



JUMMP: Job Uninterrupted Maneuverable MapReduce Platform

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Motivation [1 of 2]



- Hadoop is the de-facto implementation of the MapReduce programming model and used in many different disciplines
- Academic shared environment clusters accommodate a wide variety of research applications within a set of financial, technical, and administrative constraints.
- The Job Uninterrupted Maneuverable MapReduce Platform is an automated scheduling platform
 - Enables the integration of Hadoop into the existing large scale computational infrastructure
 - Supports high availability and continuous computing for research and education



Motivation [2 of 2]



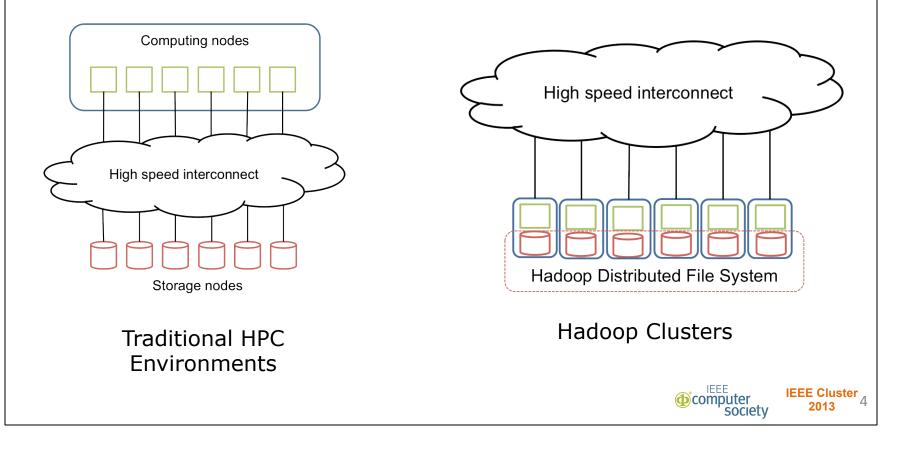
- User Considerations
 - Research that uses Hadoop as a tool for the research
 - Study on the Hadoop eco-system that consider different hardware and software configurations
 - Academic course project requiring individual clusters



Motivation [2 of 2]



- Environmental Considerations
 - Cost of hardware, space, power and cooling
 - Technology differences of Hadoop and traditional HPC configurations
 - Support of the Hadoop runtime environment



Design Objectives



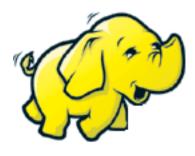
- Provision individual Hadoop clusters within an academic HPC environment:
 - As dynamic execution environment on demand by users
 - In user space without root privileges
 - Minimal interactions from system administrators
 - Exist with environment beyond reservation limitations in semi-persistent manner



Hadoop Overview [1 of 3]



- Hadoop is Apache Foundation's open-source implementation Google projects and research
- Hadoop Distributed File System (HDFS)
 - Implementation of the Google File System
 - Highly fault-tolerant distributed file system
- MapReduce Framework
 - Implementation of Google MapReduce
 - Application framework for process big data in-parallel across clusters

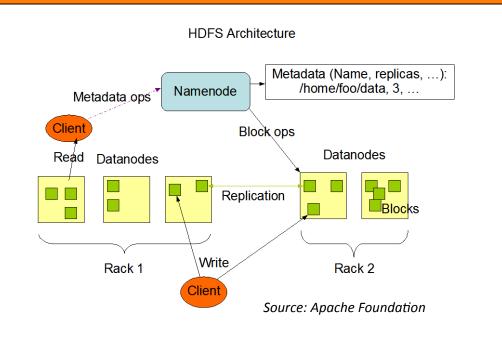




Hadoop Overview [2 of 3]



- HDFS Nodes:
 NameNode (NN)
 - Data Node (DN)
- MapReduce Nodes
 - JobTracker (JT)
 - TaskTracker (TT)



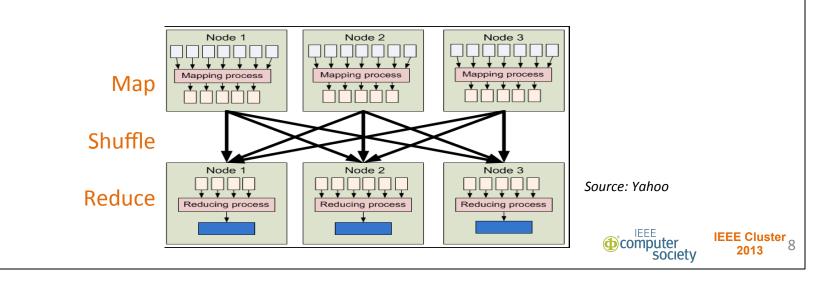
- Traditionally single NN and JT in cluster
- Multiple DNs and TTs



Hadoop Overview [3 of 3]



- Map Phase
 - Key-value pair input mapped to key-value pair outputs
- Shuffle Phase
 - Transfer of map output to reducers as input
- Reduce Phase
 - Operation on set of values for each key in map out



JUMMP Design



- JUMMP is a Hadoop cluster where DN/TT continuously move throughout HPC environment while allowing survivability of Hadoop cluster
- Configured with two variables:
 - n: number of DataNodes / TaskTracker
 - t_i: scheduled time between jumps
- Built with Palmetto HPC Cluster at Clemson
 - Uses PBS Professional Scheduler, compatible with open source versions of PBS.
- Dedicated NN/JT node outside of scheduler
- Each DN/TT reserved with its own PBS job



Jumping Node Actions



- Launch DN/TT daemons and join Hadoop Cluster
- Perform Hadoop cluster duties as normal DN/TT would
- Await trigger to "jump" (time or event based)
 - 1. Schedule replacement DN/TT PBS job
 - 2. Stop Hadoop daemons
 - 3. Decommission and blacklist itself from cluster
 - 4. End PBS job



Performance Analysis



- DataNode and TaskTracker cause distinct degradation on performance due to jumps
- DataNode jumps cause re-replication of data blocks located in the jumping node
- TaskTracker jumps cause MapReduce tasks to be rescheduled that were executing on jumping node
- TaskTracker jumps cause cluster to be "undersized" until replacement is fully operational



Experiment Design



- Pool of 96 homogenous nodes within Palmetto (shown in Table I)
- Benchmarks and Datasets from Purdue's PUMA project
- 100 runs of each experiment (shown in table II) on consistent dataset
- Baseline and three different jump times
- Recorded task times, job times, and jump times

Node	HP SL250s	
CPU	Intel Xeon E5-2665 (2)	
Cores	16	
Memory	64 GB	
Local Storage Capacity	900 GB	
Networking	Infiniband	

TABLE I: Node Configuration

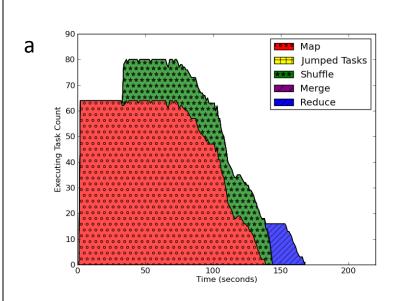
Application	Wordcount	Terasort
Dataset Size	50 GB	300 GB
Node Count	8	32
Jump Times [mins]	7/10/15	20/40/60

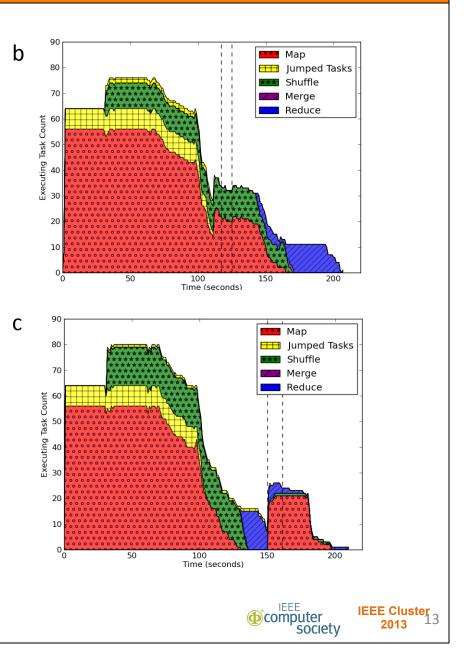
TABLE II: Experiment Parameters



Characteristics of Impact





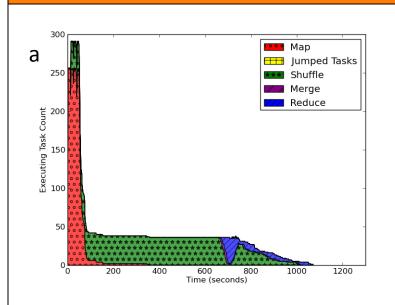


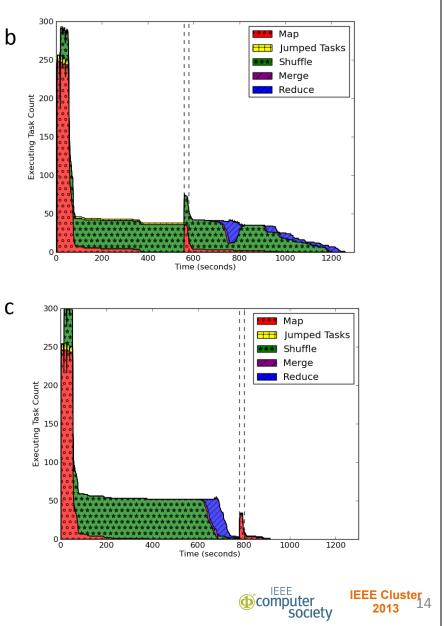
Wordcount on 50 GB with 8 nodes

(a) No jump(b) Jump during map phase(c) Jump during reduce phase

Characteristics of Impact





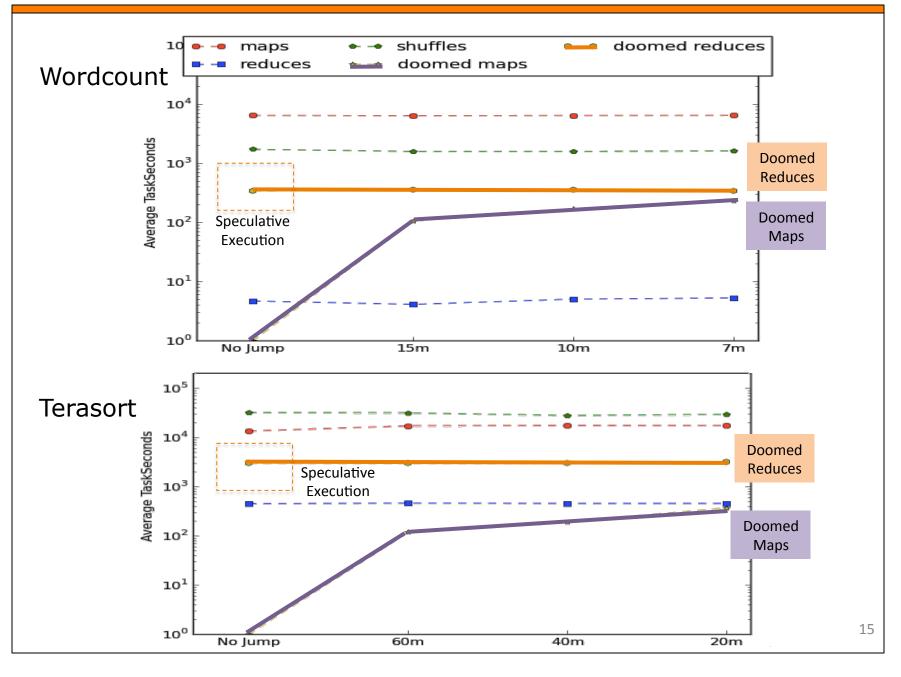


Terasort on 300 GB with 32 nodes

(a) No jump(b) Jump during shuffle phase(c) Jump during reduce phase

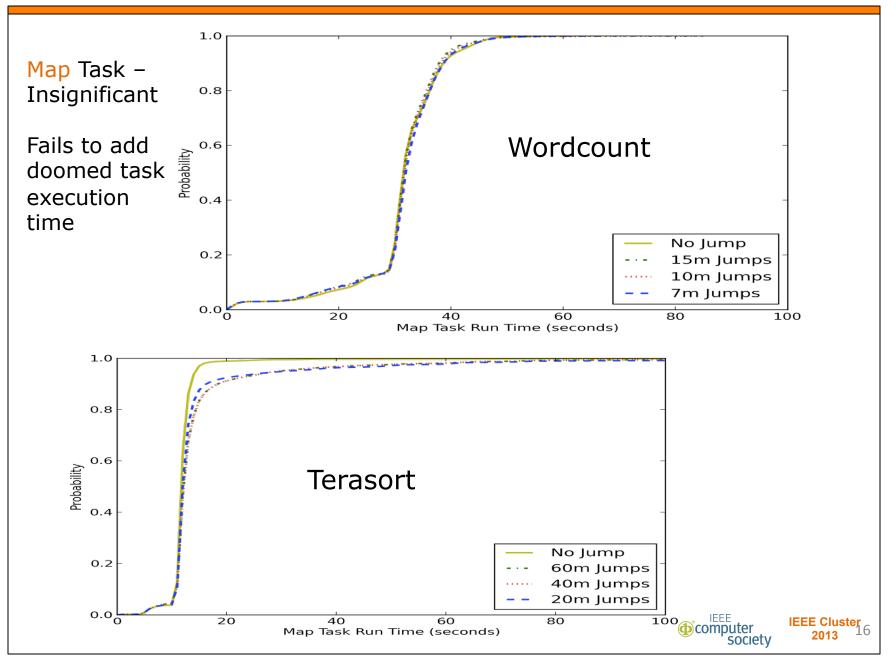
Performance Measurements





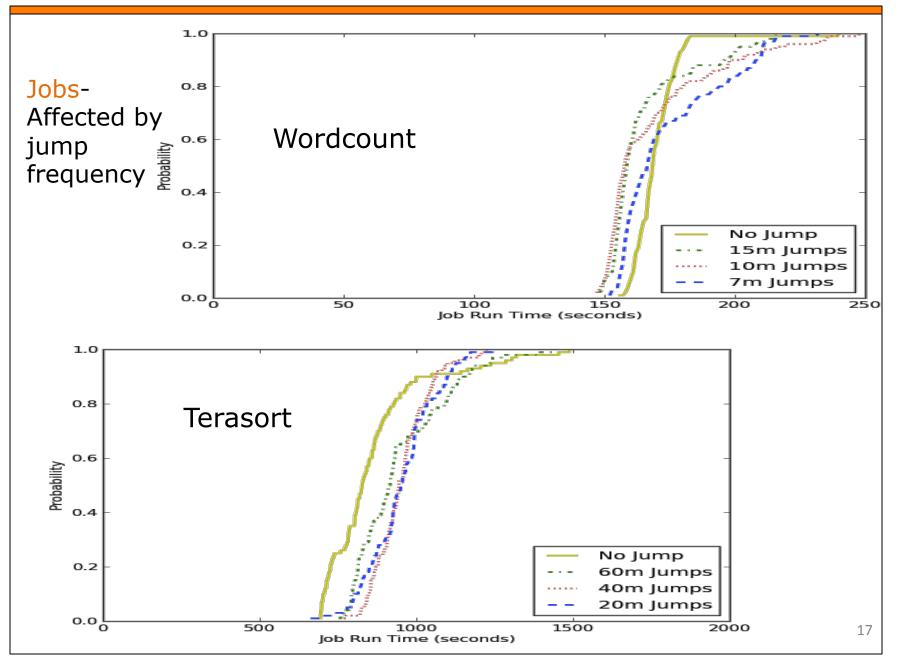
Performance Measurements





Performance Measurements





Optimizing Jump Time



- Upper Bound
 - Maximum jump times to decrease overhead
 - Maximize nodes within cluster to increase parallel processing
 - Jump time must be less than maximum reservation window divided by the number of nodes
- Lower Bound
 - Hadoop behavior contributes to jump times
 - Must be greater than time to re-replicate all datablocks
 - Must be greater than the average execution time of the reduce tasks of currently executing MapReduce jobs



Related Works



- SDSC myHadoop and Apache Hadoop-on-Demand (HOD)
 - Dynamically in userspace
 - HOD requires access to static external HDFS cluster
 - Limited to reservation limitations
 - Does not provide pseudo-persistent interactive capability
- FutureGrid and Amazon Elastic MapReduce Cloud
 - Use virtualization
 - Degraded I/O performance for data-intensive applications
- Mesos and Apache Hadoop YARN
 - New resource management mechanisms
 - Handle Hadoop and non-Hadoop processes
 - Still in development



Conclusion



- Interactive pseudo-persistent MapReduce platform within the existing administrative structure of an academic high performance computing center
- As efficient as a persistent Hadoop cluster on dedicated computing resources, depending on the jump time
- Cluster remains stable, with good performance, in the presence of jumps that occur as frequently as the average length of reduce tasks







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https://sourceforge.net/projects/jummp/

