Improving Hadoop MapReduce Performance on Supercomputers with JVM Reuse

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Supercomputers

- Expensive clusters
 - Multi-core processors
 - Large capacity of main memory
 - High-speed network
- Focus mainly on compute-intensive applications
- Data-intensive workloads are emerging as supercomputing problems
 - Graph processing
 - Pre-processing of simulation data

MapReduce

- Simple parallel paradigm to process large datasets
- Hidden parallelization & communication
- PageRank example

```
Function Mapper
Input PageA → PageB, PageC
Begin
N = outbound links
For each outbound link
output <Page, 1/N>
End
```

Shuffling
Done
automatically
(Users can
ignore)

```
Function Reducer
Input <PageA, x_1>, ...<PageA, x_n>
Begin

rank = 0

For each item x_i

rank += x_i

output <PageA, rank>
End
```

Hadoop MapReduce

- Standard of MapReduce implementation
- Provide easy-to-use MapReduce APIs
- TCP/IP-based communication
- Designed to run on commodity clusters
 - Lab clusters, or Amazon EC2
- Scalability (32,000 nodes at Yahoo) & Resilience
- Written in Java

Improving Hadoop MapReduce Performance on Supercomputers

- Hadoop MapReduce is good choice on supercomputers
 - Maturity
 - Productivity

	Supercomputer	Hadoop
Resource allocation at runtime (# of processes, memory, CPU)	Static	Dynamic
Communication	MPI	TCP/IP
Workload	Compute-intensive	Data-intensive

Our Approach

JVM Reuse

- Statically create JVM processes and dynamically allocate to Hadoop tasks
 - Enable efficient MPI communication by Hadoop tasks
 - Statically created processes can exploit efficient MPI
 - Dynamic allocation enables to use the original Hadoop implementation
 - Shorten start-up time of processes

Technique

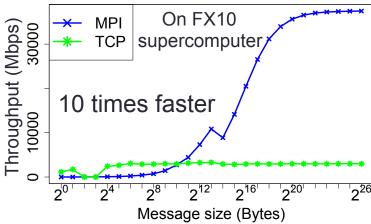
- Process pool is used to implement JVM Reuse
- Minimize changes of the original Hadoop engine

Why MPI is required for Hadoop

The de facto high-speed communication on

supercomputers

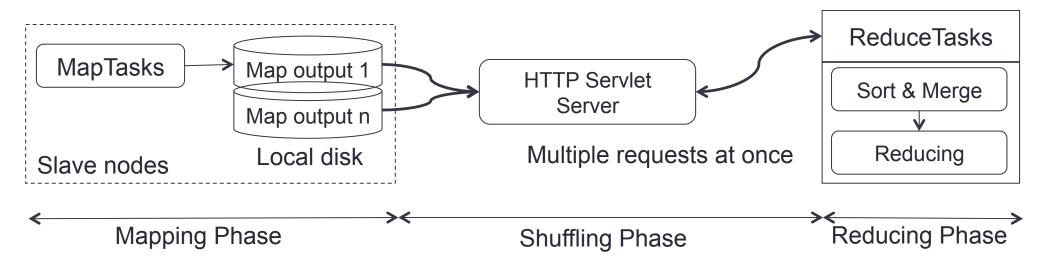
Improve slow MapReduce shuffling



- Enable Hadoop to co-host traditional MPI applications
 - Combine MPI and MapReduce models
 - Rich data analysis workflow
 - Efficient data sharing between MPI and MapReduce models
 - E.g. MPI can access data located at Hadoop file system (HDFS)

Slow MapReduce shuffling on Hadoop

- TCP/IP-based communication
 - JVM-Bypass (Wang et al., IPDPS 2013)

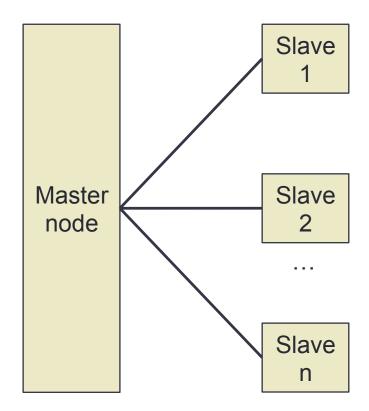


Dynamic Process Creation on MPI

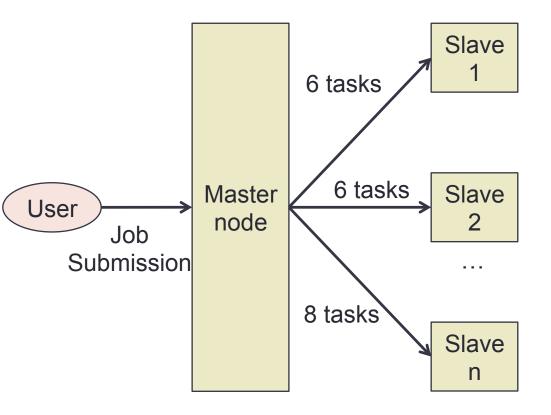
- Discouraged on supercomputers
 - Reasons of performance
 - Collective mechanism (MPISpawn)
 - Gang scheduling (error-prone if not enough resource)
 - Gerbil (Xu el al., CCGrid 2015)
 - Co-hosting MPI applications on Hadoop
 - Creating dynamically processes
 - Its experiments showed significant overhead
- Resources should be specified before running MPI applications
 - Number of processes is known (static)
 - Memory and CPU cores

- Required
 - Resources are allocated on demand to run MapReduce applications
 - Number of processes is unknown (dynamic)

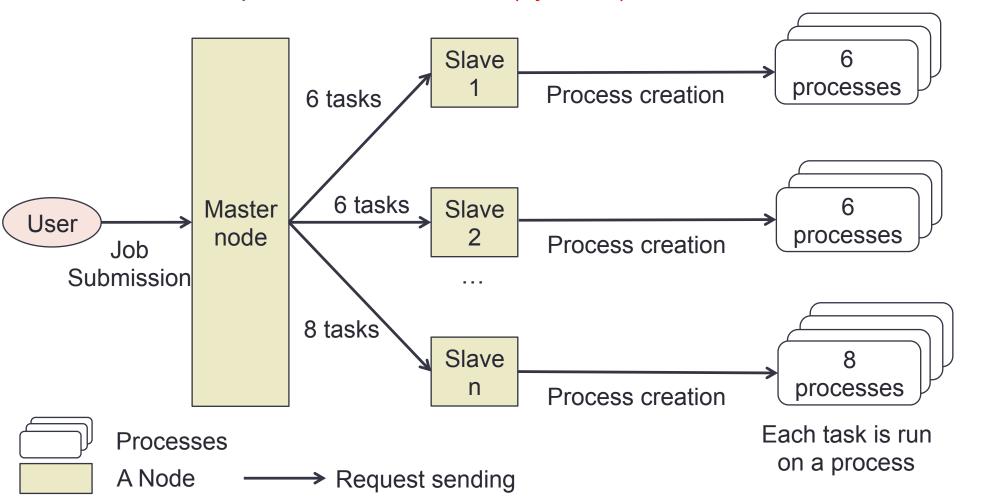
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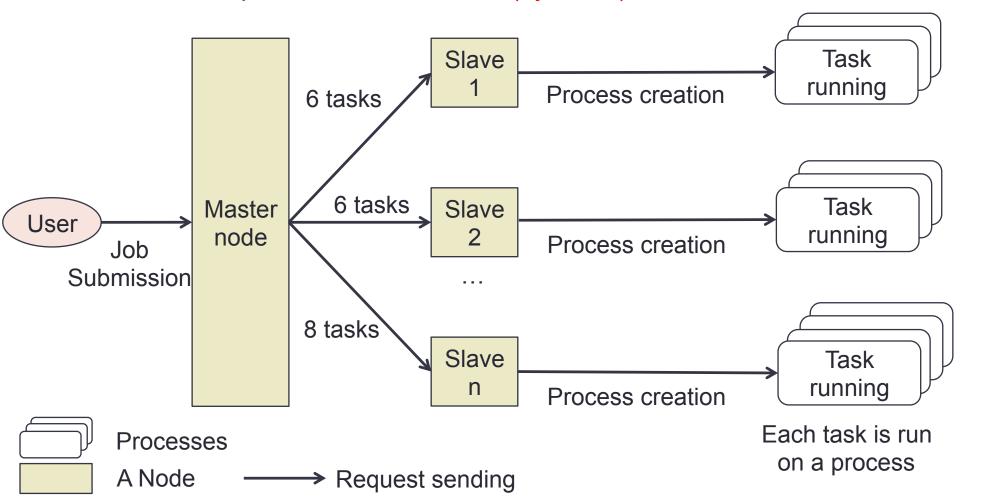
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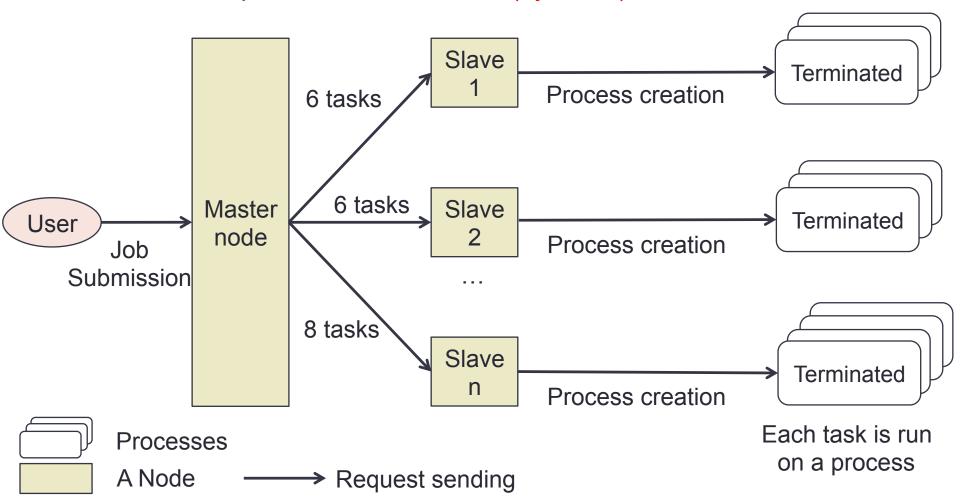
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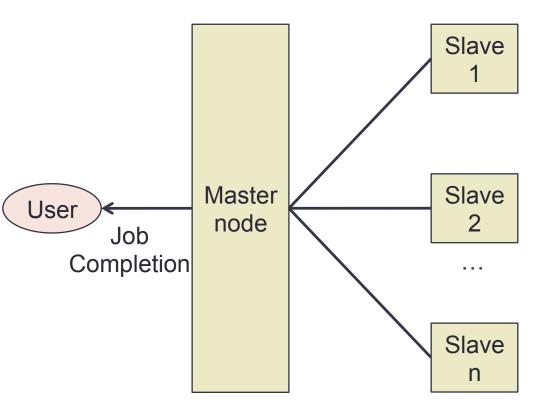
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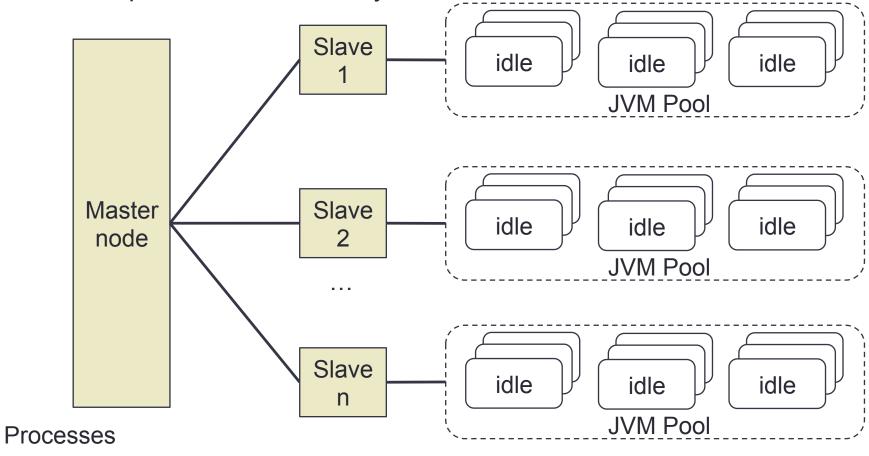


JVM Pool added

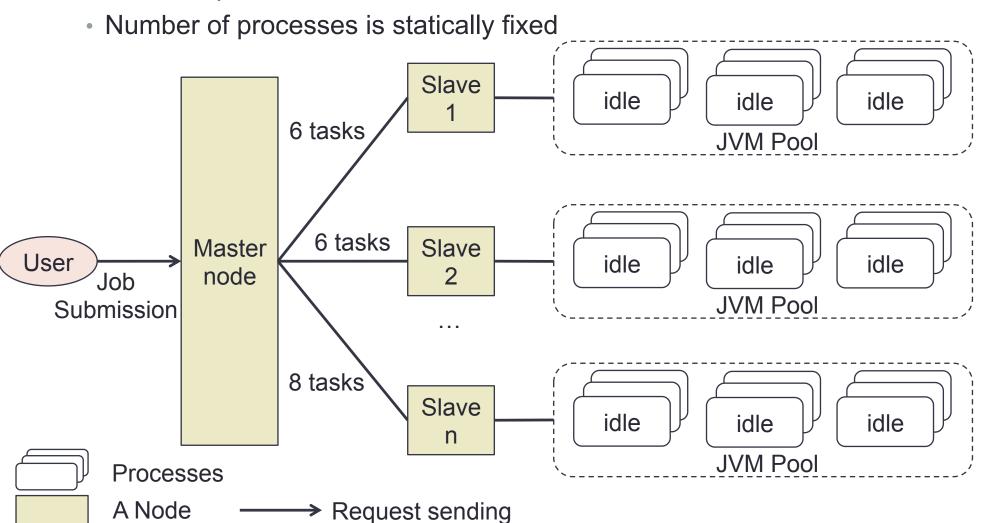
A Node

Idle JVM processes

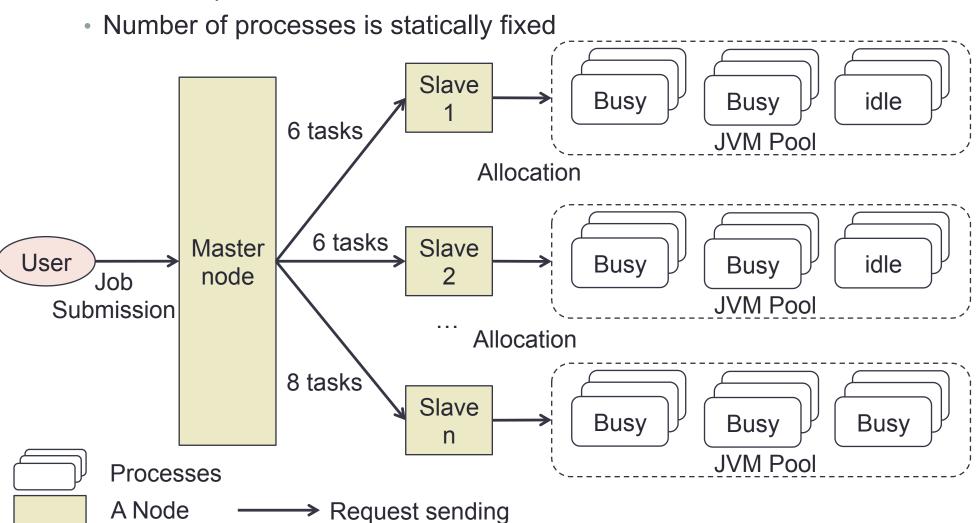
Number of processes is statically fixed



- JVM Pool added
 - Idle JVM processes



- JVM Pool added
 - Idle JVM processes



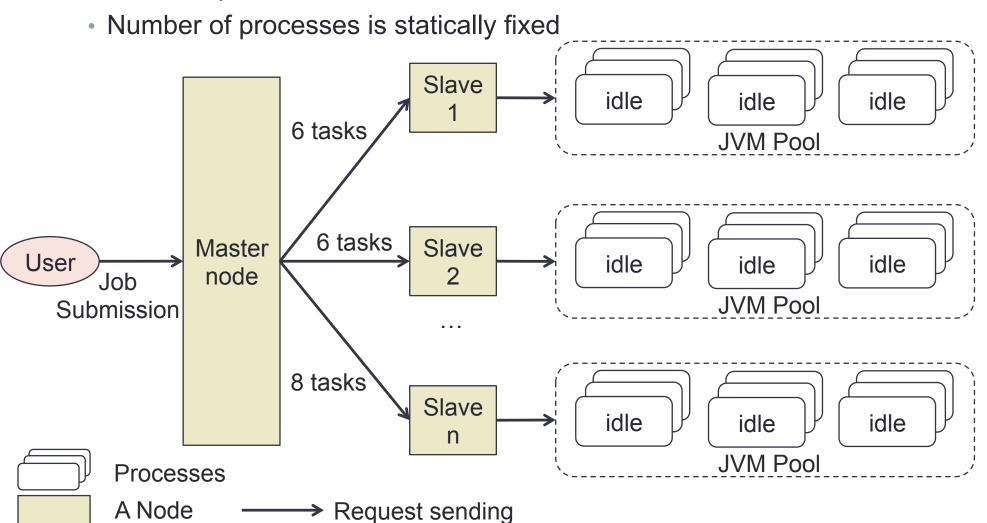
- JVM Pool added
 - Idle JVM processes

 Number of processes is statically fixed Slave Running idle Running 6 tasks JVM Pool 6 tasks Master Slave User Running idle Running node 2 Job JVM Pool Submission 8 tasks Slave Running Running Running n JVM Pool **Processes** A Node Request sending

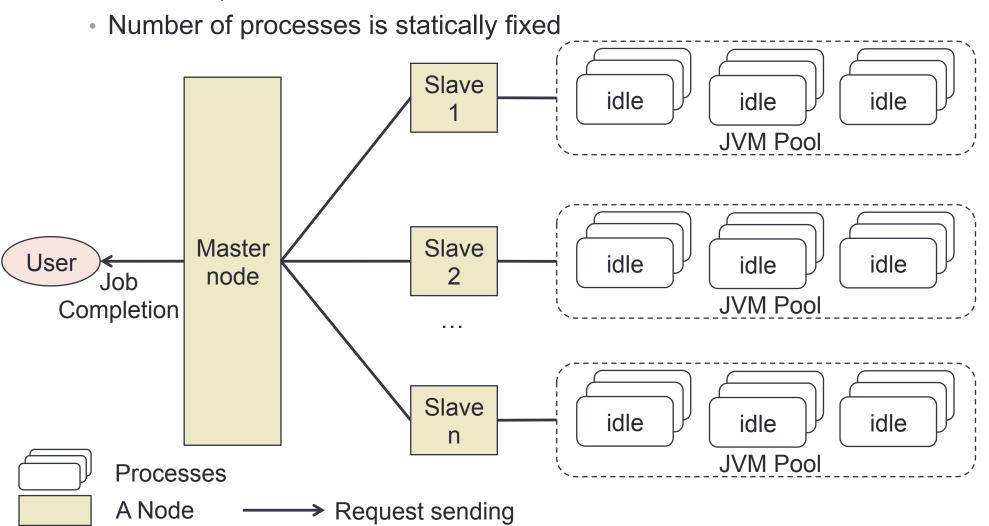
- JVM Pool added
 - Idle JVM processes

 Number of processes is statically fixed Slave Cleanup Cleanup idle 6 tasks JVM Pool 6 tasks Master Slave User Cleanup Cleanup idle node 2 Job JVM Pool Submission 8 tasks Slave Cleanup Cleanup Cleanup n JVM Pool **Processes** A Node Request sending

- JVM Pool added
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- JVM Pool added
 - Idle JVM processes

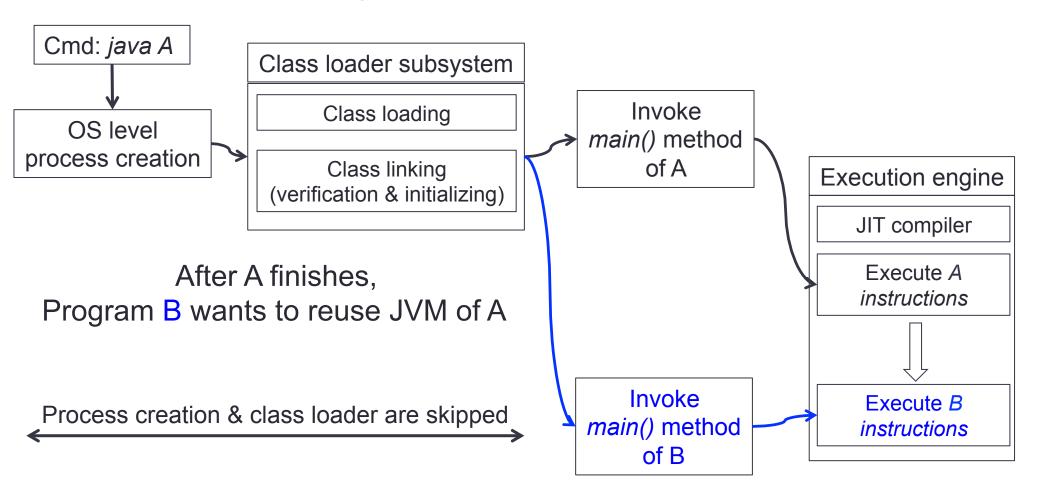


JVM Reuse enables MPI communication

- MPI communication is established at the beginning
- JVM Reuse keeps processes running
 - MPI connection is always available

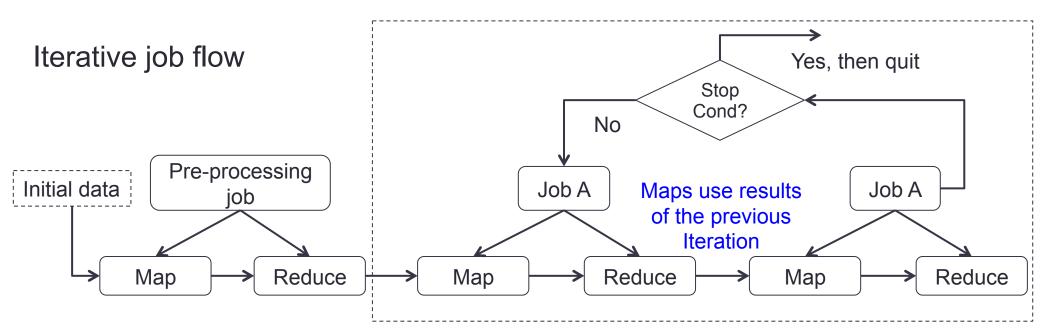
JVM Reuse shortens start-up time

JVM start-up flow of Program A

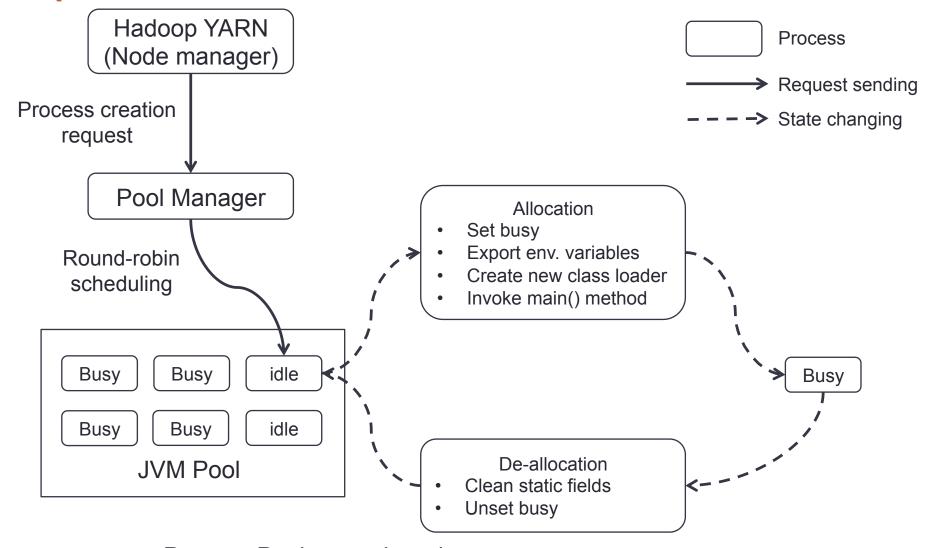


Iterative jobs benefit from JVM Reuse

- Iterative jobs
 - Many short running JVM processes
 - PageRank is an example

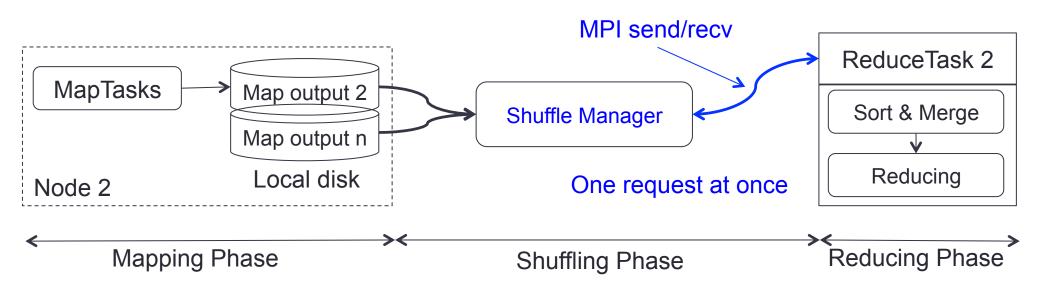


Implementation: Process Pool



Process Pool on each node

Our MPI shuffling design



Reuse's Technical Issues

- Loading user's classes
 - The original flow exports CLASSPATH before running
 - Reflection
 - Load user's classes at runtime
 - Create a new class loader for each user
 - Avoid class confliction
- Clean-up
 - Static fields
 - Security problem
 - e.g. UserGroup static field
 - Must be reset
 - Current design
 - Reset user information and job conf. static fields

Other Technical Issues

- Enable Hadoop YARN to host traditional MPI applications
 - YARN is a resource manager
 - Work in progress
 - MPI AppMaster
 - Monitor MPI ranks
 - MPI Container
 - Host a rank
- Avoid gang scheduling of MPI
 - Work in progress

Evaluation

- Hadoop version
 - v2.2.0
- Changes in our implementation
 - Line of code / total of Hadoop: ~1000 / 1,851,473
 - Number of classes / total of Hadoop: 9 / 35142

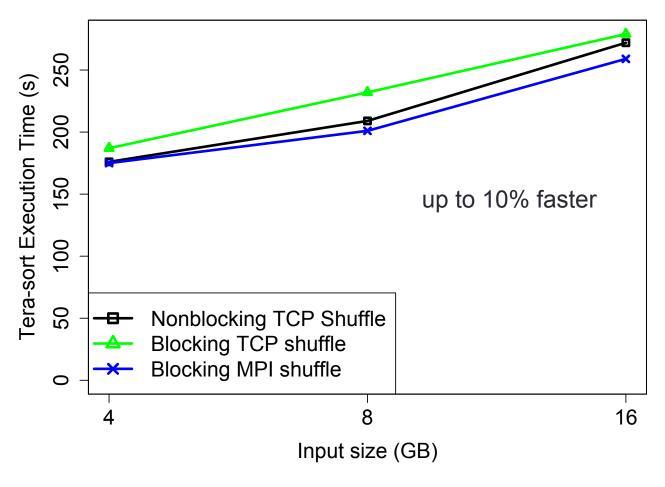
Cluster setup

- FX10 supercomputer
 - Sparc64 Ixfx 1.848 GHz (16 cores) & 32GB RAM
 - MPI over Tofu interconnection (5GB/s)
 - Central storage
- Hadoop setup
 - One master and many slaves
 - OpenJDK 7
 - HDFS is run on the central storage
- OpenMPI 1.6
 - Java MPI binding (Vega-Gisbert et al.)
 - MCA parameter: plm_ple_cpu_affinity = 0

Evaluation of JVM Reuse

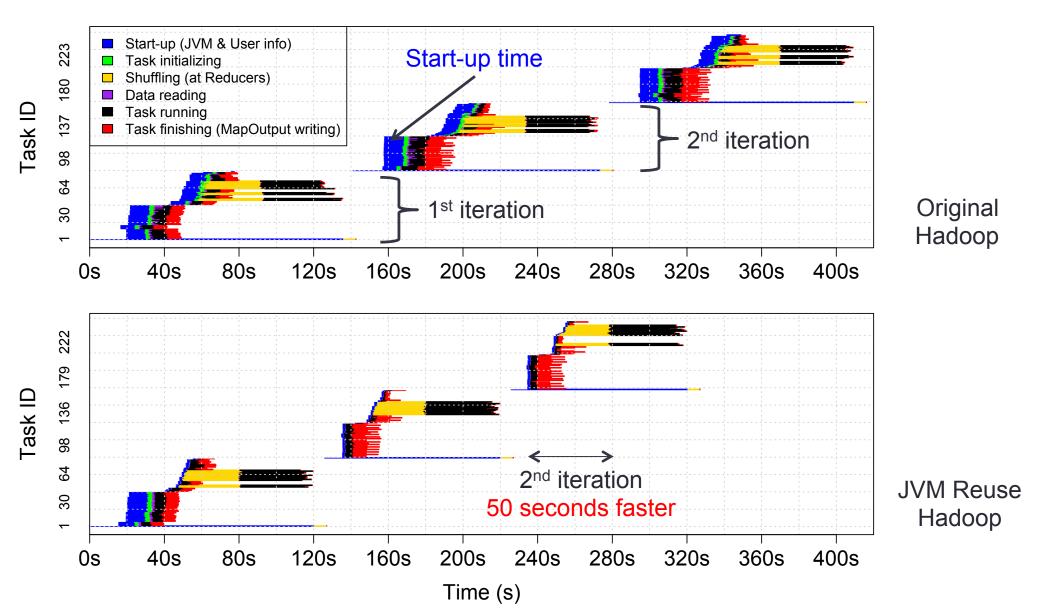
- MPI benefit
 - MPI vs. TCP/IP shuffling
 - Tera-sort job
 - Run on 32 FX10 nodes
 - 4-slot pool & -Xmx4096
- Start-up time
 - JVM Reuse vs. the original
 - PageRank iterative job
 - 400 GB wikipedia data
 - Run on 8 FX10 nodes
 - 6-slot pool & -Xmx4096

MPI vs. TCP/IP shuffling



Nonblocking : accept multi-connection at once Blocking : accept one connection at once

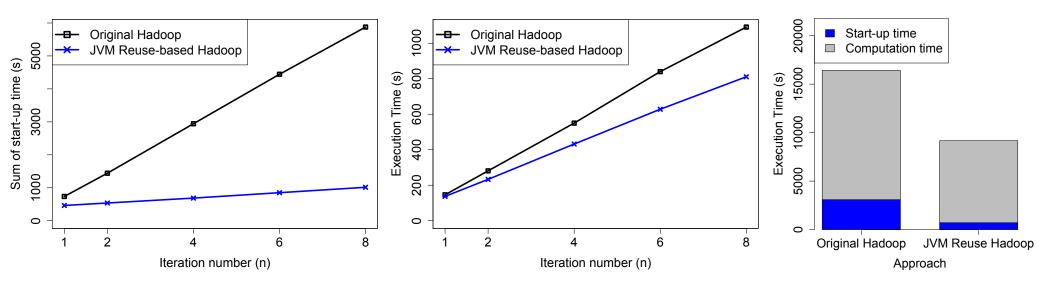
Shorten start-up time (PageRank)



Ratio

More iterations

Sum of start-up time



Execution time

Related work

- M3R (VLDB 2012)
 - Also apply JVM Reuse to enable in-memory MapReduce
 - Not providing any optimization of JVM Reuse and its evaluation
 - Hadoop MapReduce (HMR) engine is written in X10
 - We keep the original HMR engine with minimum changes
- JVM Reuse in Hadoop v1 (2012)
 - Only for a single job
 - JVM processes are terminated after their job is completed
- Gerbil: MPI + YARN (CCGrid 2015)
 - Hadoop Yarn co-hosts MPI applications
 - Long start-up time and significant overhead
- DataMPI (IPDPS 2014)
 - Hadoop-like MapReduce implementation using MPI & C
- JVM-Bypass (IPDPS 2013)
 - C-based shuffling engine & RDMA supported
 - We focus on using MPI over Hadoop processes

Summary

- Improving Hadoop MapReduce performance on supercomputers
- Approach: JVM Reuse
 - Statically create JVM processes and dynamically allocate to Hadoop tasks
 - Enable efficient MPI communication on Hadoop
 - Shorten start-up time
 - Minimum changes of the original Hadoop
- Future work
 - JVM Reuse drawback
 - Affect CPU-bound tasks
 - Co-host MPI applications more efficiently
 - Full cleanup