# ProCount: Weighted Projected Model Counting with Graded Project-Join Trees

Jeffrey M. Dudek - Vu H. N. Phan - Moshe Y. Vardi

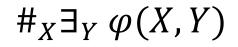
Rice University July 8, 2021



SAT 2021

# In This Talk

## **Problem:** Weighted Projected Model Counting



Applications in planning, formal verification, and infrastructure reliability

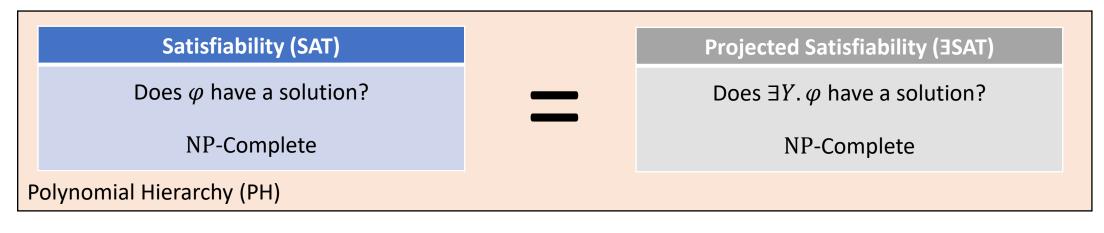
## **Our Solution:** Two-phase algorithm based on graded project-join trees

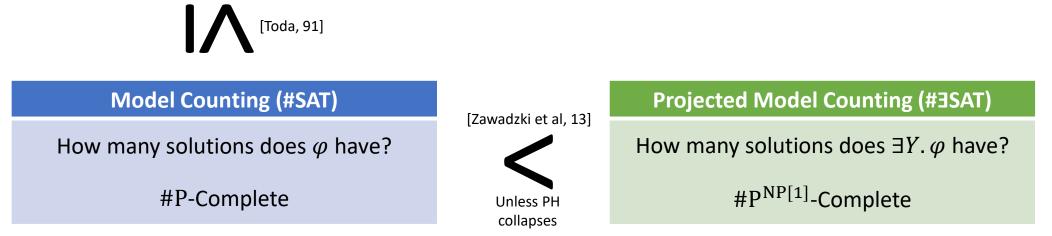
- 1. Planning: Use tree decompositions to build a graded project-join tree
- 2. Execution: Process graded project-join tree with algebraic decision diagrams (ADDs) to get the count

## **Experiments:** Our tool ProCount is fastest on 34% of solved benchmarks

https://github.com/vardigroup/DPMC

# **The Problem: Projected Model Counting**





 $\varphi$  is a CNF formula,  $Y \subseteq Vars(\varphi)$  3

# **Background: Projected Model Counting**

**Problem:** Projected Model Counting

Input: A CNF formula  $\varphi(X, Y)$  over disjoint variable sets X and YOutput:  $\#_X \exists_Y \varphi(X, Y)$ The number of  $\vec{x} \in 2^X$  s.t. there exists  $\vec{y} \in 2^Y$  where  $\varphi(\vec{x}, \vec{y}) = 1$ 

> Everything in this work generalizes to literal-weighted projected model counting.

### Example:

$$X = \{x_1, x_2\} Y = \{y_1, y_2\} \varphi(X, Y) = (x_1 \lor \neg x_2 \lor y_1) \land (x_1 \lor y_2) \land (\neg y_1 \lor \neg y_2)$$

Solutions to  $\exists Y. \varphi$  are:  $(x_1 = 0, x_2 = 0), (x_1 = 1, x_2 = 0)$ , and  $(x_1 = 1, x_2 = 1)$ 

Thus  $\#_X \exists_Y \varphi(X, Y) = 3$ 

# **Background: Projected Model Counting**

**Problem:** Projected Model Counting Input: A CNF formula  $\varphi(X, Y)$  over disjoint variable sets X and YOutput:  $\#_X \exists_Y \varphi(X, Y)$ The number of  $\vec{x} \in 2^X$  s.t. there exists  $\vec{y} \in 2^Y$  where  $\varphi(\vec{x}, \vec{y}) = 1$ 

Techniques for *exact* projected model counting:

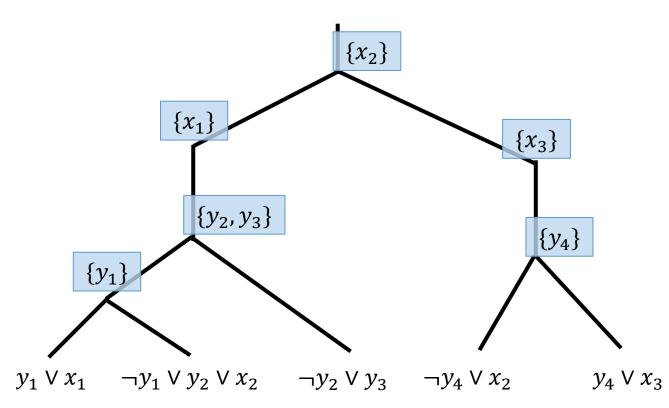
- 1. Search: Reason directly about  $\varphi$  using a SAT solver
  - projMC [Lagniez & Marquis, 19], reSSAT [Lee et al., 17]
- 2. Knowledge Compilation: Compile  $\varphi$  to a representation where counting is easy
  - D4<sub>P</sub> [Lagniez & Marquis, 19]
- 3. Dynamic Programming: Reason about the clause structure of  $\varphi$ 
  - nestHDB [Hecher et al., 20]
  - This work

There is also *approximate* projected model counting, but we focus on exact.

# **DPMC: Model Counting Algorithm**

[**Dudek** et. al, 20]

- **1. Planning**: Build a project-join tree of  $\varphi(X)$ .
- **2.** Execution: Process project-join tree from leaves up to compute  $\#_X \varphi(X)$ .

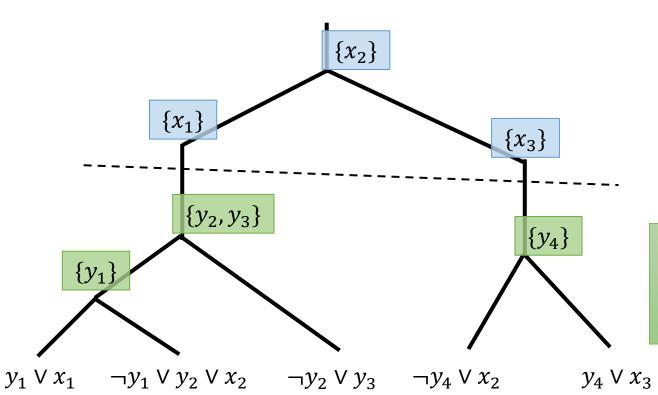


**Definition:** A *project-join tree* for  $\varphi$  is a tree where:

- 1. Each clause of  $\varphi$  is assigned a (unique) leaf node
- 2. Each variable of  $\varphi$  is assigned an internal node
- 3. For all clauses *C* and variables *z* that appear in *C*, the *z* node is an ancestor of the *C* node

# **Our Algorithm for Projected Model Counting**

- **1. Planning**: Build an (*X*,*Y*)-graded project-join tree of  $\varphi(X, Y)$ .
- 2. Execution: Process graded project-join tree from leaves up to compute  $\#_X \exists_Y \varphi(X, Y)$ .



**Definition:** A *project-join tree* for  $\varphi$  is a tree where:

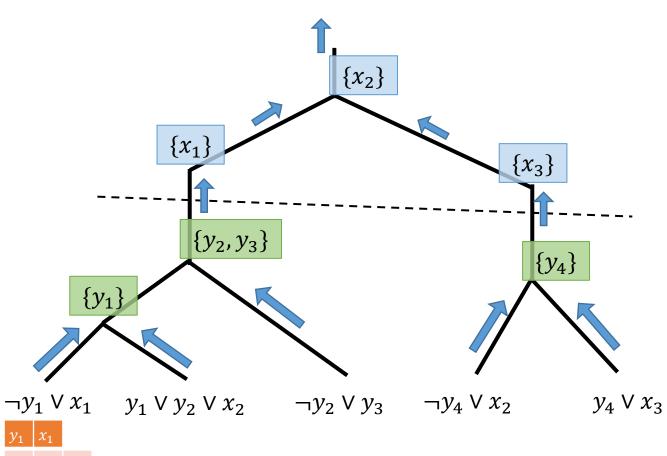
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**Definition:** A project-join tree is **(***X***,***Y***)**-*graded* if:

4. For all variables  $x \in X$  and  $y \in Y$ , the y node is not an ancestor of the x node

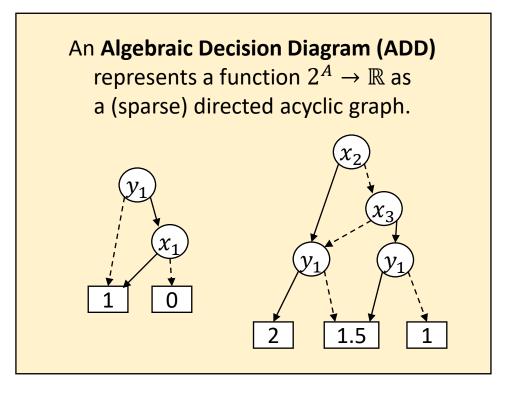
# **2. Execution**

1 1

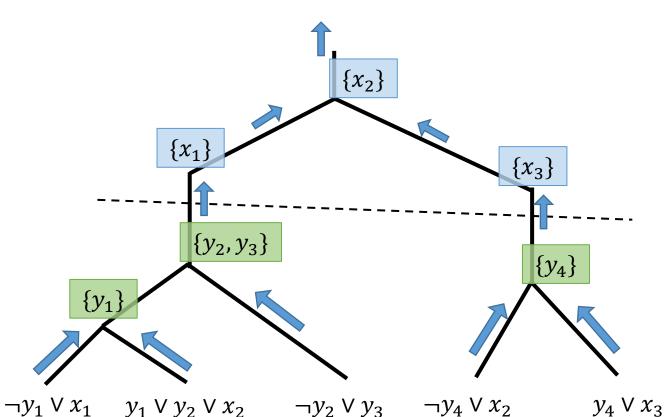


### Idea:

Pass ADDs through tree from leaves to root, projecting away variables according to the labels.



# **2. Execution: Running Time**



#### Idea:

Pass ADDs through tree from leaves to root, projecting away variables according to the labels.

Key performance measure:

- *Width* of the project-join tree
- I.e., the maximum number of variables needed for a single ADD
- Width can be computed upfront

#### Theorem: Given:

- A CNF formula  $\varphi(X, Y)$
- An (X,Y)-graded project-join tree of  $\varphi(X,Y)$  of width w

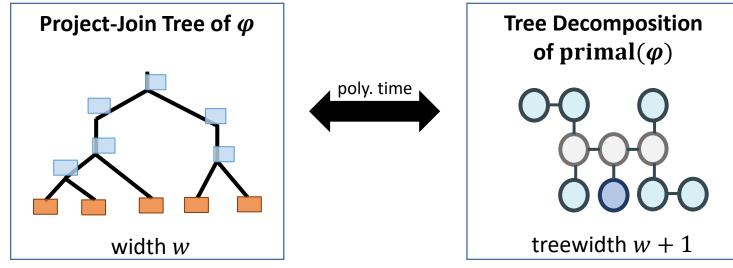
This procedure computes  $\#_X \exists_Y \varphi(X, Y)$  in time  $O(2^w \cdot \text{poly}(|\varphi|))$ .

Projected counting (parameterized by width) with *ungraded* project-join trees is  $\Omega(2^{2^w})$  assuming ETH [Fichte et al., 18]

# 1. Planning (Model Counting)

How to find a low-width project-join tree of  $\varphi$ ?

[McMahan et al., 04] [Kask et al., 05] [Markov and Shi, 05]



Decompositions show a "good" way to reason about a graph. Black-box, heuristic tree-decomposition solvers:

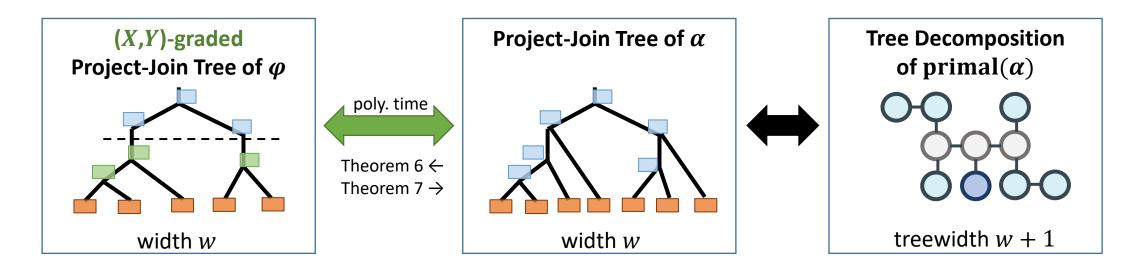
- FlowCutter [Hamann and Strasser, 18]
- Tamaki [Tamaki, 17]
- htd [Abseher et al., 17]

# 1. Planning (Projected Model Counting)

How to find a low-width (*X*,*Y*)-graded project-join tree of  $\varphi$ ?

**Possible Approach?** Modify tree-decomposition solvers to take into account different variable types. **Better Idea:** Use previous planners as a black box.

Add "virtual" clauses to  $\varphi$  to construct a new formula  $\alpha$  so that:



# **1. Planning: The Reduction**

Constructing  $\alpha$ :

1. Build the *primal graph* of  $\varphi$ 

A vertex for every variable, and an edge if two variables appear together in a clause

- 2. Examine the connected components of *Y* variables in the primal graph
- 3. For each connected component, add a "virtual clause" of the adjacent X variables to  $\alpha$

### Example:

$$\varphi = \left\{ \begin{array}{c} \neg y_1 \lor x_1 \\ y_1 \lor y_2 \lor x_2 \\ \neg y_2 \lor y_3 \\ \neg y_4 \lor x_2 \\ y_4 \lor x_3 \end{array} \right\} \qquad \begin{array}{c} x_1 & y_1 \\ x_2 & y_1 \\ x_2 & y_2 \\ y_2 & y_3 \\ x_3 & y_4 \\ y_4 \end{pmatrix}$$

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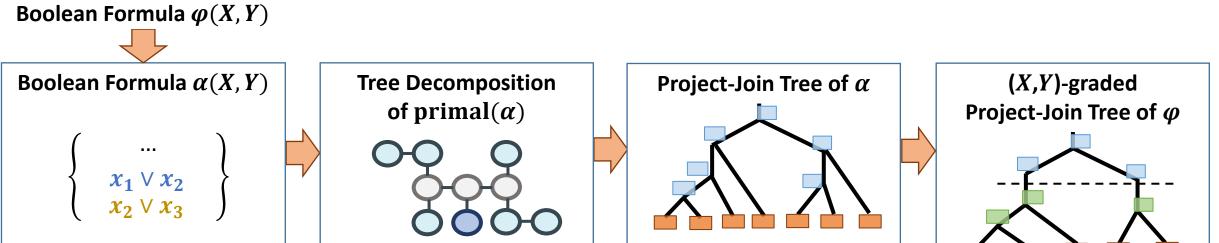
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Example:

# **Algorithm Overview: ProCount**



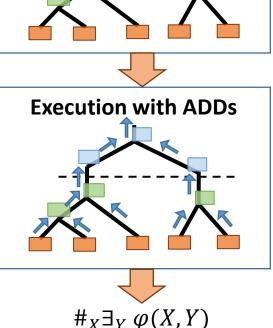
ProCount: Implemented in C++

https://github.com/vardigroup/DPMC

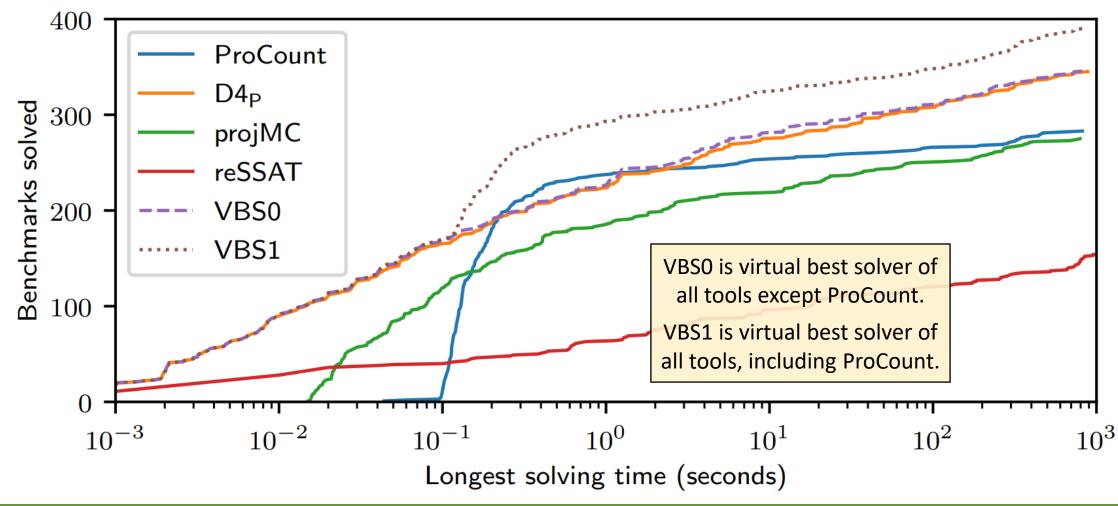
**Planning:** Black-box tree-decomposition solvers

• FlowCutter, Tamaki, htd

## **Execution:** ADDs with CUDD



## **Experimental Evaluation on Weighted #3SAT**



### Our counter ProCount is the fastest tool on 131 benchmarks (34% of solved benchmarks)

\* Run on a single 2.60 GHz core with 30 GB RAM. Used 849 #3SAT benchmarks from [Gupta et al., 19] and [Soos and Meel, 19]. 16

# **Summary and Conclusion**

**Problem:** Weighted Projected Model Counting:  $\#_X \exists_Y \varphi(X, Y)$ 

**Our Solution:** Two-phase algorithm based on graded project-join trees

- **1. Planning**: Use tree decompositions to build a graded project-join tree Using previous planning as a black box lets us use *unmodified* tree-decomposition tools
- 2. Execution: Process graded project-join tree with ADDs to get the count Using graded projected-join trees lets us avoid a double-exponential dependency on width

## **Experiments:** ProCount improves the VBS on 34% of benchmarks solved by at least one tool <u>https://github.com/vardigroup/DPMC</u>

### **Future Work:**

- More quantifier alternation (#QBF, MAP inference, FAQ problems)
- Planning with other graph decompositions
- Parallelization



