Blind Men and the Elephant: Piecing Together Hadoop for Diagnosis

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Outline

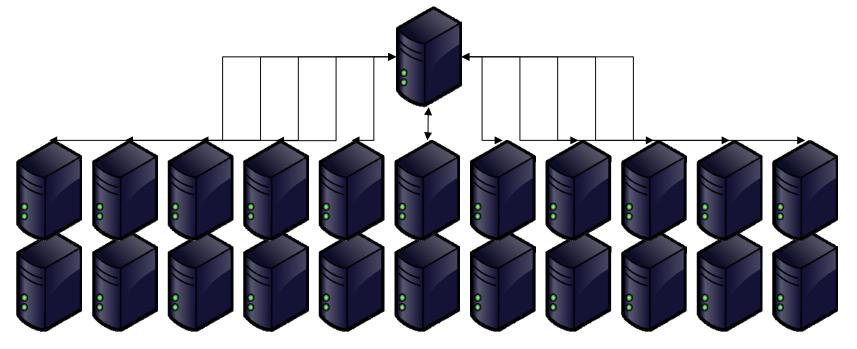
- Background
- Blimey Approach
- Instrumentation and Algorithms
- Evaluation and Results
- Conclusions

Outline

- Background
 - MapReduce and Hadoop
 - Problem Statement
- Blimey Approach
- Instrumentation and Algorithms
- Evaluation and Results
- Conclusions

Background

- MapReduce & Hadoop
 - Increasingly popular programming framework
 - Distributed, data-intensive, parallel applications
 - Cluster of machines: single master, many slaves
 - Many Map, Reduce tasks executing same code on different input data

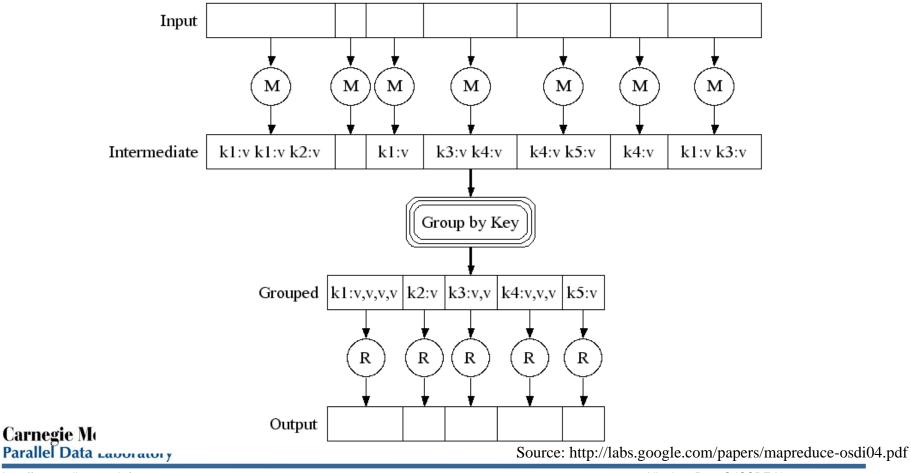


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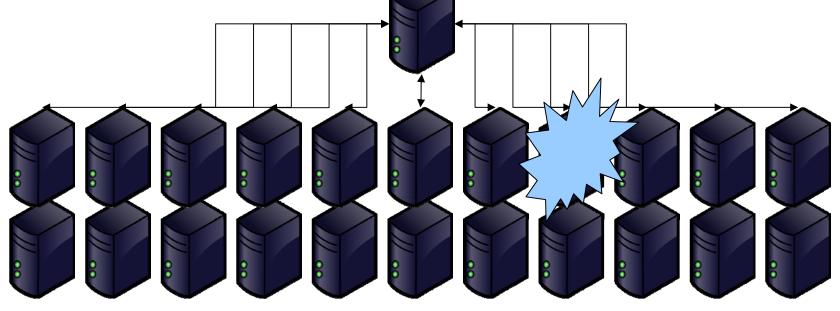
MapReduce Execution

- Inputs split into blocks
- M Map tasks; R Reduce tasks



Problem

- What to do when a failure occurs?
 - Need to quickly fix faults, but..
 - Large scale too much debugging information
 - Distributed nature difficult for human sysadmin to reason about problem
- IDEA: large distributed system also affords many points of observation!



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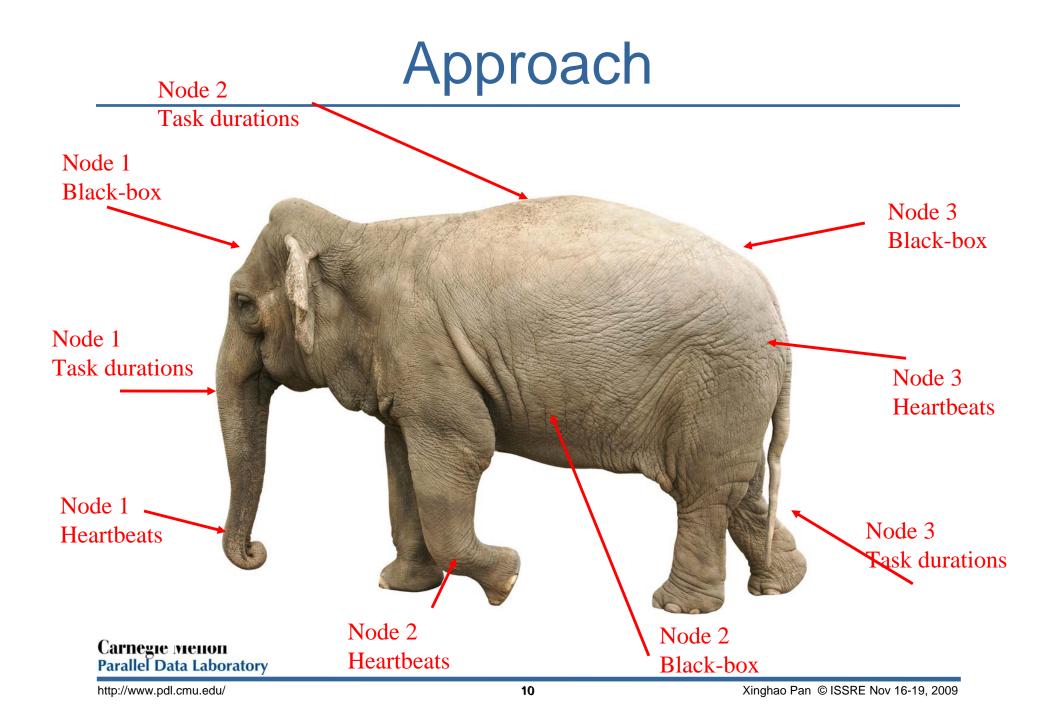
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Blimey: Blind Men and Elephant

- Diagnosis framework
- Mythological story Blind Men and the Elephant
 - Each blind man has a limited concept of the elephant
 - No blind man has complete information
 - But putting together views provide the complete picture
- Blimey
 - "Elephant": MapReduce system
 - "Blind men": Multiple perspectives into MapReduce system
- **Corroborate** and **synthesize** multiple distinct perspectives to *localize* and *identify* the fault

Blimey

- Blimey approach applied at two levels
- Corroborating instrumentation points
 - "Blind men": Instrumented data (e.g. OS metrics, logs) from the many cluster nodes serve as view into system
 - Diagnostic algorithms corroborate data to *localize* potentially faulty slave nodes
- Synthesizing outcomes from diagnostic algorithms
 - "Blind men": Diagnostic algorithms serve as secondary perspectives into system
 - We **synthesize** the secondary perspectives to *identify* the fault



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- Background
- Blimey Approach
- Instrumentation and Algorithms
 - Instrumentation: Black-Box
 - Instrumentation: White-Box
 - Instrumentation: Heartbeats
 - Synthesizing Algorithms' Outcomes
- Evaluation and Results
- Conclusions

Instrumentation: Black-Box

- Black-box: external view of application
 - OS-level performance counters collected from /proc
 - Represents instantaneous behaviour of slave node
- **IDEA:** slave nodes execute same code on blocks of input • data; slave nodes should exhibit similar behaviors
- Details of diagnostic algorithms in [ISSRE09], [HotMetrics09] ۲

	Metric	Description
	user	% CPU time in user-space
	system	% CPU time in kernel-space
	iowait	% CPU time waiting for I/O
	ctxt	Context switches per second
	runq-sz	# processes waiting to run
	plist-sz	Total # of processes and threads
	ldavg-1	system load average for the last minute
	bread	Total bytes read from disk /s
	bwrtn	Total bytes written to disk /s
	eth-rxbyt	Network bytes received /s
	eth-txbyt	Network bytes transmitted /s
	pgpgin	KBytes paged in from disk /s
	pgpgout	KBytes paged out to disk /s
	fault	Page faults (major+minor) /s
rnegie Mellon	TCPAbortOnData	# of TCP connections aborted with data in queue
rallel Data Laboratory	rto-max	Maximum TCP retransmission timeout

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Black-Box Diagnosis

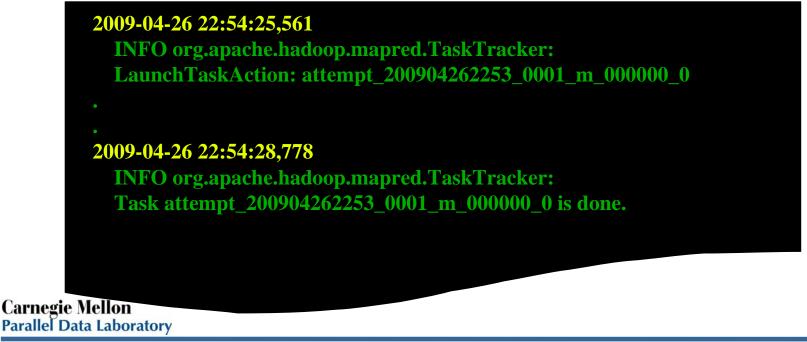
- Intuition
 - Slave nodes execute similar tasks within window of time
 - Slave nodes encountered similar workload, from black box point of view
 - Under faulty conditions, slave nodes exhibit divergent behaviour
 - Corroborate black-box metrics for diagnosis
- Details of diagnostic algorithms in [ISSRE09], [HotMetrics09]

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Instrumentation: White-Box

- White-box (Task-centric)
 - Execution states extracted from application logs
 - Of interest: Map and Reduce durations
- IDEA: Map/Reduce tasks execute same codes; Map/Reduce tasks should have similar durations
- Details of diagnostic algorithm in [ISSRE09], [WASL08]

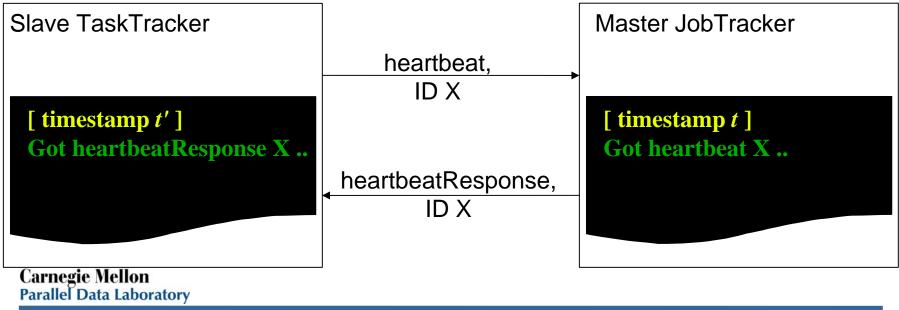


White-Box Diagnosis

- Intuition
 - Similarity of Map/Reduce tasks
 - Slave node execute subset of global set of tasks
 - Under faulty conditions, tasks on different slaves have different durations
 - Corroborate durations (as property of tasks) across slave nodes
- Details of diagnostic algorithm in [ISSRE09], [WASL08]

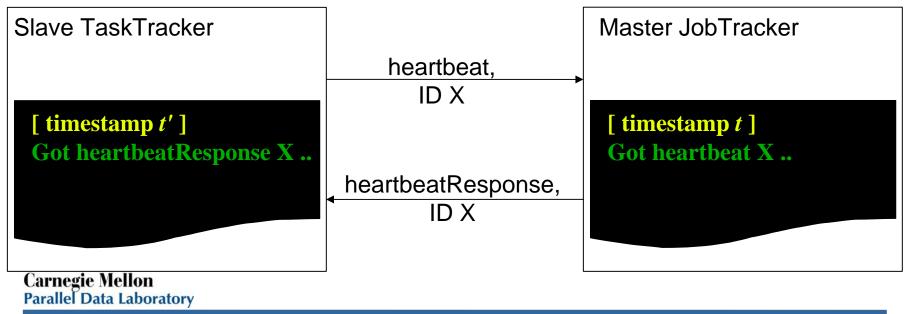
Instrumentation: Heartbeats

- Heartbeats
 - Keep-alive messages initiated by slave TaskTracker to master JobTracker
 - Sent periodically and upon task completion



Instrumentation: Heartbeats

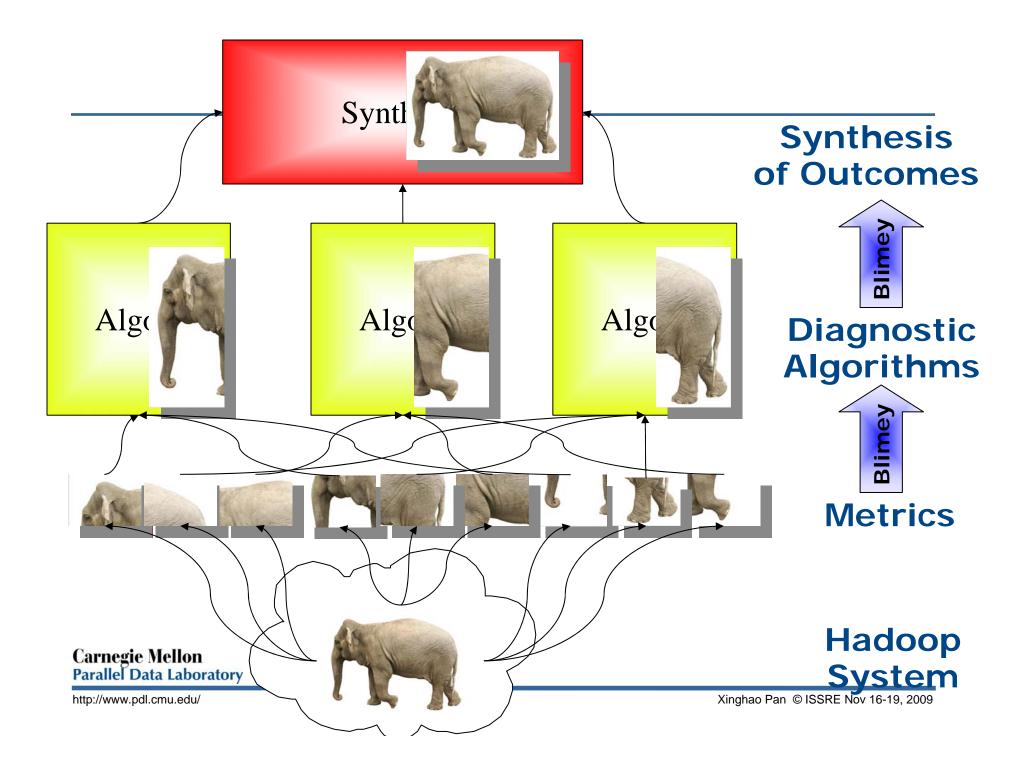
- IDEA 1: Rate of heartbeats is a coarse indicator for workload; slave nodes should have similar workloads and heartbeat rates
- IDEA 2: Timestamp difference between log messages should be constant
- Details of diagnostic algorithms in [ISSRE09]



Synthesizing Algorithms' Outcomes

- Different faults manifest differently on different metrics (e.g. CPU utilization, Map task duration, heartbeat rates, ...)
 - Fault may or may not manifest on particular metrics
 - Fault may or may not manifest in correlated fashion on particular metrics
- Different diagnostic algorithms operate on different metrics
- IDEA: Synthesize algorithms' outcomes as secondary perspectives into Hadoop system
 - Build decision tree to classify faults using outcomes from diagnostic algorithms
- Details of diagnostic algorithms in [ISSRE09]

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Evaluation: Testbed & Workload

- Inject faults into Hadoop cluster
 - Fault injected on single node in cluster
 - Test accuracy of localizing and identifying faults
- Testbed
 - Hadoop 0.18.3
 - 10- and 50-slave nodes clusters
 - Amazon's EC2, Large instances
 - 7.5 GB RAM, two dual-core CPUs, amd64 Debian/GNU Linux 4.0

• Gridmix

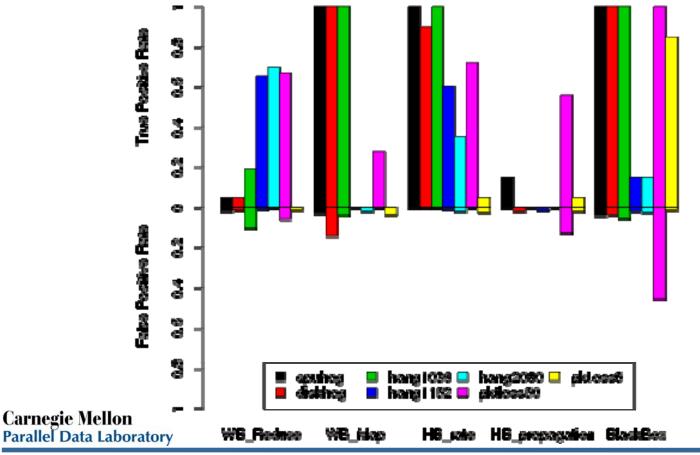
- Well-accepted, multi-workload Hadoop benchmark
- Mimics observed data-access patterns in actual user jobs in enterprise deployments

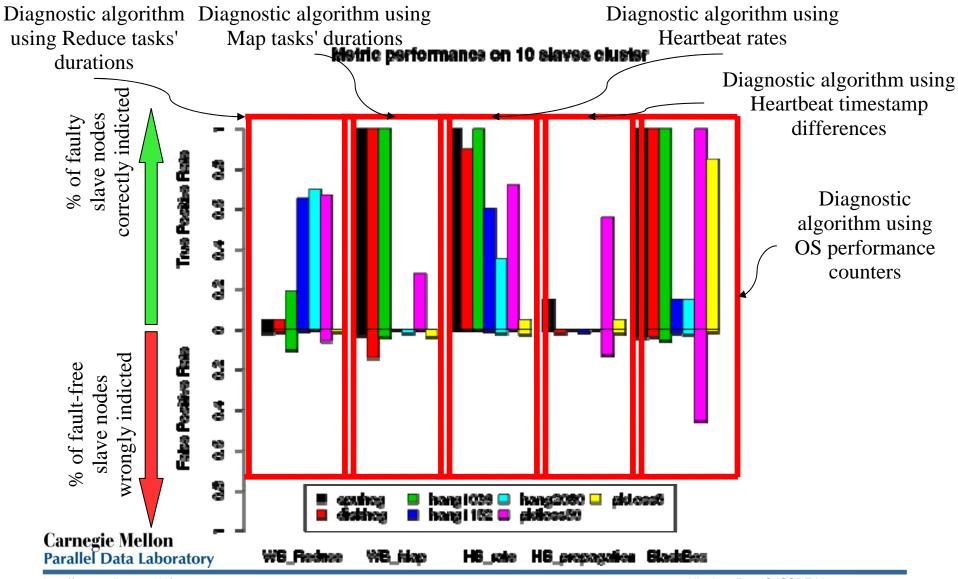
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Evaluation: Fault Injection

	Fault	Description
Fault free	Control	Control experiments with no injected faults
Resource contention	CPU hog	External process uses 70% of CPU
	Packet-loss	Drop 5% or 50% of incoming packets
	Disk hog	Repeatedly write 20GB file
Application bugs	HADOOP-1036	Maps hang due to unhandled exception
Source: Hadoop JIRA	HADOOP-1152	Reduces fail while copying map output
	HADOOP-2080	Reduces fail due to incorrect checksum
	HADOOP-2051	Jobs hang due to unhandled exception

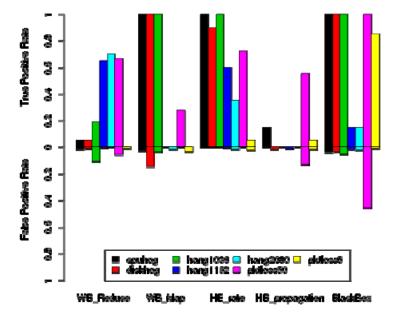
Metric performance on 10 slaves cluster



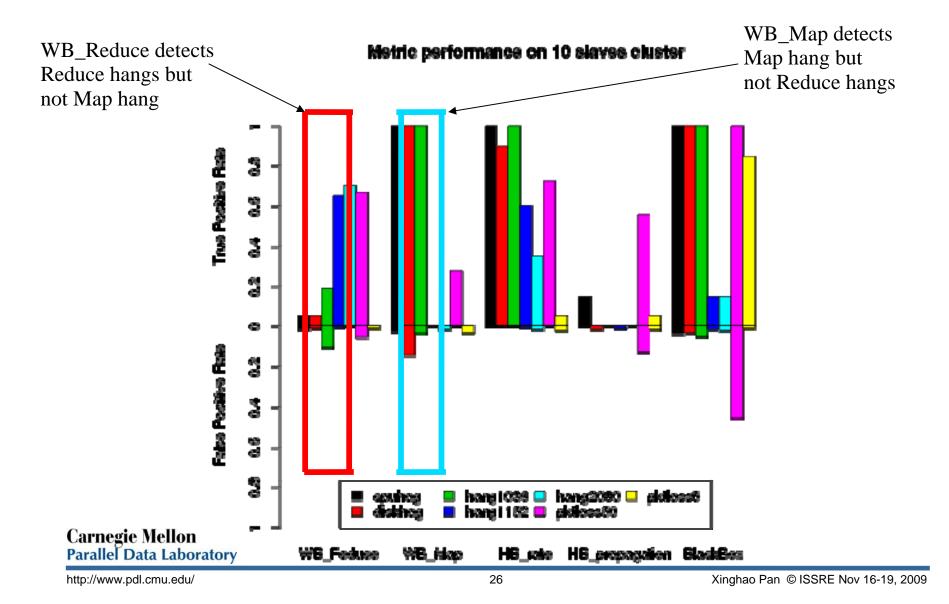


- Every fault is accurately localized by some algorithm =>
 - High true positive rate
 - Low false positive rate
- No single algorithm is perfect at localizing all faults
- Faults manifest differently
 - May or may not manifest
 - Varying degrees of correlated manifestation

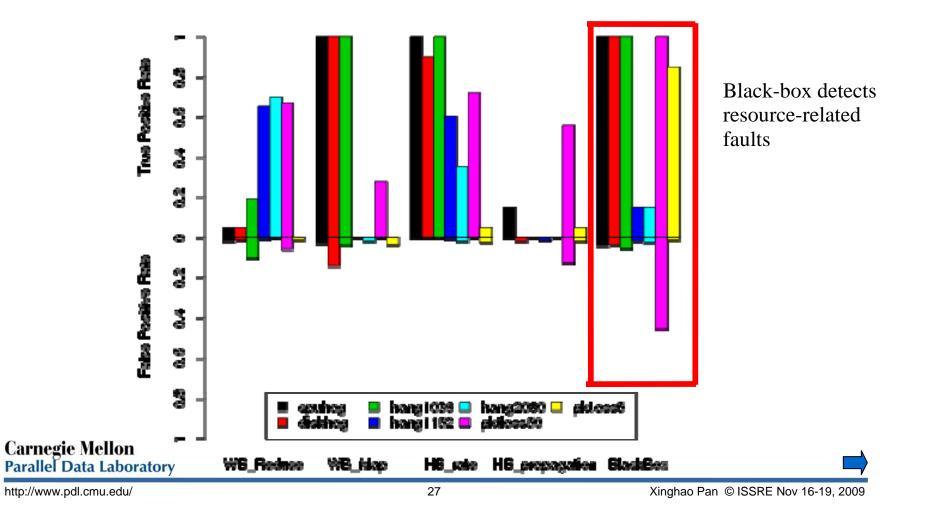
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Metric performance on 10 slaves cluster



Metric performance on 10 slaves cluster

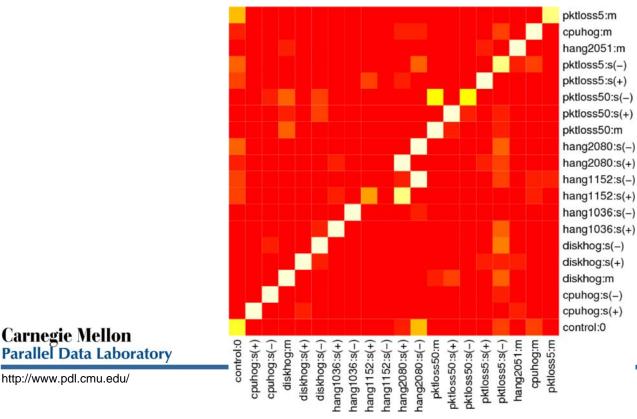


Results: Synthesis

- Confusion matrix: accuracy of fault classification
 - Rows represent true class

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- Columns represent label generated by decision tree
- Light shade means true fault likely to be given particular label
- Ideal: light diagonals, dark everywhere else



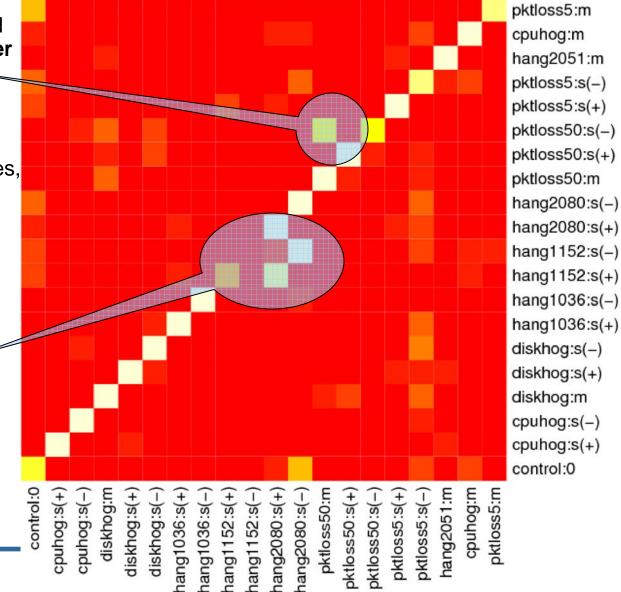
Results: Synthesis

Fault-free slave node, affected by 50% packet loss on **another slave node**, is confused as fault-free slave node, affected by 50% packet loss on the **master node**.

Both are massive packet losses, but not at the node itself.

hang1152 as hang2080 Both are Reduce hangs; location not confused.

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 - Future Work
 - Summary

Future Work

- Other instrumentation sources
- Visualization of diagnosis outcomes and data
- Fine-grain diagnosis: why and how
 - Root-cause analysis

Summary / Conclusion

- Identified multiple instrumentation points in Hadoop that can be corroborated for diagnosis
- Blimey provides a framework for failure diagnosis in MapReduce systems
 - Diagnostic algorithms corroborate multiple viewpoints to indict faulty slave nodes
 - Synthesizing secondary perspectives from algorithms, we differentiate and identify faults

Related Material

- [ISSRE09] X.Pan, J. Tan, S. Kavulya, R. Gandhi, and P. Narasimhan. Blind Men and the Elephant: Piecing Together Hadoop for Diagnosis, in 20th International Symposium on Software Reliability Engineering (ISSRE), Nov 2009.
- [HotMetrics09] X. Pan, J. Tan, S. Kavulya, R. Gandhi, and P. Narasimhan. Ganesha: Black-Box Diagnosis of MapReduce Systems, in Second Workshop on Hot Topics in Measurement and Modeling of Computer Systems, June 2009.
- [CMU-CS-09-135] X. Pan. Blind Men and the Elephant: Piecing Together Hadoop for Diagnosis. Master's Thesis, Carnegie Mellon University, 2009. Technical Report: CMU-CS-09-135. May 2009.
- [WASL08] J. Tan, X. Pan, S. Kavulya, R. Gandhi, P. Narasimhan. SALSA: Analyzing Logs as StAte Machines. First USENIX Workshop on Analysis of System Logs (WASL), San Diego, CA, Dec 2008.

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Target System: MapReduce and Hadoop

- MapReduce
 - Framework for distributed, parallel programming
 - Job = multiple copies of maps and reduces: Executed on different segments of large dataset
- Hadoop: open-source implementation
 - Distributed Runtime + Distributed FileSystem
 - Master/Slave: one Master, multiple Slaves

