Clause Sharing in Deterministic Parallel Maximum Satisfiability

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- 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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- Main algorithmic approaches:
 - Branch and Bound
 - · Extensive use of lower bounding procedures
 - Restrictive use of MaxSAT inference rules
 - $\circ~$ Linear search on the number of unsatisfied clauses
 - Each time a new restriction is found, a new constraint is added that excludes solutions with higher cost
 - Unsatisfiability-based solvers
 - Iterative identification of unsatisfiable subformulas

Our focus is on the latter two approaches since these have been shown to be more effective in Industrial instances

Best: ∞

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1$	
$\neg x_6 \lor x_8$	$x_6 \vee \neg x_8$	$x_2 \lor x_4$	$\neg x_4 \lor x_5$	
$x_7 \lor x_5$	$\neg x_7 \lor x_5$	$\neg x_5 \lor x_3$	$\neg x_3$	

Example of MaxSAT formula; Hard clauses in blue; Soft in red

Best: ∞

Add a relaxation variable to each soft clause; All clauses are now considered hard

Best: ∞

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8 \lor r_2$	$x_6 \vee \neg x_8 \vee r_3$	$x_2 \lor x_4 \lor r_4$	$\neg x_4 \lor x_5 \lor r_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5 \lor r_6$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_7$

Goal is to find an assignment that minimizes the number of relaxation variables assigned value 1

Best: 4

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8 \lor r_2$	$x_6 \vee \neg x_8 \vee r_3$	$x_2 \lor x_4 \lor r_4$	$\neg x_4 \lor x_5 \lor r_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5 \lor r_6$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_7$

Find a solution; Suppose a solution is found such that 4 relaxation variables are assigned value 1;

Best: 4

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8 \lor r_2$	$x_6 \vee \neg x_8 \vee r_3$	$x_2 \lor x_4 \lor r_4$	$\neg x_4 \lor x_5 \lor r_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5 \lor r_6$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_7$
$\sum_{i=1}^{7} r_i \leq 3$			

Add new constraint that excludes solutions with equal or higher cost;

Best: 2

$x_6 \vee x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8 \lor r_2$	$x_6 \lor \neg x_8 \lor r_3$	$x_2 \lor x_4 \lor r_4$	$\neg x_4 \lor x_5 \lor r_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5 \lor r_6$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_7$
$\sum_{i=1}^{7} r_i \leq 3$			

Find another solution; Suppose a solution is found such that 2 relaxation variables are assigned value 1;

Best: 2

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8 \lor r_2$	$x_6 \vee \neg x_8 \vee r_3$	$x_2 \lor x_4 \lor r_4$	$\neg x_4 \lor x_5 \lor r_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5 \lor r_6$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_7$
$\sum_{i=1}^{7} r_i \leq 3$	$\sum_{i=1}^{7} r_i \leq 1$		

Add new constraint that excludes solutions with equal or higher cost;

Best: 2

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8 \lor r_2$	$x_6 \vee \neg x_8 \vee r_3$	$x_2 \lor x_4 \lor r_4$	$\neg x_4 \lor x_5 \lor r_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5 \lor r_6$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_7$
$\sum_{i=1}^{7} r_i \leq 3$	$\sum_{i=1}^{7} r_i \leq 1$		

Instance is now UNSAT; Optimal solution is to have two unsatisfied soft clauses

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1$
$\neg x_6 \lor x_8$	$x_6 \vee \neg x_8$	$x_2 \lor x_4$	$\neg x_4 \lor x_5$
$x_7 \lor x_5$	$\neg x_7 \lor x_5$	$\neg x_5 \lor x_3$	¬ <i>x</i> 3

Example of MaxSAT formula; Hard clauses in blue; Soft in red;

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$x_7 \lor x_5$	$\neg x_7 \lor x_5$	$\neg x_5 \lor x_3$	¬ <i>x</i> ₃

Formula is unsat; Get Unsatisfiable subformula (Unsat Core)

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8$	$x_6 \vee \neg x_8$	$x_2 \lor x_4 \lor r_2$	$\neg x_4 \lor x_5 \lor r_3$
$x_7 \lor x_5$	$\neg x_7 \lor x_5$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_4$
$\sum_{i=1}^4 r_i \leq 1$			

Add relaxation variables to soft clauses and AtMost1 constraint

$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1$
$\neg x_6 \lor x_8$	$x_6 \vee \neg x_8$	$x_2 \lor x_4 \lor r_2$	$\neg x_4 \lor x_5 \lor r_3$
$x_7 \lor x_5$	$\neg x_7 \lor x_5$	$\neg x_5 \lor x_3$	$\neg x_3 \lor r_4$
$\sum_{i=1}^4 r_i \leq 1$			

Formula is still unsat; Get another Unsat Core

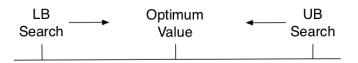
$x_6 \lor x_2$	$\neg x_6 \lor x_2$	$\neg x_2 \lor x_1$	$\neg x_1 \lor r_1 \lor r_5$
$\neg x_6 \lor x_8$	$x_6 \vee \neg x_8$	$x_2 \lor x_4 \lor r_2$	$\neg x_4 \lor x_5 \lor r_3$
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$\sum_{i=1}^4 r_i \leq 1$	$\sum_{i=5}^{7} r_i \leq 1$		

Add new relaxation variables to soft clauses in Unsat Core and AtMost1 constraint

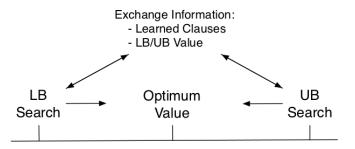
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$\sum_{i=1}^4 r_i \leq 1$	$\sum_{i=5}^{7} r_i \leq 1$		

Instance is now SAT; Algorithm Ends; Optimal solution is to have two unsatisfied soft clauses

- 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;
 - $\circ\,$ a classical linear search algorithm that searches on the upper bound.



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- Shared Clause: a clause that is *shared* by a thread to be used in other threads;
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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)

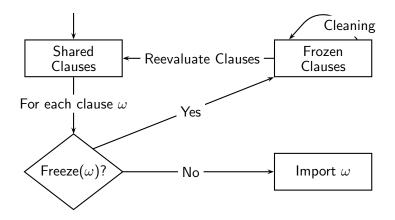
- The size of learned clauses tends to increase over time;
- Dynamic heuristics adjust the size of shared clauses during the search;
- Hamadi et al. proposed the following dynamic heuristic:
 - At every k conflicts the throughput of shared clauses is evaluated between each pair of threads $(t_i \rightarrow t_j)$;
 - 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 the variable's decision 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.

Clause Sharing Heuristics (Freezing)

Freezing procedure for importing clauses shared by other threads



Clause Sharing Heuristics (Freezing)

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.

Clause Sharing Heuristics (Freezing)

- 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;

Clause Sharing Heuristics (Evaluation)

Question: How to properly evaluate all these clause sharing heuristics?

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Observe that:

- Parallel solvers are non-deterministic due to cooperation between threads
- Cooperation is known to boost the performance of parallel solvers
- Variations might result from other factors than clause sharing procedures
- Therefore, a more stable environment is required for a fair evaluation

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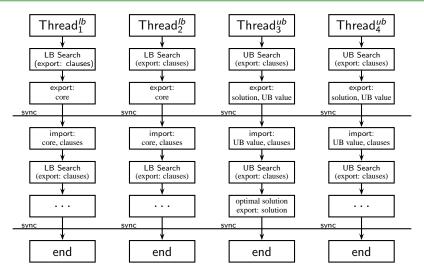
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Proposed approach: test different clause sharing heuristics in a **deterministic** parallel MaxSAT solver

Deterministic Parallel MaxSAT Solver

- Cooperation between threads must be deterministic
- Introduction of synchronization points
- Information is only exchanged at synchronization points
- When a thread reaches a synchronization point, waits until **all** other threads reach the same point
- Only when all threads stop at the synchronization point the information exchange takes place

Deterministic Parallel MaxSAT Solver



15 / 20

Deterministic Parallel MaxSAT Solver

- The definition of synchronization points must be deterministic
- Example: Synchronize after k conflicts
- If k is small, number of synchronization points is large and threads are idle more often
- If k is large, there is little cooperation between threads
- For our experiments, we defined k = 100
- New ways of defining synchronization points are being tested

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.

Experimental Results

Non-deterministic vs. Deterministic version

Solver	#Solved	Time (s)	Avg. Idle CPU (%)	Speedup
Non-Deterministic	141	13,401.88	0	1.00
Deterministic	140	21,611.21	43.12	0.62

- Deterministic version is slower
- Number of solved instances is very similiar
- Large idle times should be decreased with other synchronization techniques

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.
- Deterministic parallel MaxSAT solver is slower but is still able to solve almost all instances solved by the non-deterministic version