# GlueMiniSat2.2.5: A fast SAT solver with an aggressive acquiring strategy of glue clauses 

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## GlueMiniSat 2.2.5

Boolean satisfiability testing program (a SAT solver)

## MiniSat2.2 + Glucose1.0 $+\boldsymbol{\alpha}$

[Eén and Sörensson 03] [Audemard and Simon 09]

> Variant of LBD which is an evaluation criteria of learnt clauses
> Aggressive restart strategy to get good learnt clauses

- Application category of SAT 2011 competition
- $1^{\text {st }}$ in CPU time UNSAT class
- $2^{\text {nd }}$ in CPU time SAT+UNSAT class
- $2^{\text {nd }}$ in Wall-clock time UNSAT class (including parallel SAT solvers)


## Outline

- SAT
- CDCL (Conflict Driven Clause Learning) Algorithm [Silva 99, Bayardo 97]
- Evaluation Criteria of Learnt Clauses
- Literal Blocks Distance [Audemard 09]
- GlueMiniSat2.2 \& 2.2.5
- Experimental Results
- SAT 2009 Application
- SAT 2011 Application
- Covering Arrays
- Conclusion


## SAT

- Boolean satisfiability testing
- First NP-complete problem [Cook, 1971]
- Usually, represented in CNF formula


Boolean variables or their negations Disjunctions of literals
Purpose
Determines the satisfiability of a given formula

## Progress in SAT Solvers

- Dramatic performance improvement from the late 90s
- Can handle problems consisting of millions of variables
- Various techniques in the state-of-the-art SAT solvers
- Basic procedure: DPLL [Davis+ 62]
- Conflict driven clause learning (CDCL) [Silva+ 99, Bayardo+ 97]
- Backjumping [Silva+ 99, Bayardo+ 97]
- Fast unit propagation by watched literals [Moskewicz +01]
- Effective variable selection heuristics [Moskewicz+ 01]
- Restart strategy [Gomes+ 98, Luby+ 93]
- Phase caching [Pipatsrisawat+ 07]
- Fast identification of satisfied clauses
[J ain+ 07][Schubert+ 07][Sorensson+ 08]
- Canonical SAT solver: MiniSat [Eén+ 03]


## Problem Solving by SAT



Fast SAT solver

## Solution

A model
of SAT

- Planning / scheduling
- Hardware / software verification
- Theorem proving
- Constraint satisfaction / optimization problems
- Sugar [Tamura 08] which is a SAT-based CSP solver got first places in 3 categories of 2009 CSP solver competition


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## CDCL Algorithm [silva 99, Bayardo 97]

(1) Finds unit clauses and satisfies them (unit propagation)
(2) If no unit clause, selects an unassigned var and assigns 1 or 0
(3) If a conflict occurs, analyzes a cause of the conflict and learns the negation of the cause as a clause, and then backjumps to the level in which the learned clause becomes unit
$\{+1\} \quad \leftarrow u c$
$\{+1,+2,+4\}$
$\{+2,+9\}$
$\{-1,+4,+9\}$
$\{-2,+3\} \quad \leftarrow u c$
$\{-2,-5,+6\} \leftarrow u c$
$\{-1,-4,+5\} \leftarrow u c$
$\{-5,-3,+7\} \leftarrow u c$
$\{-6,+8\}$
$\{-7,-8\}$
Conflict!

Decision Stack

| Lv 0 | +1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| Lv 1 | +2 | +3 |  |  |
| Lv 2 | +4 | +5 | +6 | +7 |

2 is selected and assigned as true by heuristics
4 is selected and assigned as true by heuristics

## Conflict Driven Clause Learning

\{+1\}
$\{+1,+2,+4\}$
$\{+2,+9\}$
$\{-1,+4,+9\}$
$\{-2,+3\}$
$\{-2,-5,+6\}$
$\{-1,-4,+5\}$
$\{-5,-3,+7\}$
$\{-6,+8\}$
$\{-7,-8\}$
Decision Stack

| Lv 0 | +1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  | +2 | +3 |  |  |
| Lv 2 | +4 | +5 | +6 | +7 |

Implication Graph


If +2 $\wedge+5 \wedge+3$, then contradicts


Learns the clause $-2 \vee-5 \vee-3$

## Management of Learnt Clauses

- Learnt clauses are useful to prevent same conflicts
- However there is a trade-off:
> It is difficult to preserve all learnt clauses since it consumes memory and unit propagations becomes slow.
> If learnt clauses are not preserved, the search process repeats same conflicts and becomes slow
- Hence, CDCL solvers reduce learnt clauses periodically

How to select learnt clauses which will be preserved?

## Evaluation Criteria of Learnt Clauses

- Length
- Short learnt clauses have high pruning power
- Activity
> Chaff [Moskewicz+ 01] , MiniSat [Eén+ 03]
> Defines activity for each learnt clause, removes clauses whose activity is less than a certain threshold
> Activity is raised when the clause is used to produce a contradiction
> Least recently used (LRU) learnt clauses are removed
- LBD (Literal Blocks Distance)
> LBD is a measure to evaluate the possibility of use of learnt clauses in the future
> Glucose1.0 [Audemard and Simon, 09]
$\checkmark 1^{\text {st }}$ in UNSAT class and $2^{\text {nd }}$ in SAT+UNSAT class at Application category of SAT 2009 Competition


## LBD

## Learnt clause $\left\{L_{1}, L_{2}, L_{3}, L_{4}, L_{5}, L_{6}\right\}$ <br> $\begin{array}{lllllll}\text { Decision level } & 7 & 5 & 5 & 2 & 2 & 2\end{array}$

## Decision Stack



I mplication Decision

## LBD



- A set of variables assigned at the same DLV is called a block
- LBD of a clause $\boldsymbol{C}$ is defined as \# blocks in $\boldsymbol{C}$
$\checkmark$ Variables in a block have possibility that they will be assigned as false at the same time by unit propagations
$\checkmark$ LBD can be considered as a generalization of length criteria


## Glue Clauses

- Especially, clauses whose LBD=2 are called glue clauses
- Glue clauses promote unit propagations even if they are long

$\mathbf{L} \mathbf{1}$ is propagated when $\mathbf{L 2} \sim \mathbf{L 6}$ are assigned as false


## Glucose never removes glue clauses

## GlueMiniSat2.2 GlueMiniSat2.2 = MiniSat2.2 + Glucose1.0

|  | $\begin{gathered} \text { MiniSat } \\ 2.0 \end{gathered}$ | $\begin{gathered} \text { MiniSat } \\ 2.2 \end{gathered}$ | Glucose 1.0 | $\begin{gathered} \text { GlueMiniSat } \\ 2.2 \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: |
| Var selection heuristics | VSIDS | VSIDS | VSIDS | VSIDS |
| Randomness | 2\% | 0\% | 2\% | 0\% |
| Evaluation criteria of learnts | LRU | LRU | LBD | LBD |
| Reduction strategy of learnts | Exponential <br> (\#C/3)*1.1 ${ }^{\text {r }}$ | Exponential $(\# C / 3) * 1.1^{\mathrm{d}}$ | Linear 20000+500x | Linear $20000+10000 x$ |
| Restart strategy | Exponential $100 * 1.5^{r}$ | Luby | Dynamic (LBD) | Dynamic (LBD) |
| Phase caching |  | $\checkmark$ | $\checkmark$ | $\checkmark$ |
| Fast identification of satisfied clauses |  | $\checkmark$ | $\checkmark$ | $\checkmark$ |
| Memory management | malloc | Single area | malloc | Single area |

## Experimental Results in SAT 2009 Application



## From Development of GlueMiniSat 2.2

We got the following assumptions:
(a) Important to promote acquiring clauses with small LBD
(b) For unsatisfiability proof, important to preserve useful learned clauses as many as possible

(a) Aggressive restart strategy
(b) Expanding a set of preserved learnt clauses which are never removed

## Expanding Preserved Learnts

- Low performance if it holds learnts with LBD $\leqq 3$



## Details of LBD Computation

- A clause $C$ is glue
(1) when $C$ is generated from a conflict and the LBD is 2
(2) when $C$ is used in unit propagations and the LBD is 2 (LBD is recalculated by the current truth assignment)


Glue clauses

## Pseudo LBD

| When? | (1) generated from a conflict | (2) used in unit propagations |
| :---: | :---: | :---: |
| Leant clause Decision Lv | $\left\{\begin{array}{\|c\|ccc\|cc}L_{1}\end{array}, \begin{array}{c}L_{2}, \\ L_{3}, \\ \hline\end{array}\right.$ | $\left\{\begin{array}{cc\|cc\|cc}\hline L_{1}, & L_{2}, & L_{3}, & L_{4}, & L_{5}, & L_{6} \\ \mathbf{7} & \mathbf{7} & \mathbf{5} & \mathbf{5} & \mathbf{4} & \mathbf{4}\end{array}\right\}$ |
| LBD | 3 | 3 |

- GlueMiniSat holds learnt clauses with pseudo LBD $\leqq 3$
- A learnt clause from (1) always contains unit literal block. Hence, the clause somewhat promotes unit propagations even if LBD is 3
- A learnt clause from (2) may not contain unit literal block. Hence, GlueMiniSat holds learnts with pseudo LBD $\leqq 3$ (LBD $\leqq 2$ )


## LBD vs Pseudo LBD



## From Development of GlueMiniSat 2.2

We got the following assumptions:
(a) Important to promote acquiring glue clauses
(b) For unsatisfiability proof, important to preserve learned clauses which will be used in the future
(a) Aggressive restart strategy
(b) Expanding a set of preserved learnt clauses

## Restart Strategy of GlueMiniSat

- Restart strategy for DLVs

Local avg. of DLVs $* \mathbf{1 . 0}>$ Global avg. of DLVs over the last 50 conflicts

- Restart strategy for LBDs (same as Glucose1.0)

Local avg. of LBDs over the last 50 learnt clauses

* $0.8>$ Global avg. of LBDs


## Restarts if either condition is satisfied Purpose is to reduce DLVs and get small LBD clauses

## Results of Restart by DLV and LBD



## Experimental Results in SAT 2009 Application

|  | MiniSat | Glucose |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 2.0 | 1.0 | MiniSat <br> 2.2 | GlueMiniSat |  |  |
|  |  |  | 2.2 | Pseudo <br> LBD + AR |  |
| \# solved <br> (SAT/ UNSAT) | 109 <br> $(49 / 60)$ | 133 <br> $(52 / 81)$ | 141 <br> $(60 / 81)$ | 154 | 161 |
| Average time <br> [sec] | 193 | 206 | 167 | 197 | $(61 / 100)$ |
| Restart speed <br> [confs/ restart] | 14229 | 1152 | 528 | 456 | 117 |

- Enhanced the strength for UNSAT
- Restarts very aggressively

Environment: Mac mini, Core 2 Duo 1.83GHz, 2GB RAM 1000 CPU sec / instance

## SAT 2011 Application CPU Time

Gold

## SAT+UNSAT Glucose2. • Strong for UNSAT

$S$ - Weak for SAT (19th of 20 solvers in final stage)
silver
GlueMiniSat
Linaeling

GlueMiniSat Glucose2.0
GlueMiniSat Gluc
WC Time

UNSAT

Silver
Bronze

## SAT+UNSAT

SAT
UNSAT

Plingeling // CryptoMiniSat //
ppfolio // Plingeling // CryptoMiniSat // GlueMiniSat
ppfolio //
contrasat
(MiniSat hack) Plingeling //

## Conclusion

- GlueMiniSat is strong for UNSAT proof
- GlueMiniSat holds more glue clauses than Glucose
> Prevents losing useful clauses required to prove unsatisfiability
- GlueMiniSat restarts more aggressively than Glucose
> Contributes to acquire good learnt clauses


## Future Work

- Comparison with strong algorithms for SAT
- Extension from sequential to parallel

