A Hyper-heuristic for scheduling independent jobs in Computational Grids

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Overview

- Introduction and motivation
- Hyper-heuristic design
- Hyper-heuristic tests
- Conclusions
- Future work





Introduction and motivation

- Computational Grid
 - Parallel and distributed system that enables the sharing, selection and aggregation of geographically distributed resources
- Efficient scheduling of tasks in resources in a global, heterogeneous and dynamic environment
- Tasks
 - From different users
 - Executed in unique resource
 - Different types (intensive numeric computation vs data process) / (immediate vs batch)
- Resources
 - Dynamically added/dropped from the Grid
 - Can process one task at a time
 - Specialized resources (intensive numeric computation vs data process)





Introduction and motivation

- Ad-hoc heuristics:
 - Simples
 - Deterministic
 - Short execution time
- E.g.: Opportunistic Load balancing, Minimum Completion Time, Minimum Execution Time, etc...
- No one method performs best!
 - Need to select them in an accordance with grid instance to yield best performance





Problem definition (instances)

- Tasks to be scheduled
- Resources to be used in scheduling
- Workload of each task (in millions of instructions)
- Computing capacity of each resource in mips
- Ready time: ready[m] when the resource m will finish executing its scheduled tasks.
- ETC[t][m] Expected Time to Compute task t in resource m (from Simulation Model of Braun et al. 2001)





Problem definition (Objectives)

- Makespan: finishing time of latest task max{F_t: t in Tasks} or max{Cr: r in Resources}
- Flowtime: sum of finishing times of tasks sum {F_t: t in Tasks}





Ad-hoc heuristics

- Our Ad-hoc heuristics:
 - Immediate mode: tasks are scheduled as soon as they arrive in the system
 - Opportunistic Load Balancing, Minimum Completion Time, Minimum Execution Time, Switching Algorithm and K-Percent Best
 - Batch mode: The schedule is done for a set of tasks (a *batch*).
 - Min-Min, Max-Min, Sufferage and Relative-Cost





Evaluating instance characteristics

- Instance notation x_yyzz (Braun et al. 2001)
 - X : computing consistency (c-consistent, i-inconsistent and s-semiconsistent)
 - YY: Tasks heterogeneity (hi-high and lo-low)
 - ZZ: Resources heterogeneity (hi-high and lo-low)
- Preprocess input information
 - Workload variance task heterogeneity
 - Mips variance resource heterogeneity
 - ETC matrix analysis matrix consistency

Note: ETC=Expected Time to Compute





Performance of Hyper-heuristic for Braun et al.'s instances - Makespan

	OL B	МСТ	MET	SA	KPB	Min- Min	Max- Min	Suff	RC
C_HIHI		X				X			
C_HILO		X							X
C_LOHI		X				X			
C_LOL		X							X
<u>Рнн</u>		X						X	
I_HILO					X			X	
I_LOHI			X					X	
I_LOLO					X			X	
S_HIHI		X							X
S_HILO					X				X
S_LOHI					X				X
S_LOLO		X							X

X-the method was chosen most of the times out of 100 independent runs





Performance of Hyper-heuristic for Braun et al.'s instances - Flowtime

	OL B	МСТ	MET	SA	KPB	Min- Min	Max- Min	Suff	RC
C_HIHI				Χ		X			
C_HILO				Χ		X			
C_LOHI				Χ		X			
C_LOL				Χ		X			
<u>Рнн</u>			X			X			
I_HILO			X			X			
I_LOHI			X			X			
I_LOLO			X			X			
S_HIHI				Χ		X			
S_HILO				Χ		X			
S_LOHI				Χ		X			
S_LOLO				Χ		X			

X-the method was chosen most of the times out of 100 independent runs







Proposal: Hyper-heuristic design

- Parameters of the hyper-heuristic:
 - Parameter to fix task heterogeneity threshold
 - Parameter to fix resource heterogeneity threshold
 - Parameter to work with immediate or batch methods
 - Parameter to fix the measure to optimize (makespan or flowtime)





Proposal: Hyper-heuristic design

- High-level algorithm:
 - *Input*: parameters, tasks, resources, ready-times, ETC matrix
 - 1. Evaluate task heterogeneity
 - 2. Evaluate resource heterogeneity
 - 3. Examine ETC matrix to deduce its consistency
 - 4. Choose (based on parameters) the ad-hoc method to execute
 - 5. Execute ad-hoc method

Output: schedule





Performance Evaluation of Hyper-heuristic for a grid simulation environment

- We use a Grid Simulator implemented with a discrete event simulation library (*HyperSim*).
- Highly parametrizable:
 - Distributions of arriving and leaving of resources in the Grid and its mips
 - Distributions of task arrival to the Grid and its workloads
 - Initial resources/tasks in the system and maximum tasks to generate
 - Task and resource types
 - Percentage of immediate/batch tasks.
- For a schedule event the simulator calls the hyperheuristic and passes it the ETC matrix, ready times, resources and tasks that will be scheduled as input





Simulator trace example

time= 000000000.00 event= EVN NEW HOST info: Host# 00000, Mips = 00000200.00time= 000000000.00 event= EVN_NEW_HOST_info: Host# 00001, Mips = 00000200.00time= 000000000.00 event= EVN ENTER info: Task# 00000, Work = 000006000.00info: Task# 00001, time= 000000000.00 event= EVN ENTER Work = 000006000.00time= 000000000.00 event= EVN ENTER info: Task# 00002, Work = 000006000.00time= 00000000000 event= EVN ENTER info: Task# 00003, Work = 000006000.00time= 000000000.00 event= EVN SCHEDULE info: Scheduled 00004 Tasks, 00002 Hosts time= 000000000.00 event= EVN START info: Task# 00001 on Host# 00001 finishTime = 000000030.00 exeTime = 00000030.00 time= 000000000.00 event= EVN START info: Task# 00000 on Host# 00000 ۲ finishTime = 00000030.00 exeTime = 00000030.00





Simulator + Hyperheuristic







Performance Evaluation of Hyper-heuristic using the simulator: static vs dynamic

- Two environment types for tests:
 - **Static**: generate concrete instances (a priori fixed configuration)
 - Dynamic: the usual use of the simulator intended for a real environment
- Using the Simulator for generating 3 Grid types: Small, Medium and Large size
- Tests for the 2 measures: Makespan and Flowtime
- We compare the hyper-heuristic *versus* an hyperheuristic with fully random decisions
- Percentage ratio of tasks immediate/batch is modified too: 0%, 25%, 50%, 75% and 100%.





Static – Makespan

(small, medium and large size instances)







Static – Flowtime

(small, medium and large size instances)







Dynamic - Makespan







Dynamic - Flowtime







Conclusions

• As expected, the schedules done by the HH using guided decisions are better than decisions without any knowledge.

•For Makespan, we have seen that the results increase when the ratio of immediate/batch is 0.5, this indicates that both types of ad-hocs *damage* each other's strategy

•For Flowtime, when the ratio of immediate/batch is favorable to batch, better results are produced.





Future Work

- Add transmission time in our simulations
- Make the hyperheursitic more "intelligent" in decision-taking
- Evaluate the HH in a real grid:
 - Develop an interface to use it in a real grid
 - Extract the state of the net (grid characteristics, job characteristics etc.)



