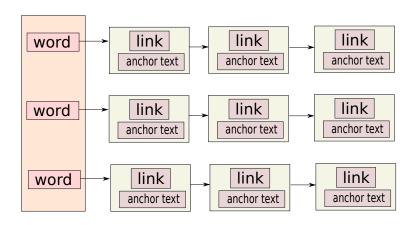
Example of a Google Search Query



Smruti Sarangi profiles | LinkedIn www.linkedin.com/pub/dir/Smruti/Sarangi

View the profiles of professionals named **Smruti Sarangi** on LinkedIn. There are 5 professionals named **Smruti Sarangi**, who use LinkedIn to exchange ...

Structure of a Web Index



The Problem of Updates

- The links in the inverted list are arranged according to their page rank.
- If the page rank of a website changes then:
 - We need to update the inverted list to reflect the change.
 - The page rank of sites that it points to need to change.
 - This problem is known as cascading update.

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Requirements of a Solution

- Should provide ACID transaction semantics (do not want to corrupt database).
- Should have high throughput, and acceptable latency.
- Should be able to handle petabytes of data.
- Traditional DBMS systems are too slow → Need new technology
- Random access to data such that changes can percolate
- Consistency Model: Snapshot Isolation

Snapshot Isolation

- Assume two concurrent updates to a linked list.
 - If they do not access the same node or its parent, then they are disjoint.
 - Disjoint accesses can continue in parallel.
 - This is different from regular transaction semantics such as serializability.

Definition :

- When a transaction starts, it takes(appears to) a consistent snapshot of the entire database.
- It then proceeds to update its private copy of the database.
- The values are committed if they have not been changed by another transaction since the snapshot.



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Design of Percolator

- Built on top of Bigtable Google's distributed storage engine
- Bigtable is a multidimensional database
 - Distributed key-value store
 - We save row, column, timestamp
 - Atomic read-modify-write operations for each row
 - Meta data is stored in separate columns
- Observer framework
 - Any row has a set of observers.
 - They run specialized functions when data in the row changes.



Model of Transactions

- Provides support for ACID transactions
 - Hard to do in such a large database
 - Required: do not want to have Google's database in an inconsistent state
 - Uses timestamp for each data item
 - The set of timestamps at the beginning of a transaction is its snapshot.
- Transactions can include multiple rows across multiple BigTable tables
- Percolator implements its own lock service
- Percolator adds a special column to save locks.



Columns in BigTable

Column	Use		
lock	contains a pointer to the lock		
write	timestamp of committed data		
data	data value		
notify	list of observers		
ack_O	last timestamp at which observer O ran		

Example

A transfers B 7\$

key	data	lock	write
Α	6:	6:	6:data@5
	5:10\$	5:	5:
В	6:	6:	6:data@5
	5:2\$	5:	5:

key	data	lock	write
	7: 3\$	7: primary	7:
Α	6:	6:	6:data@5
	5:10\$	5:	5:
В	6:	6:	6:data@5
Ь	5:2\$	5:	5:



Example - II

key	data	lock	write
	7:3\$	7: primary	7:
Α	6:	6:	6:data@5
	5:10\$	5:	5:
В	7: 9\$	7: primary@A	7:
	6:	6:	6:data@5
	5:2\$	5:	5:

key	data	lock	write
A	8:	8:	8: data @ 7
	7:3\$	7:	7:
	6:	6:	6:data@5
	5:10\$	5:	5:
В	7:9\$	7: primary@A	7:
	6:	6:	6:data@5
	5:2\$	5:	5:



Example - III

key	data	lock	write
	8:	8:	8: data @ 7
	7:3\$	7:	7:
Α	6:	6:	6:data@5
	5:10\$	5:	5:
	8:	8:	8: data @ 7
В	7:9\$	7:	7:
В	6:	6:	6:data@5
	5:2\$	5:	5:

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Algorithm: Begin Transaction

```
Algorithm 1: Begin Transaction 
startTs ← oracle.getTimeStamp()
```

2 Set(W):
writes.push(W)

Get Method

```
Get(row, column, value):
  while True do
      T \leftarrow \text{startTrans(row)}
2
      if T.hasLock(0,startTs) then
         backOffAndMaybeRemoveLock(row,col)
3
         continue
      end
4
      latestWrite ← T.read(row, [0,startTs])
5
      if !latestWrite then
         return \phi
6
      end
7
      dataTs ← latestWrite.timeStamp
8
      return (T.read(row, "data", dataTs)
 end
```

PreWrite

```
1 PreWrite(Write w, Write primary)
  Column col \leftarrow w col
  T \leftarrow \text{startTransaction}(w.row)
 if T.read(w.row, "write", [startTs, \infty]) then
      return false
4 end
 if T.read(w.row, "lock", [0, \infty]) then
      return false
6
 end
  T.write (w.row, "data", startTs, w.value)
  T.write (w.row, "lock", startTs, {primary.row, primary.col})
  return T.commit()
```

Commit - I

```
Commit()
       /* Prewrite all the entries
2 (primary, secondaries) \leftarrow (writes[0], writes[1...n])
  if !PreWrite(primary,primary) then
      return false
3
4 end
  for Write w: secondaries do
      if !PreWrite(w,primary) then
6
         return false
      end
8
  end
10 | commitTs ← oracle.getTimeStamp()
```

Commit - II

```
/ \star Commit the primary
11 T \leftarrow \text{startTransaction}(primary.row)
   /\!/\star Test to see if aborted by somebody else
12 if !T.read(primary.row, "lock", startTs) then
     return false
14 end
  /\!/ Write the primary and erase the lock
15 T.write(primary.row, "write", commitTs, "data@"+startTs)
   T.erase (primary.row, "lock", commitTs)
   /* Point of commit
16 if !T.Commit() then
      return false
18 end
```

Commit - III

```
for Write w: secondaries do
write(w.row, "write", commitTs, "data@"+startTs)
erase (w.row, "lock", commitTs)
end
return true
```

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Timestamps

- The timestamp oracle needs to be able to sustain very high throughput.
- Possible to batch several RPC calls to the oracle to reduce network load.
- Needs to give out timestamps in increasing order.
- If it fails, then it needs to recover and issue timestamps that are greater than the ones it issued earlier.

Observers

- Each observer registers a set of columns, and a function.
- The function gets invoked, if any of the columns are updated.
- Possible to do message collapsing
- At most one observer's transaction will commit per column.
- Steps in running an observer
 - After an update to a column, Percolator sets the notify column.
 - A worker thread, ultimately picks up this information, and runs an observer.
 - If the latest timestamp of an observer run (ack_O) is less than the commit timestamp of the update, then run the observer.
 - Worker threads avoid clumping by scanning random parts of the database.

Performance Improvements

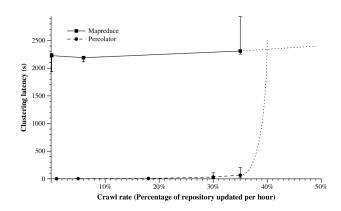
- Support for read-modify-write RPCs in BigTable.
- Create batches of RPC calls.
- Employ pre-fetching to reduce reads.
- Use blocking API calls, and a large number of threads to simplify the programming model.

Setup

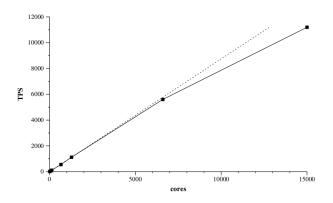
- Existing Setup:
 - Crawl billions of documents
 - Series of 100 map-reduces
 - document takes 2-3 days for getting indexed
- Percolator based indexing system Caffeine
 - 100x faster
 - Average age of documents gets reduced to 50%



Performance vs Crawl Rate



Scalability for TPC/E benchmarks



Close to Linear Scaling





Large-scale Incremental Processing Using Distributed Transactions and Notifications by Daniel Peng and Frank Dabek, OSDI, 2010