

Delay Scheduling

A Simple Technique for Achieving Locality
and Fairness in Cluster Scheduling

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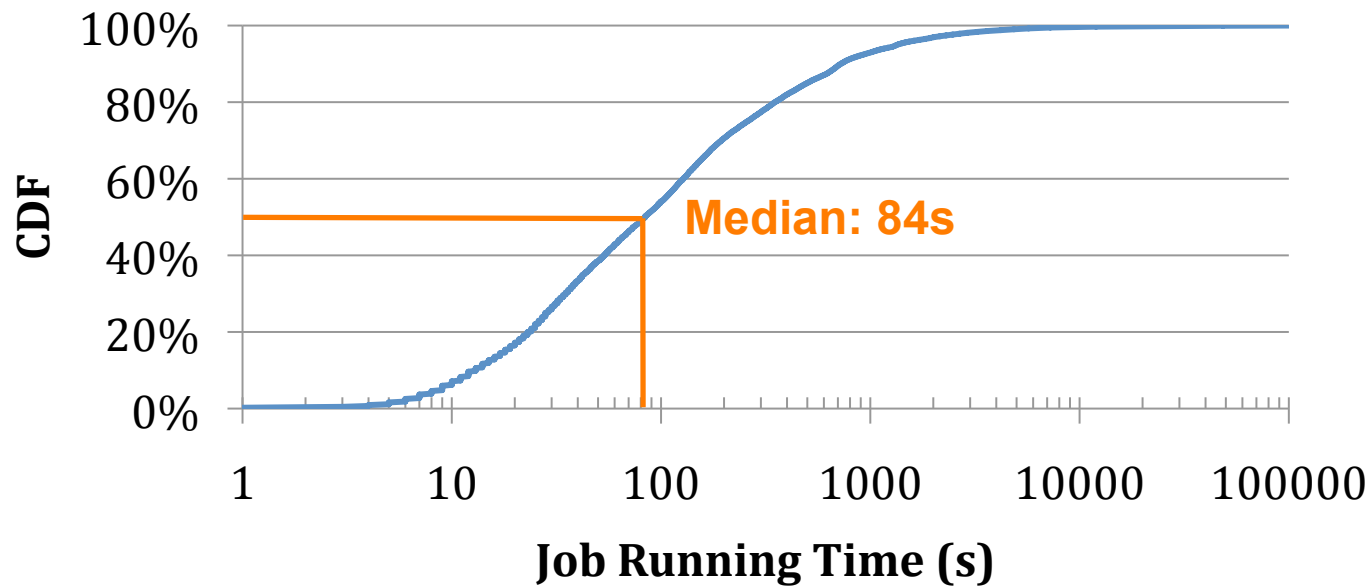


Motivation

- MapReduce / Hadoop originally designed for high throughput batch processing
 - Today's workload is far more diverse:
 - *Many users* want to share a cluster
 - Engineering, marketing, business intelligence, etc
 - Vast majority of jobs are *short*
 - Ad-hoc queries, sampling, periodic reports
 - *Response time* is critical
 - Interactive queries, deadline-driven reports
- How can we efficiently share MapReduce clusters between users?

Example: Hadoop at Facebook

- 600-node, 2 PB data warehouse, growing at 15 TB/day
- Applications: data mining, spam detection, ads
- 200 users (half non-engineers)
- 7500 MapReduce jobs / day



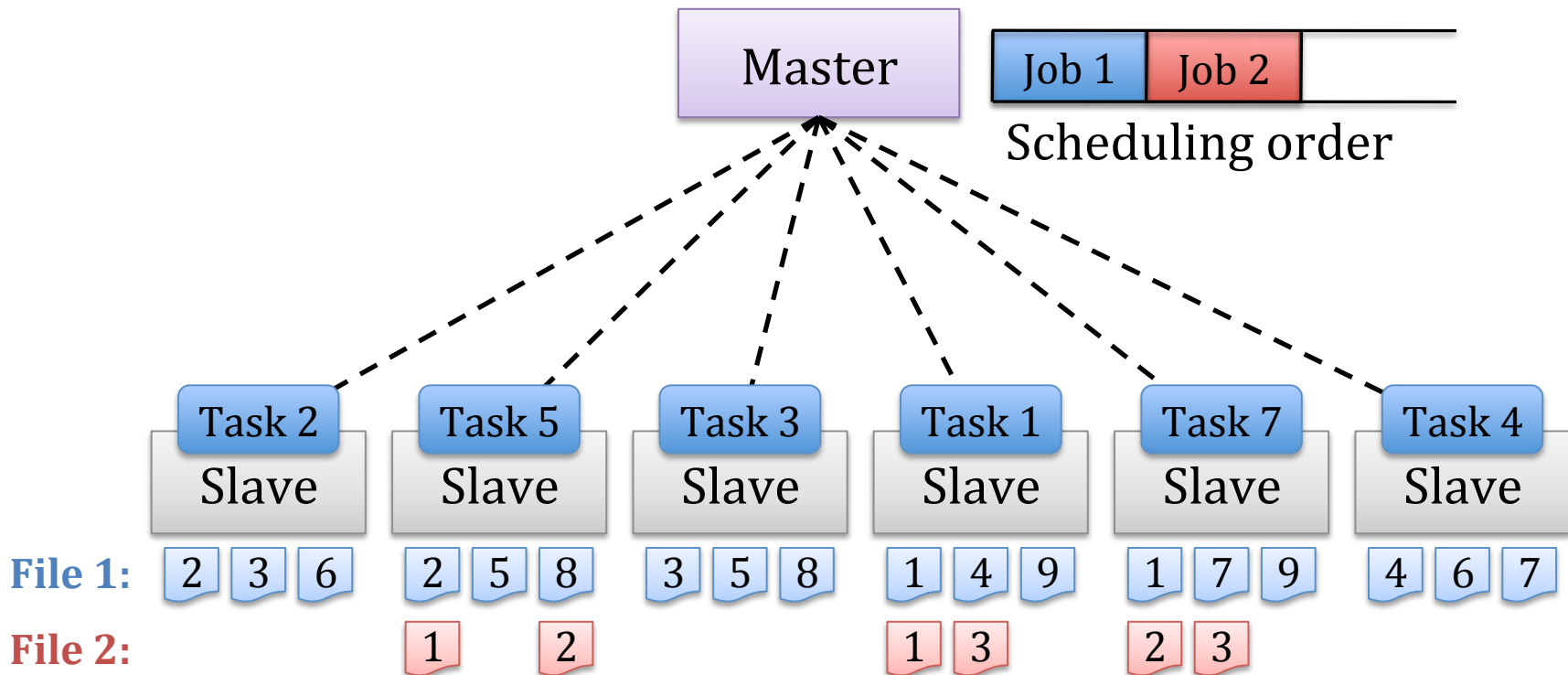
Approaches to Sharing

- Hadoop default scheduler (FIFO)
 - **Problem:** short jobs get stuck behind long ones
- Separate clusters
 - **Problem 1:** poor utilization
 - **Problem 2:** costly data replication
 - Full replication across clusters nearly infeasible at Facebook/Yahoo! scale
 - Partial replication prevents cross-dataset queries

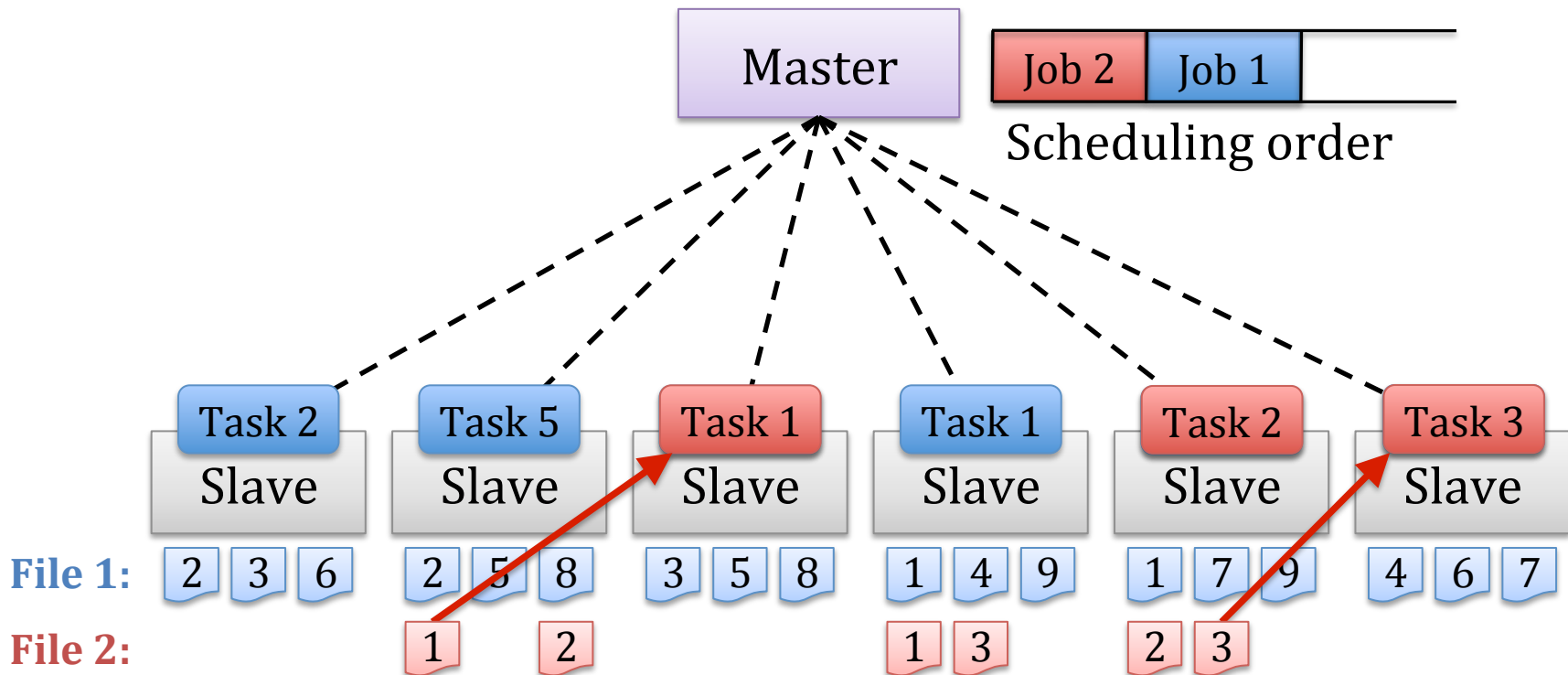
Our Work

- Hadoop Fair Scheduler
 - Fine-grained sharing at level of map & reduce tasks
 - Predictable response times and user isolation
- **Main challenge:** data locality
 - For efficiency, must run tasks near their input data
 - Strictly following *any* job queuing policy hurts locality: job picked by policy may not have data on free nodes
- **Solution:** delay scheduling
 - Relax queuing policy for limited time to achieve locality

The Problem



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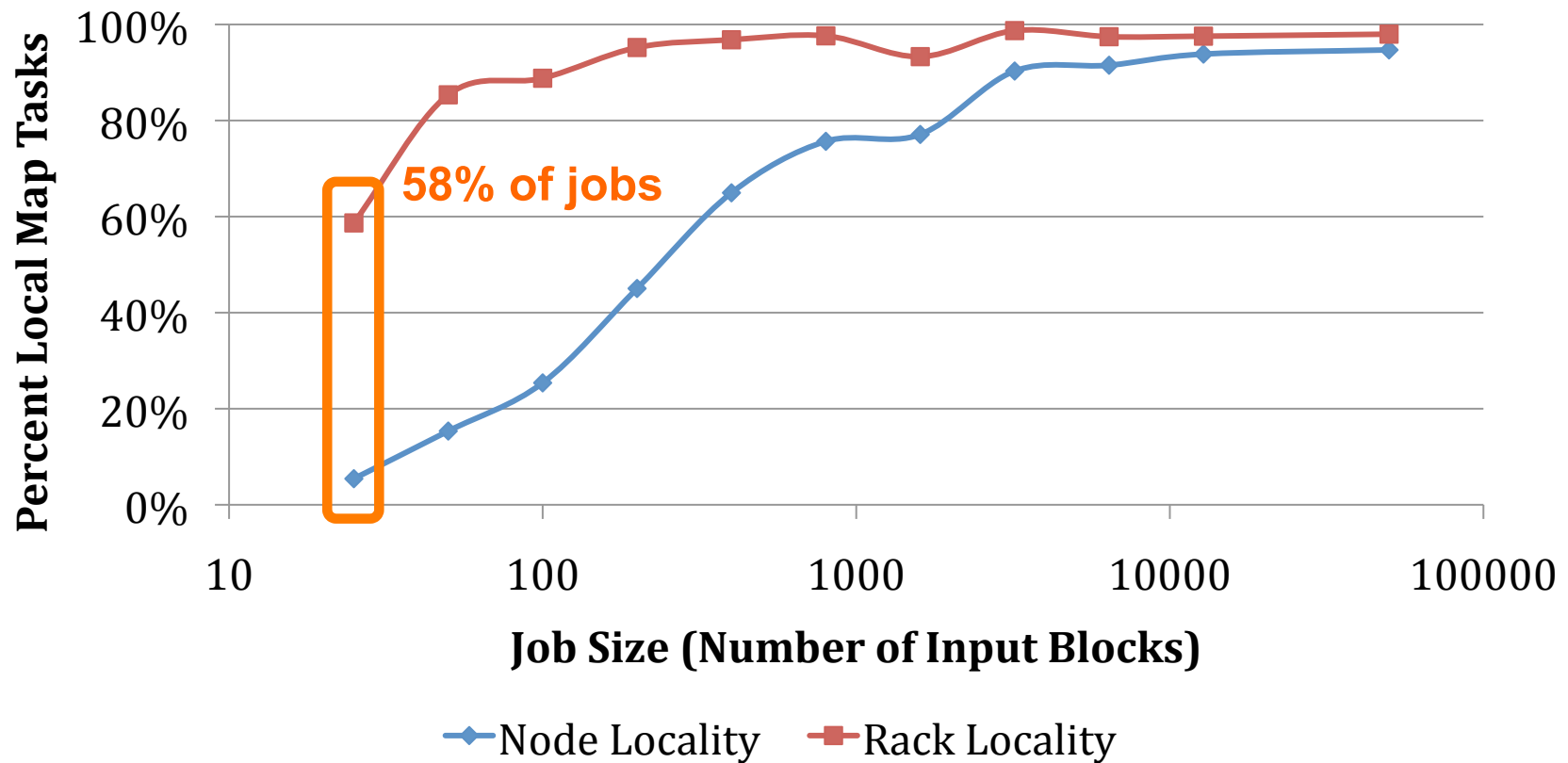


Problem: Fair decision hurts locality

Especially bad for jobs with small input files

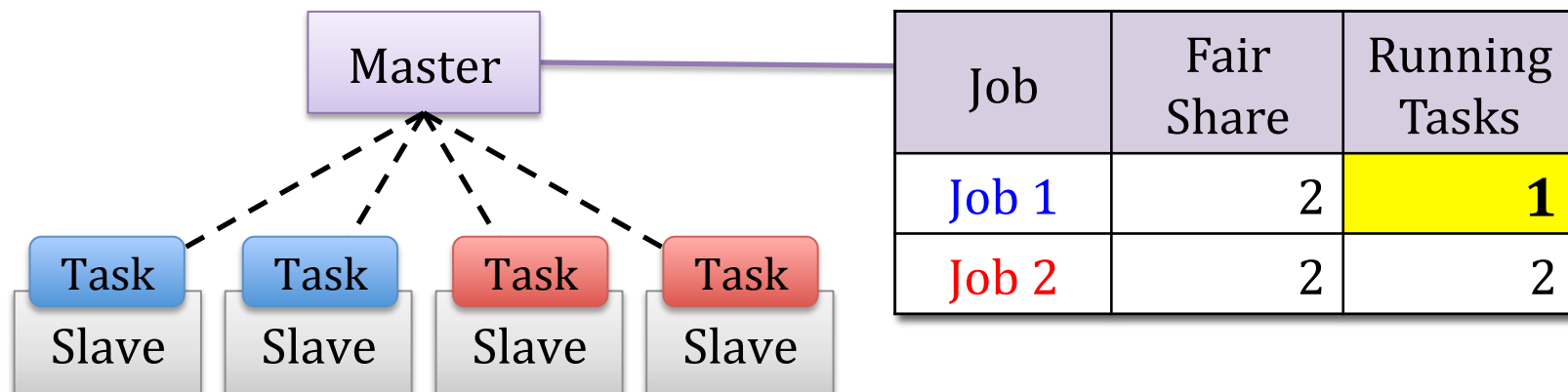
Locality vs. Job Size at Facebook

Data Locality in Production at Facebook



Special Instance: Sticky Slots

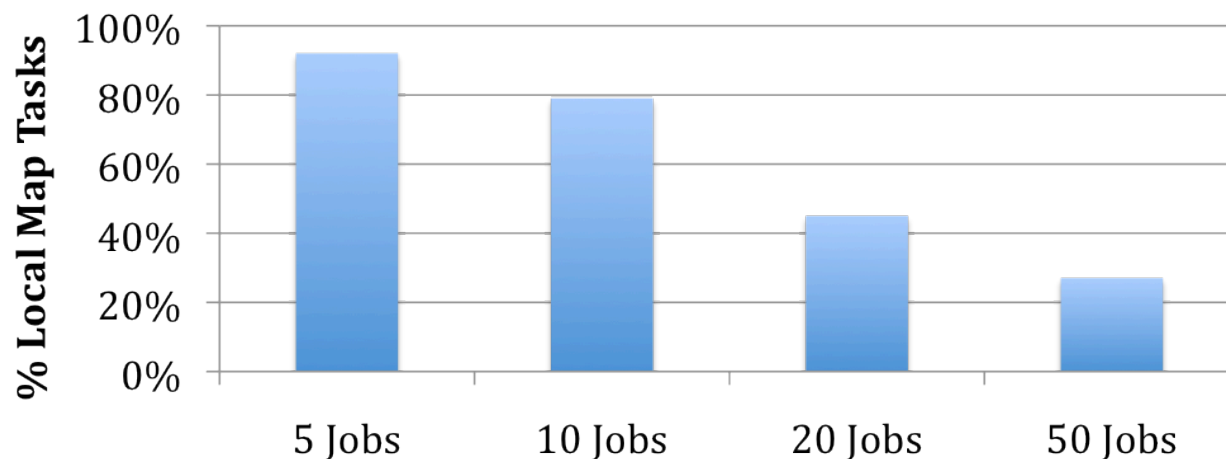
- Under fair sharing, locality can be poor even when all jobs have large input files
- Problem: jobs get “stuck” in the same set of task slots
 - When one a task in job j finishes, the slot it was running in is given back to j , because j is below its share
 - Bad because data files are spread out across *all* nodes



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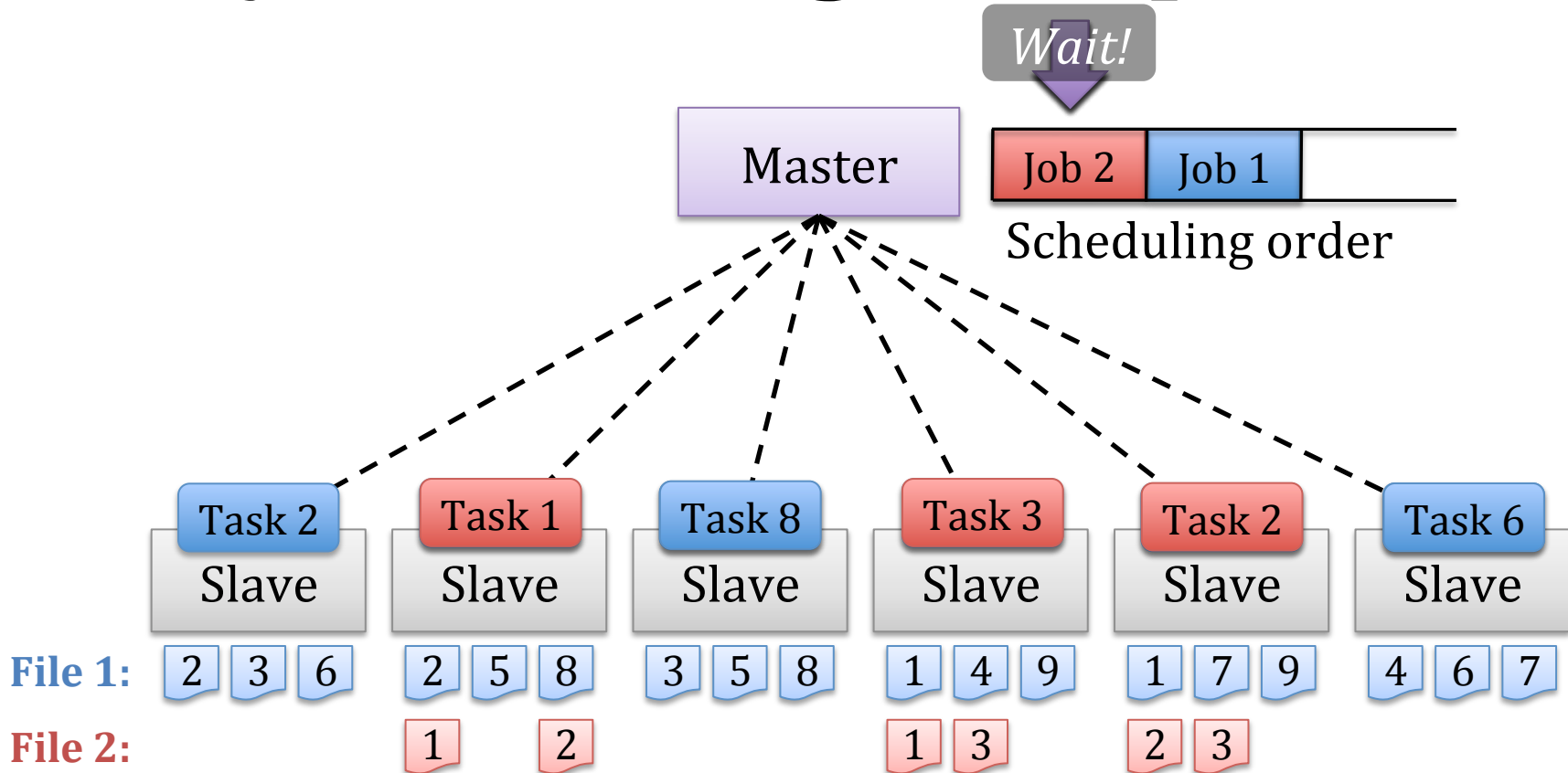
Data Locality vs. Number of Concurrent Jobs



Solution: Delay Scheduling

- *Relax queuing policy* to make jobs wait for a limited time if they cannot launch local tasks
- Result: Very short wait time (1-5s) is enough to get nearly 100% locality

Delay Scheduling Example



Idea: Wait a short time to get data-local scheduling opportunities

Delay Scheduling Details

- Scan jobs in order given by queuing policy, picking first that is permitted to launch a task
- Jobs must wait before being permitted to launch non-local tasks
 - If $\text{wait} < T_1$, only allow node-local tasks
 - If $T_1 < \text{wait} < T_2$, also allow rack-local
 - If $\text{wait} > T_2$, also allow off-rack
- Increase a job's time waited when it is skipped

Analysis

- **When is it worth it to wait, and for how long?**
 - *Waiting worth it if $E(\text{wait time}) < \text{cost to run non-locally}$*
- Simple model: Data-local scheduling opportunities arrive following Poisson process
- Expected response time gain from delay scheduling:

$$E(\text{gain}) = (1 - e^{-w/t})(d - t)$$

Wait amount

Expected time to get local scheduling opportunity

Delay from running non-locally

- *Optimal wait time is infinity*

Analysis Continued

$$E(\text{gain}) = (1 - e^{-w/t})(d - t)$$

Wait amount

Expected time to get local scheduling opportunity

Delay from running non-locally

- **Typical values of t and d :**

$$t = (\text{avg task length}) / (\underbrace{\text{file replication}}_3 \times \underbrace{\text{tasks per node}}_6)$$
$$d \geq (\text{avg task length})$$

More Analysis

- What if some nodes are occupied by long tasks?
- Suppose we have:
 - Fraction ϕ of active tasks are long
 - R replicas of each block
 - L task slots per node
- For a given data block,
P(all replicas blocked by long tasks) = ϕ^{RL}
- With $R = 3$ and $L = 6$, this is less than 2% if $\phi < 0.8$

Evaluation

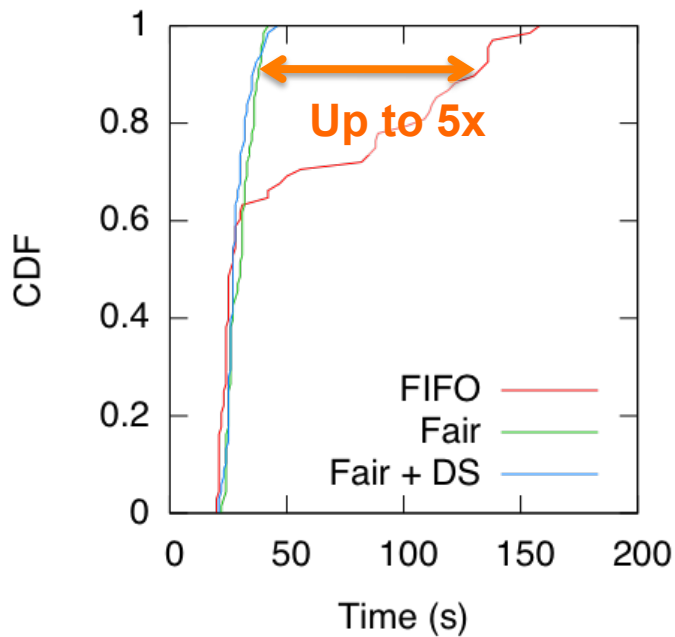
- **Macrobenchmark**
 - IO-heavy workload
 - CPU-heavy workload
 - Mixed workload
- **Microbenchmarks**
 - Sticky slots
 - Small jobs
 - Hierarchical fair scheduling
- Sensitivity analysis
- Scheduler overhead

Macrobenchmark

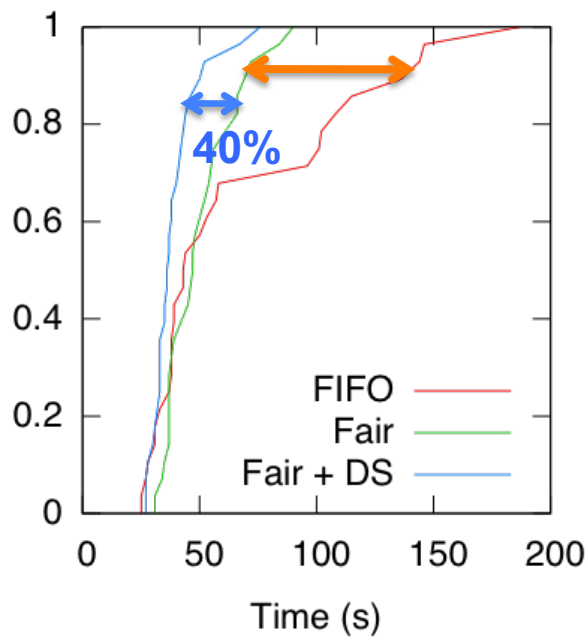
- 100-node EC2 cluster, 4 cores/node
- Job submission schedule based on job sizes and inter-arrival times at Facebook
 - 100 jobs grouped into 9 “bins” of sizes
- Three workloads:
 - IO-heavy, CPU-heavy, mixed
- Three schedulers:
 - FIFO
 - Fair sharing
 - Fair + delay scheduling (wait time = 5s)

Results for IO-Heavy Workload

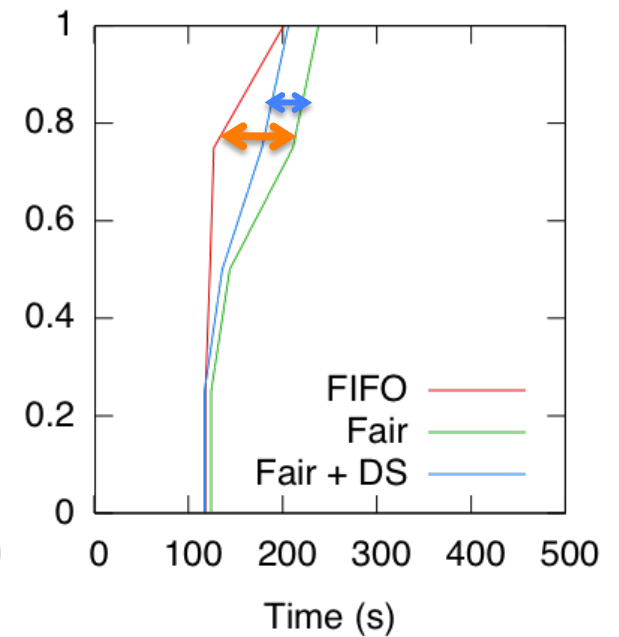
Job Response Times



Small Jobs
(1-10 input blocks)

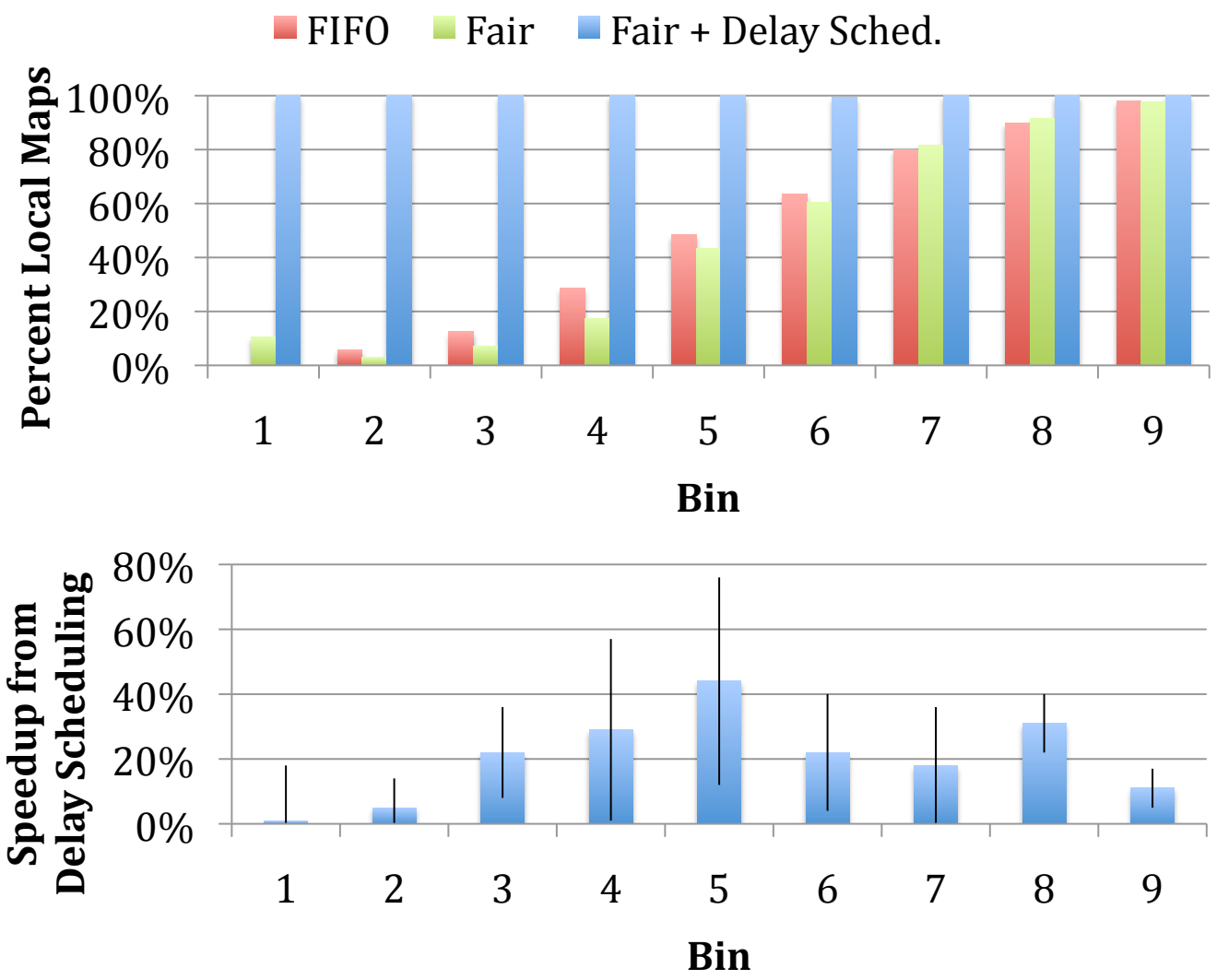


Medium Jobs
(50-800 input blocks)



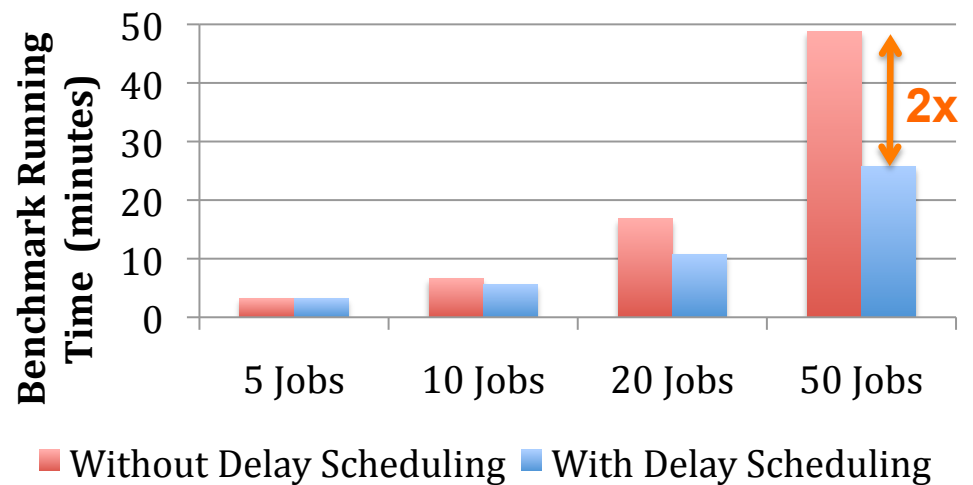
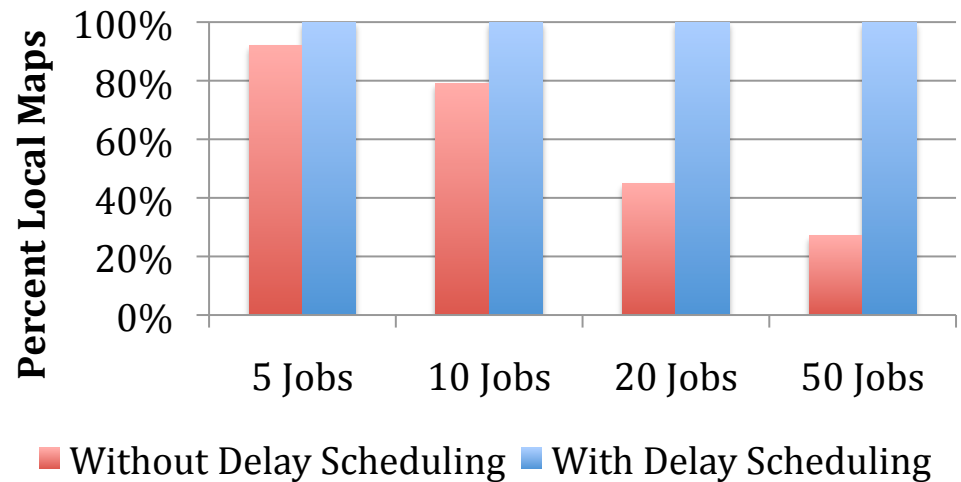
Large Jobs
(4800 input blocks)

Results for IO-Heavy Workload



Sticky Slots Microbenchmark

- 5-50 jobs on EC2
- 100-node cluster
- 4 cores / node
- 5s delay scheduling



Conclusions

- Delay scheduling works well under two conditions:
 - Sufficient fraction of tasks are *short* relative to jobs
 - There are *many locations* where a task can run efficiently
 - Blocks replicated across nodes, multiple tasks/node
- Generalizes beyond MapReduce and fair sharing
 - Applies to any queuing policy that gives ordered job list
 - Used in Hadoop Fair Scheduler to implement hierarchical policy
 - Applies to placement needs other than data locality

Thanks!

- This work is open source, packaged with Hadoop as the Hadoop Fair Scheduler
 - Old version without delay scheduling in use at Facebook since fall 2008
 - Delay scheduling will appear in Hadoop 0.21
- My email: matei@eecs.berkeley.edu

Related Work

- Quincy (SOSP '09)
 - Locality aware fair scheduler for Dryad
 - Expresses scheduling problem as min-cost flow
 - Kills tasks to reassign nodes when jobs change
 - One task slot per machine
- HPC scheduling
 - Inelastic jobs (cannot scale up/down over time)
 - Data locality generally not considered
- Grid computing
 - Focused on common interfaces for accessing compute resources in multiple administrative domains

Delay Scheduling Details

when there is a free task slot on node n :

sort $jobs$ according to queuing policy

for j in $jobs$:

if j has node-local task t on n :

$j.level := 0; j.wait := 0; \mathbf{return} t$

else if j has rack-local task t on n and $(j.level \geq 1$ or $j.wait \geq T_1)$:

$j.level := 1; j.wait := 0; \mathbf{return} t$

else if $j.level = 2$ or $(j.level = 1$ and $j.wait \geq T_2)$

 or $(j.level = 0$ and $j.wait \geq T_1 + T_2)$:

$j.level := 2; j.wait := 0; \mathbf{return} t$

else:

$j.wait +=$ time since last scheduling decision

Sensitivity Analysis

