Kepler + Hadoop: A General Architecture Facilitating Data-Intensive Applications in Scientific Workflow Systems

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Introduction

• Goals
  – Easily compose MapReduce applications in workflow
  – Easily connect MapReduce applications with other programs via workflow
  – Efficiently execute them in the Hadoop environment

• Advantages: combination of characteristics
  – Scientific workflow: GUI, component reuse and sharing, task composition
  – MapReduce: parallelism, scalability, automatic data partitioning, load balancing, fault tolerance
Background – Kepler

• **Actor-oriented Modeling**
  – All these actors inherit the same interfaces, such as `prefire()`, `fire()` and `postfire()`

• **Model of Computation**
  – Synchronous Data Flow (SDF) director: actors execute sequentially
  – Process Network (PN) director: each actor has its own execution thread and execute in parallel

• **Others**
  – Actor customization
  – Actor reuse and sharing locally or publicly through the Kepler actor repository
Background – MapReduce and Hadoop

• **MapReduce**
  – A parallel and scalable programming model for data-intensive computing
  – Input data is automatically partitioned onto multiple nodes and programs are distributed and executed in parallel on the partitioned data blocks.

  \[
  \text{map} \ (k1, v1) \Rightarrow \text{list}(k2, v2) \\
  \text{reduce} \ (k2, \text{list}(v2)) \Rightarrow \text{list}(v2)
  \]

• **Hadoop**
  – Open source implementation of MapReduce.
  – Consists of MapReduce runtime system and a distributed file system, called HDFS.
  – One Hadoop node, called *master*, dispatches tasks and manages the executions of the other Hadoop nodes, i.e., *slaves*
(a) MapReduce actor. (b) Map sub-workflow in MapReduce actor. (c) Reduce sub-workflow in MapReduce actor.
MapReduce Actor Usage
MapReduce Actor Execution in Hadoop

1. Transfer input data from local FS to HDFS.
2. MapReduce Actor execution on Hadoop.
3. Transfer output data from HDFS to local FS.

Kepler GUI

Hadoop Master

Hadoop Slaves

Map slaves with input data blocks
Reduce slaves with output data blocks

distribute Kepler execution engine with Map sub-workflow
distribute Kepler execution engine with Reduce sub-workflow
Execution Semantics in MapReduce Actor

map (k1, v1) {
    initialize Kepler execution engine for Map sub-workflow
    send k1 to Kepler engine via MapInputKey actor
    send v1 to Kepler engine via MapInputValue actor
    execute Map sub-workflow
    get list(k2, v2) from Kepler engine via MapOutputList actor
    emit list(k2, v2)
}

reduce (k2, list(v2)) {
    ...
}
Using MapReduce Actor for Word Count

Word count workflow in Kepler

Map sub-workflow

Reduce sub-workflow

Sub-workflow in IterateOverArray actor

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Experiment 1: Execution on Different Cluster Nodes

Data Sets
- protein data bank files
- 1391 files
- each with 3758 lines
- total size is 348 MB
Overhead Analysis

• Overhead Reason
  – Kepler engine initialization
  – Map/Reduce sub-workflow parsing

• Overhead in this Case
  – The overhead for each Map/Reduce sub-workflow instance takes about 10 milliseconds
  – The execution time of each Map/Reduce instance in Java is much shorter (0.3 ms for Map; 0.03 ms for Reduce)
  – The whole execution number for the Map/Reduce function invocation is about 20 million
Experiment 2: Execution with Increased Data Size in Map
Experiment 3: Execution with Increased Execution Time in MapReduce

![Graph showing total execution time vs. sleep time for Kepler Workflow in Hadoop and Java Program in Hadoop]
Using MapReduce Actor for A Parameter Sweep Application

A Parameter Sweep workflow in Kepler

Map sub-workflow

Sub-workflow in IterateOverArray actor
Conclusion and Future Work

• **Kepler + Hadoop: A General Architecture Facilitating Data-Intensive Applications**
  - Easily create MapReduce sub-workflows, connect them with other tasks using Kepler
  - Execute them efficiently and transparently via the Hadoop infrastructure

• **Future Work**
  - Refactor to enhance its capability, performance, and robustness
  - Apply to concrete domain-specific scientific problems
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• **For More Information:**
  – Distributed Execution Interest Group of Kepler: [https://dev.kepler-project.org/developers/interest-groups/distributed](https://dev.kepler-project.org/developers/interest-groups/distributed)
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