Cooperative Resource Management for Parallel and Distributed Systems

Cristian KLEIN

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Computing needs are ever increasing ...









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Big computing and data requirements

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Single-owner Computing Resources

Supercomputer



- Computers at the front-line
- Large-scale: 100,000 nodes; 1,500,000 cores
- Complex network topologies: torus, fat-tree
- Heterogeneous computing nodes
 - Blue Waters: CPU-only and CPU+GPU nodes
 - Curie: Fat, Hybrid, Thin nodes
- Top #1 (Titan): 188 M\$ + 6 M\$/yr

Single-owner Computing Resources

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Clusters



- Smaller scale
- Commodity hardware
- One cluster \rightarrow nearly homogeneous
- Multiple cluster \rightarrow heterogeneous

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Multi-owner Computing Resources

Grid Computing



- Basically a multi-cluster system
- Geographically dispersed
- Owned by multiple institutions

Cloud Computing



- Renting computing resource from a provider
- Amazon EC2 "supercomputer":
 - 1060/hr for 1,250 nodes (10,000 cores)

Sky Computing



• Renting from **multiple** providers

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Running Applications on Computing Resources



Selecting Resources

- Take into account: heterogenity, centralized / distributed, price
- Goal: minimize completion time, cost, energy

Resource Management







Resource Management System (RMS)

- Multiplexes computing nodes among multiple users
- Aims at isolating them for security and improved performance

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Dynamic allocations (à la Cloud)

 $^{1} \tt{http://blog.cyclecomputing.com/2012/04/cyclecloud-50000-core-utility-supercomputing.html}$

²http://blog.cyclecomputing.com/2011/03/cyclecloud-4096-core-cluster.html

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Dynamic allocations (à la Cloud)

Clouds

"The illusion of *infinite* computing resources available on demand"



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Infinite? Actually up to 20 nodes



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"Supercomputer" of 5,674 nodes (50,000 cores) spanning 7 Amazon EC2 regions¹

Out of capacity errors²



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Out of capacity errors²



Static allocations (à la batch schedulers)

- a.k.a. rigid jobs (node-count times duration)
- Misses opportunities for improvement (next slide)

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Moldability



Moldability



Moldability











Problem: Insufficiently supported in the state-of-the art.

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Goal of the Thesis

Improve resource management

- Resource utilization
- User-chosen criterion:
 - Application completion time
 - Energy consumption / Cost

How?

- Resource management architectures
- Cooperates with applications
- Support moldability, malleability, evolution
 - without workarounds
 - reliably
 - efficiently
- Focus is on interaction

Re-use proven scheduling algorithms as much as possible

Contributions: Resource Management Architectures



¹C. Klein, C. Pérez, An RMS Architecture for Efficiently Supporting Moldable Application, HPCC, 2011

²C. Klein, C. Pérez, *Towards Scheduling Evolving Applications*, CGWS, 2011

³C. Klein, C. Pérez, An RMS for Non-predictably Evolving Applications, Cluster, 2011

⁴ F. Camillo, E. Caron, R. Guivarch, A. Hurault, C. Klein, C. Pérez, *Diet-ethic: Fair Scheduling of Optional Computations in GridRPC Middleware*, INRIA RR-7959, 2012

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Introduction

2 CooRMv1: Moldability

- Computational Electromagnetics Application
- Architecture Description
- RMS/Application-side Scheduling
- Evaluation

3 CooRMv2: Malleability, Evolution

- Adaptive Mesh Refinement Application
- Architecture Description
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4 Conclusions and Perspectives

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Conclusions and Perspectives

Computational Electromagnetics (CEM)

CEM Application

- Part of the ANR DiscoGrid project
- Antenna performance, electromagnetic compatibility
- Traditionally executed on a single cluster
- Huge mesh (number of tetrahedra) ightarrow launch on multiple clusters



Efficient Execution of a Multi-cluster Moldable Applications

Performance of the CEM Application



Performance of the CEM Application



Devised a performance model

- cluster computation power
- inter-cluster network metrics (latency, bandwidth)
- Devised a custom resource selection algorithm















- No moldability (rigid jobs): fix node-count and duration
 - Most batch schedulers
 - Workaround: re-implement RMS's scheduling algorithm



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 - 8–16 nodes \times 2 hours
 - e.g., SLURM



- No moldability (rigid jobs): fix node-count and duration
 - Most batch schedulers
 - Workaround: re-implement RMS's scheduling algorithm
- Limited moldability: range of node-counts and a single duration
 - 8–16 nodes \times 2 hours
 - e.g., SLURM
- \bullet Moldable configurations: list of node-counts \times durations
 - 8 nodes \times 2 hours *OR* 16 nodes \times 1 hour *OR* . . .
 - e.g., OAR, Moab
 - Impractical: large number of configurations (next slide)


Number of Configurations

For a multi-cluster system:

- e.g., number of nodes on each cluster
- # configurations is large (exponential)

$\begin{array}{c} \# \text{ clusters:} & C \\ \# \text{ nodes per clusters:} & N \\ \# \text{ configurations:} & (N+1)^{\mathbf{C}} - 1 \end{array}$

For a supercomputer:

- number of CPU nodes
- number of CPU+GPU nodes
- network topology
- # configurations is large (potentially exponential)

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For a supercomputer:

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- number of CPU+GPU nodes
- network topology
- # configurations is large (potentially exponential)

Problem

What interface should the RMS expose to allow moldable applications to effectively select resources?

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How CooRM Should Work

• Applications should take a more active role in the scheduling





- Applications should take a more active role in the scheduling
- RMS gives application the resource occupation (we call this a view)
 No need to re-implement RMS's scheduling algorithm





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Architecture



Architecture





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Main Responsibilities

- Compute views
- Compute start-times for requests
- Allocate node IDs

Example Implementation

Based on Conservative Back-Filling (CBF)

Fair-start Delay and Ghosts



Initial schedule

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Application-side Scheduling CEM Application

Application-side Scheduling

CEM Application



Experimental Setup

Resource Model

- n_C clusters, each having 128 hosts
- Cluster *i* is considered $10\%, 20\%, \ldots$ faster than cluster 1
- WAN: $5\,\mathrm{ms}$ same city, $10\,\mathrm{ms}$ same country, $50\,\mathrm{ms}$ otherwise

Application Model

- 200 application arriving at $1 \operatorname{app}/\operatorname{second}$
- Mixture of
 - rigid, single-cluster moldable
 - consecutive jobs from LLNL-Atlas-2006-1.1-cln
 - 80% rigid jobs (as in traces)
 - 20% single-cluster moldable jobs (using Amdahl's law)
 - multi-cluster moldable applications (CEM)

Simulation Results



CooRMv1 Implementation

- 2,300 SLOC of Python code
- Prototype implementation using CORBA (omniORBpy)
- CPU-time vs. simulation time
- TCP payload vs. size of messages



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Adaptive Mesh Refinement Applications (AMR)

- Mesh is dynamically refined / coarsened as required by numerical precision
 - Memory requirements increase / decrease
 - Amount of parallelism increases / decreases
- Generally evolves non-predictably





End-user's Goal: maintain a given target efficiency

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Cooperative Resource Management

Problem and Goal

Problem

- Static allocations \rightarrow inefficient resource utilisation
- Dynamic allocations (à la Cloud) \rightarrow out of capacity
- Malleable jobs (KOALA, ReShape, Faucets ...)
 - \rightarrow no guarantees \Rightarrow application may crash
 - \rightarrow difficult to target custom objective

Goal

An RMS which allows non-predictably evolving applications

- To use resources efficiently
- Guarantee the availability of resources

$\rm CooRMv2:$ Additions to $\rm CooRMv1$



Overview

- Resource requests types
- Request relations
- Preemptible views

- Number of nodes, duration
- $\bullet~\mathsf{RMS}$ chooses start time \to node IDs are allocated to the application

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- Type
 - Non-preemptible (default in major RMSs, i.e., are not taken away)
 - Preemptible (i.e., can be taken away at any time)
 - Preallocation

"I do not currently need these resources, but make sure I can get them immediately if I need them."

Request Relations

Request Relations

 Dynamic applications → multiple requests + temporal constraints relatedTo an existing request relatedHow FREE, NEXT, COALLOC

• Two methods: request(), done()



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High-level Operations Spontaneous Update Announced Update $\int_{t}^{t} \int_{t}^{t} \int_{t}$

An update is guaranteed to succeed only inside a pre-allocation

Views

Views

- Apps need to adapt their requests to the availability of the resources
- Preemptible view informs when resources need to be preempted



Example Interaction



CooRMv2 RMS Implementation

Overview

- Compute views
- Compute start times for each requests
- Start requests and allocate resources

Main Idea of the Scheduling Algorithm

- Pre-allocations and non-preemptible requests
 - Conservative Back-Filling (CBF)
- Preemptible requests
 - Equi-partitioning
 - Allow unused partitions to be filled by other applications

Non-predictably Evolving: Adaptive Mesh Refinement

Application Model

- Application knows its speed-up model
- Cannot predict its data evolution
- Aim: maintain a given target efficiency

Behaviour in ${\rm CooRMv2}$

- Sends one pre-allocation
 - Simulation parameter: overcommitFactor
- Sends non-preemptible requests inside the pre-allocation

Malleable: Parameter-Sweep Application

Application Model

- Infinite number of single-node tasks
- All tasks have the same duration (known in advance)
- Aim: maximize throughput

Behaviour in ${\rm CooRMv2}$

- Send preemptible requests
- Spawn tasks if resources are available
- Kill tasks if RMS asks to (increases waste)
- Wait for task completion, if informed in due time (no waste)

Scheduling with Spontaneous Updates

Experimental Setup

- Apps: 1xAMR (target eff. = 75%), 1xPSA (task duration = 600 $\rm s)$
- Resources: number of nodes just enough to fit the AMR
- AMR uses fixed / dynamic allocations



Scheduling with Spontaneous Updates

Experimental Setup

- Apps: 1xAMR (target eff. = 75%), 1xPSA (task duration = 600 $\rm s)$
- Resources: number of nodes just enough to fit the AMR
- AMR uses fixed / dynamic allocations



Wasted 40 M against 500 k (node×seconds)

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Scheduling with Announced Updates

Experimental Setup

- Apps: 1xAMR (target eff. = 75%), 1xPSA (task duration = $600 \, \mathrm{s}$)
- Resources: number of nodes just enough to fit the AMR
- AMR uses announced updates (announce interval)


Announced Updates: Nice Resource "Filling"

Experimental Setup

- 1xAMR application
- PSA_1 : task duration = 600 s



Announced Updates: Nice Resource "Filling"

Experimental Setup

- 1xAMR application
- PSA_1 : task duration = 600 s
- PSA_2 : task duration = 60 s



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Conclusions

Goal: Improve resource management

- Proposing resource management architectures
- Promote collaboration with applications



CooRMv1 (1/2)



- Resource Management Architecture
- Efficiently support moldable applications
- Number of configurations is significantly reduced $(10^3 \text{ vs. } 10^{17})$
- New cases become practical
- Validated through simulation and prototype implementation
- Studied time needed for applications to adapt



Integration with CooRMv1

- Implemented by N. Toukourou, Engineer, INRIA
- **Results**: Easier to launch computation schemas on computing resources



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Custom Scheduling Algorithm for High-Level Waste Simulator

- Co-advised V. Lanore, ex M2 Student, ENS de Lyon
- Scheduling multi-level applications
- **Results**: Response-time improved (accepted ComPAS'13)

CooRMv2



- $\bullet~Extension$ of ${\rm CooRMv1}$
- Efficiently deal with evolving/malleable applications
- Effective resource usage improved up to 3.6 times
- Validated through simulations
- Prototype implementation is available

Other Contributions



distCooRM

- Collaboration with Y. Radenac, Myriads, INRIA
- Distributed version of CooRMv1
- Results: Shows good scalability (for a limited number of applications)

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Optional Computation Support

- Collaboration with F. Camillo, R. Guivarch, A. Hurault, IRIT
- Grid-TLSE+DIET Use-case: Multiple Threshold Pivoting
- Architecture for efficiently dealing with optional computation
- Results: Improves user satisfaction and fairness (submitted CCGrid'13)
- Transfer: DIET patches submitted upstream

Perspectives

Short-term

- Integrate CooRMv1/v2 in existing batch schedulers
 OAR, SLURM
- Validation with other applications
 - Cost, energy

Long-term

- Topology inside a supercomputer/cluster
 - Allow pre-launch topology optimization
- Economic model (à la Cloud)
 - Charge for pre-allocation?
 - Bonus for timely updates?
- distCooRM
 - Improve scalability (add a pre-selection phase)
 - Add malleable / evolving support

Backup Slides

5 Related Work

- NIST Cloud Definition
- SLA Definition





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- NIST Cloud Definition
- SLA Definition





NIST Cloud Definition

Cloud computing is a model for enabling ubiquitous, convenient, on-demand network access to a shared pool of configurable computing resources (e.g., networks, servers, storage, applications, and services) that can be rapidly provisioned and released with minimal management effort or service provider interaction.

- Essential characteristics
 - on-demand self-service
 - broad network access
 - resource pooling
 - rapid elasticity
 - measured service
- Service models: SaaS, PaaS, IaaS
- Deployment models: private, community, public, hybrid

SLA@SOI Definition

- Machine-readable contract between a customer and a provider
- Guarantee that what you ask for, you get
- Allow you to verify provisioning
- Notify violations and define appropriate automated actions/penalties

5 Related Work

- NIST Cloud Definition
- SLA Definition





Fairness

Simulation Setup

- Fair-start Delay: 5 seconds
- 1 x complex-moldable applications (CEM)
 - Simulated applications with lengthy resource selection
 - Added adaptation delay



5 Related Work

- NIST Cloud Definition
- SLA Definition





AMR Evolution

AMR Examples



AMR Model



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Cooperative Resource Management

Executing AMR applications on HPC resources

Use static allocations (rigid jobs)

- E.g., cluster, supercomputing batch schedulers
- Evolution is not known in advance
 - \rightarrow User is forced to over-allocate
 - \rightarrow Inefficient resource usage
- Example: target efficiency 75% ($\pm 10\%$)



Ideally, unused resources should be filled by other applications
 Needs support from the Resource Management System (RMS)