Knowledge Compilation for Incremental and Checkable Stochastic Boolean Satisfiability

Che Cheng [1*] Yun-Rong Lauren Luo [2*] Jie-Hong Roland Jiang [1]

[1]National Taiwan University (NTU), Taipei, Taiwan

[2]University of Michigan, Ann Arbor, MI, USA

[*]Contributed equally

The 33rd International Joint Conference on Artificial Intelligence

Motivation 1: Knowledge Compilation for SSAT

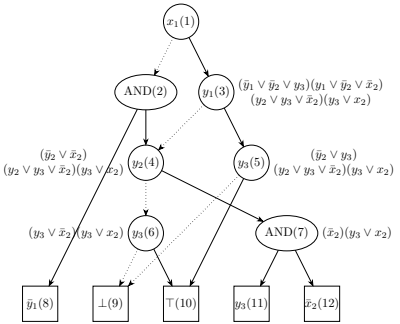
SSAT: $Q_1 v_1, \ldots, Q_n v_n . \phi$, $Q_i \in \{\bowtie^p, \exists\}$, where $\bowtie^p r$ denotes that $\Pr[r = \top] = p$

- Randomized variant of QBF
- Semantics: maximum satisfying probability $\Pr[\Phi]$

Motivation 1: Knowledge Compilation for SSAT

SSAT: $Q_1 v_1, \ldots, Q_n v_n . \phi$, $Q_i \in \{\forall^p, \exists\}$, where $\forall^p r$ denotes that $\Pr[r = \top] = p$

- Randomized variant of QBF
- Semantics: maximum satisfying probability $\Pr[\Phi]$



- Observation: the trace of a run of the SSAT solver **SharpSSAT** [FJ23] is a *dec-DNNF* [DM02].

  $\Longrightarrow$ Why not compile the dec-DNNF and use it?

Motivation 2: Incremental and Checkable SSAT

- Different but similiar SSAT formulas may result in an identical **SharpSSAT** trace
  - $\Longrightarrow$ let **SharpSSAT** re-use the trace and avoid repeated searches
  - $\Longrightarrow$ *Incremental SSAT*

Motivation 2: Incremental and Checkable SSAT

- Different but similiar SSAT formulas may result in an identical **SharpSSAT** trace
  - $\Longrightarrow$ let **SharpSSAT** re-use the trace and avoid repeated searches
  - $\Longrightarrow$ *Incremental SSAT*
- The trace is the footprint of a **SharpSSAT** run
  - $\Longrightarrow$ a proof/certificate for the run that can be independently checked
  - $\Longrightarrow$ *Checkable SSAT*

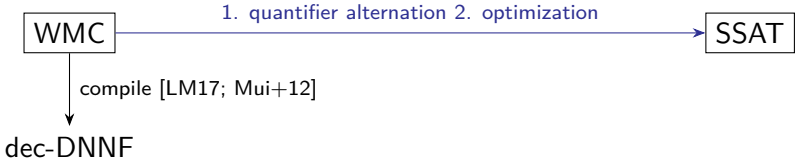
Contribution 1: Knowledge Compilation for SSAT

- Lift dec-DNNF compilation to SSAT:

SSAT

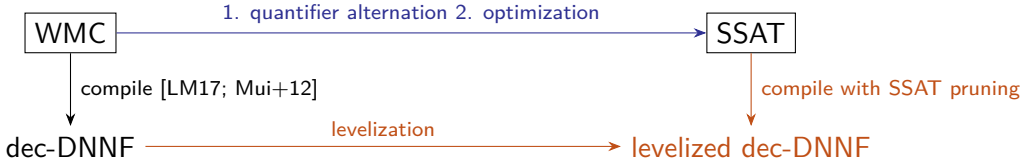
Contribution 1: Knowledge Compilation for SSAT

- Lift dec-DNNF compilation to SSAT:
  1. decision orders should follow the quantifier alternation levels in SSAT prefix.
  2. SSAT decision pruning may occur in SSAT solving due to the optimization nature of SSAT



Contribution 1: Knowledge Compilation for SSAT

- Lift dec-DNNF compilation to SSAT:
  1. decision orders should follow the quantifier alternation levels in SSAT prefix.
  2. SSAT decision pruning may occur in SSAT solving due to the optimization nature of SSAT
- Contribution: propose *levelized dec-DNNF* and compilation with SSAT pruning.



Contribution 1A: Levelized Dec-DNNF

SSAT formula: $\Phi = \mathfrak{A}^{0.4} x_1, \exists y_1, \exists y_2, \exists y_3, \mathfrak{A}^{0.6} x_2.\phi$

quantifier alternation levels: $x_1 \prec y_1 = y_2 = y_3 \prec x_2$

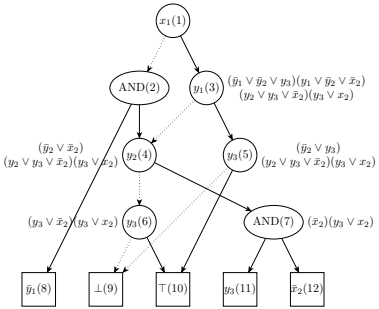


Figure: Levelized dec-DNNF G for Φ.

Definition (Levelized Dec-DNNF)

A dec-DNNF G s.t. for decision nodes $N_1 \leq N_2$ in G, their decision variables satisfy $v_1 \preceq v_2$

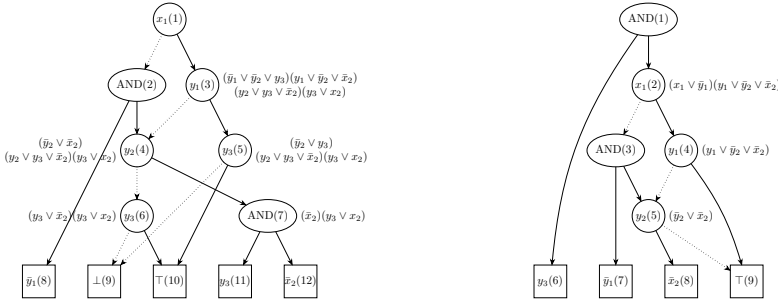
Theorem (SSAT Evaluation)

$\Pr[\Phi]$ *can be evaluated with* G *in one reversed topological traversal.*

Contribution 1B: Compilation with SSAT Pruning

Incorporate *all* pruning techniques[1] in the SSAT solver **SharpSSAT** [FJ23]

$\Longrightarrow$ a more compact levelized dec-DNNF[2] with the same SSAT evaluation result.



(a) Without pure literal detection.

(b) With pure literal detection.

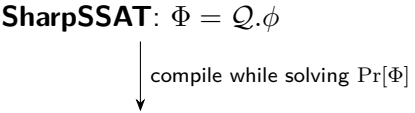
Figure: Levelized dec-DNNF graphs.

[1] pure literal detection and existential early return

[2] Whenever pruning occurs at a decision node, replace each unexplored node with the constant node $\bot$ or the other explored node.

Contribution 2: Incremental SSAT Evaluation with Levelized Dec-DNNF

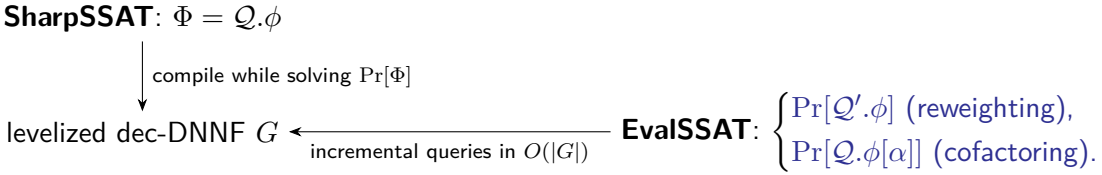
- **SharpSSAT** compiles the SSAT $\Phi = \mathcal{Q}.\phi$ into a levelized dec-DNNF G while solving $\Pr[\Phi]$



[3] Cofactoring cannot be correctly computed with SSAT pruning enabled.

Contribution 2: Incremental SSAT Evaluation with Levelized Dec-DNNF

- **SharpSSAT** compiles the SSAT $\Phi = \mathcal{Q}.\phi$ into a levelized dec-DNNF G while solving $\Pr[\Phi]$
- **EvalSSAT** performs linear incremental queries on G for maximum satisfying probability of:
  - a reweighting $\mathcal{Q}.\phi \mapsto \mathcal{Q}'.\phi$: differ from Φ in the probabilities of the randomized variables
  - a *cofactoring*[3] $\mathcal{Q}.\phi \mapsto \mathcal{Q}.\phi[\alpha]$: variables in the matrix are substituted with Boolean constants



[3] Cofactoring cannot be correctly computed with SSAT pruning enabled.

Contribution 3: SSAT Validation with Levelized Dec-DNNF

- We develop an SSAT proof framework **cert-SSAT** based on the model counting proof framework **CPOG** [Bry+23] to validate the correctness of $\Pr[\Phi]$ reported by **SharpSSAT**.
- **cert-SSAT** validates $\Pr[\Phi]$ by sandwiching: $\mathrm{LB}(\Pr[\Phi]) = \Pr[\Phi] = \mathrm{UB}(\Pr[\Phi])$[4]

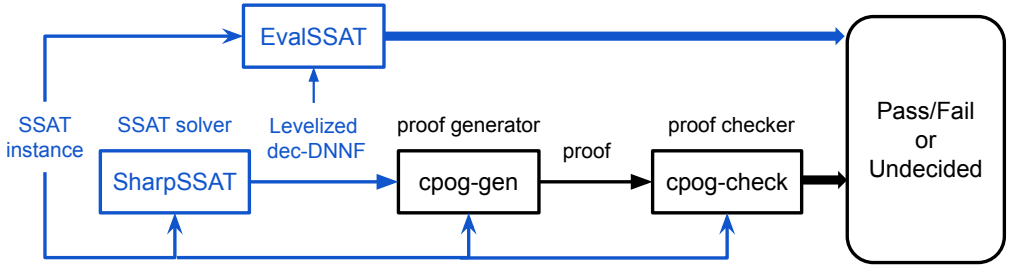


Figure: Toolchain flow for **cert-SSAT**.

[4](1) Compiles two levelized dec-DNNFs G_l, G_u from SSAT $\Phi = \mathcal{Q}.\phi$; (2) Proves $(G_l \to \phi) \wedge (\phi \to G_u)$; (3) Proves **EvalSSAT**$(\Phi, G_l) = \Pr[\Phi] =$ **EvalSSAT**(Φ, G_u)

Experimental Results 1: Incremental SSAT Evaluation

KC: solving $\Pr[\Phi]$ incrementally with knowledge compilation[5]

Baseline: solving $\Pr[\Phi]$ for each individual query from scratch[6]

[5]Time limit: 1000 sec for compilation and 100 sec for each query

[6]Time limit: 200 sec for each query

[7]SSAT benchmark: https://github.com/NTU-ALComLab/ClauSSat

Experimental Results 1: Incremental SSAT Evaluation

KC: solving $\Pr[\Phi]$ incrementally with knowledge compilation[5]

Baseline: solving $\Pr[\Phi]$ for each individual query from scratch[6]

1. Reweighting task: for each SSAT formula[7], randomly generate 10 reweighting queries
   - # instances solved: **KC: 2220** $>$ Baseline: 2212
   - PAR2: **KC: 154.22** $<$ Baseline: 157.36
2. Cofactoring task: for each SSAT formula, randomly generate 10 cofactoring queries
   - # instances solved: **Baseline: 2361** $>$ KC: 1870
   - PAR2: **Baseline: 140.03** $<$ KC: 193.71

[5] Time limit: 1000 sec for compilation and 100 sec for each query

[6] Time limit: 200 sec for each query

[7] SSAT benchmark: https://github.com/NTU-ALComLab/ClauSSat

Experimental Results 2: SSAT Validation

- Out of 236 solvable SSAT instances for **SharpSSAT**[8], **cert-SSAT**[9] validates LB($\Pr[\Phi]$) for 205 instances (86.9%) and validates $\Pr[\Phi]$ for 190 instances (80.5 %)[10]

[8] SSAT benchmark: https://github.com/NTU-ALComLab/ClauSSat

[9] Time limit: **SharpSSAT/EvalSSAT:** 1000 sec, **cpog-gen/cpog-check:** 2500 sec

[10] Validating lower and upper bounds are independent and validating a lower bound is easier.

Experimental Results 2: SSAT Validation

- Out of 236 solvable SSAT instances for **SharpSSAT**[8], **cert-SSAT**[9] validates LB($\Pr[\Phi]$) for 205 instances (86.9%) and validates $\Pr[\Phi]$ for 190 instances (80.5 %)[10]
- **cert-SSAT** helps discover one bug of **SharpSSAT**!

[8]SSAT benchmark: https://github.com/NTU-ALComLab/ClauSSat

[9]Time limit: **SharpSSAT/EvalSSAT:** 1000 sec, **cpog-gen/cpog-check:** 2500 sec

[10]Validating lower and upper bounds are independent and validating a lower bound is easier.

Conclusions

- Contributions:
  - Proposed a dec-DNNF-based knowledge compilation technique for SSAT
  - Proposed incremental and checkable SSAT solving scenarios and demonstrated their effectiveness

Conclusions

- Contributions:
  - Proposed a dec-DNNF-based knowledge compilation technique for SSAT
  - Proposed incremental and checkable SSAT solving scenarios and demonstrated their effectiveness
- Future directions:
  - More applications for levelized dec-DNNF compilers
  - More types of incremental SSAT queries
  - Extend **cert-SSAT** to support SSAT preprocessor and DSSAT solver [CJ23]

References I

[Bry+23] Randal E. Bryant et al. “Certified Knowledge Compilation with Application to Verified Model Counting”. In: *Proceedings of the International Conference on Theory and Applications of Satisfiability Testing*. Vol. 271. 2023, 6:1–6:20.

[CJ23] Che Cheng and Jie-Hong R. Jiang. “Lifting (D)QBF preprocessing and solving techniques to (D)SSAT”. In: *Proceedings of the AAAI Conference on Artificial Intelligence*. AAAI Press, 2023, pp. 3906–3914.

[DM02] Adnan Darwiche and Pierre Marquis. “A Knowledge Compilation Map”. In: *Journal of Artificial Intelligence Research* 17.1 (2002), pp. 229–264.

[FJ23] Yu-Wei Fan and Jie-Hong R. Jiang. “SharpSSAT: A Witness-Generating Stochastic Boolean Satisfiability Solver”. In: *Proceedings of the AAAI Conference on Artificial Intelligence*. Vol. 37. 2023, pp. 3949–3958.

[LM17] Jean-Marie Lagniez and Pierre Marquis. “An Improved Decision-DNNF Compiler”. In: *Proceedings of the International Joint Conference on Artificial Intelligence*. AAAI Press, 2017, pp. 667–673.

[Mui+12] Christian Muise et al. “Dsharp: Fast d-DNNF Compilation with SharpSAT”. In: *Proceedings of the Canadian Conference on Advances in Artificial Intelligence*. Springer-Verlag, 2012, pp. 356–361.