Clause Sharing in Parallel MaxSAT

Ruben Martins Vasco Manquinho Inês Lynce

IST/INESC-ID, Technical University of Lisbon, Portugal

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Maximum Satisfiability

- Maximum Satisfiability (MaxSAT):
 - Optimization version of Boolean Satisfiability (SAT);
 - **Goal:** Given a propositional formula φ , find an assignment to problem variables that maximizes (minimizes) number of satisfied (unsatisfied) clauses in φ .

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- Partial MaxSAT
 - **Goal:** Given a propositional formula $\varphi = \varphi_h \bigcup \varphi_s$, find an assignment to problem variables such that all *hard* clauses in φ_h are satisfied, while minimizing the number of unsatisfied *soft* clauses in φ_s .

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 - **Goal:** Given a propositional formula $\varphi = \varphi_h \bigcup \varphi_s$, find an assignment to problem variables such that all *hard* clauses in φ_h are satisfied, while minimizing the number of unsatisfied *soft* clauses in φ_s .
- MaxSAT solvers are now very effective in practice;
- Multicore processors are becoming predominant;
- As a result, parallel MaxSAT solvers are emerging.

- PWBO is a parallel MaxSAT solver based on having several threads running a portfolio of two orthogonal algorithms:
 - an unsatisfiability-based algorithm that searches on the lower bound of the optimal solution;
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- Not all learned clauses should be shared/imported since it could lead to an exponential blow up in memory;
- Shared clauses can be imported or discarded by the receiving thread;
- **Question:** which learned clauses should be shared/imported by the different threads?

Clause Sharing Heuristics

- Static:
 - $\circ~$ Learned clauses are shared/imported within a given cutoff.
- Dynamic:
 - Dynamic heuristics adjust the cutoff during the search.

• Freezing:

• Shared clauses are temporarily frozen until they are expected to be useful.

Clause Sharing Heuristics (Static)

• Size:

- The clause size is given by the number of literals;
- $\circ~$ Small clauses are expected to be more useful than larger clauses.
- Literal Block Distance (LBD):
 - The literal block distance corresponds to the number of different decision levels involved in a clause;
 - $\circ~$ Clauses with small LBD are considered as more relevant.
- Random:
 - Randomly decide whether to share each learned clause with a given probability.

Clause Sharing Heuristics (Dynamic)

- The size of learned clauses tends to increase over time;
- Dynamic heuristics adjust the size of shared clauses during the search;
- Hamadi et al. proposed the following dynamic heuristic:
 - At every k conflicts the throughput of shared clauses is evaluated between each pair of threads (t_i → t_j);
 - If the sharing is small, the cutoff is dynamically increased;
 - $\circ~$ If the sharing is large, the cutoff is dynamically reduced.

Clause Sharing Heuristics (Dynamic)

- The previous heuristic has been improved by Hamadi et al. by considering the quality of shared clauses:
 - A shared clause is said to have *quality* if at least half of its literals are active;
 - A literal is *active* if its VSIDS heuristic score is high, i.e. it is likely to be chosen as a decision variable in the near future;
 - If the quality is high then the increase (decrease) in the size limit of shared clauses will be larger (smaller).
- The reasoning behind this heuristic is that the information recently received from a thread t_i is qualitatively linked to the information which could be received from the same thread t_i in the near future.

Freezing procedure for importing clauses shared by other threads



The freezing heuristic:

- Considers the status of the shared clause ω in the context of the importing thread:
 - Satisfied: if at least one of its literals is satisfied;
 - Unsatisfied: if all of its literals are unsatisfied;
 - *Unit*: if all literals but one are unsatisfied and the remaining literal is unassigned;
 - Unresolved: if it is not satisfied, unsatisfied or unit.
- Freezes shared clauses ω that are not likely to be useful in the near future.

- A satisfied clause is expected to be useful in the near future if:
 - $\circ~$ It is not necessary to backtrack significantly to make the clause unit;
 - The number of unassigned literals that are not active literals is small;
- Unsatisfied clauses and unit clauses are always useful to the current search;
- An Unresolved clause is expected to be useful in the near future if:
 The number of unassigned literals that are not active literals is small;

- *level* current decision level, i.e. number of decisions since the *root* of the search tree until the current node
- $level_h(\omega)$ smallest decision level of the satisfied literals in ω ;
- $unassignedLits(\omega)$ number of unassigned literals in ω ;
- activeLits(ω) number of active literals in ω;
- ω is satisfied:
 - $(level level_h(\omega) \leq 31);$
 - $(unassignedLits(\omega) activeLits(\omega) \le 5);$
 - $\circ~$ If the above conditions are met ω is imported, otherwise it is frozen.

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- $level_h(\omega)$ smallest decision level of the satisfied literals in ω ;
- $unassignedLits(\omega)$ number of unassigned literals in ω ;
- activeLits(ω) number of active literals in ω;
- ω is unsatisfied or unit:
 - $\circ \omega$ is always imported.
- ω is unresolved:
 - (unassignedLits(ω) − activeLits(ω) ≤ 5) then the clause is imported. Otherwise, it is frozen.

Experimental Results

- Benchmarks: partial MaxSAT instances from the industrial category of the MaxSAT Evaluation 2011:
 - Instances that took less than 60 seconds to be solved were not considered;
- AMD Opteron 6172 processors (2.1 GHz with 64 GB of RAM) running Fedora Core 13;
- Timeout: 1,800 seconds (wall clock time);
- Portfolio version of PWBO with 4 threads:
 - A deterministic version of PWBO was used;
 - Information is only exchanged at synchronization points (every 100 conflicts).

Experimental Results

Comparison of the different heuristics for sharing learned clauses

	Heuristic	#Solved	Avg. #Clauses	Avg. Size	Time	Speedup
	No sharing	137	_	_	32,188.57	1.00
Static	Random 30	134	10,140.22	128.21	27,394.46	1.18
	LBD 5	137	8,947.36	9.94	25,346.69	1.27
	Size 8	137	7,529.18	5.30	25,098.85	1.28
	Size 32	138	18,027.48	11.76	25,174.29	1.28
	Dynamic	138	13,296.28	7.33	24,218.84	1.33
	Freezing	140	16,228.53	11.01	21,611.21	1.49

- Randomly sharing clauses deteriorates the performance;
- LBD and size heuristics have similar speedups;
- Dynamic heuristic outperforms the static heuristics but is outperformed by the freezing heuristic.

Conclusions

- Parallel MaxSAT solvers are now emerging:
 - $\circ~$ Sharing learned clauses boosts the performance of the solver.
- Heuristics are used for sharing learned clauses:
 - Static, Dynamic and Freezing.
- Impact of sharing learned clauses in parallel MaxSAT:
 - Number of solved instances does not increase significantly;
 - $\circ~$ Solving time is considerably reduced.
- The freezing heuristic outperforms all other heuristics both in solving time and number of solved instances.