Performance and Fairness for Users in Parallel Job Scheduling

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Motivation

- Efficient application of metaheuristics in Grid/cluster job scheduling
- Scheduling in Czech NGI MetaCentrum
 - Managed by queue-based scheduler
 - PBS-Pro, TORQUE (a form of Backfilling)
- Problems of interest
 - Performance
 - Wait time, slowdown, response time
 - Fairness
 - To keep users satisfied
 - **Scheduler's behavior** users keep asking:
 - "Why my job has not started yet?"
 - "Why my job waits when there are free resources?"

Current Approaches

- PBS, LSF, SGE, TORQUE, ...
 - Mostly (aggressive) backfilling
 - No reservations vs. EASY backfilling vs. Conservative backfilling
 - Decisions made in an ad hoc fashion
 - Fairness is very important
 - FCFS somehow fair but inefficient
 - EASY backfilling is dangerous large jobs may be delayed
 - Conservative backfilling quite fair as no job can be delayed
 - Prioritized queues by fairshare principles (balance the user's share)
 - Predictability is not usually supported
 - Advance Reservations may degrade performance
 - Cons. Backfilling is not widely used (reservations limit backfilling opportunities)

Contribution

- Realistic application of metaheuristics in Grid/cluster job scheduling
 - Flexible behavior based on applied criteria and current situation
- Real-life based problem and goals
 - Large problem instances
 - Performance
 - Fairness
 - **Fast solution** (limited runtime awareness)
 - Toward predictions
- Further work
 - Prototype implementation in actual scheduler (TORQUE)

What is "fairness"?

- Inspired by the **fairshare setup** used in MetaCentrum ullet
 - Maximize the share of mostly "penalized" user _
 - Prioritizes users with lower resource consumption —
 - Prioritizes users with higher wait time
- **Basic principles** ullet
 - Fairshare priority = normalized user wait time (NUWT) —

 - NUWT = $\frac{user wait time}{user CPU time}$
 - NUWT = "how many seconds user waits for one second of job execution"
 - Balancing NUWT values
 - Decreases the differences in the performance delivered to the users

Proposed Approach

• Combination of known "best practices"

- Use Conservative backfilling

- Conservative backfilling \rightarrow every job gets a reservation
- Reservations \rightarrow fairness (no "unlimited" delays)
- Backfill-like approach (efficient utilization)
- Predictability plan of job execution

- Use optimization

- Improve quality of execution plan (job schedule)
- Subject to schedule evaluation \rightarrow identification of inefficiencies
 - Wait time
 - Bounded slowdown
 - Response time
 - Fairness

Optimization – limited runtime

- Metaheuristics can be time consuming
 - Limited time due to the on-line problem character
- Time-efficient approach
 - (Valid) initial schedule created quickly using Conservative Backfilling (see +>)
 - Optimization is only executed when there are no higher priority events such as job arrivals or job completions (see
 - Optimization can be stopped after each iteration when necessary



Optimization – Tabu Search

- Improves initial scheduled delivered by Conservative backfilling
- Tabu search-inspired optimization algorithm (TS)
 - Tabu list prevents short cycles
 - Selective re-backfilling guided by evaluation
- Evaluation
 - Guides the optimization phase
 - Performance and Fairness related criteria
 - Wait time
 - Bounded slowdown
 - Response time
 - NUWT



Experimental Results

- Alea simulator
 - Complex job scheduling simulator built on the top of optimized GridSim toolkit
 - Functionality (scheduling algorithms, visualization, ...)
 - Speed (optimized GridSim core)
- 6 data sets from Parallel Workloads Archive
 - MetaCentrum (806 CPUs, 103,656 jobs during 5 months)
 - KTH SP2 (100 CPUs, 28,489 jobs during 11 months)
 - CTC SP2 (338 CPUs, 77,222 jobs during 11 months)
 - SDSC SP2 (128 CPUs, 59,725 jobs during 24 months)
 - SDSC BLUE (1,152 CPUs, 243,314 jobs during 34 months)
 - HPC2N (240 CPUs, 202,876 jobs during 42 months)

Algorithms

- Experimental evaluation of TS against
 - FCFS
 - Bad, offscale-high results
 - Not shown in the graphs
 - Backfilling without reservations (BF)
 - EASY backfilling (first job gets a reservation) (BF-EASY)
 - Conservative backfilling (every job get a reservation) (BF-CONS)
 - Backfilling without reservations + Fairshare (BF-FAIR)

Slowdown + Wait time



Response time



Fairness



Conclusion

- Simple but powerful extension of Conservative backfilling
 - Evaluation and optimization
 - "Controlled" re-backfilling
- Significant improvement
 - Classical criteria
 - Fairness-related criteria
 - Time efficient
- Can be used when job runtime estimates are inaccurate
 - It is only backfilling...
 - Schedule compression is needed when job completes earlier
 - Evaluation is not precise still improving solutions are found regularly

Example



Ongoing Work

- Predictability
 - Conservative backfilling is predictable
 - Due to optimization the "reservations" are changed
 - Optimization delays some jobs wrt. initial assignment
- Multi-resource fairness
 - Memory, I/O
 - Berkeley's Dominant Resource Fairness Fair Allocation of Multiple Resource Types
- Working implementation in TORQUE
 - First tests show better performance wrt. classical techniques
 - Further development toward practical usage

Runtime requirements

- Implementation in a real TORQUE scheduler
- Problem description:
 - 219 nodes with 1494 CPUs
 - Initial schedule consisting of 0..25,000 jobs
- Time needed to add 1 job
- Time needed to perform 1 iteration of TS



qstat command with prediction

