Introduction to Database Systems CSE 414

Lecture 28

Parallel Databases Wrap-up

Announcement

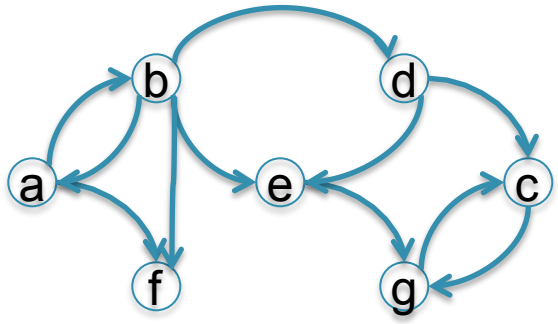
- Homework 8 (last) due on Friday night, 11pm
  NO LATE DAYS / EXTENSIONS
- Review session:
  - Sunday, 2 pm, Loew 101
- Final exam:
  - Next Monday, 2:30 pm, regular location
  - Same rules as before: open textbook, 1 sheet of paper with hand-written notes (+ midterm sheet)

Graph Analysis
In Parallel Databases

Graph Databases

Many large databases are graphs

- Examples?

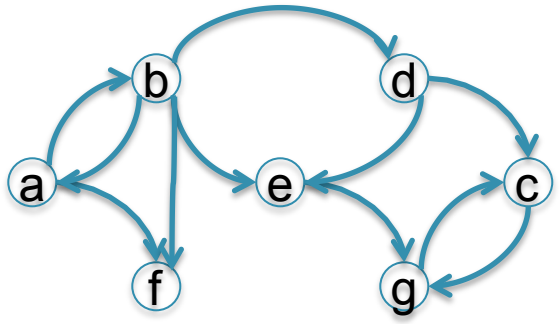


Source	Target
a	b
b	a
a	f
b	f
b	e
b	d
d	e
d	c
e	g
g	c
c	g

Graph Databases

Many large databases are graphs

- Examples:
  - The Web
  - The Internet
  - Social Networks
  - Flights between airports
  - Etc.

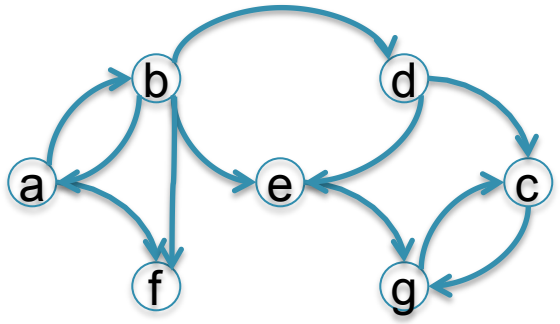


Source	Target
a	b
b	a
a	f
b	f
b	e
b	d
d	e
d	c
e	g
g	c
c	g

Data Analytics on Big Graphs

Queries expressible in SQL:

- How many nodes (edges)?
- How many nodes have > 4 neighbors?
- Which are "most connected nodes"?



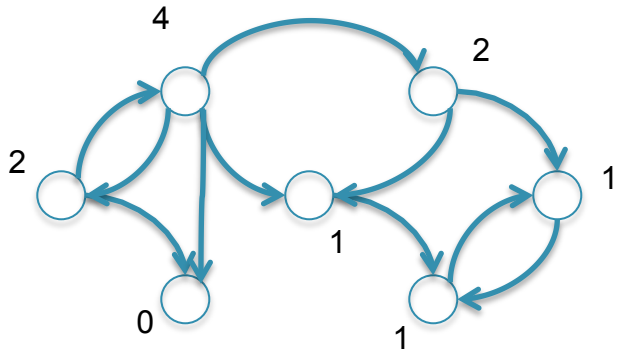
Source	Target
a	b
b	a
a	f
b	f
b	e
b	d
d	e
d	c
e	g
g	c
c	g

Queries requiring recursion:

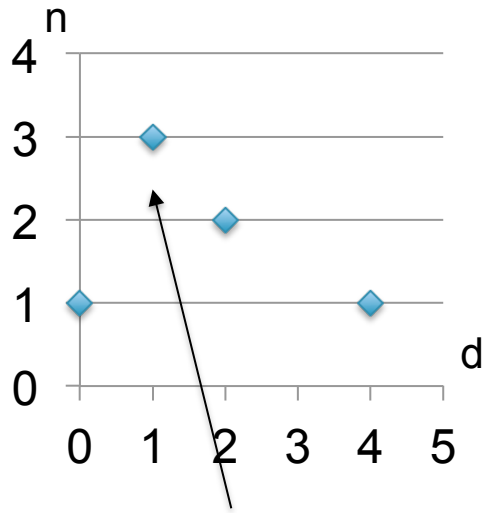
- Is the graph connected?
- What is the diameter of the graph?
- Compute *PageRank*
- Compute the *Centrality* of each node

Example: the Histogram of a Graph

- Outdegree of a node = number of outgoing edges
- For each d, let n(d) = number of nodes with oudegree d
- The outdegree histogram of a graph = the scatterplot (d, n(d))

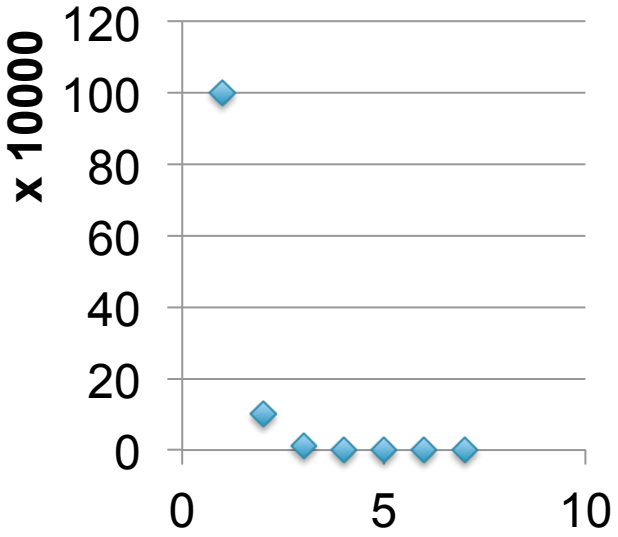


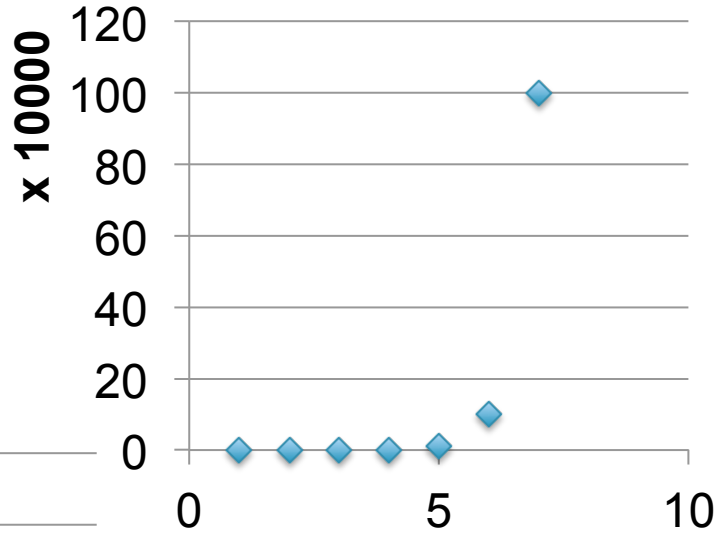
d	n(d)
0	1
1	3
2	2
3	0
4	1

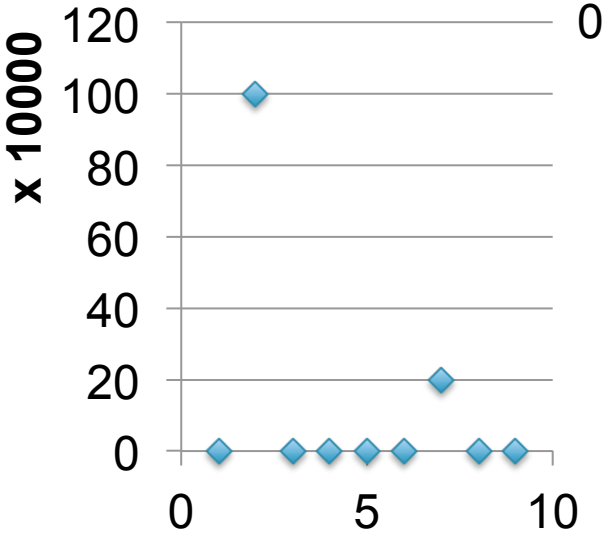


Outdegree 1 is seen at 3 nodes

Histograms Tell Us Something About the Graph





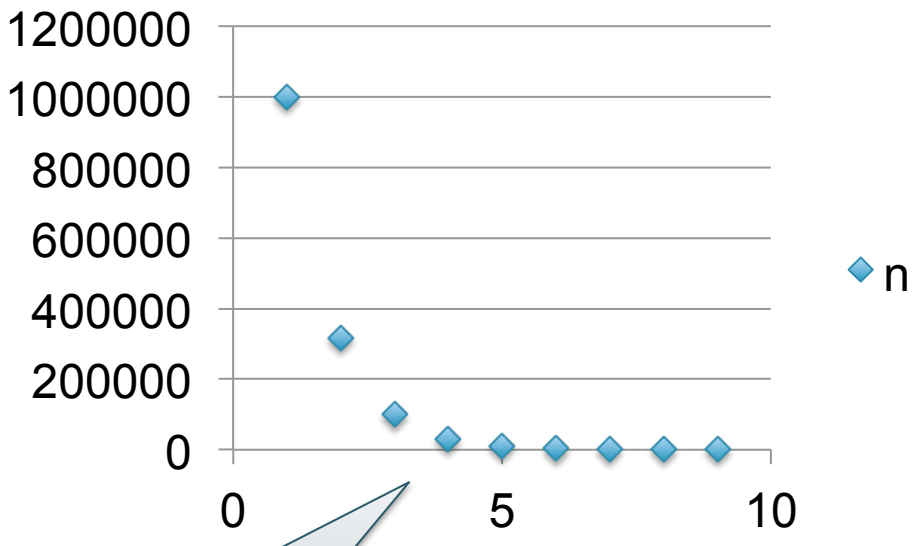


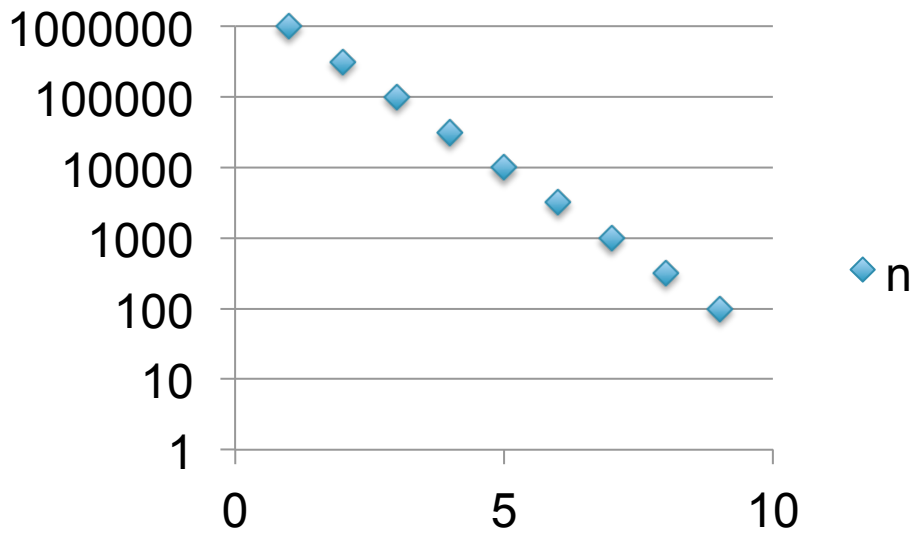
What can you say about these graphs?

Exponential Distribution

nodes with degree d

- $n(d) \cong c/2^d$ (generally, cx^d, for some $x < 1$)
- A *random graph* has exponential distribution
- Best seen when n is on a log scale

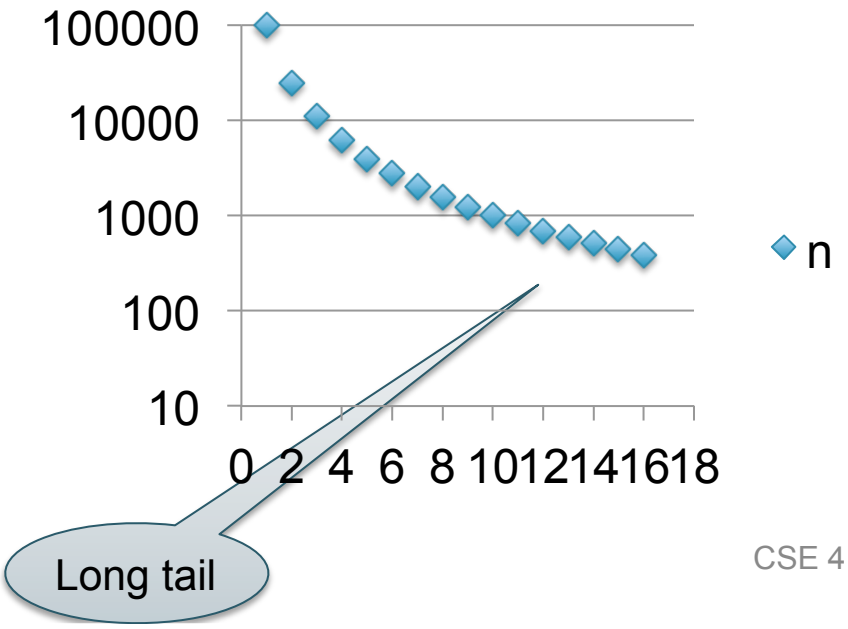


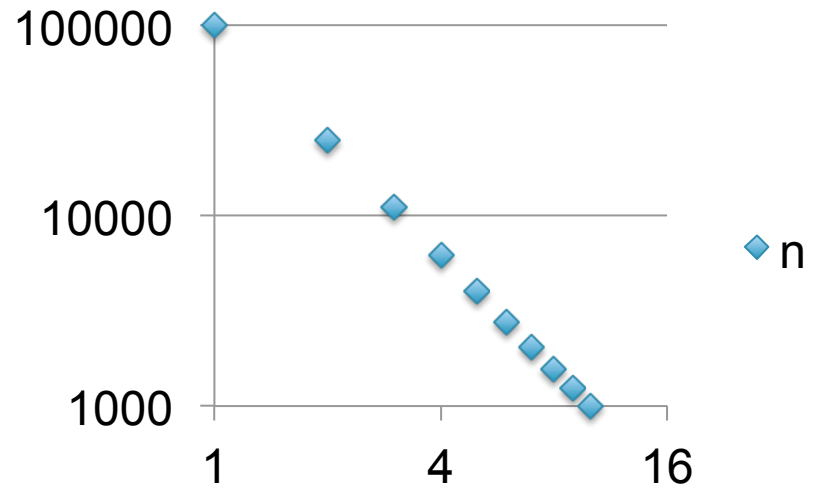


Quickly vanishing

Zipf Distribution

- $n(d) \cong 1/d^x$, for some value x>0
- Human-generated data has Zipf distribution: letters in alphabet, words in vocabulary, etc.
- Best seen in a log-log scale





The Histogram of the Web

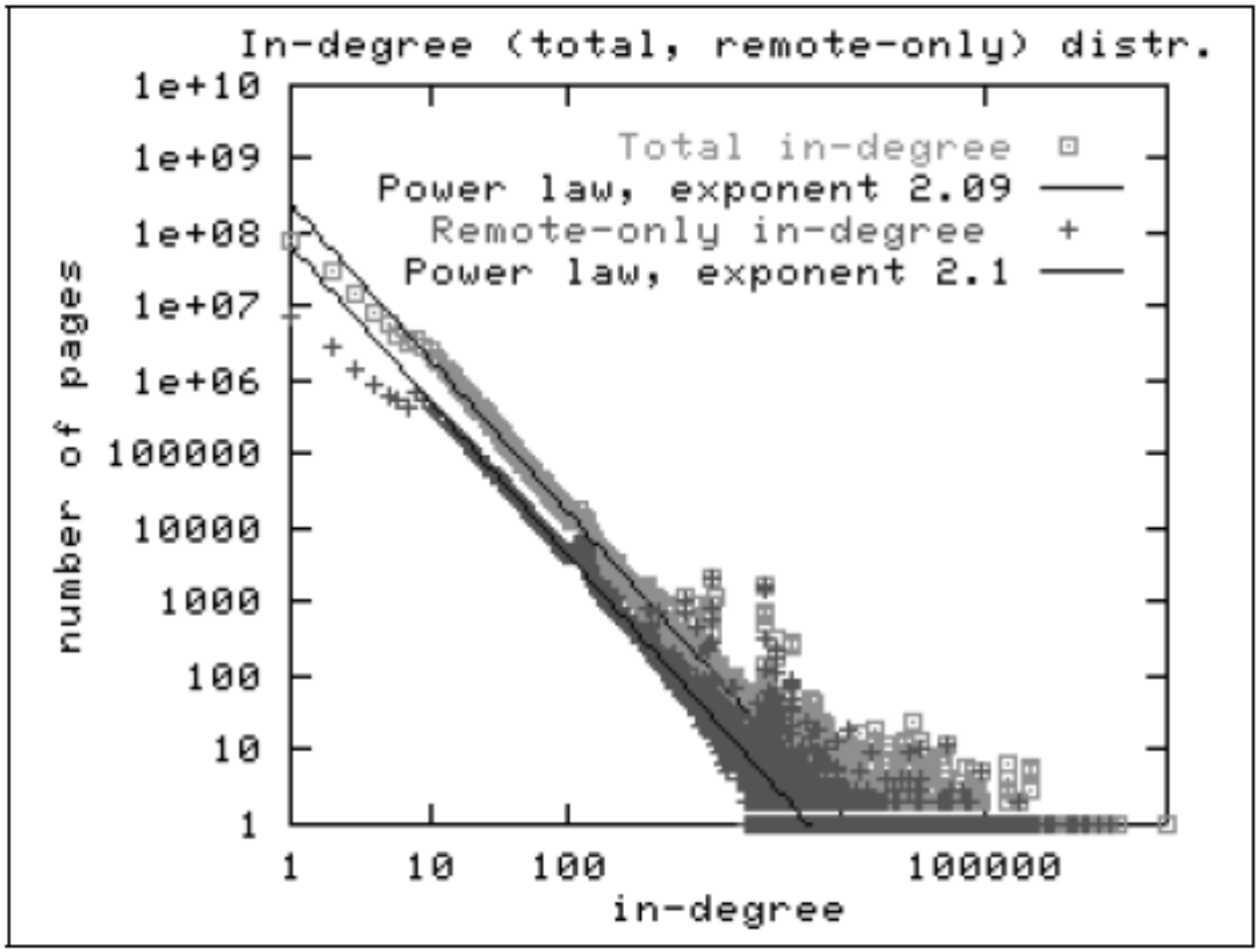


Figure 2: In-degree distribution.

Late 1990's
200M Webpages

Exponential ?

Zipf ?

The Bowtie Structure of the Web

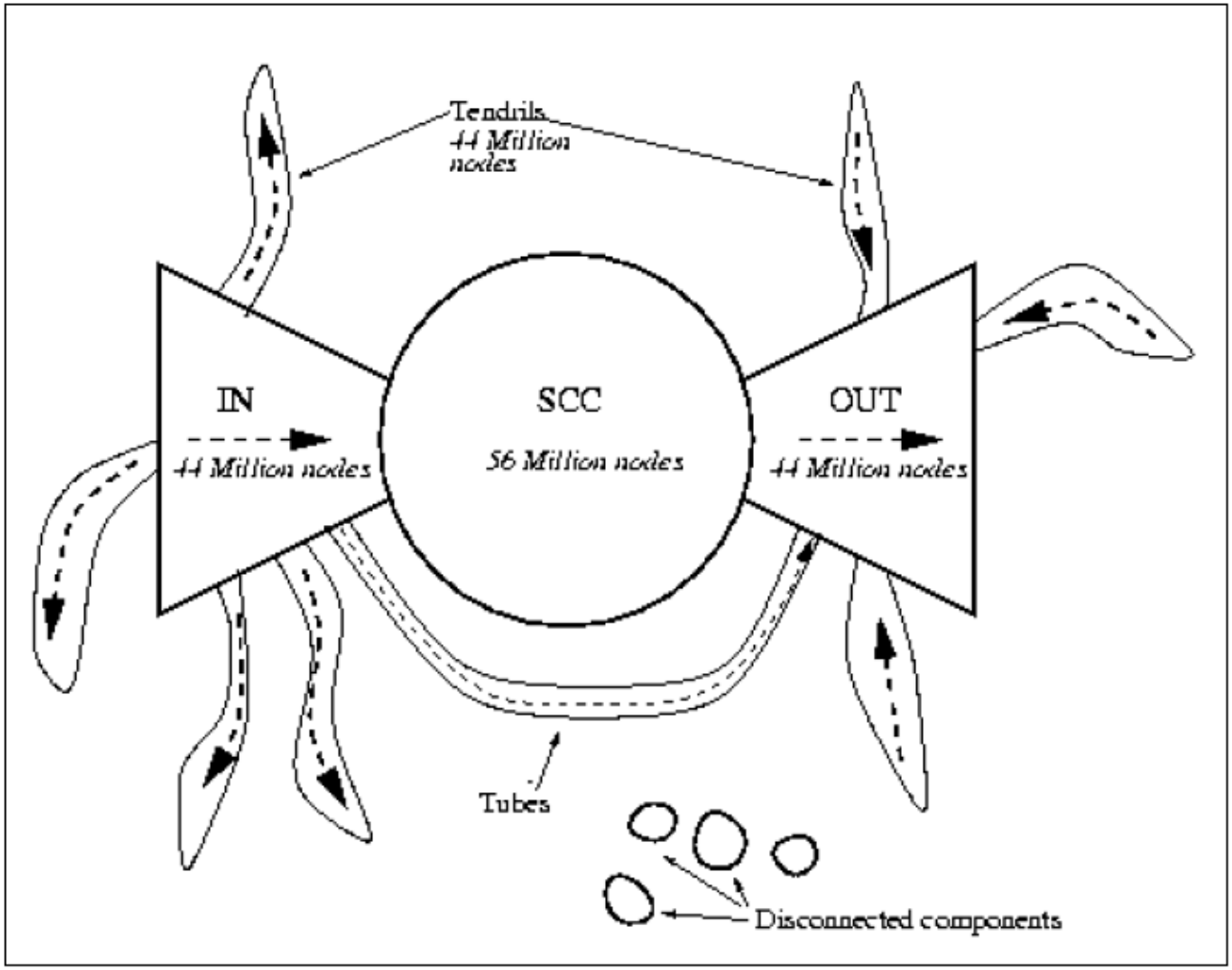


Figure 4: The web as a bowtie. SCC is a giant strongly connected component. IN consists of pages with paths to SCC, but no path from SCC. OUT consists of pages with paths from SCC, but no path to SCC. TENDRILS consists of pages that cannot surf to SCC, and which cannot be reached by surfing from SCC.

Executing a Large MapReduce Job

Anatomy of a Query Execution

- Running problem #4
- 20 nodes = 1 master + 19 workers
- Using PARALLEL 50

March 2013

Hadoop job_201303091944_0001 on domU-12-31-39-06-75-A1

User: hadoop
Job Name: PigLatin:DefaultJobName
Job File: hdfs://10.208.122.79:9000/mnt/var/lib/hadoop/tmp/mapred/staging/hadoop/.staging/job_201303091944_0001/job.xml
Submit Host: domU-12-31-39-06-75-A1.compute-1.internal
Submit Host Address: 10.208.122.79
Job-ACLs: All users are allowed
Job Setup: Successful
Status: Succeeded
Started at: Sat Mar 09 19:49:21 UTC 2013
Finished at: Sat Mar 09 23:33:14 UTC 2013
Finished in: 3hrs, 43mins, 52sec
Job Cleanup: Successful
Black-listed TaskTrackers: 1

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	100.00%	7908	0	0	7908	0	14 / 16
reduce	100.00%	50	0	0	50	0	0 / 8

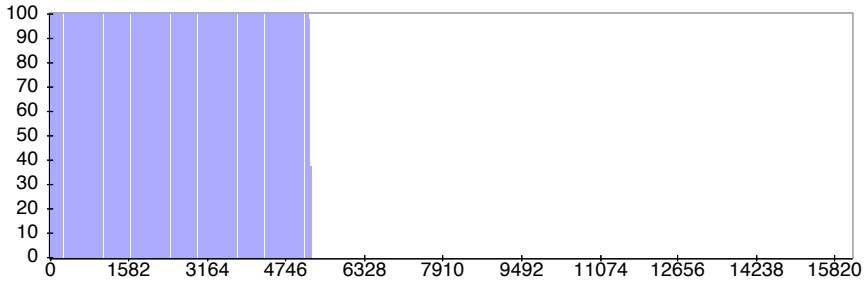
	Counter	Map	Reduce	Total
Job Counters	SLOTS_MILLIS_MAPS	0	0	454,162,761
	Launched reduce tasks	0	0	58
	Total time spent by all reduces waiting after reserving slots (ms)	0	0	0
	Rack-local map tasks	0	0	7,938
	Total time spent by all maps waiting after reserving slots	0	0	0

Some other time (March 2012)

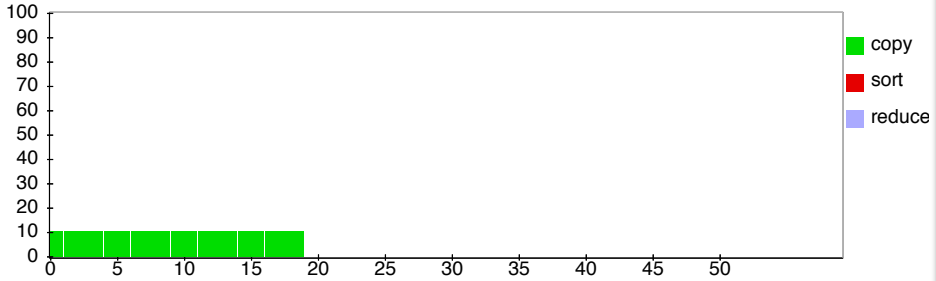
- Let's see what happened…

1h 16min

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	33.17%	15816	10549	38	5229	0	0 / 0
reduce	4.17%	50	31	19	0	0	0 / 0



luce Completion Graph - close



1h 16min

Only 19 reducers active, out of 50. Why?

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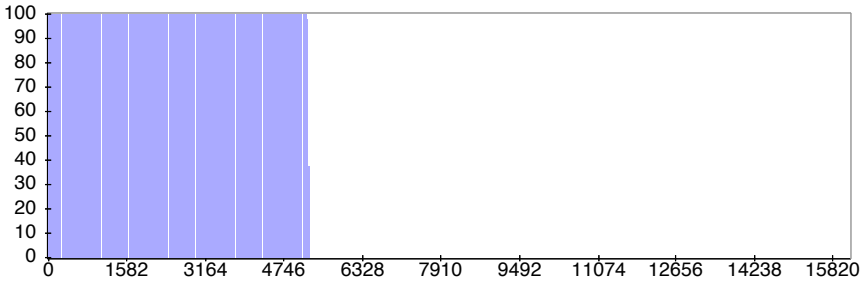

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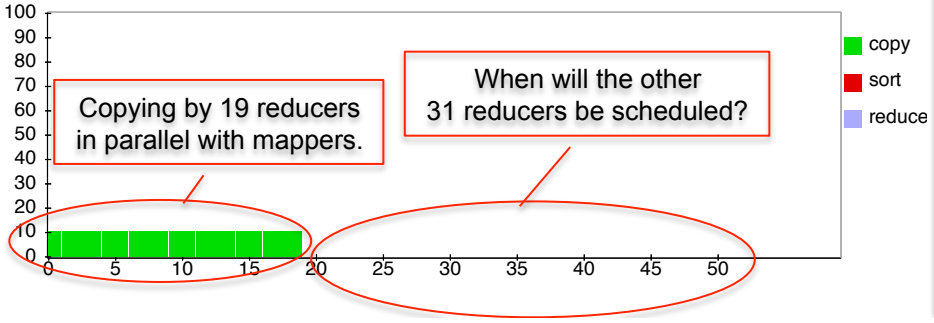

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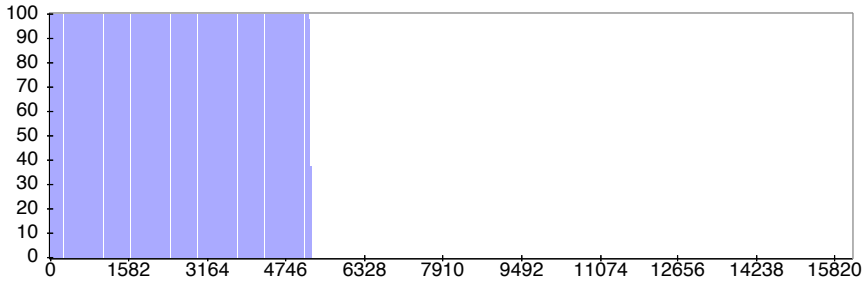

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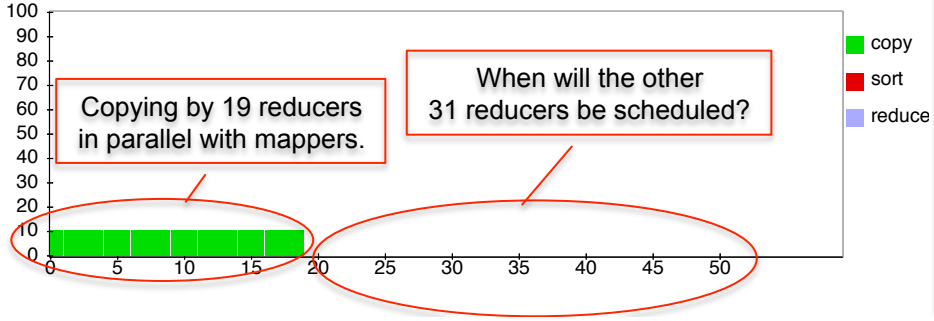
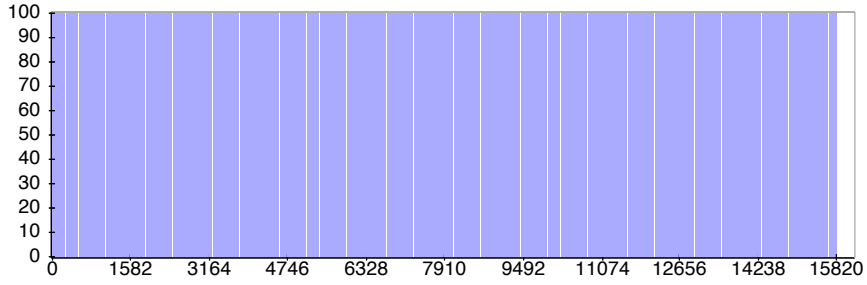


3h 50min

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	100.00%	15816	0	0	15816	0	0 / 18
reduce	32.42%	50	31	19	0	0	0 / 0



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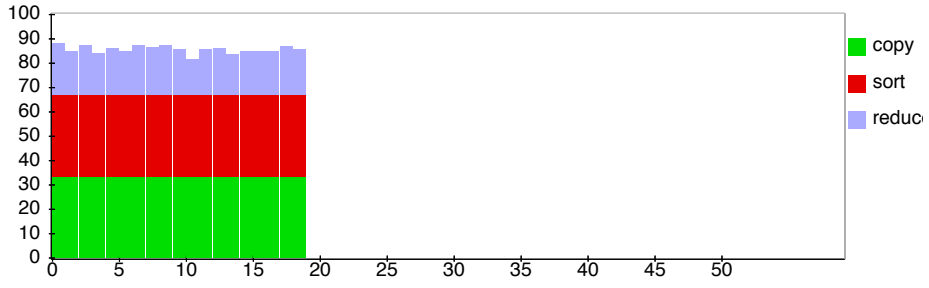

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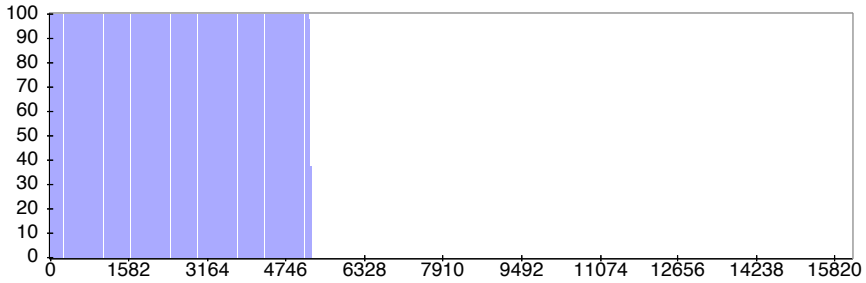

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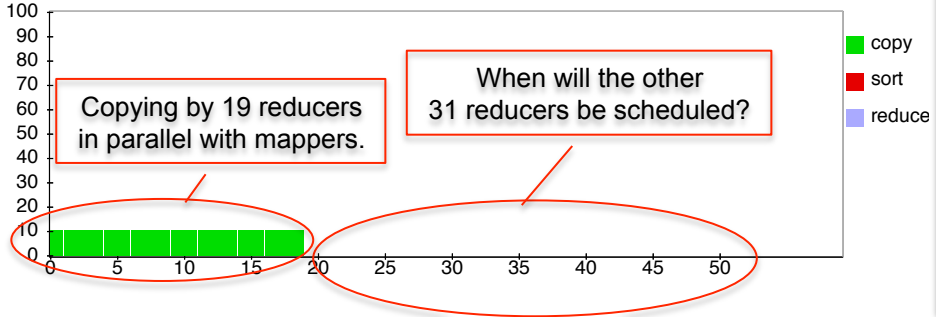

3h 50min

Speculative Execution

Completed. Sorting, and the rest of Reduce may proceed now

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
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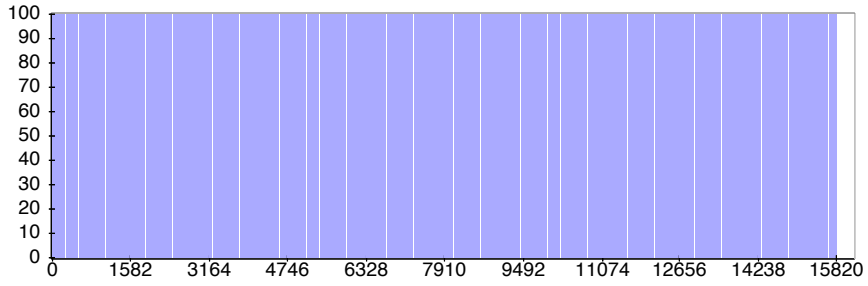
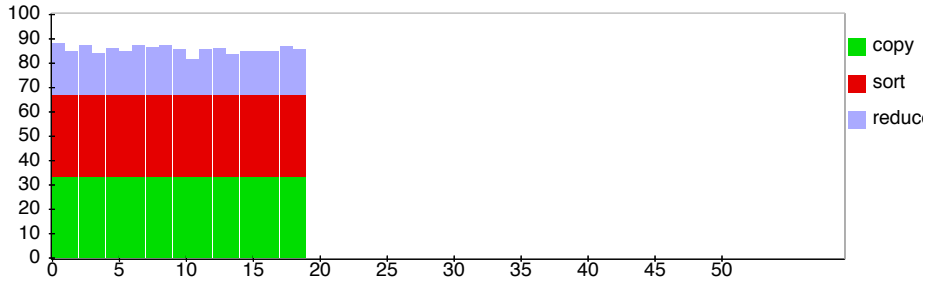
reduce	32.42%	50	31	19	0	0	0 / 0



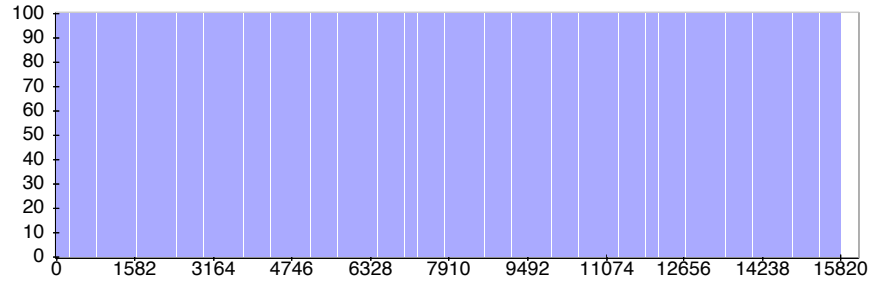
ıce Completion Graph - close



3h 51min

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	100.00%	15816	0	0	15816	0	0 / 18
reduce	37.72%	50	19	22	9	0	0 / 0

› Completion Graph - close



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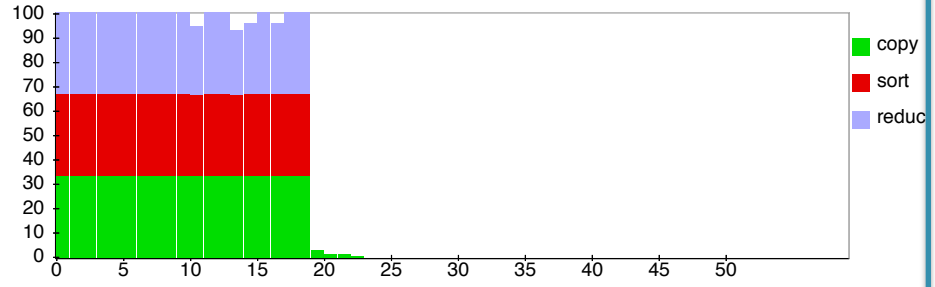

3h 51min

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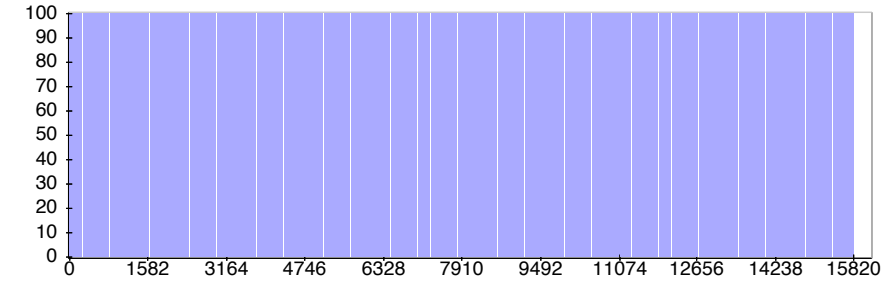
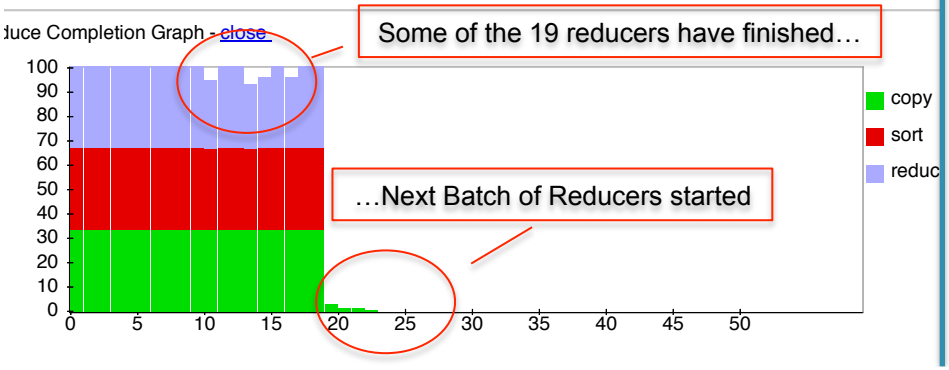


duce Completion Graph - close



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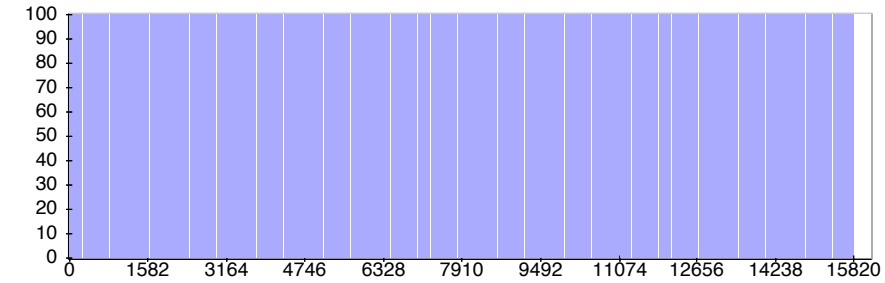
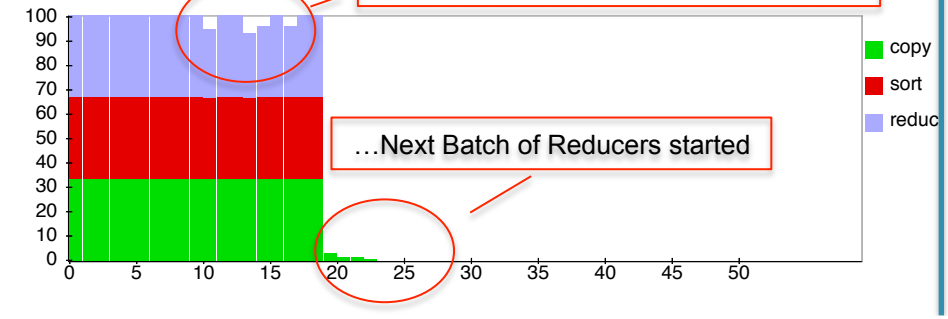


duce Completion Graph - close

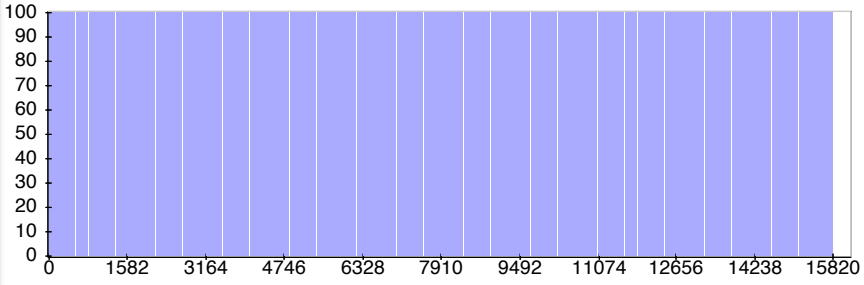
Some of the 19 reducers have finished…



…Next Batch of Reducers started

3h 52min

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	100.00%	15816	0	0	15816	0	0 / 18
reduce	42.35%	50	11	20	19	0	0 / 0



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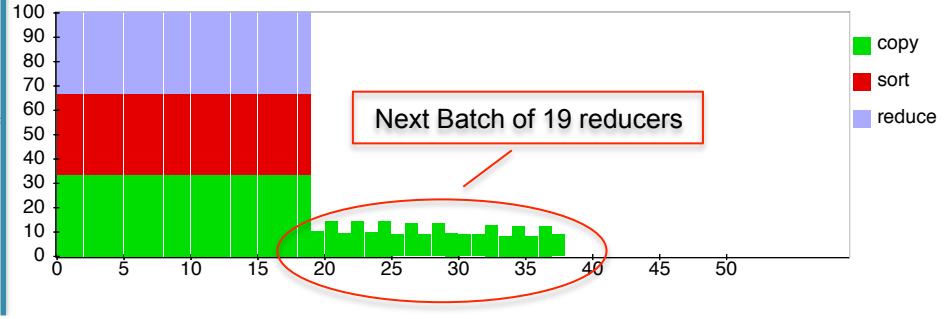
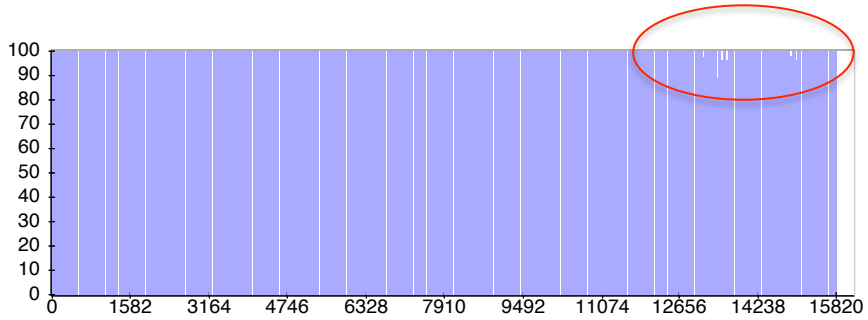


Next Batch of 19 reducers

4h 18min

Several servers failed: “fetch error”. Their map tasks need to be rerun. All reducers are waiting….

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	99.88%	15816	2638	30	13148	0	15 / 3337
reduce	48.42%	50	15	16	19	0	0 / 0



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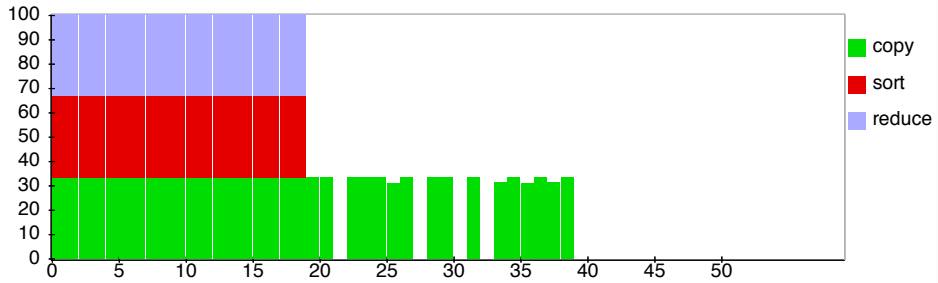

4h 18min

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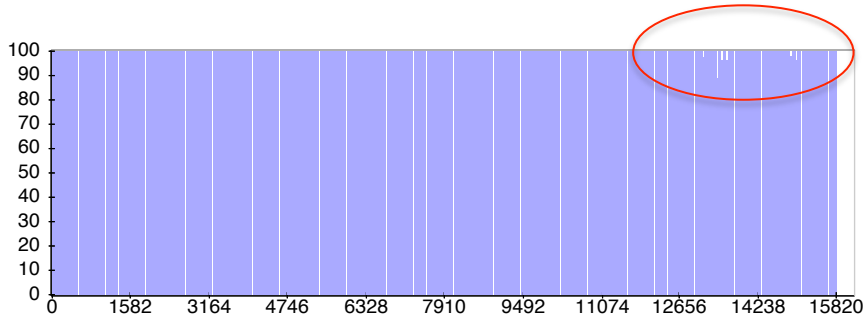

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4h 18min

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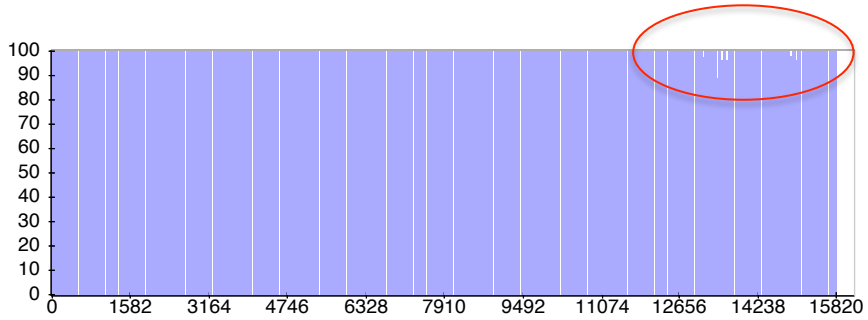

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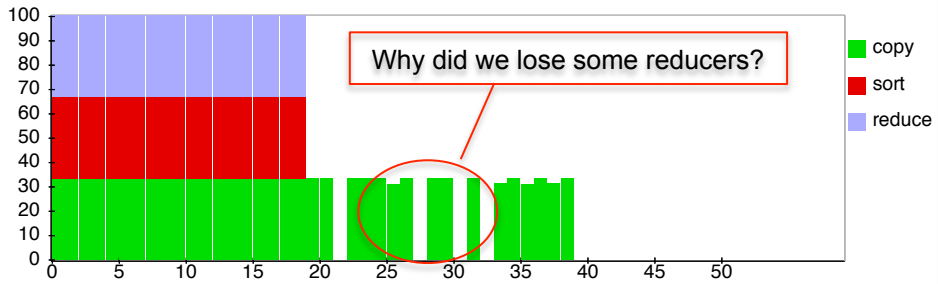
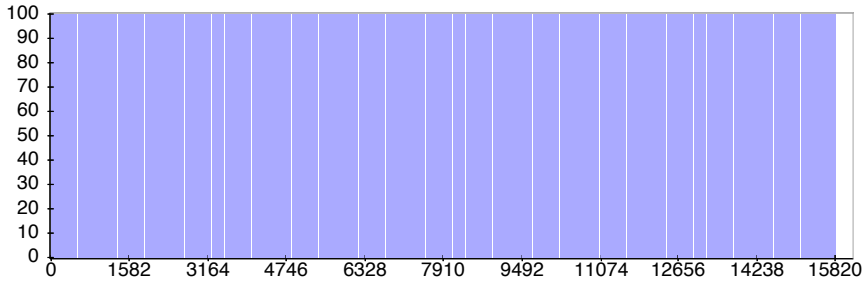


7h 10min

Mappers finished, reducers resumed.

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	100.00%	15816	0	0	15816	0	26 / 5968
reduce	94.15%	50	0	6	44	0	0 / 8



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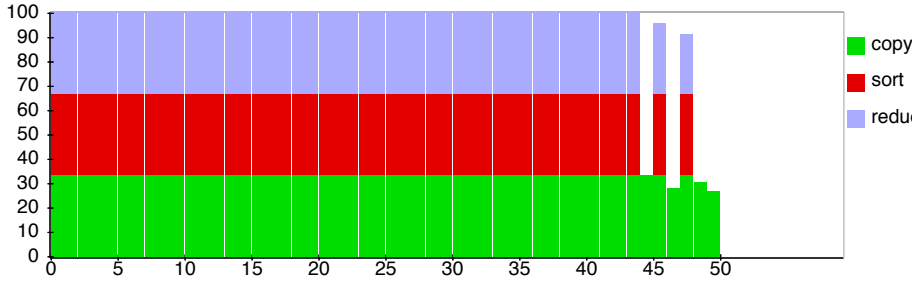


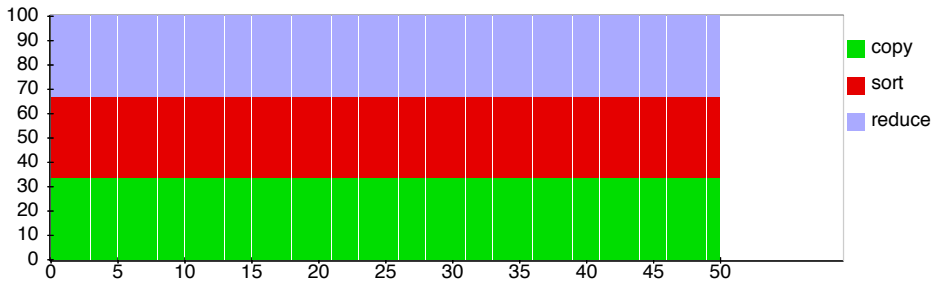
7h 20min

Success! 7hrs, 20mins.

Hadoop job_201203041905_0001 on ip-10-203-30-146

User: hadoop
Job Name: PigLatin:DefaultJobName
Job File: hdfs://10.203.30.146:9000/mnt/var/lib/hadoop/tmp/mapred/staging/hadoop/.staging/job_201203041905_0001/job.xml
Submit Host: ip-10-203-30-146.ec2.internal
Submit Host Address: 10.203.30.146
Job-ACLs: All users are allowed
Job Setup: Successful
Status: Succeeded
Started at: Sun Mar 04 19:08:29 UTC 2012
Finished at: Mon Mar 05 02:28:39 UTC 2012
Finished in: 7hrs, 20mins, 10sec
Job Cleanup: Successful
Black-listed TaskTrackers: 3

Kind	% Complete	Num Tasks	Pending	Running	Complete	Killed	Failed/Killed Task Attempts
map	100.00%	15816	0	0	15816	0	26 / 5968
reduce	100.00%	50	0	0	50	0	0 / 14



Parallel Joins in MapReduce

Hash Join in Pig

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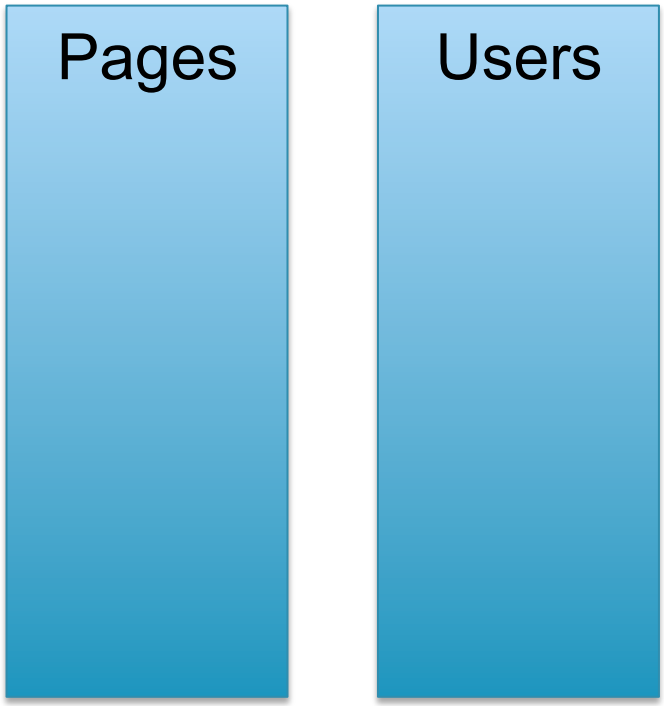
Users = load 'users' as (name, age);
Pages = load 'pages' as (user, url);
Jnd = join Users by name, Pages by user;
```



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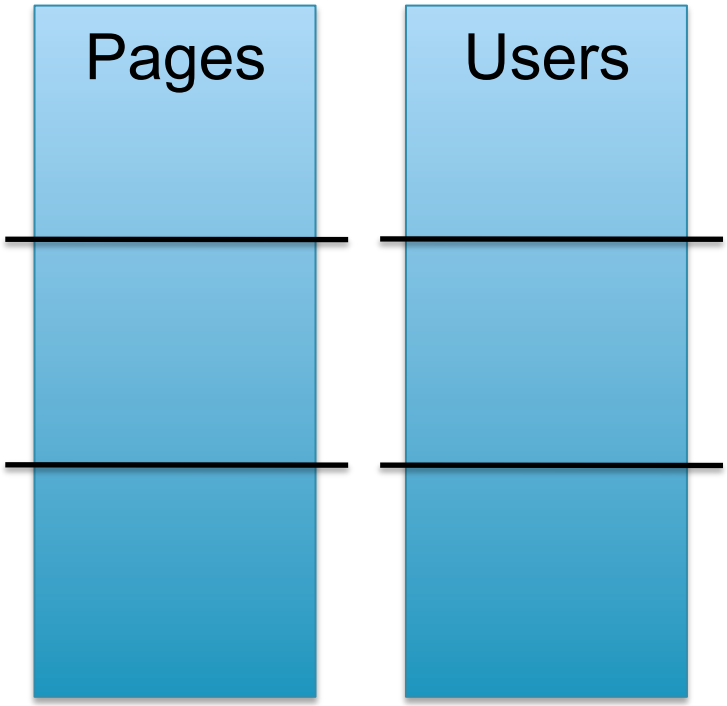

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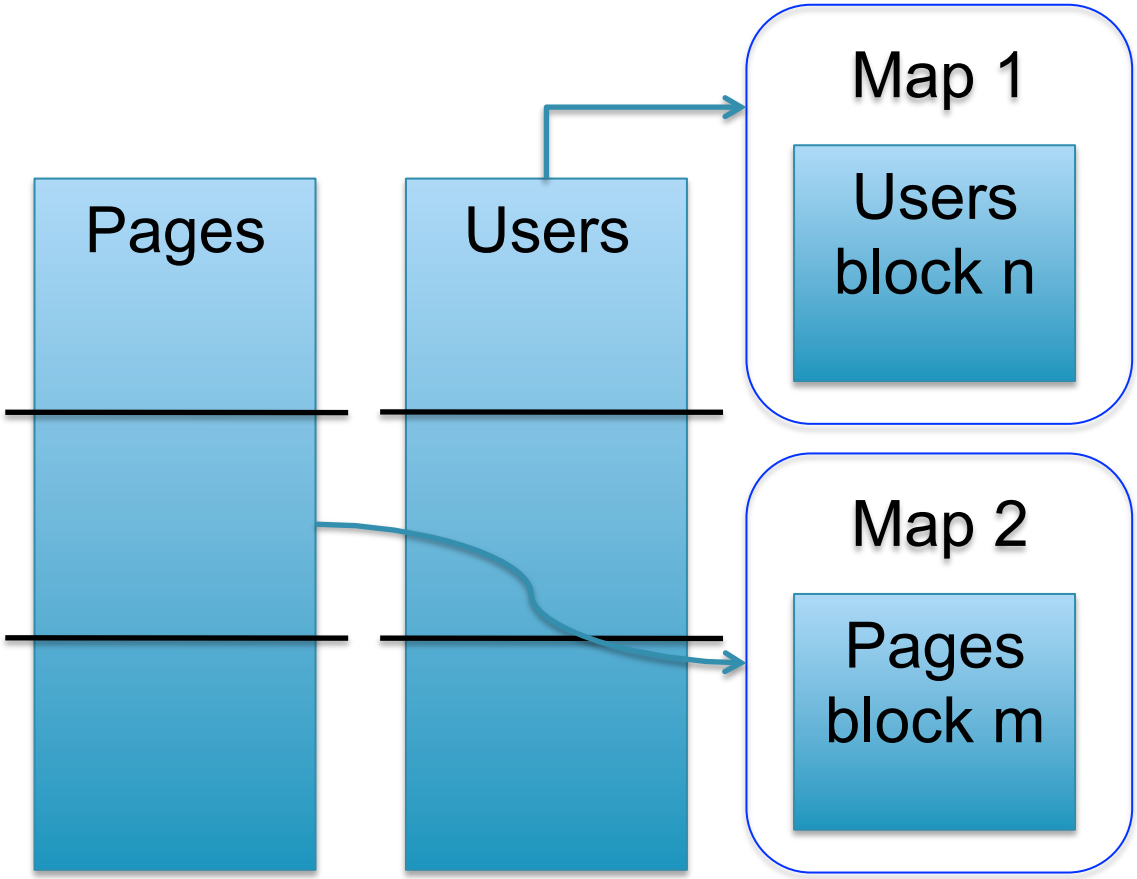

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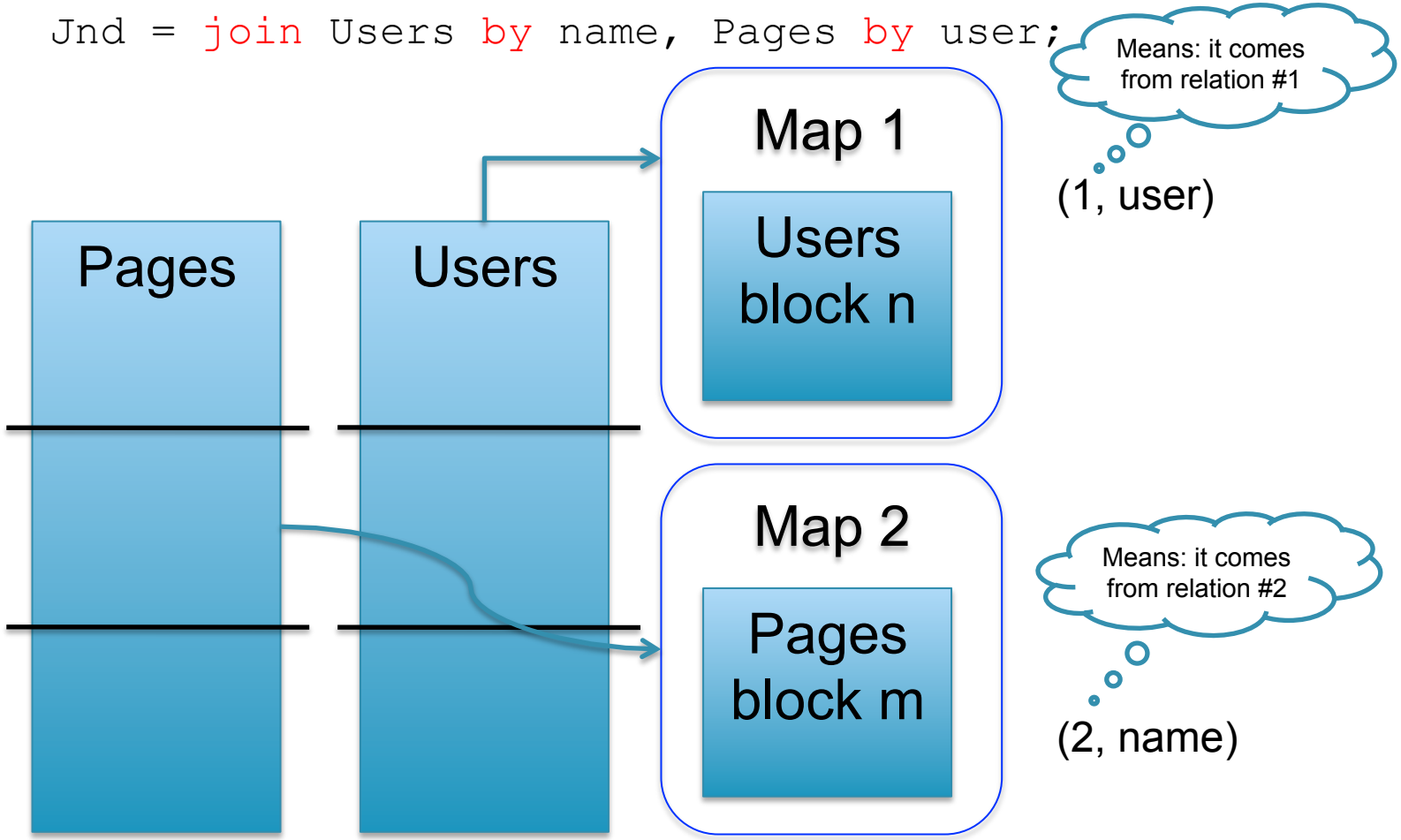

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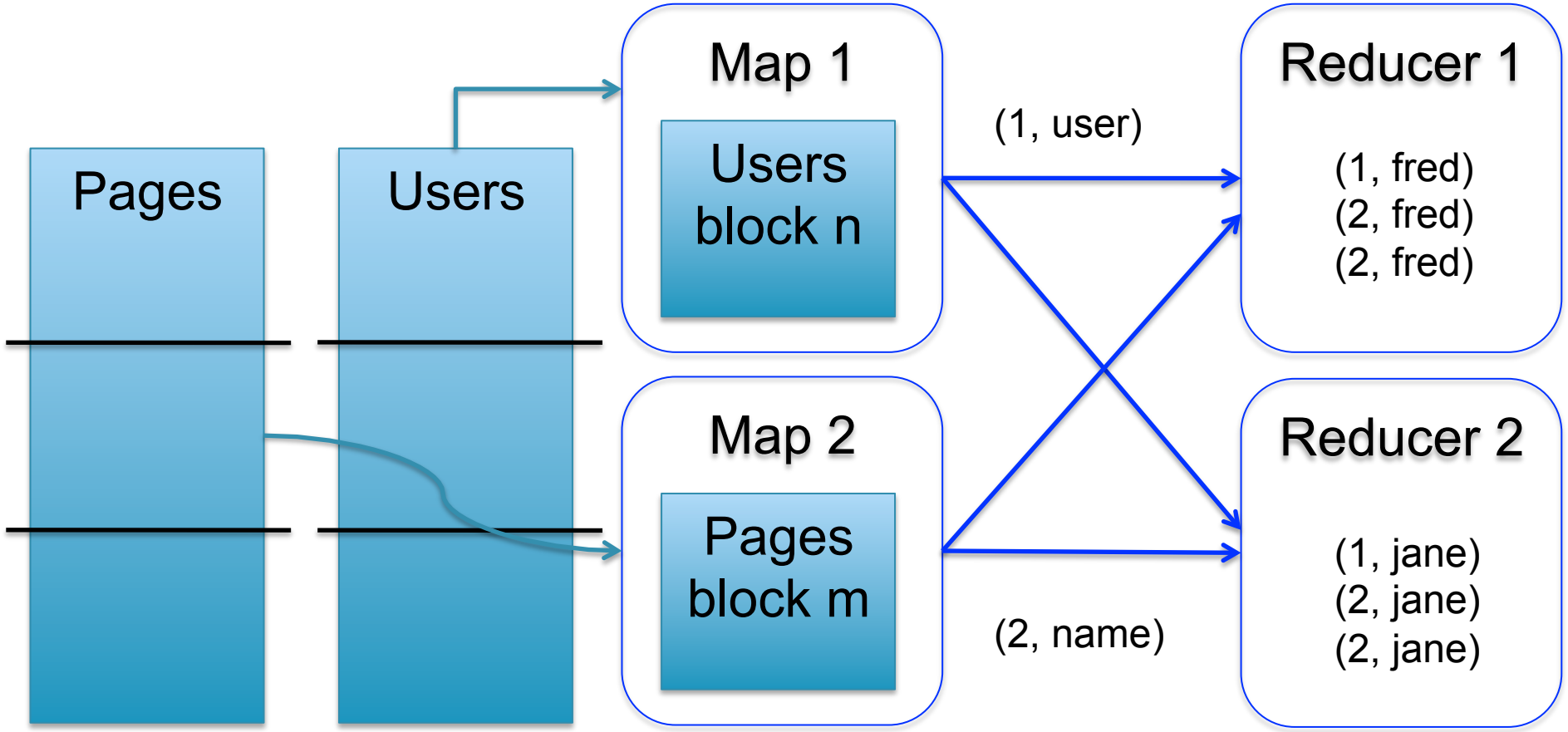

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Hash Join

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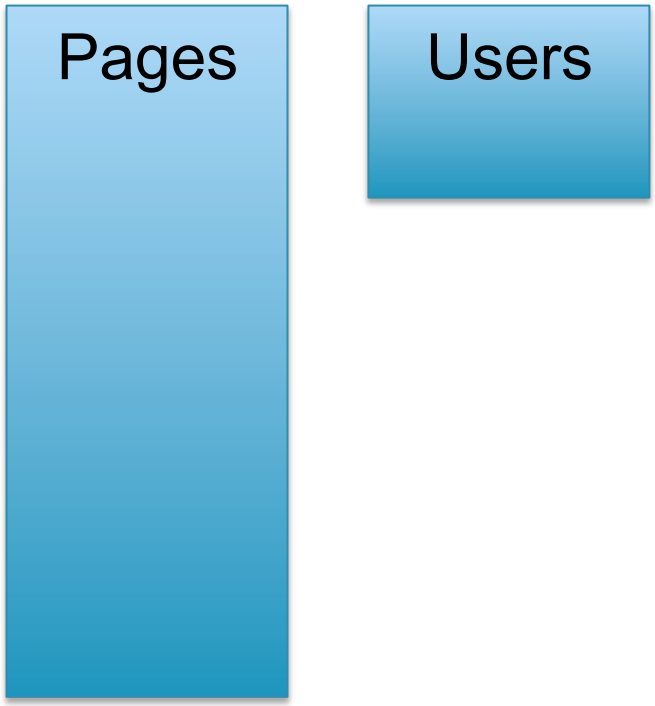
```
map(String usr, String value):
    // usr: either Users.name or Pages.user
    // value.relation is either 'Users' or 'Pages'
    if value.relation='Users':
      EmitIntermediate(usr, (1, value));
    else
      EmitIntermediate(usr, (2, value));
```

```
reduce(String usr, Iterator values):
    Users = empty;  Pages = empty;
    for each v in values:
      if v.type = 1: Users.insert(v)
      else Pages.insert(v);
    for v1 in Users, for v2 in Pages
      Emit(usr, v1,v2);
```

Broadcast Join

```
Users = load 'users' as (name, age);
Pages = load 'pages' as (user, url);
Jnd = join Pages by user, Users by name using "replicated";
```



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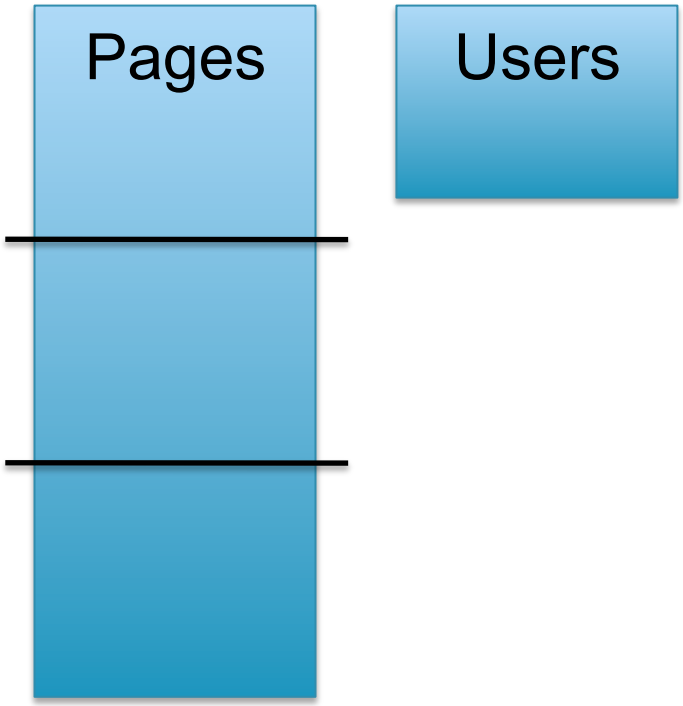

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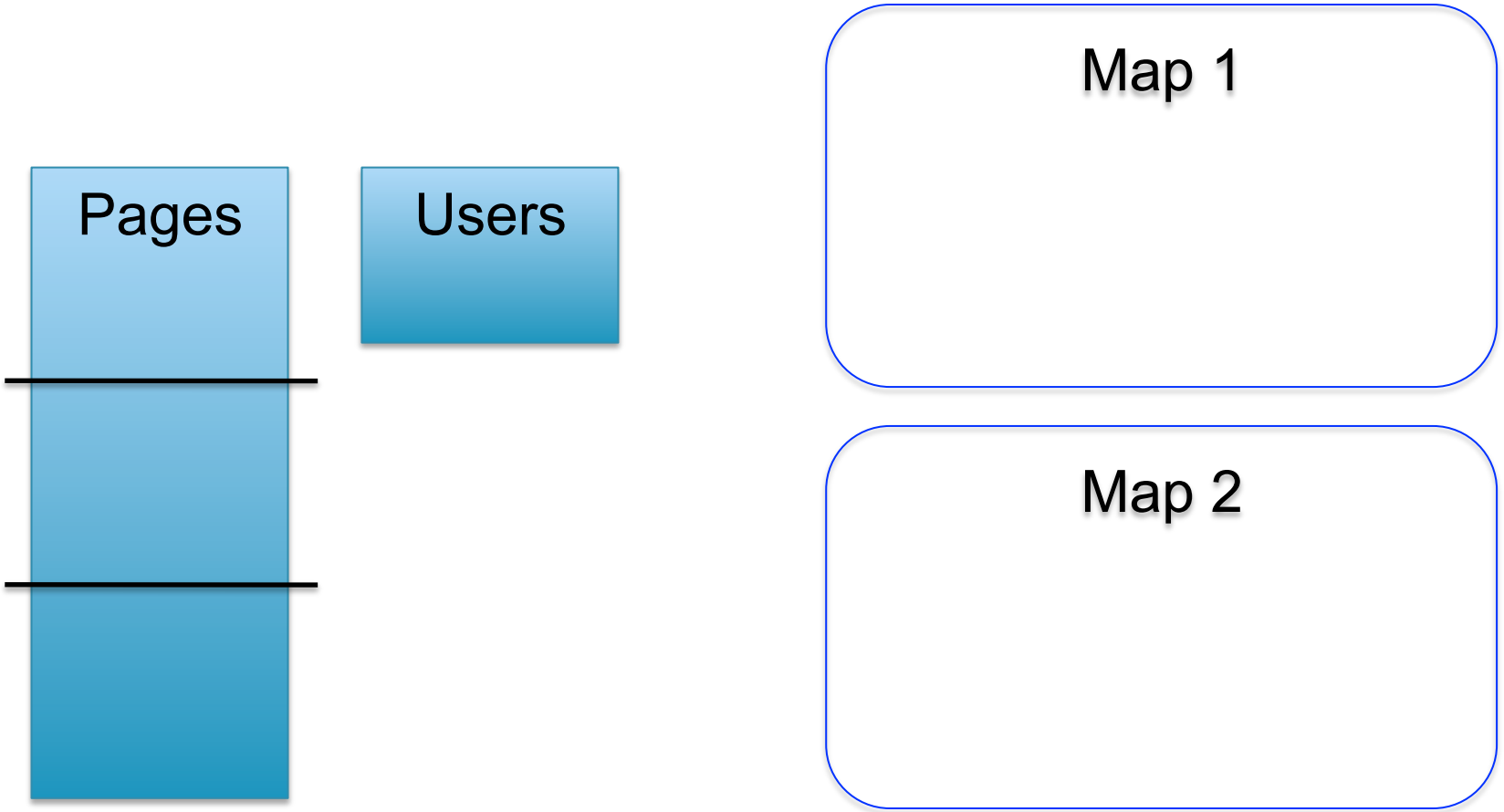

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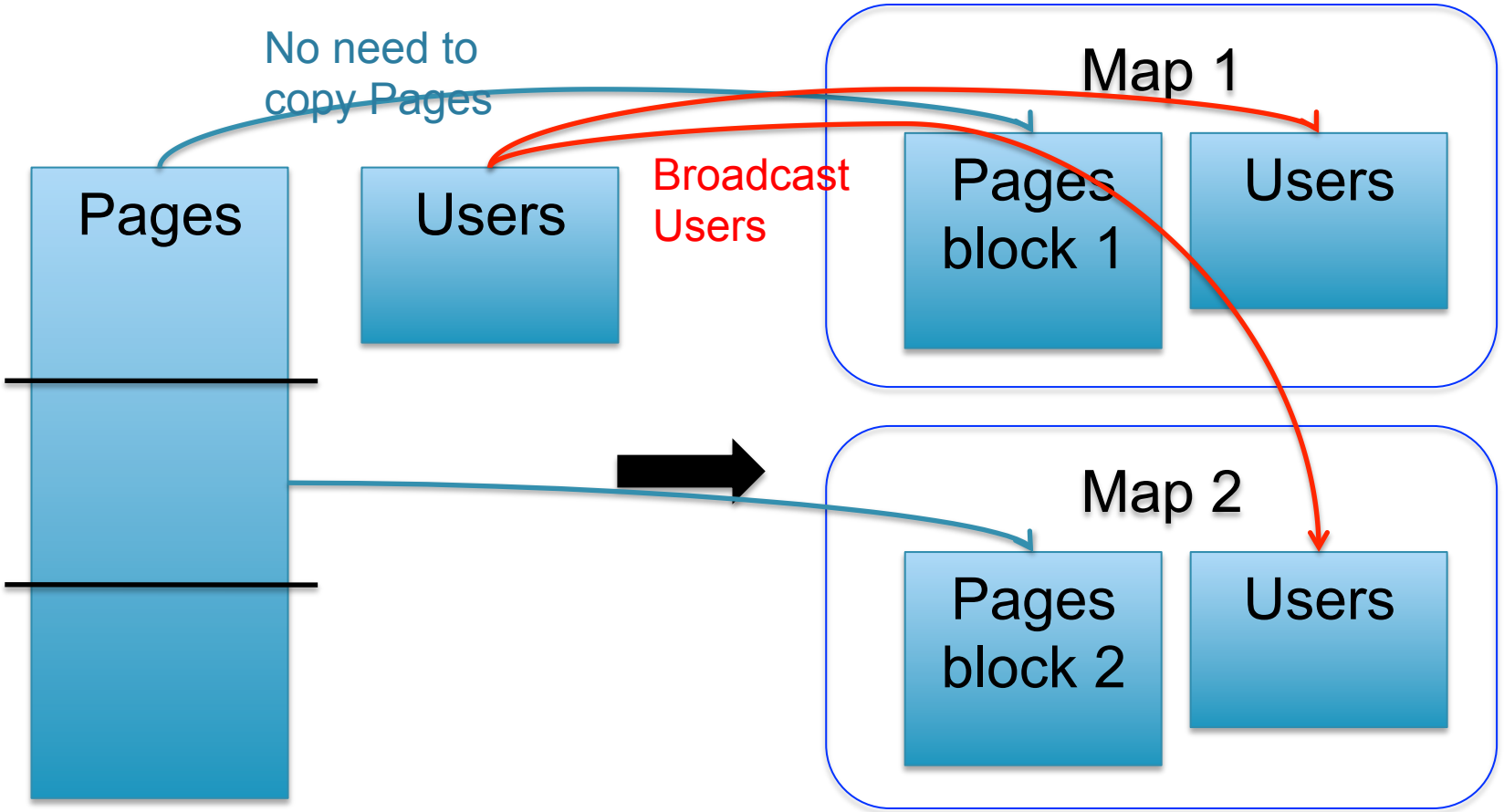

Credit: Alan Gates, Yahoo!

Parallel DBs v.s. MapReduce

Parallel DB

- Plusses

- Minuses

MapReduce

- Minuses

- Plusses

Parallel DBs v.s. MapReduce

Parallel DB

- Plusses
  - Efficient binary format
  - Indexes, physical tuning
  - Cost-based optimization
- Minuses
  - Difficult to import data
  - Lots of baggage: logging, transactions

MapReduce

- Minuses
  - Lots of time spent parsing!
  - Text files
  - “Optimizers is between your eyes and your keyboard”
- Plusses
  - Any data
  - Lightweight, easy to speedup
  - Arguably more scalable