## Dremel: Interactive Analysis of Web-Scale Datasets

Sergey Melnik, Andrey Gubarev, Jing Jing Long, Geoffrey Romer, Shiva Shivakumar, Matt Tolton, Theo Vassilakis Google, Inc.

Presenter: Rui Wang

## Background

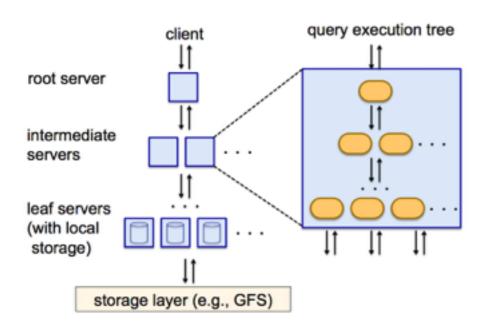
**Analytic Service** 

Parellel Processing

Distributed File System

# Is MapReduce enough for Large Scale Data Analysis?

## Dremel

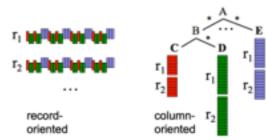


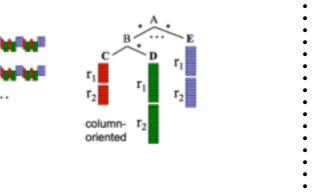
- Nested Column Store
- Serving Tree Engine
- Diverse Measures

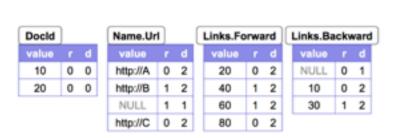
#### Data Model

## Query

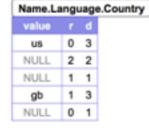
## Experiments

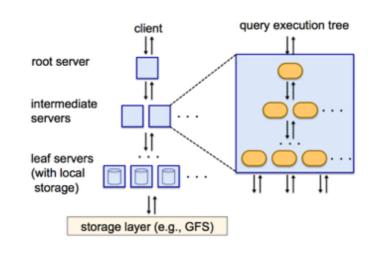






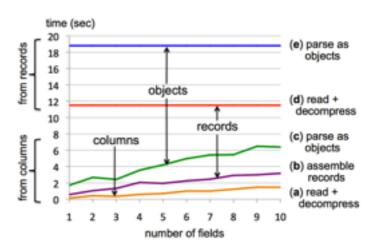
Name.La	ıngı	ıagı	e.Co
value			
en-us	0	2	
en	2	2	
NULL	1	1	
en-gb	1	2	
NULL	0	1	

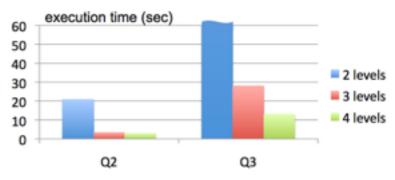


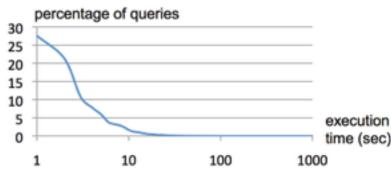


SELECT A, COUNT(B) FROM T GROUP BY A

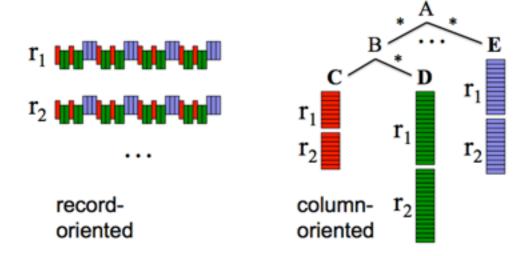
SELECT A, SUM(c) FROM ( $R_1$  UNION ALL ...  $R_n^{-1}$ ) **GROUP BY A** 







## Nested Columnar Storage



- Structured Format
- Efficient Encoding
- Fast Transform

```
\mathbf{r_1}
DocId: 10
Links
  Forward: 20
  Forward: 40
  Forward: 60
Name
  Language
    Code: 'en-us'
    Country: 'us'
  Language
    Code: 'en'
  Url: 'http://A'
Name
  Url: 'http://B'
Name
  Language
    Code: 'en-gb'
    Country: 'gb'
```

DocId: 20 Links	r <sub>2</sub>
Backward:	10
Backward:	30
Forward:	80
Name	
Url: 'http	p://C'

Docld		
value	r	d
10	0	0
20	0	0

Name.Ur	1]	
value	r	d
http://A	0	2
http://B	1	2
NULL	1	1
http://C	0	2

L	Links.Forward			
	value	r	d	
	20	0	2	
	40	1	2	
	60	1	2	
	80	0	2	

Links.Backward			L
value	r	d	
NULL	0	1	
10	0	2	
30	1	2	

Name.Language.Code			
value	r	d	
en-us	0	2	
en	2	2	
NULL	1	1	
en-gb	1	2	
NULL	0	1	

Name.Language.Country			
value	r	d	
us	0	3	
NULL	2	2	
NULL	1	1	
gb	1	3	
NULL	0	1	

DocId: 10 r <sub>1</sub>
Links
Forward: 20
Forward: 40
Forward: 60
Name
Language
Code: 'en-us'
Country: 'us'
Language
Code: 'en'
Url: 'http://A'
Name
Url: 'http://B'
Name
Language
Code: 'en-gb'
Country: 'gb'

DocId: 20	r,
Links	- 2
Backward:	10
Backward:	30
Forward:	80
Name	
Url: 'http	p://C'

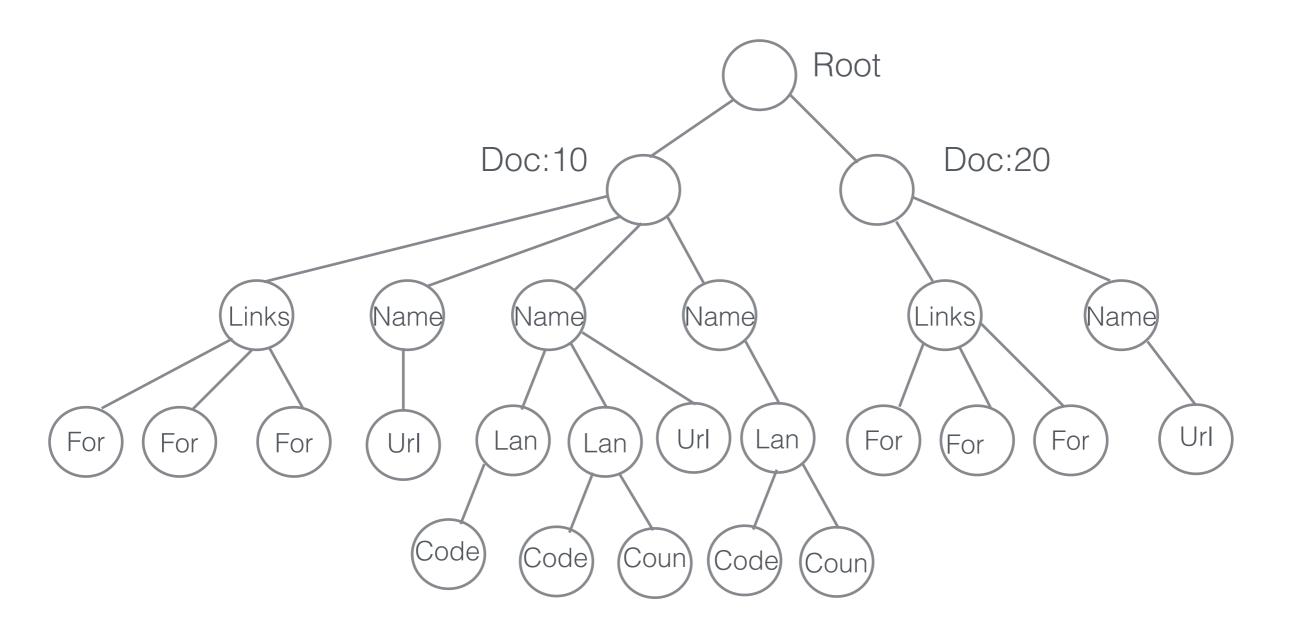
2 Language	3 Code	Repetition
1	en-us	0
2	en	2
0	NULL	NULL
1	en-gb	1
0	0	NULL
	Language  1 2 0	Language Code  1 en-us  2 en  NULL

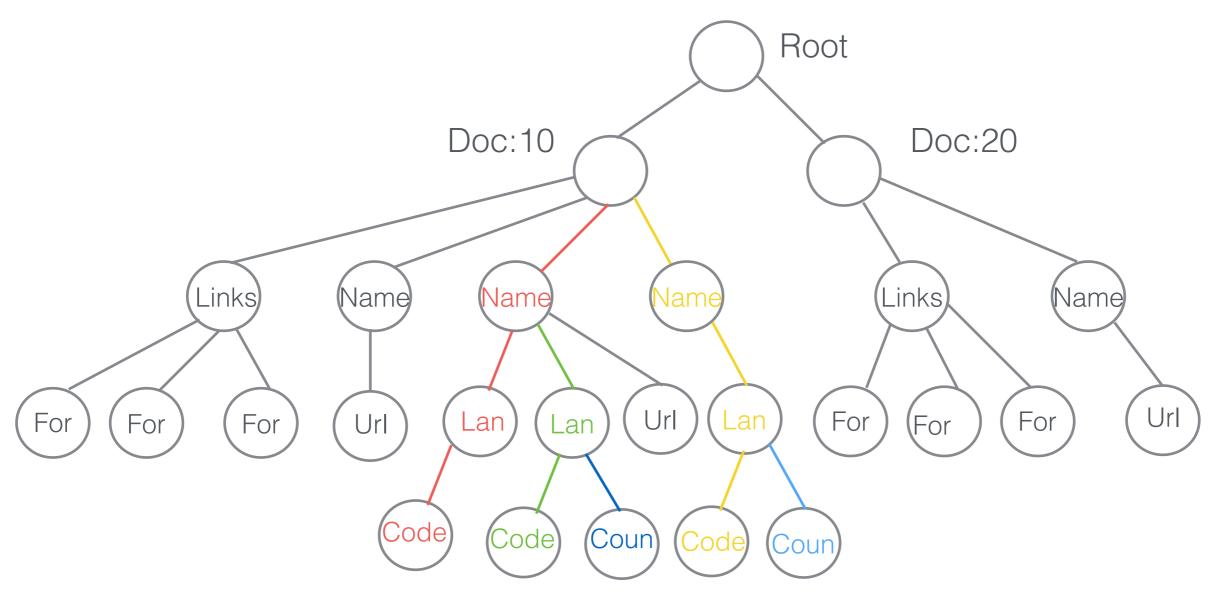
DocId: 10 r <sub>1</sub>
Links
Forward: 20
Forward: 40
Forward: 60
Name
Language
Code: 'en-us'
Country: 'us'
Language
Code: 'en'
Url: 'http://A'
Name
Url: 'http://B'
Name
Language
Code: 'en-gb'
Country: 'gb'

DOCTO: 10 11	
Links	
Forward: 20	
Forward: 40	
Forward: 60	
Name	
Language	
Code: 'en-us'	
Country: 'us'	
Language	
Code: 'en'	
Url: 'http://A'	
Name	
Url: 'http://B'	
Name	
Language	
Code: 'en-gb'	
Country: 'gb'	
ocId: 20 r <sub>2</sub>	1
inks	

DocId: 20	$\mathbf{r_2}$
Links	-
Backward: 10	
Backward: 30	
Forward: 80	
Name	
Url: 'http://	'C'

1/	2 / Cd	ode 3	
Name	Language	Country	Definition
1	1	US	3
1	2	NULL	2
2	1	NULL	1
3	1	gb	3
1	1	NULL	1





Repetition: 0 2 1

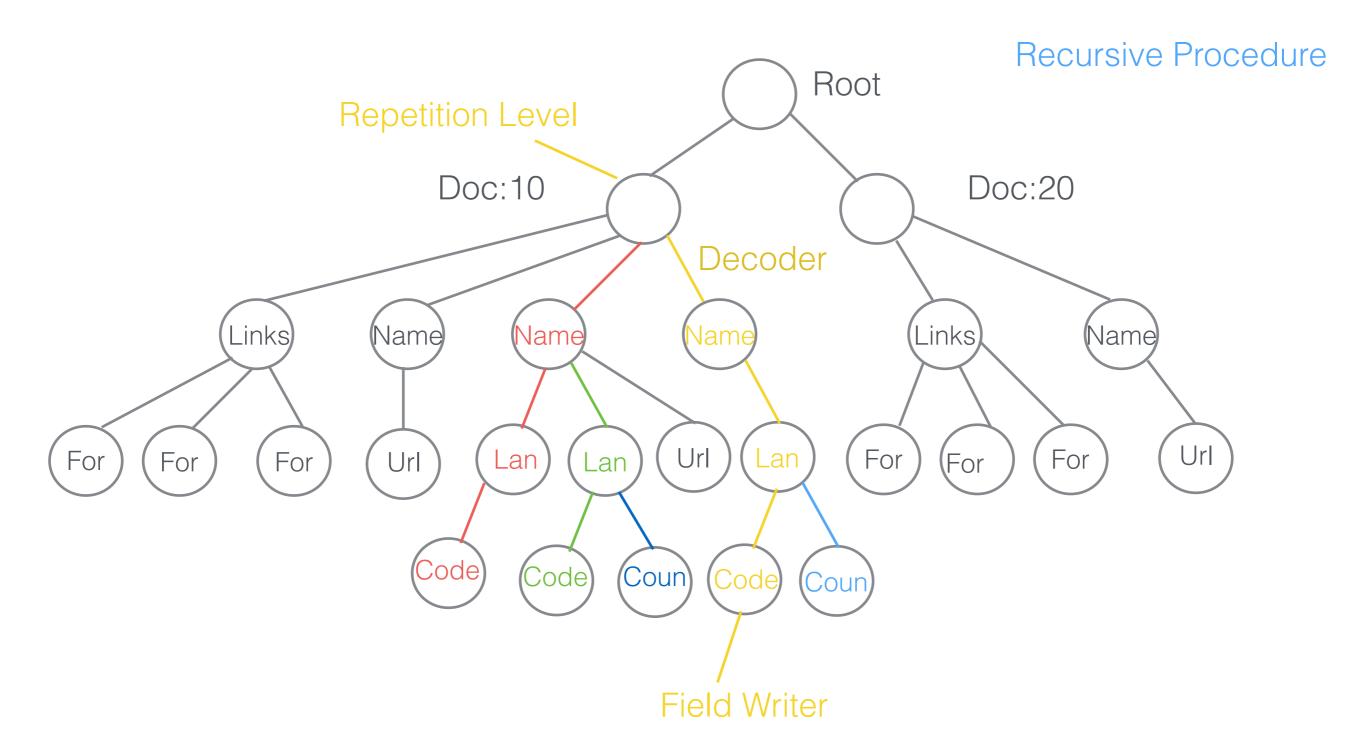
Definition: 2 2 2

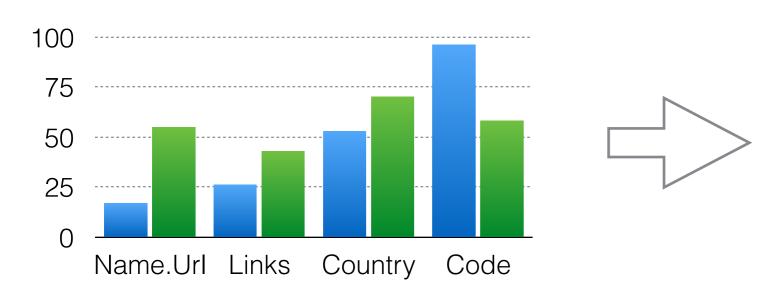
```
message Document {
  required int64 DocId;
  optional group Links {
    repeated int64 Backward;
    repeated int64 Forward; }
  repeated group Name {
    repeated group Language {
    required string Code;
    optional string Country; }
  optional string Url; }}

  Name.Url Links Country Code
```

Protocol Buffer

How to Serialize to Columns?



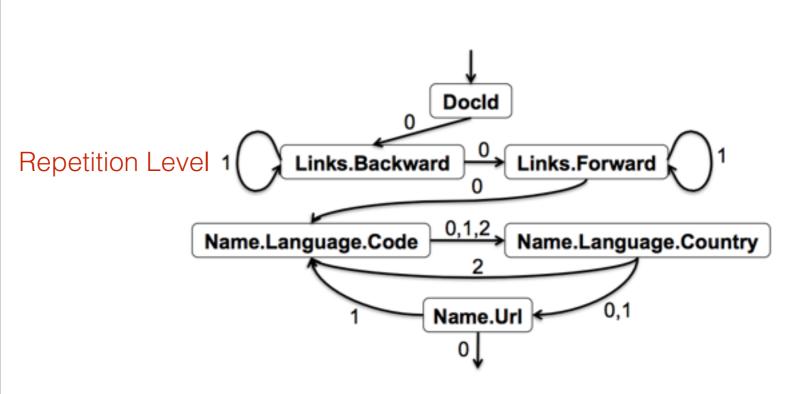


```
message Document {
   required int64 DocId;
   optional group Links {
      repeated int64 Backward;
      repeated int64 Forward; }
   repeated group Name {
      repeated group Language {
        required string Code;
        optional string Country; }
      optional string Url; }}
```

Columnar Storage

Protocol Buffer

```
\mathbf{r_1}
DocId: 10
Links
  Forward: 20
  Forward: 40
  Forward: 60
Name
  Language
    Code: 'en-us'
    Country: 'us'
  Language
    Code: 'en'
  Url: 'http://A'
Name
  Url: 'http://B'
Name
  Language
    Code: 'en-gb'
    Country: 'gb'
```

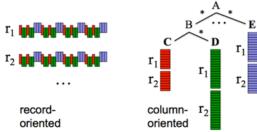


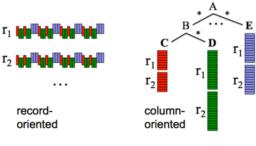
Finite State Machine

## Data Model

## Query



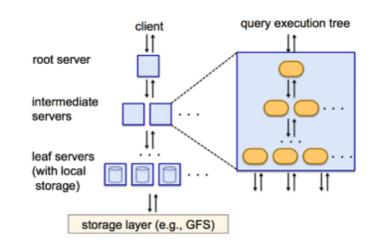




Docld Name.Url Lin				Links.Forward				Links.Backward						
value		d		value	r	d		value	r	d		value		d
10	0	0		http://A	0	2		20	0	2		NULL	0	1
20	0	0		http://B	1	2		40	1	2		10	0	2
				NULL	1	1		60	1	2		30	1	2
				http://C	0	2		80	0	2				

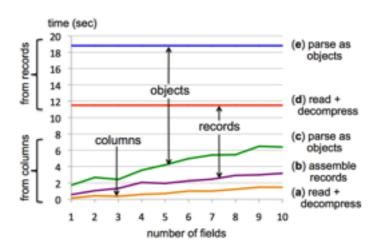
Name.Language.Code							
value	r	d					
en-us	0	2					
en	2	2					
NULL	1	1					
en-gb	1	2					
NULL	0	1					

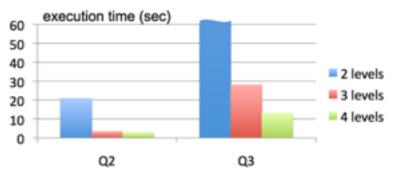
Name.Language.Country							
value		d					
us	0	3					
NULL	2	2					
NULL	1	1					
gb	1	3					
NULL	0	1					

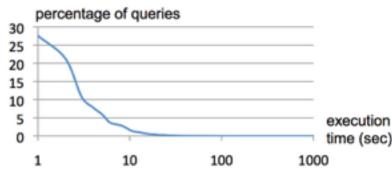


SELECT A, COUNT(B) FROM T GROUP BY A

SELECT A, SUM(c) FROM ( $R_1$  UNION ALL ...  $R_n^{-1}$ ) **GROUP BY A** 







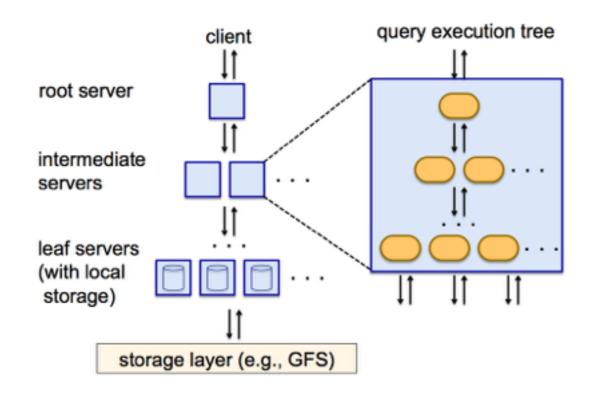
#### Query Language

```
SELECT DocId AS Id,
  COUNT(Name.Language.Code) WITHIN Name AS Cnt,
  Name.Url + ',' + Name.Language.Code AS Str
FROM t
WHERE REGEXP(Name.Url, '^http') AND DocId < 20;</pre>
```

```
message QueryResult {
   required int64 Id;
   repeated group Name {
      optional uint64 Cnt;
      repeated group Language {
         optional string Str; }}}
```

#### Query Execution

#### Serving Tree Topology



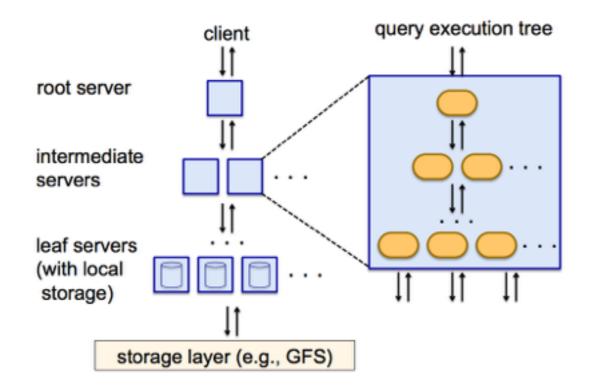
A root server receives incoming queries

Reads metadata from the tables,

Routes the queries to the next level in the serving tree

SELECT A, COUNT(B) FROM T GROUP BY A

SELECT A, SUM(c) FROM (R1 UNION ALL ...  ${\sf R}_n{}^1$  ) GROUP BY A



The amount of data processed in each query is often larger than the number of processing units available for execution, which we call slots.

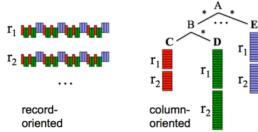
3,000 leaf servers each using 8 threads has 24,000 slots

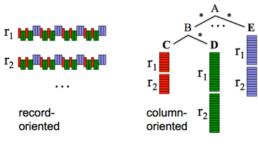
The query dispatcher honors a parameter that specifies the minimum percentage of tablets that must be scanned before returning a result

## Data Model

## Query



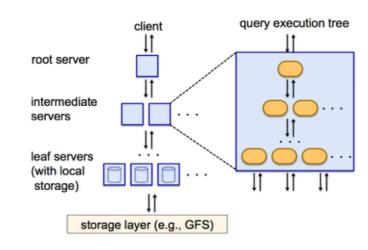




Docld			Name.Url			Links.Forward				Links.Backward			
value		d	value	r	d	value	r	d		value		d	
10	0	0	http://A	0	2	20	0	2		NULL	0	1	
20	0	0	http://B	1	2	40	1	2		10	0	2	
			NULL	1	1	60	1	2		30	1	2	
			http://C	0	2	80	0	2					

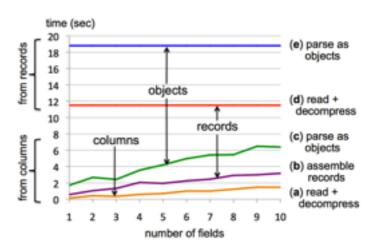
Name.Language.Code							
value	r	d					
en-us	0	2					
en	2	2					
NULL	1	1					
en-gb	1	2					
NULL	0	1					

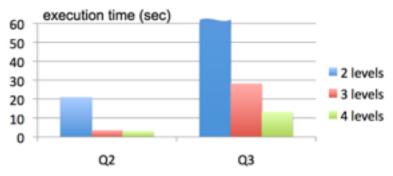
Name.Language.Country						
value		d				
us	0	3				
NULL	2	2				
NULL	1	1				
gb	1	3				
NULL	0	1				



SELECT A, COUNT(B) FROM T GROUP BY A

SELECT A, SUM(c) FROM ( $R_1$  UNION ALL ...  $R_n^{-1}$ ) **GROUP BY A** 





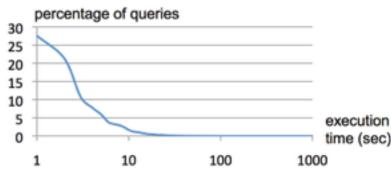


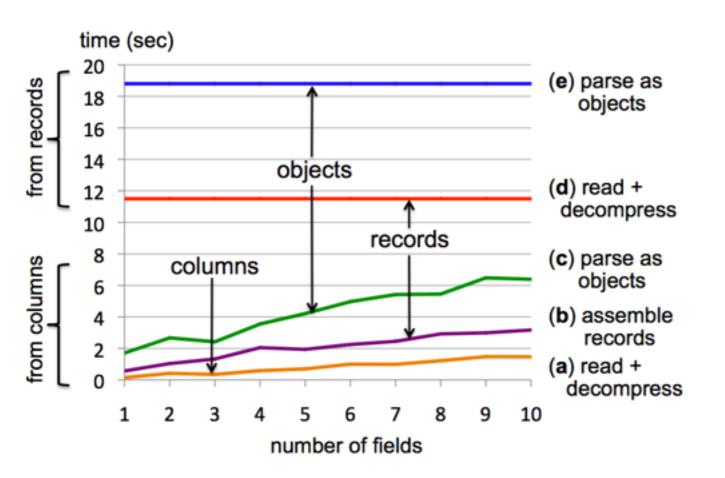
Table name	Number of records	Size (unrepl., compressed)	Number of fields	Data center	Repl. factor
T1	85 billion	87 TB	270	Α	3×
T2	24 billion	13 TB	530	Α	3×
Т3	4 billion	70 TB	1200	A	3×
T4	1+ trillion	105 TB	50	В	3×
T5	1+ trillion	20 TB	30	В	2×

Local Disk MR & Dremel Scalability

Pre-tablet Histogram Serving Tree Topology

Within-Record Aggregation Straggler

#### Local Disk



Tradeoffs of columnar vs. record-oriented storage

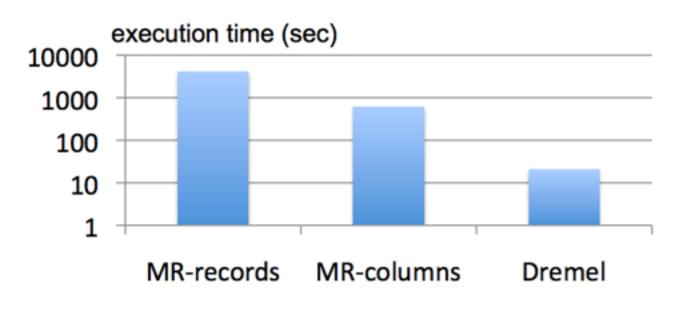
Record assembly and parsing are expensive, each potentially doubling the execution time.

A bulk of the time is spent in decompression;

The compressed data can be read from the disk in about half the time.

Columnar Storage easier to compress

#### MR & Dremel



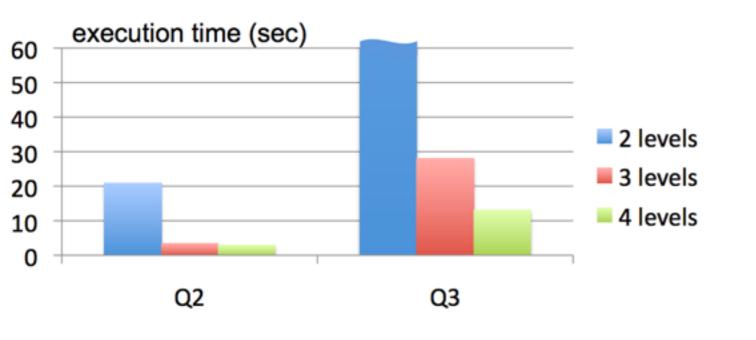
Q1: SELECT SUM(CountWords(txtField)) / COUNT(\*)
FROM T1

Both MR jobs are run on 3000 workers.

Similarly, a 3000-node Dremel instance is used to execute Query Q1.

Dremel and MR-on-columns read about 0.5TB of compressed columnar data vs. 87TB (MR-on-records)

### Serving tree Topology



2 levels: 1:2900

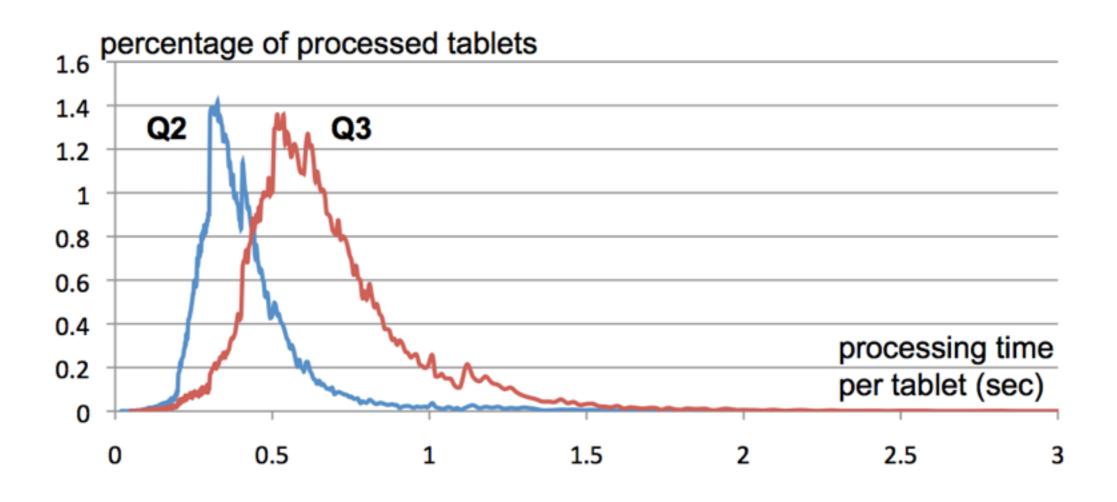
3 levels: 1:100:2900

4 levels: 1:10:100:2900

Q2: SELECT country, SUM(item.amount) FROM T2 GROUP BY country

Q3: SELECT domain, SUM(item.amount) FROM T2 WHERE domain CONTAINS '.net' GROUP BY domain

#### Per-tablet Histograms

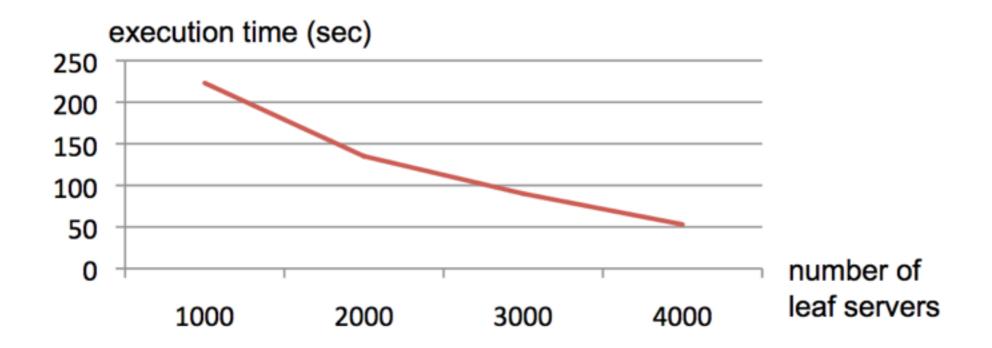


The figure shows how fast tablets get processed by the leaf servers for a specific run of Q2 and Q3.

Q2: SELECT country, SUM(item.amount) FROM T2 GROUP BY country

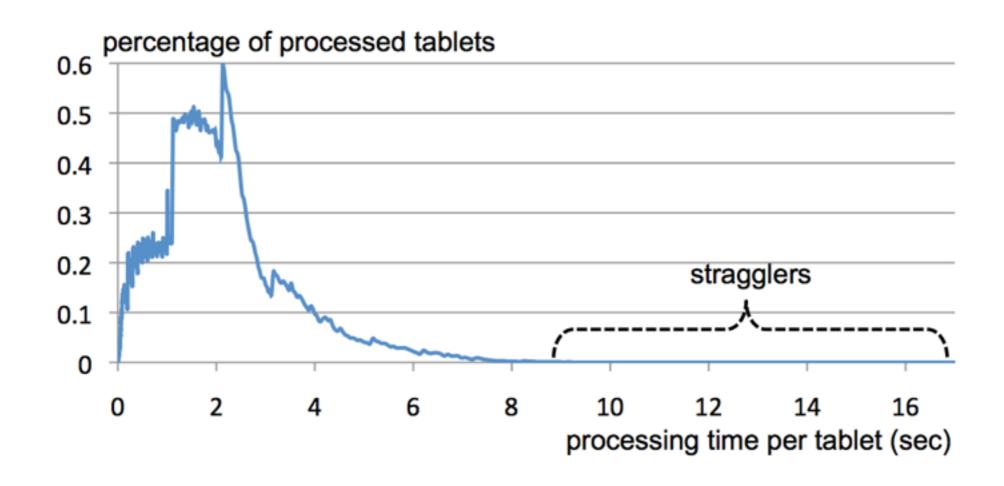
Q3: SELECT domain, SUM(item.amount) FROM T2 WHERE domain CONTAINS '.net' GROUP BY domain

#### Scalability



Near-linear scalability in the number of columns and servers is achievable for systems containing thousands of nodes

#### Stragglers

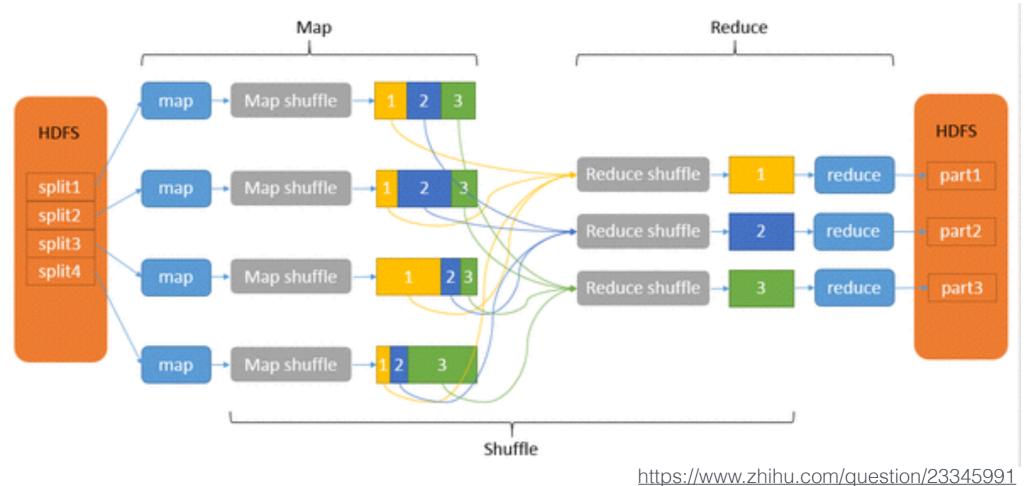


The processing time for 99% of the tablets is below 5 seconds per tablet per slot. The Query is executed on a 2500 node system.

## Summary

- 1. Scan-based queries can be executed at interactive speeds on disk-resident datasets of up to a trillion records.
- 2. In a multi-user environment, a larger system can benefit from economies of scale while offering a qualitatively better user experience.
- 3. If trading speed against accuracy is acceptable, a query can be terminated much earlier and yet see most of the data.
- 4. Column Store is naturally suitable for statistical analysis of data.

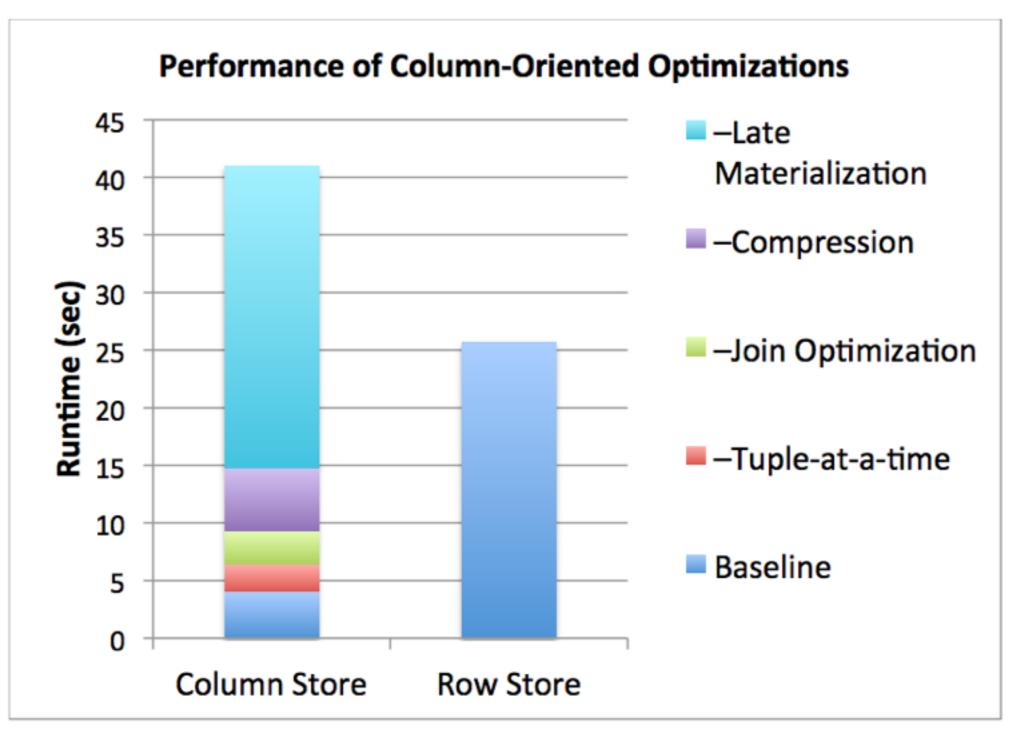
## MapReduce



nttps://www.zninu.com/question/2334599

Batch Mode

#### Columnar Storage



Data Model and Interface

**Execution Engine** 

Scheduling

Fault Tolerance

Storage Layer

#### Reflection

#### DataFlow System

The need to reduce the gap between the generation of data and the generation of analytics results over this data has led to systems that can support both OLTP and OLAP workloads in a single system. — Massively Parallel Databases and MapReduce Systems

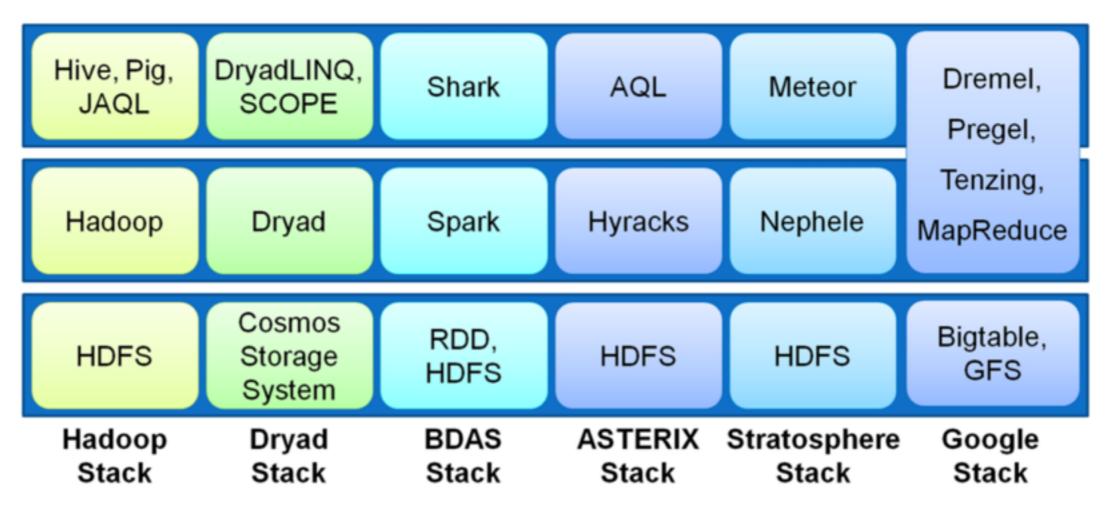
Whether could distribute the computing to Leaf Servers?

Data Format

Analytic Algorithms

### Reflection

#### **Dataflow Stack**



Massively Parallel Databases and MapReduce Systems

Thank you!

Open source Implementation of Dremel

Cloudera Impala

Apache Drill